Simulating stream flow over data sparse areas – an application of internet based data

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9 Abstract

10 Many hydrological modeling studies suffer from lack of robust station observed data, mainly rainfall and discharge. Where such a dearth of data exists, detailed modeling studies in 11 12 estimating and assessing change in water resources become difficult when models cannot be 13 compared against recorded observations. In addition, some river basins exist along trans-14 boundaries of two or more countries that problems in data sharing among them add to the 15 difficulties in such modeling studies. Nevertheless, with the advancement in the global internet 16 resources, access to such data has become easy. Whether such internet based data are good 17 substitutes for station data can be ascertained only after performing some modeling research. To 18 this end, this paper describes a hydrological modeling study that simulates the river flow of the 19 Da River across the trans-boundary regions of China and Vietnam over a 11 year period from 20 1971 to 1982. Globally available observation data used in this study include topography (from 21 SRTM - Shuttle Radar Topography Mission), land use (from GLCC - Global Land Cover 22 Characterization), soil (from FAO - Food and Agriculture Organization), precipitation (from 23 APHRODITE – Asian Precipitation Highly Resolved Observational Data Integration Towards 24 the Evaluation) and temperature (from GHCN2 – modified of Global Historical Climatology 25 Network version 2). The study employs a hydrological model to recreate the natural flow without 26 dam(s) built across the main river channel. The results of the study are promising and provide a wide scope to utilize internet based data for further research. This also has implications in the 27 28 context of climate change applications.

1 **1** Introduction

2 Water resources management is very crucial in many countries and draws an added importance 3 near the border regions between countries. In the world, international river basins that include political boundaries of two or more countries cover 45.3% of Earth's land surface, host about 4 5 40% of the world's population and account for approximately 60% of global river flow (Wolf et 6 al., 2005). Due to different governmental policies, conflicts arise in sharing of the water 7 resources, more so when the advantage is more for the upstream user country of these water 8 resources. Research by Lu and Siew (2006) showed that the series of dam constructions in China 9 are affecting water discharge and sediment flux over Lower Mekong River over the last decades. 10 The problem of water resource management comes into picture when there are no data available 11 due to poorly managed observing stations, lack of technology and resources, war time and 12 financial limitations. Data availability is also an issue when data sharing is difficult between 13 countries that are not into any formal agreement. For example, when availability of water 14 resources needed to be assessed over northern Vietnam, the data for upstream region which lies 15 over the Southern part of China are not available, as the water quantity over the downstream 16 region over Vietnam depends on the flow from the upstream China part. This is a clear case of a 17 trans-boundary problem, the issue cited earlier. Hence, this paper describes an approach to 18 resolve data requirement issues of a trans-boundary nature in managing water resources by 19 employing a hydrological model, the Soil and Water Assessment Tool (SWAT) that uses data 20 available from the internet with implications for climate change applications.

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22 Many research studies that focus on basin hydrology have used the SWAT model to simulate 23 runoff (Easton et al., 2010, Ouessar et al., 2009, Mengistu and Sorteberg, 2011, Pohlert et al., 24 2007, Cau and Paniconi, 2007, Stehr et al., 2010). Some SWAT employed studies have been 25 done in Southeast Asian (SEA) region. Victor (2009) mentioned about the potential application 26 of SWAT model for countries in SEA. The Mekong River Commission (MRC) also used the 27 SWAT model in their Digital Support Framework (DSF) for LMB Planning (John, 2008). A 28 study over the Da River catchment (Vietnam part) has been done by Nguyen et al. (2010), in 29 which they used local data to simulate soil erosion. However, SWAT is a physical based model 30 that requires the availability of spatial data like topography, land use and soil map with

1 meteorological data (precipitation and temperature) which are difficult to obtain from local 2 authorities of many countries. The use of internet based data into SWAT model was first 3 introduced by Van Griensven et al. (2007) for 3 international river basins: river Kagera (Rwanda, 4 Burundi, Uganda and Tanzania), river Blue Nile (Ethiopia) and river Ganges (India, Bangladesh, 5 Nepal) using monthly weather data to quantify the monthly river flow. Later, Rouholahnejad and 6 Abbaspour (2010) applied this approach over the Black Sea Catchment which contains 19 7 European and Asian countries, to quantify hydrological components of water resources (surface 8 runoff, deep aquifer recharge, soil water and actual evapotranspiration) on a monthly time scale. 9 In a recent study, Betri et al. (2011) applied internet based spatial data and local meteorological 10 data (rainfall and temperature) to study the sediment transport over the Blue Nile Basin on a 11 daily time scale.

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In this context, the main contribution of this paper is that it uses the spatial data on a daily time scale using available meteorological data from the internet in a first-of-its-kind study to be done over this Da River trans-boundary region, which gives the scope to be self-sufficient in data within a region to overcome dependency on other sources from which data procurement is difficult. In terms of climate change adaptations, this methodology proves useful to assess future hydrological responses to devise adaptive measures suitably.

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20 2 Model and Study Region

21 2.1 Study region

22 The Da River (known as 'Black' river in English or rivière Noire in French) originates from the 23 Yunnan Province in China and flows downstream through mountainous regions, crosses the 24 border of China-Vietnam and joins as a tributary of the Red River in Vietnam. The Da River has a total catchment area of about 53,000 km² in which 48% of the area lies on China's territory, 25 2% in Lao PDR and 50% in Vietnam (Figure 1a). It spans a total length of 1010 km of which 26 27 570 km lies in Vietnam and a population of 1.3 million people make this region their dwelling. 28 The Da River basin is formed by the ranges of high mountains in the region (Hoang Lien Son 29 mountainous area). The total precipitation ranges between 1000 to 2300 mm per annum. 30 Precipitation pattern is high over the central part of the catchment due to Hoang Lien Son

1 mountain chain and reduced to upper and lower catchment (Figure 1b). The Da River has a high annual average discharge of 1770 m³/s. During the main flood season (June to October), the total 2 3 discharge occupies around 78% of the annual discharge. Flood occurs strongly because of very 4 high rainfall concentrations over the steep topography and narrow valleys thus leading to very high peak flood of about 5000 m^3 /s. Therefore, it has high potential for hydropower over this 5 basin. There have been two huge dams built on the main river: Hoa Binh (installed capacity 1920) 6 7 MW, in operation since 1989) at the downstream part and Son La (installed capacity of 2400 8 MW, in operation since 2006) at the upstream part, in which the latter is currently the largest 9 dam in Southeast Asia (Hydroelectric power plants, 2011). Land cover of this region reduces 10 rapidly from forest to cropland, mostly because of people indulging in forest burning for 11 agriculture. The land use data collected from Global Land Cover (2000) is recent and it shows 12 that about 70% of the area is cropland whilst another 25% is forest (Figure 1c). This is the most recent version of the database available until now, which can be accessed at: 13 14 http://landcover.usgs.gov/glcc/download.php. There are two main types of soil for this regions 15 which are Ferric (Af) and Orthic (Ao) from Acrisols group, covering almost 90% of the area (Figure 1d). Due to its importance in the Vietnam hydropower system, Da River is always listed 16 17 as the first in hydropower generation. The need for modeling comes in the backdrop of main 18 issues which are: how to manage the water sources coming from China and what will be the 19 impact if there are dams constructed in China that may affect the quantity of water flowing into 20 Vietnam.

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22 2.2 SWAT model

23 SWAT is a river basin scale model, developed by the United States Department of Agriculture 24 (USDA) - Agriculture Research Service (ARS) in early 1990s. It is designated to work for a 25 large river basin over a long period of time. Its purpose is to quantify the impact of land 26 management practices on water, sediment and agriculture chemical yields with varying soil, land 27 use and management conditions. Detailed information and several related publications are available at http://swatmodel.tamu.edu. SWAT version 2005 with an ArcGIS user interface is 28 29 used in this paper. There are two methods for estimating surface runoff in the SWAT model: 30 Green & Ampt infiltration method (Green and Ampt, 1911) and the Soil Conservation Service

1 (SCS) curve number procedure (SCS Handbook, 1972) in which the latter was selected for the 2 model simulation. Retention parameter is very important in SCS method and it is defined by 3 Curve Number (CN) which is a sensitive function of the soil's permeability, land use and 4 antecedent soil water conditions. Potential evapotranspiration (PET) may be defined as the 5 evapotranspiration from a large vegetation covered land surface having adequate moisture at all times. SWAT model offers three options for estimating PET: Hargreaves (Hargreaves et al., 6 7 1985), Priestley-Taylor (Priestley and Taylor, 1972) and Penman-Monteith (Monteith, 1965). It 8 depends on the amount of required inputs that each model is preferred. While Hargreaves 9 method requires only maximum, minimum and average air temperature, the Priestley-Taylor 10 method needs solar radiation, air temperature and relative humidity. The inputs for the Penman-11 Monteith method are the same as that of Priestley-Taylor, in addition requiring the wind speed. 12 Due to limitations in the available meteorological data, the Hargreaves method is applied in this 13 study. In the SWAT model, the land area in a sub-basin is divided into what are called Hydrological Response Units (HRUs). In other words, a HRU is the smallest portion that 14 15 combines different land use and soil type by overlaying their spatial map. All processes such as 16 surface runoff, PET, lateral flow, percolation, soil erosion, nitrogen and phosphorous are usually 17 carried out for each HRU (Arnold and Fohrer, 2005).

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19 2.3 Data

20 The main spatial data used in this study are taken from many different sources from the internet.

21 The respective sources have been cited in the references.

22 The Digital Elevation Model (DEM) was taken from NASA (National Aeronautics and Space 23 Administration) SRTM 3 arc second (approx. 90m) where the digital elevation data was obtained 24 on a near global scale to generate the most complete high-resolution digital topographic database 25 of Earth (Farr et al., 2000). The land use map was taken from the Global Land Cover (2000) 26 Products having a spatial resolution of 1 km with 22 distinguished land use classes. The land use 27 for the Asia region as a whole was downloaded and cropped to the study region. Table 2 shows 28 the land use/land cover types and corresponding area percentage in the Da river basin. The land use classes are classified into 6 different categories, in which Agricultural Land-Row crops and 29 30 Forest Evergreen cover majority with respective area percentage of 62.7% and 37%. The soil

1 map was taken from the digital soil map of the world provided by the FAO of the UNESCO 2 (FAO, 2003) at 1:5.000.000 scale, in the Geographic projection (Latitude - Longitude) 3 intersected with a template containing water related features (coastlines, lakes, glaciers and 4 double-lined rivers). There are around 23 soil types with spatial resolution of 10km with soil 5 properties for 2 layers (layer 1: 0-30cm, layer 2: 30-100cm depth). The cropped region for Da river basin includes 6 types as shown in Figure 1d. Rainfall data were taken from the gridded 6 7 rainfall product called the APHRODITE (Yatagai et al., 2009) having a spatial resolution of 8 0.25° (~ 25km) on a daily time scale. This latest high resolution gridded precipitation product is 9 available for a long period from 1957 to 2007 over the Monsoon Asia, Middle East and Russia 10 and this study region is a subset of Monsoon Asia. Not many high resolution daily rainfall 11 products are available and this data is one such which is useful for hydrological simulations as 12 input. This dataset has also been found highly suitable to be used in the place of station data 13 (where there are not much station records available). Some comparisons of this data against other 14 available station data have been studied and described by Vu et al., (2011). It is to be noted that 15 the exact rainfall station locations over the Chinese region are not known and are selected to be 16 closer to the center of the sub-catchment delineated in SWAT, whilst the rainfall station 17 locations in Vietnam are known, being a part of the study region within Vietnam. Daily average, 18 maximum and minimum air temperatures were obtained from the modified Global Historical 19 Climatology Network version 2 (it has been referred to in this paper as GHCN2, for simplicity). 20 This dataset provides historical daily data over global land areas from 1950-2008 with spatial 21 resolution of 0.5°. It is a composite of climate records from numerous sources that were merged 22 and then subjected to a suite of quality assurance reviews. More information can be found in 23 Adam and Lettenmaier (2003).

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Daily river discharge data was the only source obtained from the local authority in Vietnam (Institute of Meteorology, Hydrology and Environment - IMHEN) for two gauging stations from 1971-1990: one gauge at the downstream at Hoa Binh station and the other at the upstream, named Lai Chau. The latter is located within the Vietnam territory and is 120 km away from the border of Vietnam and China which can be used to verify discharge coming from the upper part in China. The use of this observed discharge station data in this paper is to calibrate the model which used internet based input data as described above. A detailed list of input variables and
sources is tabulated in Table 1. The model setup and result is described in the following sections.

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3 Experiment Methodology and Results

5 3.1 Experiment methodology

The trans-boundary region consists of two parts: the upper part belongs to China and the lower
one belongs to Vietnam. As mentioned earlier, the input spatial data (DEM, land use and soil
map) described above is downloaded for the studied region to apply into SWAT model.

9 ASCII data has been downloaded from NASA SRTM 90m for 5 different grid boxes 10 (http://srtm.csi.cgiar.org/SELECTION/inputCoord.asp) (srtm code: 56_07, 56_08, 57_07, 57_08, 11 58_07). These datasets then have been converted to raster format then merged and projected to 12 the coordinate of Vietnam (WGS 1984, UTM zone 48N). ArcHydro model which is served as an 13 extension of ArcGIS has been applied to delineate the Da river catchment based on the 14 downloaded DEM (Figure 2a).

Land use map with spatial resolution of 1000m has been downloaded in raster format (http://bioval.jrc.ec.europa.eu/products/glc2000/products.php) from Global Land Cover 2000 product website for Asia domain. The raster data is then projected to Vietnam coordinates and clipped to the study region. The legend for land use type in original GLCC format is then modified to match with SWAT database format (Figure 2b).

Global soil map is downloaded from FAO with scale 1:5.000.000 in ESRI shapefile format (<u>http://www.fao.org/geonetwork/srv/en/metadata.show?id=14116</u>). The data is then projected to Vietnam domain, then clipped to study region, then converted to raster format as an input for SWAT model (Figure 2c).

Precipitation from APHRODITE has been downloaded from its website for Monsoon Asia (http://www.chikyu.ac.jp/precip/index.html) in daily scale, gridded spatial resolution of 0.25° in NetCDF format (Network common data format) for study period from 1970-1990. Bilinear interpolation approach has been applied using MATLAB to obtain time series daily data for 16 different stations in the study domain (Figure 2d). The same method for GHCN temperature has been applied to obtain temperature data at Dak To meteorology station. The whole methodology
 is displayed in Figure 2.

3 The whole Da River catchment was divided into 23 sub-catchments in which the discharge 4 gauging station Lai Chau is the middle of the catchment lying in Vietnam area (used for 5 validation part) and Hoa Binh, at downstream end of the catchment which serves for calibration 6 purpose (Figure 2). Daily precipitation for the whole study period was bi-linearly interpolated to 7 16 stations within the whole catchment (see Figure 2). The SWAT model takes as input, 8 measured rainfall data from gauged stations and then uses a rainfall distribution code (skewed 9 distribution or mixed exponential distribution) to generate precipitation values all over a 10 catchment (Neitsch et al., 2004). Hence, an interpolation method is required to compute the 11 station data (at a particular grid point) from the gridded observation data. Amongst different 12 interpolation methods, piecewise constant interpolation, linear interpolation, polynomial 13 interpolation and spline interpolation, the linear interpolation method is usually used by many 14 because of its simplicity and convenience. The bilinear interpolation method is an extension of 15 the linear interpolation for interpolating functions of two variables on a regular grid and hence 16 we use bilinear interpolation method to extract precipitation value for station data, at a grid point. 17 The same approach is applied for the air temperature at the meteorological station.

18 **3.2 Results and Discussion**

19 The experimental study consists of three parts: (1) set up the SWAT model for the Da River 20 basin using the internet based data and calibrate the model using APHRODITE daily rainfall, 21 GHCN2 temperature for 11 years from 1971-1982, with the first year as a warm up period. The 22 observed daily data from Hoa Binh gauging station was used as the benchmark to compare 23 SWAT performance. (2) validate the SWAT model for the basin for 7 years 1983-1989 for the 24 same region using Hoa Binh station as the benchmark for comparison of model performance. (3) 25 verify the model using upstream station at Lai Chau for the same initially calibrated period 1971-26 1982. This step was to prove that the model was performing well not only for the downstream 27 station but also for the upstream station where the stream flow source is from the runoff over the 28 Chinese region of the catchment. Figure 3 shows the locations of the 2 discharge gauging 29 stations (Lai Chau at upstream and Hoa Binh at downstream), rainfall stations and the entire 30 trans-boundary nature of the catchment in discussion. For clarity, the dotted region at downstream is the catchment framed by 2 control stations Lai Chau and Hoa Binh which lies
 over the Vietnam territory and the striped region at upstream is catchment controlled by Lai
 Chau station (in Vietnam) which measures stream flow from the China part.

4 Sensitivity analysis is a method to analyze the sensitivity of model parameters to model output 5 performance. In SWAT, there are 26 parameters sensitive to water flow, 6 parameters sensitive 6 to sediment transport and other 9 parameters sensitive to water quality. The sensitivity analysis 7 method coupled in SWAT model uses Latin Hypercube One-factor-At-a-Time method (LH-8 OAT), which accounts for the strength of the Latin Hypercube sampling (McKay et al., 1979; 9 McKay, 1988). This has also been highlighted by Griensven et al., (2006). The first column of 10 Table 3 shows the order of 10 parameters in SWAT model which are the most sensitive to model 11 output. Auto-calibration using ParaSol is applied to those most sensitive parameters to find the 12 appropriate range of parameters that yield the best result compared to observed discharge data at 13 gauging station. ParaSol is an optimization and a statistical method for the assessment of parameter uncertainty and it can be classified as being global, efficient and being able to deal 14 with multiple objectives (Van Griensven and Meixner, 2006). The Shuffled Complex Evolution 15 16 method (SCE-UA), an algorithm that optimizes model parameters (Duan et al, 1992), is used in 17 this study. This methodology has also been discussed by Stehr et al., (2010).

18 The Nash-Sutcliffe Efficiency (NSE) (Nash and Sutcliffe, 1970) and the Coefficient of Determination R^2 (Krause et al., 2005) are used as the benchmarking indices for the simulated 19 runoff. R² is the square of correlation coefficient (CC) from equation (1) and NSE is calculated 20 from equation (2) shown below. The R^2 ranges from 0 to 1 in which 1 is the best match. The 21 22 NSE shows the skill of the estimates relative to a reference and it varies from negative infinity to 23 1 (perfect match). The NSE is considered to be the most appropriate relative error or goodness-24 of-fit measures available owing to its straightforward physical interpretation (Legates and 25 McCabe, 1999).

$$26 \qquad R^{2} = \left\{ \frac{\sum_{i=1}^{n} \left[\left(o_{i} - \overline{o} \right) \left(s_{i} - \overline{s} \right) \right]}{\sqrt{\sum_{i=1}^{n} \left[\left(o_{i} - \overline{o} \right)^{2} \right] \sum_{i=1}^{n} \left[\left(s_{i} - \overline{s} \right)^{2} \right]}} \right\}^{2} \tag{1}$$

1
$$NSE = 1 - \frac{\sum_{i=1}^{n} \left[\left(o_i - s_i \right)^2 \right]}{\sum_{i=1}^{n} \left[\left(o_i - \overline{o} \right)^2 \right]}$$
 (2)

2 where o and s are observed and simulated discharge dataset respectively.

3 Results for the daily time scale calibration at the Hoa Binh station for the period 1971-1982 show 4 that the NSE and R^2 for the calibration part are quite promising with values of 0.90 and 0.91 respectively (Figure 4a). This is taken as an indicator for very good performance as such values 5 6 have been obtained for modeling at daily time scales which are usually highly variable in space 7 and time compared to monthly time scales. Results for validation at the Hoa Binh station, for the period 1983-1989 show NSE and R^2 values of 0.88 and 0.90, respectively (Figure 4b) and the 8 verification results done for the Lai Chau station show NSE and R² indices of 0.83 and 0.86, 9 10 respectively (Figure 4c). Summary of the above results are tabulated in Table 4. The very 11 promising indices for the verification part at Lai Chau gauging station imply that even there is no 12 data available for upstream region, the inputs from internet sources are good enough substitutes for station data. This is a very important finding, especially for trans-boundary area and 13 14 developing countries like in Vietnam and Southeast Asia region where lack of station data is 15 very common. These results also imply that the study can be furthered in lieu of climate change 16 studies where high resolution climate models can generate important climate variables such as 17 rainfall and temperature which can then be used for hydrological modeling. As such, these 18 internet data sources and available gridded observations can be used to validate and evaluate the 19 climate model generated variables. Such climate model derived estimates can then be used to 20 quantify stream flow changes in the future, in respect of climate change. In a similar fashion of 21 the use of bi-linear interpolation, the climate model derived variables can also be bi-linearly 22 interpolated to the station locations and then used in the SWAT model to study any hydrological 23 responses.

24 **4** Conclusions

This study simulates the stream flow of Da River in Vietnam for a daily flow over a 11 year period between 1971-1982 using the SWAT model across a trans-boundary region between China and Vietnam. The chosen period ensures that there was no dam built on the main river

which allows recalculating the existing natural flow of the river. Due to lack of spatial and 1 2 weather data from the Chinese regions, internet based data have been used in the modeling. Daily scale gridded rainfall is also used as input to the SWAT model whose results are compared 3 4 against observed gauging station data in Vietnam. Three scenarios have been run with calibration 5 and validation part for the Hoa Binh station and the verification for the Lai Chau station. The NSE and R^2 indices are very promising with values higher than 0.85 for most cases, also 6 7 showcasing very good performance of the SWAT model. The results of this study also indicate 8 that internet based data are applicable for hydrological model for large scale watersheds, especially for regions where spatial and temporal data are scarce or sensitive like the trans-9 10 boundary problem discussed here. This approach also has implications for climate change 11 applications, where the daily scale rainfall and temperature could be obtained from high resolution regional climate models for present-day and future climates from which the 12 13 hydrological responses may be ascertained.

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Variables	Resolution	Source
Topography map	90m	Shuttle Radar Topography Mission (SRTM)
Land use/Land cover map	1000m	Global Land Cover Characterization (GLCC)
Soil map	10,000m	Food and Agriculture Organization (FAO)
Precipitation	0.25°/daily	Asian Precipitation Highly Resolved Observational
		Data Integration Towards the Elevation
		(APHRODITE)
Temperature	0.5°/daily	Modified of Global Historical Climatology
		Network version 2 (GHCN2)

Land use type	Description	Area (%)
Forest Evergreen	Land dominated by evergreen forest	37.00
Range-Grasses	Land covered by natural grasses	0.02
Range-Brushes	Land covered by natural bushes	0.01
Agricultural Land-Row Crops	Land used for agriculture activities	62.70
Oil Palm	Land covered by oil palm	0.01
Rice	Land used for growing rice	0.26

1 Table 2. Land use/Land cover types and percentage of area in the study region

1 Table 3. Sensitivity analysis ranking of 10 most sensitive parameters in SWAT model to stream	n
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2 flow

Sensitivity Analysis Order	Parameter	Description	Parameter range
1	Cn2	Moisture condition II curve no	35 ~ 98
2	Alpha_Bf	Baseflow recession constant	0 ~ 1
3	Ch_K2	Effective hydraulic conductivity in main channel	-0.01 ~ 500
4	Surlag	Surface runoff lag coefficient	1 ~ 24
5	Ch_N2	Manning n value for the main channel	-0.01 ~ 0.3
6	Blai	Maximum potential leaf area index for land cover	0 ~ 8
7	Sol_Awc	Available water capacity	0 ~ 1
8	Esco	Soil evaporation compensation factor	0 ~ 1
9	Canmx	Maximum canopy storage	0 ~ 100
10	Gwqmn	Threshold water level in shallow aquifer for base flow	0 ~ 5000

1 Table 4. Model evaluation statistical indices for daily discharge for 3 simulation scenarios

Statistic	Calibration	Validation	Verification
NSE	0.90	0.88	0.83
\mathbf{R}^2	0.91	0.90	0.86



Figure 1: (a) Da River basin location and DEM (b) Total annual rainfall distribution map (c) Land use map (d) Soil map



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1	Figure 2:	Experiment	Methodology

- 2 From left to right
- 3 (a) SRTM website ASCII file downloaded and converted to raster format River basin
 4 delineation for Da river and DEM created
- (b) GLCC website Asia land use downloaded in raster format Legend table for land
 use land use projected and clipped for Da river basin to match SWAT format
- 7 (c) FAO soil website Global soil map Soil map projected and converted to raster
 8 format for Da river basin
- 9 (d) APHRODITE website to download daily gridded rainfall display of a sample for
 10 APHRODITE daily rainfall Location of 16 rainfall stations in study domain bilinear
 11 interpolation to 16 time-series for 16 rainfall locations.



Figure 3: Da River calibration and validation scenarios

