

Chapter 25

100 Years of Progress in Hydrology

CHRISTA D. PETERS-LIDARD,^a FAISAL HOSSAIN,^b L. RUBY LEUNG,^c NATE MCDOWELL,^c
MATTHEW RODELL,^d FRANCISCO J. TAPIADOR,^e F. JOE TURK,^f AND ANDREW WOOD^g

^a *Earth Sciences Division, NASA Goddard Space Flight Center, Greenbelt, Maryland*

^b *Department of Civil and Environmental Engineering, University of Washington, Seattle, Washington*

^c *Atmospheric Sciences and Global Change Division, Pacific Northwest National Laboratory, Richland, Washington*

^d *Hydrological Sciences Laboratory, NASA Goddard Space Flight Center, Greenbelt, Maryland*

^e *Department of Environmental Sciences, Institute of Environmental Sciences, University of Castilla-La Mancha, Toledo, Spain*

^f *Jet Propulsion Laboratory, California Institute of Technology, Pasadena, California*

^g *Research Applications Laboratory, National Center for Atmospheric Research, Boulder, Colorado*

ABSTRACT

The focus of this chapter is progress in hydrology for the last 100 years. During this period, we have seen a marked transition from practical engineering hydrology to fundamental developments in hydrologic science, including contributions to Earth system science. The first three sections in this chapter review advances in theory, observations, and hydrologic prediction. Building on this foundation, the growth of global hydrology, land-atmosphere interactions and coupling, ecohydrology, and water management are discussed, as well as a brief summary of emerging challenges and future directions. Although the review attempts to be comprehensive, the chapter offers greater coverage on surface hydrology and hydrometeorology for readers of this American Meteorological Society (AMS) monograph.

1. Introduction: From engineering to hydrologic science

The development of ancient civilizations along the Nile, Tigris–Euphrates, Indus, and Huang-Ho Rivers is not an accident—proximity to water for drinking, sanitation, agriculture, and navigation enriched their livelihoods. Soon after these civilizations were established, they began to measure and manage water, including constructing flood control dams, collecting stream gauge information, and building irrigation canals (Noaman and El Quosy 2017; Harrower 2008; Chandra 1990). There is evidence to support that these ancient civilizations understood the concept of the hydrologic cycle, including precipitation as the source of groundwater recharge through infiltration, and the role of solar radiation in evapotranspiration (Chow 1964; Chandra 1990). As summarized by Baker and Horton (1936),

there were competing theories of the origin of springs and rivers by ancient philosophers from Aristotle, Vitruvius, and Ovid to Seneca, who described “Forms of Water” in his “Questions Naturalis.”

Many of these ideas remained in place until the Renaissance, when Leonardo da Vinci (ca. 1500) classified hydrologic processes using hypothesis-driven science, including the hydrologic cycle (Pfister et al. 2009). Almost 200 years later during the Scientific Revolution in the sixteenth and seventeenth centuries, the French writer Pierre Perrault was the first to take a quantitative approach to understanding the nature of the hydrologic cycle (Deming 2014). In England, astronomer Edmond Halley also studied the hydrologic cycle. However, French Academy member Edme Mariotte was the first to quantitatively demonstrate a fundamental concept of hydrogeology, which is that precipitation and infiltration ultimately comprise streamflow (Deming 2017). Mariotte also made fundamental contributions to hydrostatics and applied this knowledge to engineer the water supply at Versailles. Clearly from ancient times

Corresponding author: Christa Peters-Lidard, christa.peters@nasa.gov

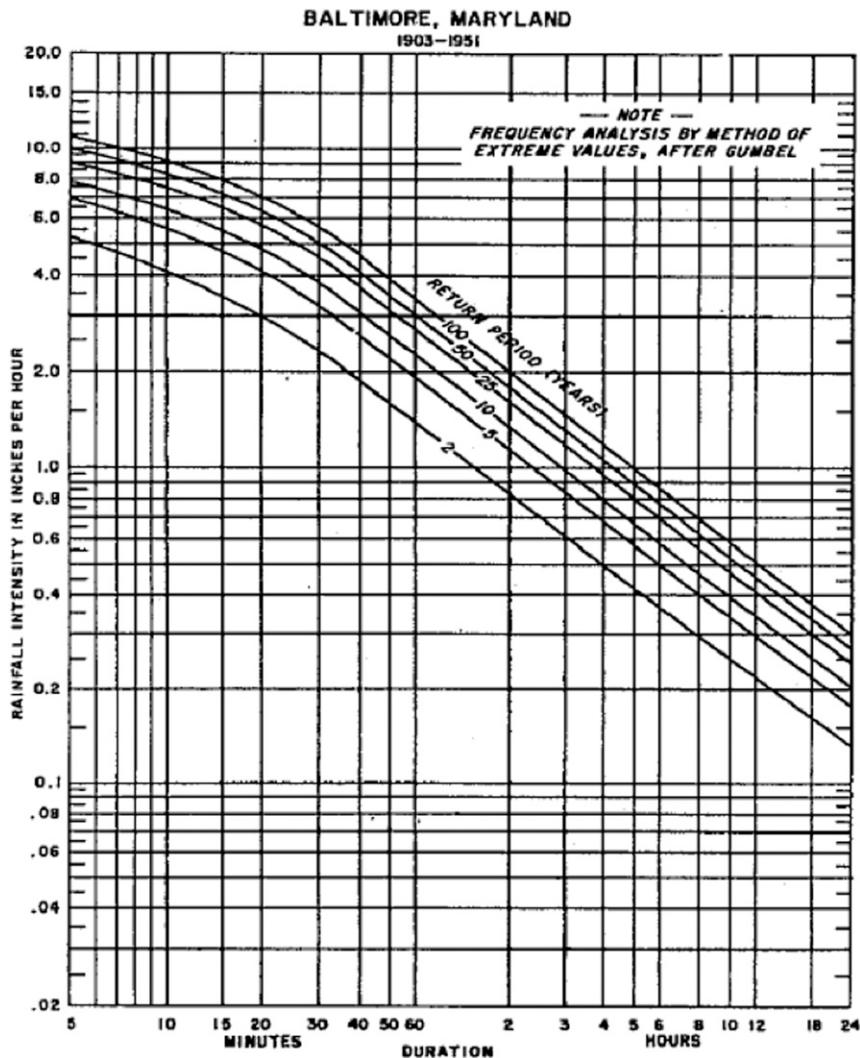


FIG. 25-1. Intensity-duration-frequency curve for Baltimore, Maryland, from TP-40 (Hershfield 1961a). [From Weather Bureau (1955); Source: NOAA/NWS.]

through the Scientific Revolution, advances in hydrologic science were often made by polymaths observing and testing their knowledge to support water management.

To meet the needs of flood design, land and forest management, and economic efficiency, national governments established programs for measuring and analyzing rainfall. In the United States, this function was carried out by the U.S. Weather Bureau, which became the National Weather Service (NWS) in 1970. The Weather Bureau conducted rainfall analyses that supported road building and the design of engineering structures such as sewers and dams. For example, since the late nineteenth century, civil engineers have utilized the so-called rational method (Kuichling 1889) for roadway drainage design. A related method from the

U.S. Department of Agriculture (USDA) Natural Resources Conservation Service [NRCS; formerly Soil Conservation Service (SCS)], known as the SCS Runoff Curve Number method (Mockus 1972), can also be used. These methods rely on “design storms” of a known intensity for a given return period and duration. In the United States, design storm intensities are derived using intensity-duration-frequency information similar to Fig. 25-1, which is based on Weather Bureau Technical Paper 40 (Hershfield 1961a). Typical return periods include 25, 50, or 100 years, depending on the importance of the roadway. Even today, such rainfall analyses underpin the design of important structures such as detention basins, roadways, bridges, and dams, and these maps continue to be periodically updated by the NWS (e.g., Bonnin et al. 2006).

The focus of this chapter is progress in hydrology for the last 100 years. As with any effort to track the progress of a field over a century, it is not quite possible to document all the advancements made across all sub-disciplines. However, to make our review scoped appropriately for this American Meteorological Society (AMS) monograph and its readers, we provide greater focus on the theoretical underpinnings of surface processes, the atmosphere above, and the interactions within the land–atmosphere interface. During the last 100 years, we have seen a marked transition that has improved practical applications of hydrology through fundamental advancements in hydrologic science, including contributions to Earth system science (Sivapalan 2018). As first described in Chow (1964) and later proposed and extended by Sivapalan and Blöschl (2017) and shown in Fig. 25-2, hydrology first progressed through the Empirical Era (1910–30), to the Rationalization Era (1930–50), to the Systems Era (1950–70). These periods were followed by the Process Era (1970–90), the Geosciences Era (1990–2010), and finally by the current Coevolution Era (2010–30). As noted in the figure, the foundations of networks, experimental basins, operations research, high-performance computing, remote sensing, and big data have advanced hydrological understanding. At the time of the founding of the AMS, there was a single journal—*Monthly Weather Review* (*MWR*)—that served as an outlet for hydrologic science in the AMS community. In the first several decades of this period (1919–59, Empirical to Rationalization Eras), *MWR* articles reflect the emergence of quantitative hydrology from empirical observation (as will be discussed in section 2 below). Examples include discussions of floods (Henry 1919, 1928; Nagler 1933), rainfall–runoff relationships (Fischer 1919; Shuman 1929; Zoch 1934), and even the potential of seasonal rainfall prediction based on snowpack (Monson 1934). Later, *MWR* articles (1960–2000; Systems to Process and Geosciences Eras) became less focused on basic hydrology, partly due to the emergence of hydrology journals including the *Hydrological Sciences Journal* (established in 1956), the *Journal of Hydrology* (established in 1963), and the American Geophysical Union’s (AGU) *Water Resources Research* (*WRR*; established in March 1965). Another factor in the change in hydrology focus for *MWR* articles in this period is related to the establishment of the *Journal of Applied Meteorology* (*JAM*) in 1962, along with the emergence of hydrometeorology and hydroclimatology, which culminated with the introduction of the *Journal of Hydrometeorology* (*JHM*) in 2000. Classic examples from *MWR* include Rasmusson’s (1967, 1968) water vapor budgets, Lettau’s (1969) evaporation climatology, Manabe’s (1969) bucket model,

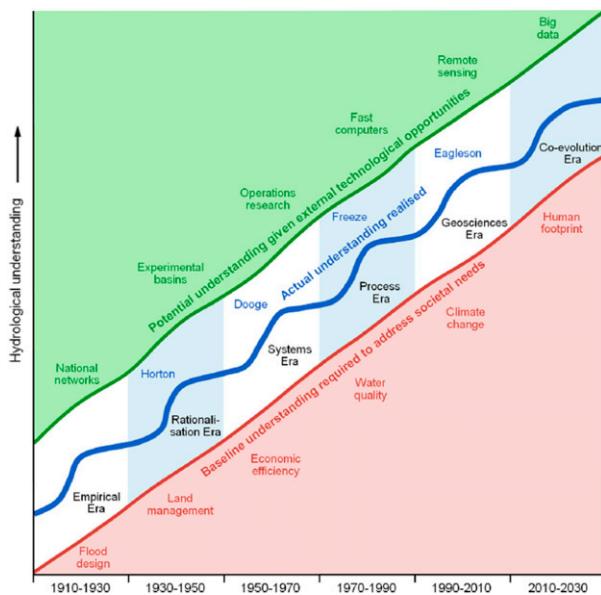


FIG. 25-2. Staircase growth of hydrological understanding sandwiched between baseline understanding required to address societal needs and potential understanding given external technological opportunities. Names in blue epitomize dominant paradigms of the eras. [From Sivapalan and Blöschl (2017).]

and Priestley and Taylor’s (1972) evaporation formulation. Examples from *JAM* include the albedo model of Idso et al. (1975), Adler and Negri’s (1988) infrared rainfall technique, Nemani et al.’s (1989) satellite-based surface resistance methodology, Dorman and Sellers’s (1989) Simple Biosphere Model parameter climatologies, Beljaars and Holtlag’s (1991) land surface flux parameterization, Daly et al.’s (1994) topography-based precipitation analysis methodology, and Hsu et al.’s (1997) neural network based satellite precipitation estimation.

An article in *WRR*’s recent fiftieth anniversary issue by Rajaram et al. (2015) identified the topics covered by the top 10 most highly cited papers of each decade since 1965. These topics mirrored the evolution of topics in *MWR* and *JAM*, from infiltration and evapotranspiration formulations to land surface hydroclimatological models and data assimilation. In addition to the evolution of scientific topics, progress in hydrology during this period is marked by the establishment of hydrology as a science rather than an “application” (e.g., Klemeš 1988). The so-called Eagleson “Blue Book” (National Research Council 1991; Eagleson 1991) was a bellwether moment in hydrology because it helped define hydrology as a distinct geoscience and recommended the establishment of research and educational programs in hydrology, hence the so-called Geosciences Era from 1990 to 2010 (Sivapalan and Blöschl 2017).

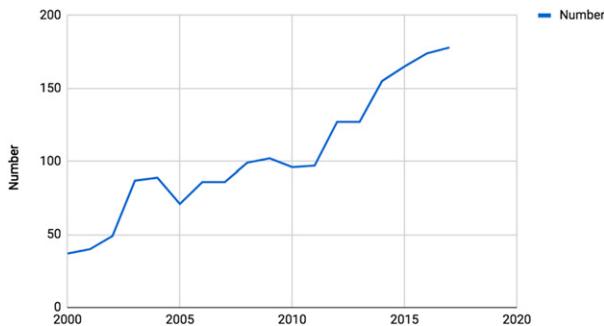


FIG. 25-3. Number of publications per year in the *Journal of Hydrometeorology* since inception in 2000.

Since *JHM* was initiated in 2000, the growth and impact of hydrologic research both within the AMS community (e.g., Fig. 25-3) and overall (Clark and Hanson 2017) has been substantial. In this monograph chapter, we will review progress in hydrology for the last 100 years, including theory, observations, and forecasting. Major themes such as the emergence of global hydrology, coupled land–atmosphere modeling including hydrometeorology and hydroclimatology, dynamical hydrologic prediction, and water resources management and water security will be reviewed. Finally, we look forward with a discussion of future directions.

2. The evolution of hydrologic understanding

As noted in chapter 1 of the *Handbook of Hydrology* (Maidment 1993), quantitative hydrology emerged in the 1850s with Mulvaney’s (1851) time of concentration concept and Darcy’s (1856) law of groundwater flow. Surface water flow equations developed shortly thereafter by Barré de Saint-Venant (1871) and Manning (1891) underpin today’s routing schemes. Infiltration models from Green and Ampt (1911) to Horton (1933) provide a physical basis for rainfall–runoff modeling that further advanced hydrologic science. In this section, we focus on advances in hydrologic theory over the past 100 years in six key areas: 1) precipitation, 2) evaporation, 3) infiltration and soil water movement, 4) groundwater, 5) streamflow and routing, and 6) hydrogeomorphology.

a. Precipitation

In the previous section, we described widely used concepts such as precipitation intensity–duration–frequency curves for engineering design for structures such as roadways, sewers, and dams. Accordingly, from the hydrology perspective, key theoretical developments with respect to precipitation were initially focused on estimating precipitation extremes, including statistical techniques and the design of rainfall measurement networks.

The reader is referred to chapter 17 of this monograph (Houze 2019) for a more detailed treatment of advances in observing and modeling mesoscale precipitation processes. In this section, we focus on advances in precipitation estimation for hydrometeorological and hydroclimatological applications.

A fundamental concept in the estimation of rainfall intensities for high-hazard structures such as dams is the probable maximum precipitation (PMP; Hershfield 1961b; WMO 2009), which is defined as “theoretically, the greatest depth of precipitation for a given duration that is physically possible over a given size storm area at a particular geographical location at a certain time of the year” (WMO 2009, p. 1). The general procedure for calculating PMP relies on estimating maximum precipitable water, convergence rate, and vertical motion, and there are numerous reports available produced by the NWS from 1963 to 1999 that provide PMP estimates for the United States (<http://www.nws.noaa.gov/oh/hdsc/studies/pmp.html>). Unfortunately, these efforts have been discontinued due to lack of funding. Despite this, PMP estimation is still an active area of research, and recent publications suggest that PMPs are changing along with climate (Kunkel et al. 2013; Chen et al. 2017).

Around the same time, the design of precipitation networks became a focus, including Dawdy and Langbein (1960), Peck and Brown (1962), and Peck (1980), who recognized the role of topography in areal precipitation estimation. Rodríguez-Iturbe and Mejía (1974a,b) further studied point–area relationships and rainfall network design. These works ultimately led to the highly cited work of Daly et al. (1994), who use observed precipitation and temperature gradients with topography to provide gridded precipitation products for the contiguous United States (CONUS).

Beyond these studies, major advances in the simulation of precipitation time series as spatially correlated random fields led to further advancements in stochastic hydrology. Understanding the space–time structure of precipitation allowed hydrologists to simulate precipitation over areas of interest, such as watersheds or river basins. Examples include Huff and Changnon (1964), Amorocho and Wu (1977), Waymire et al. (1984), Georgakakos and Bras (1984), Eagleson et al. (1987), Foufoula-Georgiou and Lettenmaier (1987), Sivapalan and Wood (1987), and Foufoula-Georgiou (1989). This work complemented studies that identified the fractal and ultimately multiscaling structure of precipitation (e.g., Zawadzki 1987).

b. Evapotranspiration

Evapotranspiration (ET) represents the combination of open water, bare soil, and canopy surface evaporation

and transpiration. Theoretically, ET represents a turbulent flux of water vapor from Earth's surface to the atmosphere resulting from the phase change of liquid water. This phase change means that ET is coupled to the surface energy balance via the latent heat of vaporization, and therefore the transfer of energy from the surface to the atmosphere due to evapotranspiration is also referred to as the latent heat flux. If the phase change is from solid to vapor, then this energy transfer must also include the latent heat of sublimation. ET can be estimated as the product of the kinematic turbulent flux of water vapor ($w'q'$), and the density of air. This is the concept behind the eddy covariance measurement technique, as will be discussed further in [section 3f](#) below.

Given that ET is a dominant term in the terrestrial water budget, there have been many efforts designed to estimate this term with limited meteorological information. An important concept in ET estimation is that of potential evaporation (PE), which is defined as “the quantity of water evaporated per unit area, per unit time from an idealized, extensive free water surface under existing atmospheric conditions” ([Maidment 1993](#), p. 4.2). This concept effectively represents atmospheric “evaporative demand.” A closely related concept is that of reference crop ET (denoted E_{To}), which is the rate of ET from an idealized grass crop with a fixed height, albedo, and surface resistance. To obtain actual ET from potential or reference ET, PE or E_{To} are multiplied by a series of coefficients representing specific crops and stress factors ([Allen et al. 1998](#); [Doorenbos and Pruitt 1977](#)).

In general, the methods for PE estimation can be classified as 1) temperature-dependent methods or 2) combination methods. The original, and still widely used, temperature-dependent method was developed by [Thornthwaite \(1948\)](#). This method has been shown by numerous authors to overestimate the sensitivity to air temperature changes (e.g., [Milly and Dunne 2016](#)), and alternative temperature-based methods have been shown to behave more like combination methods (e.g., [Blaney and Criddle 1950](#); [Hargreaves and Samani 1985](#)).

The foundation of combination methods is [Penman \(1948\)](#), who combined the energy balance approach with an aerodynamic approach to derive an estimate of E_{To} based on an implicit assumption of measurement height and roughness length ([Thom and Oliver 1977](#)). [Monteith \(1965\)](#) generalized the Penman equation by calculating leaf resistances in series with the aerodynamic resistance employed by Penman. This led to a generalized form of the reference crop E_{To} equation now known as the Penman–Monteith equation. [Priestley and Taylor \(1972\)](#) found that E_{To} could be approximated quite

accurately using a simplified form of Penman's equation requiring only the available energy (= net radiation minus ground heat flux) and a coefficient that changes depending on humid or arid climates, as defined by relative humidity.

The most general form of the Penman–Monteith equation utilizes both aerodynamic and canopy resistances in series, where the canopy resistance (or its inverse, the conductance) can be calculated using a [Jarvis \(1976\)](#) approach, which depends on both leaf area and soil moisture (e.g., [Lhomme et al. 1998](#)). A comprehensive summary of ET theory and methods is given in [Brutsaert \(1982\)](#) and [Dolman \(2005\)](#).

Modern approaches to estimating actual ET fall into three categories: energy balance (e.g., [Su et al. 2005](#); [Anderson et al. 1997, 2011](#)), combination [e.g., Penman–Monteith or Shuttleworth–Wallace ([Shuttleworth and Wallace 1985](#))] and complementary approaches (e.g., [Bouchet 1963](#)), or combinations thereof (e.g., [Mallick et al. 2013](#)), and the choice and performance depends primarily on the availability of required data ([Mueller et al. 2013](#)). Most modern land surface models used in climate models (as will be discussed in [section 6](#) below) calculate ET using a Jarvis-based energy balance approach or a coupled photosynthesis–canopy conductance energy balance approach ([Ball et al. 1987](#)), as discussed in the review by [Wang and Dickinson \(2012\)](#).

c. Infiltration and soil water movement

In the Empirical Era, infiltration estimation was primarily focused on estimating losses for runoff, and this led to the work of [Green and Ampt \(1911\)](#), which was later shown to be consistent with theory and observations, with some updates to account for antecedent moisture conditions ([Mein and Larson 1973](#)). It was also shown to be an expression of the time condensation approximation (TCA; [Sivapalan and Milly 1989](#)). [Horton \(1933\)](#) further investigated infiltration and found that a time-varying infiltration capacity can be used to better estimate infiltration excess runoff than the so-called “rational method.” In Horton's work, estimating “effective rainfall,” which is the infiltrated water available to plants, was another major objective. [Philip \(1957\)](#) derived a series solution for infiltration into a vertical, semi-infinite homogeneous soil surface, and since that time, this approach has largely been replaced by the numerical solution of the governing equation for soil water movement, as described below.

[Richards \(1931\)](#) derived an equation for capillary conduction of liquids through porous mediums, which combines Darcy's law applied to unsaturated media with the continuity equation. The “Richards equation” is the foundation for predicting both infiltration and soil water

movement, and soil moisture specifically. There are a number of works geared toward the proper form of this equation for numerical solutions (e.g., [Celia et al. 1990](#); [Zeng and Decker 2009](#)), in addition to various functional forms for the required soil water characteristic curves, which are nonlinear, hysteretic, and a function of soil texture (e.g., [Brooks and Corey 1964](#); [Clapp and Hornberger 1978](#); [Rawls et al. 1982](#); [van Genuchten 1980](#)). The Richards equation is the basis for many of today's hydrological and land surface models (e.g., [Downer and Ogden 2004](#); [Lawrence et al. 2011](#)), although there are still challenges to its application and solution ([Farthing and Ogden 2017](#)).

d. Groundwater

As with other aspects of the hydrological cycle, early efforts to characterize groundwater focused on mapping groundwater resources. The U.S. Geological Survey (USGS) was a pioneer in this area, and the foundational work of [Meinzer \(1923\)](#) defined groundwater provinces as well as the basic principles of groundwater occurrence and movement. As with soil water movement, groundwater flow in saturated porous media is governed by [Darcy's \(1856\)](#) law. Combining this constitutive relation with the continuity equation led to key theoretical developments in describing groundwater motion, for example, [Theis \(1935\)](#), [Hubbert \(1940\)](#), [Jacob \(1940\)](#), and [Hantush and Jacob \(1955\)](#). Texts such as [Bear \(1972, 125–129\)](#) and [Freeze and Cherry \(1979\)](#) provide more in-depth treatments of the topic. A breakthrough in the recognition of the role of groundwater in runoff generation came from [Dunne et al. \(1975\)](#), who showed that saturated areas intersecting the surface produced instantaneous runoff known as “saturation excess” in contrast to the Hortonian “infiltration excess” runoff produced when rainfall exceeds the infiltration capacity. This led to the observation that these saturated areas could be approximately represented via a topography-based drainage index, which led to the so-called TOPMODEL concept for parameterizing saturated areas ([Beven and Kirkby 1979](#)). Later, [Yeh \(1986\)](#) formalized inverse approaches for identifying parameters for groundwater hydrology that are now commonly used in groundwater modeling.

These theoretical developments were later translated into numerical models, the most prevalent being the USGS MODFLOW ([Trescott et al. 1980](#); [McDonald and Harbaugh 2003](#)). In the last 20 years, there has been a push to include representations of groundwater flow in land surface models, ranging from simple treatments based on the TOPMODEL concept (e.g., [Famiglietti and Wood 1994](#); [Koster et al. 2000a](#); [Niu et al. 2007](#)) to more complex treatments based on groundwater flow models

(e.g., [Maxwell and Miller 2005](#); [Fan et al. 2007](#); [Miguez-Macho et al. 2007](#)).

e. Streamflow and routing

The theory of surface water flow has been well-known since [Barré de Saint-Venant \(1871\)](#) derived the shallow water equations (also known as the Saint-Venant equations) based on the conservation of mass and momentum (i.e., the Navier–Stokes equations). Depending on which terms in the Saint-Venant equations are retained, the equations may be reduced to “kinematic wave” (e.g., [Lighthill and Whitham 1955](#); [Feldman 2000](#); [Getirana et al. 2012](#)) or “diffusion wave” approximations (e.g., [Julien et al. 1995](#)). The full 1D equations are known as “dynamic wave” (e.g., [Brunner 2016](#)) and are computationally intensive to solve, and cost–accuracy tradeoffs remain even for approximations to these equations ([Getirana et al. 2017](#)). The velocity of flow in open channels was described by [Manning \(1891\)](#), who redeveloped an earlier relationship by [Gauckler \(Hager 2001\)](#) in which velocity is related to slope, roughness, and cross-sectional area. Combining the velocity formulation with the Saint-Venant equations allows for the prediction of streamflow in open channels as well as overland flow on hillslopes (e.g., [Chow 1959](#)).

Routing refers to the prediction of changes in the height (stage) and volumetric flow rate (discharge) of water (i.e., the hydrograph) as it moves over a hillslope or through a river channel or a reservoir ([Woolhiser and Liggett 1967](#)). Hydraulic or distributed routing refers to solving both the continuity and momentum equations, while hydrologic or lumped routing refers to the continuity equation alone. One of the first approximate techniques to transform rainfall as a runoff response was [Sherman's \(1932\)](#) unit hydrograph. This technique allows the computation of a hydrograph by convolving the excess rainfall with a response function that could be derived for each watershed (e.g., [Dooge 1959](#)). This approach is still in use in engineering hydrology (e.g., [Feldman 2000](#)), although the most prevalent hydrologic routing technique is the Muskingum ([McCarthy 1940](#)) or its extension, the Muskingum–Cunge ([Cunge 1969](#); [Todini 2007](#)) method.

In addition to overland flow routing described above, there is also the concept of subsurface flow routing, which is an approximation to unsaturated and saturated groundwater flow (as described above). As discussed by [Tague and Band \(2001\)](#), the TOPMODEL ([Beven and Kirkby 1979](#)) approach can be described as an “implicit” routing approach in contrast to “explicit” approaches such as [Wigmosta et al. \(1994\)](#). The developing National Water Model, based on the WRF-Hydro system ([Gochis et al. 2015](#)), currently uses a configuration that includes

explicit subsurface routing, diffusive wave surface routing, and Muskingum–Cunge channel routing.

f. *Hydrogeomorphology*

Hydrogeomorphology “focuses on the interaction and linkage of hydrologic processes with landforms or earth materials and the interaction of geomorphic processes with surface and subsurface water in temporal and spatial dimensions” (Sidle and Onda 2004). Eminent hydrologists including Horton, Langbein (Dooge 1996), Freeze and Harlan (1969), Dooge (1973), and Beven and Kirkby (1979) have clearly recognized the linkages between the landscape and hydrologic processes. During the so-called Systems Era (1950–70), the full integration of hydrologic theories and processes was explored and led to the Freeze and Harlan “blueprint” for hydrologic modeling, as discussed recently by Clark et al. (2017). As we progressed to the process era (1970–90), hydrogeomorphological work focused on the search for universal laws (Dooge 1986); the linkages between climate, soil, and vegetation (Eagleson 1978a,b,c,d,e,f,g); and scaling and similarity (Wood et al. 1988; Blöschl and Sivapalan 1995).

As discussed in Peters-Lidard et al. (2017), one outcome of this era is the Representative Elementary Area (REA) concept (Wood et al. 1988; Fan and Bras 1995), which found that the rainfall–runoff process behaved in a much simpler manner at the roughly 1 km² scale. Later extensions to the concept include the Representative Elementary Watershed (REW) introduced by Reggiani et al. (1998, 1999, 2000, 2001) and the Representative Hillslope (RH; Troch et al. 2003; Berne et al. 2005; Hazenberg et al. 2015). The REA/REW approach is conceptually similar to Reynolds averaging and assumes that the physics are known at the smallest scale considered (e.g., Miller and Miller 1956). In another parallel, fluxes at the boundaries of model control volumes require parameterization (i.e., “closure”), with assumptions that are typically ad hoc, and may include subgrid probability distributions, scale-aware parameters, or new flux parameterizations.

Explicit “Newtonian” modeling of hillslopes at “hyper-resolutions” (Wood et al. 2011), or with clustered 2D simulations (e.g., the HydroBlocks of Chaney et al. 2016), may render the REA/REW approach obsolete, although hydrogeomorphological connections continue to be explored. For example, Maxwell and Condon (2016) found that the interplay of water table depths with rooting depths along a given hillslope exerts different controls on evaporation and transpiration, linking the water table dynamics with the land surface energy balance. Further, the concept of catchment coevolution and “Darwinian” hydrology (Sivapalan 2005; McDonnell et al. 2007,

Thompson et al. 2011; Harman and Troch 2014) has extended scale and similarity concepts to synthesize catchments across scales, places, and processes, ushering in the Coevolution Era.

3. Advances in hydrologic observations

Moving from point to areal estimates of hydrologic states and fluxes has revolutionized hydrologic science. Advances in precipitation estimation from gauges to radars to satellites, combined with similar advances in observing snow packs, soil moisture, terrestrial water storage, evapotranspiration, and stream stage and discharge, have enabled continental- and global-scale hydrology. Famiglietti et al. (2015) provide an overview of the advantages of satellite-based observation, including global coverage, near-continuity across space and time, and consistency of measurements from a given instrument. Taking the other side of the argument, Fekete et al. (2015) describe the shortcomings and dangers of overreliance on remote sensing, including errors associated with miscalibration of remote sensing retrieval algorithms, coarse spatial resolution relative to in situ measurements, inconsistencies associated with technology/instrument changes, and termination of long-term in situ measurement locations incorrectly perceived to have been made obsolete by remote sensing.

a. *Precipitation*

Rain gauges, disdrometers, and radars have long been conceptualized as the data reference for precipitation studies. The use of standardized rain gauges has been documented as early as the fifteenth century in Korea, so it is not surprising that for a long time in history they have been considered as indispensable for precipitation science (Tapiador et al. 2011). In fact, rain gauges are still considered the privileged source of reference data for precipitation estimates as they provide a direct physical record of the hydrometeors (cf. Kucera et al. 2013). Most of the global datasets of precipitation, such as the Global Precipitation Climatology Project (GPCP; Fig. 25-4; Adler et al. 2003, 2012, 2017) or the Climate Prediction Center (CPC) Merged Analysis of Precipitation (CMAP; Xie and Arkin 1997; Xie et al. 2003), include them in one way or another, and they are still used to calibrate and tune climate models. The same applies to other climate data records from multisatellite observations (Ashouri et al. 2015). Disdrometers—modern instruments that estimate not only the total precipitation but also the drop size distribution (DSD)—are also becoming an increasingly important part of ground instrumentation systems. Disdrometers are direct in that they respond to individual drops, but they have a fairly small sampling area (tens of

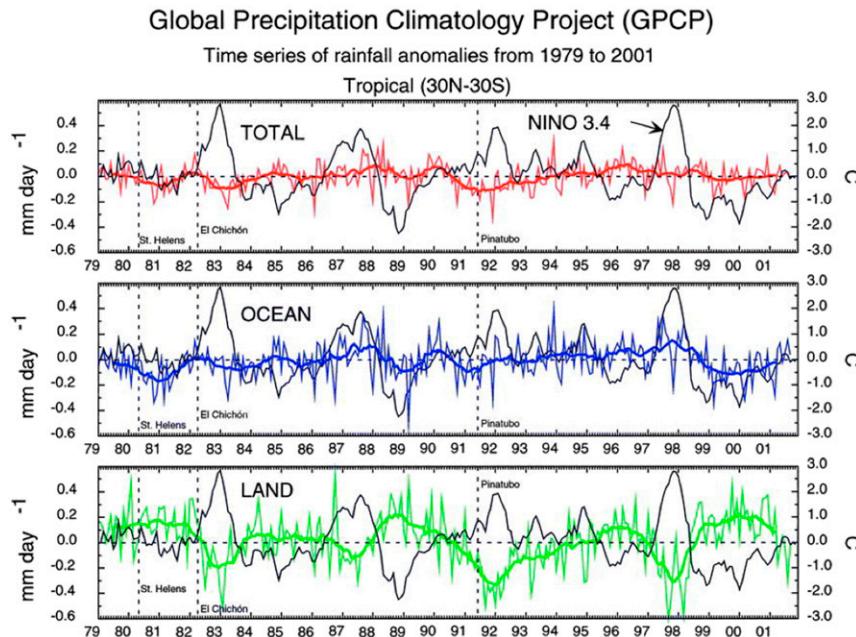


FIG. 25-4. Tropical (30°N–30°S) averages of monthly precipitation anomalies (mm day^{-1}) for (top) total, (middle) ocean, and (bottom) land. Vertical dashed lines indicate the months of significant volcanic eruptions. Black curves in all panels indicate the Niño-3.4 SST index ($^{\circ}\text{C}$). [From Adler et al. (2003).]

square centimeters), which affects the representativeness of the measurements (Tapiador et al. 2017).

In contrast, ground-based radars sample a large volume but provide an estimate of the precipitation based on the backscattered echo, an indirect observation which relates to total rainfall through the DSD. In recent times, exploiting the radar transmit/receive polarization state has enhanced radar capabilities by discriminating the phase of the hydrometeors (Brangi et al. 1990). These three ground observations of precipitation (gauges, disdrometers, and radars) were the primary input for hydrologic models for many years.

However, beginning with the inception of operational passive microwave imagery on board the operational Defense Meteorological Satellite Program (DMSP) low-Earth-orbiting (LEO) satellites in 1987, and continuing with the joint National Aeronautics and Space Administration (NASA)–Japan Aerospace Exploration Agency (JAXA) Tropical Rainfall Measuring Mission (TRMM) in 1997 (Simpson et al. 1988), a shift toward a blended or merged use of both ground- and satellite-based precipitation estimates was initiated (Adler et al. 2000; Huffman et al. 2007; Joyce et al. 2004; Behrangi et al. 2009; Kidd and Huffman 2011). With microwave-based precipitation observations on board multiple LEO satellites, complemented by the fast-refresh capabilities (30 min or less) of operational geostationary

visible and infrared imaging sensors, improved global precipitation was enabled at spatial and temporal scales relevant to hydrological models and applications (Hossain and Lettenmaier 2006; Nguyen et al. 2017, 2018). TRMM demonstrated the first spaceborne precipitation radar (Iguchi et al. 2000) and algorithms that fused passive and active measurements for global precipitation estimation over tropical and subtropical regions (Kummerow et al. 2000). TRMM was key to improving our understanding of the water cycle and the role of latent heat in the Earth system (Tao et al. 2016). The role of TRMM in improving our knowledge of the structure (and thus of the physics) of hurricanes cannot be underestimated (Hawkins and Velden 2011).

More extensive global coverage at higher latitudes (i.e., poleward of TRMM's limited 35° latitude coverage) culminated with the launch of the joint NASA–JAXA Global Precipitation Measurement (GPM) Core Observatory in 2014 (Skofronick-Jackson et al. 2017). GPM's Dual-Frequency Precipitation Radar and higher orbit inclination expands the measurements to more overland regions where much of the water cycle is driven by frozen precipitation processes (Houze et al. 2017).

Regarding the future of precipitation estimation from space, the direction in the United States points to small systems such as CubeSats (Peral et al. 2018), with an

emphasis on understanding the interplay between aerosols and precipitation, aqueous chemistry, and a better understanding of convection. Such topics are those favored by the National Academies Decadal Survey of Earth Sciences and Applications from Space ([National Academies of Sciences, Engineering, and Medicine 2018](#)) and therefore will be likely the drivers of future missions.

b. Snowfall

Quantification of frozen precipitation has been especially difficult in the past. Indeed, instrumentation designed to remotely sense snow water equivalent (SWE) rates, for example, be it via weighing gauge, radar, or disdrometer have all addressed the challenges to deal with maintenance, calibration, point-to-area representativeness, measurement error, wind, etc., in addition to the irregular shapes, sizes, and bulk density of snowfall ([Tapiador et al. 2012](#))

There are ways to retrieve SWE rates over larger areas using combinations of polarimetric radar, disdrometer, and weighing gauge data ([Brandes et al. 2007](#); [Huang et al. 2010](#)), but doing so is tedious and case specific ([Tapiador et al. 2012](#)). The advent of the GPM *Core Observatory*, with its enhanced capabilities over TRMM and the ability to measure the solid phase over the whole planet, has opened a new phase for hydrology.

c. Snowpack

Observations of snowpack properties, such as snow-covered area (SCA)/snow cover extent (SCE), snow depth, and SWE, prove challenging due to the considerable variability at fine spatial scales (e.g., [Blöschl 1999](#); [Erickson et al. 2005](#); [Dozier et al. 2016](#)). Ground-based measurement of SCA is labor intensive, although tower-mounted imaging is still a useful verification technique, and new approaches such as low-cost temperature sensors can be used to monitor seasonal SCA ([Lundquist and Lott 2010](#)). Remote sensing of SCA was among the first applications of satellite data (as discussed in the review by [Lettenmaier et al. 2015](#)), and today, the spatial and temporal evolution of SCA is monitored with multiple satellites, including 30-m-resolution Landsat ([Rosenthal and Dozier 1996](#)), 500-m-resolution Moderate Resolution Imaging Spectroradiometer (MODIS; [Hall et al. 2002](#); [Painter et al. 2009](#)), and 1000-m-resolution Advanced Very High Resolution Radiometer (AVHRR; [Ramsay 1998](#); [Romanov et al. 2000](#)). These datasets have been used to construct a Climate Data Record of Northern Hemisphere SCA from 1966 to present ([Estilow et al. 2015](#)).

Ground-based observations of snow depth can be obtained through intensive depth probe sampling

([Elder et al. 1991](#)), photographs ([Tappeiner et al. 2001](#); [König and Sturm 1998](#)), and snow pits (e.g., [Cline et al. 2004](#)) along with ground-based and airborne radar ([Machguth et al. 2006](#)) and lidar ([Deems et al. 2013](#)). Ground-based measurement of SWE is done through snow pillows with pressure transducers ([Beaumont 1965](#)). Routine airborne SWE monitoring is conducted over CONUS using gamma ray sensing ([Carroll and Carroll 1989](#); [Carroll 2001](#)). Satellite-based monitoring of snow depth has been demonstrated with passive microwave sensors ([Chang et al. 1987](#); [Kelly et al. 2003](#); [Kelly 2009](#)), despite issues of signal saturation for SWE > 200 mm and loss of signal in heavily forested areas ([Vuyovich et al. 2014](#)). To overcome some of the limitations of the passive microwave approach, the community generally favors an integrated approach among multiple satellite sensors ([Frei et al. 2012](#)). Alternatives that have been shown to work well in mountainous regions such as lidar ([Painter et al. 2016](#)) and Ka- and Ku-band radar ([Hedrick et al. 2015](#); [Liao et al. 2016](#)) are recommended in the most recent National Academies Decadal Survey of Earth Sciences and Applications from Space ([National Academies of Sciences, Engineering, and Medicine 2018](#)).

d. Soil moisture

Soil moisture is significant in its roles as an atmospheric lower-boundary condition; a regulator of near-surface temperature, evapotranspiration, and photosynthesis; an influence on flash flooding and surface runoff; an indicator of wetness conditions and drought; and the water available to shallow-rooted plants. As discussed in [Walker et al. \(2004\)](#), [Robinson et al. \(2008\)](#), and [Peng et al. \(2017\)](#), there are many techniques to measure soil moisture with ground instruments, including gravimetric methods (e.g., [Vinnikov and Yeserkepova 1991](#)), time domain reflectometry (e.g., [Robinson et al. 2003](#)), capacitance sensors (e.g., [Bogena et al. 2007](#)), neutron probes (e.g., [Hollinger and Isard 1994](#)), electrical resistivity measurements (e.g., [Samouëlian et al. 2005](#)), heat pulse sensors (e.g., [Valente et al. 2006](#)), and fiber optic sensors (e.g., [Garrido et al. 1999](#)). One of the first efforts to compile ground-based soil moisture into a global database was [Robock et al. \(2000\)](#). More recently, this effort has been expanded by [Dorigo et al. \(2011\)](#), and over the United States there is a comprehensive data collection effort described by [Quiring et al. \(2016\)](#). As described by [Crow et al. \(2012\)](#), several in situ networks have been established and expanded with the goal to evaluate satellite-based soil moisture products. While these networks are primarily focused on in situ sensors, an exciting development for measuring intermediate-scale soil moisture is via cosmic-ray neutrons ([Zreda et al. 2008](#)).

This has led to the development of the Cosmic-Ray Soil Moisture Observing System (COSMOS) soil moisture network, which has expanded from the United States to provide some international coverage.

Remote sensing of soil moisture is possible because soil moisture changes the surface emissivity and backscattering properties in microwave frequencies (Schmugge et al. 1974; Njoku and Kong 1977; Dobson et al. 1985). Surface roughness, vegetation, and the presence of rainfall confound the retrieval of soil moisture. Both active and passive microwaves respond to soil moisture signals from a shallow surface layer, with the 1.4 GHz (L band) penetrating to about 5 cm, and less at higher frequencies. Soil moisture products have been generated from numerous sensors, from the Scanning Multichannel Microwave Radiometer (SMMR) sensor on board Nimbus (launched in 1978) to the most recent European Space Agency (ESA) Soil Moisture Ocean Salinity (SMOS) mission (Kerr et al. 2016) launched in 2009 and the NASA Soil Moisture Active and Passive (SMAP) mission (Entekhabi et al. 2010) launched in 2015. Retrieval algorithms vary from the SMAP single- and dual-channel algorithms (O'Neill et al. 2015) and Land Parameter Retrieval Model (LPRM; Owe et al. 2001, 2008) to the L-band Microwave Emission of the Biosphere (L-MEB) model (Wigneron et al. 2003) of SMOS. Radiometers and scatterometers have been shown to provide highly correlated and somewhat complementary information (de Jeu et al. 2008), and radiometer algorithm performance has been shown to depend strongly on vegetation and roughness parameterizations (Mladenova et al. 2014). Recent attention has focused on providing both long-time-series soil moisture data records (Dorigo et al. 2017) and fine-spatial-scale soil moisture information (Peng et al. 2017) from multiple platforms using a variety of approaches. These approaches are increasing the relevance of remotely sensed soil moisture for drought assessment, agriculture, and vadose-zone hydrology (Mohanty et al. 2017).

e. Terrestrial water storage

Terrestrial water storage refers to the all the water stored in and on a column of land, that is, the sum of groundwater, soil moisture, surface waters, snow, ice, and wet biomass. It is the freshwater that enables life on land. It is also one of the four terms in the terrestrial water budget, that is,

$$dS = P - ET - Q, \quad (1)$$

where dS is the change in terrestrial water storage, P is precipitation, ET is evapotranspiration, and Q is runoff from a given study area. Defining the study area as a

watershed, river basin, or other closed hydrologic unit facilitates the application of Eq. (1). Of the four terms, dS and ET are the most difficult to measure, and over the years many researchers have chosen to assume that dS is negligible in order to close the water budget or to infer ET —an increasingly dubious assumption as the study period shortens. Mintz and Serafini (1992) recognized the importance of dS and were among the first to account for it in a water budget analysis. Nevertheless, due to the difficulty of obtaining coincident measurements of all the terrestrial water storage components, only a few studies paid serious attention to the left side of Eq. (1). In particular, Yeh et al. (1998) and Rodell and Famiglietti (2001) assessed variations in terrestrial water storage using in situ observations of groundwater, soil moisture, snow depth, and reservoir storage from Illinois. They showed that the interannual variations in groundwater and terrestrial water storage in Illinois were of the same order of magnitude as those of soil moisture, thereby casting doubt on studies that relied on the assumption of dS equaling zero over the course of an annual cycle. Despite advances in hydrogeophysics (Binley et al. 2015), direct measurement of this term has been most successful from the vantage point of space.

Our ability to measure terrestrial water storage changes from space is somewhat serendipitous. In the 1980s and 1990s, geodesists, who measure the precise shape of the Earth and its gravity field, had been searching for a way to improve space-based gravity measurement. The first satellite dedicated to this purpose, the Laser Geodynamics Satellite (LAGEOS), looked like a silver-colored ball; it was an uninstrumented, mirrored sphere used for laser ranging from Earth's surface. By measuring departures from its predicted orbit, the geodesists could map irregularities in Earth's static gravity field. Yoder et al. (1983) inferred that LAGEOS and other satellites were also sensitive to temporal variations in the gravity field caused by atmosphere and ocean circulations and the redistribution of water on the land. Geodesists proposed that in order to measure those orbital departures with enough precision to quantify mass changes with a useful degree of accuracy, the measurements would have to be made by another, co-orbiting satellite (Dickey et al. 1997). Thus was born the Gravity Recovery and Climate Experiment (GRACE) satellite mission, developed jointly by NASA and two German agencies. GRACE, which launched in 2002 and continued operating until 2017, used a K-band microwave system to measure the distance (~ 100 – 200 km) between two identical, co-orbiting satellites, with micron-level precision. Over the course of each month, these measurements were used to produce a new map of Earth's gravity field that was accurate enough that month to month changes in

terrestrial water storage could be estimated, after using ground-based measurements and models to remove the effects of atmospheric and oceanic mass changes (Wahr et al. 1998; Tapley et al. 2004).

Hydrologists were slow to embrace GRACE due to its data being much different from anything they had previously seen. In particular, relative to other remote sensing measurements, the spatial resolution of GRACE is extremely low—on the order of 150 000 km² at mid-latitudes (Rowlands et al. 2005; Swenson et al. 2006). GRACE only provided month-to-month changes in terrestrial water storage with a multimonth time lag, whereas most remote sensing systems provide instantaneous observations with latencies that range from a few hours to a week. Finally, GRACE provides no information on the vertical distribution of the observed terrestrial water storage changes, leaving it to hydrologists armed with auxiliary data and models to determine how much each of the water storage components contributed to an observed change. As a result of these unusual characteristics, GRACE data have been misinterpreted and misused by researchers in numerous instances, which has caused others to dismiss GRACE entirely (e.g., Darama 2014; Alley and Konikow 2015; Sahoo et al. 2016).

Nevertheless, because satellite gravimetry is currently the only remote sensing technology able to discern changes in water stored below the first few centimeters of Earth's surface, GRACE caused a revolution in water budget hydrology. Among the highlights, GRACE enabled closure of the terrestrial water budget and estimation of evapotranspiration as a residual (Rodell et al. 2004; Swenson and Wahr 2006; Rodell et al. 2011), determination of ice sheet mass changes (Luthcke et al. 2006; Velicogna and Wahr 2005, 2006), ablation of major glacier systems (Tamisiea et al. 2005; Chen et al. 2007; Luthcke et al. 2008), estimation of groundwater storage changes (Rodell et al. 2007) and trends (Rodell et al. 2009; Richey et al. 2015), enhanced drought monitoring (Houborg et al. 2012; Thomas et al. 2014), effects of terrestrial water storage changes on sea level (Boening et al. 2012; Reager et al. 2016), and improved understanding of seasonal and interannual variability of and human impacts on the global, terrestrial water cycle (van Dijk et al. 2014; Humphrey et al. 2016; Rodell et al. 2018). By 2010 the importance of GRACE to multiple disciplines became so clear that NASA promoted the GRACE Follow-On mission ahead of several other missions which had been recommended by the 2007 Decadal Survey in Earth Sciences (National Research Council 2007). GRACE Follow-On launched successfully on 22 May 2018. While it will provide only a small improvement in resolution and accuracy over GRACE,

it will ensure continuity of the terrestrial water storage data record.

f. Evapotranspiration

Measuring actual ET directly is difficult, and as discussed above, reviews of theory and methods can be found in Brutsaert (1982) and Dolman (2005). The original measurement techniques included evaporation pans and weighing lysimeters, and more recent approaches include energy balance/Bowen ratio, eddy covariance, sap flow, isotopes, and fluorescence (e.g., Wilson et al. 2001; Shuttleworth 2007). For a handy reference summarizing these techniques, see Shuttleworth (2008). Networks of flux stations (FLUXNET; Baldocchi et al. 2001) are now prevalent, and approaches to post-process these observations to address the closure problem (e.g., Twine et al. 2000) are required for comparisons among different approaches. Further, because the spatial footprint or “fetch” of these ground-based measurements is limited, empirical “upscaling” approaches (e.g., Jung et al. 2009) are required to produce gridded observations for global hydroclimatological analysis (e.g., Jung et al. 2010).

Remote sensing of ET is based on a number of techniques that depend on temperature, as summarized by Kalma et al. (2008). Popular variants of these techniques include Surface Energy Balance Algorithm for Land (SEBAL; Bastiaanssen et al. 1998), Surface Energy Balance System (SEBS; Su 2002), Mapping Evapotranspiration with Internalized Calibration (METRIC; Allen et al. 2007), Simplified Surface Energy Balance (SSEB; Senay et al. 2011), and Atmosphere–Land Exchange Inverse (ALEXI; Anderson et al. 2011). Other remote sensing–based methodologies augment the temperature signal with weather model data and vegetation information from MODIS to apply the Penman–Monteith approach (e.g., Mu et al. 2011), or augment the temperature signal with other remote sensing data, soil water, interception, and stress accounting and apply a Priestley–Taylor approach [e.g., Global Land Evaporation Amsterdam Model (GLEAM); Miralles et al. 2011; Martens et al. 2017]. Vinukollu et al. (2011) provided the first ever moderate-resolution estimates of ET on a global scale using only remote sensing–based inputs. The recent Global Energy and Water Exchanges (GEWEX) Land-Flux project evaluated multiple global ET products, ranging from models to remotely sensed, and found that no single approach outperformed the other (e.g., McCabe et al. 2016).

g. Surface water stage and discharge

In traditional hydrology, discharge (streamflow) at a stream location is considered a functional integration of

the various surface and subsurface hydrologic processes that occur in the area draining through that point. Measurement of this hydrologic variable forms the cornerstone of calibration and validation of hydrologic models, development of flow-routing schemes, and assessment of flow forecasting skill. In more recent times, as process-based understanding improved in hydrology, the importance of streamflow as a relatively easy-to-measure descriptor (using stage discharge relationship) only increased further. This is because streamflow can also exhibit signatures of the mechanistic role played by land cover change (deforestation), land surface interactions, surface-groundwater interactions, climate and weather change, drought and fluvial processes, and of course water management (see [section 5](#)).

Governments of the developed world have built extensive networks of river and stream gauges to track the flow ([Hannah et al. 2011](#)). For example, the Global Runoff Data Centre (GRDC) archives data from more than 9200 gauges worldwide, and the USGS collects real-time streamflow data at nearly 10 000 locations within the United States. Such streamflow data records have played a vital role in the continued development of hydrologic, land surface, climate, and Earth system models today.

A key limitation of conventional streamflow monitoring using stage discharge relationship is twofold. First, such networks do not exist in developing countries of the world, where most often there are no records of streamflow in rivers along the reach. Even if records were maintained, they are usually not shared openly, as is the norm in developed countries ([Hossain et al. 2014](#)). Second, point-based stage monitoring of discharge is able to only capture the flow that passes through a one-dimensional point. Such measurements do not capture the two-dimensional exchange of water mass that can occur laterally, especially in wetlands, floodplains, and braided rivers. In response to the first limitation, the remote sensing community has been using satellite radar altimeters that were originally designed for ocean monitoring to build a steady record of river heights ([Birkett 1998](#); [Gao et al. 2012](#)) and lake and reservoir elevations ([Birkett et al. 2011](#)). For both limitations, the scientific community is now eagerly awaiting a proposed satellite mission, the Surface Water Ocean Topography (SWOT; [Alsdorf et al. 2007](#)), scheduled for launch in 2021. The SWOT mission will provide the community with a more spatially distributed estimate of the flow in water bodies. With its global sampling from space of water elevation, its temporal change, and its spatial slope in fluvial environments, as well as across lakes, reservoirs, wetlands, and floodplains, ([Biancamaria et al. 2016](#)), SWOT measurements are expected to

contribute to the further understanding of global-scale hydrology.

4. Hydrologic prediction across scales

a. Background

Hydrologic forecasting is one of the earliest applications of hydrologic theory, with deep roots in engineering, and particularly civil engineering in support of water systems management ([Anderson and Burt 1985](#)). The need to anticipate and respond to hydrologic variability and the associated impacts on water uses and hazards has motivated the practice of hydrologic forecasting for much of the last century, and it is now a valuable part of operational services in most of the world's nations ([Emerton et al. 2016](#); [Adams and Pagano 2016](#)). Hydrologic forecasting applications most often target time scales from minutes to seasons and focus not only on extreme events (droughts and floods), but also the entire spectrum of hydrologic variation in which even moderate departures from normal can affect water operations, management, and planning across a broad range of sectors. Over the last century, scientific and technological advances have driven a steady growth in forecast capabilities, culminating in a modern landscape of forecasting in which the advent of high-performance computing, broadband connectivity, and global high-resolution geophysical datasets are transforming long-held traditional paradigms of operational prediction.

The concept of seamlessness (of models, data, methods, and information products across space and time) is commonly touted as a development objective for hydrological forecasting systems as well as weather and climate forecasting systems ([Wetterhall and Di Giuseppe 2018](#); [Hoskins 2013](#)). Yet, for hydrologic forecasting, the strong foundation of engineering pragmatism coupled with limitations in methods, tools, data, and scientific understanding have produced a fragmented, rapidly evolving variety of approaches and operational practices. Forecast products and services have traditionally been distinct along lead time scales and space scales—from the localized minutes to hours ahead phenomenon of flash flooding, to river stage and flow prediction at the river basin scale over periods of hours to days, to long-range seasonal runoff and drought prediction that may span multiple river basins and focus on time-space-averaged predictands (e.g., seasonal snowmelt runoff). The dominant hydrologic processes and the applicability of data, models, and methods in each area are different, and hence the operational pathways toward harnessing predictability have also differed, leading to multiple views on forecasting

development strategies. For example, despite substantial progress in groundwater modeling (as noted above in [section 2d](#)) and understanding interactions between surface water and groundwater ([Brunner et al. 2017](#)), groundwater-level short- to medium-range forecasting relies to a greater extent on statistical and machine-learning techniques (e.g., [Daliakopoulos et al. 2005](#)) than on process-based modeling (e.g., [Prudhomme et al. 2017](#)). Most surface-water hydrologic forecasting efforts supporting water management include only simplified representations of groundwater, if any.

Efforts to develop a spatially and temporally seamless paradigm over the last 15 years have yielded a proliferation of medium- to high-resolution applications of hydrologic models for continental to global-extent prediction—a trend resulting from the ready availability of high-performance computing (HPC) resources, the ease of sharing models and methods, and the accessibility of continental and global meteorological, hydrological, and extensive geophysical attributes datasets, including those derived from satellite-based remote sensing. This development has paralleled the migration of hydrology as a discipline taught in engineering schools toward an application of the geosciences and a reframing of hydrologic forecasting from an engineering practice supporting water resources management toward the view of prediction as an Earth science grand challenge. Today the fit-for-purpose traditional forecasting practice and the fledgling Earth science prediction science coexist, with the latter clearly on the rise. It is therefore timely to review the evolution of hydrologic modeling and forecasting for major phenomena, the better to understand the opportunities and challenges as new forecasting approaches arrive. In this section, we focus primarily on hydrometeorological prediction.

b. Seasonal forecasting

Hydrologic forecasts at seasonal scales (with multi-month lead times and predictand durations) predate the use of computers in hydrology (which began in the 1960s) by decades. The earliest seasonal operational forecasts, which are still today among the most economically valuable, were for peak seasonal runoff, as driven by phenomena such as snowmelt or monsoon season rainfall. Statistical methods have long been used to relate estimates of watershed variables, including snowpack and accumulated moisture from rainfall (and in the last 25 years, climate indices and forecasts), to future runoff, often achieving high skill ($r^2 \sim 0.9$). In the United States, the Department of Agriculture's SCS, now NRCS, began the practice in the mid-1930s and was joined in 1944 by the NWS in issuing forecasts ([Pagano et al. 2014](#); [Helms et al. 2008](#)). For most of their history,

seasonal predictions have been probabilistic, providing a range of uncertainty, and currently help to inform water allocation decisions in major reservoir systems.

The development and adoption of computer-based conceptual catchment hydrology models in the late 1960s and early 1970s (e.g., [Burnash et al. 1973](#)) enabled a new “dynamical” approach to seasonal hydrologic forecasting, now called ensemble streamflow prediction (ESP; [Day 1985](#); [Wood et al. 2016b](#)). ESP involves running a real-time continuous hydrologic model simulation to estimate current watershed conditions, and then using these moisture states to initialize forward simulations based on a sample of historical weather sequences to project watershed conditions ahead into the forecast period. Different forms of ESP provide the central method used operationally in countries around the world, and ESP complements statistical volume forecast techniques by providing ensembles of streamflow sequences, often at the daily or subdaily time step used in the hydrology model, from which a range of predicted variables of interest can be calculated (such as daily peak flow magnitude, timing, and probabilities). Such streamflow sequences are commonly used to estimate runoff volume probabilities or input directly into reservoir operations models to calculate probabilities of future system states and outputs, given specific release policies (e.g., [Kistenmacher and Georgakakos 2015](#)).

In the early 2000s, ensemble seasonal hydrologic forecasting began to attract attention in the land surface modeling research community ([Wood et al. 2002](#); [Luo and Wood 2008](#); [Wood and Lettenmaier 2006](#)), leading to the development of land surface model (LSM)-based national-, continental-, and global-scale forecasting systems using ESP in addition to alternatives involving downscaled climate forecast model outputs ([Duan et al. 2018](#)). These large-domain LSM-based seasonal forecast systems, however, often lack several critical elements that increase skill in the traditional, operational lumped-model seasonal forecast issued from regional or small national centers. In particular, such centers perform comprehensive model calibration as well as data assimilation (mostly manual) to reduce and correct initial hydrologic state errors and thereby improve forecasts.

The practice of using parameter estimation and optimization to support the use of low-dimensional, agile conceptual models in gauged basins and for streamflow forecasting has spurred extensive research and practical successes, yielding widely used techniques such as the shuffled complex evolution (SCE; [Duan et al. 1992](#)) method, and leading to innovative multimethod packages (e.g., [Mattot 2017](#)) and theoretical advancement toward joint parameter-state estimation approaches (e.g., [Vrugt et al. 2006](#)). Research into hydrologic

model data assimilation has also provided a broad range of promising variational and ensemble techniques that have been shown in watershed-scale research applications to benefit not only simulation but also forecasting (Liu et al. 2012). These parameter estimation and DA practices remain less developed and effective for large-domain, regional- to national-scale LSM-based hydrologic prediction systems. This is due in large part to the inability of such methods in large-domain applications to rely directly on watershed observations such as gauged streamflow versus on remotely sensed hydrologic variables such as soil moisture or snow cover. Regional parameters may be estimated through schemes that either assign or calibrate relationships between terrain attributes and model parameters (e.g., Samaniego et al. 2010), or leverage similarity concepts or proximity to transfer parameter values estimated in well-gauged locations to ungauged locations (Wagener and Wheater 2006). Both techniques are critically important to the success of large-domain hydrologic forecasting and are active areas of current research.

The heightened profile of seasonal hydrologic prediction in a research context has also spurred a new research thrust into quantifying seasonal hydrologic predictability and frameworks for the attribution of prediction skill to sources including initial hydrologic conditions (IHCs; primarily soil moisture and snow water equivalent) and future meteorological variability (Wood and Lettenmaier 2008; Paiva et al. 2012; Mahanama et al. 2012; Wood et al. 2016a). The strong role of IHCs in seasonal forecast skill argues that continued development of watershed modeling, observational monitoring, and data assimilation is critically important. To enhance skill at longer lead times and for seasons in which the role of climate variability is large, however, subseasonal to seasonal climate forecasting inputs for hydrologic prediction offer a compelling area of investigation (see the *Hydrology and Earth System Sciences* special issue “Sub-Seasonal to Seasonal Hydrological Forecasting,” https://www.hydrol-earth-syst-sci.net/special_issue824.html).

c. River flood forecasting

Modern river flood forecasting traces back to the development and application of computer-based watershed models in the late 1960s and 1970s (e.g., Crawford and Linsley 1966), and the techniques employed reflect the engineering heritage as well as limitations of the early decades of computer use. The capability began with discrete event forecasting based on relatively simple empirical or statistical models, the earliest of which were essentially graphical in nature—capturing observable relationships between rainfall, soil wetness, and future

runoff for short lead times. Early techniques such as the use of antecedent precipitation index (API) models (Fedora and Beschta 1989; O’Connell and Clarke 1981) still exist operationally in parts of the world today, but have for decades been superseded by forecasting based on continuous watershed models.

The same models and software systems that brought dynamical (model based) methods to the seasonal forecast context were among those used in the continuous river flood context, including the NWS River Forecast System (Anderson 1972), which centered on conceptual snow and soil moisture accounting models within a system providing a broad array of analytical and interactive techniques, for example, for model calibration, state updating, and postprocessing. Some national-scale operational forecasting capabilities in the United States and elsewhere (Pechlivanidis et al. 2014; Emerton et al. 2016) still rely heavily on such continuous conceptual model-based approaches, run in regional or national river forecasting centers. The most common application of these models remains relatively low-dimensional, favoring lumped discretization of small watershed areas (on the order of 10–500 km²), enabling manual effort to be applied for calibrating models and for updating their inputs, states, and outputs in real-time during the forecasting workflow.

With the strengthening connection of the operational forecasting to Earth science, the vastly improved geophysical datasets described above, and the rapid advances in computing resources, the last decade has seen an unprecedented expansion in the range and complexity of modeling approaches currently being implemented for river forecasting at short to medium ranges (out to several weeks). The development and deployment of distributed “macroscale” (on the order of 100 km² spatial resolution) hydrology models in the 2000s for continental- and global-scale domains as part of land data assimilation systems (LDASs; Mitchell et al. 2004; Rodell et al. 2004) was a notable step toward a broader convergence between parameterization approaches in watershed-scale hydrological models and global-scale LSMs used in coupled climate models and as initial conditions for numerical weather prediction (NWP). These large-domain LSMs were operationalized in research and agency settings for monitoring and prediction applications (Wood and Lettenmaier 2006; Thielen et al. 2009; Alfieri et al. 2013), adopting techniques such as ESP from traditional seasonal forecasting. At the watershed scale, high-resolution distributed process-oriented models also were implemented for forecasting applications, primarily at a local scale (Westrick et al. 2002).

Today, traditional river flood forecasting using conceptual models is joined by two new major operational

river forecasting paradigms that leverage scientific advances and technical strategies from the field of Earth system modeling (ESM). One is the application of very high-resolution watershed models (on the order of 100 m horizontal resolution) for national to continental domains, incorporating explicit hydrologic vertical and lateral fluxes at the model grid scale. Such a computationally demanding modeling is enabled by HPC and code parallelization, in some cases resourcing thousands of cores, and places a new emphasis on the need to resolve long-standing hydrologic modeling challenges. In particular, the vast increase in the distributed parameter space coupled with the cost of simulation has underscored the need for efficient parameter estimation and regionalization approaches. A recent example of such an approach is the NWS National Water Model (NWM; Fig. 25-5), which produces distributed, deterministic river flood predictions and other hydrometeorological outputs for every 250 m of the United States (Salas et al. 2018).

A second major paradigm to emerge centers on the development and application of ensemble techniques for forecasting (Duan et al. 2018), using either conceptual or intermediate-scale hydrologic models, with the objective of “completing the forecast” (National Research Council 2006)—that is, providing forecasts with uncertainty estimates that can be used for risk-based decision-making. In the last 15 years, ensemble prediction has made great strides in the United States and internationally, facilitated in part by the international Hydrologic Ensemble Prediction Experiment (HEPEX; Schaake et al. 2007) launched in 2004. A key challenge in the ensemble prediction context is the provision of probabilistic forecasts that are as accurate (sharp) as possible while maintaining statistically reliable spread (uncertainty; Werner et al. 2016). Ensemble river prediction systems now coexist with traditional forecasting systems in a number of countries; examples include the U.S. NWS Hydrologic Ensemble Forecast Service (HEFS; Demargne et al. 2014) and the European Flood Awareness System (EFAS; Thielen et al. 2009), which coordinates with national hydrometeorological services to provide forecast services across Europe.

d. Flash flood forecasting

Flash flooding is one of the most damaging water-related hazards, particularly from a human life standpoint (with over 100 lives lost per year), but for most of the last century and in many countries today, the responsibility for forecasting of flash floods has been carried out by meteorological rather than hydrological services. This organization follows from both the distributed nature of flash flooding, in that deadly torrents

of water can be generated on small creeks and washes that were not explicitly modeled in traditional river flood prediction systems, and the fact that flash flooding is proximally driven (far more than river flooding) by meteorological events, rather than a combination of rainfall inputs and subsurface fluxes (e.g., baseflow, interflow). For decades, flash flood watches and warnings have been a central alert category of the NWS, but were undertaken by weather forecast offices (WFOs) rather than river forecast centers (RFCs), and they issued products describing areas or regions of risk (polygons on a map), rather than explicit locations with a quantitative high-flow forecast.

Such products were originally generated from meteorological forecast maps alone and have steadily improved in the last four decades. Progress in several foundational scientific and technical areas has driven these advances. Rainfall monitoring has benefited from improvements in the quality of multisensor products and objective analysis techniques, and forecasting has leveraged higher-resolution and more accurate NWP. Improved satellite-based descriptions of terrain and land cover have enabled hyper-resolution digital elevation models (DEMs) and runoff routing schemes. Hydrologic simulation has evolved from lumped, conceptual models to the development and implementation of ever-finer-resolution distributed hydrologic models (Wigmosta et al. 1994; Bierkens et al. 2015) that can make use of distributed meteorological inputs.

The first of such distributed hydrology models to be applied operationally on large domains was intermediate scale, as exemplified by the the 1/8° multiagency National Land Data Assimilation System (NLDAS) project LSMs, beginning in 1998 (Mitchell et al. 2004), and the 4-km NWS Hydrology Laboratory Research Distributed Hydrologic Model (HL-RDHM; Koren et al. 2004). At a finer resolution, the 1-km distributed NWS Flooded Locations and Simulated Hydrographs (FLASH) system in 2012 also began providing 1-km, 5-min channel estimates of flash flood risk, based on the percentiles of simulated and routed flow relative to a background climatology (Gourley et al. 2014). Since 2016, the 250-m NWM has issued quantitative 0–2-day channel lead flow predictions for a channel network defined by the USGS National Hydrography Dataset Plus (NHDPlus), which includes 2.7 million river reaches.

Flash flood forecasting has commonly taken the form of Flash Flood Guidance (FFG; Carpenter et al. 1999; Georgakakos 2006), in which models—initially lumped and more recently gridded—are used to estimate the quantitative capacity of the soil to absorb rainfall, which, when combined with estimates of observed and forecasted precipitation, characterizes flood risk. In the

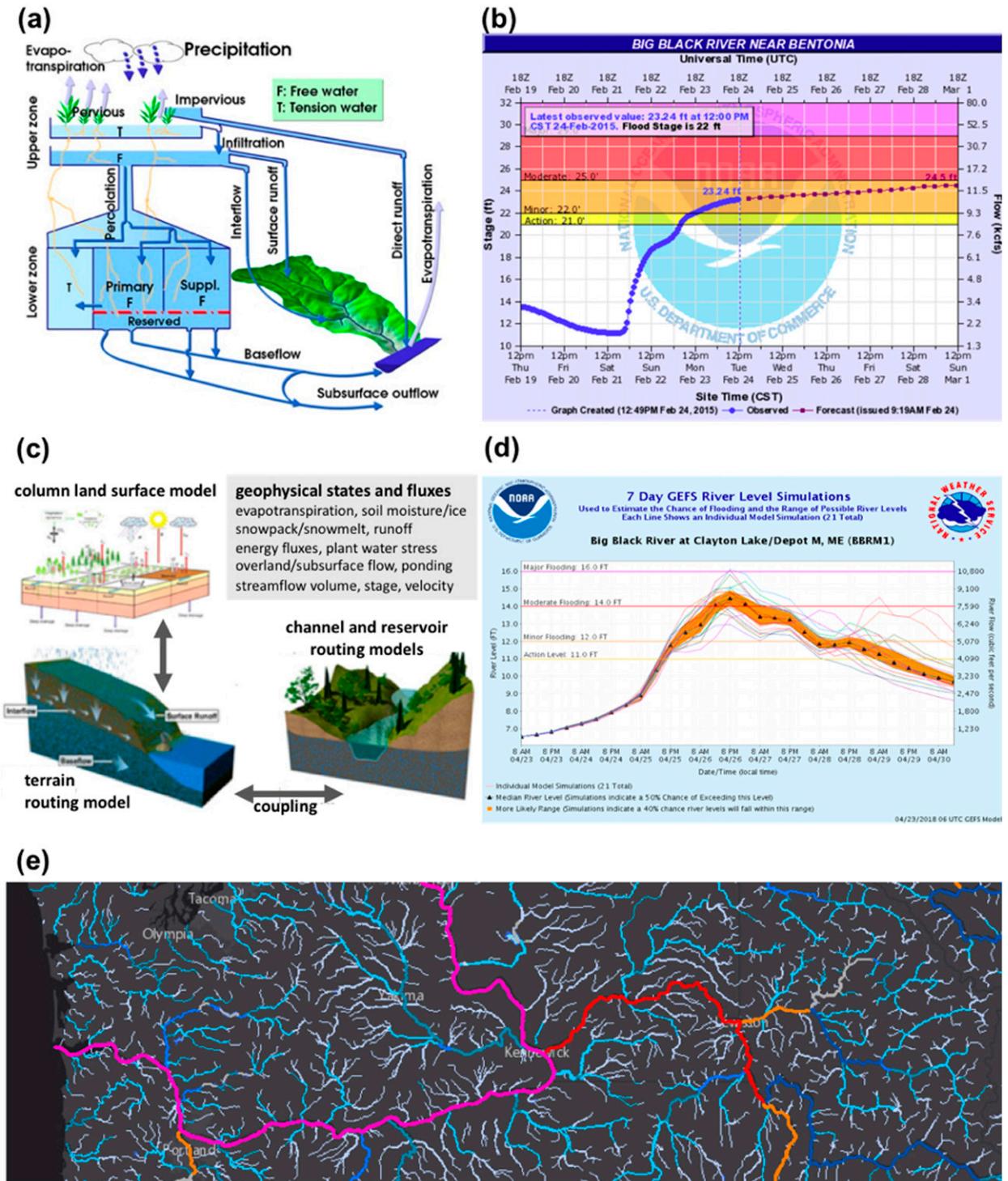


FIG. 25-5. Hydrologic forecasting evolution from (a) conceptual models [e.g., Sacramento Soil Moisture Accounting (SAC-SMA)] and (b) deterministic river flood forecasts to (c) coupled land surface, terrain, and channel routing models, (d) ensemble river flood forecasts, and (e) distributed model and stream channel predictions (e.g., the NWS NWM showing the outlet of the U.S. Columbia River basin).

distributed model context, where local calibration of runoff is not possible, a “threshold frequency” concept is applied in which modeled flows above a certain frequency threshold (e.g., 2-yr return period) relative to past model simulations are indicative of flood risk (Reed et al. 2007). Currently, both lumped and gridded FFG approaches inform operational flash flood watch and warning products, and the practice has been promoted for international adoption by the World Meteorological Organization (WMO).

The WFO areal alerts and the NWM data products represent the diversity of strategies for flash flood forecasting, from local assessment resulting from experts integrating hydrometeorological information to the automated outputs of a very high-resolution LSM. Globally, development is underway to create fine-resolution, real-time distributed flood risk maps with low latency by merging satellite-based estimates of land inundation, river-level altimetry, NWP and global NWP model runoff, and even real-time social media streams (i.e., Twitter alerts; Westerhoff et al. 2013).

e. The realization of a hydrologic prediction science

The nascent applications of LSMs and complex watershed models in prediction applications have unleashed great scientific and pragmatic enthusiasm for a new era of Earth system prediction that includes hydrological fields as well as more common meteorological ones. Yet, the ease at which data streams can be connected to models, generating real-time outputs that can be called “forecasts,” belies the substantial difficulty in producing not just distributed model output, but actionable, high-quality predictions at the local scales required for water and emergency management. Myriad, long-standing scientific challenges in hydrologic modeling remain unsolved (Clark et al. 2017) and are compounded by technical challenges that arise, particularly in the operational forecasting context. Operational river forecasters grapple with these challenges—for example, erratic or degraded data streams, model deficiencies, model state and input uncertainties—on a daily basis, and have a deep, first-hand understanding of the adequacy of hydrological methods to overcome them. Welles et al. (2007) highlighted the role that objective verification measures should play in adopting forecast process improvements. Essential techniques include parameter estimation (or model calibration; e.g., Welles and Sorooshian 2009), meteorological forecast downscaling and bias correction, hydrologic data assimilation, hydrologic forecast postprocessing, and accounting for and integrating water management—all of which reduce errors in model predictions.

Fortunately, the gap between the Earth system prediction research community and the operational river

forecasting community is narrowing. The view that current operational methods are ad hoc or (in the case of model calibration) theoretically unsound, a way of “getting the right answer for the wrong reasons,” is waning as the ESM community begins to understand the role of such methods and forecasting-related challenges are rearticulated in the language of science. For all the advances of the last decade, modern hydrological models run in a fully automated, “over the loop” paradigm, without effective, objective counterparts to traditional techniques described above, often still fail to generate outputs matching the actionable quality of those created by simpler models using traditional approaches (Smith et al. 2004; Reed et al. 2004; Smith et al. 2013; Beven et al. 2015). Today, national and global-scale hydrologic forecasting system implementations successfully reflect large-scale variability but do not bear scrutiny at the local watershed scale, where river flood impacts are most relevant. To make global-scale approaches usable locally, the Earth system prediction research community must continue to pursue scientific understanding and methods to improve engineering-based techniques in three key areas: model parameter estimation, hydrologic data assimilation of observations of all types, and representing human impacts on the hydrologic cycle.

Fortunately, over the last 15 years, a shift in perspective is leading to greater integration of traditional communities of practice and land surface modeling research. The international, multiagency HEPEx initiative has fostered collaboration between communities of hydrologic research and practice, and promoted the recognition of operational hydrologic prediction as a coherent scientific subdiscipline rather than an engineering activity. Research funding in the United States and elsewhere has increasingly supported collaborations with operational entities, and sessions on applied hydrologic prediction in national and international scientific conferences have greatly expanded. As this integration between traditional operational and research communities grows, an improved understanding of tradeoffs and limitations in forecasting system components is beginning to lead to more informed choices as Earth system research strives to create next-generation prediction systems that provide actionable information at local scales.

From a stakeholder perspective as well, the landscape of hydrological forecasting is changing dramatically. For most of the last century, for a given river location, at most one deterministic flood forecast was available from the regional forecast center, using a locally tailored conceptual watershed model. Today, multiple forecasts for a given U.S. location can be viewed and downloaded,

including one from the NWM and additional ensemble predictions from global forecasting systems run outside of the United States. For some of these systems, the runoff is even extracted from the land surface models of coupled global models (e.g., NWP systems), rather than from offline LSMs (Gaborit et al. 2017). These developments are ushering in a hydrologic forecasting future that is marked by an expansion from local and regional forecasting approaches toward national-, continental-, and global-scale hydrologic prediction systems, run in centralized over-the-loop modes.

Improvements in forecast quality will come through a range of advances: better Earth system modeling (including coupled NWP and climate prediction) and creative, efficient solutions for representing space–time heterogeneity (e.g., Peters-Lidard et al. 2017); improved observational data through data fusion and assimilation of satellite-based observations as well as nontraditional observations from social media; the adoption of reliable uncertainty frameworks; and the use of increasingly sophisticated statistical techniques (e.g., deep learning) merged in hybrid frameworks with dynamical modeling. The overarching scientific and community challenge facing the field today is the need to connect local knowledge and information to large-domain approaches, and in turn to make national to global system predictions relevant locally. With greater integration of the entire hydrologic prediction community—including those in research, in operations, and stakeholders—this challenge can be surmounted.

5. The emergence of global hydrology

a. Key milestones

As people gained the ability to travel more quickly and routinely across continents and around the world in the twentieth century, it was inevitable that they would come to recognize the global connectivity of Earth's physical processes and systems, including the water cycle. Voeikov (1884) and Murray (1887) had the foresight to assess worldwide, terrestrial precipitation, evaporation, and runoff well before global-scale meteorological measurements began to be collected systematically. Murray's (1887) global terrestrial precipitation estimate is particularly impressive, being only about 5%–10% larger than the average of estimates from the 2000s (Fig. 25-6a). Fritzsche's (1906) global terrestrial precipitation estimate was even better, but most studies misestimated evapotranspiration and/or runoff until Lvovich's (1972) published values that remain very close to the most modern estimates. Brückner (1905) and Fritzsche (1906) provided the first global ocean

precipitation and evaporation estimates. Amazingly, they were only about 10% below modern estimates (Fig. 25-6b). Mather (1969) and Baumgartner and Reichel (1973, 1975) were the first to hit the mark with both ocean evaporation and precipitation, based on recent estimates. Baumgartner and Reichel's (1975) and Budyko and Sokolov's (1978) land and ocean flux estimates, with some updates from Chahine (1992) and Oki (1999), continued to be used as benchmarks until Oki and Kanae (2006) and Trenberth et al. (2007) delivered updated global water balance assessments.

A decade after Baumgartner and Reichel's (1975) comprehensive treatise on the global water balance, Eagleson (1986) announced the “emergence of global-scale hydrology” and evaluated the state of global hydrological modeling at that time. Chahine (1992) helped to establish the global hydrology community in his review paper on the hydrological cycle and its influence on climate, declaring that, “In the short span of about 10 years, the hydrological cycle has emerged as the centerpiece of the study of climate, but ... rather than fragmented studies in engineering, geography, meteorology and agricultural science, we need an integrated program of fundamental research and education in hydrological science.” Other important milestones included Berner and Berner's (1987) thorough physical and chemical description of the water cycle, and Mintz and Serafini (1992) recognizing the importance of water storage in the land when they published a global, monthly climatology of world water balance.

b. Remote sensing of the global water cycle

In 1958 NASA's *Explorer 1* satellite launched and provided imagery of clouds and snow cover that revolutionized the way scientists thought about the water cycle. Other satellites soon began to improve our ability to observe Earth from space, and in 1972 the Blue Marble photograph from *Apollo 17* inspired a new generation of Earth scientists and conservationists. During that seminal period of space exploration, the study of hydrology at continental to global scales began to accelerate. The difficulty in extrapolating limited point observations to those scales soon became clear, which was one of the motivations for satellite remote sensing of the global water cycle, and, more broadly, the entire Earth system (Famiglietti et al. 2015).

The first several satellites that were useful for studying the water cycle were not designed for that purpose. The *TIROS-1* satellite delivered fuzzy, black-and-white images of Earth in 1960, which elucidated the large-scale patterns of cloud and snow cover. The NASA–USGS Landsat series of satellites, which began in 1972, has provided increasingly higher-resolution imagery that

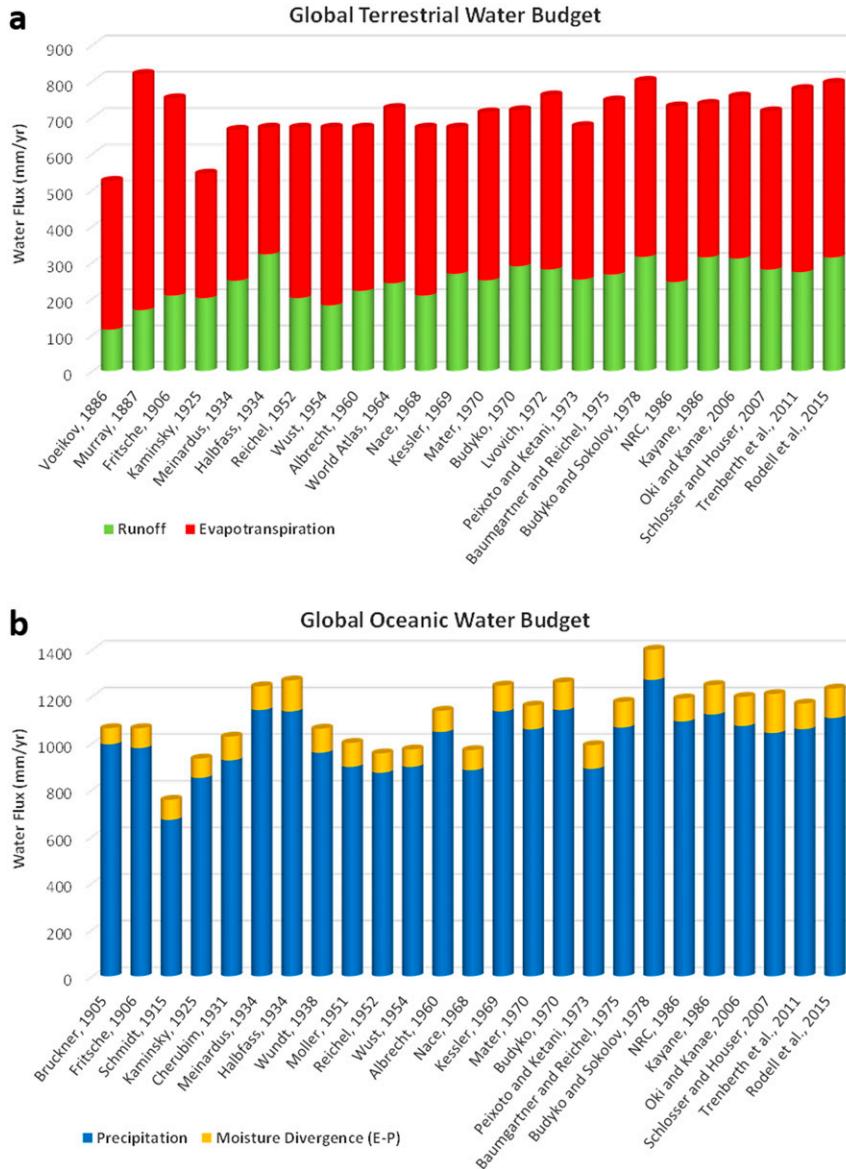


FIG. 25-6. The evolution of global, mean annual water flux estimates (mm yr^{-1}) during the past 100+ years, assuming long-term changes in atmospheric, terrestrial, and oceanic storages are negligible. (a) Global terrestrial runoff (green), evapotranspiration (red), and precipitation (the entire bar). (b) Global oceanic precipitation (blue), moisture flux divergence (orange; this is equivalent to evaporation minus precipitation and to terrestrial runoff to the oceans), and evaporation (the entire bar). Note the y axes of the two panels differ, and that the fluxes in (a) and (b) are not directly comparable due to the difference in the areas of the global land and global ocean.

has been useful for delineating and monitoring the extent of snow cover, glaciers, surface water bodies, different types of vegetation and land cover, and irrigation, and for estimating evapotranspiration. The first satellite in NOAA’s Geostationary Operational Environmental Satellites (GOES) series was launched in 1975, providing visible and infrared imagery that have been essential for

weather forecasting and similarly valuable for hydrometeorological studies. As described in section 3, satellite remote sensing now enables global-scale observation of precipitation, soil moisture, terrestrial water storage, snow cover depth and snow water equivalent, evapotranspiration, lake elevation, and soon river discharge. The integration of these capabilities has transformed

research on the global water cycle. [Lettenmaier et al. \(2015\)](#) provides a detailed overview of the contributions of remote sensing to hydrologic science, while [McCabe et al. \(2017\)](#) describes future prospects in this area.

c. Global-scale hydrological modeling

Global modeling of the water cycle is motivated by multiple considerations. For one, we cannot currently observe the water cycle globally with adequate resolution, accuracy, and continuity. While remote sensing can provide global coverage, the observations themselves typically are derived using retrieval algorithms that have to be calibrated and that require simplifying assumptions, both of which introduce error. Second, global models can be used to investigate different climate change or paleoclimate scenarios and to test sensitivities to natural properties and anthropogenic influences. Third, our proficiency in modeling the water cycle at the global scale provides insight into our understanding of the Earth system. Fourth, global land surface models can be coupled to Earth system models or used to integrate data from multiple observing systems. Finally, running a hydrological model at the global scale complements remote sensing or in situ observing systems that are limited by both cost and technology in their ability to provide continuous spatial and temporal coverage. Such models, which range from extremely simple water budget equations to physically based, coupled land-atmosphere-ocean models comprising tens of thousands of lines of code, have their own weaknesses. In particular, they are constructed using our sometimes-flawed understanding of physical processes, they rely on their own simplifying assumptions, and their accuracy is limited by that of the input parameters and meteorological variables. Nevertheless, global hydrological models have supported a huge number of water cycle studies over the years, and they enable sensitivity studies and the analysis of scenarios that could never be tested in the real world.

Global hydrological models were originally developed in order to improve the lower-boundary condition for atmospheric models. One of the first was [Manabe's \(1969\)](#) "bucket model," which he incorporated into the Geophysical Fluid Dynamics Laboratory's general circulation model, yielding estimates of the global rates of land surface evaporation. By the mid-1980s, it was understood that simplistic land surface representations were creating systematic errors in simulated evapotranspiration and hence the overlying atmosphere, leading to the development of more sophisticated schemes ([Dickinson 1984](#)). In particular, the Simple Biosphere model (SiB; [Sellers et al. 1986](#)) and the Biosphere-Atmosphere Transfer Scheme (BATS; [Dickinson et al.](#)

[1986](#)) enabled simulation of the transfers of mass, energy, and momentum between the atmosphere and the land surface. Thirty-two years later elements of SiB and BATS, the first soil-vegetation-atmosphere transfer schemes, are obvious in the code of many modern land surface models. Incorporating the influence of topography on runoff generation and other processes was one of the next key milestones ([Famiglietti and Wood 1991](#)) along with ways to represent such spatially and vertically distributed processes statistically ([Famiglietti and Wood 1994](#)). Around the same time, [Koster and Suarez \(1992\)](#) demonstrated a "tiling" approach to modeling multiple different vegetation types within a single grid pixel. An explosion of new land surface models ensued, which brought about the need for model intercomparison projects. The Global Soil Wetness Projects (GSWP) 1 and 2 ([Dirmeyer et al. 1999, 2006a](#)) focused on soil moisture and included 10 and 15 models, respectively, each with their own unique set of advances and simplifications. The Water Model Intercomparison Project (WaterMIP; [Haddeland et al. 2011](#)) emphasized water cycle fluxes and included five coupled models as well as six offline (land only) models.

Owing to exponential increases in computing power, incremental improvements in our understanding of water and energy cycle processes, and the availability of more accurate and higher-resolution forcing and parameter datasets, many global land surface models now run routinely (e.g., [Alcamo et al. 2003](#); [Rodell et al. 2004](#)) and most weather forecasting agencies have implemented advanced land surface modules into their operational systems. Software packages like the Land Information System (LIS; [Kumar et al. 2006](#)) now allow nonexpert users to configure and run multiple land surface models for both scientific and practical applications. A handful of land surface models have benefitted disproportionately from a community development approach and/or implementation by multiple operational agencies that foster their continued improvement. Examples include the Noah LSM with multiparameterization options (Noah-MP; [Niu et al. 2011](#)), the Community Land Model ([Oleson et al. 2010](#)), and the Joint U.K. Land Environment Simulator (JULES; [Best et al. 2011](#)). [Kumar et al. \(2017\)](#) concluded that this may be causing a convergence of the output from different models.

d. Community water cycle research initiatives

In addition to these individual efforts, GEWEX and other international programs have facilitated community initiatives aimed at improving understanding of the global water cycle and its components. [Kinter and Shukla \(1990\)](#) suggested a framework for utilizing ground- and space-based observations during the first phase of GEWEX toward the goal of improved

understanding of the global water and energy cycles. Subprojects within GEWEX have included global hydrology as an explicit component, such as the GSWP 1 and 2, LandFlux, the Coordinated Enhanced Observing Period (CEOP), and the GEWEX Hydroclimatology Panel (GHP). The International Geosphere–Biosphere Programme (IGBP; 1987–2015) similarly included projects relevant to global water cycle and water resources research, such as the Integrated Land Ecosystem–Atmosphere Processes Study (iLEAPS) and the Global Water System Project (GWSP). However, the Intergovernmental Panel on Climate Change (IPCC), which is perhaps the best known international community Earth science initiative, has focused largely on ground and near-surface air temperature variations and trends, while water cycle impacts have been considered secondarily. Further, a persistent obstacle for studies that are conceived within GEWEX and other community initiatives is that the initiatives themselves typically have little or no funding to support the research. NASA’s Energy and Water Cycle Study (NEWS) program has sought to overcome that issue and combine integrative community research with funding support, toward the goal of quantifying water cycle consequences of global climate change. Another example of that approach was the European Union’s Water and Global Change project (EU-WATCH; 2007–11), which aimed to bring together scientists from the hydrology and related communities to improve quantification and understanding of global hydrological processes.

e. Recent advances in global water cycle science

Many of the major advances in global water cycle science in the twenty-first century have involved 1) assessing changes in the water cycle and the distribution of water resources, 2) science enabled by satellite remote sensing, and/or 3) science enabled by data integrating numerical models with ever increasing spatial resolution and sophistication of process representation. Regarding the first, [Vörösmarty et al. \(2000\)](#) used climate model predictions together with hydrologic and socioeconomic information to assess the vulnerability of water resources to climate change and population growth, with startling results. Many related studies followed, including [Vörösmarty et al.’s \(2010\)](#) reassessment that also considered threats to biodiversity. [Allen et al. \(2002\)](#) analyzed variability of the hydrological cycle during the twentieth century in order to evaluate the range of possible twenty-first-century changes. [Milly et al. \(2002\)](#) reported an increasing risk of great floods due to climate change. [Bosilovich et al. \(2005\)](#) and [Held and Soden \(2006\)](#) analyzed climate model output to identify evidence of “intensification” of the water cycle,

which refers to the prediction of more intense and rapid cycling of water fluxes in a warming environment. [Milly et al. \(2008\)](#) warned that water management, which has heretofore relied on the assumption of stationarity—natural systems fluctuating within an unchanging range of variability—is imperiled by both direct human disturbances and climate change. [Brown and Robinson \(2011\)](#) used a combination of ground, airborne, and satellite datasets to estimate that March and April Northern Hemisphere snow cover extent decreased at a rate of ~ 0.8 million km^2 per decade during 1970–2010. GRACE has been used in combination with other data sources to quantify groundwater depletion around the world ([Rodell et al. 2009](#); [Wada et al. 2012](#); [Döll et al. 2014](#); [Richey et al. 2015](#); [Chen et al. 2016](#); [Rodell et al. 2018](#)). Nevertheless, [Reager et al. \(2016\)](#) used GRACE data to show that there was a net increase in nonfrozen terrestrial water storage during 2002–14, which reduced the rate of sea level rise by 15%. [Rodell et al. \(2015\)](#) used an objective optimization approach to combine ground- and space-based observational datasets with data integrating model output, covering the first decade of the millennium, while simultaneously closing the water and energy budgets at multiple scales. The result was a physically, spatially, and temporally consistent set of estimates of the major fluxes and storages of the water cycle at continental, ocean basin, and global scales ([Fig. 25-7](#)). This analysis is useful as a baseline for assessing future changes in the water cycle and for global model evaluations. Other projects have produced global water cycle accountings through the assimilation of data into global coupled or offline models (e.g., [Rienecker et al. 2011](#); [van Dijk et al. 2014](#); [Gelaro et al. 2017](#); [Zhang et al. 2018](#)).

Modeling improvements and the unprecedented availability of satellite-based observations have benefitted global water cycle science enormously, but questions and uncertainty remain. For example, while many have predicted an increasing occurrence of drought in a warming environment, [Sheffield et al. \(2012\)](#) reported no significant change in drought over 60 years, setting off an intense debate in the hydroclimate community. Similarly, when [Jasechko et al. \(2013\)](#) used an isotope analysis to estimate that transpiration accounts for 80%–90% of evapotranspiration globally, the community responded with a slew of alternate interpretations and analyses (e.g., [Sutanto et al. 2014](#); [Coenders-Gerrits et al. 2014](#)).

Future breakthroughs in global water cycle science will continue to be fueled by advances in remote sensing and modeling. Expansion of remote sensing data records is already enabling studies of global change that are less dependent on the sparse network of in situ observations. By 2030, many of these records will be long enough to

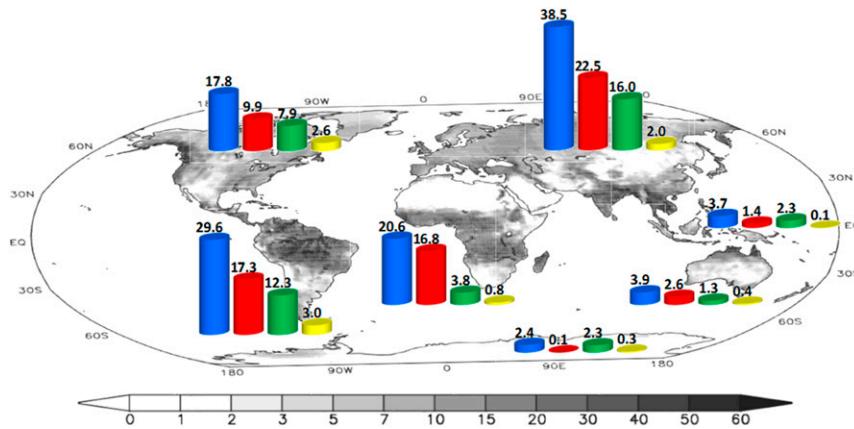


FIG. 25-7. Optimized annual-mean fluxes ($1000 \text{ km}^3 \text{ yr}^{-1}$) for North America (including Greenland), South America, Africa, Eurasia, the islands of Australasia and Indonesia, mainland Australia, and Antarctica: precipitation (blue), evapotranspiration (red), runoff (green), and annual amplitude of terrestrial water storage (yellow). The background grayscale image shows GRACE-based amplitude (max minus min) of the annual cycle of terrestrial water storage (cm). [From Rodell et al. (2015).]

generate climatologies and to identify trends. This is already happening with soil moisture (e.g., Owe et al. 2008; Dorigo et al. 2012) in addition to snow cover. While there are serious concerns about the decline of ground-based networks that are crucial for both long-term temporal continuity and calibration of remote sensing retrievals (Fekete et al. 2015), an optimistic perspective is that implementation of advanced observational approaches, including measurements using signals of opportunity and remote sensing from CubeSats and unmanned aerial vehicles, will fill gaps and provide a more complete view of the water cycle (McCabe et al. 2017).

6. Coupling of hydrology with the atmosphere and ecosystem

By redistributing surface and subsurface moisture in space and time, hydrological processes have an important control over evapotranspiration at the surface and moisture available to plants. Hence, hydrological science plays a critical role in understanding and modeling land-atmosphere interactions and ecohydrology, the topics of this section discussed below.

a. Coupling of hydrology with the atmosphere: Land-atmosphere interactions

Land and atmosphere can interact through exchanges of water, energy, momentum, and biogeochemistry that are influenced by many processes across a wide range of temporal and spatial scales. Understanding and modeling surface fluxes of precipitation, evapotranspiration,

sensible and latent heat, momentum, and aerosol particles and trace gases, as well as the processes that control these fluxes, are all important for advancing the study of land-atmosphere interactions. In the context of this monograph on progress in hydrology, this section focuses mainly on land-atmosphere interactions related to soil hydrological processes such as soil moisture, groundwater, and lateral flow. We note, however, that the land surface can interact with the atmosphere importantly through surface albedo, surface roughness, and biogeochemical processes that influence the net energy input to land and surface flux exchanges.

Traditionally, hydrological science has focused on understanding the hydrologic response to atmospheric forcing, as mediated by the landscapes at watershed and basin scales, while atmospheric science has focused on understanding atmospheric dynamical and physical processes that are influenced, to some degrees, by the surface fluxes. Hence, in hydrology, the atmosphere was considered an external forcing and was provided as an input to hydrologic modeling while in atmospheric modeling, land surface processes were ignored or simplified to provide lower-boundary conditions for atmospheric general circulation models (GCMs). For example, Manabe (1969) developed a bucket model to represent surface hydrology in GCMs, which allows time-evolving surface evapotranspiration to be calculated as lower-boundary conditions for GCMs.

The need to improve the lower-boundary conditions in GCMs became more recognized in the 1980s from studies that investigated the atmospheric response to land surface conditions (Shukla and Mintz 1982;

Rowntree and Bolton 1983; Mintz 1984) and impacts of deforestation on climate (Dickinson and Henderson-Sellers 1988). Efforts to develop land surface models with more physically based representations enabled the role of the land surface on climate to be better understood. For example, including a physical representation of soil moisture variability in a fully coupled ocean–atmosphere–land model, Delworth and Manabe (1993) noted that the presence of an interactive soil moisture reservoir increases the variance and adds memory to near-surface atmospheric variables such as humidity. Studies of precipitation recycling in the 1990s using gridded observations and analyses further established the role of the land surface in providing important sources of moisture for continental precipitation in certain regions (Brubaker et al. 1993; Trenberth 1999). Advances in modeling such as development of BATS (Dickinson et al. 1986) and SiB (Sellers et al. 1986) and observations such as the First International Satellite and Surface Climatology Project (ISLSCP) Field Experiment (FIFE) (Sellers and Hall 1992) and the Boreal Ecosystem Atmosphere Study (BOREAS; Sellers et al. 1995) provided an impetus for studying land–atmosphere interactions. Readers are referred to Garratt (1993), Entekhabi (1995), Eltahir and Bras (1996), and Betts et al. (1996) for reviews of advances in land–atmosphere interaction research through the 1990s.

With increasing availability of in situ and remotely sensed observations, gridded global and regional analyses and land data assimilation products, more complex modeling tools, and larger and faster computers, studies of land–atmosphere interactions have advanced more rapidly since the 2000s. More specifically, the role of soil moisture on precipitation, or soil moisture–precipitation feedback, has been investigated extensively using observations and modeling. GCM experiments (Koster et al. 2000b) and observations (Yoon and Leung 2015) showed that in some midlatitude continental areas during summer, the impacts of the oceans on precipitation can be small relative to the impacts of soil moisture, suggesting that soil moisture memory may provide important predictability for summer precipitation from weather to seasonal time scales.

Locally, soil moisture can influence precipitation through its impacts on the lower-level moist static energy (MSE) and the partitioning of surface energy flux between sensible and latent heat fluxes (Betts et al. 1996; Schär et al. 1999). Wetter soils increase the evaporative fraction (EF defined as the ratio of latent heat flux to the sum of the latent and sensible heat fluxes) to moisten the atmospheric planetary boundary layer (PBL) and lower the levels of lifting condensation and free convection, which may trigger convection and enhance

precipitation (Findell and Eltahir 2003). Conversely, sensible heat flux is enhanced relative to latent heat flux over drier soils to deepen the PBL and dilute the moist static energy within the PBL. As the PBL grows, more rigorous entrainment of drier air from above the PBL further reduces the MSE within the PBL and reduces the likelihood of convective triggering and precipitation. Changes in cloud cover may further enhance the positive soil moisture–precipitation feedback as increased formation of convective clouds over wetter soils may increase the net radiation at the surface if the reduction of outgoing longwave radiation by the high clouds overcompensates for the reduction of solar radiation due to cloud cover, thus increasing the MSE and convection (Schär et al. 1999; Pal and Eltahir 2001).

Importantly, the sign of the soil moisture–precipitation feedback may depend not only on the partitioning of the surface fluxes, which depends on soil moisture (i.e., surface control), but also on the atmospheric conditions that determine the levels of lifting condensation and free convection (i.e., atmospheric control). The more rapid growth of the PBL over drier soils may allow the PBL to reach the level of free convection and trigger convection and precipitation while convection is prohibited over wetter soils because the level of lifting condensation may never reach the top of the shallow PBL, despite enhanced moisture and MSE by the increased latent heat flux over wetter soils (Findell and Eltahir 2003). Soil moisture can also influence atmospheric circulation through changes in the thermal gradients near the surface that induce sea level pressure gradients. Changes in sea level pressure can influence mesoscale circulation such as the Great Plains low-level jet (Fast and McCorcle 1990) and large-scale circulation systems such as the monsoon systems (e.g., Douville et al. 2001). As soil moisture affects convection, it can also induce changes in the large-scale circulation through its impacts on convection and latent heating in the atmosphere.

To quantify the strength of land–atmosphere coupling, Koster et al. (2004, 2006) designed the Global Land–Atmosphere Coupling Experiment (GLACE) that provided an ensemble of GCM simulations following the same simulation protocol. In GLACE, each GCM was used to perform 16 simulations in which soil moisture varies in each simulation based on the precipitation produced by the model (i.e., land–atmosphere interactions are active). In another set of 16 simulations, geographically varying time series of subsurface soil moisture was forced to be the same across the simulations (i.e., land–atmosphere interactions are disabled). Comparison of the intraensemble variance of precipitation between the two sets of simulations yields an estimation of land–atmosphere coupling strength for each model.

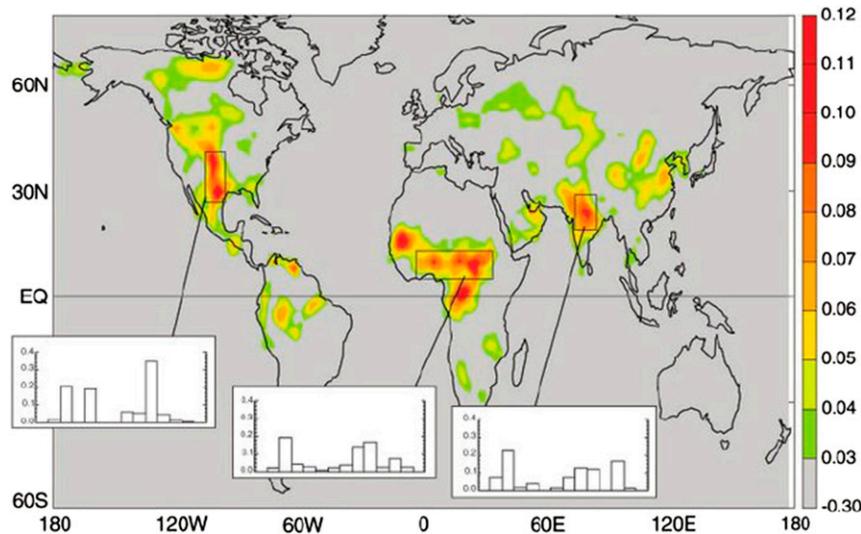


FIG. 25-8. The land–atmosphere coupling strength diagnostic for boreal summer (dimensionless, describing the impact of soil moisture on precipitation), averaged across the 12 models participating in GLACE. (insets) Areal averaged coupling strengths for the 12 individual models over the outlined, representative hot spot regions. No signal appears in southern South America or at the southern tip of Africa. [From [Koster et al. \(2004\)](#); reprinted with permission from AAAS.]

Since GCM representations of atmospheric and land surface processes vary, results from an ensemble of 12 GCMs that participated in GLACE provide a more robust estimate of land–atmosphere coupling strengths. [Koster et al. \(2004\)](#) identified the central United States, the Sahel, and India as hot spot regions of land–atmosphere coupling (Fig. 25-8). However, the use of 12 GCMs in GLACE also reveals large uncertainty in model estimates of land–atmosphere coupling strengths ([Koster et al. 2006](#); [Guo et al. 2006](#)).

The GLACE experiments motivated many follow-on studies to estimate the land–atmosphere coupling strength using observations and modeling experiments. For example, using long-term in situ measurements from the U.S. Department of Energy Atmospheric Radiation Measurement (ARM) Program in the Southern Great Plains and FLUXNET sites in the United States and Europe ([Baldocchi et al. 2001](#)), [Dirmeyer et al. \(2006b\)](#) compared the local covariability of key atmospheric and land surface variables similar to [Betts \(2004\)](#) in model simulations and in situ measurements. They found that most models do not reproduce the observed relationships between surface and atmospheric state variables and fluxes, partly due to systematic biases in near-surface temperature and humidity. Despite the large intermodel spread and biases in individual models, the multimodel mean captures behaviors quite comparably to that observed. The international GEWEX Global Land–Atmosphere System Study (GLASS) panel has formed

the Local Land–Atmosphere Coupling (LoCo) project to focus on understanding and quantifying these processes in nature and evaluating them with standardized coupling metrics ([Santanello et al. 2018](#)).

With availability of global surface soil moisture and precipitation data from satellites, [Taylor et al. \(2012\)](#) evaluated the soil moisture–precipitation feedback, focusing particularly on the least well understood aspect of the feedback loop—the response of daytime moist convection to soil moisture anomalies. They analyzed the location of afternoon rain events relative to the underlying antecedent soil moisture using global daily and 3-hourly gridded soil moisture and precipitation data at $0.25^\circ \times 0.25^\circ$ resolution to determine whether rain is more likely over soils that are wetter or drier than the surrounding areas. Across all six continents studied, they found that afternoon rain falls preferentially over soils that are relatively dry compared to the surrounding area, implying that enhanced afternoon moist convection is driven by increased sensible heat flux over drier soils and/or increased mesoscale variability in soil moisture, and hence a negative soil moisture–precipitation feedback. In contrast, a positive feedback dominates in six global weather and climate models analyzed, which may contribute to the excessive droughts simulated by the models.

The challenge of modeling land–atmosphere interactions was elucidated by [Hohenegger et al. \(2009\)](#), who compared cloud-resolving simulations at 2.2-km grid

spacing in which deep convection is explicitly resolved with simulations at 25-km grid spacing with a cumulus parameterization. In their 2.2-km simulations with cumulus parameterization turned off, dry initial soil moisture conditions yield more vigorous thermals that more easily break through the stable air barrier. In contrast, a stable layer setting on top of the PBL that develops over wet initial soil inhibits deep convection. Hence, the 2.2-km simulations produce a negative soil moisture–precipitation feedback, but in the 25-km simulations with parameterized convection, deep convection is much less sensitive to the stable layer on top of the PBL because of the design of the convective parameterization, so simulations initialized with wet soil moisture produce stronger convection and a positive soil moisture–precipitation feedback. These results highlight the sensitivity of land–atmosphere interactions to model resolution and convection parameterizations.

Land–atmosphere interactions in weather and climate models are also sensitive to representations of land surface processes. With advances in land surface modeling incorporating more complete hydrological processes, the role of surface water–groundwater interactions on land–atmosphere interactions has been studied using models that include representations of groundwater table dynamics (e.g., Anyah et al. 2008; Yuan et al. 2008; Jiang et al. 2009; Leung et al. 2011; Martinez et al. 2016). These studies found that groundwater table variations can induce soil moisture anomalies that subsequently influence ET and precipitation through land–atmosphere interactions. Integrated hydrology model featuring land surface models coupled to detailed three-dimensional groundwater/surface-water models have also been used to investigate the role of groundwater dynamics and land–atmosphere feedbacks (e.g., Maxwell and Miller 2005). Applying such models to the southern Great Plains, Maxwell and Kollet (2008) found very strong correlations between groundwater table depth and land surface response in a critical zone between 2 and 5 m below the surface, which could then influence land–atmosphere interactions. Miguez-Macho and Fan (2012) found that groundwater can buffer the dry season soil moisture stress in the Amazon basin, with important effects on the dry season ET. With the long memory, groundwater table can potentially provide an important source of predictability for precipitation and other water cycle processes.

Besides groundwater dynamics, the impacts of lateral flow on land–atmosphere interactions have also been investigated using detailed hydrology models, as lateral flow is typically ignored in one-dimensional land surface models used in weather and climate models. Subsurface lateral flow can have important effects on ET and the

partitioning of ET between transpiration and bare ground evaporation through spatial redistribution of soil moisture and groundwater table. Using a continental-scale integrated hydrology model, Maxwell and Condon (2016) found that including lateral subsurface flow in models increases transpiration partitioning from 47% to 62% over the conterminous United States. With integrated hydrology coupled to an atmosphere model in regional domains, the impacts of groundwater dynamics and lateral flow have been investigated in recent studies. In idealized simulations, terrain effects dominate the PBL development during the morning, but heterogeneity of soil moisture and water table can overcome the effects of terrain on PBL in the afternoon and influence the convective boundary layer strongly in wet-to-dry transition zones (Rihani et al. 2015). In case studies of strong convective precipitation events, modeling using coupled atmosphere-integrated hydrology model shows that groundwater table dynamics can affect atmospheric boundary layer height, convective available potential energy, and precipitation through its coupling with soil moisture and energy fluxes (Rahman et al. 2015). Recognizing the importance of subsurface processes on land–atmosphere interactions, groundwater dynamics are now commonly included in land surface models used in climate models, but lateral subsurface flow is still mostly ignored (Clark et al. 2015), though some efforts have begun to introduce parameterizations of lateral flow in land surface models (e.g., Miguez-Macho et al. 2007; Maquin et al. 2017).

Research over the last few decades has greatly advanced understanding and modeling of land–atmosphere interactions. The impacts of initial soil moisture conditions on weather forecast skill (e.g., Trier et al. 2004; Sutton et al. 2006) and seasonal forecast skill (e.g., Fennessy and Shukla 1999; Douville and Chauvin 2000; Ferranti and Viterbo 2006; Della-Marta et al. 2007; Vautard et al. 2007; Koster et al. 2010; Hirsch et al. 2014) through land–atmosphere interactions have been demonstrated. Land–atmosphere interactions have also been found to have important effects on extreme events such as droughts (Hong and Kalnay 2000; Schubert et al. 2004) and floods (Beljaars et al. 1996) as soil moisture anomalies and precipitation anomalies may be amplified through positive soil moisture–precipitation feedback (Findell et al. 2011; Gentile et al. 2013; Ford et al. 2015; Guillod et al. 2014, 2015; Taylor 2015; Hsu et al. 2017). Land–atmosphere interactions can also contribute to summer heat waves (Fischer et al. 2007) as anomalous warm temperatures reduce soil moisture through enhanced ET, and drier soils may subsequently intensify and prolong the heat waves.

Understanding land–atmosphere interactions is also important for understanding the impacts of land-use and

land-cover change (LULCC). Based on global observations of forest cover and land surface temperature, forest losses have been shown to significantly alter ET and amplify the diurnal temperature variation and increase the mean and maximum air temperature (Alkama and Cescatti 2016). Consistent findings are obtained using modeling, showing that conversion of midlatitude forests to cropland and pastures increases the occurrence of hot, dry summers (Findell et al. 2017). The impacts of afforestation in the midlatitude on climate have also been studied, with results showing the important control of soil moisture on the response (Swann et al. 2012). In water-limited regions in which latent heat flux is not able to compensate for the increase in surface temperature due to increase in solar absorption by the darker forest, afforestation can lead to large warming. The latter can induce changes in remote circulation and precipitation by perturbing the meridional energy transport that shifts the tropical rainbelt. Irrigation can have important effects on precipitation locally through its impacts on soil moisture, ET, and surface cooling (e.g., Kueppers et al. 2007; Bonfils and Lobell 2007), and remotely through its impacts on atmospheric moisture transport (e.g., DeAngelis et al. 2010; Lo and Famiglietti 2013; Yang et al. 2017).

Land–atmosphere interactions can also play an important role in modulating the impacts of global warming. For example, land–atmosphere interactions can enhance interannual variability of summer climate such as summer temperatures because climate regimes may shift as a result of greenhouse warming. The latter can create new wet-to-dry transitional climate zones with strong land–atmosphere coupling (Seneviratne et al. 2006), and GCMs provided some evidence that land–atmosphere interactions will be enhanced in a warmer climate (Dirmeyer et al. 2012). Climate models projected an increase in global aridity in the future. This response has been attributed to the larger warming over land relative to the ocean, which increases the saturation vapor deficit over land as moisture over land is mainly supplied by moisture evaporated from the ocean surface, which increases at a lower rate due to the smaller warming (Sherwood and Fu 2014). However, the GLACE–CMIP5 (phase 5 of the Coupled Model Intercomparison Project) experiments show that the increase in aridity under global warming can be substantially amplified by land–atmosphere interactions through changes in soil moisture and CO₂ effects on plant water use efficiency (Berg et al. 2016).

In summary, land–atmosphere interactions have important implications to weather and climate forecast skill, understanding and predicting extreme events including floods, droughts, and heat waves, and projecting

future changes in surface climates such as surface temperature variability and drought and aridity over land. Although both complexity and resolution have increased over time, models still struggle to reproduce the observed surface fluxes (Dirmeyer et al. 2018), suggesting more efforts are needed to improve modeling of the behaviors of the coupled land–atmosphere system. Coordinated modeling experiments such as GLACE, GLACE-2, and GLACE–CMIP5 have provided valuable insights on land–atmosphere interactions and their role in predictability and climate change impacts. Coordinated efforts to design experiments such as Clouds Above the United States and Errors at the Surface (CAUSES; Morcrette et al. 2018) focusing particularly on understanding model biases, combined with more systematic use of process-oriented diagnostics and designing observing approaches targeting the data needs for characterizing land–atmosphere interactions (e.g., Wulfmeyer et al. 2018), could prove useful for advancing understanding and modeling of land–atmosphere interactions. Cloud-resolving and large-eddy simulations constrained by observations could provide detailed information for improving understanding of the complex processes involved in land–atmosphere interactions in different climate regimes.

b. Coupling of hydrology with ecosystems: Ecohydrology

Ecohydrology is the study of the interactions between ecosystems and the hydrological cycle (e.g., Porporato and Rodríguez-Iturbe 2002). Building upon theory and approaches from both hydrology and ecology, ecohydrology is an extension of the study of the water cycle to include its impacts and feedbacks with other ecosystem processes such as biogeochemistry, plant ecology, and climate (Hannah et al. 2004). Though this discipline arose from hydrologic and ecosystem science that dates back more than a century (Rodríguez-Iturbe 2000; Vose et al. 2011), the explicit focus on understanding the processes that couple and feedback between hydrology and vegetation has allowed ecohydrology to make significant advances. This relatively young discipline (e.g., Zalewski 2000) has in common among these advances the theme of integrated, multidisciplinary research focused on the interactions between the biota and the hydrologic cycle.

Ecohydrology has benefitted science and society through improved understanding of the hydrologic cycle, ecosystem function, and how climate change, management, and disturbances impact resources of human value (e.g., Adams et al. 2012; McDowell et al. 2018). Because of the inherent interdisciplinary nature of ecohydrology, it has resulted in significant knowledge

gains in fields ranging from biogeochemistry to plant physiology to climate impacts. Ecohydrological research spans from arid environments where water–ecosystem coupling is strongly evident in part through water scarcity (e.g., [Newman et al. 2006](#)), to humid environments where ecohydrological conditions result in ecosystems sustaining high biomass and stature (e.g., [Brooks et al. 2010](#)).

Ecohydrology is rooted strongly in observations of both hydrologic and ecosystem parameters that respond to each other. Classical hydrologic measurements such as streamflow, precipitation, and evaporation remain a central component of ecohydrology as they are in hydrology, but are often coupled with measurements of plant water sources, transpiration, and ecosystem biogeochemical fluxes to better understand the interacting systems. A frequent focus is on vegetation–hydrology feedbacks with the goal of understanding where and how plants obtain water, and how such water acquisition subsequently impacts the local water cycle (e.g., [Brantley et al. 2017](#)). A global review of depth of plant acquisition of water revealed a wide range of rooting depths, with a surprisingly large fraction derived from groundwater ([Evaristo and McDonnell 2017](#); [Fan et al. 2017](#); [Fig. 25-9](#)). In addition to a large groundwater support of plant transpiration ([Fig. 25-9](#)), ecohydrology has also revealed an additional surprise, that plants use water that is from a distinct pool from the source of stream water (e.g., [McDonnell 2017](#)). The two water worlds hypothesis that has emerged from these observations has yielded significant improvements in our understanding of ecosystem function ([Berry et al. 2018](#)) and has major implications for how we understand, model, and manage catchment hydrologic cycles. This observation fundamentally improves our knowledge of the hydrologic cycle and its control, and simultaneously informs us on vegetation function.

One focal area of ecohydrology has been to understand how management of ecosystem properties impacts subsequent ecohydrological processes across all terrestrial ecosystems ([Wilcox 2010](#)). Ecohydrological management applies to estuaries and coastal waters ([Wolanski et al. 2004](#)) to the management of forests to maximize water-based resources ([Ford et al. 2011](#)). There is a long-history of investigation into runoff responses to forest harvest (e.g., [Bosch and Hewlett 1982](#); [Beschta et al. 2000](#); [Holmes and Likens 2016](#)), much of which falls into the category of ecohydrology due to the coupled nature of the investigation into the interactions of hydrology and vegetation disturbance. Ecohydrology of agricultural and other managed lands is also a critical issue given our growing demand for food and fuel production and the tight coupling between hydrology and crop yields ([Hatfield et al. 2011](#)). Future

land management can benefit from ecohydrological knowledge and forecasts to better mitigate the consequences of warming temperatures, drought, and associated disturbances on both ecological and hydrological functions of human value, for example, crop production, water yields, and energy supply ([Vose et al. 2011](#); [McDowell et al. 2018](#)).

Our changing climate has provided a large impetus to understand how ecohydrological functions may change under future conditions ([Vose et al. 2011](#); [Wei et al. 2011](#)). Rising temperature is forcing greater evaporative demand (e.g., [Trenberth et al. 2014](#)), resulting in greater water stress for ecosystems ([Williams et al. 2014](#)). Integration of hydrologic formulations such as Darcy’s law into ecohydrologic frameworks suggests that vegetation stature must decline under increasing evaporative demand, even with no change in the frequency of precipitation droughts ([McDowell and Allen 2015](#)); this theory is supported by experimental, observational, and simulation evidence ([Allen et al. 2015](#); [Bennett et al. 2015](#)). However, rising carbon dioxide is also increasing water use efficiency (but not growth; e.g., [Peñuelas et al. 2011](#); [van der Slepen et al. 2015](#)), resulting in a shift in the balance of carbon uptake per water consumed that has significant potential hydrologic impacts on soil water content and streamflow (though climate and land use may have larger impacts; e.g., [Piao et al. 2007](#)). Thus, ecohydrologic approaches will be valuable for understanding the net impacts of future global change on the hydrologic cycle and its feedbacks with ecosystem functions.

Vegetation disturbances, and their dependence and feedbacks upon hydrology, have become an important ecohydrology research focus in recent years (e.g., [Adams et al. 2010](#)). Watershed-scale measurements and process modeling are revealing both increasing and decreasing streamflow responses to vegetation loss via disturbance (reviewed in [McDowell et al. 2018](#)). Multiple possible ecohydrologic impacts and feedbacks appear to be underlying these variable responses of disturbances on hydrology. The removal of transpiring vegetation by wildfire, logging, or insect outbreaks are expected to increase streamflow due to reductions in net transpiration from the ecosystem, however, shifts in interception and albedo can allow net infiltration responses to go in the opposite direction, resulting in complex streamflow responses to disturbances (e.g., [Molotch et al. 2009](#); [Adams et al. 2012](#); [Bennett et al. 2018](#)). Other global change factors that are a growing focus of ecohydrological research include the impacts of invasive species and land-use change ([Vose et al. 2011](#)).

Models play a large role in our understanding and prediction of ecohydrology. Next-generation models of

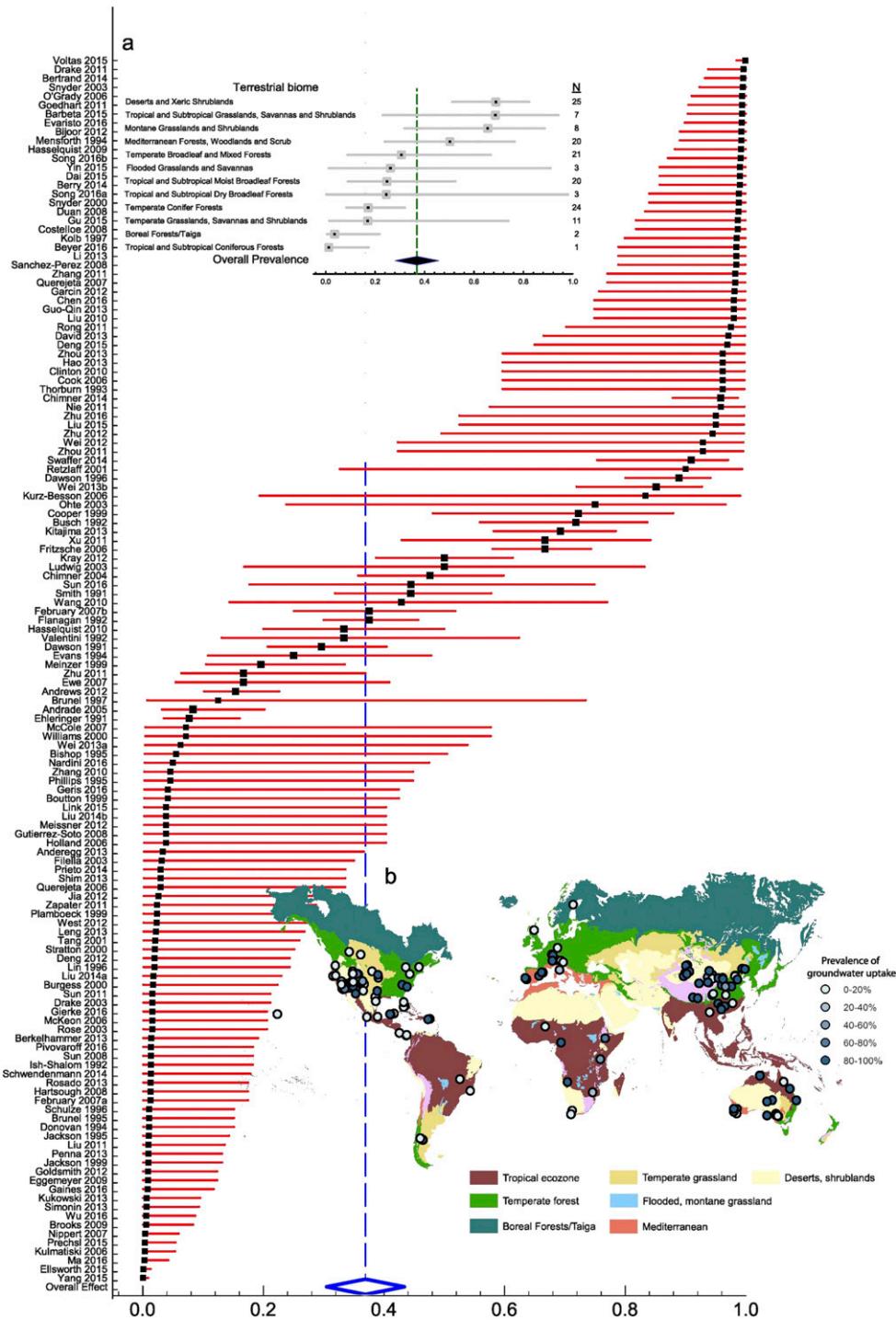


FIG. 25-9. Prevalence of groundwater influence (x axis: 0 corresponds to no groundwater influence). Main plot: prevalence estimates grouped by source paper (first author–year format). Filled black squares are prevalence point estimates, error bars are 95% confidence intervals (CI, red horizontal lines). Open diamond represents overall prevalence value and its 95% CI is represented by the width of diamond. (a) Prevalence estimates grouped by terrestrial biome with *N* representing corresponding number of sites. (b) Map of prevalence estimates in 162 sites in the global meta-analysis database [terrestrial biomes delineated by The Nature Conservancy <http://www.nature.org>; map was generated using ArcMap 10.2 (http://services.arcgis.com/arcgis/rest/services/World_Street_Map/MapServer)]. [From Evaristo and McDonnell (2017).]

the coupling of water, vegetation, and biogeochemistry are emerging that capitalize on the simulation strengths from each discipline. For example, inclusion of rigorous plant hydraulics knowledge from empirical physiology work has allowed much improved representation of plant transpiration and its dependence on rooting depth (e.g., Mackay et al. 2015) and can now be fully coupled to photosynthesis (Sperry et al. 2017). Such models are now being employed to understand how regional drought kills trees (e.g., Johnson et al. 2018). Likewise, the growing frequency and severity of terrestrial disturbances, such as insect outbreaks and wildfires, have driven significant ecohydrological advancement in recent years. The ecohydrologic consequences of these disturbances are large, including vegetation removal, accelerated sediment transport, and changes in the timing and amount of water yields (Adams et al. 2010; Penn et al. 2016; McDowell et al. 2018; Bennett et al. 2018). Using process models, we can better understand how disturbances impact the water cycle (Bearup et al. 2016).

Representation of hydrology in Earth system models remains challenged by integration of hydrologic and land surface processes (Clark et al. 2015), and thus an ecohydrologic approach is required to advance model representation. This is particularly true under a nonstationary climate, in which the feedbacks and interactions between climate, hydrology, and vegetation are complex and difficult to test. An important component to bridging the gap between modeling hydrology and ecosystems is the use of ecohydrological benchmarks (Kollet et al. 2017). For example, the most rigorous tests of Earth system models will require not only hydrologic benchmarks (e.g., streamflow, soil water content) but also of vegetation function (e.g., transpiration, growth). Utilization of new tools such as the International Land–Atmosphere Modeling Benchmarking (ILAMB; Hoffman et al. 2017) and the Protocol for the Analysis of Land Surface Models (PALS) Land Surface Model Benchmarking Evaluation Project (PLUMBER; Best et al. 2015) should greatly accelerate both the rate and knowledge gained through benchmarking of both water and nonwater parameters simulated by models. Ultimately, benchmarking against multiple data constraints crossing multiple biogeochemical cycles (e.g., water, carbon, nutrients) forces models to get the right answers for the right reasons, and is thus a powerful direction forward for ecohydrological modeling (e.g., Nearing et al. 2016).

The interactions between hydrology and biogeochemistry, specifically the water, carbon, and nutrient cycles, are a central component of ecohydrology. Nutrient availability, for example, is critical to growth of aquatic biota, soil microbes, and vegetation, and is simultaneously highly responsive to the hydrologic cycle

(Liu et al. 2008; Wang et al. 2015). The movement of nutrients such as nitrogen across land–water gradients is of growing concern, particularly with land-cover and climate changes (Burt et al. 2010). Such changes can have cascading impacts on trophic systems and water quality (Krause et al. 2011). Disturbances are particularly threatening to impact the nitrogen and other elemental cycles (Sollins and McCorison 1981), and thus a strong need for integrated research for prediction and mitigation is required under a future disturbance regime (McDowell et al. 2018).

Future ecohydrological research will benefit hydrology not only in addressing the linkages between vegetation, nutrient cycles, and water, but through an explicit focus on understanding ecosystem/watershed-scale mechanisms driving our observations. To achieve this, ecohydrology must continue to utilize cutting-edge techniques including remote sensing (described in section 3 of this paper), fine- and coarse-resolution models, and advanced monitoring and experimental techniques. The long history of cause-and-effect experiments (e.g., catchment disturbances) must continue to play a strong role, but can be refined to address future ecohydrological threats such as wildfires, insect outbreaks, and climate warming. With these advances, we can expect ecohydrology to continue to advance our knowledge and mitigation options for water and nonwater resources of human value under increasing future pressure.

7. Water management and water security

a. *The origins*

As already noted in the first section, the concept of the hydrologic cycle appears to have been known since ancient civilizations. As the population increased, so did the demand for a steady and reliable source of water. Water as a resource has thus been artificially “managed” ever since there was such demand for mankind. However, until the advent of hydrology as a proper scientific discipline, most water management practices around the world were relatively ad hoc and lacked sound hydrologic principles. For example, in ancient India, the amount rain in an area was recorded for each year and used as a proxy for estimating food production and taxation rate for the following year (Srinivasan 2000). In Sri Lanka, giant-sized reservoirs were built in the first century BC during the reign of King Wasabha (67–111 BC). According to historical records, the king built 11 large reservoirs and two irrigation canals of what is known today as perhaps the world’s oldest and surviving rainwater harvesting project (de Silva et al. 1995). Thus, the history of hydrology in water management is

long and has always been driven by societal needs for maintaining a steady supply of water.

b. Water management today

Today, water management owes its foundation to pioneers who developed hydrology as a science during early twentieth century. One particular pioneer who must be mentioned for his seminal role in spurring water management is Robert Horton, who performed scientific investigations to solve real-world problems. Horton (1941) had written about infiltration and runoff production that is commonly used today to express runoff generation process from precipitation in many of today's watershed management models. He had also written on erosion, geomorphology, basin response—all of which have directly contributed to the evolution toward physical hydrology-based engineering design of water management systems.

In the current computer era, the first use of digital computing in hydrology, although driven primarily by scientific investigations, was in the Stanford Watershed Model (Crawford and Linsley 1966) and the MIT Catchment Model (Harley 1971). These computer models offered hydrologists and water managers, an opportunity to look at the complex behavior of a river basin more holistically for decision making. Since these early computer models, there have been numerous others developed for hydrologic prediction in water management decision making for flood management (Abbott et al. 1986), irrigation management (Singh et al. 1999), reservoir operations (Yeh 1985), and water quality management (Abbaspour et al. 2007). For a historical overview of current computer models used for watershed management, Singh and Woolhiser (2002) provide a very comprehensive review.

With the advent of “dynamic hydrology” (Eagleson 1970) as a discipline, hydrology evolved after the 1980s as a more interdisciplinary topic with closer links to atmospheric science, groundwater science, plant biology, and climate (see sections 2 and 5). Land–atmosphere interactions were recognized for their importance and land surface hydrologic (computer) models were developed. These models, such as the Variable Infiltration Capacity (VIC) model (Liang et al. 1994), Noah (Chen et al. 1996; Ek et al. 2003), Noah-MP (Niu et al. 2011), and the Common Land Model (CLM; Dai et al. 2003), among others, opened doors for water managers to explore the role of climate and weather on water management. Unlike traditional hydrologic models, the atmospheric forcings are integrated with the land's response through energy and water fluxes. Such land surface models, including those that are coupled with water management models (e.g., Haddeland et al. 2006;

Hanasaki et al. 2006; Voisin et al. 2013), have thus been used in identifying best practices for land or irrigation management (Pielke et al. 2011; Ozdogan et al. 2010), water development and adaptation policy for climate change (Kundzewicz et al. 2008), and reservoir management (Hamlet 2011), just to name a few.

Most recently, with the advent of remote sensing from ground or space platforms (section 3) that can now provide estimates of key hydrologic variables on a global scale, hydrology has begun to experience much broader and more global application in water management. This is primarily because remote sensing from satellites is the only way to monitor changing fluxes of the water cycle in difficult or ungauged regions of the world. In what follows next, water management is broken down thematically into societal application topics.

1) RESERVOIR MANAGEMENT

Dams and artificial reservoirs are built to trap a sufficiently large amount of water from the hydrologic cycle to make up for a shortfall when demand for water exceeds the variable supply from nature. Using advancements in hydrologic science, much is now known about the management of postdam effects on aquatic ecology (e.g., Ligon et al. 1995; Richter et al. 1996; Zhang et al. 2019), geomorphology (e.g., Graf 2006), floods (e.g., Wang et al. 2017) and droughts (e.g., Wan et al. 2017), and sediment trapping by reservoirs (Graf et al. 2010). Such understanding has consequently improved water management practices for regulated river basins. Yeh (1985) provides a thorough review of the progress of quantitative water management practices that remain a cornerstone for practitioners today even after three decades.

In designing a dam's physical dimensions, the inflow design flood (IDF) is a major parameter that is derived from analyzing probability of occurrence of flood and precipitation events using historical hydrologic records (Hossain et al. 2010). Also, most of the large dams, especially the hazardous ones located upstream of population centers, are often designed considering the standard Probable Maximum Flood (PMF; Yigzaw et al. 2013). PMF, by its definition, is the hydrologic response as flow to the previously introduced concept of Probable Maximum Precipitation (PMP). WMO (2009) suggests several methods for PMP estimation: statistical method, generalized method, transposition method, and moisture maximization method (Rakhecha and Singh 2009). Ever since the wider availability of numerical atmospheric models and reanalysis data of the atmosphere, the dam design and reservoir management community is increasingly marching toward more atmospheric, science-based approaches to predict changing risks associated with

PMP and PMF (Chen and Hossain 2018; Rastogi et al. 2017; Chen et al. 2017; Rouhani and Leconte 2016; Ohara et al. 2011).

2) REMOTE SENSING APPLICATIONS IN WATER, FOOD, AND DISASTER MANAGEMENT

As indicated in section 3, from the early days of satellite precipitation remote sensing driven mostly by weather and climate science (Griffith et al. 1978; Arkin and Meisner 1987) to the modern era of the Global Precipitation Measurement (GPM) mission (Hou et al. 2014), the scientific community has made great strides in reducing uncertainty and improving resolution. Consequently, this has opened up a diverse set of applications over the last decade. The global nature of coherent and more accurate satellite precipitation products have now improved water management in river basins where rainfall is abundant but in situ measurement networks are generally inadequate. Building on the success of past satellite remote sensing missions for precipitation, we can now perform global-scale runoff/flood prediction (Wu et al. 2012, 2014), monitor drought/crop yield (Funk and Verdin 2010; McNally et al. 2017), provide irrigation advisory services (Hossain et al. 2017), and monitor landslide risks (Kirschbaum et al. 2012). Remote sensing applications and decision support systems have also been utilized in monitoring water supplies stored in snowpacks, drought impacts on agricultural production, and groundwater depletion (Schumann et al. 2016).

Transboundary flood forecasting is another area that has recently benefited from application of hydrologic prediction driven by remote sensing, particularly in developing countries (Hossain and Katiyar 2006). This is because in transboundary river basins, the lack of knowledge about the real-time hydrological state of the upstream nations makes floods more catastrophic than other places. Bakker (2009) has shown that the number of the international river basin floods (i.e., transboundary flood) is only 10% of the total riverine floods. With this small number of occurrences, transboundary floods are responsible for 32% of total casualties, and the affected individuals could be high as 60%. UN-Water (2008) reported that 40% of the global population lives in the 263 shared or transboundary lake or river basins. For transboundary basins, flood forecasting based on satellite remote sensing and hydrologic models has become one of the most economic and effective ways to mitigate floods (Wu et al. 2014). Given the plethora of satellite nadir altimetry sensors that can now measure river levels (*Jason-3*, *Sentinel-3A* and *-3B*, *IceSat-2*, *AltiKa*), it appears that altimetry usage with conventional flood forecasting systems will further

improve the management of floods. The impacts of food security are felt most seriously in developing countries where people practice subsistence farming. This is where accurate monitoring of growing season conditions can significantly help mitigate the effects of food insecurity in the developing world. These assessments are now being done using remotely sensed monitoring data for precipitation, crop water requirements, and vegetation indices, using hydrologic models and monitoring systems (Budde et al. 2010). For example, MODIS satellite data are now used in developing vegetation indices that provide consistent spatial and temporal comparisons of vegetation properties used to track drought conditions that may threaten subsistence agriculture (Budde et al. 2010).

c. Emerging issues

The current trend of expanding human settlements, economic activity, population increase, and climate change mean that water will continue to get redistributed and artificially managed to the extent that there will be no pristine river basin left today without the human footprint caused by water diversions, barrages, dams, and irrigation projects (Zarfl et al. 2015; Kumar 2015). The evidence is already there. For example, USGS records indicate an increase in irrigation acreage from 35 million acres (1950) to 65 million acres (in 2005) in the United States alone (Kenny et al. 2009). The latter is equivalent to a withdrawal of 144 million acre-feet (or 177 km³) of surface and groundwater per year. Similarly, there are about 75 000 artificial reservoirs built in the United States alone during the last century with a total storage capacity almost equaling one year's mean runoff (Graf 1999). Around the world, the number of impoundments in populated regions is more staggering and exploding due to needs for economic development (Zarfl et al. 2015).

Studies now clearly show that the regulation of rivers by dams built by upstream nations and the ensuing lack of connectivity between river reaches or the increased time water remains stagnant (in reservoirs) will be most severe in the mid-twenty-first century (Grill et al. 2015). However, the impact on availability of freshwater, which also drives food and energy production, cannot be monitored and managed by downstream nations of such transboundary river basin using conventional approaches to water management.

8. Future directions

As we move forward in the twenty-first century, the expansion of human settlements, economic activity, and increasing population mean that water availability will

continue to increase in importance. As already evident in the previous sections, scientists, engineers, planners, and decision-makers will be dealing more and more with a “human–water cycle.” This human–water cycle will represent active interplay between humans and nature, therefore inviting new and exciting dimensions to hydrology in the coming decades (Wheater and Gober 2015). Within hydrologic science, there is increasing recognition of the coevolution of natural and anthropogenic landscape features and the hydrological response of catchments, and this concept has been termed “catchment coevolution” (Sivapalan and Blöschl 2017), with the coevolution era projected from 2010 to 2030. Advances in hydrologic modeling will continue, supported by more observational data available to constrain the model, improved understanding of hydrologic processes and incorporation of key processes and new approaches, and comprehensive benchmarking of models (Clark et al. 2015). Coincident with these trends is the new era of “big data” in which computational and theoretical advances are ushering in new learning opportunities, as discussed in Peters-Lidard et al. (2017). Combining big data with new observational platforms, as described in McCabe et al. (2017), will yield important new insights and societal benefits.

In the 50-year anniversary celebration of *WRR*, Alberto Montanari et al. state that “Water science will play an increasingly important role for the benefit of humanity during the next decades, as water will be the key to ensuring adequate food and energy resources for future generations” (Montanari et al. 2015). They go on to exhort the community that the “target for hydrology in the 21st century must be ambitious. There are relevant and global water problems to solve and there is a compelling need to ensure sustainable development of the human community.”

We are already witnessing some of this “ambition” to solve grand challenge societal problems through the assimilation of climate, weather, numerical modeling, and remote sensing into tangible solutions for society. Some of the most exciting prospects for advancing hydrologic science exist at the interfaces with other scientific disciplines, for example, plant biology and ecology for crop yield and ecosystem modeling, oceanography for estuarine process modeling, biogeochemistry for understanding the interactions between carbon and water cycles, and socioeconomics for integrating human and water systems (Vogel et al. 2015). All of these advances will likely converge to improve understanding and modeling of the Earth system, leading to improvements in weather and climate predictions that exploit land memory from a spectrum of interconnected processes of surface and subsurface hydrology, vegetation,

biogeochemistry, and human activities (e.g., irrigation). In the climate projection arena, Bierkens (2015) posits that physically based continental Earth system models (PBCESMs) will converge to support integrated assessments, including, for example, groundwater (e.g., Fan 2015). For example, in the water management area, there now exists operational satellite remote sensing–based transboundary flood forecasting systems that provide valuable updates of flood risk around the world (Wu et al. 2014; Alfieri et al. 2013).

However, the human–water cycle is not the only area that needs to experience growth for the future of hydrology. In a recent review of progress and future directions for hydrologic modeling, Singh (2018) cites other areas that need to be studied, such as hydrologic impacts of hydraulic fracturing, transport of biochemical and microorganisms in the soil, hydrology of hurricanes and atmospheric rivers, and sociohydrology. The review goes on to state that “For management of hydrologic systems, political, economic, legal, social, cultural, and management aspects will need to be integrated” where “both hydrologic science and engineering applications are equally emphasized.”

At the fundamental process level, studies involving isotopes have revealed a complexity of the movement and distribution of water particles in time and space where many of the dynamic connections and disconnections of water stored in the ground remain unexplained today (McDonnell 2017). For example, at the hillslope scale, the movement of water is often compartmentalized. Runoff from snowmelt can often be from precipitation snowpack that occurred several years earlier. There is clear evidence that plants often remove water through transpiration from immobile pools underground that are not tightly coupled to the infiltration and groundwater recharge processes being modeled today (Brooks et al. 2010). With such process-based questions on hydrology remaining unexplained today, McDonnell argues that future directions in hydrology should also require thinking of newer frameworks that can track both flow and the age of water.

One likely direction toward which hydrology seems to be already evolving is in the area of “nexuses” of resources or themes—such as food–energy–water (the FEW nexus) or climate–energy–water (the CEW nexus) and even the sociology–hydrology nexus (sociohydrology). There is no doubt that the future direction of hydrology will be increasingly more multi- and interdisciplinary and draw in fields that have traditionally never interacted with hydrology. For example, freshwater access and nutrition are the foundation pillars of public health. Lack of safe water and sanitation access and malnutrition are intricately linked to water and food

security and can be critical factors behind child mortality and morbidity anywhere. To tackle challenges due to compounding factors of lack of sanitation/safe water and nutrition, health management will need to partner closely with agricultural and water management and naturally require strong collaboration from the hydrologic community.

In addressing the ensuing challenges for managing the water, a piecemeal approach to hydrology research or investigation will not suffice anymore. To keep water management practices timely and relevant, hydrologic research will most likely be converted into language that encourages uptake by policy-makers, practitioners, and the public in the real world out of sheer necessity in the twenty-first century (Cosgrove and Loucks 2015; Wheeler and Gober 2015).

Acknowledgments. The authors gratefully acknowledge the support from their institutions, in addition to NASA's Energy and Water Cycle Study (NEWS) program (MR), and Spain's Ministry of Economy and Competitiveness (FT). LRL was supported by the U.S. Department of Energy Office of Science Biological and Environmental Research as part of the Regional and Global Climate Modeling Program. NGM was supported by Pacific Northwest National Labs' LDRD program. Pacific Northwest National Laboratory is operated by Battelle for the U.S. Department of Energy under Contract DE-AC05-76RLO1830. The work by FJT was carried out at the Jet Propulsion Laboratory, California Institute of Technology, under a contract with NASA. CP-L gratefully acknowledges Ms. Brynne Norton of the NASA/GSFC Library for her assistance with compiling reference databases. Further, we acknowledge the support of AMS publications staff for their assistance with reference checking and figure permissions.

REFERENCES

- Abbaspour, K. C., J. Yang, I. Maximov, R. Siber, K. Bogner, J. Mieleitner, J. Zobrist, and R. Srinivasan, 2007: Modelling hydrology and water quality in the pre-alpine/alpine Thur watershed using SWAT. *J. Hydrol.*, **333**, 413–430, <https://doi.org/10.1016/j.jhydrol.2006.09.014>.
- Abbott, M. B., J. C. Bathurst, J. A. Cunge, P. E. O'Connell, and J. Rasmussen, 1986: An introduction to the European Hydrological System — Systeme Hydrologique Europeen, "SHE", 1: History and philosophy of a physically-based, distributed modelling system. *J. Hydrol.*, **87**, 45–59, [https://doi.org/10.1016/0022-1694\(86\)90114-9](https://doi.org/10.1016/0022-1694(86)90114-9).
- Adams, H. D., A. K. MacAlady, D. D. Breshears, C. D. Allen, N. L. Stephenson, S. R. Saleska, T. E. Huxman, and N. G. McDowell, 2010: Climate-induced tree mortality: Earth system consequences. *Eos, Trans. Amer. Geophys. Union*, **91**, 153–154, <https://doi.org/10.1029/2010EO170003>.
- , C. H. Luce, D. D. Breshears, C. D. Allen, M. Weiler, V. C. Hale, A. M. S. Smith, and T. E. Huxman, 2012: Ecohydrological consequences of drought- and infestation- triggered tree die-off: Insights and hypotheses. *Ecohydrology*, **5**, 145–159, <https://doi.org/10.1002/eco.233>.
- Adams, T. E., and T. C. Pagano, 2016: *Flood Forecasting: A Global Perspective*. Elsevier, 478 pp.
- Adler, R. F., and A. J. Negri, 1988: A Satellite infrared technique to estimate tropical convective and stratiform rainfall. *J. Appl. Meteor.*, **27**, 30–51, [https://doi.org/10.1175/1520-0450\(1988\)027<0030:ASITTE>2.0.CO;2](https://doi.org/10.1175/1520-0450(1988)027<0030:ASITTE>2.0.CO;2).
- , G. J. Huffman, D. T. Bolvin, S. Curtis, and E. J. Nelkin, 2000: Tropical rainfall distributions determined using trmm combined with other satellite and rain gauge information. *J. Appl. Meteor.*, **39**, 2007–2023, [https://doi.org/10.1175/1520-0450\(2001\)040<2007:TRDDUT>2.0.CO;2](https://doi.org/10.1175/1520-0450(2001)040<2007:TRDDUT>2.0.CO;2).
- , and Coauthors, 2003: The version-2 Global Precipitation Climatology Project (GPCP) monthly precipitation analysis (1979–present). *J. Hydrometeorol.*, **4**, 1147–1167, [https://doi.org/10.1175/1525-7541\(2003\)004<1147:TVGPCP>2.0.CO;2](https://doi.org/10.1175/1525-7541(2003)004<1147:TVGPCP>2.0.CO;2).
- , G. Gu, and G. J. Huffman, 2012: Estimating climatological bias errors for the Global Precipitation Climatology Project (GPCP). *J. Appl. Meteor. Climatol.*, **51**, 84–99, <https://doi.org/10.1175/JAMC-D-11-052.1>.
- , —, M. Sapiano, J.-J. Wang, and G. J. Huffman, 2017: Global precipitation: Means, variations and trends during the satellite era (1979–2014). *Surv. Geophys.*, **38**, 679–699, <https://doi.org/10.1007/s10712-017-9416-4>.
- Alcamo, J., P. Döll, T. Henrichs, F. Kaspar, B. Lehner, T. Rösch, and S. Siebert, 2003: Development and testing of the WaterGAP 2 global model of water use and availability. *Hydrol. Sci. J.*, **48**, 317–338, <https://doi.org/10.1623/hysj.48.3.317.45290>.
- Alfieri, L., P. Burek, E. Dutra, B. Krzeminski, D. Muraro, J. Thielen, and F. Pappenberger, 2013: GloFAS-global ensemble streamflow forecasting and flood early warning. *Hydrol. Earth Syst. Sci.*, **17**, 1161–1175, <https://doi.org/10.5194/hess-17-1161-2013>.
- Alkama, R., and A. Cescatti, 2016: Biophysical climate impacts of recent changes in global forest cover. *Science*, **351**, 600–604, <https://doi.org/10.1126/science.aac8083>.
- Allen, C. D., D. D. Breshears, and N. G. McDowell, 2015: On underestimation of global vulnerability to tree mortality and forest die-off from hotter drought in the Anthropocene. *Ecosphere*, **6**, <https://doi.org/10.1890/ES15-00203.1>.
- Allen, M. R., W. J. Ingram, and D. Stainforth, 2002: Constraints on future changes in climate and the hydrologic cycle. *Nature*, **419**, 224–232.
- Allen, R. G., L. S. Pereira, D. Raes, and M. Smith, 1998: Crop evapotranspiration: Guidelines for computing crop water requirements. FAO Irrigation and Drainage Paper 56, 300 pp., <http://www.fao.org/3/X0490E/X0490E00.htm>.
- , and Coauthors, 2007: Satellite-based energy balance for Mapping Evapotranspiration with Internalized Calibration (METRIC)—Applications. *J. Irrig. Drain. Eng.*, **133**, 395–406, [https://doi.org/10.1061/\(ASCE\)0733-9437\(2007\)133:4\(395\)](https://doi.org/10.1061/(ASCE)0733-9437(2007)133:4(395)).
- Alley, W. M., and L. F. Konikow, 2015: Bringing GRACE down to Earth. *Groundwater*, **53**, 826–829, <https://doi.org/10.1111/gwat.12379>.
- Alsdorf, D. E., E. Rodríguez, and D. P. Lettenmaier, 2007: Measuring surface water from space. *Rev. Geophys.*, **45**, RG2002, <https://doi.org/10.1029/2006RG000197>.
- Amorochio, J., and B. Wu, 1977: Mathematical models for the simulation of cyclonic storm sequences and precipitation

- fields. *J. Hydrol.*, **32**, 329–345, [https://doi.org/10.1016/0022-1694\(77\)90025-7](https://doi.org/10.1016/0022-1694(77)90025-7).
- Anderson, E. A., 1972: National Weather Service River Forecast System forecast procedures. NOAA Tech. Memo. NWS HYDRO-14, 264 pp.
- Anderson, M. C., J. M. Norman, G. R. Diak, W. P. Kustas, and J. R. Mecikalski, 1997: A two-source time-integrated model for estimating surface fluxes using thermal infrared remote sensing. *Remote Sens. Environ.*, **60**, 195–216, [https://doi.org/10.1016/S0034-4257\(96\)00215-5](https://doi.org/10.1016/S0034-4257(96)00215-5).
- , and Coauthors, 2011: Mapping daily evapotranspiration at field to continental scales using geostationary and polar orbiting satellite imagery. *Hydrol. Earth Syst. Sci.*, **15**, 223–239, <https://doi.org/10.5194/hess-15-223-2011>.
- Anderson, M. G., and T. P. Burt, 1985: *Hydrological Forecasting*. Wiley, 604 pp.
- Anyah, R. O., C. P. Weaver, G. Miguez-Macho, Y. Fan, and A. Robock, 2008: Incorporating water table dynamics in climate modeling: 3. Simulated groundwater influence on coupled land-atmosphere variability. *J. Geophys. Res.*, **113**, D07103, <https://doi.org/10.1029/2007JD009087>.
- Arkin, P. A., and B. N. Meisner, 1987: The relationship between large-scale convective rainfall and cold cloud over the Western Hemisphere during 1982–84. *Mon. Wea. Rev.*, **115**, 51–74, [https://doi.org/10.1175/1520-0493\(1987\)115<0051:TRBLSC>2.0.CO;2](https://doi.org/10.1175/1520-0493(1987)115<0051:TRBLSC>2.0.CO;2).
- Ashouri, H., K. Hsu, S. Sorooshian, D. K. Braithwaite, K. R. Knapp, L. D. Cecil, B. R. Nelson, and O. P. Prat, 2015: PERSIANN-CDR: Daily precipitation climate data record from multisatellite observations for hydrological and climate studies. *Bull. Amer. Meteor. Soc.*, **96**, 69–83, <https://doi.org/10.1175/BAMS-D-13-00068.1>.
- Baker, M. N., and R. E. Horton, 1936: Historical development of ideas regarding the origin of springs and ground-water. *Eos, Trans. Amer. Geophys. Union*, **17**, 395–400, <https://doi.org/10.1029/TR017i002p00395>.
- Bakker, M. H. N., 2009: Transboundary river floods: Examining countries, international river basins and continents. *Water Policy*, **11**, 269–288, <https://doi.org/10.2166/wp.2009.041>.
- Baldocchi, D., and Coauthors, 2001: FLUXNET: A new tool to study the temporal and spatial variability of ecosystem-scale carbon dioxide, water vapor, and energy flux densities. *Bull. Amer. Meteor. Soc.*, **82**, 2415–2434, [https://doi.org/10.1175/1520-0477\(2001\)082<2415:FANTTS>2.3.CO;2](https://doi.org/10.1175/1520-0477(2001)082<2415:FANTTS>2.3.CO;2).
- Ball, J. T., I. E. Woodrow, and J. A. Berry, 1987: A model predicting stomatal conductance and its contribution to the control of photosynthesis under different environmental conditions. *Progress in Photosynthesis Research*, Vol. 4, J. Biggins, Ed., Springer, 221–224, https://doi.org/10.1007/978-94-017-0519-6_48.
- Barré de Saint-Venant, A. J. C., 1871: Theorie du mouvement non permanent des eaux, avec application aux crues des rivières et à l'introduction de mares dans leurs lits. *C. R. Hebd. Seances Acad. Sci.*, **73**, 148–154.
- Bastiaanssen, W. G. M., H. Pelgrum, J. Wang, Y. Ma, J. F. Moreno, G. J. Roerink, and T. van der Wal, 1998: A remote sensing Surface Energy Balance Algorithm for Land (SEBAL). *J. Hydrol.*, **212–213**, 213–229, [https://doi.org/10.1016/S0022-1694\(98\)00254-6](https://doi.org/10.1016/S0022-1694(98)00254-6).
- Baumgartner, A., and E. Reichel, 1973: Eine neue Bilanz des globalen Wasserkreislaufes. *Umschau*, **20**, 631–632.
- , and —, 1975: *The World Water Balance: Mean Annual Global, Continental and Maritimes Precipitation, Evaporation and Run-Off*. Elsevier, 179 pp.
- Bear, J., 1972: *Dynamics of Fluids in Porous Media*. Elsevier, 764 pp.
- Bearup, L. A., R. M. Maxwell, and J. E. Mccray, 2016: Hillslope response to insect-induced land-cover change: An integrated model of end-member mixing. *Ecohydrology*, **9**, 195–203, <https://doi.org/10.1002/eco.1729>.
- Beaumont, R. T., 1965: Mt. Hood pressure pillow snow gage. *J. Appl. Meteor.*, **4**, 626–631, [https://doi.org/10.1175/1520-0450\(1965\)004<0626:MHPPSG>2.0.CO;2](https://doi.org/10.1175/1520-0450(1965)004<0626:MHPPSG>2.0.CO;2).
- Behrangi, A., K. Hsu, B. Imam, S. Sorooshian, G. J. Huffman, and R. J. Kuligowski, 2009: PERSIANN-MSA: A precipitation estimation method from satellite-based multispectral analysis. *J. Hydrometeorol.*, **10**, 1414–1429, <https://doi.org/10.1175/2009JHM1139.1>.
- Beljaars, A. C. M., and A. A. M. Holtslag, 1991: Flux parameterization over land surfaces for atmospheric models. *J. Appl. Meteor.*, **30**, 327–341, [https://doi.org/10.1175/1520-0450\(1991\)030<0327:FPOLSF>2.0.CO;2](https://doi.org/10.1175/1520-0450(1991)030<0327:FPOLSF>2.0.CO;2).
- , P. Viterbo, M. J. Miller, and A. K. Betts, 1996: The anomalous rainfall over the United States during July 1993: Sensitivity to land surface parameterization and soil moisture anomalies. *Mon. Wea. Rev.*, **124**, 362–383, [https://doi.org/10.1175/1520-0493\(1996\)124<0362:TAROTU>2.0.CO;2](https://doi.org/10.1175/1520-0493(1996)124<0362:TAROTU>2.0.CO;2).
- Bennett, A. C., N. G. McDowell, C. D. Allen, and K. J. Anderson-Teixeira, 2015: Larger trees suffer most during drought in forests worldwide. *Nat. Plants*, **1**, <https://doi.org/10.1038/nplants.2015.139>.
- Bennett, K. E., T. J. Bohn, K. Solander, N. G. McDowell, C. Xu, E. Vivoni, and R. S. Middleton, 2018: Climate-driven disturbances in the San Juan River sub-basin of the Colorado River. *Hydrol. Earth Syst. Sci.*, **22**, 709–725, <https://doi.org/10.5194/hess-22-709-2018>.
- Berg, A., and Coauthors, 2016: Land-atmosphere feedbacks amplify aridity increase over land under global warming. *Nat. Climate Change*, **6**, 869–874, <https://doi.org/10.1038/nclimate3029>.
- Berne, A., R. Uijlenhoet, and P. A. Troch, 2005: Similarity analysis of subsurface flow response of hillslopes with complex geometry. *Water Resour. Res.*, **41**, W09410, <https://doi.org/10.1029/2004WR003629>.
- Berner, E. K., and R. A. Berner, 1987: *The Global Water Cycle: Geochemistry and Environment*. Prentice-Hall, 397 pp.
- Berry, Z. C., and Coauthors, 2018: The two water worlds hypothesis: Addressing multiple working hypotheses and proposing a way forward. *Ecohydrology*, **11**, e1843, <https://doi.org/10.1002/eco.1843>.
- Beschta, R. L., M. R. Pyles, A. E. Skaugset, and C. G. Surfleet, 2000: Peakflow responses to forest practices in the western cascades of Oregon, USA. *J. Hydrol.*, **233**, 102–120, [https://doi.org/10.1016/S0022-1694\(00\)00231-6](https://doi.org/10.1016/S0022-1694(00)00231-6).
- Best, M. J., and Coauthors, 2011: The Joint UK Land Environment Simulator (JULES), model description – Part 1: Energy and water fluxes. *Geosci. Model Dev.*, **4**, 677–699, <https://doi.org/10.5194/gmd-4-677-2011>.
- , and Coauthors, 2015: The plumbing of land surface models: Benchmarking model performance. *J. Hydrometeorol.*, **16**, 1425–1442, <https://doi.org/10.1175/JHM-D-14-0158.1>.
- Betts, A. K., 2004: Understanding hydrometeorology using global models. *Bull. Amer. Meteor. Soc.*, **85**, 1673–1688, <https://doi.org/10.1175/BAMS-85-11-1673>.
- , J. H. Ball, A. C. M. Beljaars, M. J. Miller, and P. A. Viterbo, 1996: The land surface-atmosphere interaction: A review based on observational and global modeling perspectives. *J. Geophys. Res.*, **101**, 7209–7225, <https://doi.org/10.1029/95JD02135>.

- Beven, K. J., and M. J. Kirkby, 1979: A physically based, variable contributing area model of basin hydrology. *Hydrol. Sci. Bull.*, **24**, 43–69, <https://doi.org/10.1080/02626667909491834>.
- , H. Cloke, F. Pappenberger, R. Lamb, and N. Hunter, 2015: Hyperresolution information and hyperresolution ignorance in modelling the hydrology of the land surface. *Sci. China Earth Sci.*, **58**, 25–35, <https://doi.org/10.1007/s11430-014-5003-4>.
- Biancamaria, S., D. P. Lettenmaier, and T. M. Pavelsky, 2016: The SWOT mission and its capabilities for land hydrology. *Surv. Geophys.*, **37**, 307–337, <https://doi.org/10.1007/s10712-015-9346-y>.
- Bierkens, M. F. P., 2015: Global hydrology 2015: State, trends, and directions. *Water Resour. Res.*, **51**, 4923–4947, <https://doi.org/10.1002/2015WR017173>.
- , and Coauthors, 2015: Hyper-resolution global hydrological modelling: what is next? *Hydrol. Processes*, **29**, 310–320, <https://doi.org/10.1002/hyp.10391>.
- Binley, A., S. S. Hubbard, J. A. Huisman, A. Revil, D. A. Robinson, K. Singha, and L. D. Slater, 2015: The emergence of hydrogeophysics for improved understanding of subsurface processes over multiple scales. *Water Resour. Res.*, **51**, 3837–3866, <https://doi.org/10.1002/2015WR017016>.
- Birkett, C. M., 1998: Contribution of the TOPEX NASA radar altimeter to the global monitoring of large rivers and wetlands. *Water Resour. Res.*, **34**, 1223–1239, <https://doi.org/10.1029/98WR00124>.
- , C. Reynolds, B. Beckley, and B. Doorn, 2011: From research to operations: The USDA global reservoir and lake monitor. *Coastal Altimetry*, S. Vignudelli et al., Eds., Springer, 19–50.
- Blaney, H. F., and W. D. Criddle, 1950: Determining water requirements in irrigated areas from climatological and irrigation data. USDA Rep. SCS-TP-96, 50 pp.
- Blöschl, G., 1999: Scaling issues in snow hydrology. *Hydrol. Processes*, **13**, 2149–2175, [https://doi.org/10.1002/\(SICI\)1099-1085\(199910\)13:14/15<2149::AID-HYP847>3.0.CO;2-8](https://doi.org/10.1002/(SICI)1099-1085(199910)13:14/15<2149::AID-HYP847>3.0.CO;2-8).
- , and M. Sivapalan, 1995: Scale issues in hydrological modeling: A review. *Hydrol. Processes*, **9**, 251–290, <https://doi.org/10.1002/hyp.3360090305>.
- Boening, C., J. K. Willis, F. W. Landerer, S. Nerem, and J. Fasullo, 2012: The 2011 La Niña: So strong, the oceans fell. *Geophys. Res. Lett.*, **39**, L19602, <https://doi.org/10.1029/2012GL053055>.
- Bogena, H. R., J. A. Huisman, C. Oberdörster, and H. Vereecken, 2007: Evaluation of a low-cost soil water content sensor for wireless network applications. *J. Hydrol.*, **344**, 32–42, <https://doi.org/10.1016/j.jhydrol.2007.06.032>.
- Bonfils, C., and D. Lobell, 2007: Empirical evidence for a recent slowdown in irrigation-induced cooling. *Proc. Natl. Acad. Sci. USA*, **104**, 13 582–13 587, <https://doi.org/10.1073/pnas.0700144104>.
- Bonnin, G. M., D. Martin, B. Lin, T. Parzybok, M. Yekta, and D. Riley, 2006: Version 3.0: Delaware, District of Columbia, Illinois, Indiana, Kentucky, Maryland, New Jersey, North Carolina, Ohio, Pennsylvania, South Carolina, Tennessee, Virginia, West Virginia. Vol. 2, Precipitation-Frequency Atlas of the United States, NOAA Atlas 14, 295 pp., http://www.nws.noaa.gov/oh/hdsc/PF_documents/Atlas14_Volume2.pdf.
- Bosch, J. M., and J. D. Hewlett, 1982: A review of catchment experiments to determine the effect of vegetation changes on water yield and evapotranspiration. *J. Hydrol.*, **55**, 3–23, [https://doi.org/10.1016/0022-1694\(82\)90117-2](https://doi.org/10.1016/0022-1694(82)90117-2).
- Bosilovich, M. G., S. D. Schubert, and G. K. Walker, 2005: Global changes of the water cycle intensity. *J. Climate*, **18**, 1591–1608, <https://doi.org/10.1175/JCLI3357.1>.
- Bouchet, R. J., 1963: Évapotranspiration réelle et potentielle signification climatique. *IAHS Publ.*, **62**, 134–162.
- Brandes, E. A., K. Ikeda, G. Zhang, M. Schönhuber, and R. M. Rasmussen, 2007: A statistical and physical description of hydrometeor distributions in Colorado snowstorms using a video disdrometer. *J. Appl. Meteor. Climatol.*, **46**, 634–650, <https://doi.org/10.1175/JAM2489.1>.
- Brantley, S. L., and Coauthors, 2017: Reviews and syntheses: On the roles trees play in building and plumbing the critical zone. *Biogeosciences*, **14**, 5115–5142, <https://doi.org/10.5194/bg-14-5115-2017>.
- Bringi, V. N., V. Chandrasekar, N. Balakrishnan, and D. S. Zrníc, 1990: An examination of propagation effects in rainfall on radar measurements at microwave frequencies. *J. Atmos. Oceanic Technol.*, **7**, 829–840, [https://doi.org/10.1175/1520-0426\(1990\)007<0829:AEOPEI>2.0.CO;2](https://doi.org/10.1175/1520-0426(1990)007<0829:AEOPEI>2.0.CO;2).
- Brooks, J. R., H. R. Barnard, R. Coulombe, and J. J. McDonnell, 2010: Ecohydrologic separation of water between trees and streams in a Mediterranean climate. *Nat. Geosci.*, **3**, 100–104, <https://doi.org/10.1038/ngeo722>.
- Brooks, R. H., and A. T. Corey, 1964: Hydraulic properties of porous media. Colorado State University Hydrology Paper 3, 27 pp., https://dspace.library.colostate.edu/bitstream/handle/10217/61288/HydrologyPapers_n3.pdf?sequence=1.
- Brown, R. D., and D. A. Robinson, 2011: Northern Hemisphere spring snow cover variability and change over 1922–2010 including an assessment of uncertainty. *Cryosphere*, **5**, 219–229, <https://doi.org/10.5194/tc-5-219-2011>.
- Brubaker, K. L., D. Entekhabi, and P. S. Eagleson, 1993: Estimation of continental precipitation recycling. *J. Climate*, **6**, 1077–1089, [https://doi.org/10.1175/1520-0442\(1993\)006<1077:EOCPR>2.0.CO;2](https://doi.org/10.1175/1520-0442(1993)006<1077:EOCPR>2.0.CO;2).
- Brückner, E., 1905: Die Bilanz des Kreislaufs des Wassers auf der Erde. *Geogr. Z.*, **11**, 436–445.
- Brunner, G. W., 2016: HEC-RAS River Analysis System: Hydraulic Reference Manual, version 5.0. USACE Doc. CPD-69, 538 pp., [http://www.hec.usace.army.mil/software/hec-ras/documentation/HEC-RAS 5.0 Reference Manual.pdf](http://www.hec.usace.army.mil/software/hec-ras/documentation/HEC-RAS%205.0%20Reference%20Manual.pdf).
- Brunner, P., R. Therrien, P. Renard, C. T. Simmons, and H. J. H. Franssen, 2017: Advances in understanding river-groundwater interactions. *Rev. Geophys.*, **55**, 818–854, <https://doi.org/10.1002/2017RG000556>.
- Brutsaert, W., 1982: *Evaporation into the Atmosphere*. Springer, 302 pp.
- Budde, M. E., J. Rowland, and C. C. Funk, 2010: Agriculture and food availability – Remote sensing of agriculture for food security monitoring in the developing world. *Earthzine*, <https://earthzine.org/agriculture-and-food-availability-remote-sensing-of-agriculture-for-food-security-monitoring-in-the-developing-world/>.
- Budyko, M. I., and A. A. Sokolov, 1978: Water balance of the Earth. *World Water Balance and Water Resources of the Earth*, V. I. Korzun, Ed., UNESCO, 586–591.
- Burnash, R. J. C., R. L. Ferral, and R. A. McGuire, 1973: A generalized streamflow simulation system: Conceptual models for digital computers. Joint Federal and State River Forecast Center, U.S. National Weather Service, and California Department of Water Resources Tech. Rep., 204 pp.
- Burt, T., G. Pinay, and S. Sabater, 2010: What do we still need to know about the ecohydrology of riparian zones? *Ecohydrology*, **3**, 373–377, <https://doi.org/10.1002/eco.140>.
- Carpenter, T. M., J. A. Sperfslage, K. P. Georgakakos, T. Sweeney, and D. L. Fread, 1999: National threshold runoff estimation utilizing GIS in support of operational flash flood warning

- systems. *J. Hydrol.*, **224**, 21–44, [https://doi.org/10.1016/S0022-1694\(99\)00115-8](https://doi.org/10.1016/S0022-1694(99)00115-8).
- Carroll, S. S., and T. R. Carroll, 1989: Effect of uneven snow cover on airborne snow water equivalent estimates obtained by measuring terrestrial gamma radiation. *Water Resour. Res.*, **25**, 1505–1510, <https://doi.org/10.1029/WR025i007p01505>.
- Carroll, T., 2001: Airborne Gamma Radiation Snow Survey Program: A User's guide version 5.0. National Operational Hydrologic Remote Sensing Center, 14 pp., <https://www.nohrsc.noaa.gov/special/tom/gamma50.pdf>.
- Celia, M. A., E. T. Bouloutas, and R. L. Zarba, 1990: A general mass-conservative numerical solution for the unsaturated flow equation. *Water Resour. Res.*, **26**, 1483–1496, <https://doi.org/10.1029/WR026i007p01483>.
- Chahine, M. T., 1992: The hydrological cycle and its influence on climate. *Nature*, **359**, 373–380, <https://doi.org/10.1038/359373a0>.
- Chandra, S., 1990: *Hydrology in Ancient India*. National Institute of Hydrology, 106 pp.
- Chaney, N. W., P. Metcalfe, and E. F. Wood, 2016: HydroBlocks: a field-scale resolving land surface model for application over continental extents. *Hydrol. Processes*, **30**, 3543–3559, <https://doi.org/10.1002/hyp.10891>.
- Chang, A. T. C., J. L. Foster, and D. K. Hall, 1987: Nimbus-7 SMMR derived global snow cover parameters. *Ann. Glaciol.*, **9**, 39–44, <https://doi.org/10.1017/S0260305500000355>.
- Chen, F., and Coauthors, 1996: Modeling of land surface evaporation by four schemes and comparison with FIFE observations. *J. Geophys. Res.*, **101**, 7251–7268, <https://doi.org/10.1029/95JD02165>.
- Chen, J. L., C. R. Wilson, B. D. Tapley, D. D. Blankenship, and E. R. Ivins, 2007: Patagonia icefield melting observed by Gravity Recovery and Climate Experiment (GRACE). *Geophys. Res. Lett.*, **34**, L22501, <https://doi.org/10.1029/2007GL031871>.
- , J. S. Famiglietti, B. R. Scanlon, and M. Rodell, 2016: Groundwater storage changes: Present status from GRACE observations. *Surv. Geophys.*, **37**, 397–417, <https://doi.org/10.1007/s10712-015-9332-4>.
- Chen, X., and F. Hossain, 2018: Understanding model-based probable maximum precipitation estimation as a function of location and season from atmospheric reanalysis. *J. Hydrometeorol.*, **19**, 459–475, <https://doi.org/10.1175/JHM-D-17-0170.1>.
- , —, and L. R. Leung, 2017: Probable maximum precipitation in the U.S. Pacific Northwest in a changing climate. *Water Resour. Res.*, **53**, 9600–9622, <https://doi.org/10.1002/2017WR021094>.
- Chow, V. T., 1959: *Open-Channel Hydraulics*. McGraw Hill, 680 pp.
- , 1964: *Handbook of Applied Hydrology: A Compendium of Water-Resources Technology*. McGraw-Hill, 1468 pp.
- Clapp, R. B., and G. M. Hornberger, 1978: Empirical equations for some soil hydraulic properties. *Water Resour. Res.*, **14**, 601–604, <https://doi.org/10.1029/WR014i004p00601>.
- Clark, M. P., and R. B. Hanson, 2017: The citation impact of hydrology journals. *Water Resour. Res.*, **53**, 4533–4541, <https://doi.org/10.1002/2017WR021125>.
- , and Coauthors, 2015: Improving the representation of hydrologic processes in Earth system models. *Water Resour. Res.*, **51**, 5929–5956, <https://doi.org/10.1002/2015WR017096>.
- , and Coauthors, 2017: The evolution of process-based hydrologic models: Historical challenges and the collective quest for physical realism. *Hydrol. Earth Syst. Sci.*, **21**, 3427–3440, <https://doi.org/10.5194/hess-21-3427-2017>.
- Cline, D., R. Armstrong, R. Davis, K. Elder, and G. Liston, 2004: CLPX Snow Pit Measurements. National Snow and Ice Data Center, accessed 10 April 2018, https://nsidc.org/data/clpx/clpx_pits.gd.html.
- Coenders-Gerrits, A. M. J., R. J. van der Ent, T. A. Bogaard, L. Wang-Erlandsson, M. Hrachowitz, and H. H. G. Savenije, 2014: Uncertainties in transpiration estimates. *Nature*, **506**, E1–E2, <https://doi.org/10.1038/nature12925>.
- Cosgrove, W. J., and D. P. Loucks, 2015: Water management: Current and future challenges and research directions. *Water Resour. Res.*, **51**, 4823–4839, <https://doi.org/10.1002/2014WR016869>.
- Crawford, N. H., and R. K. Linsley, 1966: Digital simulation in hydrology; Stanford watershed model IV. Tech. Rep. 39, Dept. of Civil Engineering, Stanford University, 210 pp.
- Crow, W. T., and Coauthors, 2012: Upscaling sparse ground-based soil moisture observations for the validation of coarse-resolution satellite soil moisture products. *Rev. Geophys.*, **50**, RG2002, <https://doi.org/10.1029/2011RG000372>.
- Cunge, J. A., 1969: On the subject of a flood propagation computation method (Muskingum method). *J. Hydraul. Res.*, **7**, 205–230, <https://doi.org/10.1080/00221686909500264>.
- Dai, Y., and Coauthors, 2003: The Common Land Model. *Bull. Amer. Meteor. Soc.*, **84**, 1013–1023, <https://doi.org/10.1175/BAMS-84-8-1013>.
- Daliakopoulos, I. N., P. Coulibaly, and I. K. Tsanis, 2005: Groundwater level forecasting using artificial neural networks. *J. Hydrol.*, **309**, 229–240, <https://doi.org/10.1016/j.jhydrol.2004.12.001>.
- Daly, C., R. P. Neilson, and D. L. Phillips, 1994: A statistical-topographic model for mapping climatological precipitation over mountainous terrain. *J. Appl. Meteor.*, **33**, 140–158, [https://doi.org/10.1175/1520-0450\(1994\)033<0140:ASTMFM>2.0.CO;2](https://doi.org/10.1175/1520-0450(1994)033<0140:ASTMFM>2.0.CO;2).
- Darama, Y., 2014: Comment on “Groundwater depletion in the Middle East from GRACE with implications for transboundary water management in the Tigris-Euphrates-Western Iran Region” by Katalyn A. Voss et al. *Water Resour. Res.*, **50**, 754–757, <https://doi.org/10.1002/2013WR014084>.
- Darcy, H., 1856: *Les fontaines publiques de la ville de Dijon: exposition et application des principes à suivre et des formules à employer dans les questions de distribution d'eau*. Victor Dalmon, 647 pp.
- Dawdy, D. R., and W. B. Langbein, 1960: Mapping mean areal precipitation. *Int. Assoc. Sci. Hydrol. Bull.*, **5**, 16–23, <https://doi.org/10.1080/02626666009493176>.
- Day, G. N., 1985: Extended streamflow forecasting using NWSRFS. *J. Water Resour. Plann. Manage.*, **111**, 157–170, [https://doi.org/10.1061/\(ASCE\)0733-9496\(1985\)111:2\(157\)](https://doi.org/10.1061/(ASCE)0733-9496(1985)111:2(157)).
- DeAngelis, A., F. Dominguez, Y. Fan, A. Robock, M. D. Kustu, and D. Robinson, 2010: Evidence of enhanced precipitation due to irrigation over the Great Plains of the United States. *J. Geophys. Res.*, **115**, D15115, <https://doi.org/10.1029/2010JD013892>.
- Deems, J. S., T. H. Painter, and D. C. Finnegan, 2013: Lidar measurement of snow depth: A review. *J. Glaciol.*, **59**, 467–479, <https://doi.org/10.3189/2013JoG12J154>.
- de Jeu, R. A. M., W. Wagner, T. R. H. Holmes, A. J. Dolman, N. C. Giesen, and J. Friesen, 2008: Global soil moisture patterns observed by space borne microwave radiometers and scatterometers. *Surv. Geophys.*, **29**, 399–420, <https://doi.org/10.1007/s10712-008-9044-0>.

- Della-Marta, P. M., J. Luterbacher, H. von Weissenfluh, E. Xoplaki, M. Brunet, and H. Wanner, 2007: Summer heat waves over western Europe 1880–2003, their relationship to large-scale forcings and predictability. *Climate Dyn.*, **29**, 251–275, <https://doi.org/10.1007/s00382-007-0233-1>.
- Delworth, T., and S. Manabe, 1993: Climate variability and land-surface processes. *Adv. Water Resour.*, **16**, 3–20, [https://doi.org/10.1016/0309-1708\(93\)90026-C](https://doi.org/10.1016/0309-1708(93)90026-C).
- Demargne, J., and Coauthors, 2014: The Science of NOAA's Operational Hydrologic Ensemble Forecast Service. *Bull. Amer. Meteor. Soc.*, **95**, 79–98, <https://doi.org/10.1175/BAMS-D-12-00081.1>.
- Deming, D., 2014: Pierre Perrault, the hydrologic cycle and the scientific revolution. *Groundwater*, **52**, 156–162, <https://doi.org/10.1111/gwat.12138>.
- , 2017: Edme Mariotte and the beginning of quantitative hydrogeology. *Groundwater*, **56**, 350–355, <https://doi.org/10.1111/gwat.12609>.
- de Silva, T. K. N. P., D. L. O. Mendis, V. E. A. Wickramanayake, and C. G. Urugoda, 1995: Scientific principles seen in ancient artefacts and monumental structures. Natural Resources Energy and Science Authority of Sri Lanka, 54 pp.
- Dickey, J. O., and Coauthors, 1997: *Satellite Gravity and the Geosphere*. National Academies Press, 126 pp., <https://doi.org/10.17226/5767>.
- Dickinson, R. E., 1984: Modeling evapotranspiration for three-dimensional global climate models. *Climate Processes and Climate Sensitivity, Geophys. Monogr.*, Vol. 29, Amer. Geophys. Union, 58–72.
- , and A. Henderson-Sellers, 1988: Modelling tropical deforestation: A study of GCM land-surface parametrizations. *Quart. J. Roy. Meteor. Soc.*, **114**, 439–462, <https://doi.org/10.1002/qj.49711448009>.
- , —, P. J. Kennedy, and M. F. Wilson, 1986: Biosphere–Atmosphere Transfer Scheme (BATS) for the NCAR Community Climate Model. NCAR Tech. Note NCAR/TN-275+STR, 72 pp., <https://doi.org/10.5065/D6668B58>.
- Dirmeyer, P. A., A. J. Dolman, and N. Sato, 1999: The pilot phase of the Global Soil Wetness Project. *Bull. Amer. Meteor. Soc.*, **80**, 851–878, [https://doi.org/10.1175/1520-0477\(1999\)080<0851:TPPOTG>2.0.CO;2](https://doi.org/10.1175/1520-0477(1999)080<0851:TPPOTG>2.0.CO;2).
- , X. Gao, M. Zhao, Z. Guo, T. Oki, and N. Hanasaki, 2006a: GSWP-2: Multimodel analysis and implications for our perception of the land surface. *Bull. Amer. Meteor. Soc.*, **87**, 1381–1397, <https://doi.org/10.1175/BAMS-87-10-1381>.
- , R. D. Koster, and Z. Guo, 2006b: Do global models properly represent the feedback between land and atmosphere? *J. Hydrometeorol.*, **7**, 1177–1198, <https://doi.org/10.1175/JHM532.1>.
- , and Coauthors, 2012: Evidence for enhanced land–atmosphere feedback in a warming climate. *J. Hydrometeorol.*, **13**, 981–995, <https://doi.org/10.1175/JHM-D-11-0104.1>.
- , and Coauthors, 2018: Verification of land–atmosphere coupling in forecast models, reanalyses, and land surface models using flux site observations. *J. Hydrometeorol.*, **19**, 375–392, <https://doi.org/10.1175/JHM-D-17-0152.1>.
- Dobson, M. C., F. T. Ulaby, M. T. Hallikainen, and M. A. El-Rayes, 1985: Microwave dielectric behavior of wet soil-Part II: Dielectric mixing models. *IEEE Trans. Geosci. Remote Sens.*, **GE-23**, 35–46, <https://doi.org/10.1109/TGRS.1985.289498>.
- Döll, P., H. Müller Schmied, C. Schuh, F. T. Portmann, and A. Eicker, 2014: Global-scale assessment of groundwater depletion and related groundwater abstractions: Combining hydrological modeling with information from well observations and GRACE satellites. *Water Resour. Res.*, **50**, 5698–5720, <https://doi.org/10.1002/2014WR015595>.
- Dolman, A. J., 2005: Actual evaporation. *Encyclopedia of Hydrological Sciences*, M. G. Anderson and J. J. McDonnell, Eds., Wiley, 647–656, <https://doi.org/10.1002/0470848944.hsa048>.
- Dooge, J. C. I., 1959: A general theory of the unit hydrograph. *J. Geophys. Res.*, **64**, 241, <https://doi.org/10.1029/JZ064i002p00241>.
- , 1973: Linear theory of hydrologic systems. USDA-ARS Tech. Bull. 1468, 337 pp.
- , 1986: Looking for hydrologic laws. *Water Resour. Res.*, **22**, 46S–58S, <https://doi.org/10.1029/WR022i09Sp0046S>.
- , 1996: Walter Langbein and the emergence of scientific hydrology. *Water Resour. Res.*, **32**, 2969–2977, <https://doi.org/10.1029/96WR00273>.
- Doorenbos, J., and W. O. Pruitt, 1977: Guidelines for predicting crop water requirements. FAO Irrigation and Drainage Paper 24, 154 pp., <http://www.fao.org/3/a-f2430e.pdf>.
- Dorigo, W., and Coauthors, 2011: The International Soil Moisture Network: A data hosting facility for global in situ soil moisture measurements. *Hydrol. Earth Syst. Sci.*, **15**, 1675–1698, <https://doi.org/10.5194/hess-15-1675-2011>.
- , R. de Jeu, D. Chung, R. Parinussa, Y. Liu, W. Wagner, and D. Fernández-Prieto, 2012: Evaluating global trends (1988–2010) in harmonized multi-satellite surface soil moisture. *Geophys. Res. Lett.*, **39**, L18405, <https://doi.org/10.1029/2012GL052988>.
- , and Coauthors, 2017: ESA CCI Soil Moisture for improved Earth system understanding: State-of-the art and future directions. *Remote Sens. Environ.*, **203**, 185–215, <https://doi.org/10.1016/j.rse.2017.07.001>.
- Dorman, J. L., and P. J. Sellers, 1989: A global climatology of albedo, roughness length and stomatal resistance for atmospheric general circulation models as represented by the Simple Biosphere Model (SiB). *J. Appl. Meteor.*, **28**, 833–855, [https://doi.org/10.1175/1520-0450\(1989\)028<0833:AGCOAR>2.0.CO;2](https://doi.org/10.1175/1520-0450(1989)028<0833:AGCOAR>2.0.CO;2).
- Douville, H., and F. Chauvin, 2000: Relevance of soil moisture for seasonal climate predictions: A preliminary study. *Climate Dyn.*, **16**, 719–736, <https://doi.org/10.1007/s003820000080>.
- , —, and H. Broqua, 2001: Influence of soil moisture on the Asian and African monsoons. Part I: Mean monsoon and daily precipitation. *J. Climate*, **14**, 2381–2403, [https://doi.org/10.1175/1520-0442\(2001\)014<2381:IOSMOT>2.0.CO;2](https://doi.org/10.1175/1520-0442(2001)014<2381:IOSMOT>2.0.CO;2).
- Downer, C. W., and F. L. Ogden, 2004: GSSHA: Model to simulate diverse stream flow producing processes. *J. Hydrol. Eng.*, **9**, 161–174, [https://doi.org/10.1061/\(ASCE\)1084-0699\(2004\)9:3\(161\)](https://doi.org/10.1061/(ASCE)1084-0699(2004)9:3(161)).
- Dozier, J., E. H. Bair, and R. E. Davis, 2016: Estimating the spatial distribution of snow water equivalent in the world's mountains. *Wiley Interdiscip. Rev.: Water*, **3**, 461–474, <https://doi.org/10.1002/wat2.1140>.
- Duan, Q., S. Sorooshian, and V. Gupta, 1992: Effective and efficient global optimization for conceptual rainfall-runoff models. *Water Resour. Res.*, **28**, 1015–1031, <https://doi.org/10.1029/91WR02985>.
- , F. Pappenberger, A. Wood, H. L. Cloke, and J. Schaake, Eds., 2018: *Handbook of Hydrometeorological Ensemble Forecasting*. Springer, 1528 pp.
- Dunne, T., T. R. Moore, and C. H. Taylor, 1975: Recognition and prediction of runoff-producing zones in humid regions. *Hydrol. Sci. Bull.*, **20** (3), 305–327.
- Eagleson, P. S., 1970: *Dynamic Hydrology*. McGraw-Hill, 462 pp.
- , 1978a: Climate, soil, and vegetation: 1. Introduction to water balance dynamics. *Water Resour. Res.*, **14**, 705–712, <https://doi.org/10.1029/WR014i005p00705>.

- , 1978b: Climate, soil, and vegetation: 2. The distribution of annual precipitation derived from observed storm sequences. *Water Resour. Res.*, **14**, 713–721, <https://doi.org/10.1029/WR014i005p00713>.
- , 1978c: Climate, soil, and vegetation: 3. A simplified model of soil moisture movement in the liquid phase. *Water Resour. Res.*, **14**, 722–730, <https://doi.org/10.1029/WR014i005p00722>.
- , 1978d: Climate, soil, and vegetation: 4. The expected value of annual evapotranspiration. *Water Resour. Res.*, **14**, 731–739, <https://doi.org/10.1029/WR014i005p00731>.
- , 1978e: Climate, soil, and vegetation: 5. A derived distribution of storm surface runoff. *Water Resour. Res.*, **14**, 741–748, <https://doi.org/10.1029/WR014i005p00741>.
- , 1978f: Climate, soil, and vegetation: 6. Dynamics of the annual water balance. *Water Resour. Res.*, **14**, 749–764, <https://doi.org/10.1029/WR014i005p00749>.
- , 1978g: Climate, soil, and vegetation: 7. A derived distribution of annual water yield. *Water Resour. Res.*, **14**, 765–776, <https://doi.org/10.1029/WR014i005p00765>.
- , 1986: The emergence of global-scale hydrology. *Water Resour. Res.*, **22**, 6S–14S, <https://doi.org/10.1029/WR022i09Sp0006S>.
- , 1991: Hydrologic science: A distinct geoscience. *Rev. Geophys.*, **29**, 237–248, <https://doi.org/10.1029/90RG02615>.
- , N. M. Fennessey, W. Qinliang, and I. Rodríguez-Iturbe, 1987: Application of spatial Poisson models to air mass thunderstorm rainfall. *J. Geophys. Res.*, **92**, 9661–9678, <https://doi.org/10.1029/JD092iD08p09661>.
- Ek, M. B., K. E. Mitchell, Y. Lin, E. Rogers, P. Grunmann, V. Koren, G. Gayno, and J. D. Tarpley, 2003: Implementation of Noah land surface model advances in the National Centers for Environmental Prediction operational mesoscale Eta model. *J. Geophys. Res.*, **108**, 8851, <https://doi.org/10.1029/2002JD003296>.
- Elder, K., J. Dozier, and J. Michaelsen, 1991: Snow accumulation and distribution in an Alpine Watershed. *Water Resour. Res.*, **27**, 1541–1552, <https://doi.org/10.1029/91WR00506>.
- Eltahir, E. A. B., and R. L. Bras, 1996: Precipitation recycling. *Rev. Geophys.*, **34**, 367–378, <https://doi.org/10.1029/96RG01927>.
- Emerton, R. E., and Coauthors, 2016: Continental and global scale flood forecasting systems. *Wiley Interdiscip. Rev.: Water*, **3**, 391–418, <https://doi.org/10.1002/wat2.1137>.
- Entekhabi, D., 1995: Recent advances in land-atmosphere interaction research. *Rev. Geophys.*, **33**, 995–1003, <https://doi.org/10.1029/95RG01163>.
- , and Coauthors, 2010: The Soil Moisture Active Passive (SMAP) mission. *Proc. IEEE*, **98**, 704–716, <https://doi.org/10.1109/JPROC.2010.2043918>.
- Erickson, T. A., M. W. Williams, and A. Winstral, 2005: Persistence of topographic controls on the spatial distribution of snow in rugged mountain terrain, Colorado, United States. *Water Resour. Res.*, **41**, W04014, <https://doi.org/10.1029/2003WR002973>.
- Estilow, T. W., A. H. Young, and D. A. Robinson, 2015: A long-term Northern Hemisphere snow cover extent data record for climate studies and monitoring. *Earth Syst. Sci. Data*, **7**, 137–142, <https://doi.org/10.5194/essd-7-137-2015>.
- Evaristo, J., and J. J. McDonnell, 2017: Prevalence and magnitude of groundwater use by vegetation: A global stable isotope meta-analysis. *Sci. Rep.*, **7**, 44110, <https://doi.org/10.1038/srep44110>.
- Famiglietti, J. S., and E. F. Wood, 1991: Evapotranspiration and runoff from large land areas: Land surface hydrology for atmospheric general circulation models. *Surv. Geophys.*, **12**, 179–204, <https://doi.org/10.1007/BF01903418>.
- , and —, 1994: Multiscale modeling of spatially variable water and energy balance processes. *Water Resour. Res.*, **30**, 3061–3078, <https://doi.org/10.1029/94WR01498>.
- , A. Cazenave, A. Eicker, J. T. Reager, M. Rodell, and I. Velicogna, 2015: Satellites provide the big picture. *Science*, **349**, 684–685, <https://doi.org/10.1126/science.aac9238>.
- Fan, Y., 2015: Groundwater in the Earth's critical zone: Relevance to large-scale patterns and processes. *Water Resour. Res.*, **51**, 3052–3069, <https://doi.org/10.1002/2015WR017037>.
- , and R. L. Bras, 1995: On the concept of a representative elementary area in catchment runoff. *Hydrol. Processes*, **9**, 821–832, <https://doi.org/10.1002/hyp.3360090708>.
- , G. Miguez-Macho, C. P. Weaver, R. Walko, and A. Robock, 2007: Incorporating water table dynamics in climate modeling: 1. Water table observations and equilibrium water table simulations. *J. Geophys. Res.*, **112**, D10125, <https://doi.org/10.1029/2006JD008111>.
- , —, E. G. Jobbágy, R. B. Jackson, and C. Otero-Casal, 2017: Hydrologic regulation of plant rooting depth. *Proc. Natl. Acad. Sci. USA*, **114**, 10572–10577, <https://doi.org/10.1073/pnas.1712381114>.
- Farthing, M. W., and F. L. Ogden, 2017: Numerical solution of Richards' equation: A review of advances and challenges. *Soil Sci. Soc. Amer. J.*, **81**, 1257–1269, <https://doi.org/10.2136/sssaj2017.02.0058>.
- Fast, J. D., and M. D. McCorcle, 1990: A two-dimensional numerical sensitivity study of the Great Plains low-level jet. *Mon. Wea. Rev.*, **118**, 151–164, [https://doi.org/10.1175/1520-0493\(1990\)118<0151:ATDNSS>2.0.CO;2](https://doi.org/10.1175/1520-0493(1990)118<0151:ATDNSS>2.0.CO;2).
- Fedora, M. A., and R. L. Beschta, 1989: Storm runoff simulation using an antecedent precipitation index (API) model. *J. Hydrol.*, **112**, 121–133, [https://doi.org/10.1016/0022-1694\(89\)90184-4](https://doi.org/10.1016/0022-1694(89)90184-4).
- Fekete, B. M., R. D. Robarts, M. Kumagai, H. P. Nachtnebel, E. Odada, and A. V. Zhulidov, 2015: Time for in situ renaissance. *Science*, **349**, 685–686, <https://doi.org/10.1126/science.aac7358>.
- Feldman, A. D., 2000: Hydrologic modeling system HEC-HMS: Technical reference manual. USACE Doc. CPD-74B, 145 pp., [http://www.hec.usace.army.mil/software/hec-hms/documentation/HEC-HMS_Technical_Reference_Manual_\(CPD-74B\).pdf](http://www.hec.usace.army.mil/software/hec-hms/documentation/HEC-HMS_Technical_Reference_Manual_(CPD-74B).pdf).
- Fennessey, M. J., and J. Shukla, 1999: Impact of initial soil wetness on seasonal atmospheric prediction. *J. Climate*, **12**, 3167–3180, [https://doi.org/10.1175/1520-0442\(1999\)012<3167:IOISWO>2.0.CO;2](https://doi.org/10.1175/1520-0442(1999)012<3167:IOISWO>2.0.CO;2).
- Ferranti, L., and P. Viterbo, 2006: The European summer of 2003: Sensitivity to soil water initial conditions. *J. Climate*, **19**, 3659–3680, <https://doi.org/10.1175/JCLI3810.1>.
- Findell, K. L., and E. A. B. Eltahir, 2003: Atmospheric controls on soil moisture–boundary layer interactions. Part I: Framework development. *J. Hydrometeor.*, **4**, 552–569, [https://doi.org/10.1175/1525-7541\(2003\)004<0552:ACOSML>2.0.CO;2](https://doi.org/10.1175/1525-7541(2003)004<0552:ACOSML>2.0.CO;2).
- , P. Gentine, B. R. Lintner, and C. Kerr, 2011: Probability of afternoon precipitation in eastern United States and Mexico enhanced by high evaporation. *Nat. Geosci.*, **4**, 434–439, <https://doi.org/10.1038/ngeo1174>.
- , A. Berg, P. Gentine, J. P. Krasting, B. R. Lintner, S. Malyshev, J. A. Santanello, and E. Shevliakova, 2017: The impact of anthropogenic land use and land cover change on regional climate extremes. *Nat. Commun.*, **8**, 1–10, <https://doi.org/10.1038/s41467-017-01038-w>.
- Fischer, E. M., S. I. Seneviratne, D. Lüthi, and C. Schär, 2007: Contribution of land-atmosphere coupling to recent European

- summer heat waves. *Geophys. Res. Lett.*, **34**, L06707, <https://doi.org/10.1029/2006GL027992>.
- Fischer, K., 1919: Precipitation and run-off in the drainage basin of the Oder. *Mon. Wea. Rev.*, **47**, 743–744, [https://doi.org/10.1175/1520-0493\(1919\)47<743b:PARITD>2.0.CO;2](https://doi.org/10.1175/1520-0493(1919)47<743b:PARITD>2.0.CO;2).
- Ford, C. R., S. H. Laseter, W. T. Swank, and J. M. Vose, 2011: Can forest management be used to sustain water-based ecosystem services in the face of climate change? *Ecol. Appl.*, **21**, 2049–2067, <https://doi.org/10.1890/10-2246.1>.
- Ford, T. W., A. D. Rapp, and S. M. Quiring, 2015: Does afternoon precipitation occur preferentially over dry or wet soils in Oklahoma? *J. Hydrometeorol.*, **16**, 874–888, <https://doi.org/10.1175/JHM-D-14-0005.1>.
- Foufoula-Georgiou, E., 1989: A probabilistic storm transposition approach for estimating exceedance probabilities of extreme precipitation depths. *Water Resour. Res.*, **25**, 799–815, <https://doi.org/10.1029/WR025i005p00799>.
- , and D. P. Lettenmaier, 1987: A Markov renewal model for rainfall occurrences. *Water Resour. Res.*, **23**, 875–884, <https://doi.org/10.1029/WR023i005p00875>.
- Freeze, R. A., and R. L. Harlan, 1969: Blueprint for a physically-based, digitally-simulated hydrologic response model. *J. Hydrol.*, **9**, 237–258, [https://doi.org/10.1016/0022-1694\(69\)90020-1](https://doi.org/10.1016/0022-1694(69)90020-1).
- , and J. A. Cherry, 1979: *Groundwater*. Prentice-Hall, 604 pp.
- Frei, A., M. Tedesco, S. Lee, J. Foster, D. K. Hall, R. Kelly, and D. A. Robinson, 2012: A review of global satellite-derived snow products. *Adv. Space Res.*, **50**, 1007–1029, <https://doi.org/10.1016/j.asr.2011.12.021>.
- Fritzsche, R., 1906: Niederschlag, Abfluss und Verdunstung auf den Landflächen der Erde. *Z. Gewässerkunde*, **7** (6), 321–370.
- Funk, C., and J. P. Verdin, 2010: Real-time decision support systems: The Famine Early Warning System network. *Satellite Rainfall Applications for Surface Hydrology*, F. Hossain, Ed., Springer, 295–320.
- Gaborit, E., and Coauthors, 2017: A hydrological prediction system based on the SVS land-surface scheme: Efficient calibration of GEM-Hydro for streamflow simulation over the Lake Ontario basin. *Hydrol. Earth Syst. Sci.*, **21**, 4825–4839, <https://doi.org/10.5194/hess-21-4825-2017>.
- Gao, H., C. Birkett, and D. P. Lettenmaier, 2012: Global monitoring of large reservoir storage from satellite remote sensing. *Water Resour. Res.*, **48**, W09504, <https://doi.org/10.1029/2012WR012063>.
- Garratt, J. R., 1993: Sensitivity of climate simulations to land-surface and atmospheric boundary-layer treatments—A review. *J. Climate*, **6**, 419–449, [https://doi.org/10.1175/1520-0442\(1993\)006<0419:SOCSSTL>2.0.CO;2](https://doi.org/10.1175/1520-0442(1993)006<0419:SOCSSTL>2.0.CO;2).
- Garrido, F., M. Ghodrati, and M. Chendorain, 1999: Small-scale measurement of soil water content using a fiber optic sensor. *Soil. Sci. Soc. Amer. J.*, **63**, 1505–1512, <https://doi.org/10.2136/sssaj1999.6361505x>.
- Gelaro, R., and Coauthors, 2017: The Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2). *J. Climate*, **30**, 5419–5454, <https://doi.org/10.1175/JCLI-D-16-0758.1>.
- Gentine, P., A. A. M. Holtslag, F. D’Andrea, and M. Ek, 2013: Surface and atmospheric controls on the onset of moist convection over land. *J. Hydrometeorol.*, **14**, 1443–1462, <https://doi.org/10.1175/JHM-D-12-0137.1>.
- Georgakakos, K. P., 2006: Analytical results for operational flash flood guidance. *J. Hydrol.*, **317**, 81–103, <https://doi.org/10.1016/j.jhydrol.2005.05.009>.
- , and R. L. Bras, 1984: A hydrologically useful station precipitation model. *Water Resour. Res.*, **20**, 1597–1610, <https://doi.org/10.1029/WR020i011p01597>.
- Getirana, A., A. Boone, D. Yamazaki, B. Decharme, F. Papa, and N. Mognard, 2012: The Hydrological Modeling and Analysis Platform (HyMAP): Evaluation in the Amazon basin. *J. Hydrometeorol.*, **13**, 1641–1665, <https://doi.org/10.1175/JHM-D-12-021.1>.
- , C. Peters-Lidard, M. Rodell, and P. D. Bates, 2017: Trade-off between cost and accuracy in large-scale surface water dynamic modeling. *Water Resour. Res.*, **53**, 4942–4955, <https://doi.org/10.1002/2017WR020519>.
- Gochis, D. J., W. Yu, and D. Yates, 2015: The WRF-Hydro model technical description and user’s guide, version 3.0. NCAR Tech. Doc., 120 pp., http://www.ral.ucar.edu/projects/wrf_hydro/.
- Gourley, J. J., Z. L. Flamig, Y. Hong, and K. W. Howard, 2014: Evaluation of past, present and future tools for radar-based flash-flood prediction in the USA. *Hydrol. Sci. J.*, **59**, 1377–1389, <https://doi.org/10.1080/02626667.2014.919391>.
- Graf, W. L., 1999: Dam nation: A geographic census of American dams and their large-scale hydrologic impacts. *Water Resour. Res.*, **35**, 1305–1311, <https://doi.org/10.1029/1999WR900016>.
- , 2006: Downstream hydrologic and geomorphic effects of large dams on American rivers. *Geomorphology*, **79**, 336–360, <https://doi.org/10.1016/j.geomorph.2006.06.022>.
- , E. Wohl, T. Sinha, and J. L. Sabo, 2010: Sedimentation and sustainability of western American reservoirs. *Water Resour. Res.*, **46**, W12535, <https://doi.org/10.1029/2009WR008836>.
- Green, W. H., and G. A. Ampt, 1911: Studies on soil physics. *J. Agric. Sci.*, **4**, 1–24, <https://doi.org/10.1017/S0021859600001441>.
- Griffith, C. G., W. L. Woodley, P. G. Grube, D. W. Martin, J. Stout, and D. N. Sikdar, 1978: Rain estimation from geosynchronous satellite imagery—Visible and infrared studies. *Mon. Wea. Rev.*, **106**, 1153–1171, [https://doi.org/10.1175/1520-0493\(1978\)106<1153:REFGSI>2.0.CO;2](https://doi.org/10.1175/1520-0493(1978)106<1153:REFGSI>2.0.CO;2).
- Grill, G., B. Lehner, A. E. Lumsdon, G. K. MacDonald, C. Zarfl, and C. Reidy Liermann, 2015: An index-based framework for assessing patterns and trends in river fragmentation and flow regulation by global dams at multiple scales. *Environ. Res. Lett.*, **10**, 015001, <https://doi.org/10.1088/1748-9326/10/1/015001>.
- Guilod, B. P., and Coauthors, 2014: Land-surface controls on afternoon precipitation diagnosed from observational data: Uncertainties and confounding factors. *Atmos. Chem. Phys.*, **14**, 8343–8367, <https://doi.org/10.5194/acp-14-8343-2014>.
- , B. Orłowsky, D. G. Miralles, A. J. Teuling, and S. I. Seneviratne, 2015: Reconciling spatial and temporal soil moisture effects on afternoon rainfall. *Nat. Commun.*, **6**, 6443, <https://doi.org/10.1038/ncomms7443>.
- Guo, Z., and Coauthors, 2006: GLACE: The Global Land–Atmosphere Coupling Experiment. Part II: Analysis. *J. Hydrometeorol.*, **7**, 611–625, <https://doi.org/10.1175/JHM511.1>.
- Haddeland, I., T. Skaugen, and D. P. Lettenmaier, 2006: Anthropogenic impacts on continental surface water fluxes. *Geophys. Res. Lett.*, **33**, L08406, <https://doi.org/10.1029/2006GL026047>.
- , and Coauthors, 2011: Multimodel estimate of the global terrestrial water balance: Setup and first results. *J. Hydrometeorol.*, **12**, 869–884, <https://doi.org/10.1175/2011JHM1324.1>.
- Hager, W. H., 2001: Gauckler and the GMS formula. *J. Hydraul. Eng.*, **127**, 635–638, [https://doi.org/10.1061/\(ASCE\)0733-9429\(2001\)127:8\(635\)](https://doi.org/10.1061/(ASCE)0733-9429(2001)127:8(635)).

- Hall, D. K., G. A. Riggs, V. V. Salomonson, N. E. DiGirolamo, and K. J. Bayr, 2002: MODIS snow-cover products. *Remote Sens. Environ.*, **83**, 181–194, [https://doi.org/10.1016/S0034-4257\(02\)00095-0](https://doi.org/10.1016/S0034-4257(02)00095-0).
- Hamlet, A. F., 2011: Assessing water resources adaptive capacity to climate change impacts in the Pacific Northwest region of North America. *Hydrol. Earth Syst. Sci.*, **15**, 1427–1443, <https://doi.org/10.5194/hess-15-1427-2011>.
- Hanasaki, N., S. Kanae, and T. Oki, 2006: A reservoir operation scheme for global river routing models. *J. Hydrol.*, **327**, 22–24, <https://doi.org/10.1016/j.jhydrol.2005.11.011>.
- Hannah, D. M., P. J. Wood, and J. P. Sadler, 2004: Ecohydrology and hydroecology: A “new paradigm”? *Hydrol. Processes*, **18**, 3439–3445, <https://doi.org/10.1002/hyp.5761>.
- , S. Demuth, H. A. J. van Lanen, U. Looser, C. Prudhomme, G. Rees, K. Stahl, and L. M. Tallaksen, 2011: Large-scale river flow archives: Importance, current status and future needs. *Hydrol. Processes*, **25**, 1191–1200, <https://doi.org/10.1002/hyp.7794>.
- Hantush, M. S., and C. E. Jacob, 1955: Non-steady radial flow in an infinite leaky aquifer. *Eos, Trans. Amer. Geophys. Union*, **36**, 95–100, <https://doi.org/10.1029/TR036i001p00095>.
- Hargreaves, G. H., and Z. A. Samani, 1985: Reference crop evapotranspiration from temperature. *Appl. Eng. Agric.*, **1**, 96–99, <https://doi.org/10.13031/2013.26773>.
- Harley, B. M., 1971: A modular distributed model of catchment dynamics. Ph.D. dissertation, Massachusetts Institute of Technology, 537 pp.
- Harman, C., and P. A. Troch, 2014: What makes Darwinian hydrology “Darwinian”? Asking a different kind of question about landscapes. *Hydrol. Earth Syst. Sci.*, **18**, 417–433, <https://doi.org/10.5194/hess-18-417-2014>.
- Harrower, M. J., 2008: Hydrology, ideology, and the origins of irrigation in ancient southwest Arabia. *Curr. Anthropol.*, **49**, 497–510, <https://doi.org/10.1086/587890>.
- Hatfield, J. L., K. J. Boote, B. A. Kimball, L. H. Ziska, R. C. Izaurralde, D. Ort, A. M. Thomson, and D. Wolfe, 2011: Climate impacts on agriculture: Implications for crop production. *Agron. J.*, **103**, 351–370, <https://doi.org/10.2134/agronj2010.0303>.
- Hawkins, J., and C. Velden, 2011: Supporting meteorological field experiment missions and postmission analysis with satellite digital data and products. *Bull. Amer. Meteor. Soc.*, **92**, 1009–1022, <https://doi.org/10.1175/2011BAMS3138.1>.
- Hazenber, P., Y. Fang, P. Broxton, D. Gochis, G. Y. Niu, J. D. Pelletier, P. A. Troch, and X. Zeng, 2015: A hybrid-3D hillslope hydrological model for use in Earth system models. *Water Resour. Res.*, **51**, 8218–8239, <https://doi.org/10.1002/2014WR016842>.
- Hedrick, A., H. P. Marshall, A. Winstral, K. Elder, S. Yueh, and D. Cline, 2015: Independent evaluation of the SNODAS snow depth product using regional-scale lidar-derived measurements. *Cryosphere*, **9**, 13–23, <https://doi.org/10.5194/tc-9-13-2015>.
- Held, I. M., and B. J. Soden, 2006: Robust responses of the hydrologic cycle to global warming. *J. Climate*, **19**, 5686–5699, <https://doi.org/10.1175/JCLI3990.1>.
- Helms, D., S. E. Phillips, and P. F. Reich, Eds., 2008: The history of snow survey and water supply forecasting: Interviews with U.S. Department of Agriculture pioneers. USDA NRCS, 321 pp., https://www.nrcs.usda.gov/Internet/FSE_DOCUMENTS/stelprdb1043910.pdf.
- Henry, A. J., 1919: Floods on lower Rio Grande. *Mon. Wea. Rev.*, **47**, 742–743, [https://doi.org/10.1175/1520-0493\(1919\)47<742:FOLRG>2.0.CO;2](https://doi.org/10.1175/1520-0493(1919)47<742:FOLRG>2.0.CO;2).
- , 1928: Horton and Grunsky on the hydrology of the Great Lakes. *Mon. Wea. Rev.*, **56**, 11–14, [https://doi.org/10.1175/1520-0493\(1928\)56<11:HAGOTH>2.0.CO;2](https://doi.org/10.1175/1520-0493(1928)56<11:HAGOTH>2.0.CO;2).
- Hershfield, D. M., 1961a: Rainfall frequency atlas of the United States: For durations from 30 minutes to 24 hours and return periods from 1 to 100 years. U.S. Weather Bureau Technical Paper 40, 65 pp., http://www.nws.noaa.gov/oh/hdsc/PF_documents/TechnicalPaper_No40.pdf.
- , 1961b: Estimation of the probable maximum precipitation. *ASCE J. Hydraul. Div.*, **87**, 99–116.
- Hirsch, A. L., J. Kala, A. J. Pitman, C. Carouge, J. P. Evans, V. Haverd, and D. Mocko, 2014: Impact of land surface initialization approach on subseasonal forecast skill: A regional analysis in the Southern Hemisphere. *J. Hydrometeorol.*, **15**, 300–319, <https://doi.org/10.1175/JHM-D-13-05.1>.
- Hoffman, F. M., and Coauthors, 2017: 2016 International Land Model Benchmarking (ILAMB) Workshop Report. Rep. DOE/SC-0186, 159 pp., <https://doi.org/10.2172/1330803>.
- Hohenegger, C., P. Brockhaus, C. S. Bretherton, and C. Schär, 2009: The soil moisture-precipitation feedback in simulations with explicit and parameterized convection. *J. Climate*, **22**, 5003–5020, <https://doi.org/10.1175/2009JCLI2604.1>.
- Hollinger, S. E., and S. A. Isard, 1994: A soil moisture climatology of Illinois. *J. Climate*, **7**, 822–833, [https://doi.org/10.1175/1520-0442\(1994\)007<0822:ASMCOI>2.0.CO;2](https://doi.org/10.1175/1520-0442(1994)007<0822:ASMCOI>2.0.CO;2).
- Holmes, R. T., and G. E. Likens, 2016: *Hubbard Brook: The Story of a Forest Ecosystem*. Yale University Press, 271 pp.
- Hong, S.-Y., and E. Kalnay, 2000: Role of sea surface temperature and soil-moisture feedback in the 1998 Oklahoma–Texas drought. *Nature*, **844**, 842–844, <https://doi.org/10.1038/35048548>.
- Horton, R. E., 1933: The role of infiltration in the hydrologic cycle. *Eos, Trans. Amer. Geophys. Union*, **14**, 446–460, <https://doi.org/10.1029/TR014i001p00446>.
- , 1941: An approach toward a physical interpretation of infiltration-capacity. *Soil. Sci. Soc. Amer. J.*, **5**, 399–417, <https://doi.org/10.2136/sssaj1941.036159950005000C0075x>.
- Hoskins, B., 2013: The potential for skill across the range of the seamless weather-climate prediction problem: A stimulus for our science. *Quart. J. Roy. Meteor. Soc.*, **139**, 573–584, <https://doi.org/10.1002/qj.1991>.
- Hossain, F., and N. Katiyar, 2006: Improving flood forecasting in international river basins. *Eos, Trans. Amer. Geophys. Union*, **87**, 49, <https://doi.org/10.1029/2006EO050001>.
- , and D. P. Lettenmaier, 2006: Flood prediction in the future: Recognizing hydrologic issues in anticipation of the Global Precipitation Measurement mission. *Water Resour. Res.*, **42**, W11301, <https://doi.org/10.1029/2006WR005202>.
- , I. Jeyachandran, and R. Pielke Sr., 2010: Dam safety effects due to human alteration of extreme precipitation. *Water Resour. Res.*, **46**, W03301, <https://doi.org/10.1029/2009WR007704>.
- , and Coauthors, 2014: Crossing the “valley of death”: Lessons learned from implementing an operational satellite-based flood forecasting system. *Bull. Amer. Meteor. Soc.*, **95**, 1201–1207, <https://doi.org/10.1175/BAMS-D-13-00176.1>.
- , N. Biswas, M. Ashraf, and B. A. Zeeshan, 2017: Growing more with less in the Indus Valley using cell phones and satellite data. *Eos, Trans. Amer. Geophys. Union*, **98**, <https://doi.org/10.1029/2017EO075143>.
- Hou, A. Y., and Coauthors, 2014: The Global Precipitation Measurement mission. *Bull. Amer. Meteor. Soc.*, **95**, 701–722, <https://doi.org/10.1175/BAMS-D-13-00164.1>.
- Houborg, R., M. Rodell, B. Li, R. Reichle, and B. Zaitchik, 2012: Drought indicators based on model assimilated GRACE

- terrestrial water storage observations. *Water Resour. Res.*, **48**, W07525, <https://doi.org/10.1029/2011WR011291>.
- Houze, R. A., Jr., 2019: 100 years of research on mesoscale convective systems. *A Century of Progress in Atmospheric and Related Sciences: Celebrating the American Meteorological Society Centennial*, Meteor. Monogr., No. 59, Amer. Meteor. Soc., <https://doi.org/10.1175/AMSMONOGRAPHS-D-18-0001.1>.
- , and Coauthors, 2017: The Olympic Mountains Experiment (Olympex). *Bull. Amer. Meteor. Soc.*, **98**, 2167–2188, <https://doi.org/10.1175/BAMS-D-16-0182.1>.
- Hsu, H., M. H. Lo, B. P. Guillod, D. G. Miralles, and S. Kumar, 2017: Relation between precipitation location and antecedent/subsequent soil moisture spatial patterns. *J. Geophys. Res. Atmos.*, **122**, 6319–6328, <https://doi.org/10.1002/2016JD026042>.
- Hsu, K., X. Gao, S. Sorooshian, and H. V. Gupta, 1997: Precipitation Estimation from Remotely Sensed Information Using Artificial Neural Networks. *J. Appl. Meteor.*, **36**, 1176–1190, [https://doi.org/10.1175/1520-0450\(1997\)036<1176:PEFRSI>2.0.CO;2](https://doi.org/10.1175/1520-0450(1997)036<1176:PEFRSI>2.0.CO;2).
- Huang, G. J., V. N. Bringi, R. Cifelli, D. Hudak, and W. A. Petersen, 2010: A methodology to derive radar reflectivity-liquid equivalent snow rate relations using C-band radar and a 2D video disdrometer. *J. Atmos. Oceanic Technol.*, **27**, 637–651, <https://doi.org/10.1175/2009JTECHA1284.1>.
- Hubbert, M. K., 1940: The theory of ground-water motion. *J. Geol.*, **48**, 785–944, <https://doi.org/10.1086/624930>.
- Huff, F. A., and S. A. Changnon, 1964: A model 10-inch rainstorm. *J. Appl. Meteor.*, **3**, 587–599, [https://doi.org/10.1175/1520-0450\(1964\)003<0587:AMIR>2.0.CO;2](https://doi.org/10.1175/1520-0450(1964)003<0587:AMIR>2.0.CO;2).
- Huffman, G. J., and Coauthors, 2007: The TRMM Multisatellite Precipitation Analysis (TMPA): Quasi-global, multiyear, combined-sensor precipitation estimates at fine scales. *J. Hydrometeorol.*, **8**, 38–55, <https://doi.org/10.1175/JHM560.1>.
- Humphrey, V., L. Gudmundsson, and S. I. Seneviratne, 2016: Assessing global water storage variability from GRACE: Trends, seasonal cycle, subseasonal anomalies and extremes. *Surv. Geophys.*, **37**, 357–395, <https://doi.org/10.1007/s10712-016-9367-1>.
- Idso, S. B., R. D. Jackson, R. J. Reginato, B. A. Kimball, and F. S. Nakayama, 1975: The dependence of bare soil albedo on soil water content. *J. Appl. Meteor.*, **14**, 109–113, [https://doi.org/10.1175/1520-0450\(1975\)014<0109:TDOBSA>2.0.CO;2](https://doi.org/10.1175/1520-0450(1975)014<0109:TDOBSA>2.0.CO;2).
- Iguchi, T., T. Kozu, R. Meneghini, J. Awaka, and K. Okamoto, 2000: Rain-profiling algorithm for the TRMM Precipitation Radar. *J. Appl. Meteor.*, **39**, 2038–2052, [https://doi.org/10.1175/1520-0450\(2001\)040<2038:RPAFTT>2.0.CO;2](https://doi.org/10.1175/1520-0450(2001)040<2038:RPAFTT>2.0.CO;2).
- Jacob, C. E., 1940: On the flow of water in an elastic artesian aquifer. *Trans. Amer. Geophys. Union*, **21**, 574, <https://doi.org/10.1029/TR021i002p00574>.
- Jarvis, P. G., 1976: The interpretation of the variations in leaf water potential and stomatal conductance found in canopies in the field. *Philos. Trans. Roy. Soc. London*, **273B**, 593–610, <https://doi.org/10.1098/rstb.1976.0035>.
- Jasechko, S., Z. Sharp, J. Gibson, S. Birks, Y. Yi, and P. Fawcett, 2013: Terrestrial water fluxes dominated by transpiration. *Nature*, **496**, 347–350, <https://doi.org/10.1038/nature11983>.
- Jiang, X., G. Y. Niu, and Z. L. Yang, 2009: Impacts of vegetation and groundwater dynamics on warm season precipitation over the central United States. *J. Geophys. Res.*, **114**, D06109, <https://doi.org/10.1029/2008JD010756>.
- Johnson, D. M., and Coauthors, 2018: Co-occurring woody species have diverse hydraulic strategies and mortality rates during an extreme drought. *Plant Cell Environ.*, **41**, 576–588, <https://doi.org/10.1111/pce.13121>.
- Joyce, R. J., J. E. Janowiak, P. A. Arkin, and P. Xie, 2004: CMORPH: A method that produces global precipitation estimates from passive microwave and infrared data at high spatial and temporal resolution. *J. Hydrometeorol.*, **5**, 487–503, [https://doi.org/10.1175/1525-7541\(2004\)005<0487:CAMTPG>2.0.CO;2](https://doi.org/10.1175/1525-7541(2004)005<0487:CAMTPG>2.0.CO;2).
- Julien, P. Y., B. Saghaian, and F. L. Ogden, 1995: Raster-based hydrologic modeling of spatially-varied surface runoff. *J. Amer. Water Resour. Assoc.*, **31**, 523–536, <https://doi.org/10.1111/j.1752-1688.1995.tb04039.x>.
- Jung, M., M. Reichstein, and A. Bondeau, 2009: Towards global empirical upscaling of FLUXNET eddy covariance observations: Validation of a model tree ensemble approach using a biosphere model. *Biogeosciences*, **6**, 2001–2013, <https://doi.org/10.5194/bg-6-2001-2009>.
- , and Coauthors, 2010: Recent decline in the global land evapotranspiration trend due to limited moisture supply. *Nature*, **467**, 951–954, <https://doi.org/10.1038/nature09396>.
- Kalma, J. D., T. R. McVicar, and M. F. McCabe, 2008: Estimating land surface evaporation: A review of methods using remotely sensed surface temperature data. *Surv. Geophys.*, **29**, 421–469, <https://doi.org/10.1007/s10712-008-9037-z>.
- Kelly, R. E., 2009: The AMSR-E snow depth algorithm: Description and initial results. *J. Remote Sens. Soc. Japan*, **29**, 307–317, <https://doi.org/10.11440/rssj.29.307>.
- , A. T. Chang, L. Tsang, and J. L. Foster, 2003: A prototype AMSR-E global snow area and snow depth algorithm. *IEEE Trans. Geosci. Remote Sens.*, **41**, 230–242, <https://doi.org/10.1109/TGRS.2003.809118>.
- Kenny, J., N. Barber, S. Hutson, K. S. Linsey, J. K. Lovelace, and M. Maupin, 2009: Estimated use of water in the United States in 2005. USGS Circular 1344, 52 pp., <https://pubs.usgs.gov/circ/1344/pdf/c1344.pdf>.
- Kerr, Y. H., and Coauthors, 2016: Overview of SMOS performance in terms of global soil moisture monitoring after six years in operation. *Remote Sens. Environ.*, **180**, 40–63, <https://doi.org/10.1016/j.rse.2016.02.042>.
- Kidd, C., and G. Huffman, 2011: Global precipitation measurement. *Meteor. Appl.*, **18**, 334–353, <https://doi.org/10.1002/met.284>.
- Kinter, J. L., and J. Shukla, 1990: The Global Hydrologic and Energy Cycles: Suggestions for studies in the pre-Global Energy and Water Cycle Experiment (GEWEX) period. *Bull. Amer. Meteor. Soc.*, **71**, 181–189, [https://doi.org/10.1175/1520-0477\(1990\)071<0181:TGHAEC>2.0.CO;2](https://doi.org/10.1175/1520-0477(1990)071<0181:TGHAEC>2.0.CO;2).
- Kirschbaum, D., R. Adler, D. Adler, C. Peters-Lidard, and G. Huffman, 2012: Global distribution of extreme precipitation and high-impact landslides in 2010 relative to previous years. *J. Hydrometeorol.*, **13**, 1536–1551, <https://doi.org/10.1175/JHM-D-12-02.1>.
- Kistenmacher, M., and A. P. Georgakakos, 2015: Assessment of reservoir system variable forecasts. *Water Resour. Res.*, **51**, 3437–3458, <https://doi.org/10.1002/2014WR016564>.
- Klemeš, V., 1988: A hydrological perspective. *J. Hydrol.*, **100**, 3–28, [https://doi.org/10.1016/0022-1694\(88\)90179-5](https://doi.org/10.1016/0022-1694(88)90179-5).
- Kollet, S., and Coauthors, 2017: The integrated hydrologic model intercomparison project, IH-MIP2: A second set of benchmark results to diagnose integrated hydrology and feedbacks. *Water Resour. Res.*, **53**, 867–890, <https://doi.org/10.1002/2016WR019191>.
- König, M., and M. Sturm, 1998: Mapping snow distribution in the Alaskan Arctic using aerial photography and topographic relationships. *Water Resour. Res.*, **34**, 3471–3483, <https://doi.org/10.1029/98WR02514>.

- Koren, V., S. Reed, M. Smith, Z. Zhang, and D.-J. Seo, 2004: Hydrology laboratory research modeling system (HL-RMS) of the US national weather service. *J. Hydrol.*, **291**, 297–318, <https://doi.org/10.1016/j.jhydrol.2003.12.039>.
- Koster, R. D., and M. J. Suarez, 1992: Modeling the land surface boundary in climate models as a composite of independent vegetation stands. *J. Geophys. Res.*, **97**, 2697–2715, <https://doi.org/10.1029/91JD01696>.
- , —, A. Ducharne, M. Stieglitz, and P. Kumar, 2000a: A catchment-based approach to modeling land surface processes in a general circulation model: 1. Model structure. *J. Geophys. Res.*, **105**, 24 809–24 822, <https://doi.org/10.1029/2000JD900327>.
- , —, and M. Heiser, 2000b: Variance and predictability of precipitation at seasonal-to-interannual timescales. *J. Hydrometeorol.*, **1**, 26–46, [https://doi.org/10.1175/1525-7541\(2000\)001<0026:VAPOPA>2.0.CO;2](https://doi.org/10.1175/1525-7541(2000)001<0026:VAPOPA>2.0.CO;2).
- , and Coauthors, 2004: Regions of strong coupling between soil moisture and precipitation. *Science*, **305**, 1138–1140, <https://doi.org/10.1126/science.1100217>.
- , and Coauthors, 2006: GLACE: The Global Land–Atmosphere Coupling Experiment. Part I: Overview. *J. Hydrometeorol.*, **7**, 590–610, <https://doi.org/10.1175/JHM510.1>.
- , and Coauthors, 2010: Contribution of land surface initialization to subseasonal forecast skill: First results from a multi-model experiment. *Geophys. Res. Lett.*, **37**, L02402, <https://doi.org/10.1029/2009GL041677>.
- Krause, S., and Coauthors, 2011: Inter-disciplinary perspectives on processes in the hyporheic zone. *Ecohydrology*, **4**, 481–499, <https://doi.org/10.1002/eco.176>.
- Kucera, P. A., E. E. Ebert, F. J. Turk, V. Levizzani, D. Kirschbaum, F. J. Tapiador, A. Loew, and M. Borsche, 2013: Precipitation from space: Advancing earth system science. *Bull. Amer. Meteor. Soc.*, **94**, 365–375, <https://doi.org/10.1175/BAMS-D-11-00171.1>.
- Kueppers, L. M., M. A. Snyder, and L. C. Sloan, 2007: Irrigation cooling effect: Regional climate forcing by land-use change. *Geophys. Res. Lett.*, **34**, L03703, <https://doi.org/10.1029/2006GL028679>.
- Kuichling, E., 1889: The relation between the rainfall and the discharge of sewers in populous districts. *Trans. Amer. Soc. Civ. Eng.*, **20**, 1–56.
- Kumar, P., 2015: Hydrocomplexity: Addressing water security and emergent environmental risks. *Water Resour. Res.*, **51**, 5827–5838, <https://doi.org/10.1002/2015WR017342>.
- Kumar, S. V., and Coauthors, 2006: Land information system: An interoperable framework for high resolution land surface modeling. *Environ. Modell. Software*, **21**, 1402–1415, <https://doi.org/10.1016/j.envsoft.2005.07.004>.
- , S. Wang, D. M. Mocko, C. D. Peters-Lidard, and Y. Xia, 2017: Similarity assessment of land surface model outputs in the North American Land Data Assimilation System. *Water Resour. Res.*, **53**, 8941–8965, <https://doi.org/10.1002/2017WR020635>.
- Kummerow, C., and Coauthors, 2000: The status of the Tropical Rainfall Measuring Mission (TRMM) after two years in orbit. *J. Appl. Meteor.*, **39**, 1965–1982, [https://doi.org/10.1175/1520-0450\(2001\)040<1965:TSOTTR>2.0.CO;2](https://doi.org/10.1175/1520-0450(2001)040<1965:TSOTTR>2.0.CO;2).
- Kundzewicz, Z. W., L. J. Mata, N. W. Arnell, P. Döll, B. Jimenez, K. Miller, T. Oki, Z. Şen, and I. Shiklomanov, 2008: The implications of projected climate change for freshwater resources and their management. *Hydrol. Sci. J.*, **53**, 3–10, <https://doi.org/10.1623/hysj.53.1.3>.
- Kunkel, K. E., T. R. Karl, D. R. Easterling, K. Redmond, J. Young, X. Yin, and P. Hennon, 2013: Probable maximum precipitation and climate change. *Geophys. Res. Lett.*, **40**, 1402–1408, <https://doi.org/10.1002/grl.50334>.
- Lawrence, D. M., and Coauthors, 2011: Parameterization improvements and functional and structural advances in Version 4 of the Community Land Model. *J. Adv. Model. Earth Syst.*, **3**, M03001, <https://doi.org/10.1029/2011MS000045>.
- Lettau, H., 1969: Evaporation climatology. *Mon. Wea. Rev.*, **97**, 691–699, [https://doi.org/10.1175/1520-0493\(1969\)097<0691:IANATN>2.3.CO;2](https://doi.org/10.1175/1520-0493(1969)097<0691:IANATN>2.3.CO;2).
- Lettenmaier, D. P., D. Alsdorf, J. Dozier, G. J. Huffman, M. Pan, and E. F. Wood, 2015: Inroads of remote sensing into hydrologic science during the WRR era. *Water Resour. Res.*, **51**, 7309–7342, <https://doi.org/10.1002/2015WR017616>.
- Leung, L. R., M. Huang, Y. Qian, and X. Liang, 2011: Climate-soil-vegetation control on groundwater table dynamics and its feedbacks in a climate model. *Climate Dyn.*, **36**, 57–81, <https://doi.org/10.1007/s00382-010-0746-x>.
- Lhomme, J. P., E. Elguero, A. Chehbouni, and G. Boulet, 1998: Stomatal control of transpiration: Examination of Monteith's formulation of canopy resistance. *Water Resour. Res.*, **34**, 2301–2308, <https://doi.org/10.1029/98WR01339>.
- Liang, X., D. P. Lettenmaier, E. F. Wood, and S. J. Burges, 1994: A simple hydrologically based model of land surface water and energy fluxes for general circulation models. *J. Geophys. Res.*, **99**, 14 415, <https://doi.org/10.1029/94JD00483>.
- Liao, L., R. Meneghini, A. Tokay, and L. F. Bliven, 2016: Retrieval of snow properties for Ku- and Ka-band dual-frequency radar. *J. Appl. Meteor. Climatol.*, **55**, 1845–1858, <https://doi.org/10.1175/JAMC-D-15-0355.1>.
- Lighthill, M. J., and G. B. Whitham, 1955: On kinematic waves. I. Flood movement in long rivers. *Proc. Roy. Soc. London*, **229A**, 281–316, <https://doi.org/10.1098/rspa.1955.0088>.
- Ligon, F. K., W. E. Dietrich, and W. J. Trush, 1995: Downstream ecological effects of dams. *BioScience*, **45**, 183–192, <https://doi.org/10.2307/1312557>.
- Liu, F., R. Parmenter, P. D. Brooks, M. H. Conklyn, and R. C. Bales, 2008: Seasonal and interannual variation of streamflow pathways and biogeochemical implications in semi-arid, forested catchments in Valles Caldera, New Mexico. *Ecohydrology*, **1**, 239–252, <https://doi.org/10.1002/eco.22>.
- Liu, Y., and Coauthors, 2012: Advancing data assimilation in operational hydrologic forecasting: Progresses, challenges, and emerging opportunities. *Hydrol. Earth Syst. Sci.*, **16**, 3863–3887, <https://doi.org/10.5194/hess-16-3863-2012>.
- Lo, M. H., and J. S. Famiglietti, 2013: Irrigation in California's Central Valley strengthens the southwestern U.S. water cycle. *Geophys. Res. Lett.*, **40**, 301–306, <https://doi.org/10.1002/grl.50108>.
- Lundquist, J. D., and F. Lott, 2010: Using inexpensive temperature sensors to monitor the duration and heterogeneity of snow-covered areas. *Water Resour. Res.*, **46**, W00D16, <https://doi.org/10.1029/2008WR007035>.
- Luo, L., and E. F. Wood, 2008: Use of Bayesian merging techniques in a multimodel seasonal hydrologic ensemble prediction system for the eastern United States. *J. Hydrometeorol.*, **9**, 866–884, <https://doi.org/10.1175/2008JHM980.1>.
- Lutcke, S. B., and Coauthors, 2006: Recent Greenland ice mass loss by drainage system from satellite gravity observations. *Science*, **314**, 1286–1289, <https://doi.org/10.1126/science.1130776>.
- , A. A. Arendt, D. D. Rowlands, J. J. McCarthy, and C. F. Larsen, 2008: Recent glacier mass changes in the Gulf of Alaska region from GRACE mascon solutions. *J. Glaciol.*, **54**, 767–777, <https://doi.org/10.3189/002214308787779933>.
- Lvovich, M. I., 1972: World water balance (general report). *IAHS Publ.*, **93**, 401–415.

- Machguth, H., O. Eisen, F. Paul, and M. Hoelzle, 2006: Strong spatial variability of snow accumulation observed with helicopterborne GPR on two adjacent Alpine glaciers. *Geophys. Res. Lett.*, **33**, L13503, <https://doi.org/10.1029/2006GL026576>.
- Mackay, D. S., D. E. Roberts, B. E. Ewers, J. S. Sperry, N. G. McDowell, and W. T. Pockman, 2015: Interdependence of chronic hydraulic dysfunction and canopy processes can improve integrated models of tree response to drought. *Water Resour. Res.*, **51**, 6156–6176, <https://doi.org/10.1002/2015WR017244>.
- Mahanama, S., B. Livneh, R. Koster, D. Lettenmaier, and R. Reichle, 2012: Soil moisture, snow, and seasonal streamflow forecasts in the United States. *J. Hydrometeorol.*, **13**, 189–203, <https://doi.org/10.1175/JHM-D-11-046.1>.
- Maidment, D. R., 1993: *Handbook of Hydrology*. McGraw-Hill, 1424 pp.
- Mallick, K., A. Jarvis, J. B. Fisher, K. P. Tu, E. Boegh, and D. Niyogi, 2013: Latent heat flux and canopy conductance based on Penman–Monteith, Priestley–Taylor equation, and Bouchet’s complementary hypothesis. *J. Hydrometeorol.*, **14**, 419–442, <https://doi.org/10.1175/JHM-D-12-0117.1>.
- Manabe, S., 1969: Climate and the ocean circulation. I. The atmospheric circulation and the hydrology of the Earth’s surface. *Mon. Wea. Rev.*, **97**, 739–774, [https://doi.org/10.1175/1520-0493\(1969\)097<0739:CATOC>2.3.CO;2](https://doi.org/10.1175/1520-0493(1969)097<0739:CATOC>2.3.CO;2).
- Manning, R., 1891: On the flow of water in open channels and pipes. *Trans. Inst. Civ. Eng. Irel.*, **20**, 161–207.
- Maquin, M., E. Mouche, C. Mügler, M.-C. Pierret, and D. Viville, 2017: A soil column model for predicting the interaction between water table and evapotranspiration. *Water Resour. Res.*, **53**, 5877–5898, <https://doi.org/10.1002/2016WR020183>.
- Martens, B., and Coauthors, 2017: GLEAM v3: Satellite-based land evaporation and root-zone soil moisture. *Geosci. Model Dev.*, **10**, 1903–1925, <https://doi.org/10.5194/gmd-10-1903-2017>.
- Martinez, J. A., F. Dominguez, and G. Miguez-Macho, 2016: Effects of a groundwater scheme on the simulation of soil moisture and evapotranspiration over southern South America. *J. Hydrometeorol.*, **17**, 2941–2957, <https://doi.org/10.1175/JHM-D-16-0051.1>.
- Mather, J. R., 1969: The average annual water balance of the world. *Proc. Symp. on Water Balance in North America*, Banff, AB, Canada, American Water Resource Association, 29–40.
- Matott, L. S., 2017: OSTRICH: An optimization software tool, documentation and user’s guide, version 17.12.19. University at Buffalo Center for Computational Research, 79 pp., <http://www.eng.buffalo.edu/~lsmatott/Ostrich/OstrichMain.html>.
- Maxwell, R. M., and N. L. Miller, 2005: Development of a coupled land surface and groundwater model. *J. Hydrometeorol.*, **6**, 233–247, <https://doi.org/10.1175/JHM422.1>.
- , and S. J. Kollet, 2008: Interdependence of groundwater dynamics and land-energy feedbacks under climate change. *Nat. Geosci.*, **1**, 665–669, <https://doi.org/10.1038/ngeo315>.
- , and L. E. Condon, 2016: Connections between groundwater flow and transpiration partitioning. *Science*, **353**, 377–380, <https://doi.org/10.1126/science.aaf7891>.
- McCabe, M. F., A. Ershadi, C. Jimenez, D. G. Miralles, D. Michel, and E. F. Wood, 2016: The GEWEX LandFlux project: Evaluation of model evaporation using tower-based and globally gridded forcing data. *Geosci. Model Dev.*, **9**, 283–305, <https://doi.org/10.5194/gmd-9-283-2016>.
- , and Coauthors, 2017: The future of Earth observation in hydrology. *Hydrol. Earth Syst. Sci.*, **21**, 3879–3914, <https://doi.org/10.5194/hess-21-3879-2017>.
- McCarthy, G. T., 1940: Flood routing. *Flood Control*, Engineer School Fort Belvoir, 127–177.
- McDonald, M. G., and A. W. Harbaugh, 2003: The history of MODFLOW. *Groundwater*, **41**, 280–283, <https://doi.org/10.1111/j.1745-6584.2003.tb02591.x>.
- McDonnell, J. J., 2017: Beyond the water balance. *Nat. Geosci.*, **10**, 396, <https://doi.org/10.1038/ngeo2964>.
- , and Coauthors, 2007: Moving beyond heterogeneity and process complexity: A new vision for watershed hydrology. *Water Resour. Res.*, **43**, W07301, <https://doi.org/10.1029/2006WR005467>.
- McDowell, N. G., and C. D. Allen, 2015: Darcy’s law predicts widespread forest mortality under climate warming. *Nat. Climate Change*, **5**, 669–672, <https://doi.org/10.1038/nclimate2641>.
- , S. T. Michaletz, K. E. Bennett, K. C. Solander, C. Xu, R. M. Maxwell, and R. S. Middleton, 2018: Predicting chronic climate-driven disturbances and their mitigation. *Trends Ecol. Evol.*, **33**, 15–27, <https://doi.org/10.1016/j.tree.2017.10.002>.
- McNally, A., and Coauthors, 2017: A land data assimilation system for sub-Saharan Africa food and water security applications. *Sci. Data*, **4**, 170012, <https://doi.org/10.1038/sdata.2017.12>.
- Mein, R. G., and C. L. Larson, 1973: Modeling infiltration during a steady rain. *Water Resour. Res.*, **9**, 384–394, <https://doi.org/10.1029/WR009i002p00384>.
- Meinzer, O. E., 1923: The occurrence of principles. Geological Survey Water-Supply Paper 489, 321 pp., <https://pubs.er.usgs.gov/publication/wsp489>.
- Miguez-Macho, G., and Y. Fan, 2012: The role of groundwater in the Amazon water cycle: 1. Influence on seasonal streamflow, flooding and wetlands. *J. Geophys. Res.*, **117**, D15113, <https://doi.org/10.1029/2012JD017539>.
- , —, C. P. Weaver, R. Walko, and A. Robock, 2007: Incorporating water table dynamics in climate modeling: 2. Formulation, validation, and soil moisture simulation. *J. Geophys. Res.*, **112**, D13108, <https://doi.org/10.1029/2006JD008112>.
- Miller, E. E., and R. D. Miller, 1956: Physical theory for capillary flow phenomena. *J. Appl. Phys.*, **27**, 324–332, <https://doi.org/10.1063/1.1722370>.
- Milly, P. C. D., and K. A. Dunne, 2016: Potential evapotranspiration and continental drying. *Nat. Climate Change*, **6**, 946–949, <https://doi.org/10.1038/nclimate3046>.
- , R. T. Wetherald, K. A. Dunne, and T. L. Delworth, 2002: Increasing risk of great floods in a changing climate. *Nature*, **415**, 514–517, <https://doi.org/10.1038/415514a>.
- , J. Betancourt, M. Falkenmark, R. M. Hirsch, Z. W. Kundzewicz, D. P. Lettenmaier, and R. J. Stouffer, 2008: Stationarity is dead: Whither water management? *Science*, **319**, 573–574, <https://doi.org/10.1126/science.1151915>.
- Mintz, Y., 1984: The sensitivity of numerically simulated climates to land-surface conditions. *The Global Climate*, J. H. Houghton, Ed., Cambridge University Press, 79–105.
- , and Y. V. Serafini, 1992: A global monthly climatology of soil moisture and water balance. *Climate Dyn.*, **8**, 13–27, <https://doi.org/10.1007/BF00209340>.
- Miralles, D. G., T. R. H. Holmes, R. A. M. De Jeu, J. H. Gash, A. G. C. A. Meesters, and A. J. Dolman, 2011: Global land-surface evaporation estimated from satellite-based observations. *Hydrol. Earth Syst. Sci.*, **15**, 453–469, <https://doi.org/10.5194/hess-15-453-2011>.
- Mitchell, K. E., and Coauthors, 2004: The multi-institution North American Land Data Assimilation System (NLDAS): Utilizing multiple GCIP products and partners in a continental distributed hydrological modeling system. *J. Geophys. Res.*, **109**, D07S90, <https://doi.org/10.1029/2003JD003823>.

- Mladenova, I. E., and Coauthors, 2014: Remote monitoring of soil moisture using passive microwave-based techniques—Theoretical basis and overview of selected algorithms for AMSR-E. *Remote Sens. Environ.*, **144**, 197–213, <https://doi.org/10.1016/j.rse.2014.01.013>.
- Mockus, V., 1972: Section 4 Hydrology: Chapter 21 Design Hydrographs. *SCS National Engineering Handbook*, 127 pp., <https://www.wcc.nrcs.usda.gov/ftpref/wntsc/H&H/NEHhydrology/ch21.pdf>.
- Mohanty, B. P., M. H. Cosh, V. Lakshmi, and C. Montzka, 2017: Soil moisture remote sensing: State-of-the-science. *Vadose Zone J.*, **16**, <https://doi.org/10.2136/vzj2016.10.0105>.
- Molotch, N. P., P. D. Brooks, S. P. Burns, M. Litvak, R. K. Monson, J. R. McConnell, and K. Musselman, 2009: Ecohydrological controls on snowmelt partitioning in mixed-conifer sub-alpine forests. *Ecohydrology*, **2**, 129–142, <https://doi.org/10.1002/eco.48>.
- Monson, O. W., 1934: The snow survey as an index to summer precipitation. *Mon. Wea. Rev.*, **62**, 322–330, [https://doi.org/10.1175/1520-0493\(1934\)62<322:TSSAAI>2.0.CO;2](https://doi.org/10.1175/1520-0493(1934)62<322:TSSAAI>2.0.CO;2).
- Montanari, A., J. Bahr, G. Blöschl, X. Cai, D. S. Mackay, A. M. Michalak, H. Rajaram, and G. Sander, 2015: Fifty years of *Water Resources Research*: Legacy and perspectives for the science of hydrology. *Water Resour. Res.*, **51**, 6797–6803, <https://doi.org/10.1002/2015WR017998>.
- Monteith, J. L., 1965: Evaporation and environment. *Symp. Soc. Exp. Biol.*, **19**, 205–234.
- Morcrette, C. J., and Coauthors, 2018: Introduction to CAUSES: Description of weather and climate models and their near-surface temperature errors in 5 day hindcasts near the Southern Great Plains. *J. Geophys. Res. Atmos.*, **123**, 2655–2683, <https://doi.org/10.1002/2017JD027199>.
- Mu, Q., M. Zhao, and S. W. Running, 2011: Improvements to a MODIS global terrestrial evapotranspiration algorithm. *Remote Sens. Environ.*, **115**, 1781–1800, <https://doi.org/10.1016/j.rse.2011.02.019>.
- Mueller, B., and Coauthors, 2013: Benchmark products for land evapotranspiration: LandFlux-EVAL multi-data set synthesis. *Hydrol. Earth Syst. Sci.*, **17**, 3707–3720, <https://doi.org/10.5194/hess-17-3707-2013>.
- Mulvaney, T. J., 1851: On the use of self-registering rain and flood gauges in making observations of the relations of rainfall and flood discharges in a given catchment. *Proc. Inst. Civ. Eng. Irel.*, **4**, 18–33.
- Murray, J., 1887: On the total annual rainfall on the land of the globe, and the relation of rainfall to the annual discharge of rivers. *Scott. Geogr. Mag.*, **3**, 65–77, <https://doi.org/10.1080/14702548708554511>.
- Nagler, F. A., 1933: Certain flood-flow phenomena of Iowa Rivers. *Mon. Wea. Rev.*, **61**, 5–7, [https://doi.org/10.1175/1520-0493\(1933\)61<5:CFPOIR>2.0.CO;2](https://doi.org/10.1175/1520-0493(1933)61<5:CFPOIR>2.0.CO;2).
- National Academies of Sciences, Engineering, and Medicine, 2018: *Thriving on Our Changing Planet: A Decadal Strategy for Earth Observation from Space*. National Academies Press, 700 pp., <https://doi.org/10.17226/24938>.
- National Research Council, 1991: *Opportunities in the Hydrologic Sciences*. National Academies Press, 371 pp., <https://doi.org/10.17226/1543>.
- , 2006: *Completing the Forecast: Characterizing and Communicating Uncertainty for Better Decisions Using Weather and Climate Forecasts*. National Academies Press, 124 pp., <https://doi.org/10.17226/11699>.
- , 2007: *Earth Science and Applications from Space: National Imperatives for the Next Decade and Beyond*. National Academies Press, 454 pp., <https://doi.org/10.17226/11820>.
- Nearing, G. S., D. M. Mocko, C. D. Peters-Lidard, S. V. Kumar, and Y. Xia, 2016: Benchmarking NLDAS-2 soil moisture and evapotranspiration to separate uncertainty contributions. *J. Hydrometeorol.*, **17**, 745–759, <https://doi.org/10.1175/JHM-D-15-0063.1>.
- Nemani, R. R., S. W. Running, R. R. Nemani, and S. W. Running, 1989: Estimation of regional surface resistance to evapotranspiration from NDVI and thermal-IR AVHRR data. *J. Appl. Meteorol.*, **28**, 276–284, [https://doi.org/10.1175/1520-0450\(1989\)028<0276:EORSRT>2.0.CO;2](https://doi.org/10.1175/1520-0450(1989)028<0276:EORSRT>2.0.CO;2).
- Newman, B. D., and Coauthors, 2006: Ecohydrology of water-limited environments: A scientific vision. *Water Resour. Res.*, **42**, W06302, <https://doi.org/10.1029/2005WR004141>.
- Nguyen, P., and Coauthors, 2017: Exploring trends through “RainSphere”: Research data transformed into public knowledge. *Bull. Amer. Meteor. Soc.*, **98**, 653–658, <https://doi.org/10.1175/BAMS-D-16-0036.1>.
- , A. Thorstensen, S. Sorooshian, K. Hsu, A. Aghakouchak, H. Ashouri, H. Tran, and D. Braithwaite, 2018: Global precipitation trends across spatial scales using satellite observations. *Bull. Amer. Meteor. Soc.*, **99**, 689–697, <https://doi.org/10.1175/BAMS-D-17-0065.1>.
- Niu, G.-Y., Z.-L. Yang, R. E. Dickinson, L. E. Gulden, and H. Su, 2007: Development of a simple groundwater model for use in climate models and evaluation with Gravity Recovery and Climate Experiment data. *J. Geophys. Res.*, **112**, D07103, <https://doi.org/10.1029/2006JD007522>.
- , and Coauthors, 2011: The community Noah land surface model with multiparameterization options (Noah-MP): 1. Model description and evaluation with local-scale measurements. *J. Geophys. Res.*, **116**, D12109, <https://doi.org/10.1029/2010JD015139>.
- Njoku, E. G., and J.-A. Kong, 1977: Theory for passive microwave remote sensing of near-surface soil moisture. *J. Geophys. Res.*, **82**, 3108–3118, <https://doi.org/10.1029/JB082i020p03108>.
- Noaman, M. N., and D. El Quosy, 2017: Hydrology of the Nile and ancient agriculture. *Irrigated Agriculture in Egypt: Past, Present and Future*, M. Staoh and S. Aboulroos, Eds., Springer, 9–28, https://doi.org/10.1007/978-3-319-30216-4_2.
- O’Connell, P. E., and R. T. Clarke, 1981: Adaptive hydrological forecasting—A review. *Hydrol. Sci. Bull.*, **26**, 179–205, <https://doi.org/10.1080/02626668109490875>.
- Ohara, N., M. L. Kavvas, S. Kure, Z. Q. Chen, S. Jang, and E. Tan, 2011: Physically based estimation of maximum precipitation over American River watershed, California. *J. Hydrol. Eng.*, **16**, 351–361, [https://doi.org/10.1061/\(ASCE\)HE.1943-5584.0000324](https://doi.org/10.1061/(ASCE)HE.1943-5584.0000324).
- Oki, T., 1999: The Global Water Cycle. *Global Energy and Water Cycles*, K. A. Browning and R. J. Gurney, Eds., Cambridge University Press, 10–29.
- , and S. Kanae, 2006: Global hydrological cycles and water resources. *Science*, **313**, 1068–1072, <https://doi.org/10.1126/science.1128845>.
- Oleson, K. W., and Coauthors, 2010: Technical description of version 4.0 of the Community Land Model (CLM). NCAR Tech. Note NCAR/TN-478+STR, 257 pp., <https://doi.org/10.5065/D6FB50WZ>.
- O’Neill, P. E., S. Chan, E. G. Njoku, T. J. Jackson, and R. Bindlish, 2015: Soil Moisture Active Passive (SMAP) Algorithm Theoretical Basis Document Level 2 & 3 Soil Moisture (Passive) Data Products. JPL D-66480, 80 pp., <https://smap.jpl.nasa.gov/documents/>.
- Owe, M., R. de Jeu, and J. Walker, 2001: A methodology for surface soil moisture and vegetation optical depth retrieval using

- the microwave polarization difference index. *IEEE Trans. Geosci. Remote Sens.*, **39**, 1643–1654, <https://doi.org/10.1109/36.942542>.
- , —, and T. Holmes, 2008: Multisensor historical climatology of satellite-derived global land surface moisture. *J. Geophys. Res.*, **113**, F01002, <https://doi.org/10.1029/2007JF000769>.
- Ozdogan, M., M. Rodell, H. K. Beaudoin, and D. L. Toll, 2010: Simulating the effects of irrigation over the United States in a land surface model based on satellite-derived agricultural data. *J. Hydrometeorol.*, **11**, 171–184, <https://doi.org/10.1175/2009JHM1116.1>.
- Pagano, T., A. Wood, K. Werner, and R. Tama-Sweet, 2014: Western U.S. water supply forecasting: A tradition evolves. *Eos, Trans. Amer. Geophys. Union*, **95**, 28–29, <https://doi.org/10.1002/2014EO030007>.
- Painter, T. H., K. Rittger, C. McKenzie, P. Slaughter, R. E. Davis, and J. Dozier, 2009: Retrieval of subpixel snow covered area, grain size, and albedo from MODIS. *Remote Sens. Environ.*, **113**, 868–879, <https://doi.org/10.1016/j.rse.2009.01.001>.
- , and Coauthors, 2016: The Airborne Snow Observatory: Fusion of scanning lidar, imaging spectrometer, and physically-based modeling for mapping snow water equivalent and snow albedo. *Remote Sens. Environ.*, **184**, 139–152, <https://doi.org/10.1016/j.rse.2016.06.018>.
- Paiva, R. C. D., W. Collischonn, M. P. Bonnet, and L. G. G. De Gonçalves, 2012: On the sources of hydrological prediction uncertainty in the Amazon. *Hydrol. Earth Syst. Sci.*, **16**, 3127–3137, <https://doi.org/10.5194/hess-16-3127-2012>.
- Pal, J. S., and E. A. B. Eltahir, 2001: Pathways relating soil moisture conditions to future summer rainfall within a model of the land–atmosphere system. *J. Climate*, **14**, 1227–1242, [https://doi.org/10.1175/1520-0442\(2001\)014<1227:PRSMCT>2.0.CO;2](https://doi.org/10.1175/1520-0442(2001)014<1227:PRSMCT>2.0.CO;2).
- Pechlivanidis, I. G., T. Bosshard, H. Spångmyr, G. Lindström, D. Gustafsson, and B. Arheimer, 2014: Uncertainty in the Swedish operational hydrological forecasting systems. *Vulnerability, Uncertainty, and Risk*, American Society of Civil Engineers, 253–262, <https://doi.org/10.1061/9780784413609.026>.
- Peck, E. L., 1980: Design of precipitation networks. *Bull. Amer. Meteor. Soc.*, **61**, 894–906, <https://doi.org/10.1175/1520-0477-61.8.894>.
- , and M. J. Brown, 1962: An approach to the development of isohyetal maps for mountainous areas. *J. Geophys. Res.*, **67**, 681–694, <https://doi.org/10.1029/JZ067i002p00681>.
- Peng, J., A. Loew, O. Merlin, and N. E. C. Verhoest, 2017: A review of spatial downscaling of satellite remotely sensed soil moisture. *Rev. Geophys.*, **55**, 341–366, <https://doi.org/10.1002/2016RG000543>.
- Penman, H. L., 1948: Natural evaporation from open water, bare soil and grass. *Proc. Roy. Soc. London*, **193A**, 120–145, <https://doi.org/10.1098/rspa.1948.0037>.
- Penn, C. A., L. A. Bearup, R. M. Maxwell, and D. W. Clow, 2016: Numerical experiments to explain multiscale hydrological responses to mountain pine beetle tree mortality in a headwater watershed. *Water Resour. Res.*, **52**, 3143–3161, <https://doi.org/10.1002/2015WR018300>.
- Peñuelas, J., M. F. Garbulsky, and I. Filella, 2011: Photochemical reflectance index (PRI) and remote sensing of plant CO₂ uptake. *New Phytol.*, **191**, 596–599, <https://doi.org/10.1111/j.1469-8137.2011.03791.x>.
- Peral, E., and Coauthors, 2018: Radar technologies for Earth remote sensing from CubeSat Platforms. *Proc. IEEE*, **106**, 404–418, <https://doi.org/10.1109/JPROC.2018.2793179>.
- Peters-Lidard, C. D., and Coauthors, 2017: Scaling, similarity, and the fourth paradigm for hydrology. *Hydrol. Earth Syst. Sci.*, **21**, 3701–3713, <https://doi.org/10.5194/hess-21-3701-2017>.
- Pfister, L., H. H. G. Savenije, and F. Fenicia, 2009: *Leonardo Da Vinci's Water Theory: On the Origin and Fate of Water*. International Association of Hydrological Sciences, 92 pp.
- Philip, J. R., 1957: The theory of infiltration: 1. The infiltration equation and its solution. *Soil Sci.*, **83**, 345–357, <https://doi.org/10.1097/00010694-195705000-00002>.
- Piao, S., P. Friedlingstein, P. Ciais, N. de Noblet-Ducoudre, D. Labat, and S. Zaehle, 2007: Changes in climate and land use have a larger direct impact than rising CO₂ on global river runoff trends. *Proc. Natl. Acad. Sci. USA*, **104**, 15 242–15 247, <https://doi.org/10.1073/pnas.0707213104>.
- Pielke, R. A., Sr., and Coauthors, 2011: Land use/land cover changes and climate: Modeling analysis and observational evidence. *Wiley Interdiscip. Rev.: Climate Change*, **2**, 828–850, <https://doi.org/10.1002/wcc.144>.
- Porporato, A., and I. Rodríguez-Iturbe, 2002: Ecohydrology—A challenging multidisciplinary research perspective. *Hydrol. Sci. J.*, **47**, 811–821, <https://doi.org/10.1080/02626660209492985>.
- Priestley, C. H. B., and R. J. Taylor, 1972: On the assessment of surface heat flux and evaporation using large-scale parameters. *Mon. Wea. Rev.*, **100**, 81–92, [https://doi.org/10.1175/1520-0493\(1972\)100<0081:OTAOSH>2.3.CO;2](https://doi.org/10.1175/1520-0493(1972)100<0081:OTAOSH>2.3.CO;2).
- Prudhomme, C., and Coauthors, 2017: Hydrological outlook UK: An operational streamflow and groundwater level forecasting system at monthly to seasonal time scales. *Hydrol. Sci. J.*, **62**, 2753–2768, <https://doi.org/10.1080/02626667.2017.1395032>.
- Quiring, S. M., T. W. Ford, J. K. Wang, A. Khong, E. Harris, T. Lindgren, D. W. Goldberg, and Z. Li, 2016: The North American Soil Moisture Database: Development and applications. *Bull. Amer. Meteor. Soc.*, **97**, 1441–1459, <https://doi.org/10.1175/BAMS-D-13-00263.1>.
- Rahman, M., M. Sulis, and S. J. Kollet, 2015: The subsurface-land surface-atmosphere connection under convective conditions. *Adv. Water Resour.*, **83**, 240–249, <https://doi.org/10.1016/j.advwatres.2015.06.003>.
- Rajaram, H., and Coauthors, 2015: A reflection on the first 50 years of *Water Resources Research*. *Water Resour. Res.*, **51**, 7829–7837, <https://doi.org/10.1002/2015WR018089>.
- Rakhecha, P. R., and V. P. Singh, 2009: Greatest point and areal rainfalls. *Applied Hydrometeorology*, Springer, 163–189.
- Ramsay, B., 1998: The interactive multisensor snow and ice mapping system. *Hydrol. Processes*, **12**, 1537–1546, [https://doi.org/10.1002/\(SICI\)1099-1085\(199808/09\)12:10<11<1537::AID-HYP679>3.0.CO;2-A](https://doi.org/10.1002/(SICI)1099-1085(199808/09)12:10<11<1537::AID-HYP679>3.0.CO;2-A).
- Rasmusson, E. M., 1967: Atmospheric water vapor transport and the water balance of North America. Part I: Characteristics of the water vapor flux field. *Mon. Wea. Rev.*, **95**, 403–426, [https://doi.org/10.1175/1520-0493\(1967\)095<0403:AWVTAT>2.3.CO;2](https://doi.org/10.1175/1520-0493(1967)095<0403:AWVTAT>2.3.CO;2).
- , 1968: Atmospheric water vapor transport and the water balance of North America. Part II: Large-scale water balance investigations. *Mon. Wea. Rev.*, **96**, 720–734, [https://doi.org/10.1175/1520-0493\(1968\)096<0720:AWVTAT>2.0.CO;2](https://doi.org/10.1175/1520-0493(1968)096<0720:AWVTAT>2.0.CO;2).
- Rastogi, D., and Coauthors, 2017: Effects of climate change on probable maximum precipitation: A sensitivity study over the Alabama-Coosa-Tallapoosa River Basin. *J. Geophys. Res. Atmos.*, **122**, 4808–4828, <https://doi.org/10.1002/2016JD026001>.
- Rawls, W. J., D. L. Brakensiek, and K. E. Saxton, 1982: Estimation of Soil Water Properties. *Trans. ASAE*, **25**, 1316–1320, <https://doi.org/10.13031/2013.33720>.

- Reager, J. T., A. S. Gardner, J. S. Famiglietti, D. N. Wiese, A. Eicker, and M. H. Lo, 2016: A decade of sea level rise slowed by climate-driven hydrology. *Science*, **351**, 699–703, <https://doi.org/10.1126/science.aad8386>.
- Reed, S., V. Koren, M. Smith, Z. Zhang, F. Moreda, D.-J. Seo, and DMIP Participants, 2004: Overall distributed model intercomparison project results. *J. Hydrol.*, **298**, 27–60, <https://doi.org/10.1016/j.jhydrol.2004.03.031>.
- , J. Schaake, and Z. Zhang, 2007: A distributed hydrologic model and threshold frequency-based method for flash flood forecasting at ungauged locations. *J. Hydrol.*, **337**, 402–420, <https://doi.org/10.1016/j.jhydrol.2007.02.015>.
- Reggiani, P., M. Sivapalan, and S. Majid Hassanizadeh, 1998: A unifying framework for watershed thermodynamics: Balance equations for mass, momentum, energy and entropy, and the second law of thermodynamics. *Adv. Water Resour.*, **22**, 367–398, [https://doi.org/10.1016/S0309-1708\(98\)00012-8](https://doi.org/10.1016/S0309-1708(98)00012-8).
- , S. M. Hassanizadeh, M. Sivapalan, and W. G. Gray, 1999: A unifying framework for watershed thermodynamics: constitutive relationships. *Adv. Water Resour.*, **23**, 15–39, [https://doi.org/10.1016/S0309-1708\(99\)00005-6](https://doi.org/10.1016/S0309-1708(99)00005-6).
- , M. Sivapalan, and S. M. Hassanizadeh, 2000: Conservation equations governing hillslope responses: Exploring the physical basis of water balance. *Water Resour. Res.*, **36**, 1845–1863, <https://doi.org/10.1029/2000WR900066>.
- , —, —, and W. G. Gray, 2001: Coupled equations for mass and momentum balance in a stream network: theoretical derivation and computational experiments. *Proc. Roy. Soc. London*, **457A**, 157–189, <https://doi.org/10.1098/rspa.2000.0661>.
- Richards, L. A., 1931: Capillary conduction of liquids through porous mediums. *Physics*, **1**, 318–817, <https://doi.org/10.1063/1.1745010>.
- Richey, A. S., B. F. Thomas, M. H. Lo, J. S. Famiglietti, S. Swenson, and M. Rodell, 2015: Uncertainty in global groundwater storage estimates in a Total Groundwater Stress framework. *Water Resour. Res.*, **51**, 5198–5216, <https://doi.org/10.1002/2015WR017351>.
- Richter, B. D., J. V. Baumgartner, J. Powell, and D. P. Braun, 1996: a method for assessing hydrologic alteration within ecosystems. *Conserv. Biol.*, **10**, 1163–1174, <https://doi.org/10.1046/j.1523-1739.1996.10041163.x>.
- Rienecker, M. M., and Coauthors, 2011: MERRA: NASA's Modern-Era Retrospective Analysis for Research and Applications. *J. Climate*, **24**, 3624–3648, <https://doi.org/10.1175/JCLI-D-11-00015.1>.
- Rihani, J. F., F. K. Chow, and R. M. Maxwell, 2015: Isolating effects of terrain and soil moisture heterogeneity on the atmospheric boundary layer: Idealized simulations to diagnose land-atmosphere feedbacks. *J. Adv. Model. Earth Syst.*, **7**, 915–937, <https://doi.org/10.1002/2014MS000371>.
- Robinson, D. A., S. B. Jones, J. M. Wraith, D. Or, and S. P. Friedman, 2003: A review of advances in dielectric and electrical conductivity measurement in soils using time domain reflectometry. *Vadose Zone J.*, **2**, 444, <https://doi.org/10.2136/vzj2003.4440>.
- , and Coauthors, 2008: Soil moisture measurement for ecological and hydrological watershed-scale observatories: A review. *Vadose Zone J.*, **7**, 358, <https://doi.org/10.2136/vzj2007.0143>.
- Robock, A., K. Y. Vinnikov, G. Srinivasan, J. K. Entin, S. E. Hollinger, N. A. Speranskaya, S. Liu, and A. Namkhai, 2000: The Global Soil Moisture Data Bank. *Bull. Amer. Meteor. Soc.*, **81**, 1281–1299, [https://doi.org/10.1175/1520-0477\(2000\)081<1281:TGSMDB>2.3.CO;2](https://doi.org/10.1175/1520-0477(2000)081<1281:TGSMDB>2.3.CO;2).
- Rodell, M., and J. S. Famiglietti, 2001: An analysis of terrestrial water storage variations in Illinois with implications for the Gravity Recovery and Climate Experiment (GRACE). *Water Resour. Res.*, **37**, 1327–1340, <https://doi.org/10.1029/2000WR900306>.
- , and Coauthors, 2004: The Global Land Data Assimilation System. *Bull. Amer. Meteor. Soc.*, **85**, 381–394, <https://doi.org/10.1175/BAMS-85-3-381>.
- , J. Chen, H. Kato, J. S. Famiglietti, J. Nigro, and C. R. Wilson, 2007: Estimating groundwater storage changes in the Mississippi River basin (USA) using GRACE. *Hydrogeol. J.*, **15**, 159–166, <https://doi.org/10.1007/s10040-006-0103-7>.
- , I. Velicogna, and J. S. Famiglietti, 2009: Satellite-based estimates of groundwater depletion in India. *Nature*, **460**, 999–1002, <https://doi.org/10.1038/nature08238>.
- , E. B. McWilliams, J. S. Famiglietti, H. K. Beaudoin, and J. Nigro, 2011: Estimating evapotranspiration using an observation based terrestrial water budget. *Hydrol. Processes*, **25**, 4082–4092, <https://doi.org/10.1002/hyp.8369>.
- , and Coauthors, 2015: The observed state of the water cycle in the early twenty-first century. *J. Climate*, **28**, 8289–8318, <https://doi.org/10.1175/JCLI-D-14-00555.1>.
- , J. S. Famiglietti, D. N. Wiese, J. T. Reager, H. K. Beaudoin, F. W. Landerer, and M.-H. Lo, 2018: Emerging trends in global freshwater availability. *Nature*, **557**, 651–659, <https://doi.org/10.1038/s41586-018-0123-1>.
- Rodríguez-Iturbe, I., 2000: Ecohydrology: A hydrologic perspective of climate-soil-vegetation dynamics. *Water Resour. Res.*, **36**, 3–9, <https://doi.org/10.1029/1999WR900210>.
- , and J. M. Mejía, 1974a: On the transformation of point rainfall to areal rainfall. *Water Resour. Res.*, **10**, 729–735, <https://doi.org/10.1029/WR010i004p00729>.
- , and —, 1974b: The design of rainfall networks in time and space. *Water Resour. Res.*, **10**, 713–728, <https://doi.org/10.1029/WR010i004p00713>.
- Romanov, P., G. Gutman, and I. Csizsar, 2000: Automated monitoring of snow cover over North America with multispectral satellite data. *J. Appl. Meteor.*, **39**, 1866–1880, [https://doi.org/10.1175/1520-0450\(2000\)039<1866:AMOSCO>2.0.CO;2](https://doi.org/10.1175/1520-0450(2000)039<1866:AMOSCO>2.0.CO;2).
- Rosenthal, W., and J. Dozier, 1996: Automated mapping of montane snow cover at subpixel resolution from the Landsat Thematic Mapper. *Water Resour. Res.*, **32**, 115–130, <https://doi.org/10.1029/95WR02718>.
- Rouhani, H., and R. Leconte, 2016: A novel method to estimate the maximization ratio of the Probable Maximum Precipitation (PMP) using regional climate model output. *Water Resour. Res.*, **52**, 7347–7365, <https://doi.org/10.1002/2016WR018603>.
- Rowlands, D. D., S. B. Luthcke, S. M. Klosko, F. G. R. Lemoine, D. S. Chinn, J. J. McCarthy, C. M. Cox, and O. B. Anderson, 2005: Resolving mass flux at high spatial and temporal resolution using GRACE intersatellite measurements. *Geophys. Res. Lett.*, **32**, L04310, <https://doi.org/10.1029/2004GL021908>.
- Rowntree, P. R., and J. A. Bolton, 1983: Simulation of the atmospheric response to soil moisture anomalies over Europe. *Quart. J. Roy. Meteor. Soc.*, **109**, 501–526, <https://doi.org/10.1002/qj.49710946105>.
- Sahoo, S., T. Russo, and U. Lall, 2016: Comment on “Quantifying renewable groundwater stress with GRACE” by Alexandra S. Richey et al. *Water Resour. Res.*, **52**, 4184–4187, <https://doi.org/10.1002/2015WR018085>.
- Salas, F. R., and Coauthors, 2018: Towards real-time continental scale streamflow simulation in continuous and discrete space.

- J. Amer. Water Resour. Assoc.*, **54**, 7–27, <https://doi.org/10.1111/1752-1688.12586>.
- Samaniego, L., R. Kumar, and S. Attinger, 2010: Multiscale parameter regionalization of a grid-based hydrologic model at the mesoscale. *Water Resour. Res.*, **46**, W05523, <https://doi.org/10.1029/2008WR007327>.
- Samouëlian, A., I. Cousin, A. Tabbagh, A. Bruand, and G. Richard, 2005: Electrical resistivity survey in soil science: A review. *Soil Tillage Res.*, **83**, 173–193, <https://doi.org/10.1016/j.still.2004.10.004>.
- Santanello, J. A., and Coauthors, 2018: Land–atmosphere interactions: The LoCo perspective. *Bull. Amer. Meteor. Soc.*, **99**, 1253–1272, <https://doi.org/10.1175/BAMS-D-17-0001.1>.
- Schaake, J. C., T. M. Hamill, R. Buizza, and M. Clark, 2007: HEPEX: The Hydrological Ensemble Prediction Experiment. *Bull. Amer. Meteor. Soc.*, **88**, 1541–1548, <https://doi.org/10.1175/BAMS-88-10-1541>.
- Schär, C., D. Lüthi, U. Beyerle, and E. Heise, 1999: The soil–precipitation feedback: A process study with a regional climate model. *J. Climate*, **12**, 722–741, [https://doi.org/10.1175/1520-0442\(1999\)012<0722:TSPFAP>2.0.CO;2](https://doi.org/10.1175/1520-0442(1999)012<0722:TSPFAP>2.0.CO;2).
- Schmugge, T., P. Gloersen, T. Wilheit, and F. Geiger, 1974: Remote sensing of soil moisture with microwave radiometers. *J. Geophys. Res.*, **79**, 317–323, <https://doi.org/10.1029/JB079i002p00317>.
- Schubert, S. D., M. J. Suarez, P. J. Pegion, R. D. Koster, and J. T. Bacmeister, 2004: On the cause of the 1930s Dust Bowl. *Science*, **303**, 1855–1859, <https://doi.org/10.1126/science.1095048>.
- Schumann, G., D. Kirschbaum, E. Anderson, and K. Rashid, 2016: Role of Earth observation data in disaster response and recovery: From science to capacity building. *Earth Science Satellite Applications*, F. Hossain, Ed., Springer, 119–146, https://doi.org/10.1007/978-3-319-33438-7_5.
- Sellers, P. J., and F. G. Hall, 1992: FIFE in 1992: Results, scientific gains, and future research directions. *J. Geophys. Res.*, **97**, 19 091–19 109, <https://doi.org/10.1029/92JD02173>.
- , Y. Mintz, Y. C. Sud, and A. Dalcher, 1986: A Simple Biosphere model (SiB) for use within general circulation models. *J. Atmos. Sci.*, **43**, 505–531, [https://doi.org/10.1175/1520-0469\(1986\)043<0505:ASBMFU>2.0.CO;2](https://doi.org/10.1175/1520-0469(1986)043<0505:ASBMFU>2.0.CO;2).
- , and Coauthors, 1995: The Boreal Ecosystem—Study (BOREAS): An overview and early results from the 1994 field year. *Bull. Amer. Meteor. Soc.*, **76**, 1549–1577, [https://doi.org/10.1175/1520-0477\(1995\)076<1549:TBESAO>2.0.CO;2](https://doi.org/10.1175/1520-0477(1995)076<1549:TBESAO>2.0.CO;2).
- Senay, G. B., M. E. Budde, and J. P. Verdin, 2011: Enhancing the Simplified Surface Energy Balance (SSEB) approach for estimating landscape ET: Validation with the METRIC model. *Agric. Water Manage.*, **98**, 606–618, <https://doi.org/10.1016/j.agwat.2010.10.014>.
- Seneviratne, S. I., D. Lüthi, M. Litschi, and C. Schär, 2006: Land–atmosphere coupling and climate change in Europe. *Nature*, **443**, 205–209, <https://doi.org/10.1038/nature05095>.
- Sheffield, J., E. F. Wood, and M. L. Roderick, 2012: Little change in global drought over the past 60 years. *Nature*, **491**, 435–438, <https://doi.org/10.1038/nature11575>.
- Sherman, L. K., 1932: Streamflow from rainfall by the unit-hydrograph method. *Eng. News-Rec.*, **108**, 501–505.
- Sherwood, S., and Q. Fu, 2014: A drier future? *Science*, **343**, 737–739, <https://doi.org/10.1126/science.1247620>.
- Shukla, J., and Y. Mintz, 1982: Influence of land-surface evapotranspiration on the Earth’s climate. *Science*, **215**, 1498–1501, <https://doi.org/10.1126/science.215.4539.1498>.
- Shuman, J. W., 1929: Correlation between rainfall and run-off. *Mon. Wea. Rev.*, **57**, 179–184, [https://doi.org/10.1175/1520-0493\(1929\)57<179:CBRAR>2.0.CO;2](https://doi.org/10.1175/1520-0493(1929)57<179:CBRAR>2.0.CO;2).
- Shuttleworth, W. J., 2007: Putting the “vap” into evaporation. *Hydrol. Earth Syst. Sci.*, **11**, 210–244, <https://doi.org/10.5194/hess-11-210-2007>.
- , 2008: Evapotranspiration measurement methods. *SouthWest Hydrology*, Vol. 7, SAHRA, The University of Arizona, Tucson, AZ, 22–23, http://web.sahra.arizona.edu/swhydro/archive/V7_N1/feature3.pdf.
- , and J. S. Wallace, 1985: Evaporation from sparse crops - An energy combination theory. *Quart. J. Roy. Meteor. Soc.*, **111**, 839–855, <https://doi.org/10.1002/qj.49711146910>.
- Sidele, R. C., and Y. Onda, 2004: Hydrogeomorphology: Overview of an emerging science. *Hydrol. Processes*, **18**, 597–602, <https://doi.org/10.1002/hyp.1360>.
- Simpson, J., R. F. Adler, and G. R. North, 1988: A proposed Tropical Rainfall Measuring Mission (TRMM) satellite. *Bull. Amer. Meteor. Soc.*, **69**, 278–295, [https://doi.org/10.1175/1520-0477\(1988\)069<0278:APTRMM>2.0.CO;2](https://doi.org/10.1175/1520-0477(1988)069<0278:APTRMM>2.0.CO;2).
- Singh, R., K. Subramanian, and J. C. Refsgaard, 1999: Hydrological modelling of a small watershed using MIKE SHE for irrigation planning. *Agric. Water Manage.*, **41**, 149–166, [https://doi.org/10.1016/S0378-3774\(99\)00022-0](https://doi.org/10.1016/S0378-3774(99)00022-0).
- Singh, V. P., 2018: Hydrologic modeling: Progress and future directions. *Geosci. Lett.*, **5**, 15, <https://doi.org/10.1186/s40562-018-0113-z>.
- , and D. A. Woolhiser, 2002: Mathematical modeling of watershed hydrology. *J. Hydrol. Eng.*, **7**, 270–292, [https://doi.org/10.1061/\(ASCE\)1084-0699\(2002\)7:4\(270\)](https://doi.org/10.1061/(ASCE)1084-0699(2002)7:4(270)).
- Sivapalan, M., 2005: Pattern, process and function: Elements of a unified theory of hydrology at the catchment scale. *Encyclopedia of Hydrological Sciences*, M. G. Anderson and J. J. McDonnell, Eds., 1–27, John Wiley & Sons, Ltd, <https://doi.org/10.1002/0470848944.hsa012>.
- , 2018: From engineering hydrology to Earth system science: Milestones in the transformation of hydrologic science. *Hydrol. Earth Syst. Sci.*, **22**, 1665–1693, <https://doi.org/10.5194/hess-2017-670>.
- , and E. F. Wood, 1987: A multidimensional model of non-stationary space-time rainfall at the catchment scale. *Water Resour. Res.*, **23**, 1289–1299, <https://doi.org/10.1029/WR023i007p01289>, <https://doi.org/10.1002/2017WR021396>.
- , and P. C. D. Milly, 1989: On the relationship between the time condensation approximation and the flux concentration relation. *J. Hydrol.*, **105**, 357–367, [https://doi.org/10.1016/0022-1694\(89\)90113-3](https://doi.org/10.1016/0022-1694(89)90113-3).
- , and G. Blöschl, 2017: The growth of hydrological understanding: technologies, ideas, and societal needs shape the field. *Water Resour. Res.*, **53**, 8137–8146.
- Skofronick-Jackson, G., and Coauthors, 2017: The Global Precipitation Measurement (GPM) mission for science and society. *Bull. Amer. Meteor. Soc.*, **98**, 1679–1695, <https://doi.org/10.1175/BAMS-D-15-00306.1>.
- Smith, M. B., K. P. Georgakakos, and X. Liang, 2004: The distributed model intercomparison project (DMIP). *J. Hydrol.*, **298**, 1–3, <https://doi.org/10.1016/j.jhydrol.2004.05.001>.
- , and Coauthors, 2013: The distributed model intercomparison project – Phase 2: Experiment design and summary results of the western basin experiments. *J. Hydrol.*, **507**, 300–329, <https://doi.org/10.1016/j.jhydrol.2013.08.040>.
- Sollins, P., and F. M. McCorison, 1981: Nitrogen and carbon solution chemistry of an old growth coniferous forest watershed

- before and after cutting. *Water Resour. Res.*, **17**, 1409–1418, <https://doi.org/10.1029/WR017i005p01409>.
- Sperry, J. S., M. D. Venturas, W. R. L. Anderegg, M. Mencuccini, D. S. Mackay, Y. Wang, and D. M. Love, 2017: Predicting stomatal responses to the environment from the optimization of photosynthetic gain and hydraulic cost. *Plant Cell Environ.*, **40**, 816–830, <https://doi.org/10.1111/pce.12852>.
- Srinivasan, T. N., 2000: *Eight Lectures on India's Economic Reforms*. T. N. Srinivasan, Ed., Oxford University Press, 114 pp.
- Su, H., M. F. McCabe, E. F. Wood, Z. Su, and J. H. Prueger, 2005: Modeling Evapotranspiration during SMACEX: Comparing two approaches for local- and regional-scale prediction. *J. Hydrometeorol.*, **6**, 910–922, <https://doi.org/10.1175/JHM466.1>.
- Su, Z., 2002: The Surface Energy Balance System (SEBS) for estimation of turbulent heat fluxes. *Hydrol. Earth Syst. Sci.*, **6**, 85–100, <https://doi.org/10.5194/hess-6-85-2002>.
- Sutanto, S. J., and Coauthors, 2014: HESS Opinions “A perspective on isotope versus non-isotope approaches to determine the contribution of transpiration to total evaporation.” *Hydrol. Earth Syst. Sci.*, **18**, 2815–2827, <https://doi.org/10.5194/hess-18-2815-2014>.
- Sutton, C., T. M. Hamill, and T. T. Warner, 2006: Will perturbing soil moisture improve warm-season ensemble forecasts? A proof of concept. *Mon. Wea. Rev.*, **134**, 3174–3189, <https://doi.org/10.1175/MWR3248.1>.
- Swann, A. L. S., I. Y. Fung, and J. C. H. Chiang, 2012: Mid-latitude afforestation shifts general circulation and tropical precipitation. *Proc. Natl. Acad. Sci. USA*, **109**, 712–716, <https://doi.org/10.1073/pnas.1116706108>.
- Swenson, S., and J. Wahr, 2006: Estimating large-scale precipitation minus evapotranspiration from GRACE satellite gravity measurements. *J. Hydrometeorol.*, **7**, 252–270, <https://doi.org/10.1175/JHM478.1>.
- , P. J.-F. Yeh, J. Wahr, and J. Famiglietti, 2006: A comparison of terrestrial water storage variations from GRACE with in situ measurements from Illinois. *Geophys. Res. Lett.*, **33**, L16401, <https://doi.org/10.1029/2006GL026962>.
- Tague, C. L., and L. E. Band, 2001: Evaluating explicit and implicit routing for watershed hydro-ecological models of forest hydrology at the small catchment scale. *Hydrol. Processes*, **15**, 1415–1439, <https://doi.org/10.1002/hyp.171>.
- Tamisiea, M. E., E. W. Leuliette, J. L. Davis, and J. X. Mitrovica, 2005: Constraining hydrological and cryospheric mass flux in southeastern Alaska using space-based gravity measurements. *Geophys. Res. Lett.*, **32**, L20501, <https://doi.org/10.1029/2005GL023961>.
- Tao, W.-K., and Coauthors, 2016: TRMM latent heating retrieval: Applications and comparisons with field campaigns and large-scale analyses. *Multiscale Convection-Coupled Systems in the Tropics: A Tribute to Dr. Michio Yanai*, *Meteor. Monogr.*, No. 56, Amer. Meteor. Soc., <https://doi.org/10.1175/AMSMONOGRAPHS-D-15-0013.1>.
- Tapiador, F. J., C. F. Angelis, N. Viltard, F. Cuartero, and M. de Castro, 2011: On the suitability of regional climate models for reconstructing climatologies. *Atmos. Res.*, **101**, 739–751, <https://doi.org/10.1016/j.atmosres.2011.05.001>.
- , and Coauthors, 2012: Global precipitation measurement: Methods, datasets and applications. *Atmos. Res.*, **104–105**, 70–97, <https://doi.org/10.1016/j.atmosres.2011.10.021>.
- , and Coauthors, 2017: On the optimal measuring area for pointwise rainfall estimation: A dedicated experiment with 14 laser disdrometers. *J. Hydrometeorol.*, **18**, 753–760, <https://doi.org/10.1175/JHM-D-16-0127.1>.
- Tapley, B. D., S. Bettadpur, J. C. Ries, P. F. Thompson, and M. M. Watkins, 2004: GRACE measurements of mass variability in the Earth system. *Science*, **305**, 503–505, <https://doi.org/10.1126/science.1099192>.
- Tappeiner, U., G. Tappeiner, J. Aschenwald, E. Tasser, and B. Ostendorf, 2001: GIS-based modelling of spatial pattern of snow cover duration in an alpine area. *Ecol. Modell.*, **138**, 265–275, [https://doi.org/10.1016/S0304-3800\(00\)00407-5](https://doi.org/10.1016/S0304-3800(00)00407-5).
- Taylor, C. M., 2015: Detecting soil moisture impacts on convective initiation in Europe. *Geophys. Res. Lett.*, **42**, 4631–4638, <https://doi.org/10.1002/2015GL064030>.
- , R. A. M. De Jeu, F. Guichard, P. P. Harris, and W. A. Dorigo, 2012: Afternoon rain more likely over drier soils. *Nature*, **489**, 423–426, <https://doi.org/10.1038/nature11377>.
- Theis, C. V., 1935: The relation between the lowering of the Piezometric surface and the rate and duration of discharge of a well using groundwater storage. *Eos, Trans. Amer. Geophys. Union*, **16**, 519–524, <https://doi.org/10.1029/TR016i002p00519>.
- Thielen, J., J. Bartholmes, M.-H. Ramos, and A. de Roo, 2009: The European Flood Alert System – Part 1: Concept and development. *Hydrol. Earth Syst. Sci.*, **13**, 125–140, <https://doi.org/10.5194/hess-13-125-2009>.
- Thom, A. S., and H. R. Oliver, 1977: On Penman's equation for estimating regional evaporation. *Quart. J. Roy. Meteor. Soc.*, **103**, 345–357, <https://doi.org/10.1002/qj.49710343610>.
- Thomas, A. C., J. T. Reager, J. S. Famiglietti, and M. Rodell, 2014: A GRACE-based water storage deficit approach for hydrological drought characterization. *Geophys. Res. Lett.*, **41**, 1537–1545, <https://doi.org/10.1002/2014GL059323>.
- Thompson, S. E., and Coauthors, 2011: Patterns, puzzles and people: Implementing hydrologic synthesis. *Hydrol. Processes*, **25**, 3256–3266, <https://doi.org/10.1002/hyp.8234>.
- Thornthwaite, C. W., 1948: An approach toward a rational classification of climate. *Geogr. Rev.*, **38**, 55, <https://doi.org/10.2307/210739>.
- Todini, E., 2007: A mass conservative and water storage consistent variable parameter Muskingum-Cunge approach. *Hydrol. Earth Syst. Sci.*, **11**, 1645–1659, <https://doi.org/10.5194/hess-11-1645-2007>.
- Trenberth, K. E., 1999: Atmospheric moisture recycling: Role of advection and local evaporation. *J. Climate*, **12**, 1368–1381, [https://doi.org/10.1175/1520-0442\(1999\)012<1368:AMRROA>2.0.CO;2](https://doi.org/10.1175/1520-0442(1999)012<1368:AMRROA>2.0.CO;2).
- , L. Smith, T. Qian, A. Dai, and J. Fasullo, 2007: Estimates of the global water budget and its annual cycle using observational and model data. *J. Hydrometeorol.*, **8**, 758–769, <https://doi.org/10.1175/JHM600.1>.
- , A. Dai, G. Van Der Schrier, P. D. Jones, J. Barichivich, K. R. Briffa, and J. Sheffield, 2014: Global warming and changes in drought. *Nat. Climate Change*, **4**, 17–22, <https://doi.org/10.1038/nclimate2067>.
- Trescott, P. C., G. F. Pinder, and S. P. Larson, 1980: Finite difference model for aquifer simulation in two dimensions with results of numerical experiments. USGS Water-Resources Investigations, Book 7, Chapter C1, 116 pp., https://pubs.usgs.gov/twri/twri7-c1/html/pdf/TWRI_7-C1.pdf.
- Trier, S. B., F. Chen, and K. W. Manning, 2004: A study of convection initiation in a mesoscale model using high-resolution land surface initial conditions. *Mon. Wea. Rev.*, **132**, 2954–2976, <https://doi.org/10.1175/MWR2839.1>.

- Troch, P. A., C. Paniconi, and E. Emiel van Loon, 2003: Hillslope-storage Boussinesq model for subsurface flow and variable source areas along complex hillslopes: 1. Formulation and characteristic response. *Water Resour. Res.*, **39**, 1316, <https://doi.org/10.1029/2002WR001728>.
- Twine, T. E., and Coauthors, 2000: Correcting eddy-covariance flux underestimates over a grassland. *Agric. For. Meteorol.*, **103**, 279–300, [https://doi.org/10.1016/S0168-1923\(00\)00123-4](https://doi.org/10.1016/S0168-1923(00)00123-4).
- UN-Water, 2008: Transboundary waters: Sharing benefits, sharing responsibilities. Task Force on Transboundary Waters Thematic Paper, 16 pp., <https://www.unwater.org/publications/transboundary-waters-sharing-benefits-sharing-responsibilities/>.
- Valente, A., R. Morais, A. Tuli, J. W. Hopmans, and G. J. Kluitenberg, 2006: Multi-functional probe for small-scale simultaneous measurements of soil thermal properties, water content, and electrical conductivity. *Sens. Actuators A*, **132**, 70–77, <https://doi.org/10.1016/j.sna.2006.05.010>.
- van der Sleen, P., and Coauthors, 2015: No growth stimulation of tropical trees by 150 years of CO₂ fertilization but water-use efficiency increased. *Nat. Geosci.*, **8**, 24–28, <https://doi.org/10.1038/ngeo2313>.
- van Dijk, A. I. J. M., L. J. Renzullo, Y. Wada, and P. A. Tregoning, 2014: Global water cycle reanalysis (2003–2012) merging satellite gravimetry and altimetry observations with a hydrological multi-model ensemble. *Hydrol. Earth Syst. Sci.*, **18**, 2955–2973, <https://doi.org/10.5194/hess-18-2955-2014>.
- van Genuchten, M. T., 1980: A closed-form equation for predicting the hydraulic conductivity of unsaturated soils. *Soil. Sci. Soc. Amer. J.*, **44**, 892–898, <https://doi.org/10.2136/sssaj1980.03615995004400050002x>.
- Vautard, R., and Coauthors, 2007: Summertime European heat and drought waves induced by wintertime Mediterranean rainfall deficit. *Geophys. Res. Lett.*, **34**, L07711, <https://doi.org/10.1029/2006GL028001>.
- Velicogna, I., and J. Wahr, 2005: Greenland mass balance from GRACE. *Geophys. Res. Lett.*, **32**, L18505, <https://doi.org/10.1029/2005GL023955>.
- , and —, 2006: Measurements of time-variable gravity show mass loss in Antarctica. *Science*, **311**, 1754–1756, <https://doi.org/10.1126/science.1123785>.
- Vinnikov, K., and I. Yeserkepova, 1991: Soil moisture: Empirical data and model results. *J. Climate*, **4**, 66–79, [https://doi.org/10.1175/1520-0442\(1991\)004<0066:SMEDAM>2.0.CO;2](https://doi.org/10.1175/1520-0442(1991)004<0066:SMEDAM>2.0.CO;2).
- Vinukollu, R. K., E. F. Wood, C. R. Ferguson, and J. B. Fisher, 2011: Global estimates of evapotranspiration for climate studies using multi-sensor remote sensing data: Evaluation of three process-based approaches. *Remote Sens. Environ.*, **115**, 801–823, <https://doi.org/10.1016/j.rse.2010.11.006>.
- Voikov, A. I., 1884: *Klimaty Zemnego Shara, v Osobennosti Rossii (The Earth's Climate with Special Reference to Russia)*. A. Ilyin, 672 pp.
- Vogel, R. M., U. Lall, X. Cai, B. Rajagopalan, P. K. Weiskel, R. P. Hooper, and N. C. Matalas, 2015: Hydrology: The interdisciplinary science of water. *Water Resour. Res.*, **51**, 4409–4430, <https://doi.org/10.1002/2015WR017049>.
- Voisin, N., H. Li, D. Ward, M. Huang, M. Wigmosta, and L. R. Leung, 2013: On an improved sub-regional water resources management representation for integration into earth system models. *Hydrol. Earth Syst. Sci.*, **17**, 3605–3622, <https://doi.org/10.5194/hess-17-3605-2013>.
- Vörösmarty, C. J., P. Green, J. Salisbury, and R. B. Lammers, 2000: Global water resources: Vulnerability from climate change and population growth. *Science*, **289**, 284–288, <https://doi.org/10.1126/science.289.5477.284>.
- , and Coauthors, 2010: Global threats to human water security and river biodiversity. *Nature*, **467**, 555–561, <https://doi.org/10.1038/nature09440>.
- Vose, J. M., G. Sun, C. R. Ford, M. Bredemeier, K. Otsuki, X. Wei, Z. Zhang, and L. Zhang, 2011: Forest ecohydrological research in the 21st century: What are the critical needs? *Ecohydrology*, **4**, 146–158, <https://doi.org/10.1002/eco.193>.
- Vrugt, J. A., H. V. Gupta, B. Nualláin, and W. Bouten, 2006: Real-time data assimilation for operational ensemble streamflow forecasting. *J. Hydrometeorol.*, **7**, 548–565, <https://doi.org/10.1175/JHM504.1>.
- Vuyovich, C. M., J. M. Jacobs, and S. F. Daly, 2014: Comparison of passive microwave and modeled estimates of total watershed SWE in the continental United States. *Water Resour. Res.*, **50**, 9088–9102, <https://doi.org/10.1002/2013WR014734>.
- Wada, Y., L. P. H. Van Beek, and M. F. P. Bierkens, 2012: Non-sustainable groundwater sustaining irrigation: A global assessment. *Water Resour. Res.*, **48**, W00L06, <https://doi.org/10.1029/2011WR010562>.
- Wagener, T., and H. S. Wheatler, 2006: Parameter estimation and regionalization for continuous rainfall-runoff models including uncertainty. *J. Hydrol.*, **320**, 132–154, <https://doi.org/10.1016/j.jhydrol.2005.07.015>.
- Wahr, J., M. Molenaar, and F. Bryan, 1998: Time variability of the Earth's gravity field: Hydrological and oceanic effects and their possible detection using GRACE. *J. Geophys. Res.*, **103**, 30 205–30 229, <https://doi.org/10.1029/98JB02844>.
- Walker, J. P., G. R. Willgoose, and J. D. Kalma, 2004: In situ measurement of soil moisture: a comparison of techniques. *J. Hydrol.*, **293**, 85–99, <https://doi.org/10.1016/j.jhydrol.2004.01.008>.
- Wan, W., and Coauthors, 2017: Hydrological drought in the Anthropocene: Impacts of local water extraction and reservoir regulation in the U.S. *J. Geophys. Res. Atmos.*, **122**, 11 313–11 328, <https://doi.org/10.1002/2017JD026899>.
- Wang, K., and R. E. Dickinson, 2012: A review of global terrestrial evapotranspiration: Observation, modeling, climatology, and climatic variability. *Rev. Geophys.*, **50**, RG2005, <https://doi.org/10.1029/2011RG000373>.
- Wang, L., S. Manzoni, S. Ravi, D. Riveros-Iregui, and K. Caylor, 2015: Dynamic interactions of ecohydrological and biogeochemical processes in water-limited systems. *Ecosphere*, **6**, 133, <https://doi.org/10.1890/ES15-00122.1>.
- Wang, W., H. Lu, L. R. Leung, H. Li, J. Zhao, K. Yang, F. Tian, and K. Sothea, 2017: Dam construction in Lancang-Mekong river basin could mitigate future flood risk from warming-induced intensified rainfall. *Geophys. Res. Lett.*, **44**, 10 378–10 386, <https://doi.org/10.1002/2017GL075037>.
- Waymire, E., V. K. Gupta, and I. Rodriguez-Iturbe, 1984: A spectral theory of rainfall intensity at the meso- β scale. *Water Resour. Res.*, **20**, 1453–1465, <https://doi.org/10.1029/WR020i010p01453>.
- Wei, X., G. Sun, J. M. Vose, K. Otsuki, Z. Zhang, and K. Smetterm, 2011: Forest ecohydrological processes in a changing environment. *Ecohydrology*, **4**, 143–145, <https://doi.org/10.1002/eco.218>.
- Welles, E., and S. Sorooshian, 2009: Scientific verification of deterministic river stage forecasts. *J. Hydrometeorol.*, **10**, 507–520, <https://doi.org/10.1175/2008JHM1022.1>.
- , —, G. Carter, and B. Olsen, 2007: Hydrologic verification: A call for action and collaboration. *Bull. Amer. Meteor. Soc.*, **88**, 503–511, <https://doi.org/10.1175/BAMS-88-4-503>.
- Werner, K., J. S. Verkade, and T. C. Pagano, 2016: Application of hydrological forecast verification information. *Handbook of*

- Hydrometeorological Ensemble Forecasting*, Q. Duan et al., Eds., Springer, 1–21, https://doi.org/10.1007/978-3-642-40457-3_7-1.
- Westerhoff, R. S., M. P. H. Kleuskens, H. C. Winsemius, H. J. Huizinga, G. R. Brakenridge, and C. Bishop, 2013: Automated global water mapping based on wide-swath orbital synthetic-aperture radar. *Hydrol. Earth Syst. Sci.*, **17**, 651–663, <https://doi.org/10.5194/hess-17-651-2013>.
- Westrick, K. J., P. Storck, and C. F. Mass, 2002: Description and evaluation of a hydrometeorological forecast system for mountainous watersheds. *Wea. Forecasting*, **17**, 250–262, [https://doi.org/10.1175/1520-0434\(2002\)017<0250:DAEOAH>2.0.CO;2](https://doi.org/10.1175/1520-0434(2002)017<0250:DAEOAH>2.0.CO;2).
- Wetterhall, F., and F. Di Giuseppe, 2018: The benefit of seamless forecasts for hydrological predictions over Europe. *Hydrol. Earth Syst. Sci.*, **22**, 3409–3420, <https://doi.org/10.5194/hess-22-3409-2018>.
- Wheater, H. S., and P. Gober, 2015: Water security and the science agenda. *Water Resour. Res.*, **51**, 5406–5424, <https://doi.org/10.1002/2015WR016892>.
- Wigmosta, M. S., L. W. Vail, and D. P. Lettenmaier, 1994: A distributed hydrology-vegetation model for complex terrain. *Water Resour. Res.*, **30**, 1665–1679, <https://doi.org/10.1029/94WR00436>.
- Wigneron, J. P., J. C. Calvet, T. Pellarin, A. A. Van De Griend, M. Berger, and P. Ferrazzoli, 2003: Retrieving near-surface soil moisture from microwave radiometric observations: Current status and future plans. *Remote Sens. Environ.*, **85**, 489–506, [https://doi.org/10.1016/S0034-4257\(03\)00051-8](https://doi.org/10.1016/S0034-4257(03)00051-8).
- Wilcox, B. P., 2010: Transformative ecosystem change and ecohydrology: Ushering in a new era for watershed management. *Ecohydrology*, **3**, 126–130, <https://doi.org/10.1002/eco.104>.
- Williams, I. N., M. S. Torn, W. J. Riley, and M. F. Wehner, 2014: Impacts of climate extremes on gross primary production under global warming. *Environ. Res. Lett.*, **9**, 094011, <https://doi.org/10.1088/1748-9326/9/9/094011>.
- Wilson, K. B., P. J. Hanson, P. J. Mulholland, D. D. Baldocchi, and S. D. Wullschlegel, 2001: A comparison of methods for determining forest evapotranspiration and its components: Sapflow, soil water budget, eddy covariance and catchment water balance. *Agric. For. Meteorol.*, **106**, 153–168, [https://doi.org/10.1016/S0168-1923\(00\)00199-4](https://doi.org/10.1016/S0168-1923(00)00199-4).
- WMO, 2009: Manual on Estimation of Probable Maximum Precipitation (PMP). 2nd ed. Operational Hydrology Rep. 1, WMO-332, 259 pp. http://library.wmo.int/pmb_ged/wmo_332.pdf.
- Wolanski, E., L. A. Boorman, L. Chícharo, E. Langlois-Saliou, R. Lara, A. J. Plater, R. J. Uncles, and M. Zalewski, 2004: Ecohydrology as a new tool for sustainable management of estuaries and coastal waters. *Wetlands Ecol. Manage.*, **12**, 235–276, <https://doi.org/10.1007/s11273-005-4752-4>.
- Wood, A. W., and D. P. Lettenmaier, 2006: A test bed for new seasonal hydrologic forecasting approaches in the western United States. *Bull. Amer. Meteor. Soc.*, **87**, 1699–1712, <https://doi.org/10.1175/BAMS-87-12-1699>.
- , and —, 2008: An ensemble approach for attribution of hydrologic prediction uncertainty. *Geophys. Res. Lett.*, **35**, L14401, <https://doi.org/10.1029/2008GL034648>.
- , E. P. Maurer, A. Kumar, and D. P. Lettenmaier, 2002: Long-range experimental hydrologic forecasting for the eastern United States. *J. Geophys. Res.*, **107**, 4429, <https://doi.org/10.1029/2001JD000659>.
- , T. Hopson, A. Newman, L. Brekke, J. Arnold, and M. Clark, 2016a: Quantifying streamflow forecast skill elasticity to initial condition and climate prediction skill. *J. Hydrometeorol.*, **17**, 651–668, <https://doi.org/10.1175/JHM-D-14-0213.1>.
- , T. Pagano, and M. Roos, 2016b: Tracing the Origins of ESP. HEPEX Historical Hydrology Series, Edition 1, HEPEX: A Global Community in Hydrological Ensemble Prediction, accessed 30 April 2018, <https://hexex.irstea.fr/tracing-the-origins-of-esp/>.
- Wood, E. F., M. Sivapalan, K. Beven, and L. Band, 1988: Effects of spatial variability and scale with implications to hydrologic modeling. *J. Hydrol.*, **102**, 29–47, [https://doi.org/10.1016/0022-1694\(88\)90090-X](https://doi.org/10.1016/0022-1694(88)90090-X).
- , and Coauthors, 2011: Hyperresolution global land surface modeling: Meeting a grand challenge for monitoring Earth's terrestrial water. *Water Resour. Res.*, **47**, W05301, <https://doi.org/10.1029/2010WR010090>.
- Woolhiser, D. A., and J. A. Liggett, 1967: Unsteady, one-dimensional flow over a plane—The rising hydrograph. *Water Resour. Res.*, **3**, 753–771, <https://doi.org/10.1029/WR003i003p00753>.
- Wu, H., R. F. Adler, Y. Hong, Y. Tian, and F. Policelli, 2012: Evaluation of global flood detection using satellite-based rainfall and a hydrologic model. *J. Hydrometeorol.*, **13**, 1268–1284, <https://doi.org/10.1175/JHM-D-11-087.1>.
- , —, Y. Tian, G. J. Huffman, H. Li, and J. Wang, 2014: Real-time global flood estimation using satellite-based precipitation and a coupled land surface and routing model. *Water Resour. Res.*, **50**, 2693–2717, <https://doi.org/10.1002/2013WR014710>.
- Wulfmeyer, V., and Coauthors, 2018: A new research approach for observing and characterizing land–atmosphere feedback. *Bull. Amer. Meteor. Soc.*, **99**, 1639–1667, <https://doi.org/10.1175/BAMS-D-17-0009.1>.
- Xie, P., and P. A. Arkin, 1997: Global precipitation: A 17-year monthly analysis based on gauge observations, satellite estimates, and numerical model outputs. *Bull. Amer. Meteor. Soc.*, **78**, 2539–2558, [https://doi.org/10.1175/1520-0477\(1997\)078<2539:GPAYMA>2.0.CO;2](https://doi.org/10.1175/1520-0477(1997)078<2539:GPAYMA>2.0.CO;2).
- , J. E. Janowiak, P. A. Arkin, R. Adler, A. Gruber, R. Ferraro, G. J. Huffman, and S. Curtis, 2003: GPCP pentad precipitation analyses: An experimental dataset based on gauge observations and satellite estimates. *J. Climate*, **16**, 2197–2214, <https://doi.org/10.1175/2769.1>.
- Yang, Z., F. Dominguez, X. Zeng, H. Hu, H. Gupta, and B. Yang, 2017: Impact of irrigation over the California Central Valley on regional climate. *J. Hydrometeorol.*, **18**, 1341–1357, <https://doi.org/10.1175/JHM-D-16-0158.1>.
- Yeh, P. J. F., M. Irizarry, and E. A. Eltahir, 1998: Hydroclimatology of Illinois: A comparison of monthly evaporation estimates based on atmospheric water balance and soil water balance. *J. Geophys. Res.*, **103**, 19 823–19 837, <https://doi.org/10.1029/98JD01721>.
- Yeh, W. W.-G., 1985: Reservoir management and operations models: A state-of-the-art review. *Water Resour. Res.*, **21**, 1797–1818, <https://doi.org/10.1029/WR021i012p01797>.
- , 1986: Review of parameter identification procedures in groundwater hydrology: The inverse problem. *Water Resour. Res.*, **22**, 95–108, <https://doi.org/10.1029/WR022i002p00095>.
- Yizgaw, W., F. Hossain, and A. Kalyanapu, 2013: Comparison of PMP-driven probable maximum floods with flood magnitudes due to increasingly urbanized catchment: The case of American River Watershed. *Earth Interact.*, **17**, <https://doi.org/10.1175/2012EI000497.1>.
- Yoder, C. F., J. G. Williams, J. O. Dickey, B. E. Schutz, R. J. Eanes, and B. D. Tapley, 1983: Secular variation of Earth's gravitational harmonic J₂ coefficient from Lageos and nontidal acceleration of Earth rotation. *Nature*, **303**, 757–762, <https://doi.org/10.1038/303757a0>.

- Yoon, J.-H., and L. R. Leung, 2015: Assessing the relative influence of soil moisture and ENSO SST on precipitation predictability over the contiguous United States. *Geophys. Res. Lett.*, **42**, 5005–5013, <https://doi.org/10.1002/2015GL064139>.
- Yuan, X., Z. Xie, J. Zheng, X. Tian, and Z. Yang, 2008: Effects of water table dynamics on regional climate: A case study over East Asian monsoon area. *J. Geophys. Res.*, **113**, D21112, <https://doi.org/10.1029/2008JD010180>.
- Zalewski, M., 2000: Ecohydrology—The scientific background to use ecosystem properties as management tools toward sustainability of water resources. *Ecol. Eng.*, **16**, 1–8, [https://doi.org/10.1016/S0925-8574\(00\)00071-9](https://doi.org/10.1016/S0925-8574(00)00071-9).
- Zarfl, C., A. E. Lumsdon, J. Berlekamp, L. Tydecks, and K. Tockner, 2015: A global boom in hydropower dam construction. *Aquat. Sci.*, **77**, 161–170, <https://doi.org/10.1007/s00027-014-0377-0>.
- Zawadzki, I., 1987: Fractal structure and exponential decorrelation in rain. *J. Geophys. Res.*, **92**, 9586–9590, <https://doi.org/10.1029/JD092iD08p09586>.
- Zeng, X., and M. Decker, 2009: Improving the numerical solution of soil moisture-based Richards equation for land models with a deep or shallow water table. *J. Hydrometeorol.*, **10**, 308–319, <https://doi.org/10.1175/2008JHM1011.1>.
- Zhang, S., Y. Xie, F. Counillon, X. Ma, P. Yu, and Z. Jing, 2018: Regional coupled model and data assimilation. *Adv. Meteor.*, **2018**, 9434102, <https://doi.org/10.1155/2018/9434102>.
- Zhang, X., H.-Y. Li, Z. D. Deng, L. R. Leung, J. R. Skalski, and S. J. Cooke, 2019: On the variable effects of climate change on Pacific salmon. *Ecol. Modell.*, **397**, 95–106, <https://doi.org/10.1016/j.ecolmodel.2019.02.002>.
- Zoch, R. T., 1934: On the relation between rainfall and stream flow. *Mon. Wea. Rev.*, **62**, 315–322, [https://doi.org/10.1175/1520-0493\(1934\)62<315:OTRBR>2.0.CO;2](https://doi.org/10.1175/1520-0493(1934)62<315:OTRBR>2.0.CO;2).
- Zreda, M., D. Desilets, T. P. A. Ferré, and R. L. Scott, 2008: Measuring soil moisture content non-invasively at intermediate spatial scale using cosmic-ray neutrons. *Geophys. Res. Lett.*, **35**, L21402, <https://doi.org/10.1029/2008GL035655>.