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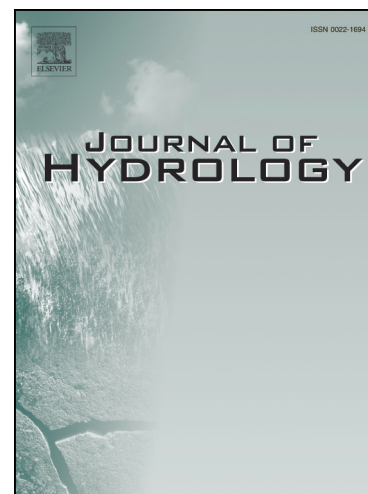
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1 An integrated modeling system for estimating glacier and snow melt driven
2 streamflow from remote sensing and earth system data products in the Himalayas

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21 Abstract

22 Quantification of the contribution of the hydrologic components (snow, ice and rain) to
 23 river discharge in the Hindu Kush Himalayan (HKH) region is important for decision-

DRAFT

making in water sensitive sectors, and for water resources management and flood risk reduction. In this area, access to and monitoring of the glaciers and their melt outflow is challenging due to difficult access, thus modeling based on remote sensing offers the potential for providing information to improve water resources management and decision making. This paper describes an integrated modeling system developed using downscaled NASA satellite based and earth system data products coupled with in-situ hydrologic data to assess the contribution of snow and glaciers to the flows of the rivers in the HKH region. Snow and glacier melt was estimated using the Utah Energy Balance (UEB) model, further enhanced to accommodate glacier ice melt over clean and debris-covered tongues, then meltwater was input into the USGS Geospatial Stream Flow Model (GeoSFM). The two model components were integrated into Better Assessment Science Integrating point and Nonpoint Sources modeling framework (BASINS) as a user-friendly open source system and was made available to countries in high Asia. Here we present a case study from the Langtang Khola watershed in the monsoon-influenced Nepal Himalaya, used to validate our energy balance approach and to test the applicability of our modeling system. The snow and glacier melt model predicts that for the eight years used for model evaluation (Oct 2003 to Sep 2010), the total surface water input over the basin was 9.43 m, originating as 62% from glacier melt, 30% from snowmelt and 8% from rainfall. Measured streamflow for those years were 5.02 m, reflecting a runoff coefficient of 0.53. GeoSFM simulated streamflow was 5.31m indicating reasonable correspondence between measured and model confirming the capability of the integrated system to provide a quantification of water availability.

Keywords: Himalayas, glacier melt, energy balance, stream flow

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2 **1. Introduction**

3 The Hindu Kush Himalayan (HKH) region possesses a large resource of snow and ice,
4 which act as a freshwater reservoir for irrigation, domestic water consumption or hydro-
5 electric power for billions of people in Asia. Snow and glacier-melt represent a
6 significant source of surface water and influence many aspects of hydrology including
7 water supply, erosion and flood control (National Research Council, 2012). With
8 projected climate-induced changes in snow and ice and population growth, the region is
9 at risk of experiencing water stress in the coming years (Immerzeel et al., 2010;
10 Immerzeel et al., 2012; Kaser et al., 2010). There are, in particular, concerns about the
11 effect of climate change on snow water equivalent, snowmelt runoff, glacier melt runoff
12 and total streamflow and their distribution due to mean global temperature increases.

13 Some recent studies published the impact of temperature increase on glacier melt runoff
14 high altitude basins (Barnett et al., 2005; Singh and Kumar, 1997a; Singh and Kumar,
15 1997b), but these tend to be over-estimated or conducted on small basins (Savoskul and
16 Smakhtin, 2013). Research has shown that there are significant differences across the
17 HKH region with regard to the contribution of glacier and snow melt to hydrological
18 systems, changes in the timing or amount of snowmelt due to increasing temperatures or
19 decreasing winter precipitation due to climate change. These changes may have far-
20 reaching societal consequences, particularly in Asia (Bookhagen and Burbank, 2010;
21 Immerzeel et al., 2012). More information is needed to monitor and anticipate snow and
22 glacier ice melt runoff at larger scales to improve water resources management and flood
23 protection (Jeuland et al., 2013).

DRAFT

1 The monitoring capability of hydrologic resources in this region is challenged by the
2 difficulty of installing and maintaining a climate and hydrologic monitoring network, due
3 to limited transportation and communication infrastructure and difficult access to glaciers.
4 As a result of the high, rugged topographic relief, ground observations in the region are
5 extremely sparse (Lo et al., 2011). For example, only a few glaciers are currently
6 monitored for mass balance measurements in the Himalaya (Dobhal et al., 2008; Wagon,
7 2007; Wagon et al., 2013). In the recent years, remote sensing-based modeling has
8 helped provide has been increasingly used in recent years to estimate water resources
9 (Thayyen and Gergan, 2010) and thus has helped improve water resources decision
10 making and management in these data-scarce areas.

11 While some progress has been made in understanding the contribution of snow and ice
12 melt to streamflow in the Himalaya using degree-day or simple ablation models
13 (Immerzeel et al., 2010; Immerzeel et al., 2012; Racoviteanu et al., 2013), a region-wide
14 estimate of water resources is hampered by the fact that these models are not in public
15 domain, coupled with lack of access or technical expertise of local institutions. Another
16 significant barrier for decision makers in monitoring and understanding the impact of
17 climate-induced changes in snow and glaciers on water resources is the scientific
18 disciplinary divide that isolates glacier experts and hydrological analysts, as well as lack
19 of integrated tools that allow institutional actors trusted by decision makers to conduct
20 analyses themselves. Developing a tool that can address all these limitations by
21 integrating snow, ice and precipitation information from both satellite and local sources is
22 critical to allow appropriate response to natural hazards to the population (Wisner et al.,
23 2004), and is a focus of this study.

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2 Here we present an integrated modeling capability developed to meet the water resources
3 planning needs of this broad, multi-nation region taking full advantage of NASA Earth
4 Science data and modeling products. We developed a hydrologic tool that can be used at
5 basin or sub-basin scale in an easy-to-use graphical user interface framework accessible
6 to most users. The modeling needs were addressed by integrating an enhanced, gridded
7 version of the Utah Energy Balance (UEB) snowmelt model, denoted here as UEBGrid
8 and the Geospatial Stream Flow Model (GeoSFM) into the Better Assessment Science
9 Integrating point and Nonpoint Sources (BASINS) modeling system. We downscaled
10 various NASA gridded remote sensing and climate products and combined them with
11 higher-resolution data for use in these models.

12 The resulting tool, publicly available to both the hydrological and cryospheric
13 communities online (http://hspf.com/pub/HIMALA_BASINS/) is referred here as
14 HIMALA BASINS, and will be maintained in conjunction with the U.S. Environmental
15 Protection Agency's (EPA) open-source BASINS tool. The HIMALA BASINS model
16 grew out of a decade of collaborative work between USAID's Office of Foreign Disaster
17 Assistance, USGS and ICIMOD and was initially funded through the Asia Flood
18 Network, a project designed to produce satellite-derived rainfall data products used to
19 drive hydrological models. Since 2003, the NOAA Climate Prediction Centre's RFE2
20 (Xie et al., 2002) products have been validated by ICIMOD (Shrestha, 2010) within the
21 hydrological model GeoSFM (Artan et al., 2007). ICIMOD has been training
22 collaborators from its member countries to use these products. Here, we address the need
23 to augment the GeoSFM with a capability to model water melt from snow and glacier ice,

DRAFT

1 which was previously not taken into account in earlier versions of this model.

2 The novelty of HIMALA BASINS tool consists in allowing the user to isolate various
3 components of streamflow (rainfall, snow and glacier ice melt) in a cost-free, open-
4 source graphical-user interface-based system that can be used for government and
5 institutional decision-making. Given the limitations in the spatial frequency, temporal
6 resolution and accuracy of satellite data, this study does not claim to provide the most
7 accurate estimate of the streamflow components at large scales in the HKH region.

8 Rather, we focus on developing and validating a tool that is capable of integrating glacier
9 melt components as well as high-resolution climate data when available. In this paper we
10 focus on the methodology used to integrate UEB and GeoSFM into a seamless product,
11 illustrated for eight years of simulations. A more thorough description of the model
12 results is presented in another study (Gupta et al., 2014).

13 Our study addresses two needs: (1) to improve the understanding of the contribution of
14 snow and ice to Himalayan water resource and (2) to assist with improved management
15 of water resources, evaluation of projected of climate change impacts on water resources,
16 and advanced modeling and data assimilation capability available to users in the
17 Himalayan region. Here we present results from a very high altitude, highly glaciated
18 region where we document the contribution of snow and ice melt to streamflow and
19 demonstrate the importance using an integrated model for these regions. The paper is
20 structured as follows: we first describe the integration of the two models (UEB and
21 GeoSFM) into the BASINS modeling framework; we then describe the satellite data
22 products, downscaling algorithms and glacier mapping methods, and finally we present
23 model results and discussion for the Langtang Khola case study in the Nepal Himalaya.

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2. Study site

The test site for the HIMALA BASINS methodology is the Langtang Khola Watershed in Nepal Himalaya, covering a surface of 360 km², with an elevation range of 3737 m to 7174 m (Figure 1). Various research studies focused on the contribution of snow and ice-melt to streamflow in the Langtang Khola watershed (Immerzeel et al., 2010; Immerzeel et al., 2012; Racoviteanu et al., 2013). We chose Langtang Khola as a validation site due to the availability of climate records from Nepal Department of Hydrology and Meteorology (DHM), as well as for comparison with these past studies, using different methodology. The prototype system was tested on this watershed by the HIMALA team while further evaluation of the system is being conducted by ICIMOD and the partners from regional member countries in larger basins in the HKH region (Narayani, Manas and Jhelum basins).

3. Modeling framework: HIMALA BASINS

In this study, we linked a snow melt model with a stream flow model within a version of the BASINS software developed and maintained at U.S EPA. BASINS consists of a pre-existing suite of hydrological models and supporting tools, and was chosen for this project due to its implementation using the MapWindow Geographic Information Systems (GIS). HIMALA BASINS incorporates two models: the UEBGrid (for estimating meltwater components) and GeoSFM models (for hydrologic modeling and routing), both of which were implemented as BASINS plugins by AQUA TERRA Consultants, the prime contractor for development and support of BASINS. The HIMALA BASINS version of BASINS is fully available to users via free download at http://hspf.com/pub/HIMALA_BASINS/. The HIMALA BASINS User's Manual

DRAFT

(located in the documentation folder at the same URL) guides the user through each component so that one component builds on another beginning with the preprocessing and downscaling routines, building UEBGrid-required layers in GeoSFM, running the UEBGrid, and finally running the GeoSFM program. The acquisition of data is discussed when introducing each component within the manual so that the input data required for each model are not confused. Running the UEBGrid and GeoSFM independently is also described in the manual.

3.1 Data sources and downscaling algorithms

To estimate the effectiveness of using both the UEB and GeoSFM within the same framework, we combined NASA gridded climate data and remote sensing products with higher resolution elevation data for use with the model (Figure 2). Below we describe these data sources as well as the preprocessing scripts that were developed to extract and downscale NASA products into the format required for these models.

3.1.1 Climate data

Climate data used to drive the snow and ice melt model was derived from MERRA and RFE2 data sources (Table 1). We developed downscaling methods for temperature, precipitation, wind speed, relative humidity, shortwave and longwave radiation (Figure 3) to match the 90-m scale of the melt model, chosen based on the Shuttle Radar Topography Mission (SRTM) elevation data described below. This high resolution is required to estimate glacier and snow melt appropriately in our high altitude glaciated basin. The downscaling methods were implemented in R (Bates et al., 2012).

DRAFT

1 Because of the lack of comprehensive meteorological data in the basin, we chose to use
2 MERRA inputs. MERRA is a recent near-real time global climate reanalysis product
3 developed at NASA, based on the Goddard Earth Observing System version 5 (GEOS-5),
4 NASA general circulation model (Rienecker et al., 2011; Suarez et al., 2008) and
5 National Centers for Environmental Prediction (NCEP) Gridpoint Statistical Interpolation
6 (GSI) analysis (Wu et al., 2002). MERRA temperature, wind speed and relative humidity
7 are reported at a height of 2 m above ground, at a spatial resolution of $2/3^\circ$ longitude by
8 $1/2^\circ$ latitude, and hourly time resolution. Incoming shortwave and longwave radiation are
9 reported at the surface, at coarser resolution of 1.0° by 1.25° and 3-hourly time step
10 (Lucchesi, 2012). All MERRA records are available from 1979 to present.

11
12 MERRA hourly temperature data were averaged into three hour blocks, bilinearly
13 interpolated and projected to 90-m using functions in the R raster library (Hijmans et al.,
14 2013). They were also adjusted for elevation differences between the effective elevation
15 determined from the geo-potential height that MERRA used and SRTM DEM elevation
16 using a monthly lapse rate from Liston and Elder (2006). Relative humidity was
17 calculated from MERRA specific humidity using a monthly dew point lapse rate, also
18 from Liston and Elder (2006) and the same elevation differences as for temperature.
19 Horizontal wind speed magnitude was obtained from eastward and northward wind
20 components from MERRA and was interpolated and projected to 90-m resolution.
21 MERRA reports three hourly incoming solar radiation at an elevation corresponding to
22 the MERRA geo-potential height instead of the actual elevation from sea level. A
23 pressure based atmospheric attenuation coefficient was calculated for each time step and

DRAFT

used to adjust MERRA incoming solar-radiation to the grid SRTM DEM elevation using a standard atmosphere pressure elevation relationship. Incoming longwave radiation was calculated based on downscaled air temperature following the methods of Liston and Elder (2006).

We used daily total precipitation estimates from RFE2 data. These records were constructed using four observational input data sources, namely: approximately 280 GTS stations, geostationary infrared cloud top temperature fields, polar orbiting satellite precipitation estimate data from SSM/I, and AMSU-B microwave sensors (Xie et al., 2002). Near-real time daily rainfall estimations are available for the Southern Asian domain (70°-110° East; 5°-35° North) at a spatial resolution of 0.1° by 0.1° beginning on May 01, 2001. The RFE2 generally underestimates intense rainfall events and overestimates rainfall in rainshadow and arid areas (Shrestha, 2010). Comparing with gauge data, the RFE2 has daily rainfall bias of -1.1 mm/day over the period 2003 to 2006, with a root mean square error from gauges over the whole of Nepal of -4.0 mm/day (Shrestha, 2010). RFE2 daily precipitation data was divided into three-hourly precipitation increments assuming uniform precipitation within the day and bilinearly interpolated to the 90 m spatial resolution.

3.1.2 Elevation and glacier data

Elevation data came from the SRTM v.4 (CGIAR), a hydrologically-sound, void-filled DEM (CGIAR-CSI, 2004). The vertical accuracy of the SRTM DEM in this area, was reported as $31 \text{ m} \pm 10 \text{ m}$ (Racoviteanu et al., 2013). An orthorectified ASTER scene

DRAFT

from Oct 30, 2003 covering the entire Trishuli basin was used as a basis for delineating glacier outlines and variables needed as input in the melt model. The scene had high contrast over glaciers, minimal cloud cover, and was acquired at the end of the ablation season, so it was well-suited for computing a glacier albedo in absence of seasonal snow. Glacier outlines needed for the melt model were derived using semi-automatic methods (band ratios $\frac{3}{4}$ with a threshold of 2.0) described in detail elsewhere (Racoviteanu et al., 2009; Racoviteanu et al., 2008). Debris-covered ice was delineated manually using on-screen digitizing on false color composites (ASTER 321 and 543) and texture filters. Substrate albedo values for the glacier surface were determined from ASTER satellite reflectance values on a cell-by-cell basis using single-band to broad-band conversion algorithms (Greuell and Oerlemans, 2004; Greuell et al., 2002).

3.1.3 Land cover and soil data

Land cover data came from the MODIS Land Cover Type Yearly L3 Global 500-meter version 5.1 product (Figure 3). The product is comprised of five classification schemes constructed from a year of MODIS Terra and Aqua observations using a supervised decision-tree classification method (Friedl et al., 2010). Land Cover Type 1, the International Geosphere-Biosphere Programme (IGBP) global vegetation classification scheme consisting of 17 classes, was used to derive appropriate land-cover based layers (canopy height, canopy structure, canopy cover fraction, leaf area index) needed to run UEBGrid and to compute basins response and flow distribution in GeoSFM. Since the rate of runoff generation and the rate of overland flow transport are influenced by land

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cover (Asante et al., 2008), these data are used in the model to compute vegetation roughness and overland velocity.

The GeoSFM requires soil data for basin characterization and model parameterization. Soil texture and depth, hydraulic conductivity, soil water holding capacity, maximum impervious area, and runoff curve number were estimated from the Digital Soil Map of the World produced by the U.N. Food and Agriculture Organization (FAO) and the U.N. Educational, Scientific and Cultural Organization (UNESCO) (FAO, 1971-1981). Both land cover and soil data are used in GeoSFM to derive runoff curve numbers. These numbers are used to determine the amount of incident precipitation that becomes surface runoff (Asante et al., 2008).

The UEB requires canopy height, canopy coverage, leaf area index and canopy structure parameters in vegetated areas. A look up table is used to assign each of these for each grid cell based on the IGBP MODIS land cover classes, and parameter values are written to netCDF files for each parameter. The majority of the Langtang Khola watershed is barren or glacier. Altitudes are too high for significant vegetation except in some of the lowest valleys where for 1.6% of the watershed MODIS landcover was classified as mixed forest (Figure 3).

3.2 UEBGrid melt model

Snow and glacier ice melt was computed in this study using an enhanced version of the Utah Energy Balance (UEB) snowmelt model to which the capability to compute glacier melt has been added. UEB is a parsimonious, physically-based model that can be driven

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1 by readily available inputs and applied with no (or minimal) calibration. UEB was
2 initially developed for the prediction of snowmelt rates that produce stream and river
3 flows during the spring and summer (Tarboton et al., 1995; Tarboton, 1994) and
4 evaluated at locations in California, Idaho, Utah and Colorado (Luce, 2000; Luce and
5 Tarboton, 2004; Luce and Tarboton, 2010; Luce et al., 1998; Tarboton et al., 2000; You,
6 2004). In its initial form, the UEB model used a lumped representation of the snowpack
7 with two primary state variables: snow water equivalent and energy content relative to a
8 reference state of water in the ice phase at or below freezing. This energy content was
9 used to determine snowpack average temperature or liquid fraction. Snow surface age
10 was retained as a third state variable, and used for the calculation of surface albedo
11 (Tarboton and Luce, 1996).

12
13 A vegetation component was developed for UEB to enable the evaluation of snowmelt in
14 forested areas (Mahat and Tarboton, 2012; Mahat and Tarboton, 2013; Mahat et al.,
15 2013). The vegetation components were tested at the TW Daniel Experimental Forest
16 (TWDEF), located about 30 miles North–East of Logan, Utah. With the addition of the
17 vegetation component, a state variable quantifying intercepted snow water equivalent was
18 added. The vegetation component describes physical processes of snow-vegetation-
19 atmosphere interactions including parameterizations for the representation of
20 transmission and attenuation of radiation through a forest canopy, precipitation
21 interception and unloading, snowmelt and sublimation of intercepted snow, and turbulent
22 energy exchanges between the ground surface, canopy, and atmosphere in the initial
23 model. The addition of this component in the model has enhanced the models' physically

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1 based capability for modeling snow accumulation, melt, interception, sublimation, and
2 unloading in a forested environment. UEB has been used in various hydrological studies
3 including estimating snowmelt and sublimation in the high Atlas mountains in Morocco
4 (Schulz and De Jong, 2004), analyzing potential climate change impacts in
5 Sacramento/San Joaquin watershed (Knowles and Cayan, 2002) in the Western US, and
6 assessing the surface meteorological variables most critical for snowmelt (Raleigh et al.,
7 2008).

8
9 Glacier melt is driven by the balance of energy at the interface between glacier and
10 atmosphere, which is controlled by meteorological conditions (temperature and radiation)
11 above the glacier and the physical properties of the glacier itself. On one side, the
12 atmosphere supplies energy for melt, and on the other side, the glacier surface influences
13 air temperature due to snow/glacier properties and their variability, mainly the albedo
14 effect (Hock, 2005). The relationship between the melt rate and short- and long-term
15 mean temperature provides the basis for degree-day models, widely used to determine
16 glacier melt in data-scarce areas of the world (Hock, 2003; Immerzeel et al., 2009;
17 Immerzeel et al., 2012; Kayastha et al., 1999; Kayastha et al., 2000; Singh et al., 2000a;
18 Takeuchi et al., 2000). While simple degree-day models are useful for estimating melt
19 based on temperature only, they have a limitation in that they do not take into account
20 topographic effects (slope, aspect, and shading) that influence melt rates on a glacier.

21 Energy balance models overcome these limitations. UEB was chosen for this study
22 because it is a relatively simple energy balance model that parameterizes the snowpack
23 using lumped (depth averaged) state variables so as to avoid having to model the complex

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processes that occur within a snowpack, while using a modified force-restore parameterization to capture physical differences between bulk (depth averaged) properties and the surface properties, which are important for calculating surface energy exchanges.

We extended the representations of surface energy balance fluxes in UEB to include the capability to quantify glacier melt on the basis of substrate type. Substrate is represented as one of: 0- Ground/non-glacier, 1-Clean glacier ice, 2-Debris covered glacier ice and 3- Glacier accumulation zone (snow). The amount of snow/ice melt is determined as follows: In the case of bare ground/non-glacier substrate type, the model tracks seasonal snow accumulation and ablation. In the glacier accumulation zone (snow surface) above the equilibrium line altitude, no melt is generated, as all precipitation is presumed to add to glacier accumulation. In the glacier ablation zone (comprised of clean or debris-covered ice), snow may accumulate on the glacier surface and then melt during the melting season. The model tracks seasonal snow accumulation and ablation in this area of the glacier in form of snow water equivalent. At each time step, the model computes the snowmelt, which is referred to as surface water input from snow melt (SWISM) until the seasonal snow on the ablation zone completely disappears. When seasonal snow water equivalent reaches zero, the surface energy balance is used to calculate the amount of glacier ice melt, which then becomes a component of the surface water input. The melted glacier ice is referred to as surface water input from glacier melt (SWIGM). Glacier ice melt is generated at the ice substrate only once seasonal snow covering glacier ice has melted. In addition, rain may occur both on bare ground as well as the glacierized parts of

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the watersheds. Surface water generated from rainfall is referred to as surface water input from rain (SWIR). The difference in functionality between debris covered and clean glacier ice surface is due only to the substrate albedo which is provided as an input.

This parameterization of glacier melt provides a simple, yet practical way to quantify energy balance driven glacier melt given the information available. It neglects a number of physical processes for which there is limited information. Debris cover on glaciers, influences the melt rates in two ways: a thick debris cover ($>$ a few centimeters, or “critical thickness”) reduces the ablation rates of the ice underneath due to the low thermal conductivity of debris (Foster et al., 2012; Mihalcea et al., 2008), whereas a thin debris cover ($<$ a few centimeters) accelerates the ice melt rates due to the lower albedo of the supra-glacial debris compared to clean ice (Kayastha et al., 2000; Singh et al., 2000b). The “critical thickness” is the thickness above which ice melt is substantially reduced due to insulation of the supraglacial debris (Brock et al., 2010). Parameterizing melt under the debris cover is therefore difficult due to lack of debris cover thickness measurements, and thus the modeling approach used here does not consider melt under the debris cover. When seasonal snow has melted over debris covered glaciers, melt is generated due to the debris covered glacier albedo, which is generally significantly lower than that of clean glacier ice- which in general would result in larger energy inputs and higher melt rates. Thus, we estimate this approach to work best for glacier ablation areas which are covered by a thin debris cover (less than the critical thickness), where melt is governed by albedo.

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For this study, the UEB model was reconfigured to run on a distributed grid to explicitly represent spatial variability in the inputs across a basin, and the enhanced model is referred to in this paper as UEBGrid. The new gridded version of the UEB model facilitates coupling with EPA BASINS and the forcing by inputs from NASA remote sensing and earth science data products such as, satellite data (MODIS, ASTER, SRTM), reanalysis data (MERRA) and climate model output (RFE2). UEBGrid was integrated into the HIMALA BASINS software to facilitate the linking to other models, such as GeoSFM, and to take advantage of BASINS' capability to manage input data and visualize results. UEBGrid has adopted a structured file-based input/output format using text and NetCDF files to facilitate its use and incorporation into the HIMALA BASINS software. The UEB model has never before had a graphical user interface and its incorporation into BASINS here has provided graphical user interface capability for UEB thereby making it accessible to a broader group of users.

In UEBGrid, a watershed is divided into a mesh of grid cells and the model runs individually for each grid cell and computes snow or glacier melt. Outflow is then aggregated over sub-basins derived from a Digital Elevation Model (DEM) and used as input into a hydrologic model (in this case, GeoSFM described below, section 3.3).

3.3 GeoSFM streamflow model

The Geospatial Stream Flow Model (GeoSFM), originally developed to work within ESRI's ArcView 3.x software, is a streamflow model developed at the U.S. Geological Survey for the monitoring of hydrologic conditions and the identification of streamflow

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anomalies (Artan et al., 2007). The monitoring activities include topographic analysis, data assimilation, and time series processing and analysis. In data-sparse regions of the world, GeoSFM is designed to use remotely sensed meteorological data, many of which are in raster format, requiring the adoption of a customizable geographic information system with raster functionality (Artan et al., 2007). Input data consist of elevation, topography, land cover, and soil information to derive and parameterize the sub-basins. Forcing data for the model includes daily estimates of precipitation and potential evapotranspiration to predict daily streamflow at in-situ gauge stations. GeoSFM consists of four modules- preprocessing, hydrologic analysis, parameter calibration, and post-processing. The components of the GeoSFM that are easily accessed through a series of tabs in HIMALA BASINS are: *Terrain Analysis*, *Basin Characteristics*, *Basin Response*, *Rain/Evap Data*, *Compute Soil Water Balance*, *Compute Stream Flow*, *Sensitivity Analysis*, *Model Calibration*, and *Output Results*. HIMALA BASINS allows access to the other hydrological models currently supported by BASINS, along with extending some of the pre-processed datasets to a global extent to enable users outside of the United States to benefit from the modeling framework. To find out more about BASINS, visit <http://water.epa.gov>.

4. Results and discussion: Langtang Khola case study

4.1 Model set-up

The integrated HIMALA BASINS model was run for the Langtang Khola watershed in Nepal Himalaya (Figure 1). Terrain analysis was performed to delineate the sub-basins and flow layers based on the SRTM elevation data using the built-in MapWindow

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TauDEM tools (Tarboton and Ames, 2001). The Langtang Khola watershed was divided into 24 sub-basins based on hydrologic tools and the SRTM DEM. Figure 4 indicates the glacier subtype for each sub-basin, with a detailed insets for sub-basins 3, which is an upper glacierized basin containing Langtang glacier. Basin characterization and hydrograph response steps follow to account for the soil and land cover types that fall within each sub-basin and influence the hydrology of that corresponding area. The process continues to the hydrologic analysis stage, including the selection of precipitation/melt and potential evapotranspiration (PET) time series data assigned to the corresponding sub-basin based on sub-basin ID. In this case, the UEBGrid-derived SWIT (total outflow) was selected as the forcing dataset and the PET was automatically estimated using downscaled MERRA air temperature. The soil water balance is then computed using either a linear or a non-linear soil model. For the Langtang case study, the non-linear option was used since it is more suitable for high spatial or temporal resolution analysis and when a model needs to be well calibrated with observed data. Once this is complete, streamflow is computed using one of three routing options (Simple Lag, Diffusion Analog, and Muskingum Cunge). The Muskingum Cunge method was used in this study since it has more parameters and it can diffuse the hydrograph. HIMALA BASINS provides the user not only the option to run both UEBGrid and GeoSFM separately, but to run both models seamlessly in an integrated fashion using the UEBGrid melt output as input for modeling streamflow in GeoSFM.

The GeoSFM-simulated streamflow was estimated using eight years of data, October 1, 2003 – September 30, 2010 hydrologic year, and was compared to observed gauge stream

DRAFT

flow data from Kyangjing station (3,920 m) provided by the Nepal Department of Hydrology and Meteorology (DHM), Nepal. Once the calibrated streamflow run was complete, the post-processing module steps were implemented through the calculation of flow statistics, and the creation of flow maps and hydrographs.

4.2 Model Results

Modeled results were mapped within HIMALA BASINS depicting status of streamflow and soil water conditions. ASTER-derived substrate type (glacier or ground) and albedo are shown in Figure 3. The substrate albedo over the clean glacier parts ranged from 0.17 to 0.87 with an average of 0.56, while the substrate albedo over the debris covered glacier parts ranged from 0.15 to 0.71, with an average of 0.26.

In Figures 5a, we illustrate the UEBGrid derived outflow components: glacier melt, snowmelt and rainfall, as well as GeoSFM simulated streamflow for test sub-basin 3. Figure 5a illustrates patterns of glacier and snowmelt occurring in sub-basin 3, a highly glacierized upper sub-basin at the headwaters of Langtang glacier. We note that periods of snowmelt occur throughout the year in this region of the Himalaya, whereas glacier melt occurs from May to September, with low flow from October to April. In Figure 5b, hydrologic components aggregated from all of the sub-basins are shown for the 2004 water year. In Table 3, we present hydrologic components aggregated from all of the sub-basins and averaged across the eight years of simulation (2003-2010).

DRAFT

1 The annual average aggregate surface water input of 1.18 m/year (Table 3) originated
 2 30% from snow melt (SWISM), 62% from glacier melt (SWIGM - both debris covered
 3 and clean ice) and 8% from rain on bare ground or glacier (SWIR). Under the
 4 assumption that the conversion from surface water input to streamflow retains the same
 5 proportions, this is in relative agreement with other recent estimates based on a degree-
 6 day or ablation model (Racoviteanu et al, 2013; Immerzeel et al, 2012). Racoviteanu et al.
 7 (2013) reported 58.3% of streamflow measured at Kyangjing to glacier snow and ice melt,
 8 using a simple ablation model. UEB results are also in close agreement with another
 9 recent study (Pradhanga et al., 2014), which estimated a contribution of snow and ice
 10 melt of 54.3% for the period 1993–2006 in the same basin, using a degree-day model. It
 11 is notable that while only 8% of the total watershed is covered by supra-glacial debris, ,
 12 in our model debris-covered ice contributed about 52% of total surface water, with this
 13 52% originating 47% as glacier melt, 1.4% as rain on the glacier surface and 3.6% from
 14 melting of snow on the debris covered surface. This is in contrast with results from
 15 Racoviteanu et al (2013), which obtained a smaller contribution of debris-covered glacier
 16 ice (17.7%) compared to clean ice (40.6%) in Langtang Khola. Our study shows a
 17 significant excess of meltwater from debris-covered areas, which we speculate it is due to
 18 the occurrence of debris covered glacier tongues at low elevations, and particularly due to
 19 the low albedo of the debris covered surfaces. This has been recently confirmed in a
 20 study by Fujita and Sakai (2014), which also obtained a large amount of meltwater from
 21 debris covered areas in Nepal Himalaya (Tso Rolpa glacier). A sensitivity analysis
 22 showed that the excess of meltwater in their study was indeed generated at lower
 23 elevations over debris covered tongues, and was accelerated due to low albedo. Our

DRAFT

1 results are consistent with Fujita and Sakai (2014). We also note that patterns on glacier
2 melt of debris-covered tongues are highly variable from one area to the other, depending
3 on debris thickness and albedo. As discussed in section 3.2, in this study we do not take
4 into consideration melt under the debris cover, and the potential insulating effect of the
5 debris cover, which would reduce the ice melt, as noted in other studies (Mihalcea et al,
6 2008, Brock et al, 2010, Fujita and Sakai, 2014).

7
8 In spite of these limitations, we note on Figure 6 that aggregated simulated discharge
9 from GeoSFM (5.31 m) is in agreement with measured discharge (5.02 m), indicating
10 that the combined models produce a reasonable aggregate water balance. The runoff ratio,
11 based on measured discharge and modeled surface water inputs (5.02/9.43) is 0.53 with
12 the remaining surface water input being the combination of loss to evaporation and
13 change in storage, or due to errors in the surface water input. Furthermore, a comparison
14 between simulated streamflow and observed streamflow for the entire Langtang Khola
15 watershed (Figure 6) shows that overall, the model captures the hydrologic pattern in this
16 area of the Himalaya, with low flow during the winter months (November to March), and
17 high flow during the monsoon months (June to September). Both the observed and
18 simulated streamflow curves show periods of low and high flows; however, the simulated
19 runs are generally higher than the observed ones, and produce some spikes of high flow
20 during the monsoon, that are not present in the observed data. These overestimates may
21 be due to inaccuracies in the input climate data or problems with the UEB or GeoSFM
22 model, or a combination of these. The largest likely source of error is the climate data,
23 which is entirely derived from satellite and modeled data products.

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1

2 **5. Summary and Conclusions**

3 In this study, we presented a modeling system (HIMALA BASINS) developed within the
4 framework of HIMALA project to better understand the current contribution of snow and
5 ice melt to streamflow and enable an assessment of the impact of changing climate in the
6 high Himalaya. The UEB snowmelt model was extended to include a capability to
7 simulate glacier melt. It was reconfigured to run on a grid for integration into BASINS.
8 The UEBGrid model and GeoSFM have been added to the BASINS toolset and coupled
9 to estimate the contributions of glacier, snow melt and rain to streamflow in a seamless
10 fashion.

11

12 This integrated modeling system was demonstrated using a case study in the Langtang
13 Khola watershed, where model inputs were taken entirely from downscaled remote
14 sensing products available over a remote south Asian region where meteorological
15 observations are scarce. The results indicate the high fraction of contribution to water
16 input from glacier melt (62%) even though this is a small fraction of the watershed area.
17 The discharge from GeoSFM driven by UEBGrid inputs compares favorably to the total
18 discharge measured. There are discrepancies in the detail of the hydrograph between
19 modeled and measured. However considering the scarcity of data and the modeling
20 system being totally driven by remote sensing and global or regional climate products the
21 comparisons are reasonable.

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1 The augmentation of the GeoSFM model with a capability to capture water melt from
2 snow and glacier ice is important for ICIMOD and its member countries, since water
3 stored during the winter as ice and snow is a significant contributor to rivers during the
4 spring and summer in many basins in South Asia (Immerzeel et al., 2009; Immerzeel et
5 al., 2010). Scarcity of data for meteorological parameters necessary to run hydrological
6 models is a problem in regions at risk of natural hazards such as landslides, floods,
7 droughts and food insecurity (Artan et al., 2007; Brown and Funk, 2008; Sanyal and Lu,
8 2004). If downstream communities are not aware of excessive temperatures causing
9 snow and ice melt, or of extreme precipitation events, then they will be poorly prepared
10 for changes in river flow after the fact. Using satellite-derived datasets to drive
11 hydrological models in regions without universally available ground observations helps
12 overcome this problem.

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1 Agency for International Development.

2

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1 **Table Captions**

2 Table 1. Input data sources for BASINS UEB and GeoSFM models

3 Table 2. UEBgrid model inputs, outputs, and state variables. The inputs include static
4 distributed parameters and dynamic meteorological data.

5 Table 3. Langtang Khola watershed hydrologic components estimated by the UEB model
6 averaged for 2003-2010.

7 **Figure Captions**

8 **Figure 1.** Langtang Khola watershed in the Koshi in Nepal where HIMALA BASINS
9 test data was developed and evaluated.

10 **Figure 2.** General modeling framework of HIMALA BASINS system, integrating UEB
11 with GeoSFM.

12 **Figure 3.** Downscaled (90-m) UEBGrid model input: MERRA climatologic variables for
13 a single 3-hour time step; RFE2 precipitation for a single 3-hour time step; ASTER-
14 derived glacier subtype and albedo; CGIAR SRTM elevation data.

15 **Figure 4.** A map of Langtang Khola showing glacier subtype characteristics for each
16 subbasin with an inset for subbasins 3, an upper glacierized basin containing Langtang
17 glacier. The Kyangjing station is labeled as well to indicate the stream flow gauge
18 location used during GeoSFM model calibration.

19 **Figure 5. (a)** Hydrograph produced in HIMALA BASINS depicting the breakdown of
20 UEBGrid-derived contributions for Langtang Khola glacierized subbasin 3 for 2004
21 water year. Purple lines depict glacier-melt (SWIGM); green lines depict snowmelt
22 (SWISM), and red lines depict surface water input due to rainfall (SWIR). **(b)**

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1 Hydrographs produced in HIMALA BASINS depicting the cumulative aggregate of
2 UEBGrid-derived surface water contributions for all Langtang Khola subbasins in the
3 form of glacier-melt, snowmelt, and rainfall. Simulated streamflow from GeoSFM is
4 presented on the upper portion of the graph for both figures.

5 **Figure 6.** Hydrograph from 2003 to 2010 produced in HIMALA BASINS comparing
6 GeoSFM-derived simulated streamflow with observed (gauge) data collected at the
7 Kyangjing station indicated in Figure 4.

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1 Table 1.

Data	Type/Resolution	Source	Frequency	Period
Satellite Data Products				
MODIS Land cover	MOD12Q1 500m	NASA	Annual	2000 to current
Precipitation (RFE-2)	Satellite/ground 0.1°x0.1°	NOAA	Daily	2001 to present
Leaf Area Index	MOD15, 1km	NASA	Daily	2000 to current
Percent Canopy Cover	MOD15, 500m	NASA	Daily	2000 to current
Glaciers				
Glacier Outlines 2000	ASTER and Landsat 30m	ICIMOD Calculated by team	Decadal	2000
Debris versus Ice information				
Hydrological Parameters				
Digital elevation Model	SRTM, 90m	USGS	One time	2000
Soil Type	1 to 5 million resolution	FAO	One time	1960
Evapotranspiration	Calculated from MERRA temperature data	NASA	Daily	1979 to current
Modeled data from GEOS-5 (MERRA)				
Air temperature	0.5° x 0.66° resolution	NASA GMAO	Hourly	1979 to current
Horizontal wind speed,	0.5° x 0.66°	NASA GMAO	Hourly	1979 to current
Relative humidity	0.5° x 0.66°	NASA GMAO	Hourly	1979 to current
Short-wave radiation	1° x 1.25°	NASA GMAO	Hourly	1979 to current
Long-wave radiation	1° x 1.25°	NASA GMAO	Hourly	1979 to current
In situ				
Stream flow data at basin outlet	Langtang Khola observations	Nepal Meteorological Agency	Daily	2001 to current

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Table 2:

Dynamic Inputs	Static Inputs	Output Fluxes	State Variables
Incoming shortwave radiation	Elevation	Latent heat flux	Snow energy content
Incoming longwave radiation	Vegetation cover	Sensible heat flux	Snow water content
Air temperature	Vegetation height	Ground heat flux	Snow age
Average wind speed	Soil bulk density	Snow temperature	
Precipitation		Melt advected energy	
Relative humidity		Melt outflow flux	
Atmospheric pressure			

Table 3

Component	meters/yr
Surface water input from rain SWIR	0.09
Surface water input from snow melt SWISM	0.35
Surface water input from glacier melt SWIGM	0.74
Total surface water input SWIT	1.18
Spatially averaged snow water equivalent SWE accumulation	0.18
Precipitation	0.80
Measured streamflow	0.63
Simulated streamflow	0.66
Sublimation	0.20

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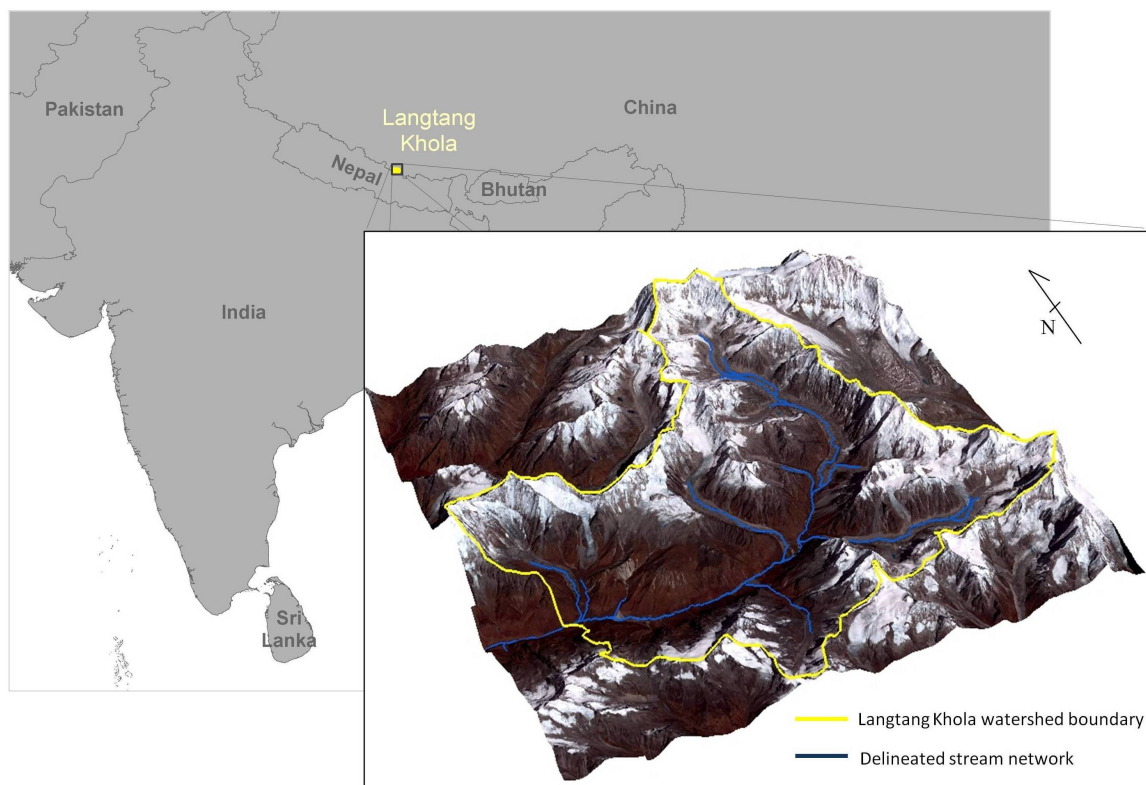


Fig. 1

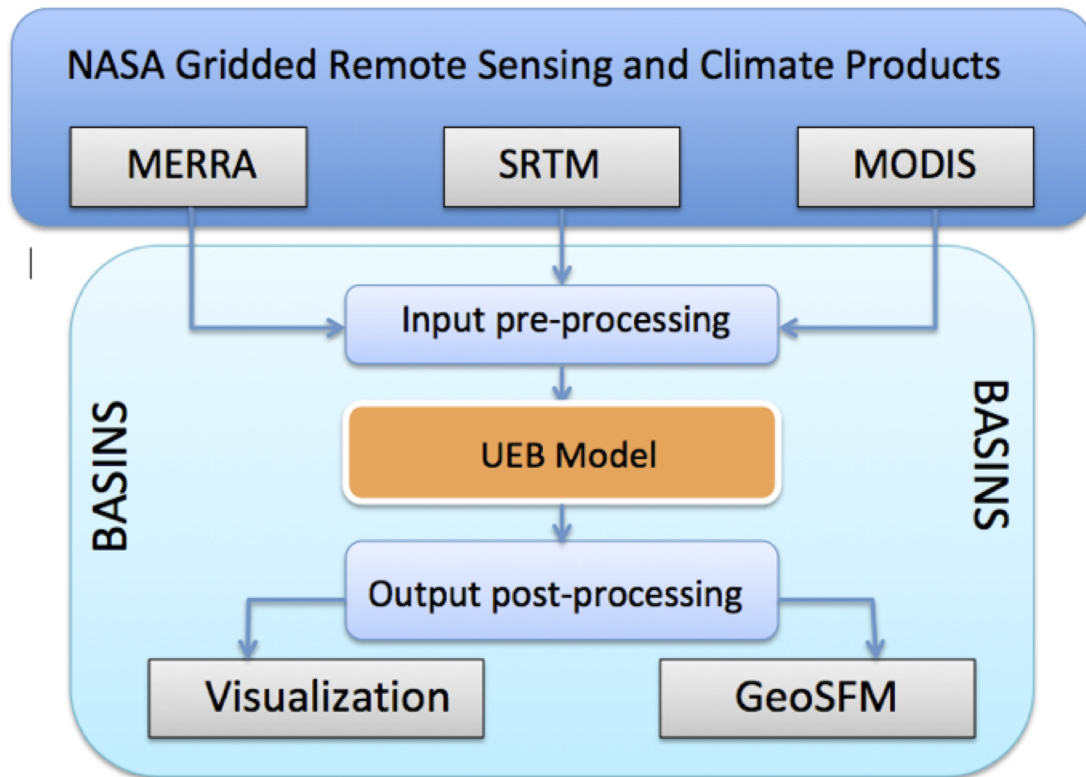
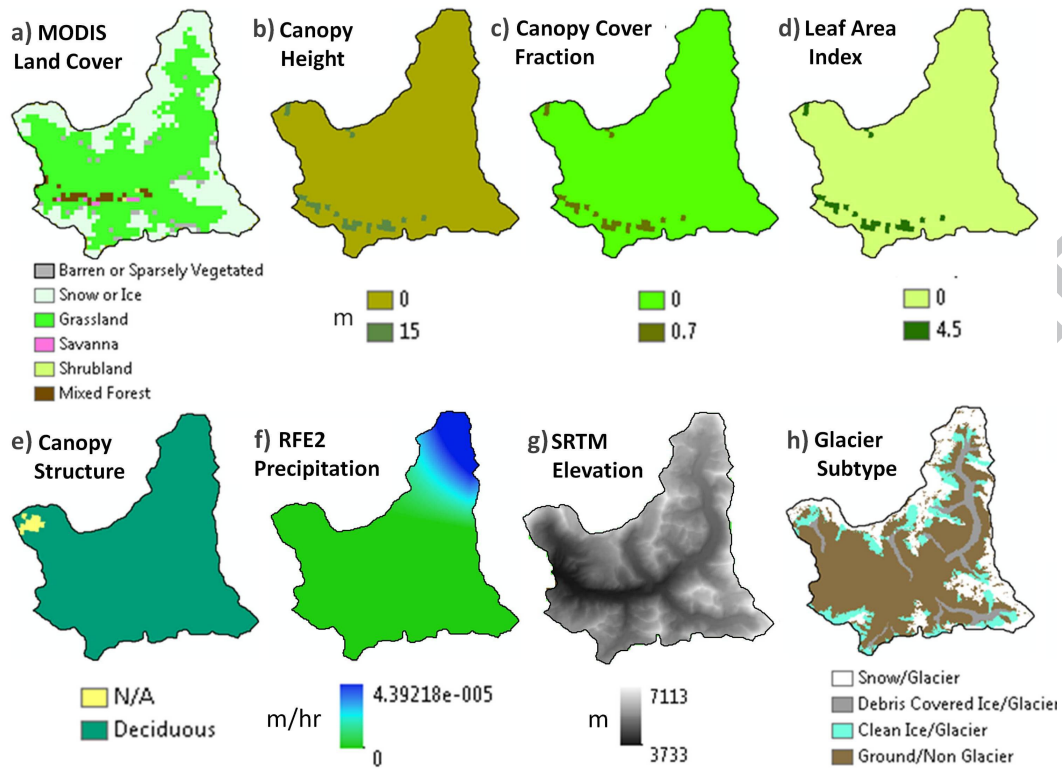


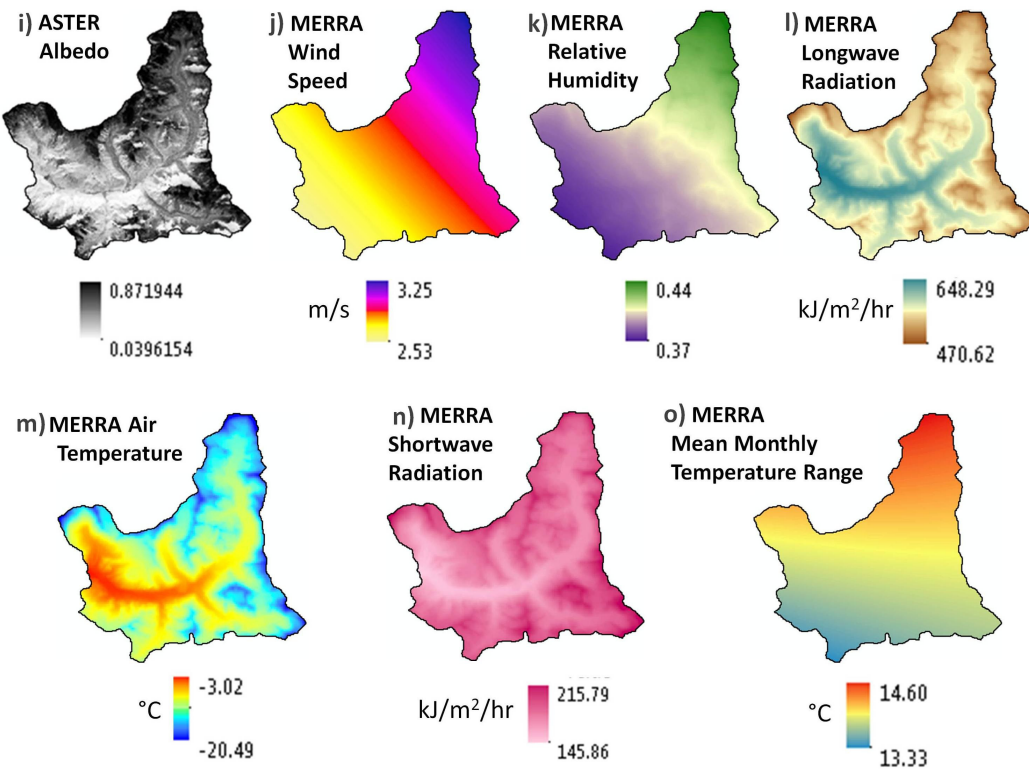
Fig. 2

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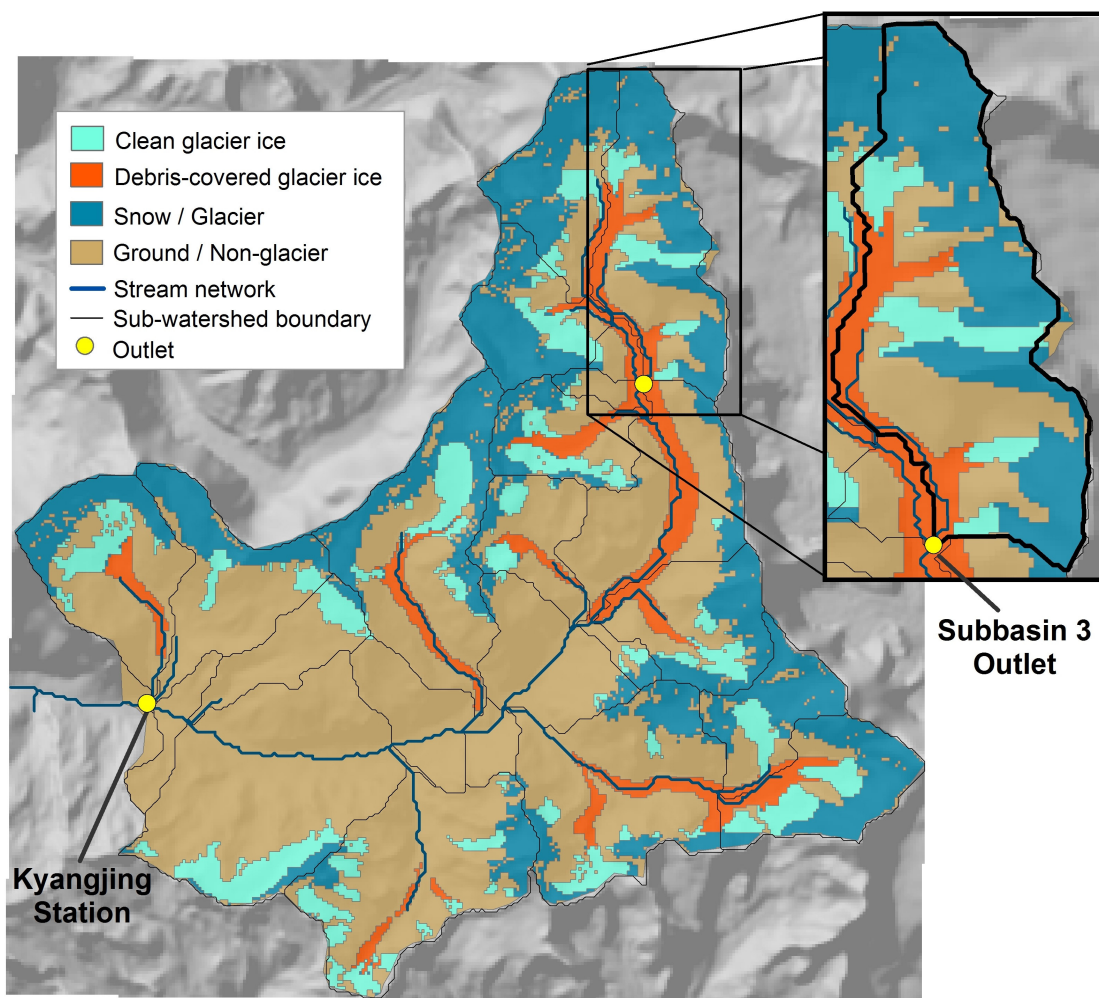
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2



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1

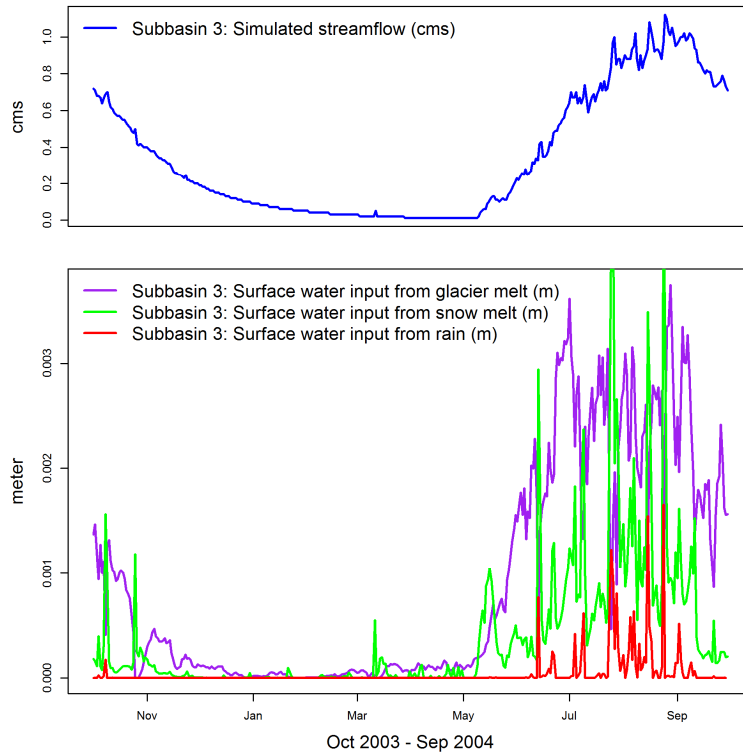
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3 Fig. 4

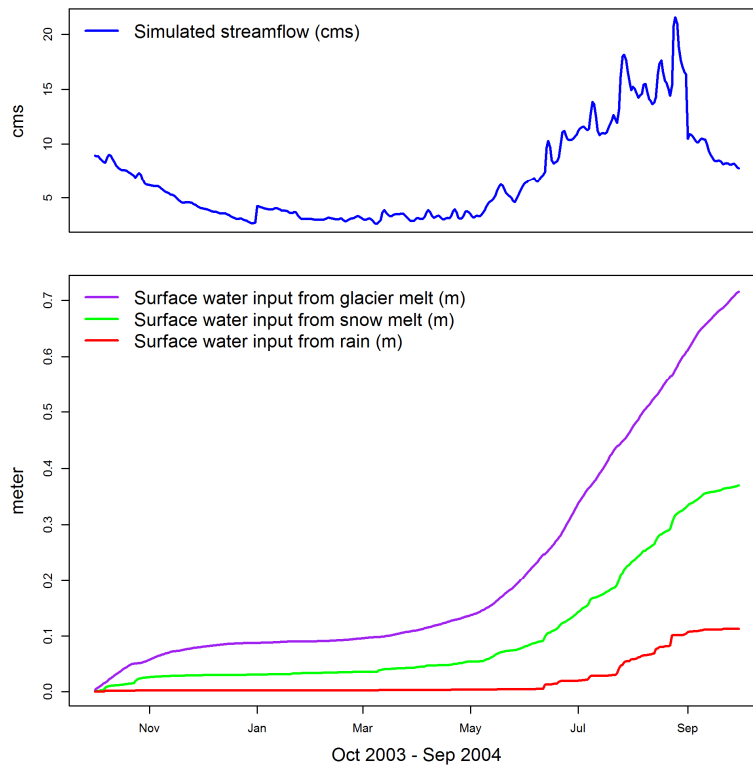
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5

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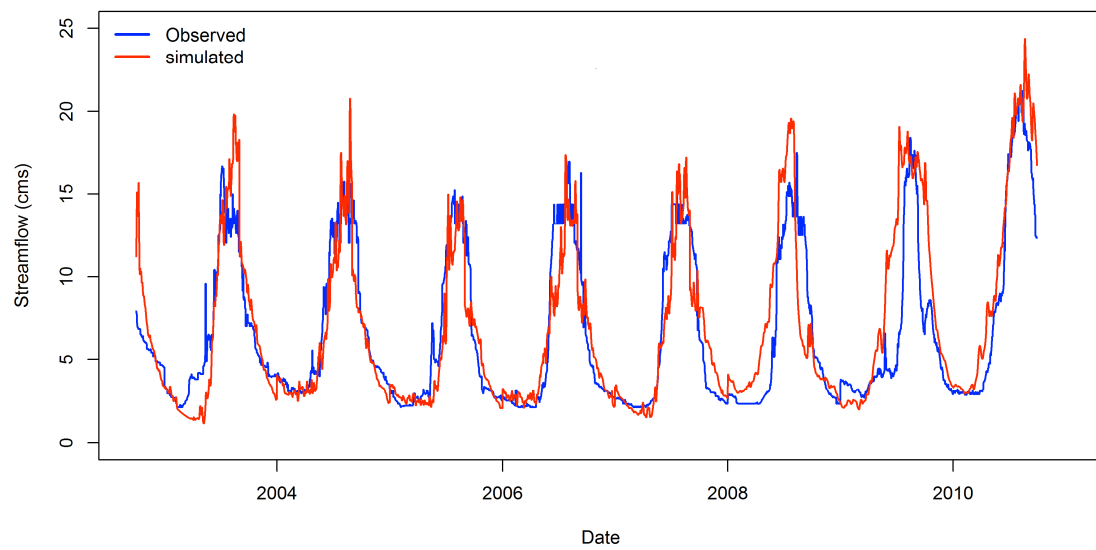
1



2

3 Fig. 5

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1

2

3 Fig. 6

4

5

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Highlights

- Integration of snow, glacier and hydrologic modeling in a data sparse region
- Modeling the contribution of snow and glacier melt to streamflow from Himalaya basins
- Inputs derived primarily from NASA remote sensing and climate data products
- Open source energy balance snow and glacier melt model now part of EPA BASINS