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Climate Change Science Program

Uses and Limitations of Observations, Data, Forecasts, and Other Projections in Decision Support for Selected Sectors and Regions

Chapter 5. “Decision Support for Water Resources Management”

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1. Introduction

Water resource managers have long been incorporating information related to climate in their decisions. The tremendous, regionally ubiquitous, investments in infrastructure to reduce flooding (e.g., levees and reservoirs) or assure reliable water supplies (e.g., reservoirs, groundwater development, irrigation systems, water allocation, and transfer agreements) reflect societal goals to mitigate the impacts of climate variability at multiple time and space scales. As the financial, political, social, and environmental costs of infrastructure options have become less tractable, water management institutions have undergone comprehensive reform, shifting their focus to optimizing operations of existing projects and managing increasingly diverse, and often conflicting, demands on the services provided by water resources (Bureau of Reclamation [BOR], 1992; Beard, 1993; Congressional Budget Office, 1997; Stakhiv, 2003; National Research Council [NRC], 2004). Governments have also made substantial investments to improve climate information and understanding over the past decades through satellites, *in situ* measuring networks, supercomputers, and research programs. National and international programs have explicitly identified as an important objective ensuring that improved data products, conceptual models, and predictions are useful to the water resources management community (Endreny et al., 2003; Lawford et al., 2005). Although exact accounting is difficult, potential values associated with appropriate use of accurate hydrometeorologic predictions generally range from the millions to the billions of dollars (e.g., National Hydrologic Warning Council, 2002). There are also non-monetary values associated with more efficient, equitable, and environmentally sustainable decisions related to water resources.

1993 Droughts, floods, and increasing demands on available water supplies continue to create concern, and even
1994 crises, for water resources management. Many communities have faced multiple hydrologic events that were earlier
1995 thought to have low probabilities of occurrence (e.g., NRC, 1995), and long-term shifts in streamflows have been
1996 observed (Lettenmaier et al., 1994; Lins and Slack, 1999; Douglas et al., 2000), leading to questions about the relative
1997 impacts of shifts in river hydraulics, land use, and climate conditions.

1998 Until the last two decades, climate was viewed largely as a collection of random processes, and this paradigm
1999 informed much of the water resource management practices developed over the past 50 years that persist today.
2000 However, climate is now recognized as a chaotic process, shifting among distinct regimes with statistically significant
2001 differences in average conditions and variability (Hansen et al., 1997). As instrumental records have grown longer and
2002 extremely long time-series of paleoclimatological indicators have been developed (Ekwurzal, 2005), they increasingly
2003 belie one of the fundamental assumptions behind most extant water resources management—stationarity. Stationary
2004 time series have time-invariant statistical characteristics (e.g., mean or variance), meaning that different parts of the
2005 historical record can be considered equally likely. Within the limits posed by sampling, statistics computed from
2006 stationary time series can be used to define a probability distribution that will also then faithfully represent expectations
2007 for the future (Salas, 1993).

2008
2009 Further, prospects for climate change due to global warming have moved from the realm of speculation to
2010 general acceptance (Intergovernmental Panel on Climate Change [IPCC] 1990, 1995a, 2001a, 2007). The potential
2011 impacts of climate on water resources, and their implications for management, have been central topics of concern in
2012 climate change assessments (e.g., EPA, 1989; IPCC, 1995b, 2001b; National Assessment Synthesis Team, 2000; Gleick
2013 and Adams, 2000; Barnett *et al.*, 2004). These studies are becoming increasingly confident in their conclusions that the
2014 future portends statistically significant changes in hydroclimatic averages and variability.

2015 There has been persistent and broad disappointment in the extent to which improvements in hydroclimatic
2016 science from large-scale research programs have affected resource management practices in general (Pielke, 1995,
2017 2001; NRC, 1998a, 1999a) and water resource management in particular (NRC, 1998b, 1999b,c). For example,
2018 seasonal climate outlooks have been slow to be entered into the water management decision processes, even though
2019 they have improved greatly over the past 20 years (Hartmann et al., 2002a, 2003). Water managers have been even more
2020 resistant to incorporating notions of hydrologic non-stationarity in general and climate change in particular in decision

2021 processes. Until recently, hydrologic analysis techniques have been seen as generally sufficient (e.g., Matalas, 1997;
2022 Lins and Stakhiv, 1998), especially in the context of slow policy and institutional evolution (Stakhiv, 2003). However,
2023 an inescapable message for the water resource management community is the inappropriateness of the stationarity
2024 assumption in the face of climate change.

2025 Several ongoing efforts are leading the way forward to establish more effective ways of incorporating climate
2026 understanding and earth observations into water resources management (Pulwarty, 2002; Office of Global Programs,
2027 2004; NASA, 2005). While diverse in their details, these efforts seek to link hydroclimatological variability, analytical
2028 and predictive technologies, and water management decisions within an end-to-end context extending from
2029 observational data through large-scale analyses and predictions, uncertainty evaluation, impacts assessment,
2030 applications, and evaluations of applications (e.g., Young, 1995; Miles et al., 2000). Some end-to-end efforts focus on
2031 cultivating information and management networks; designing processes for recurrent interaction among research,
2032 operational product generation, management, and constituent communities; and developing adaptive strategies for
2033 accommodating climate variability, uncertainty, and change. Other end-to-end efforts focus on the development of
2034 decision support tools (DST) that embody unique resource management circumstances to enable formal and more
2035 objective linkages between meteorological, hydrologic, and institutional processes. Typically, end-to-end DST
2036 applications are developed for organizations making decisions with high-impact (e.g., state or national agencies) or
2037 high-economic value (e.g., hydropower production) and that possess the technical and managerial abilities to efficiently
2038 exploit research advances (e.g., Georgakakos et al, 1998, 2004, 2005; Georgakakos, 2006). If linked to socioeconomic
2039 models incorporating detailed information about the choices open to decision-makers and their tolerance for risk, these
2040 end-to-end tools could also enable explicit assessment of the impacts of scientific and technological research advances.

2041 This chapter describes a river management DST, RiverWare, which facilitates coordinated efforts among the
2042 research, operational product generation, and water management communities. RiverWare emerged from an early and
2043 sustained effort by several federal agencies to develop generic tools to support the assessment of water resources
2044 management options in river basins with multiple reservoirs and multiple management objectives (Frevert et al., 2006).
2045 RiverWare was selected for use as a case study because it has been used in a variety of settings, by multiple agencies,
2046 over a longer period than many other water management DSTs. Furthermore, RiverWare can explicitly accommodate a
2047 broad range of resource management concerns (e.g., flood control, recreation, navigation, water supply, water quality,
2048 and power production). RiverWare can also consider perspectives ranging from day-to-day scheduling of operations to

2049 long-range planning and can accommodate a variety of climate observations, forecasts, and even climate change
2050 projections. RiverWare can incorporate hydrologic risk, whereby event consequences and their magnitudes are mediated
2051 by their probability of occurrence, in strategic planning applications and design studies, which can offer a way forward
2052 for decision makers reluctant to shift away from use of traditional, stationarity-based, statistical analysis of historical
2053 data (Lee, 1999; Davis and Pangburn, 1999).

2054

2055 **2. Description of RiverWare**

2056 RiverWare is a software framework used to develop detailed models of how water moves and is managed
2057 throughout complex river basin systems. RiverWare applications include physical processes (e.g., streamflow, bank
2058 storage, and solute transport), infrastructure (e.g., reservoirs, hydropower generating turbines, spillways, and diversion
2059 connections), and policies (e.g., minimum instream flow requirements and trades between water users) (Zagona et al.,
2060 2001, 2005). At a minimum, RiverWare applications require streamflow hydrographs as input for multiple locations
2061 throughout a river system. While hydrographs can be generated within the DST, they can also be input from other
2062 sources, with the latter approach being especially important in advanced end-to-end assessments. Detailed discussion of
2063 the role of observations and considerations of global change using RiverWare are discussed in later sections. RiverWare
2064 can be applied to address diverse water management concerns, including real-time operations, strategic planning for
2065 seasonal to interannual variability in water supplies and demands, and examining impacts of hydrologic non-
2066 stationarity. Because infrastructure, management rules, and policies can be easily changed, RiverWare also allows
2067 examination of alternative options for achieving management objectives over short-, medium-, and long-term planning
2068 horizons.

2069 RiverWare was developed by the University of Colorado-Boulder's Center for Advanced Decision Support for
2070 Water and Environmental Systems (CADSWES) in collaboration with the BOR, Tennessee Valley Authority, and the
2071 Army Corps of Engineers (Frevert et al., 2006). CADSWES continues to develop and maintain the RiverWare software,
2072 as well as offer training and support for RiverWare users (see <http://cadswes.colorado.edu>). According to CADSWES,
2073 RiverWare is used by more than 75 federal and state agencies, private sector consultants, universities and research
2074 institutes, and water districts, among others.

2075

2076 *Example Applications*

2077 Consistent with the intent of its original design, the use of RiverWare varies widely, depending on the specific
2078 application. An early application was its use for scheduling reservoir operations by the Tennessee Valley Authority
2079 (Eshenbach et al., 2001). In that application, RiverWare was used to define the physical and economic characteristics of
2080 the multi-reservoir system, including power production economics, to prioritize the policy goals that governed the
2081 reservoir operations and to specify parameters for linear optimization of system objectives. In another application,
2082 RiverWare was used to balance the competing priorities of minimum instream flows and consumptive water use in the
2083 operation of the Flaming Gorge Reservoir in Colorado (Wheeler et al., 2002).

2084 While day-to-day scheduling of reservoir operations is more a function of weather than climate, the use of
2085 seasonal climate forecasts to optimize reservoir operations has long been a goal for water resources management.
2086 RiverWare is being implemented for the Truckee-Carson River basin in Nevada to investigate the impact of
2087 incorporating climate outlooks into an operational water management framework that prioritizes irrigation water
2088 supplies, interbasin diversions, and fish habitat (Grantz et al., 2007). Another example application to the Truckee-
2089 Carson River using a hypothetical operating policy indicated that fish populations could benefit from purchases of water
2090 rights for reservoir releases to mitigate warm summer stream temperatures resulting from low flows and high air
2091 temperatures (Neumann et al., 2006).

2092 RiverWare has also been used to evaluate politically charged management strategies, including water transfers
2093 proposed in California's Quantification Settlement Agreement and the BOR's Inadvertant Overrun Policy, maintaining
2094 instream flows sufficient to restore biodiversity in the Colorado River delta, and conserving riparian habitat while
2095 accommodating future water and power development in the BOR Multiple Species Conservation Program (Wheeler et
2096 al., 2002). RiverWare also played a key role in negotiations by seven western states concerning how the Colorado River
2097 should be managed and the river flow should be distributed among the states during times of drought. The BOR
2098 implemented a special version of the RiverWare model of the Colorado River and its many reservoirs, diversions, and
2099 watersheds (Jerla, 2005). The model was used to provide support to the Basin States Modeling Work Group Committee
2100 over an 18-month period, as they assessed different operational strategies under different hydrologic scenarios,
2101 including extreme drought (U.S. Department of Interior, 2007).

2102

2103 *Implementation*

2104 RiverWare requirements are multi-dimensional. A specific river system and its infrastructure operating policies
2105 are defined by data files supplied to RiverWare. This allows incorporation of new basin features (e.g., reservoirs),
2106 operating policies, and hydroclimatic conditions without users having to write software code. Utilities within RiverWare
2107 enable users to automatically execute many simulations, including accessing external data or exporting results of model
2108 runs. Users can also write new modules that CADSWES can integrate into RiverWare for use in other applications. For
2109 example, in an application for the Pecos River in New Mexico, engineers developed new methods and software code for
2110 realistic downstream routing of summer monsoon-related flood waves (Boroughs and Zagona, 2002). RiverWare is
2111 implemented for use on Windows or Unix Solaris systems, as described in the requirements document
2112 (<http://cadswes.colorado.edu/PDF/RiverWare/RecommendedMinimumSystemsRequirements.pdf>). An extensive
2113 manual is also available (<http://cadswes.colorado.edu/PDF/ReleaseNotes/RiverWareHelp.pdf>).

2114 RiverWare applications can be implemented by any group that can pay for access, both in terms of finances
2115 and educational effort. Development of RiverWare applications requires a site license from CADSWES. Significant
2116 investment is required to learn to use RiverWare as well. CADSWES offers two 3-day RiverWare training courses, an
2117 initial class covering general simulation modeling, managing scenarios, and incorporating policy options through rule-
2118 based simulation, and a second class covering rule-based simulation in more detail, creating basin policies, and
2119 examining water policy options. Costs for the original license, annual renewals, technical support, and training require
2120 several thousand dollars. The costs of licensing and learning RiverWare mean that small communities and civic groups
2121 are unlikely to implement their own applications for assessing water management options. Rather, large agencies with
2122 technical staff or the financial means to fund university research or consultants are the most frequent users of
2123 RiverWare. The agencies then mediate the access of stakeholders to assessments of water management options through
2124 traditional public processes (e.g., U.S. Department of Interior, 2007). Conflicts may arise in having academic research
2125 groups conduct analyses funded by stakeholder groups, with inherent tensions between the open publication of research
2126 required by academia and the limited access to results required by strategic negotiations among interest groups.

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2128 **3. Current and Future Use of Observations**

2129 The specific combination of observations used by a RiverWare application depends on both the decision
2130 context and the use of other models and DSTs to provide input to RiverWare that more comprehensively or accurately
2131 describes the character, conditions, and response of the river basin system. Figure 1 illustrates the information flow

2132 linking observations, RiverWare, other models and DSTs, and water management decisions; it shows that RiverWare
2133 has tremendous flexibility in the kinds of observations that could be useful in hydrologic modeling and river system
2134 assessment and management. The types of observations that may ultimately feed into RiverWare applications also
2135 depend on the timescale of the situation.

2136 A detailed discussion of the role of satellite observations in RiverWare applications and selected input models
2137 and DSTs (e.g., the BOR's ET Toolbox and Precipitation Runoff Modeling System [PRMS]) is given by the
2138 "Evaluation Report for AWARDS ET Toolbox and RiverWare Decision Support Tools" (Hydrological Sciences
2139 Branch, 2007). Briefly, RiverWare can use a combination of observations from multiple sources, including satellites,
2140 products derived from land-atmosphere or hydrologic models, and combinations of both. Satellite observations can
2141 assist models in estimating evapotranspiration, precipitation, snow water equivalent, soil moisture, groundwater storage
2142 and aquifer volumes, reservoir storage, and water quality, among other variables. Measurements from sensors aboard a
2143 variety of satellites are being considered for their usefulness within DST contexts and their impacts on reducing water
2144 management uncertainty, including the MODIS (Moderate Resolution Imaging Spectroradiometer) sensor aboard the
2145 Earth Observing System (EOS) Terra and Aqua satellites, Landsat TM data, Advanced Spaceborne Thermal Emission
2146 and Reflection Radiometer (ASTER), Shuttle Radar Topography Mission (SRTM), Advanced Microwave Scanning
2147 Radiometer-EOS (AMSR-E), Gravity Recovery and Climate Experiment (GRACE), and Tropical Rainfall Mapping
2148 Mission (TRMM), among others. Future and planned satellites with hydrologically relevant sensors and measurements
2149 include CloudSAT, the Global Precipitation Mission (GPM), and the National Polar-Orbiting Operational
2150 Environmental Satellite (NPOESS). Use of these observations can be enhanced by assimilating them into land surface
2151 models to produce spatially-distributed estimates of snowpack, soil moisture, evapotranspiration, energy fluxes, and
2152 runoff, which then provide inputs to RiverWare to base a more comprehensive assessment of river basin conditions. The
2153 land surface models include the Community Land Model (CLM), Mosaic, Noah, and VIC, among others, supported by
2154 NASA's Land Data Assimilation System (LDAS) and Land Information System (LIS) (NASA, 2006a).

2155 NASA has several pilot projects specifically focused on assessing the impact of satellite observations in a
2156 variety of hydrologic models and DSTs as they feed into RiverWare applications (NASA, 2005, 2006b, 2007). For
2157 example, one project is comparing Terra and Aqua MODIS snow cover products for the Yakima-Columbia River basins
2158 with land-based snow telemetry measurements, testing their use for LIS simulations that also use the North American
2159 LDAS, connecting assimilated snow data with the Modular Modeling System (MMS) Precipitation-Runoff Modeling

2160 System (PRMS), and then supplying the simulated runoff as inputs to RiverWare. Another project on the Rio Grande
2161 River basin is assessing MODIS and Landsat data to improve evapotranspiration estimates generated by the BOR DST,
2162 the Agricultural Water Resources Decision Support (AWARDS) ET Toolbox, which then provides water demand time
2163 series to RiverWare. While application of specific hydrologic models and observations depend on the specific
2164 RiverWare application, significant processing of both model and observations are required and can be resource
2165 intensive (e.g., calibration and aggregation/disaggregation).

2166 Operational scheduling of reservoir releases depend on orders of water from downstream users (e.g., irrigation
2167 districts) that are largely affected by day-to-day weather conditions as well as seasonally varying demands. In these
2168 cases, the important observations are the near real-time estimates of conditions within the river basin system (e.g., soil
2169 moisture or infiltration capacity), which affect the transformation of precipitation into runoff into the river system,
2170 relative to constraints on system operation (e.g., reservoir storage levels or water temperatures at specific river
2171 locations). Meteorological prospects are mediated by those placing the water orders or through short-term weather
2172 forecasts that may affect operations when the system is near some constraint (e.g., flood flows when reservoir levels are
2173 near peak storage capacity). In these situations, the important observations are recent extreme precipitation events and
2174 their location, which may be provided, separately or in some combination, by *in situ* monitoring networks, radar, or
2175 satellites.

2176 For mid-range applications, such as strategic planning for operations over the next season or year, outlooks of
2177 total seasonal water supplies are routinely used in making commitments for water deliveries, determining industrial and
2178 agricultural water allocation, and carrying out reservoir operations. In these applications, it is also important for water
2179 managers to keep track of the current state of the watershed. Such observations are often used as input to one of the
2180 many independent hydrologic models that can provide input to a specific RiverWare application. In these situations, the
2181 important observations are those that provide boundary or forcing conditions for the independent hydrologic models,
2182 including snowpack moisture storage, soil moisture, precipitation (intensity, duration, and spatial distribution), air
2183 temperature, humidity, winds, and other meteorological conditions.

2184 For long-term planning and design applications, observations are less important because the effects of recent
2185 conditions have less impact on long-term outcomes than future meteorological uncertainty, or even institutions at multi-
2186 decadal time scales. In these applications, accurate representation of anticipated natural hydroclimatological variability
2187 is important. In many western U.S. applications, observed streamflows are adjusted to remove the effects of reservoir

2188 management, interbasin diversions, and water withdrawals. The adjusted flows, termed “naturalized flows.” may be
2189 used as input to RiverWare applications to assess the impact of different management options. Use of naturalized flows
2190 is fraught with problems. A central issue is poor monitoring of actual human impacts, especially withdrawals,
2191 diversions, and return flows (e.g., from irrigation). Alternative approaches include the use of proxy streamflows (e.g.,
2192 from paleoclimatological indicators) or output from hydrologic modeling studies (Hartmann, 2005). For example,
2193 Tarboton (1995) developed hydrologic scenarios for severe sustained drought in the Colorado River basin based on
2194 streamflows reconstructed from centuries of tree-ring records; the scenarios were used in an assessment of management
2195 options using a precursor to the current RiverWare application to the Colorado River system.

2196 The usefulness of the observations used within RiverWare depends on the specific implementation, as well as
2197 the quality of the information itself. For example, one direct use of climate information for long-term planning includes
2198 hydrologic and hydraulic routing of “design storms” of various magnitudes and likelihoods, with the storms based on
2199 analyses of the available instrumental record (Urbanas and Roesner, 1993). However, those instrumental records have
2200 often been too short to adequately express climate variability and resulting impacts, regardless of the specific DSTs
2201 used to do the hydrologic or hydraulic routing. In short- and mid-range forecasting applications, the use of observations
2202 is mediated by the hydrologic model or DST that transforms weather and climate into streamflows, evaporative water
2203 demands, and other hydrologic processes. In these situations, from an operational perspective, the stream of
2204 observational inputs must be dependable, without downtime or large data gaps, and data processing, model simulation,
2205 and creation of forecast products must be fast and efficient. The usefulness of observations may be limited by other
2206 issues as well. The water resources management milieu is complex and diverse, and climate influences are only one
2207 factor among many affecting water management policies and practices. Factors limiting the use of observations or
2208 subsequent hydrologic model input to RiverWare for actual water management include lack of familiarity with the
2209 available information, disconnects between the specific information available (e.g., variables and spatiotemporal scales)
2210 and their relevance to decision makers, skepticism about the quality and applicability of information, conservative
2211 decision preferences due to accountability for poor consequences, and institutional impediments such as the inflexible
2212 nature of many multi-jurisdictional water management agreements (Changnon, 1990; Kenney, 1995; Pulwarty and
2213 Redmond, 1997; Pagano et al., 2001, 2002; Jacobs, 2002; Jacobs and Pulwarty, 2003; Rayner et al., 2005).

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2216 **4. Uncertainty**

2217 The reliability of observations for driving hydrologic models that may provide input to RiverWare applications
2218 is the subject of much ongoing research. The hydrologic models, because they incompletely describe the physical
2219 relationships among important watershed components (e.g., vegetation processes that link the atmosphere and different
2220 levels of soil and surface and groundwater interactions), are themselves the subject of much research to determine their
2221 reliability. Streamflow and other hydrologic variables are intimately responsive to atmospheric factors, especially
2222 precipitation, that drive a watershed's behavior; however, errors in precipitation estimates are often amplified in the
2223 hydrologic response (Oudin et al., 2006).

2224 Obtaining quality precipitation estimates is a formidable challenge, especially in the western U.S. where
2225 orographic effects produce large spatial variability and where there is a scarcity of real-time precipitation gage data and
2226 radar beam blockage by mountains. In principal, outputs from atmospheric models can serve as surrogates for
2227 observations, as well as providing forecasts of meteorological variables that can be used to drive hydrologic models.
2228 One issue in integrating atmospheric model output into hydrologic models for small watersheds (<1000 km²) is that the
2229 spatial resolution of atmospheric models is lower than the resolution of hydrologic models. For example, quantitative
2230 precipitation forecasts (QPF) produced by some atmospheric models may cover several thousand square kilometers, but
2231 the hydrologic models used for predicting daily streamflows require precipitation to be downscaled to precipitation
2232 fields for watersheds covering only tens or hundreds of square kilometers. One approach to produce output consistent
2233 with the needs of hydrologic models is to use nested atmospheric models, whereby outputs from large scale but coarse
2234 resolution models are used as boundary conditions for models operating over smaller domains with higher resolution.
2235 However, the error characteristics of atmospheric model products (e.g., bias in precipitation and air temperature) also
2236 can have significant effects on subsequent streamflow forecasts. Bias corrections require knowledge of the
2237 climatologies (i.e., long-term distributions) of both modeled and observed variables.

2238 Although meteorological uncertainty may be high for the periods addressed by streamflow forecasts, accurate
2239 estimates of the state of watershed conditions prior to the forecast period are important because they are used to
2240 initialize hydrologic model states, with significant consequences for forecast results. However, watershed conditions
2241 can be difficult to measure, especially when streamflow forecasts must be made quickly, as in the case of flash flood
2242 forecasts. One option is to continuously update watershed states by running the hydrologic models continuously and by
2243 using inputs from recent meteorological observations and/or atmospheric models. Regardless of the source of inputs,

2244 Westrick et al. (2002) found it essential to obtain observational estimates of initial conditions to keep streamflow
2245 forecasts realistic; storm-by-storm corrections of model biases determined over extended simulation periods were
2246 insufficient. Recent experimental end-to-end forecasts of streamflow produced in a simulated operational setting (Wood
2247 et al., 2001) highlighted the critical role of quality estimates of spring and summer soil moisture used to initialize
2248 hydrologic model states for the eastern U.S.

2249 Where streamflows may be largely comprised of snowmelt runoff, quality estimates of snow conditions are
2250 important. The importance of reducing errors in the timing and magnitude of snowmelt runoff are especially acute in
2251 regions where a large percentage of annual water supplies derive from snowmelt runoff, snowmelt impacts are highly
2252 non-linear with increasing deviation from long-term average supplies, and reservoir storage is smaller than interannual
2253 variation of water supplies. However, resources for on-site monitoring of snow conditions have diminished rather than
2254 grown, relative to the increasing costs of errors in hydrologic forecasts (Davis and Pangburn, 1999). Research activities
2255 of the NWS National Office of Hydrology Remote Sensing Center (NOHRSC) have long been directed at improving
2256 estimates of snowpack conditions through aerial and satellite remote sensing (Carroll, 1985). However, the cost of aerial
2257 flights prohibits routine use (T. Carroll, NOHRSC, personal communication, 1999), while satellite estimates have
2258 qualitative limitations (e.g., not considering fractional snow coverage over large regions) and have not found broad use
2259 operationally.

2260 Multiple techniques exist to more accurately represent the uncertainty inherent in understanding and predicting
2261 potential hydroclimatic variability. Stochastic hydrology techniques use various forms of autoregressive models to
2262 generate multiple synthetic streamflow time series with statistical characteristics matching available observations. For
2263 example, in estimating the risk of low flows for the Sacramento River Basin in California, the BOR (Frevert et al.,
2264 1989) generated 20 one-thousand-year streamflow time series matching selected statistics of observed flows (adjusted to
2265 compensate for water management impacts on natural flows); the non-exceedance probabilities of low flows were
2266 computed by counting the occurrences of low flows within 1- through 10-year intervals for all 20 one-thousand-year
2267 sequences. The U.S. Army Corps of Engineers (1992) used a similar approach to estimate flood magnitudes with return
2268 periods exceeding 1,000 years, using Monte Carlo sampling from within the 95% confidence limits of a Log Pearson III
2269 distribution developed by synthesizing multiple streamflow time series.

2270 The ability to automatically execute many model runs within RiverWare, including accessing data from
2271 external sources and exporting model results, facilitates using stochastic hydrology approaches for representing

2272 uncertainty. For example, Carron et al. (2006) demonstrated RiverWare's capability to identify and quantify significant
2273 sources of uncertainty in projecting river and reservoir conditions, using a first-order, second-moment (FOSM)
2274 algorithm that is computationally more efficient than more traditional Monte Carlo approaches. The FOSM processes
2275 uncertainties in inputs and models to provide estimates of uncertainty in model results that can be used directly within a
2276 risk management decision framework. The case study presented by Carron et al. (2006) evaluated the uncertainties
2277 associated with meeting goals for reservoir water levels beneficial for recovering endangered fish species within the
2278 lower Colorado River.

2279 With regard to RiverWare applications concerned with mid-range planning and use of hydrologic forecasts, at
2280 the core of any forecasting system is the predictive model, whether a simple statistical relationship or a complex
2281 dynamic numerical model. Advances in hydrologic modeling have been notable, especially those associated with the
2282 proper identification of a model's parameters (e.g., Duan et al., 2002) and the development of models that consider the
2283 spatially distributed characteristics of watersheds, rather than treating entire basins as a single point (Grayson and
2284 Bloschl, 2000). Conceptual rainfall-runoff models offer some advantages over statistical techniques in support of long-
2285 range planning for water resources management. These models represent, with varying levels of complexity, the
2286 transformation of precipitation and other meteorological forcing variables (e.g., air temperature and humidity) to
2287 watershed runoff and streamflow, including accounting for hydrologic storage conditions (e.g., snowpack, soil moisture,
2288 and groundwater). These models can be used to assess the impacts and implications of various climate scenarios by
2289 using historic meteorological time series as input, generating hydrologic time series, and then using those hydrologic
2290 scenarios as input to RiverWare. This approach enables consideration of current landscape and river channel conditions,
2291 which may be quite different than recorded in early instrumental records and which can dramatically alter a watershed's
2292 hydrologic behavior (Vorosmarty et al., 2004). Furthermore, the use of multiple input time series, system
2293 parameterizations, or multiple models, enables a probabilistic assessment of an ensemble of scenarios. The Hydrological
2294 Ensemble Prediction Experiment (HEPEX) (Schaake et al., 2007) aims to address the unique challenges of expressing
2295 uncertainty associated with ensemble forecasts for water resources management.

2296 An additional concern for mid- and long-range planning is that, as instrumental records have grown longer,
2297 they often show trends (e.g., Baldwin and Lall, 1999; Olsen et al., 1999; Andreadis and Lettenmaier, 2006) or persistent
2298 regimes (i.e., periods characterized by distinctly different statistics) (e.g., Angel and Huff, 1995; Quinn, 1981, 2002),
2299 with consequences for estimation of hydrologic risk (Olsen et al., 1998). Observed regimes and trends can have multiple

2300 causes, including climatic changes, watershed and river transformations, and management impacts (e.g., irrigation
2301 return flows and trans-basin water diversions). These issues enter into RiverWare applications directly through the use
2302 of naturalized flows, which are notoriously unreliable. For example, in assessments of water management options on the
2303 San Juan River in Colorado and New Mexico, the reliability of naturalized flows was considered to be affected by the
2304 inconsistent accounting of consumptive uses between irrigation and non-irrigation data, use of reservoir evaporation
2305 rates with no year-to-year variation, neglecting time lags in the accounting of return flows from irrigation to the river,
2306 errors in river gage readings that underestimated flows in critical months, and the lack of documentation of diversions
2307 that reduce river flows as well as subsequent adjustments to data used to compute naturalized flows.

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2309 **5. Global Change Information and RiverWare**

2310 *Climate Variability*

2311 Decision makers increasingly recognize that climate is an important source of uncertainty and potential
2312 vulnerability in long-term planning for the sustainability of water resources (Hartmann, 2005). With the appropriate
2313 investment in site licenses, training of personnel, implementation for a specific river system, and assessment efforts,
2314 RiverWare is capable of supporting climate-related water resources management decisions by U.S. agencies. However,
2315 technology alone is insufficient to resolve conflicts among competing water uses. Early in the development of
2316 RiverWare, Reitsma et al. (1996) investigated its potential role as a DST within complex negotiations between
2317 hydroelectric, agricultural, and flood control interests. Results indicated that while DSTs can help identify policies that
2318 can satisfy specific management requirements and constraints, as well as expand the range of policy options considered,
2319 they are of limited value in helping decision makers understand interactions within the river system. Furthermore, the
2320 burdens of direct use by decision makers of a DST that embodies a complex system are significant; a more useful
2321 approach is to have specialists support decision makers by making model runs and presenting the results in an iterative
2322 manner. This is the approach used by the Bureau of Reclamation in the application of RiverWare to support interstate
2323 negotiations concerning the sharing of Colorado River water supply shortages during times of drought (Jerla, 2005; U.S.
2324 Department of Interior, 2007).

2325 From the perspective of mid-range water management issues, the use of forecasts within RiverWare
2326 applications constitutes an important pathway for supporting climate-related decision making. Each time a prediction is
2327 made, science has an opportunity to address and communicate the strengths and limitations of current understanding.

2328 Each time a decision is made, managers have an opportunity to confront their understanding of scientific information
2329 and forecast products. Furthermore, each prediction and decision provides opportunities for interaction between
2330 scientists and decision makers and for making clear the importance of investments in scientific research. Perceptions of
2331 poor forecast quality are a significant barrier to more effective use of hydroclimatic forecasts (Changnon, 1990; Pagano
2332 et al., 2001, 2002; Rayner et al., 2005); however, recent advances in modeling and predictive capabilities naturally lead
2333 to speculation that hydroclimatic forecasts can be used to improve the operation of water resource systems.

2334 Great strides have been made in monitoring, understanding, and predicting interannual climate phenomena
2335 such as the El Nino-Southern Oscillation (ENSO). This improved understanding has resulted in long-lead (up to about a
2336 year) climate forecast capabilities that can be exploited in streamflow forecasting. Techniques have been developed to
2337 directly incorporate variable climate states into probabilistic streamflow forecast models based on linear discriminant
2338 analysis (LDA) with various ENSO indicators, (e.g., the Southern Oscillation Index [SOI]) (Peichota and Dracup, 1999;
2339 Piechota et al., 2001). Recent improved understanding of decadal-scale climate variability also has contributed to
2340 improved interannual hydroclimatic forecast capabilities. For example, the Pacific Decadal Oscillation (PDO) (Mantua
2341 et al., 1997) has been shown to modulate ENSO-related climate signals in the West. Experimental streamflow
2342 forecasting systems for the Pacific Northwest have been developed based on long-range forecasts of both PDO and
2343 ENSO (Hamlet and Lettenmaier, 1999). In the U.S., the Pacific Northwest, California, and the Southwest are strong
2344 candidates for the use of long-lead forecasts because ENSO and PDO signals are particularly strong in these regions and
2345 each region's water supplies are closely tied to accumulation of winter snowfall, amplifying the impacts of climatic
2346 variability.

2347 While many current water management decision processes use single-value deterministic approaches,
2348 probabilistic forecasts enable quantitative estimation of the inevitable uncertainties associated with weather and climate
2349 systems. From a decision maker's perspective, probabilistic forecasts are more informative because they explicitly
2350 communicate uncertainty and are more useful because they can be directly incorporated into risk-based calculations.
2351 Probabilistic forecasts of water supplies can be created by overlaying a single prediction with a normal distribution of
2352 estimation error determined at the time of calibration of the forecast equations (Garen, 1992). However, to account for
2353 future meteorological uncertainty, new developments have focused on ensembles, whereby multiple possible futures
2354 (each termed an ensemble trace) are generated; statistical analysis of the ensemble distribution then provides the basis
2355 for a probabilistic forecast.

2356 Changnon (2000), Rayner et al. (2005), and Pagano et al. (2002) found that improved climate prediction
2357 capabilities are initially incorporated into water management decisions informally, using subjective, ad hoc procedures
2358 on the initiative of individual water managers. While improvised, those decisions are not necessarily insignificant. For
2359 example, the Salt River Project, among the largest water management agencies in the Colorado River Basin and primary
2360 supplier to the Phoenix metropolitan area, decided in August 1997 to substitute groundwater withdrawals with reservoir
2361 releases, expecting increased surface runoff during a wet winter related to El Nino. With that decision, they risked
2362 losses exceeding \$4 million in an attempt to realize benefits of \$1 million (Pagano et al., 2002). Because these informal
2363 processes are based in part on confidence in the predictions, overconfidence in forecasts can be even more problematic
2364 than lack of confidence, as a single incorrect forecast that provokes costly shifts in operations can devastate user
2365 confidence in subsequent forecasts (e.g., Glantz, 1982).

2366 The lack of verification of hydroclimatic forecasts is a significant barrier to their application in water
2367 management, but it is not easy to resolve with traditional research efforts, because the level of acceptable skill varies
2368 widely depending on the intended use (Hartmann et al., 2002a; Pagano et al., 2002). Information on forecast
2369 performance has rarely been available to, and framed for, decision makers, although hydrologic forecasts are reviewed
2370 annually by the issuing agencies in the U.S (Hartmann et al., 2002b). Hydrologic forecast verification is an expanding
2371 area of research (Franz et al., 2003; Hartmann et al., 2003; Bradley et al, 2004; Pagano et al., 2004; Kruger et al., 2007),
2372 but much work remains and could benefit from approaches developed within the meteorological community (Welles et
2373 al., 2007). Because uncertainty exists in all phases of the forecast process, forecast systems designed to support risk-
2374 based decision making need to explicitly quantify and communicate uncertainties from the entire forecast system and
2375 from each component source, including model parameterization and initialization, meteorological forecast uncertainty at
2376 the multiple spatial and temporal scales at which they are issued, adjustment of meteorological forecasts (e.g., through
2377 downscaling) to make them usable for hydrologic models, implementation of ensemble techniques, and verification of
2378 hydrologic forecasts.

2379

2380 *Climate Change*

2381 From the perspective of long-range water management issues, the potential impacts of climate change on water
2382 resources, and their implications for management, are central topics of concern. Estimates of prospective impacts of
2383 climate change on precipitation have been mixed, leading, in many cases, to increasing uncertainty about the reliability

2384 of future water supplies. However, where snow provides a large fraction of annual water supplies, prospective
2385 temperature increases dominate hydrologic impacts, leading to stresses on water resources and increased hydrologic
2386 risk. Higher temperatures effectively shift the timing of the release of water stored in the snowpack “reservoir” to earlier
2387 in the year, reducing supplies in summer when demands are greatest, while also increasing the risk of floods due to rain-
2388 on-snow events. While not using RiverWare, several river basin studies have assessed the risks of higher temperatures
2389 on water supplies and management challenges. The near universal analytical approach has been one of sensitivity
2390 analysis (Lettenmaier, 2003):

- 2391 1) downscaling outputs from a dynamic general circulation model of the global land-atmosphere-ocean system to
2392 generate regional- or local-scale meteorological time series over many decades,
- 2393 2) using the meteorological time series as input to rainfall-runoff models to generate hydrologic time series,
- 2394 3) using the hydrologic scenarios as input to water management models, and
- 2395 4) assessing differences among baseline and change scenarios using a variety of metrics.

2396 Early assessments of warming impacts on large river basins generally showed extant water management systems to
2397 be effective for all but the most severe scenarios (Hamlet and Lettenmaier, 1999; Lettenmaier et al., 1999), with a
2398 notable exception being the Great Lakes system where increased lake heat storage was tied to loss of ice cover,
2399 increased winter lake evaporation, lower lake levels, and potential failure to meet Lake Ontario regulation objectives
2400 under extant operating rules (Croley, 1990; Hartmann, 1990; Lee et al., 1994; Lee et al., 1997; Sousounis et al., 2000;
2401 Lofgren et al., 2002).

2402 Extensive detailed studies of the ability of existing reservoir systems and operational regulation rules to meet water
2403 management goals under changed climates are fairly recent (e.g., Saunders and Lewis, 2003; Christensen, et. al, 2004;
2404 Payne et. al, 2004; VanRheenan et. al, 2004; Maurer, 2007). However, there is a rapidly growing literature on broad
2405 considerations of climate change in water resources management (Frederick et al., 1997; Gamble et al., 2003;
2406 Lettenmaier, 2003; Loomis et al., 2003; Snover et al., 2003; Stakhiv, 2003; Ward et al., 2003; Vicuna et al., 2007).
2407 Some (Matalas, 1997) that contend that existing approaches are sufficient for water resource management planning and
2408 risk assessment because they contain safety factors; however, an inescapable message for the water resource
2409 management community is the inappropriateness of the stationarity assumption in the face of climate change. While
2410 precipitation changes may remain too uncertain for consideration in the near term, temperature increases are more
2411 certain and can have strong hydrologic consequences.

2412 Cognitively, climate change information is difficult to integrate into water resources management. First, within
2413 the water resources engineering community, the stationarity assumption is a fundamental element of professional
2414 training. Second, the century timescales of climate change exceed typical planning and infrastructure design horizons
2415 and are remote from human experience. Third, even individuals trying to stay up-to-date can face confusion in
2416 conceptually melding the burgeoning climate change impacts literature. Assessments are often repeated as general
2417 circulation and hydrologic model formulations advance or as new models become available throughout the research
2418 community. Furthermore, assessments can employ a variety of techniques for downscaling. Transposition techniques
2419 (e.g., Croley et al., 1998) are more intuitive than the often mathematically complex statistical and dynamical
2420 downscaling techniques (e.g., Clark et al., 1999; Westrick and Mass, 2001; Wood et al., 2002; Benestad, 2004).

2421 GCMs and their downscaled corollaries provide one unique perspective on long-term trends related to global
2422 change. Another unique perspective is provided by tree-ring reconstructions of paleo-streamflows, which, for example,
2423 indicate that in the U.S. Southwest droughts over the past several hundred years have been more intense, regionally
2424 extensive, and persistent than those reflected in the instrumental record (Woodhouse and Lukas, 2006). Decision makers
2425 have expressed interest in combining the perspectives of paleoclimatological information and GCMs. While some
2426 studies have linked instrumental records to paleoclimatological information (e.g., Prairie, 2006) and others with GCMs
2427 (e.g., Christensen and Lettenmaier, 2006), few link all three (an exception is Smith et al., 2007).

2428 Conceptual integration of climate change impacts assessment results in a practical water management context
2429 is complicated by the multiplicity of scenarios and vague attribution of their prospects for occurrence, which depend so
2430 strongly on feedbacks among social, economic, political, technological, and physical processes. For decision makers, a
2431 critical issue concerns the extent to which the various scenarios reflect the actual uncertainty of the relevant risks versus
2432 the uncertainty due to methodological approaches and biases in underlying models. The difficulties facing decision
2433 makers in reconciling disparate climate change impact assessments are exemplified by the Upper Colorado River Basin,
2434 where reductions in naturalized flow by the mid-21st century have been estimated to range from about 45% by Hoerling
2435 and Eischeid (2007), 10 to 25% by Milly et al (2005), about 18% by Christensen et al. (2004), and about 6% by
2436 Christensen and Lettenmaier (2006). Furthermore, using the difference between precipitation and evapotranspiration as
2437 a proxy for runoff, Seager et al. (2007) suggest an “imminent transition to a more arid climate in southwestern North
2438 America.”

2439 However, in the face of circumstances nearing or exceeding the effectiveness of existing management
2440 paradigms, individuals can become more cognizant of the need to consider climate change. In the U.S. Southwest, over
2441 1999–2004, Lake Powell levels declined faster than previously considered in scenarios of extreme sustained drought
2442 (e.g., Harding et al., 1995; Tarboton, 1995), from full to only 38% capacity in November 2004 (BOR, 2004). Resource
2443 managers, policymakers, and the general public are now actively seeking scientific guidance in exploring how
2444 management practices can be more responsive to the uncertainties associated with a changing climate.

2445

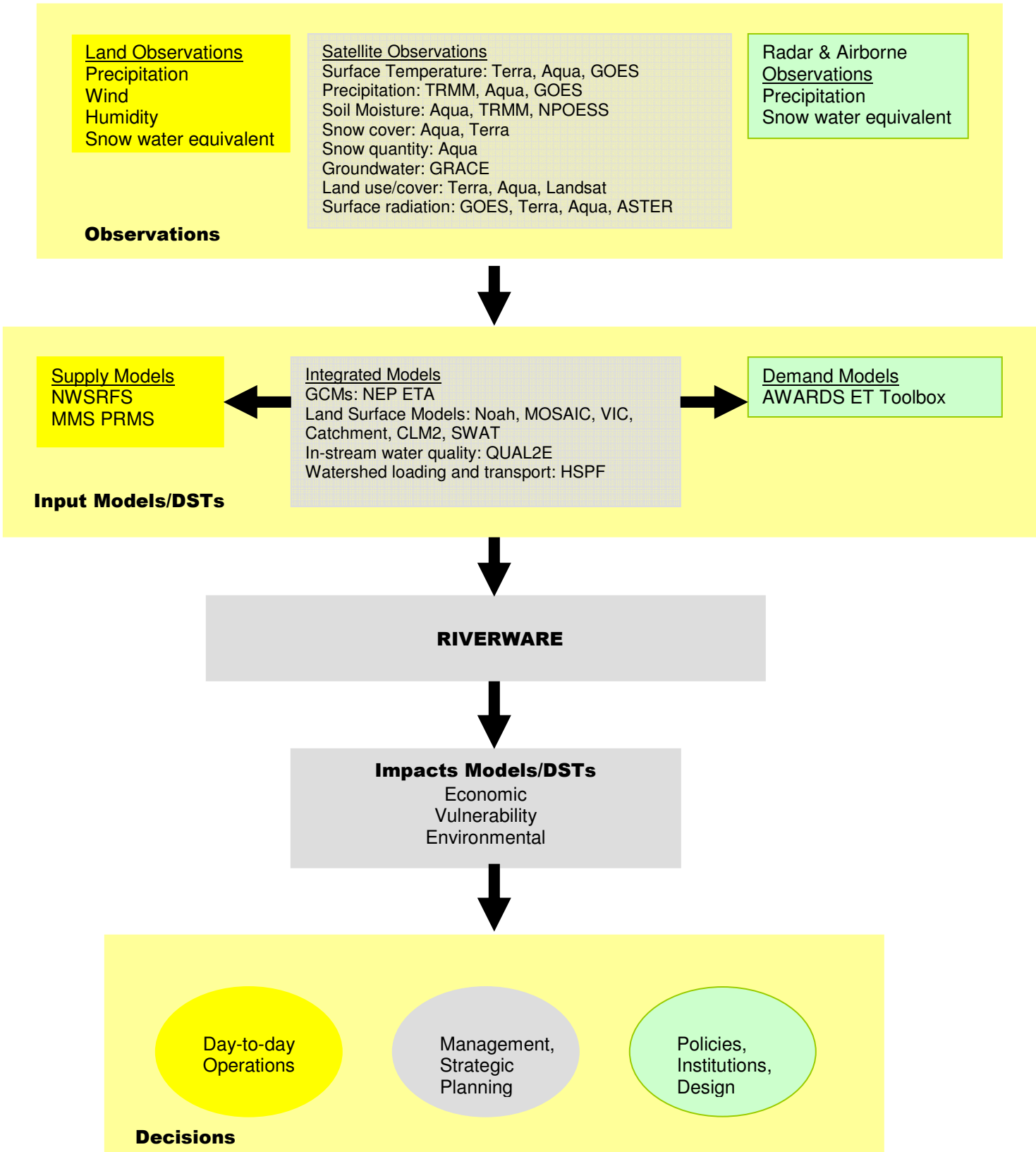


Figure 1: Illustration depicting the flow of information