

## Hydrological Modeling in India

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Human alteration of the natural environment continues nearly unabated in many parts of the world. This has resulted in exploitation of natural resources including water, over and above their limits of natural replenishment. Therefore, the International Association of Hydrologic Sciences (IAHS) has dedicated the decade 2013-2022 to explore the interactions between human society and hydrology. Due to its hydro-climatic and socio-economic diversity, India is a microcosm of the world where observations pertaining to physical hydrology as well as socio-hydrology can be made to develop generalizable insights. This has enabled a number of studies in the last few years advancing surface and groundwater modeling, process understanding, use of satellite-based or low-cost sensing for model development or improvement, inclusion of anthropogenic influences in hydrologic models, and uncertainty estimation methods. Further research in these directions is expected to continue to enable better decision making for management of water resources.

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### Introduction

Rapid transformations in infrastructure and standards of living over the last few years have left India's water resources in a precarious position. Water managers are finding it difficult to cope with the pace at which these changes are occurring and several large scale projects are being implemented to deal with foreseeable long-term water scarcity as well as hydrologic extremes of floods and droughts. This situation warrants a simultaneous advancement in hydrologic sciences so that decisions related to water management are made on a sound footing. A large number of studies exploring different aspects of Indian hydrology have been carried out in the last decadewhose synthesis allows the hydrologic community to appraise previous advancements and carve a path for the future. To this end, Mondal *et al.*

(2016) provided a comprehensive assessment of advances in hydrologic science in India. This report builds on that effort and summarizes advances in hydrologic sciences since then. It is encouraging to see that a few thrust areas of research suggested by Mondal *et al.* (2016) have been actively researched, but some challenges still persist. This report highlights advances in surface and groundwater modeling, process understanding, application of satellite-based or low-cost sensing for model development or improvement, and methods to quantify uncertainty in predicted hydrologic variables. Studies span various spatio-temporal scales and have a sound coverage of the country geographically.

### Advances in Surface Water Modeling

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(IAHS) designated the decade from 2003-2012 for improving the science of Prediction in Ungauged Basin (PUB) (Sivapalan *et al.*, 2003). This triggered a considerable amount of research interest in India due to challenges in data availability and accessibility within the country. The recent decade saw hydrologic data becoming available online from local (e.g. INDIA-WRIS, BHUVAN) as well as from global sources (e.g. APHRODITE, GLOVIS, GRDC). Although hydrological observational networks have expanded and improved across India in the last decade under the National Hydrology Project (NHP), still there is a need to improve and strengthen the quantity and quality of observation network across the country. The water cycle is also impacted due to the rapidly changing human systems. The existing observational networks are not necessarily explicitly designed to monitor changes brought about by the society. Further, hydrological models also need to be improved upon to capture the connections and interactions between hydrology and society and their co-evolution. Hence, following decade of PUB, the scientific decade 2013-2022 is dedicated to developing a comprehensive framework to the two-way feedback between society and hydrology (Socio-hydrology) entitled "Panta Rhei-Everything Flows" (Montanari *et al.*, 2013).

### Developments in Model Algorithms

In order to make better use of gridded information from the satellite directly, the Space Application Centre (SAC), Ahmedabad has developed a satellite-based hydrological model (SHM) (Paul *et al.*, 2019). SHM is a parsimonious model, runs on a 5km X 5km grid cell and has a modular structure. Grid-to-grid routing was implemented in the model based on the time variant spatially distributed direct hydrograph (SDDH) method and has been tested at the Kabini dam measurement site (Paul *et al.*, 2018a). SHM model was applied to the Subarnarekha river basin and was subjected to a robust hierarchical operational testing scheme with different criteria. The results showed that the SHM model structure, forced by satellite based inputs, is capable of realistically simulating the hydrology under different geographical and climatological conditions that exist across the country.

Real-time implementations of hydrologic models are now being attempted for monitoring the hydrologic conditions of river basins for which accurate initial

soil moisture settings is essential. Understanding the sensitivity of hydrologic simulations to initial soil moisture conditions is very limited in Indian river basins. However, this information is much needed for determining model spin-up requirements for operational hydrologic forecasts. Niroula *et al.* (2018) studied the effect of perturbations in initial soil moisture conditions on streamflow simulations during the summer monsoon season in the Ganga River basin using VIC. The study showed that streamflows converge quickly for high rainfall years as well as for wet spells within a season. The convergence takes longer for regions with low rainfall, during break period and in regions with deeper soil layers. The Ganga River basin showed high spatial variations in convergence time due to its highly varying climatic regimes and soils. Hence, model spin-up time needs to be chosen carefully for such large basins to overcome the errors in initial soil moisture. Surface layer soil moisture data assimilation described later (Patil and Ramsankaran, 2017; Patil and Ramsankaran, 2018) could have the potential to reduce the model spin-up time and needs to be explored in future research.

### Application of Existing Models

There are numerous applications of existing hydrologic models to Indian river basins. Commonly used hydrological models and their recent applications are being discussed here.

Variable Infiltration Capacity (VIC) model is a semi-distributed macroscale hydrologic model that solves water and energy balance equations at each grid cell. It is a physically-based model that can be coupled with climate models. Shah and Mishra (2016) used the VIC model to observe changes in hydroclimatic variables in the Indian sub-continental basins at long time scales. Recently, Hengade and Eldho (2016) applied the VIC model on Ashti Catchment (sub-catchment of Godavari Basin in India) to assess the impacts of LULC changes and rainfall trends on hydrological variables. Additionally, Chawla and Mujumdar (2015) employed the VIC model to separate the impacts of land use and climate change in the upper Ganga river basin. Srivastava *et al.* (2017) applied the water-budget method of the VIC-3L land surface model in the Kangsabati River Basin in eastern India to estimate evapotranspiration

(ET) indirectly using the Penman-Monteith (PM) equation. More recently, Shah *et al.* (2019) applied the VIC model in India to estimate the role of irrigation on water and energy budgets. Other than water budget estimates, the VIC model has also been applied to estimate climate change impacts on hydropower in India (Ali *et al.*, 2018).

Other than the VIC model, Noah and Community Land Model (CLM) have been used to study land surface hydrology in India. Noah is a land surface model (LSM) that simulates land-atmosphere interaction processes. It has a 1-layer canopy, 3-layer snow, and 4-layer soil as its major components. The Noah LSM has been used for soil moisture assimilation over Indian Subcontinent (Nair and Indu, 2016). The Community Land Model (CLM) represents various physical, chemical and biological land surface processes within the Community Earth System Model (CESM). Mishra *et al.* (2016) and Asoka *et al.* (2017) used an ensemble of VIC, NOAH, CLM, and MOSAIC LSMs available from the Global Land Data Assimilation System (GLDAS) for surface water storage estimation (sum of soil moisture, surface water, and canopy storage). Soil moisture was simulated from three LSMs (VIC, NOAH, and CLM) and agricultural drought events over India from 1951 to 2015 were reconstructed by Mishra *et al.* (2018). Kumar *et al.* (2017b) used the CLM Crop version 4 (CLM4) to simulate the land surface temperature (LST). Irrigation and urban biome can be incorporated into a grid cell in this model to identify the impact of irrigation. Coupling of CLM3.5 with Regional Climate Model (RegCM4) showed better skill in simulating Indian Summer Monsoon (ISM) (Maity *et al.*, 2017).

Soil Water Assessment Tool (SWAT) is a physically based, semi-distributed and continuous time model that simulates the water and sediment yield in basins over long periods. It can be applied to small watersheds as well as large river basins. Mishra and Lilhare (2016) found satisfactory performance in simulating monthly streamflow for most of the selected basins in the Indian subcontinent using the SWAT model. By using SWAT with a unique combination of slope, soil, and land cover classes, Pandey *et al.* (2017) estimated various water balance components under future climatic scenarios. Goyal *et al.* (2018) evaluated the performance of SWAT in two different catchments under different resolution (30 m to 300 m) of Digital

Elevation Model (DEM) from three sources namely SRTM, ASTER, and CartoDEM. SWAT was used for seasonal water budget analysis in the Betwa river basin (Suryavanshi *et al.*, 2017), and for runoff and sediment yield prediction in Ken basin of Central India (Himanshu *et al.*, 2017). Anand *et al.* (2018a) used SWAT to study trend in streamflow in Ganges basin.

HEC-HMS (Hydrologic Engineering Centre-Hydrologic Modelling System) can be used for both continuous and event-based hydrologic modelling. The model simulates precipitation-runoff processes within watersheds and is capable of simulating runoff at daily, monthly, and seasonal time scales. Mandal and Chakrabarty (2016) analysed different hydraulic models to detect flash flood probability using HEC-RAS and HEC-HMS software. MIKE SHE is a deterministic, distributed and physically based integrated catchment model capable of simulating surface water and groundwater interactions along with other hydrological processes. An integrated hydrologic model (MIKE SHE/MIKE 11) was used to evaluate the availability of groundwater and surface water and the effects of the future expansion of irrigated agriculture in a semi-arid watershed (Sishodia *et al.*, 2017). Storm Water Management Model (SWMM) is a dynamic simulation rainfall-runoff model often used for studying flood in urban areas. Bisht *et al.* (2016) designed a systematic drainage system using MIKE URBAN and SWMM model. 2D MIKE URBAN model overcomes the shortcomings of the 1D SWMM model for flood simulation and inundation. Rai *et al.* (2017) developed a GIS-based SWMM model to simulate streamflow for flood modelling in the Brahmani river delta.

### ***Incorporating Anthropogenic Interventions***

Rapid increases in urbanization, industrialization, and agricultural expansion over the last few decades have likely left palpable imprints on the water resources of the country. Consequently, several studies in the last few years explore methods to quantify the impact of anthropogenic interventions on various hydrologic fluxes. Most studies use a combination of hydrologic models, climate change scenarios, and socio-economic scenarios to assess how human interventions may affect long-term water availability or likelihood of floods and droughts (Anand *et al.*, 2018b; Mittal *et al.*, 2016; Wagner *et al.*, 2016; Zope *et al.*, 2016).

Wagner *et al.* (2016) assessed the joint impact of climate and land use change in medium sized Mula and Mutha River catchments draining the city of Pune. They found that the impact of land use change is evident in streamflow simulations of smaller sub-basins but not at the basin scale. Using the SWAT model, Mittal *et al.* (2016) assessed the impact of climate change and dam construction on different indicators of hydrologic alteration such as frequency, timing, and duration of high and low flows for the Kangsabati River. They found that the combined effect of climate change and dam construction is significantly greater than the individual impact of dam or climate change. Zope *et al.* (2016) performed an event-based analysis of flood peaks and flood runoff volumes for a small urban catchment in Mumbai and reported a marginal increase in both variables for floods of 2-yr and 100-yr return periods despite a 74.84% increase in the built-up area of the catchment over the period of analysis. Dey and Mishra (2017) provide a comprehensive review of different methods that can be used to separate the impacts of climate change and human interventions on hydrologic fluxes. Madhusoodhanan *et al.* (2016) provide a review of challenges in assessing the joint impact of climate and other anthropogenic interventions on water resources of the country, highlighting the need to develop fine-resolution datasets for both hydro-meteorological and water use data.

### Advances in Groundwater Modeling

Groundwater is a significant source of domestic and irrigation water in many parts of India. As the demand for water rises, this invisible resource is diminishing in both quality and quantity. This section discusses recent advances in the estimation of groundwater depletion rates, coastal groundwater modeling, modeling of water quality and groundwater management.

#### *Estimating Groundwater Depletion Rates*

Population growth and climate extremes (e.g., droughts) have contributed to increased groundwater withdrawals, which has led to significant groundwater depletion (GWD) in many areas. Environmental impacts of GWD are diverse such as decrease in baseflows resulting in drying-up of wetlands and rivers, land subsidence, saltwater intrusion, and declining water supplies. Traditional approaches for

assessing GWD rely on groundwater level data from wells that are relatively sparse or inaccessible in many aquifers, preventing a consistent assessment of GWD over major aquifers that are extremely important for agricultural production.

The Northwest India Aquifer (NWIA) has been shown to have the highest GWD rate globally, threatening crop production and sustainability of groundwater resources. Long *et al.* (2016) assessed GWD in the three-state region (i.e., Punjab, Haryana & Delhi, and Rajasthan with a total area of 438,296 km<sup>2</sup>) of NWIA. GWD rates were evaluated over the NWIA using a variety of approaches, including constrained forward modeling resulting in a GWD rate of  $3.1 \pm 0.1$  cm/a (or  $14 \pm 0.4$  km<sup>3</sup>/a) for Jan 2005-Dec 2010, consistent with the GWD rate (2.8 cm/a or 12.3 km<sup>3</sup>/a) from groundwater level monitoring data. As the quality of GRACE data and related processing techniques improve, GRACE satellites will become more valuable in assessing GWD.

Using multiple data sources (GRACE, well observations, PCR-GLOBWB model, precipitation, and sea surface temperature) and methods such as regression and dominance analysis, Asoka *et al.* (2017) concluded that long-term changes in monsoon precipitation are driving groundwater storage variability in most parts of India either directly by changing recharge or indirectly by changing abstraction. It was found that groundwater storage has declined in northern India at the rate of 2 cm yr<sup>-1</sup> and increased by 1 to 2 cm yr<sup>-1</sup> in southern India between 2002 and 2013.

#### *Coastal Groundwater Modeling*

Indiscriminate and unplanned groundwater withdrawal for fulfilling the growing freshwater needs of coastal regions leads to the problem of saltwater intrusion. Numerical modelling for variable density flow and transport is a useful tool in helping hydrologists to understand and predict how saltwater intrusion occurs in coastal aquifers. SEAWAT module in MODFLOW based on finite differences scheme has been used by various researchers to investigate the saltwater intrusion process in coastal aquifers in India (Dunlop *et al.*, 2019; Gopinath *et al.*, 2016; Lathashri and Mahesha, 2015; Maheswaran *et al.*, 2016; Surinaidu *et al.*, 2016).

Besides numerical models, data-based models, such as the artificial neural network (ANN), support vector machine (SVM), genetic programming (GP) and extreme learning machine (ELM), have also been utilized to approximate 3D variable density flow and transport processes in coastal aquifers. Yadav *et al.* (2018) used SEAWAT to generate data required for the training and testing of the data-based models. Four data-based models (ANN, SVM, GP, and ELM) were considered as proxy simulators to simulate variable density flow and transport. The selected models were also compared based on their computational ability, and results showed that ELM was the fastest technique taking just 0.5s to simulate the dataset; however, SVM was the most accurate, with Nash-Sutcliffe efficiency  $\geq 0.95$  and correlation coefficient  $\geq 0.92$  for all the wells. Unlike process-based models, SVM and ELM do not require basin information and other physical parameters in the modelling process. Thus, a data-based model such as SVM and ELM can also be utilized for saltwater intrusion modelling provided the training dataset is large enough so that the model architectures can capture the involved complexities adequately.

### **Modeling of Groundwater Quality**

Groundwater contamination has serious effects as it persists for a long time due to the high residence time of groundwater. The unsaturated zone acts as a conduit for transport of contaminants to the groundwater. Beegum *et al.* (2018) developed a modified unsaturated flow and transport package and linked it to the groundwater flow model MODFLOW and groundwater solute transport model MT3DMS. In addition to water flow in the vadose zone, the new package can also simulate solute transport involving many biogeochemical processes and reactions, including first-order degradation, volatilization, linear or nonlinear sorption, one-site kinetic sorption, two-site sorption, and two-kinetic sites sorption. Sharma *et al.* (2016) developed a numerical model for the mobile-immobile phase advective-dispersive transport equation including equilibrium sorption and the first-order degradation. The numerical model was also used to simulate experimental breakthrough curves (BTCs) for transport of chloride and fluoride through heterogeneous soil column using constant, linear, and exponential distance-dependent dispersion models. It was shown that the behavior of the concentration

profile produced with a constant dispersion model is similar to the distance-dependent dispersion model.

Radioactive wastes are one of the potential sources of radiation causing risk to the environment and human health. Recent studies have concluded that “geological isolation” is the safest long-term option for storing high level radioactive waste. Thus, waste disposal facilities aim to isolate these wastes from the environment. A fully implicit finite difference numerical model was developed by Bagalkot and Kumar (2016) to simulate two species radionuclide transport in a coupled single fracture-matrix system with variable fracture aperture and results so obtained were compared with a parallel plate model. Two distinct geometric profiles namely, sinusoidal and logarithmic have been used to capture the variation of aperture width. The dependence of advection, hydrodynamic dispersion, linear sorption, and matrix diffusion on aperture width was considered in the analysis of radionuclides transport. Geetha Manjari and Sivakumar Babu (2018) modelled the radionuclide transport from the barrier using FEFLOW, capable of simulating groundwater flow and contaminant transport problems in porous and fractured media. A three-dimensional domain with a decaying source concentration was modelled and the concentration was measured at different distances up to 50 and 200 m away from the source for short-lived (strontium ( $^{90}\text{Sr}$ ), cesium ( $^{137}\text{Cs}$ ) and long-lived radionuclides (carbon ( $^{14}\text{C}$ ) and iodine ( $^{129}\text{I}$ )). A code was developed using the built-in python interface of FEFLOW to run the simulations and make the model computationally efficient. The development of code using python interface bridged the deterministic FEM software with a probabilistic analysis.

Using an implicit finite difference scheme, Mohanadhas and Govindarajan (2018) developed a one-dimensional numerical model to describe the transport of Uranium  $^{238}\text{U}$  and its progenies in the vadose zone from uranium tiling pond situated on the soil surface. The numerical results show that the  $^{238}\text{U}$  and its progenies are migrating up to a depth of 90 m and 800 m after 10 years in silty and sandy soil, respectively. Essentially, silt may reduce the risk of contamination in the groundwater for a longer time span and at the deeper depths. In general, a coupled effect of sorption and hydro-geological parameters (soil type, moisture context and hydraulic conductivity)

decides the resultant uranium transport in the subsurface environment. Similarly, Renu and Kumar (2016) developed a finite difference numerical model to simulate the transport of aqueous benzene concentration along a fracture in a saturated fracture-matrix system. Wagh *et al.* (2018) employed the ANN model for prediction of nitrate concentration in groundwater of Kadava River basin, Nashik District, Maharashtra.

### ***Groundwater Management and Vulnerability Assessment***

Three broad methods can be used for assessing groundwater contamination risk: process-based models, statistical methods employing laboratory and site-specific data, and overlay and index methods that use the spatial representation of hydrogeologic features (Ahada and Suthar, 2018). The DRASTIC method uses several hydrogeological parameters to estimate aquifer vulnerability to pollution. Generally, DRASTIC employs seven hydrogeological factors within a GIS framework: depth to water table, net recharge, aquifer media, soil media, topography (slope), material comprising vadose zone, and hydraulic conductivity. These factors are weighted to arrive at a combined 'DRASTIC' index that indicates vulnerability. Applications of the DRASTIC method and its modifications include analyses by Ghosh and Kanchan (2016) for the Bengal alluvial tract; Kumar *et al.* (2016a) for Fatehgarh Sahib district in Punjab; Joshi and Gupta (2018) for Ajmer District in Rajasthan, and Karan *et al.* (2018) for Jharia Coalfield in Dhanbad district of Jharkhand. Using the analysis conducted for the cultivable belt of Malwa in Punjab, Ahada and Suthar (2018) reported that parts of aquifers of eastern and western Malwa region are at greater risk of contamination. On further analysis, it was noticed that two factors: vadose zone and depth to groundwater table determine the aquifer vulnerability in the Malwa region.

Several studies attempt correlated groundwater quality as indicated by the well data and the DRASTIC-predicted vulnerability index (Ghosh and Kanchan, 2016; Karan *et al.*, 2018). Ghosh and Kanchan (2016) found that the DRASTIC vulnerability index and observed arsenic concentrations in groundwater together indicated that the eastern portion of Bengal alluvial tract is more

susceptible to arsenic pollution. However, individual factors that are used to estimate the DRASTIC index were found to have varying degrees of correlation with observed arsenic concentrations, indicating that some factors are more important than others when assessing vulnerability w.r.t specific contaminants. Kumar *et al.* (2016a) found that higher vulnerability zones as indicated by the DRASTIC index were generally related to the high groundwater pollution. However, the contaminants responsible for poor groundwater quality varied across the high vulnerability zones. They, therefore, recommend that additional information related to anthropogenic activities such as land cover types, urban settlements, mining and exploration activities, etc. should also be incorporated in the DRASTIC framework to enable site-specific applicability.

Problems pertaining to groundwater management range from estimation of aquifer response to changes in recharge or pumping rates to conjunctive use of surface and groundwater. Sahoo and Jha (2017) developed a conceptual model of groundwater flow for the multi-layered aquifer system of the Kushabhadra-Bhargavi inter-basin in the Mahanadi River delta. Using this model, they evaluated the response of the aquifer system to various pumping scenarios and found that the impact of changes in pumping rates was greater for the confined aquifer when compared with that of the unconfined aquifer. They also found that multi-layered aquifers can show a nonlinear response to changes in recharge and pumping rates, with various layers displaying quite different behavior. This should be considered while planning for management strategies of such aquifers. Using a similar model-based analysis, Sashikkumar *et al.* (2017) explored the impact of artificial recharge on groundwater levels in the Kodaganar River basin in Dindigul district of Tamil Nadu. They found significant spatial variations in response to groundwater levels to artificial recharge structures.

### **Advances in Process Understanding**

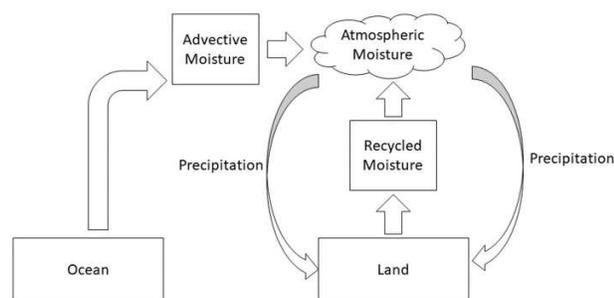
#### ***Land-Atmosphere Feedback***

Analysis of feedback processes between land and atmosphere is still a neglected area for hydro-meteorological analyses in the Indian sub-continent.

Monsoon circulations are traditionally believed to be associated with large scale circulations (Kumar *et al.*, 1999; Saji *et al.*, 1999) and among local factors aerosols have got significant attention (Bollasina *et al.*, 2011; Sarangi *et al.*, 2017). There has been growing evidence (Pathak *et al.*, 2017a; Pathak *et al.*, 2017b) on the feedback from hydrologic processes to monsoon rainfall specifically the Indian Summer Monsoon Rainfall (ISMR). The existing literature on land-atmosphere interactions emphasizing feedback processes may be classified into two broad categories:

- i. quantification of moisture supply from land sources using backward trajectory approach, and,
- ii. use of coupled land-atmosphere simulations to perform multiple experiments.

The literature on the backward trajectory approach uses a modelling technique known the dynamic recycling modelling. The atmospheric moisture over a region has two sources (Fig. 1). The first one is the advective moisture coming from distant sources, mostly the oceans. The other fraction is through evapotranspiration (ET) from land, known as recycled moisture and the associated precipitation fraction is known as recycled precipitation (Dominguez *et al.*, 2006; Eltahir and Bras, 1996). The contribution of this recycled moisture to the precipitation in the same region is known as “Precipitation Recycling” and it is characterized by the ratio of recycled



**Fig. 1: Conceptualization of recycled and advective moisture precipitation to the total precipitation.**

Pathak *et al.* (2017b) found that recycled precipitation not only affects the seasonal rainfall but also its intraseasonal variations. Intraseasonal oscillations of Indian summer monsoon rainfall is associated with the northward propagation of high

and low rainfall bands resulting in active and break periods. This also results in a north-south asymmetry of rainfall. Pathak *et al.* (2017a) showed that along with this north-south asymmetry, there also exists an east-west asymmetry that can be attributed to the recycled moisture. This land contribution mostly comes from the vegetation over the Gangetic basin.

The approach related to coupled land-atmosphere simulations involves simulation of the water cycle with coupled models either at the global or regional scale. Regional simulations have been performed by Paul *et al.* (2016) using the Weather Research and Forecasting (WRF) model coupled with CLM. They found that the changing pattern of monsoon is sensitive to the land use land cover changes. For example, deforestation along with an increase in cropland area results in a declining trend of monsoon and this is mostly restricted over the North East India and the Ganga basin. Paul *et al.* (2018b) further found that the moisture generated from the vegetation of Western Ghats contributes a significant fraction of rainfall in the state of Tamilnadu during the summer monsoon season.

Though land processes play a major role in monsoon rainfall, the operational monsoon model in India, the Climate Forecast System Version 2 (CFSv2) still does not consider land processes adequately (Sahana *et al.*, 2018). CFSv2 also has a very high dry bias over the Gangetic basin, i.e., it underestimates the rainfall amount. Devanand *et al.* (2018) hypothesized that poor representation of land processes is responsible for the dry bias in CFSv2. Devanand *et al.* (2018) performed regional coupled land-atmosphere simulations using WRF-CLM with CFSv2 boundary conditions and have found a significant decrease in the dry bias with improvements in the seasonal monsoon simulations. In the regional coupled framework, simulations of ET get significantly improved and produce high recycled precipitation. Further to this, a better representation of the Himalayas controls the circulation patterns and produces reliable estimates of monsoon rainfall.

One of the major limitations of the above-mentioned studies is that they do not explicitly consider the irrigation over Indian cropland, which may affect significantly the water cycle. The irrigation practices are also different in India, for which improvisation of

models is required. Consideration of India specific irrigation, analysing its impacts on monsoon rainfall and understanding the predictability of coupled processes dominated by feedback may be considered as a potential area for future research.

### ***Integrating Eco-Hydrological Processes***

The hydrological cycle and the biogeochemical cycle of the earth are highly interconnected. In the past 25 years, across the world, the earth science community has established Critical Zone Observatories (CZO) to study through detailed long-term observations, the impact of land-use changes and climate variability on the hydrological and biogeochemical fluxes of different ecosystems across spatiotemporal scales (Banwart *et al.*, 2013). There is only one CZO located in India in Kabini, a subbasin of Cauvery. The specific aim of the Kabini CZO is to provide insights into the influence of climate variability and agriculture on water and biogeochemical cycles and has been monitored since 2003 (Sekhar, 2016). This has been a test bed for several studies related to a) soil moisture retrieval from Microwave remote sensing, b) algorithms to retrieve latent heat flux and ET estimation, c) retrieval of soil hydraulic properties, and d) development and testing of the groundwater model AMBHAS-GW (Sekhar, 2016). India with its diverse ecosystems due to its climatic variability, topographic and geographical setting in the tropics needs to have more CZOs to better understand the interactions and feedbacks due to landuse, climate and human interventions in different agro-climatic settings. Indian contributions to the Earth system science from such CZOs near the tropics could help considerably to improve process representations in hydrologic models and their reliability as tools for water resources planning and management.

### ***Hillslope Experiments***

The process of runoff generation in hillslopes is complex due to the requirement of detailed hillslope conditions for process conceptualization. Chouksey *et al.* (2017) attributed the influence of subsurface flow in the runoff process to soil and macropore characteristics in the Western Himalayan hillslopes. They developed an empirical hillslope hydrological model by observing runoff generation process using controlled rainfall simulation (drop size, velocity, kinetic energy, intensity ranges, uniformity and continuity of

flow) over vegetated hillslopes and estimating vertical profile of soil macropore fraction, moisture variability and soil characteristics. A dye infiltration test on soil column for macropore fraction variability was also carried out. Chouksey *et al.* (2017) reported that for long duration rainfall, bottom soil layer at a depth of 600-1000 mm became permanently saturated with the passage of time. The results of Chouksey *et al.* (2017) highlight the importance of AMC and subsurface conditions on runoff generation processes. Furthermore, they also estimated surface runoff, subsurface runoff and soil/macropore storage components with high accuracy.

### ***Surfacewater Groundwater Interactions***

Mukherjee *et al.* (2018a) attributed the observed reduction in summer flows in lower reaches of Ganges River in the period 1999-2013 to nearly 59% reduction in baseflow contributions from the Gangetic aquifers. They used a combination of in-situ and remotely sensed data along with numerical modeling techniques to quantify the interaction between the Gangetic aquifers and flow in different reaches of the river. Sahoo and Jha (2017) explored aquifer-river interaction using a conceptual model of the multi-layered aquifer system of the Kushabhadra-Bhargavi inter-basin in the Mahanadi River delta. They found that during the post-monsoon season the middle and lower reaches of the Kushabhadra and Bhargavi rivers recharge the adjoining aquifer (4.41 Mm<sup>3</sup>), while the aquifer provides baseflow to the rivers (0.85 Mm<sup>3</sup>) during pre-monsoon season. A similar analysis by Surinaidu *et al.* (2016) uses two physically-based models, SWAT, and MODFLOW, to simulate aquifer-stream interactions in the Ramganga sub-basin of the Ganges river basin. They use this semi-coupled modeling framework to assess the technical feasibility of conjunctive use of surface and groundwater in the Ganges river basin so that the aquifer itself can act as a storage unit, diminishing the need for large infrastructural investments in surface storages. They found that controlled pumping may help create subsurface storage in the aquifer, which combined with appropriate artificial recharge structures can simultaneously maintain groundwater levels. Although baseflow will be reduced in such a scenario unless additional natural recharge from rainfall is available.

## Advances in Observations

### Application of Remote Sensing Data

In the present decade, understanding the link between Water, Energy and Nutrient cycles is of growing interest around the globe for sustainable hydro-ecological management practices. Especially in different agro-climatic regions in India, varying climate from tropical to temperate, Himalayan to plains and, rainforests to deserts has significant control over the hydrological processes at local and global scales. Moreover, growing anthropogenic activities and, interventions to the natural systems are prominent in the present day, showing significant impact on these cycles. Recent advancement in geo-spatial technologies and global satellite platforms provide synoptic information to understand and monitor the changes in such coupled cycles (Fig. 2). The availability of global data sets, GIS techniques, and, efficient data analysis algorithms have been helping researchers to analyze these cycles at different scales. Table 1 lists the recent research work carried for Indian River basins using advanced satellite data.

### Precipitation Estimates

Precipitation plays an important role in the hydrological cycle and acts as a dominant source of water for all living beings. Accurate estimation of precipitation is crucial for agriculture dependent countries like India, where precipitation impacts socio-economic conditions of the society. Satellite derived rainfall estimates has proven successful for representing the spatial distribution of rainfall over conventional methods. For example, Mondal *et al.* (2018) showed successful applicability of TRMM Multisatellite Precipitation Analysis (TMPA) data in climate and hydrological studies. However, a similar product (TRMM2A25-V7) in Northeast Indian region largely underestimated rainfall when validated with a dense rain gauge network (Terao *et al.*, 2017). INSAT-3D derived rainfall by Hydro-Estimator Method was found to estimate heavy rainfall episodes more accurately when compared to light rain events (Mitra *et al.*, 2018). Global Precipitation Measurement is a new data source and provides relatively higher spatial ( $0.1^\circ \times 0.1^\circ$ ) and temporal (30 min) resolution for hydro-climatological studies (Skofronick-Jackson *et al.*, 2017).

**Table 1: Application of remote sensing input for hydrological studies in the Indian river basins**

S.No.	Hydrological Parameter	Study Area	Satellite Data Used	Reference
1	Precipitation	Indian Monsoon Region, Northeastern Indian Subcontinent	INSAT, TRMM, IMR, IMERG	Terao <i>et al.</i> (2017), Mitra <i>et al.</i> (2018), Shah and Mishra (2015)
2	Snow Cover	Indian Himalayan Glaciers, Pindari glacier, Himalayan River Basin	TanDEM-X InSAR, GRACE, Landsat Imagery, MODIS, Sentinel-1	Wulf <i>et al.</i> (2016), Pandey <i>et al.</i> (2018), Snapir <i>et al.</i> (2019)
3	Evapotranspiration	Kangsabati Basin, Ganga Basin	MODIS	Srivastava <i>et al.</i> (2017), Shah and Mishra (2016)
4	Soil Moisture	Bundelkhand Region, Indian Subcontinent	MODIS, SMOS,	Padhee <i>et al.</i> (2017), Varikoden and Revadekar (2018), Patil and Ramsankaran (2018)
5	Water Level and Ground water storage	North-West India, Ganga-Brahmaputra, Godavari, Krishna and Mahanadi Basin, Western and Southern Part of India, Kosi River Basin	GRACE, ENVISAT, AVHRR, MODIS, Jason-2, TRMM, AMSR-E, AMSR-2	Bhanja <i>et al.</i> (2017), Mukherjee <i>et al.</i> (2018a), Chembolu <i>et al.</i> (2018)
6	Water Quality	Brahmani Basin, Beas Basin,	MODIS, Landsat-8	Swain and Sahoo (2017), Kumar <i>et al.</i> (2016b)

\*TRMM - Tropical Rainfall Measurement Mission, IMR – INSAT Multi-spectral rainfall, IMERG -Integrated Multisatellite Retrievals for GPM, GPM - Global Precipitation Measurement, GRACE - Gravity Recovery And Climate Experiment, MODIS - Moderate Resolution Imaging Spectroradiometer, ENVISAT - Environmental Satellite, AVHRR - Advanced Very High Resolution Radiometer, AMSR - Advanced Microwave Scanning Radiometer

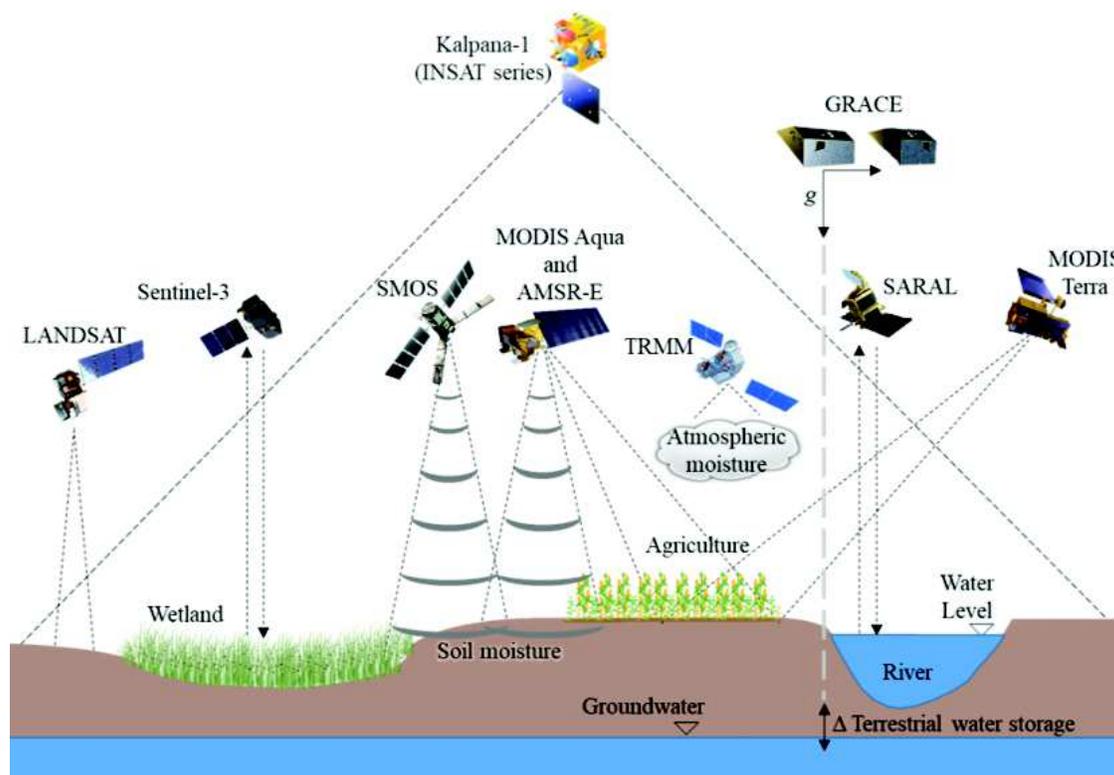


Fig. 2: Active satellites of the present decade (2010-2020) used in different fields of hydrology

### ***Snow and Ice Cover Mapping***

In the global hydrological cycle, snow and glaciers hold one-sixth of water and are a key environmental indicator for both local and global climate change. AWIFS, MODIS, LISS, Landsat, Sentinel and DEM products like ASTER, Cartosat and SRTM are the different dataset used for mapping snow cover and glacier mass changes (Behera *et al.*, 2018; Gaddam *et al.*, 2016; Joshi *et al.*, 2018; Mukherjee *et al.*, 2018b; Varade and Dikshit, 2019). Using data from MODIS Terra, Shukla *et al.* (2017) mapped snow cover variability and estimated its relationship with topographical features in the Satluj River basin. Pandey *et al.* (2018) identified the snow area and snout point for different years by using a series of multi-date Landsat imagery with manual digitization. Snapir *et al.* (2019) produced monthly maps of dry and wet snow areas by applying data fusion techniques to MODIS fractional snow cover and Sentinel-1 wet snow mask. A distributed hydrological model applied in the Sutlej Valley showed 45% of snow and glacier contribution to annual discharge of the river at the mountain front revealing the importance of snow melting runoff (Wulf *et al.*, 2016).

### ***Evapotranspiration***

In the global water cycle, evapotranspiration (ET) affects the climate feedback mechanism and availability of freshwater resources. ET can be estimated using satellite inputs either by energy balance methods (eg. SABEL, 3T-Model, etc.) or hydrological models (eg. VIC, SWAT, etc.). A comparative analysis between ET estimates of VIC-3L model and MODIS sensor, with Penman-Monteith based *in-situ* ET estimates revealed underestimation of MODIS-ET values due to cloud cover and leaf shadowing effects (Srivastava *et al.*, 2017). Using *in-situ* and satellite based observations and the VIC model with an irrigation scheme, it was showed that the presence of irrigation substantially alters water budget and land surface temperature (LST) in the sub-continental river basins like Indus and Ganga basin (Shah *et al.*, 2019).

### ***Soil Moisture***

Soil moisture plays a crucial role in the partitioning of rainfall into runoff and infiltration, thus eventually affecting ET and groundwater recharge. Hence, any

error in the simulation of soil moisture by hydrologic models, due to errors in rainfall observations, soil hydraulic parameters or structural error in process representation, would affect streamflow simulations. Patil and Ramsankaran (2017 and 2018) implemented an EnKF based assimilation of SMOS soil moisture to study the improvements in streamflow simulations of SWAT model in Munneru catchment of the Krishna River basin. Although they found that improvements in streamflow simulations are only moderate, probably due to errors in other model inputs and/or model structure, the method demonstrates significant potential for large-scale implementation of real-time hydrologic model simulations for the entire country.

A time-based function of space-borne soil moisture proposed by Padhee *et al.* (2017) was found to have a better association with crop yield than vegetation based indices in arid and semi-arid areas of Bundelkhand region in India due to the direct influence of soil moisture in irrigated and rainfed conditions. Soil moisture can also be used as an efficient indicator in the hydrology because of its strong association with rainfall. This association was explained by Varikoden and Revadekar (2018) using a soil moisture product derived from four passive and two active microwave sensors spanning over the period 1979-2013 from the European Space Agency and gridded daily rainfall data from the Indian Meteorological Department.

### ***Ground Water Storage and Water Level***

Groundwater is a vital resource to sustain agricultural, industrial, and domestic activities in populated countries like India or arid regions (Middle East and North Africa). It is also an important component of the global water cycle. Chembolu *et al.* (2018) successfully used altimetry based water level data to develop wetland storage-elevation curves and to improve understanding of river-flood plain wetland interactions for the Kosi River. Bhanja *et al.* (2017) found that a shift in groundwater withdrawal pattern and management practices for sustainable water utilization appeared to slow recharge to aquifers.

### ***Development of in-situ Observational Techniques***

Collection of in-situ data adopting traditional methods is often expensive and time consuming, for example, bathymetry data of water bodies such as tanks and

reservoirs. Although bathymetry data of some large reservoirs are available to derive the stage-storage relationships, often such data are not available for large irrigation tanks and lakes which play a significant role in the hydrology of the basin. Hence, modelling the hydrology of such systems with limited data involves large uncertainties. Young *et al.* (2017) developed and demonstrated a small, low-cost robot-assisted surface data collection method at three different sites within the Arkavathy Basin located near Bangalore, Karnataka. Further, an automated workflow to integrate high-resolution bathymetry data through photogrammetry from the Unmanned Aerial Vehicle (UAV) over dry regions of the tank bed and the bathymetry of the submerged portions derived from the Unmanned Surface Vehicle (USV) using sonar sensors was evaluated. The study showed that apart from being low cost, this technology enables collecting highly accurate data rapidly for large regions to improve the hydrologic understanding and modelling of highly human impacted watersheds.

### **Uncertainty Modeling**

Hydrologic modelling is burdened with uncertainties at several levels. Even when the model is physically consistent with the processes occurring at different spatial and time scales, its application in a given context typically requires model calibration and validation, both of which introduce uncertainties due to parameter uncertainty. Limitations due to quality and quantity of data – which is rather unfortunately common in the Indian context – render the model parameters particularly uncertain. The processes such as evapotranspiration, streamflow, infiltration and groundwater recharge are modelled differently in different hydrologic models and thus may result in varying inferences. Additionally, when the hydrologic models are used to draw inferences about how the hydrologic future is likely to unfold under climate change, uncertainties introduced by scale mismatch between the global climate models (GCMs) providing the climate projections and the processes modelled by the hydrologic models, along with the climate change scenarios used in arriving at the climate projections introduce a significant uncertainty. Quantifying such uncertainties has been a major thrust of research in the country in the last about five years.

Chawla and Mujumdar (2018) used the VIC

model to segregate the contribution to overall uncertainty in hydrologic projections from GCMs, emission scenarios, land use scenarios, stationarity assumption of the hydrologic model, and internal variability of the processes, with Upper Ganga Basin (UGB). For modelling urban floods, Intensity-Duration-Frequency (IDF) relationships are often used to obtain precipitation intensity. Change in IDF relationship due to climate change (and other forcings) will affect hydrologic designs to mitigate the impacts of floods. Estimation of such change needs to account for the significant uncertainty due to the choice of climate models, scenarios and parameters of the extreme value distributions. Chandra and Mujumdar (2018) used a Bayesian hierarchical model to obtain spatial maps and to quantify the associated uncertainties in the return levels of precipitation extremes in urban areas.

Climate change poses serious implications for the water resource in the Indian sub-continent. Majority of the sub-continental river basins in the country are projected to witness wetter and warmer climate in the 21st century thereby affecting water-availability in the region (Mishra and Lihare, 2016). In addition to climate change, changes in land use/land cover alter the hydrologic cycle and ultimately impact water resources. Hydrological model simulations may have uncertainties due to model structure, model parameters, and model input data.

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Several methods exist in the literature for modeling the uncertainty in projections from a hydrological model. For example, uncertainty analysis (parameter and model predictive uncertainty) of the simulations from the SWAT model has been done using Sequential Uncertainty Fitting (SUFI-2)).

## Concluding Remarks

The increasing availability of hydro-climatic datasets in the last few years has seen tremendous growth in the number of studies on modeling of hydrological processes in medium sized to large river basins. Efforts have also been made to develop hydrological models that are tailor-made to datasets and hydro-climatic conditions in Indian basins, which should be further consolidated. Modeling studies still need to overcome the challenges of incorporating snow/glacial melt processes, non-stationarity, and anthropogenic influences. Strong efforts are also needed to develop socio-hydrological models across a range of socio-economic and hydrological settings to address the needs of a diverse society. As advances in hydrology depend strongly on observations, development of low-cost sensors and crowd sourcing approaches for data retrieval need to be further explored. Finally, understanding the diversity of eco-hydrological processes and surface-groundwater interactions across the country should be aided via development of critical zone observatories across a hydro-climatic gradient.

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