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Article type : Technical Paper

Comparative Analysis of Inundation Mapping Approaches for the 2016 Flood in the Brazos River, Texas

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Abstract: Accurate and timely flood inundation maps serve as crucial information for hydrologists, first-responders and decision makers of natural disaster management agencies. In this study, two modeling approaches are applied to estimate the inundation area for a large flooding event occurring in May of 2016 in the Brazos River: (1) Height Above the Nearest Drainage combined with National Hydrograph Dataset Plus (NHDPlus-HAND) and (2) International River Interface Cooperative— Flow and Sediment Transport with Morphological Evolution of Channels (iRIC-FaSTMECH). The inundation extents simulated from these two modeling approaches are then compared against the observed inundation extents derived from a **This is the author manuscript accepted for publication and has undergone full peer review but has not been through the copyediting, typesetting, pagination and proofreading process, which may lead to differences between this version and the [Version of Record](#). Please cite this article as [doi: 10.1111/1752-1688.12623-17-0035](https://doi.org/10.1111/1752-1688.12623-17-0035)**

28 Landsat-8 Satellite image. The simulated results from NHDPlus-HAND and iRIC- FaSTMECH
29 show 56% and 70% of overlaps with the observed flood extents, respectively. A modified
30 version of the NHDPlus-HAND model, considering networked catchment behaviors, is also
31 tested with an improved fitness of 67%. This study suggests NHDPlus-HAND has the potential
32 for real-time continental inundation forecast due to its low computational cost and ease to couple
33 with the NWM. Better performance of NHDPlus-HAND can be achieved by considering the
34 inter-catchment flows during extreme riverine flood events. Overall, this study presents a
35 comprehensive examination made of remote sensing compared with HAND-based inundation
36 mapping in a region of complex topography.

37

38 **(Key Terms:** Flooding; Inundation; NHDPlus-HAND; iRIC; Simulation; Observation; Remote
39 Sensing.)

40

INTRODUCTION

41 Flooding is one of the leading causes of natural disaster related deaths worldwide
42 (Federal Emergency Management Agency, 1992; Conrad *et al.*, 1998; Merwade *et al.*, 2008;
43 Cook and Merwade, 2009). According to a study on flood damage in the United States, flood
44 damage increases over time due to rapidly growing population and urban development (Pielke *et*
45 *al.*, 2002). Accurate and timely inundation maps not only provide first-hand information for
46 rescuing and emergency operations during floods, but also potentially improve flood risk
47 management and better estimate flood insurance rates (Merwade *et al.*, 2008; Cook and
48 Merwade, 2009; Fang *et al.*, 2011). In the United States, most major river systems have flood-
49 risk maps delineated by the Federal Emergency Management Agency (FEMA) through the
50 National Flood Insurance Program (NFIP). Although FEMA has produced approximately
51 100,000 flood-risk maps based on 100-year return period flows (NFIP, 2002), inundation maps
52 for real events are unavailable or limited by uncertainties in data sources or modeling (Christian
53 *et al.*, 2013).

54 Flood inundation modeling approaches are essentially to convert flows from either
55 hydrologic models or observation gages into inundation extent/depth based upon topographic
56 information. In general, inundation models can be classified as terrain-based and dynamic-based
57 approaches. Terrain-based approaches refer to the methods employing topography and
58 simplifying the fluid mechanics process to predict inundation extents. Intersecting topography

59 surface with a planar water surface is normally defined as the simplest terrain-based approach to
60 generate inundated area (Priestnall *et al.*, 2000). Some terrain-based models are also known as
61 storage cell models by treating the floodplain as many storage cells and solving uniform flow
62 formulas like Manning's and weir-type equations for floodplain routing (Cunge *et al.*, 1976;
63 Estrela, 1994; Romanowicz *et al.*, 1996, Bates and De Roo, 2000). Another type of terrain-based
64 approach calculates the elevation difference between each grid cell and its nearest flowpath grid
65 based on topographic information (Nobre *et al.*, 2011). In summary, the terrain-based approaches
66 aim to reduce computational cost while generating satisfactory inundation results.

67 Dynamic-based approaches include hydraulic/hydrodynamic models which are generally
68 categorized as one-, two- and three-dimensional models. One-dimensional hydraulic models
69 consider fluid continuity and momentum, which solve one-dimensional St. Venant equations.
70 One example of such hydraulic models is the Hydrologic Engineering Center's River Analysis
71 System (HEC-RAS) developed by the U.S. Army Corps of Engineers (USACE). The general
72 steps of inundation mapping using one-dimensional hydraulic models involve: (1) obtaining
73 discharge information from gage observation or a calibrated hydrologic model; (2) developing
74 perpendicular cross sections along the flow path based on Digital Elevation Model (DEM) or
75 survey information with hydraulic parameters (e.g. surface roughness); (3) calculating the water
76 surface elevations based on the discharge and cross-sectional information from the previous steps;
77 (4) comparing the water surface elevations with DEMs, and the area where water surface is
78 higher than terrain elevation is defined as inundated (IACWD 1982; Maidment and Djokic, 2000;
79 Noman *et al.*, 2001; FEMA 2003; Merwade *et al.*, 2008). Two-dimensional hydraulic models use
80 finite-element mesh as a calculation unit and have capability to simulate the lateral unsteady flow
81 dynamics including backflow condition (Crowder and Diplas, 2000; Merwade *et al.*, 2008).
82 Three-dimensional hydraulic models can fully represent the comprehensive form of the Navier-
83 Stokes equations (White, 1974; Lane *et al.*, 1999). Since three-dimensional approaches might be
84 unnecessarily complex and computationally expensive (Bates and De Roo, 2000; Horritt and
85 Bates, 2001; Hunter *et al.*, 2007), one- and two-dimensional models are the primarily used in
86 floodplain prediction to date (Hunter *et al.*, 2007). Table 1 shows the comparison between
87 dynamic-based and terrain-based inundation mapping approaches.

88

[INSERT TABLE 1 HERE]

89

[Table 1. Comparison of dynamic-based and terrain-based inundation approaches]

90

Height Above the Nearest Drainage combined with National Hydrograph Dataset Plus (NHDPlus-HAND) is a terrain-based flood inundation model. The Height Above the Nearest Drainage (HAND) concept was first introduced by Rennó *et al.* (2008). The HAND model normalizes topography based on relative heights found along the nearest drainage network (Nobre *et al.*, 2011). The HAND raster is generated by subtracting the elevation of each grid cell from the elevation of its nearest stream grid cell. Nobre *et al.* (2016) has validated the HAND method using a flood event in Southern Brazil with a finding that HAND can be used to predict inundation extents. Since HAND rasters are computed based on topography and flowpath information, accurate Digital Elevation Model (DEM) and flowline are essential components in establishing a HAND model. The National Hydrograph Dataset Plus (NHDPlus) is an integrated geo-spatial, hydrologic dataset built by the U.S. Environmental Protection Agency (USEPA) Office of Water and the US Geological Survey (USGS). The NHDPlus Version 2 dataset provides a reliable stream network consisting of approximately 2.7 million reaches in the continental United States (<http://www.horizon-systems.com/nhdplus/>). Liu *et al.* (2016) calculated HAND rasters for the contiguous United States using a 10-m resolution DEM combined with NHDPlus streamlines (termed NHDPlus-HAND). For delineating inundation maps, HAND needs discharge and rating curve information from hydrologic/hydraulic models. Brought into operations in August of 2016, the National Water Model (NWM) is a high-resolution hydrologic model simulating discharge for 2.7 million NHDPlus (Version 2) stream reaches over the continental United States (NOAA, 2016). The NWM is developed based on the Weather Research and Forecasting Model Hydrological (WRF-Hydro) framework, which utilizes meteorological forcing from the operational High Resolution Rapid Refresh (HRRR) model and precipitation forcing from the Multi-Radar/Multi-Sensor System (MRMS). To obtain discharge information, the NWM utilizes a vector-based channel routing module based on the NHDPlus reaches, which was firstly demonstrated using the Routing Application of the Parallel Computation of Discharge (RAPID) in 2015 (Maidment, 2017; Lin *et al.*, 2017), and then evolved into the Muskingum-Cunge routing method in 2016 (NOAA, 2016). This study uses a set of the pre-operational NWM discharge data that ingests the streamflow data assimilation

117

118 capability at ~7000 gauge stations (NOAA, 2016) on top of the Muskingum-Cunge routing. The
119 NHDPlus-HAND is chosen as the terrain-based model for the study due to its ease of application
120 to be coupled with the existing hydrologic model (NWM).

121 A dynamic-based model, International River Interface Cooperative - Flow and Sediment
122 Transport with Morphological Evolution of Channels (iRIC-FaSTMECH) is a two-dimensional
123 hydrodynamic model employing a channel-fitted coordinate system (cylindrical coordinate
124 system), where the curvature follows the stream direction (FaSTMECH Model Note. Accessed
125 March 1, 2017, <http://i-ric.org/en/downloads>). It provides information of velocity and water
126 surface elevation for a given discharge and roughness by hydrostatic-distribution pressure and a
127 quasi-steady approximation (Nelson and McDonald, 1996), which allows the discharges to vary
128 in time, and simplifies unsteady terms in the equations of motion. The iRIC system includes
129 different models with less restrictive assumptions and more applicability but requires more
130 extensive calibration data (Nelson *et al.*, 2016, www.i-ric.org). The iRIC model framework is
131 upgraded from the Multi-Dimensional-Surface Water System Modeling System (MD-SWMS)
132 (McDonald *et al.*, 2001, 2005), which employs a finite difference approach on a curvilinear grid
133 to solve the depth and Reynolds-averaged Navier-Stokes equations (Nelson *et al.*, 2003). Not
134 only can the iRIC-FaSTMECH provide the maximum inundation area at peak stage but also be
135 used for simulating water level, flow velocity distribution for the floodplain, etc. (Ku and Kim,
136 2014; Son *et al.*, 2014; Kail *et al.*, 2015). Kenney and Freeman (2011) suggested that iRIC-
137 FaSTMECH give a fair spatial understanding of water-surface elevation, velocities, and sheer
138 stress associated with high flows. Son *et al.* (2014) showed that iRIC-FaSTMECH well
139 simulated water-surface levels in South Korea. Due to its simulation efficiency and utility in
140 predicting water-surface elevation during the flood event, iRIC-FaSTMECH is selected as the
141 dynamic-based model in this study.

142 Lack of reliable observed spatial extents of flood inundation limits the validation and
143 utility of both approaches in flood inundation mapping. Fortunately, the advent of satellite-based
144 remote sensing technology has become a key tool for flood monitoring (e.g. Dartmouth Flood
145 Observatory: <http://floodobservatory.colorado.edu>). Such applications of satellite imagery for
146 river inundation (e.g. the U.S. Flood Inundation Map Repository: <http://sdml.ua.edu/usfimr>)
147 serve as observations on flooding areal extents within the region of interest (Khan *et al.*, 2011).

148 In this regard, the authors are motivated to investigate the performances of various
149 inundation simulations based on observation derived using remote sensing techniques. A better
150 understanding of mechanisms of terrain- and dynamic-based inundation approaches can then be
151 achieved. Furthermore, deeper insights are gained to improve the timeliness and accuracy of
152 real-time flood inundation mapping. This study is conducted to achieve the following objectives:

153 1. To validate the NHDPlus-HAND's simulation using streamflow information from the
154 NWM along with the observation derived from the satellite imagery.

155 2. To evaluate the terrain-based (NHDPlus-HAND) and dynamic-based (iRIC-
156 FaSTMECH) models with respect to modeling accuracy, modeling efficiency (running time),
157 and feasibility in real-time mode.

158 3. To provide suggestions for future model development and explore potential efficient
159 ways to improve NHDPlus-HAND towards accurate large-scale inundation mapping.

160 **STUDY AREA**

161 Based on a request for presidential disaster declaration by Abbott (2016), 12 counties in
162 Texas with a population of 3.9 million were impacted by a flood event occurring in the Brazos
163 River in May of 2016. There were over 11,000 people evacuated from their impacted homes
164 along the Brazos River (Abbott, 2016). Due to the severity of the flood, the May of 2016 event is
165 particularly selected for this study to seek useful information for future decision making under
166 severe weather conditions.

167 The study area (Figure 1) is a section of the Brazos River spanning 27 km of main-stem
168 distance upstream of the USGS gage near Hempstead (ID: 8111500). The inundation of the May
169 2016 event over the study area was captured in a relatively cloud-free Landsat 8 image. As
170 shown in Figure 2B, the majority of the area appears to be flooded in comparison with the pre-
171 flood condition (Figure 2A). The USGS gage (ID: 8111500) recorded a total rainfall of 255.8
172 mm within 25 hours, as well as the peak stage of 16.78 m (above the datum) at 3 p.m. on May
173 27th, 2016 (CDT). The Landsat 8 imagery (Figure 2B) was captured at 12 p.m. on May 28th
174 (CDT), 21 hours after the peak stage occurred. However, at the moment when satellite imagery
175 was taken, it shows that the stage elevation decreased from the peak stage elevation only by 2.5%
176 (Figure 3), implying the slow recession of the river after the peak stage occurred. Because of

177 such a small difference, the flood extents captured by satellite imagery is assumed to represent
178 the peak inundation in this study.

179 **[INSERT FIGURE 1 HERE]**

180 [Figure 1. The study area and stream reaches in the Brazos River, Texas]

181 **[INSERT FIGURE 2 HERE]**

182 [Figure 2. Satellite imageries showing A. pre-flood (12 p.m. on March 25th, 2016(CDT)) and B.
183 post-flood condition (12 p.m. on May 28th, 2016(CDT)) in the study area]

184 **[INSERT FIGURE 3 HERE]**

185 [Figure 3. Stage hydrograph and rainfall hyetograph shows the target timing of model simulation
186 and the timing of satellite observation]

187 **METHODOLOGY**

188 To better understand and evaluate two inundation modeling methods and their
189 corresponding performances, the authors conduct a series of comparisons of simulated
190 inundation from both approaches, i.e. terrain-based (NHDPlus-HAND and the modified HAND)
191 and dynamic-based (iRIC-FaSTMECH), with observed inundation from the satellite imagery for
192 the study area during the May 2016 event. For simplicity, NHDPlus-HAND and iRIC-
193 FaSTMECH are referred to as HAND and iRIC respectively in the following sections. The
194 methodology consists of three major parts: HAND, iRIC, and remote sensing. The HAND model
195 uses hydrologic model's outcome as the input discharge to delineate flood inundation. As a
196 supplementary approach, a modified version of HAND is also tested for inundation mapping.
197 The iRIC is a two-dimensional hydraulic model, to simulate flood inundation maps. Satellite
198 imagery is used as observation for further comparisons with simulations in the third part. Figure
199 4 illustrates the data and workflow of the methodology with more details in the following
200 sections.

201 **[INSERT FIGURE 4 HERE]**

202 [Figure 4. Schematic diagram of this study]

203 ***NHDPlus-HAND***

204 NHDPlus-HAND generates inundation maps through the procedure as shown in Figure 4.
205 First, the NWM output provides hourly discharge information for each NHDPlus flowline
206 (identified by a unique ID number - ComID). In general, the discharge from the NWM
207 undergoes nudging-based data assimilation wherever/whenever gage-observed values become
208 available, which leads to a close match between the NWM discharge and observed discharge as
209 shown in Figure 5. Then these NWM flow information is converted into stage height using rating
210 curves (stage-discharge relationship). The rating curve for each reach is generated from channel
211 properties in the HAND model based on Manning's equation. These interpreted stage height
212 information allows us to determine inundation extents from the HAND raster. In each catchment,
213 HAND cells with lower elevation than the interpreted stage height information can be classified
214 as 'wet' cells. For example if the calculated stage for a given catchment (ComID) is 5 m, all
215 HAND cells with a value between 0 and 5 will be classified as 'wet' cells. A simple GIS-based
216 Raster Calculator equation is applied to identify the 'wet' cells. The whole methodology of
217 creating inundation map using HAND is automated with a Python script utilizing a number of
218 ArcGIS tools for a better efficiency.

219 **[INSERT FIGURE 5 HERE]**

220 [Figure 5. U.S. Geological Survey gage observed hydrograph and the National Water Model
221 hydrograph with nudging-based data assimilation]

222 ***Modified HAND***

223 The NHDPlus-HAND method typically enables users to determine inundated area based
224 on individual NHDPlus catchment with corresponding water depth. However, with the
225 assumption of applying uniform water depth for each catchment, NHDPlus-HAND cannot
226 consider inter-catchment flow mechanisms, i.e. the flow transfer between adjacent catchments.
227 To tackle this deficiency, McGehee *et al.* (2016) developed the modified HAND method by
228 taking stream orders into account when determining the nearest drainage. In essence, the
229 modified HAND method provides an approach to re-define channel network, i.e. deleting
230 streams with low stream orders. McGehee *et al.* (2016) stated that such a modification, if applied
231 locally where the original HAND overlooks catchment interaction, can potentially improve the
232 accuracy of inundation mapping (as demonstrated later). Therefore, the modified HAND method

233 is used to improve the HAND simulation in this study and compare with other inundation
234 mapping results.

235 *iRIC-FaSTMECH*

236 The terrain used to develop iRIC-FaSTMECH model is 10-meter resolution DEM
237 obtained from USGS database (<https://earthexplorer.usgs.gov>), and the NHDPlus-HAND is
238 developed using DEM of the same resolution (10 meters). Table 2 shows the final model settings
239 for generating inundation by iRIC, and specific procedures are illustrated in Figure 4. The
240 researchers first assign a water surface elevation as the initial upstream condition and the peak
241 discharge (4,445.7 m³/s) with stage information (49.7 m) from USGS gage (8111500) as the
242 downstream boundary condition, respectively. After numerous iterations, the iRIC model is able
243 to compute a converged solution as upstream discharge (4,300 m³/s). During the iterations, drag
244 coefficient is updated as it is a function of Manning's coefficient and water depth which varies
245 with upstream discharge. Three Manning's coefficients (0.03, 0.05 and 0.035) are used to
246 corresponding land cover types (main channel, brushy and cultivated) (Chow 1959). Finally,
247 inundation maps are delineated using the optimal drag coefficient and discharge after 1,500
248 iterations. Suggested by Nelson (2016), re-wetting option is turned during the modeling process
249 on since it can improve inundation estimations in large and shallow areas by re-evaluating the
250 wet/dry status of each node during the simulation.

251 [INSERT TABLE 2 HERE]

252 [Table 2. iRIC model settings]

253 *Satellite-based Flood Inundation Mapping*

254 Landsat Satellite missions have been applied in delineating floodplain boundaries over a
255 few regions under different conditions in climate, morphology and land use since 1972 (Rango *et*
256 *al.*, 1975; Hollyday, 1976; Sollers *et al.*, 1978; Smith, 1997; Ho *et al.*, 2010). Amongst the many
257 different techniques of identifying water pixels using a suite of the Landsat Satellites, Supervised
258 Classification has been proven as a robust method to classify features of interest (Frazer and
259 Page, 2000; Shalaby and Tateishi, 2007). The Supervised Classification technique allows users to
260 select sample pixels (end members) in an image as representatives of a specific spectral signature
261 (e.g. water). Image processing software is then used to classify all the image pixels based on the

262 maximum likelihood that these pixels' spectral signature is similar to that of a specific end
263 member.

264 The remote sensing imagery used in this study is obtained from the Landsat 8-
265 Operational Land Imager (OLI) multispectral database (<http://earthexplorer.usgs.gov>). The pre-
266 flood (March 25, 2016) and post-flood (May 28, 2016) images are classified via Erdas Imagine®
267 2015 Image processing software (Hexagon Geospatial, Norcross, GA, USA) for pre-processing
268 and subsequent data manipulation. The resultant Geometrically and Radiometrically rectified
269 imagery is subject to Supervised Classification of flooded pixels based on the maximum
270 likelihood classifier. Typically, the presence of clouds is a common problem in remote sensing
271 imagery, which hinders the identification of flooded water pixels beneath the clouds, leading to
272 under-representation of flood water extents in the study domain. In an attempt to alleviate this
273 problem, the DEM of the flooded region is used to identify the height of the pixels beneath the
274 clouded areas. Each pixel with a lower height than that of the lowest height of an apparent
275 flooded pixel in the neighborhood of the cloud is considered as 'wet'. Spatial filling techniques
276 are then applied to convert these pixels into water pixels. Accuracy assessments are finally
277 performed on the classified imagery subsequent to being post-processed through a 3×3 high pass
278 kernel (Zhang *et al.*, 2016). A high pass kernel has the effect of highlighting boundaries between
279 features (e.g., where water body meets the vegetated land), thus water features can be easily
280 classified by sharpening edges between water and non-water pixels.

281 *Advanced Fitness index (AFI)*

282 In order to evaluate the correspondence between the simulated and observed inundation,
283 the advanced fitness index (AFI) method is applied in this study. The AFI accounts for the
284 match in terms of both inundated and non-inundated area, as shown in **Equation 1**:

$$\text{Advanced Fitness (\%)} = \frac{IA_{obs} \cap IA_{model} + NIA_{obs} \cap NIA_{model}}{A_{total}} \times 100 \quad (1)$$

285 where IA_{obs}/NIA_{obs} is inundated/non-inundated area of the observation; IA_{model}/NIA_{model} is
286 inundated/non-inundated area of the model simulation, and A_{total} is the total area of the study
287 region.

288 The simulated and observed inundation results cannot be compared directly because of
289 their different resolution and patchy nature of the classified water body in a satellite image. To

290 tackle this technical difficulty, the simulated inundation is further resampled to 30 m x 30 m
291 grids and aligned to be consistent with the observed. To evaluate the mapping performances
292 based on corresponding AFI values, the raster calculator function in ArcGIS is used to quantify
293 the inundated/non-inundated area with number of pixels. The following section demonstrates the
294 results from aforementioned methods with an evaluation on their performances.

295 **RESULTS AND DISCUSSION**

296 Inundation maps for the 2016 May flood event in the Brazos River using terrain-based,
297 physical modeling methods and remote sensing classification technique are shown in Figures 6A,
298 6B, 6C and 6D, respectively. In general, inundation areas derived from the iRIC and modified
299 HAND methods have more overlaps with the observed inundation than the original HAND
300 simulation. Also, the HAND simulation clearly misses a few areas considered inundated by all
301 other two methods. As the advanced fitness indices (AFI) indicated, the HAND model has a 56%
302 match with the observation, while iRIC and the modified HAND have higher AFI values of 70%
303 and 67%, respectively. The modified HAND appears to delineate larger inundated area than
304 iRIC, but still generates inferior AFI. As a supplementary approach to HAND, the modified
305 HAND method is found to improve the AFI value by 11% with capturing a few missed areas.
306 Interestingly, the inundation classified from satellite imagery has the least area of only 41.3 km²,
307 compared to the other three modeling approaches (55.9 km² for iRIC, 41.7 km² for HAND and
308 70.8 km² for modified HAND). Although as shown in Figure 6D, the inundated area appears to
309 have some discontinuous features or gaps, leading to the reduced inundation area, the authors
310 consider the inundation from satellite imagery as the 'true' flooded area during the study.

311 **[INSERT FIGURE 6 HERE]**

312 [Figure 6. Results of inundation maps and fitness indices from simulation of (A) HAND, (B)
313 iRIC, (C) Modified HAND, and (D) Satellite Observation]

314 Further investigation is performed to discover why HAND misses certain areas that are
315 captured by the other two methods. Compared to HAND (Figures 7A and 7B), an in-depth
316 illustration for the improvement of the modified HAND simulation is shown in Figures 7C and
317 7D. The left two panels (7A and 7C) respectively show inundated areas of the HAND and
318 modified HAND simulations on top of the topography. The two highlighted NHDPlus

319 catchments 1 and 2 are chosen to exemplify the difference between the HAND and modified
320 HAND methods. It is found that discrepancy between the results of two methods mainly takes
321 place in catchment 2. Catchment 1 is located along the Brazos River main stem with a higher
322 stream order than that of catchment 2 which drains into the main stream as tributary. The cross-
323 sectional views of selected cutting line across catchment 1 and 2 with corresponding simulated
324 water surface elevation (blue lines in the right panels) are shown in Figures 7B and 7D,
325 respectively. The disconnected water surface shown in Figure 7B indicates that the catchment-
326 based calculation by HAND overlooks the inter-catchment flow. However, the water surface
327 would be more realistically simulated as shown in Figure 7D if the modified HAND is
328 strategically applied to the problematic areas. Therefore, the results show that the modified
329 HAND can essentially replace the water depth of the low-stream-order catchment (catchment 2
330 herein) with that of the adjacent catchment with high stream order (catchment 1), instead of
331 determining the inundation of the catchments based on their individual water depth solely.

332 **[INSERT FIGURE 7 HERE]**

333 [Figure 7. (A) Inundation map derived from HAND on top of DEM, (B) Cross-sectional view of
334 the selected area and water surface simulated in HAND, (C) Inundation map derived from the
335 modified HAND on top of DEM, (D) Cross-sectional view of the selected area and water surface
336 simulated in modified HAND]

337 Overall, iRIC is found to generate the best match with the observation in this case study.
338 Bates and De Roo (2000) also reported in their study that two-dimensional models would
339 perform better than terrain-based models when their resolution was similar. If the problematic
340 catchments in HAND simulation are excluded from the comparison, the HAND and iRIC models
341 would generate comparable AFI values (65% and 68% respectively). Such results indicate that
342 aside from HAND's deficiency to model inter-catchment inundation, it has equivalent capability
343 as iRIC in terms of matching the observation. This study suggests that simplification on the
344 intricacy of flow dynamics employed by terrain-based models has relatively minor influence on
345 peak inundation prediction. This finding confirms that using a simple terrain-based model could
346 adequately simulate the flood inundation area as discovered by Bates and De Roo (2000).

347 Although the dynamic-based model represented by iRIC in this study provides more
348 accurate estimates under high flow conditions, the model needs intensive calibrations using

349 various historical data to achieve a reliable performance (Nobre *et al.*, 2016). Promisingly,
350 terrain-based model, as represented by HAND in this study, demonstrates a unique inundation
351 mapping capability for a river section without much historical data. Given the fact that solving
352 the St.Venant equations for hydraulic/hydrodynamic models through iterating process is very
353 computationally expensive, their utility is limited in real-time flood prediction (Fang *et al.*, 2008).
354 In our case, the iRIC model needs approximately 90 minutes (5,400 seconds) to run 1,500
355 iterations, while the HAND model only takes about 2 minutes (120 seconds) for 37 NHDPlus
356 river reaches to generate the inundation. For a large-scale hydrologic forecasting system like the
357 NWM, the HAND inundation mapping approach clearly shows benefits in coupling with the
358 model as demonstrated in this study, potentially addressing the real-time continental-scale
359 inundation mapping problem in an efficient way.

360 However, the HAND calculation does not explicitly reflect interactions between the main
361 stem and its tributaries. This issue likely becomes more pronounced for larger riverine floods. In
362 addition to the modified HAND, one remedy to this problem is to incorporate a mass balance
363 process into the modeling framework as suggested by Bates and De Roo (2000). The authors
364 think that while not observed in this study, the accuracy of the HAND model would be inferior to
365 a traditional hydraulic model if the flooded area has complex urban hydraulic components like
366 culverts, pipes and bridges. Therefore, extra caution should be taken when choosing proper
367 inundation models for flood risk prediction due to the uncertainties in data requirement,
368 computation demand, accuracy, types of land cover, etc.

369 CONCLUSIONS AND FUTURE WORK

370 This paper demonstrates a unique analysis of using terrain-based (NHDPlus-HAND and
371 the modified HAND) and physical (iRIC-FaSTMECH) models to simulate the maximum
372 inundation extents during the May 2016 flood event in the Brazos River, TX. A supervised
373 classification method is used to classify water from Landsat 8 satellite imagery and generate an
374 observed inundation map. To better understand and evaluate the performances of three methods,
375 the goodness of overlapping between the simulated and observed is quantified via the advanced
376 fitness index (AFI). The main conclusions from this study are summarized as follows:

377 1. NHDPlus-HAND, the modified HAND and iRIC generated a fair (> 50% of AFI) fit
378 with the satellite imagery. iRIC performed a slight better (~ 70% in AFI) than other two methods
379 (NHDPlus-HAND and the modified HAND) during this extreme flood event.

380 2. Although the NHD catchment-based calculation does not allow NHDPlus-HAND to
381 explicitly account for inter-catchment flows between the main stem and its tributaries, the
382 modified HAND method provides a remedy to this issue when strategically applied to the areas
383 overlooked by NHDPlus-HAND.

384 3. For extreme events, simplification on the intricacy of flow dynamics has relatively
385 minor influence on predictions, which can positively justify the utility of NHDPlus-HAND for
386 large-scale inundation mapping.

387 4. Even though the current version of NHDPlus-HAND may not be a superior choice for
388 handling accurate inundation mapping for urban areas, its low computational cost and ease to
389 couple with the National Water Model (NWM) provide great potential to support real-time
390 continental inundation forecast in the future.

391 The authors think that there is room for future investigation in uncertainty analysis of
392 observations using multiple sources of raw imagery along with various classification techniques.
393 Potential sources of raw imagery will be used including synthetic aperture radar (SAR),
394 unmanned aerial vehicle (UAV) and so on; while classification methods like Delta-cue change
395 detection on pre/during flooding scenarios, normalized difference water index and image fusion
396 techniques will be also used to generate inundation extents. The results of the future research will
397 be reported in a forthcoming paper.

398 Overall, this study presents a comprehensive examination made of remote sensing
399 compared with HAND-based inundation mapping in a region of complex topography. Findings
400 from this paper can also help identify potential improvements for HAND-based simulation. In
401 light of frequent floods occurring in the nations, the information provided from this study is
402 valuable for the scientific/engineering communities, floodplain managers, emergency personnel
403 and governmental entities that were impacted by the storm and/or had a vested interest in the
404 region.

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ACKNOWLEDGMENTS

This research was conducted during the 2016 NOAA National Water Center (NWC) Summer Institute administered by the Consortium of Universities for the Advancement of Hydrologic Science, Inc. (CUAHSI). The authors would like to thank Peirong Lin and Adnan Rajib for their coordination and all the advisory faculty, theme leaders and coordinators involved in the 2016 Summer Institute. The authors would especially like to thank, Dr. David Maidment, Dr. Sarah Praskievicz and Dr. Richard McDonald for their generous support and thoughtful comments during the period of our project work. Fang and Zhang would like to thank the Tarrant Regional Water District (TRWD) and the National Science Foundation (NSF) under Grant No. CyberSEES-1442735 for their support. Tsang and Huang would like to thank Dr. Yi-Leng Chen for his assistance to the workshop travel and application. Cohen and Munasinghe would like to thank the University Corporation for Atmospheric Research (UCAR) and the NWC for their support via the COMET Program Cooperative Project Grant under Cooperative Agreement No. Z16-23487 with the National Oceanic and Atmospheric Administration (NOAA).

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Table 1. Comparison of dynamic-based and terrain-based inundation approaches

Model Type	Reference	Routing	Data requirement	Run time	Calibration	Validation	
Dynamic-based	1D	Fread (1984); Fread (1993)	Full solution of the 1D St. Venant equations	Discharge, DEM, Cross-sections, Channel parameters, Land use/type	Fast	Necessary	Yes
	2D	Bates <i>et al.</i> (1992)	Full solution of the 2D St. Venant equations with turbulence closure	Discharge, DEM, Land use/type	Moderate	Contingent	Yes
	3D	White (1974); Lane <i>et al.</i> (1999)	Full solution of the Navier-Stokes equations	3D DEM, 3D bathymetry, 3D velocity, Land use/type	Slow	Contingent	Yes
Terrain-based	Planar water surface	Priestnall <i>et al.</i> (2000)	None	DEM, water surface elevation	Fast	Limited	Yes
	Storage cell	Cunge <i>et al.</i> (1976); Estrela (1994); Romanowicz <i>et al.</i>	Uniform flow formula (Manning equation or weir-type equations);	DEM, discharge, Initial channel flow depth, Land	Fast	Limited	Yes

	(1996); Bates and De Roo (2000);	Kinematic wave and Manning equation	use/type			
HAND	Nobre <i>et al.</i> (2011)	Uniform flow formula (Manning equation)	DEM, discharge/water surface elevation, Land use/type	Fast	Limited	Yes

Table 2. iRIC model settings

Setting Menu	Description
Initial Condition	Initial Water Surface Elevation: 1D step-backwater
Boundary Condition	Downstream Peak Discharge: 4445.7 m^3/s Downstream Peak Stage: 49.7 m
Iteration	1500
Upstream Discharge	4300 m^3/s
Upstream Stage	Constant (time-invariant)
Drag Coefficient	Variable
Re-wetting	On

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FIGURE CAPTIONS586 **Figure 1.** The study area and stream reaches in the Brazos River, Texas587 **Figure 2.** Satellite imageries showing A. pre-flood (12 p.m. on March 25th, 2016(CDT)) and B.
588 post-flood condition (12 p.m. on May 28th, 2016(CDT)) in the study area589 **Figure 3.** Stage hydrograph and rainfall hyetograph shows the target timing of model simulation
590 and the timing of satellite observation591 **Figure 4.** Schematic diagram of this study592 **Figure 5.** U.S. Geological Survey gage observed hydrograph and the National Water Model
593 hydrograph with nudging-based data assimilation594 **Figure 6.** Results of inundation maps and fitness indices from simulation of (A) HAND, (B)
595 iRIC, (C) Modified HAND, and (D) Satellite Observation596 **Figure 7.** (A) Inundation map derived from HAND on top of DEM, (B) Cross-sectional view of
597 the selected area and water surface simulated in HAND, (C) Inundation map derived
598 from the modified HAND on top of DEM, (D) Cross-sectional view of the selected
599 area and water surface simulated in modified HAND

Figure 1

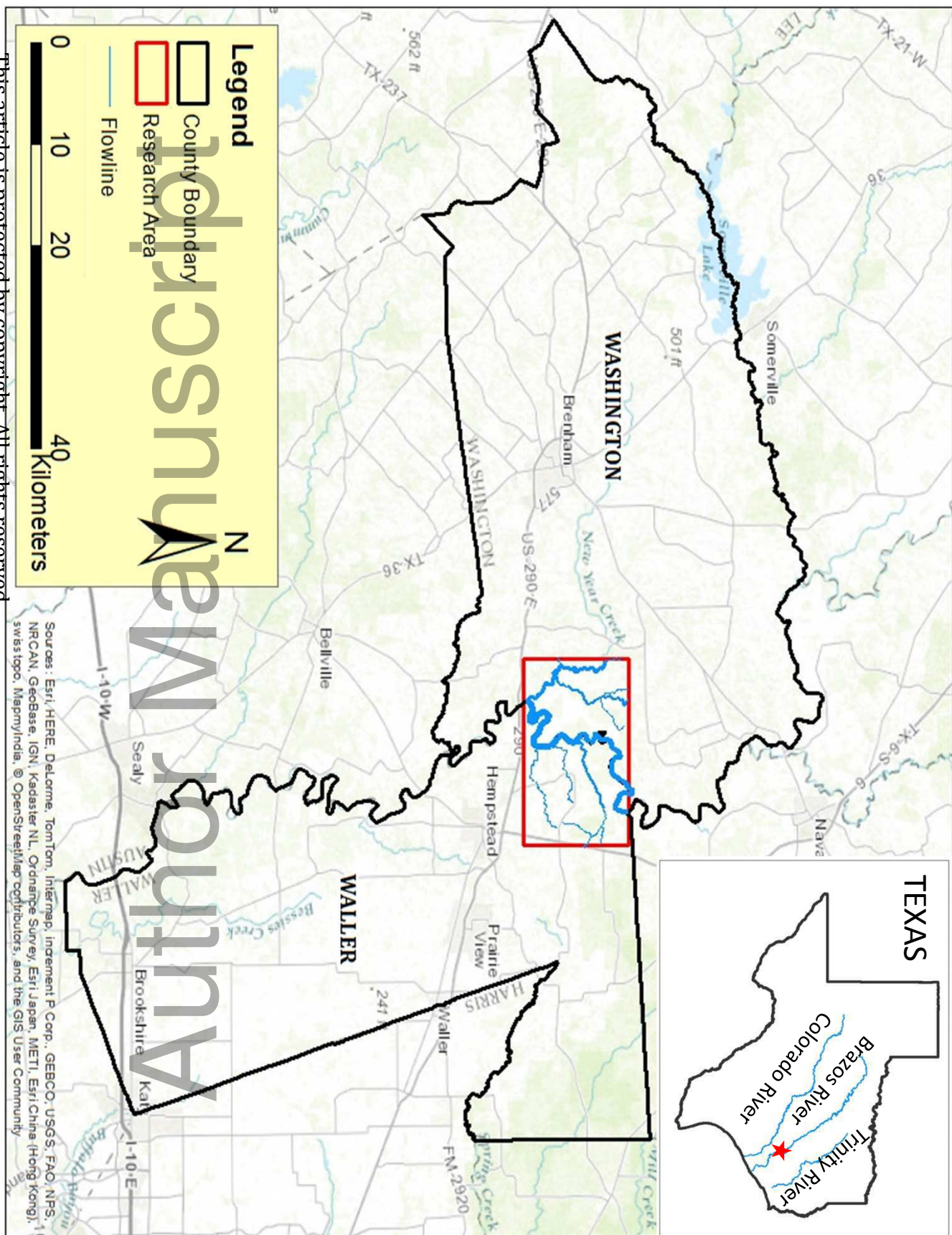


Figure 2

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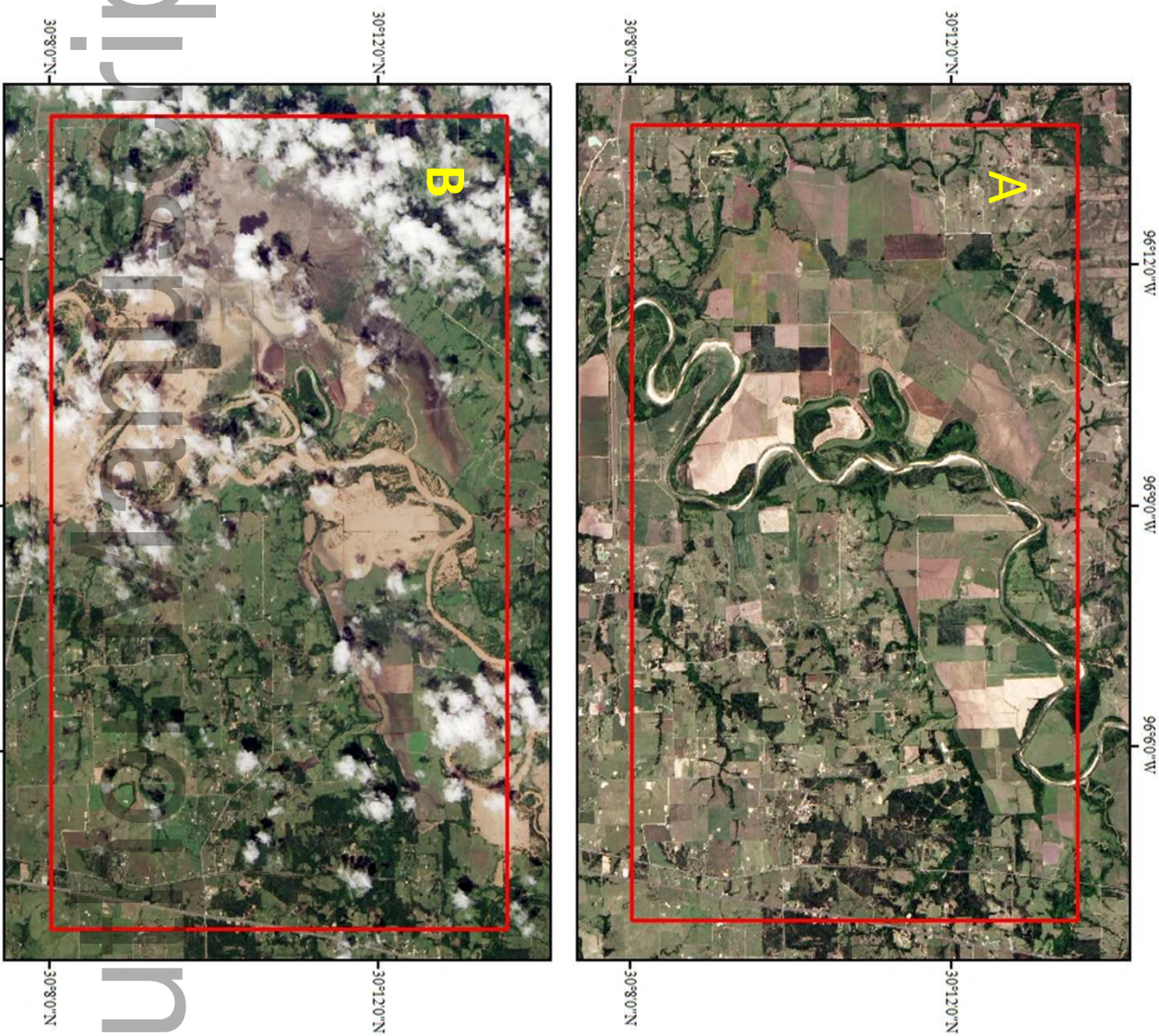


Figure 3

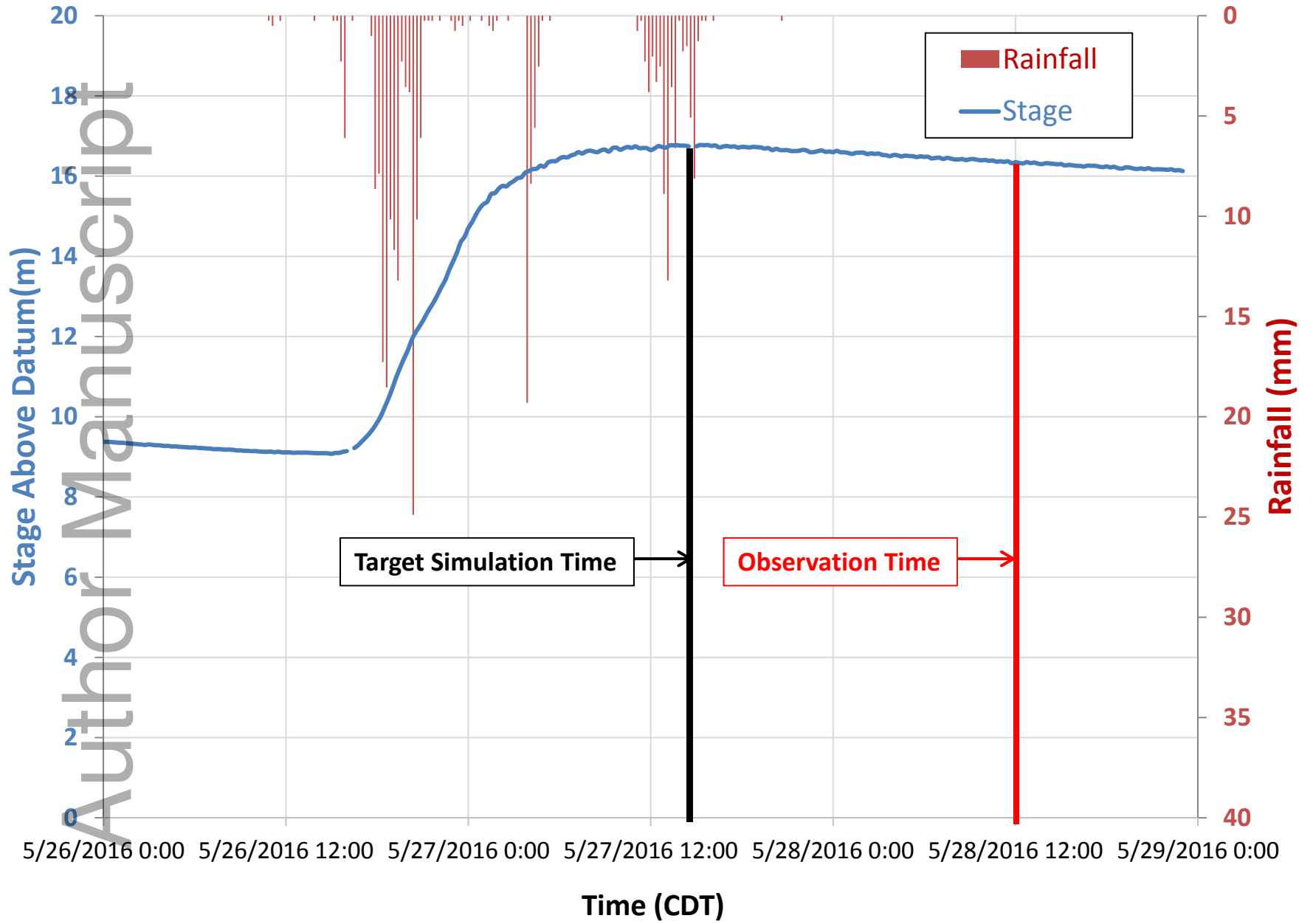


Figure 4

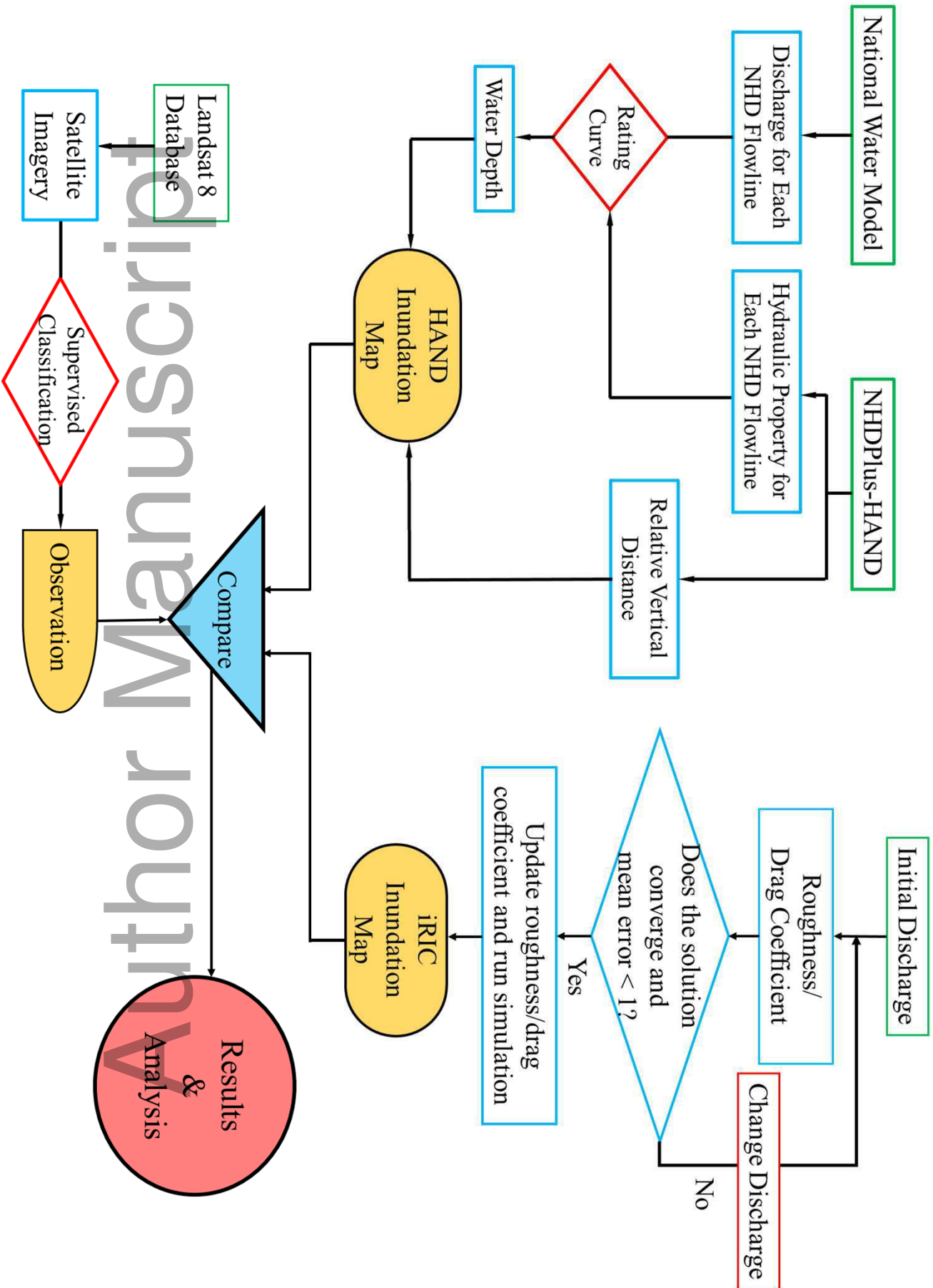


Figure 5

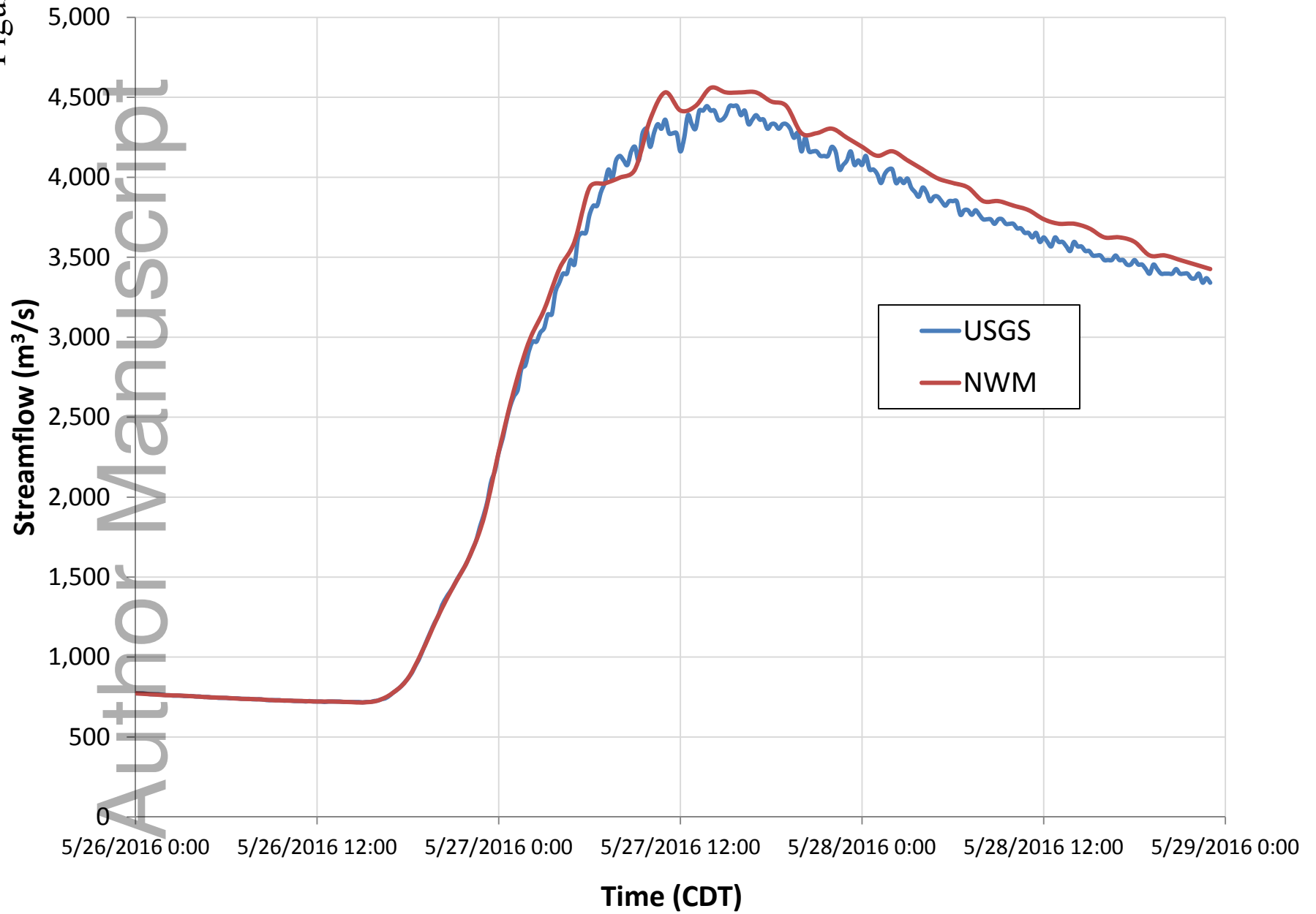


Figure 6

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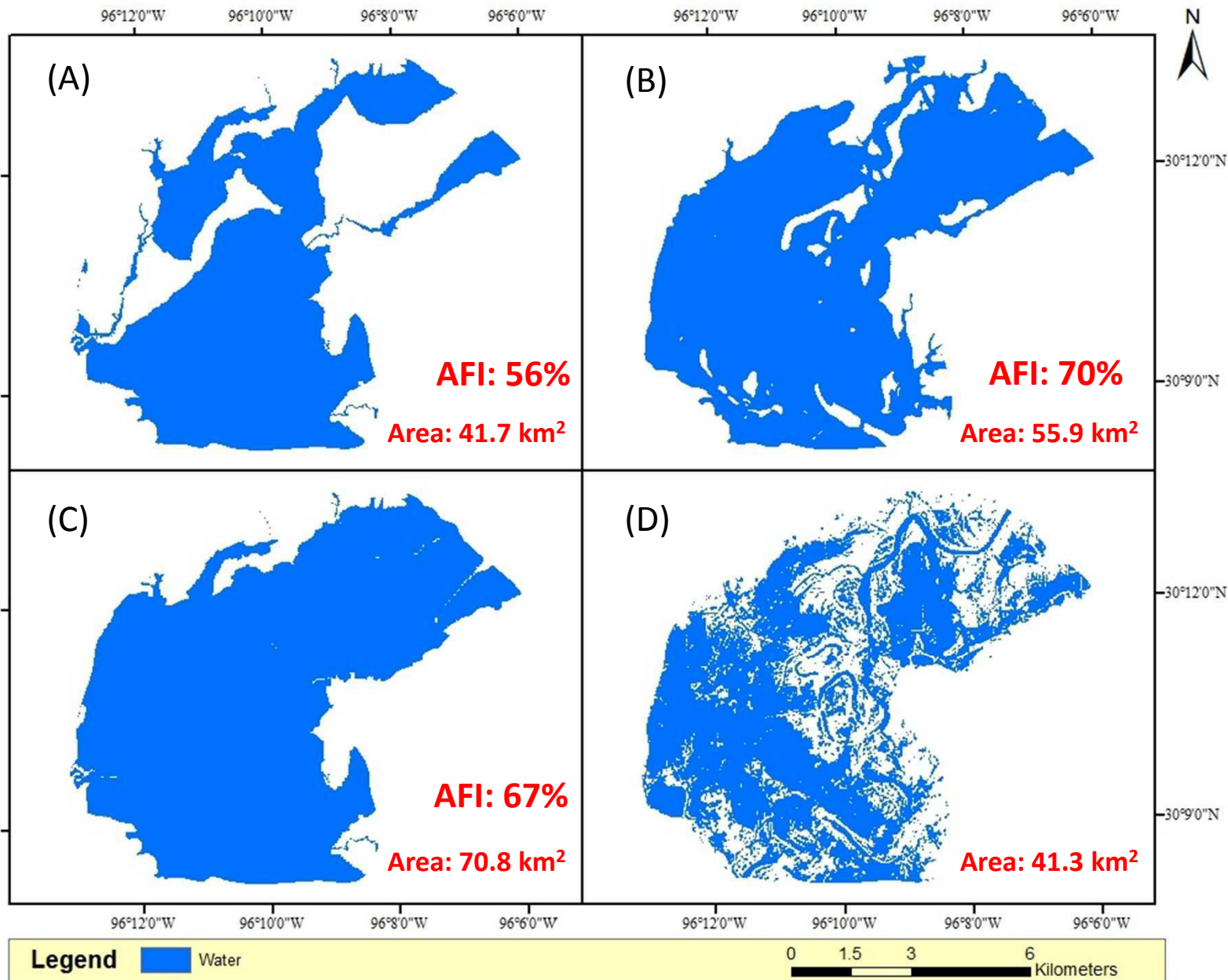
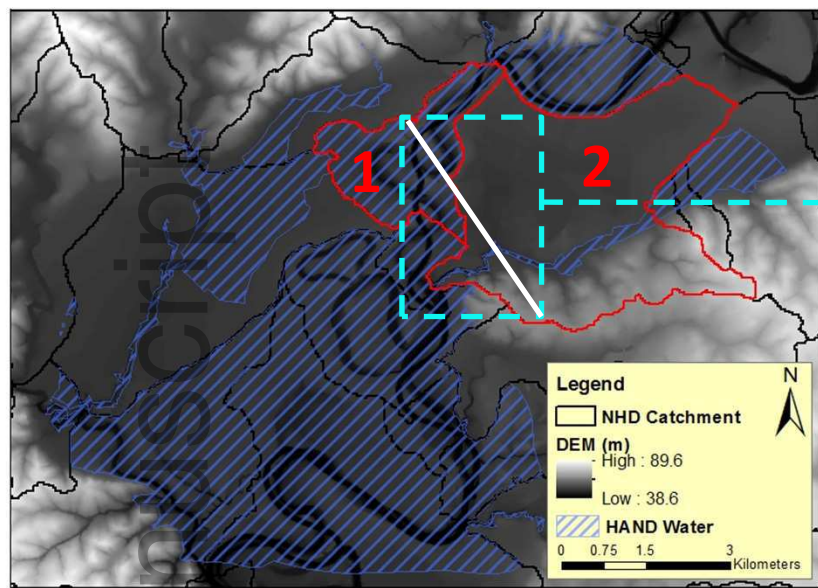
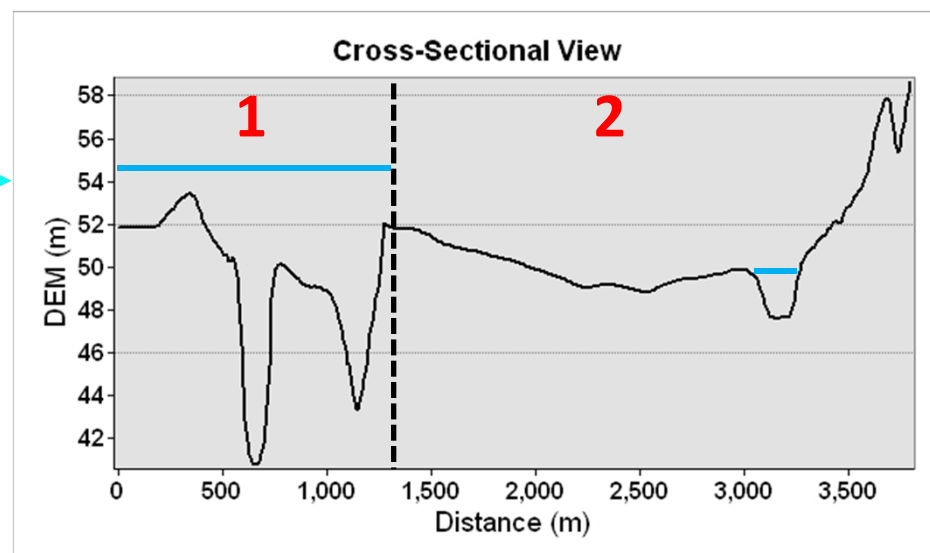


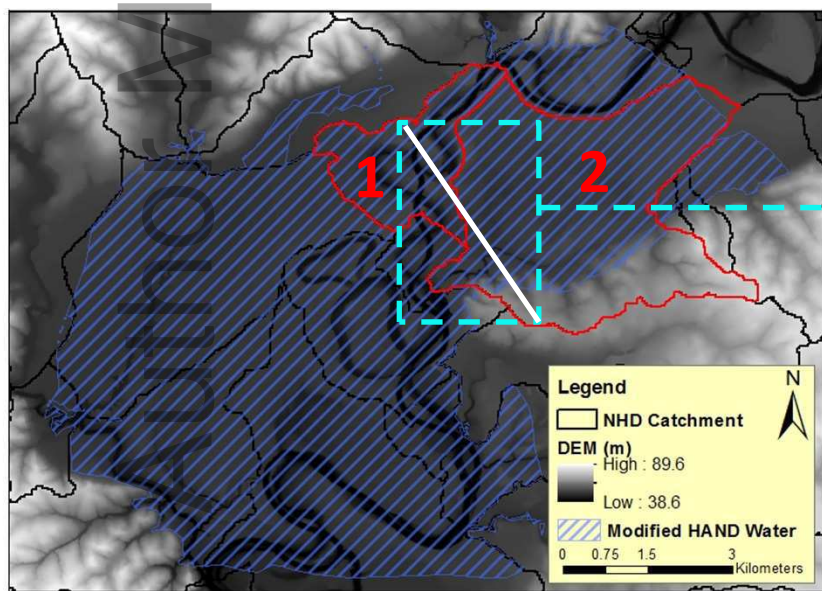
Figure 7



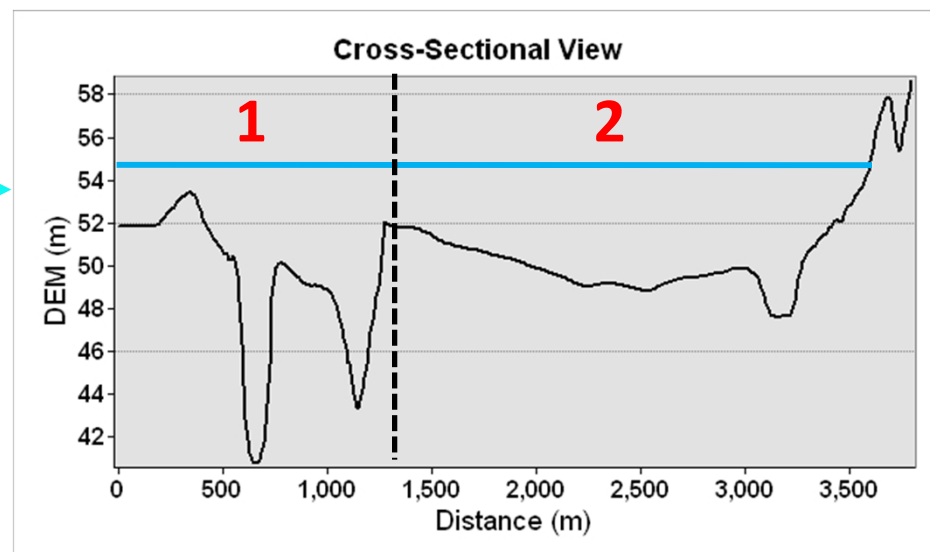
(A)



(B)



(C)



(D)