

Flood Inundation Mapping and Hazard Assessment for Mitigation Analysis of Local Adaption Measures in Upper Ping River Basin, Thailand

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2 **Flood inundation mapping and hazard assessment for mitigation analysis of**
3 **local adaption measures in Upper Ping River Basin, Thailand**

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ABSTRACT

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Annual monsoon flooding phenomena have caused disastrous impacts on the Upper Ping River basin and its inhabitants over the years. The existing administration set-up for flood mitigation and adaptation measures lacks effective utilization of locally available resources for complete flood protection. This study addressed this gap by flood hazard assessment at a lower administration scale (sub-district level) and performance evaluation of local adaptation measures was performed. 1D and 2D hydrodynamic models were developed and calibrated against observed discharge and water level (1D) and flood extent (2D), respectively. Flood inundation and hazard maps were reproduced and categorized into different classes based on defined critical depths for 2, 5, 10, 25, 50 and 100 years return periods. A maximum flood inundated area of 996.9 km² (3.94% of total basin area) was simulated in a 100-year return period. The flood hazard results represent that the largest flooded area categorized under “high hazard”, followed by “very high hazard” and “low hazard” categories, and least flooded area was classified under “medium hazard” category. The current administration set-up for flood adaptation and mitigation needs to update based on an integrated flood management approach to improve its effectiveness for future flood protection.

38 Keywords: Flood hazard; Mitigation; Upper Ping River Basin; Adaptation; Flood inundation

39 Introduction

40 Annual pluvial and fluvial floods occurred during monsoon period in Thailand leads to disastrous
41 impacts on people, their properties, government and commercial infrastructure, and socio-
42 economic activities (Gale and Saunders 2013; Jular 2017). Besides these impacts, the floods
43 severely disturbed daily routine life of local people in rural and urban areas at a larger scale.
44 Generally, the flooding phenomena occurs in this region because of heavy monsoon rainfalls,
45 climate change impacts, conventional flood control and management strategies and resources,
46 topographical characteristics of floodplain areas and low level of education and technical training
47 of local residents about flood disaster preparedness (Haraguchi and Lall 2015). Considering local
48 structural and non-structural flood management resources, there is need to improve local flood
49 control and management framework based on available resources in order to improve its
50 effectiveness with the help of combined effort of all available resources.

51

52 Upper Ping River basin, a sub-region of Chao Phraya basin located in north of Thailand is home
53 to about ~11.5 million population, rice dominated region with a high economic growth rate. Large
54 fraction of population (~90%) in the basin lives in rural areas and exposed to river flooding due to
55 storage of flood water in rice fields (UN 2003; Reda et al. 2012; Komsai et al. 2016). The flooding
56 in Ping basin is associated to heavy monsoon rainfall, tropical storms, typhoons, and land-use
57 changes (Wood and Ziegler 2008; Gale and Saunders 2013; Komsai et al. 2016; Takata and
58 Hanasaki 2020). Available literature on climate change assessment in the basin though very
59 limited, reports significant increase in temperature extremes and insignificant decrease in mean
60 and extreme precipitation in the basin (Sharma and Babel 2013; Masud et al. 2016; Saengsawang
61 et al. 2017). Several other studies on floods highlight the occurrence of extreme events in past in
62 1994, 1995, 1996, 2001, 2003, 2005, and 2011 (Wood and Ziegler 2008; Gale and Saunders 2013;
63 Komsai et al. 2016). For example, the 2011 flood in Thailand caused heavy damage to agriculture,
64 transport, fishery, tourism and industry. The overall economic losses in all sectors amount to
65 around \$200 million (Taesombat and Sriwongsitanon 2010). Such huge losses necessitate
66 evaluation of the hazard potential to flood depth and spatial extent.

67

68 Flood hazard potential can be mapped in multiple ways such as by conducting field surveys for
69 historical flood investigations or using remotely sensed imagery, employing hydrological and
70 hydrodynamic models, or both (Shehata and Mizunaga 2018). However hydrodynamic modeling
71 approach is more promising, it uses terrain data to simulate spatiotemporal variations of depth of
72 flood water (hazard potential) (Yalcin 2020). Hydrodynamic modeling can be one-dimensional
73 (1D) or two dimensional (2D) (Abdessamed and Abderrazak 2019; Mahmood et al. 2019; Nastiti
74 et al. 2018; Masood and Takeuchi 2012). One dimensional is most commonly used for flood hazard
75 mapping however possess one major limitation that it calculates spatial extent of the flood water
76 between two cross-sections using linear interpolation that introduces uncertainty in the interpolated
77 maps (De Kok and Grossmann 2010; Sarhadi et al. 2016; Yin et al. 2013; Zhang et al. 2021). Two
78 dimensional models have capability to overcome this problem using sophisticated numerical
79 methods such as finite difference and finite element. They can calculate flood depth, spatial extent
80 and velocity for each node and time step (Bates and De Roo 2000; Mani et al. 2014; Patel et al. 2017;
81 Farooq et al. 2019).

82

83 Flood adaptation and mitigation measures are normally applied under different administration
84 scales (e.g., National Government Level, Local Government Level, Local Community Level) in
85 Thailand and mostly they only work within their jurisdiction. This conventional flood control and
86 management set-up are not effectively working for potential flood risk reduction over the
87 vulnerable floodplain areas, as mostly, one administration department or stakeholder completes
88 his assigned task only, due to which the flood risk is transferred or redistributed to neighboring
89 regions or lower floodplain areas (Lebel et al. 2011; Manuta et al. 2006; Lebel and Sinh 2009;
90 Nikitina 2007). Considering this situation, there is a need to re-think about flood control and
91 management strategies that should cover this deficiency and produce combined effect of all
92 resources for potential flood risk reduction over the region. To cover this gap, this study developed
93 and applied 1D-2D hydrodynamic models for flood inundation mapping and hazard assessment
94 for evaluation of available mitigation options in Upper Ping River basin.

95

96 Synthesis of the literature available on flood hazard mapping and evaluation of mitigation options
97 in Ping River basin reveals a major drawback. The scope of almost all the studies was limited to
98 assess flood characteristics using simple rainfall-runoff and 1D hydrodynamic models (e.g., Wood

99 and Ziegler 2008; Mapiam and Sriwongsitanon 2009; Sriwongsitanon 2010; Boonrawd and
100 Jothityangkoon 2015; Komsai et al. 2016) and no study has yet been conducted for flood hazard
101 mapping at sub-district scale for improvement in potential of locally available mitigation options.
102 The authors argue that “integrated flood management framework” is the absolute omission of
103 current flood adaptation and mitigation application system that can mislead future flood mitigation
104 planning and limit the efficiency of flood protection system. This is actually happening in the basin
105 during past one decade and bringing public concerns about the performance of the existing
106 protection measures (Jarungrattanapong and Manasboonphempool 2011). Since integrated flood
107 management approach has potential to boost the current flood protection system and its impacts
108 would be beneficial for local residents for future floods.

109
110 Considering its importance this study evaluates the flood hazard potential at sub-district level in
111 Upper Ping River basin using 1D and 2D hydrodynamic models for mitigation assessment of local
112 available flood mitigation resources and recommendations for these resources for improvement in
113 future floods. The paper is divided into five sections; section 1 details characteristics of the Upper
114 Ping River basin; section 2 comprehends materials and methods; section 3 describes results and
115 discussion parts, and section 4 includes conclusions.

116

117 **Study area and data description**

118 The Ping river is one of the main tributaries among four rivers of Chao Phraya river that originates
119 from Chiang Dao district in the north of Thailand, flows through the central and lower region of
120 Thailand before discharging into the Gulf of Thailand. The Upper Ping River basin is located
121 between Wang and Salween river basin of Myanmar (Jothityangkoon et al. 2013). The Basin
122 covers a catchment area of 25300 km² with several mountain ranges in Chiang Mai and Lamphun
123 provinces (Figure 1). The Ping river passes from north to south in the center of Chiang Mai city,
124 the most popular tourist place in the northern region of Thailand. The Bhumipol Dam that is
125 constructed in Tak province, divides the upper and lower ping river basin, has a maximum storage
126 capacity of 13462 million m³. The basin generates an average annual runoff of 6815 million m³
127 (Mapiam and Sriwongsitanon 2009) with average annual rainfall and potential evaporation of
128 1174mm and 1661mm. Most of the basin’s climate is tropical and has a long rainy season (May –
129 October), and 90% of the annual rainfall occurs during rainy season. The weather of the study

130 basin is influenced by the Southwest and Northeast monsoon. The topography of the basin is based
131 on steep and complex mountainous regions with elevation ranges from 224m to 2575m. The
132 highest proportion of land-use consists of forest (70%), followed by urban (12%), and only 6%
133 area is covered under plantation. Multiple flood episodes have been observed over the last two
134 decades during the monsoon period in the Upper ping basin including 1994, 1995, 2001, 2005,
135 2009, 2010, and 2011 with the highest flood level of 4.94m in 2011 year at P.1 station (Boonrawd
136 and Jothityangkoon 2015).

137 The basin has a well-distributed network of rain gauges (90) and streamflow stations (44), but
138 many of them are not functional or do not have sufficient data, and only the meteorological and
139 hydrological data from 17 rain gauges and 2 flow gauges (respectively) have used for the flood
140 hazard assessment and recommendations of appropriate flood mitigation and adaption measures
141 at this region that could be practically implementable considering locally available resources, as
142 represented in Figure 1. The detailed description of meteorological and hydrological stations is
143 mentioned in Table 1 and Table 2. The digital elevation model (DEM), which is used to prepare
144 for floodplain bathymetry for the 2D-hydrodynamic model, is downloaded from
145 <https://earthexplorer.usgs.gov/> with a resolution of 30m × 30m. Due to the larger floodplain area
146 and to optimize computational time, the 30m dem resolution was resampled into 85m for the
147 preparation of floodplain bathymetry.

148

149 [Figure 1.]

150

151

152 [Table 1.]

153

154 [Table 2.]

155 **Methods**

156 **Methodological Framework**

157 The methodology of the study based on the development of hydrological and hydrodynamic
158 models, forcing of input data, followed by simulations, calibration and validation, and analysis of
159 post-simulation flood results and its evaluation for further scenarios. The methodological
160 framework consists of three sections (a) 1D-2D coupled model development, (b) 1D and 2D model
161 calibration and validation, and (c) flood hazard assessment and mitigation measures. Figure 2
162 represents the step-by-step methodological framework adopted in this study.

163 [Figure 2.]

164 1D Hydrodynamic Model Set-up

165 The 1D hydrodynamic model (MIKE 11) applied in this study, which has six editors are mutually
166 interlinked with each other and used for river model development by defining sub-basins, cross-
167 sections, boundary conditions, network layout, hydrodynamic parameters and simulation
168 specifications (Tansar et al. 2020). All editors of the MIKE 11 model are dynamically interlinked
169 and controlled by the simulation editor. Also, this editor used to define simulation specifications
170 including simulation model, simulation mode, runoff and hotstart files, simulation time-period and
171 time-step, and saving time-step. The river network is defined in network editor. In upper ping river
172 basin, the ping river is the main tributary that starts from Doi Thuai in Chiang Dao district, passes
173 from Chaing Mai, Lamphun and Tak provinces and inflowed to Bhumipol Dam. The main rivers
174 including Ping river, Mae Nam Taeng, Mae Nam Rim, Mae Nam NGAT, Mae Nam Kuang, Mae
175 Wang, Mae Klang, Mae Nam Tha and Mae Li's have defined with their total river lengths of 283
176 km, 155 km, 56 km, 82 km, 115 km, 116 km, 46 km, 78 km and 212 km, respectively. The river
177 geometry is defined as X-Z coordinates (from left bank) in the cross-section editor and the mean
178 distance between two adjacent cross-sections was 2 km. The 19 sub-basins with corresponding
179 rainfall stations (17), and rainfall and evaporation data are defined in runoff editor for runoff
180 generation. The simulated runoff of each sub-basin is laterally distributed along the corresponding
181 river in network editor, where the runoff and network editors are dynamically interlinked by
182 defining sub-basins along with corresponding rivers with upstream and downstream boundaries.
183 The upstream river boundaries as discharge are defined at "0" chainage and only one downstream
184 boundary as water level is defined at the end of Ping river (P.73 station, Figure 1). The
185 hydrodynamic parameters including water depth, discharge, bed resistance, and output parameters
186 with their initial conditions are defined in the hydrodynamic editor. The bed resistance is the main

187 parameter that is defined as local (horizontal and vertical) and global roughness coefficient
188 (Manning' n) for discharge and water level calibration.

189 **2D Hydrodynamic Model Set-up**

190 The floodplain boundary is defined in MIKE 21 model as rectangular bathymetry with 85 m × 85
191 m resolution, considering the river network and 100-year observed flood extent. With this
192 resolution, the 2D hydrodynamic model simulates discharge and water level at every 85 m
193 distance. Apart from bathymetry as the main input parameter, several other parameters including
194 boundaries, simulation specifications, initial surface elevation, flood and dry depths, resistance
195 and eddy viscosity and infiltration are required for flood inundation simulation. Only the flood
196 plain resistance as roughness coefficient (n) of 0.04 was used for flood extent calibration and
197 remaining parameters were considered as default. The simulation time step and results saving time
198 step were 5 seconds and 12 hours, respectively.

199 **1D-2D Coupled Modeling Package**

200 The calibrated 1D-hydrodynamic model (MIKE 11) and 2D-hydrodynamic model (MIKE 21) are
201 dynamically coupled through MIKE Flood to reproduce flood inundation results. In order to
202 capture flood inundations in MIKE Flood, the weir equation is defined at the lateral links between
203 both river banks (left and right) and floodplain, and the flood volume is overflowed to the
204 floodplain through these links when the water level rises over the bank elevation in the 1D-
205 hydrodynamic model. The flood extent and depth are based on the magnitude of flooded volume,
206 overflow location and topography of floodplain located around both river banks. In addition, the
207 overflow can travel from river to floodplain and vice versa in MIKE Flood based on the
208 topographical characteristics of the river banks, floodplain and the magnitude of flooded volume.

209 The detailed theoretical background of 1D-hydrodynamic model (MIKE 11), 2D-hydrodynamic
210 model (MIKE 21) and coupling package (MIKE Flood) are explained in Tansar et al., 2020.

211 **Flood Frequency Analysis**

212 The Gumbel probability distribution was applied on meteorological, hydrological and water level
213 data of past 30 years to find the extreme values for different return periods. The gumble distribution
214 was selected among other probability distributions (e.g. normal, log-normal, Pearson type-III, log-

215 Pearson Type-III and extreme value Type I (EV I) because of its better performance and it has
216 been extensively applied and reported over the last few years in different published case studies of
217 Thailand (Tingsanchali and Karim 2010; Shrestha and Lohpaisankrit 2017; Tansar et al. 2020)

218 **Gumbel Distribution**

219 The cumulative distribution function (cdf) of the of Gumbel distribution (maximum) is defined as
220 represented in Eq. (1).

221

$$222 \quad F(x) = \exp[-\exp(-y)] = e^{-e^{-y}} \dots (1)$$

223

224 Where function $F(x)$ denotes the probability distribution function of random variable x and y
225 represents the reduced variable.

226 Eq. (2) denotes the relation between probability of exceedance of excluded events and return
227 period (T).

228

$$229 \quad F_1(x) = 1 - F(x) = \frac{1}{T} \dots (2)$$

230 Eq. (3) is established by following Eq. (1) and (2).

231

$$232 \quad \frac{1}{T} = 1 - e^{-e^{-y}} \dots (3)$$

233

234 The reduced variable y is calculated by using Eq. (4).

235

$$236 \quad y = -\ln \left\{ \ln \left[\frac{T}{(T-1)} \right] \right\} \dots (4)$$

237

238 The statistical variate is computed with Eq. (5).

239

$$240 \quad y = \bar{y}_n + K \cdot \sigma_n \dots (5)$$

241

242 Eq. (6) is applied to calculate frequency factor (K) in which \bar{y}_n is a Gumbel's reduced mean
243 variable and σ_n is the standard deviation.

244

$$245 \quad K = \frac{y - \bar{y}_n}{\sigma_n} = \frac{\left(-\ln\left\{\ln\left[\frac{T}{(T-1)}\right]\right\}\right) - \bar{y}_n}{\sigma_n} \dots (6)$$

246

247 More technical details of Gumbel distribution method and its application can be found in Onen
248 and Bagatur 2017.

249 **Design Peak Hydrograph**

250 Design peak hydrographs of meteorological and hydrological data of different return periods are
251 derived based on the annual maximum values estimated by frequency analysis of past 30 years of
252 measurements. The design peak flood hydrographs of different return periods (2, 5, 10, 25, 50, and
253 100-year) are developed based on non-dimensional hydrographs (Tingsanchali and Karim 2010).
254 The 2011 flood year is selected for the development of designed hydrograph. A single annual
255 maximum value is provided by frequency analysis of selected flood year that is used to generate
256 the continuous time series of the entire flood year. The time series of selected flood year is divided
257 by maximum observed value to normalize the non-dimensional hydrograph. Following this
258 procedure, the peak of the non-dimensional value should be equal to unity. The time series of non-
259 dimensional hydrograph is multiplied with the annual maximum value to generate the designed
260 peak flood hydrograph. The same procedure was repeated for each return period for the
261 hydrographs' construction of all stations and boundary conditions (Brunner et al. 2016; Mediero
262 et al. 2010).

263 **Categorization of Flood hazard Assessment**

264 The flood hazard has been classified into different categories for spatial flood risk assessment, and
265 this information leads to in-depth understanding of severity of flood problems that can be avoided
266 in future by following proper flood mitigation and adaptation measures (Shrestha and
267 Lohpaisankrit 2017). Multiple factors such as flood depth, duration of flood and flood velocity are
268 considered for estimation of flood hazard assessment during flood study. These parameters are

269 considered based on the topographical characteristics of study area, especially of floodplain
270 regions and also flood conditions. The present study only consider flood depth for classification
271 of flood hazard as the topography of flooding region is quite flat and flood wave velocity can be
272 negligible.

273 The detailed classification of flood hazard categories along with their indexes and description has
274 been presented in Table 3. The critical flood depths 0.8, 1.0 and 3.5m that applied for classification
275 of flood hazard assessment established based on the floor level of local buildings, neighboring
276 commercial infrastructure, and also elevation of single-storey building (Tingsanchali and Karim
277 2005, 2010). The present study adapted liner scale among linear, geometric, and exponential scales
278 for hazard classification as the linear scale has performed the best among others and also has been
279 recommended by previous published research studies. Eq. (6) is used to calculate hazard factor
280 H_F .

$$281 \quad H_F(i) = \left(\sum_{j=1}^{N_D} A(i,j) H_I(j) \right) / \sum_{j=1}^{N_D} A(i,j) \dots (7)$$

282 Where, i , j , N_D and H_I represent land unit identification number, depth category, total number of
283 depth categories and hazard index, respectively, whereas $A(i,j)$ belongs to land unit i under depth
284 category j .

285 [Table 3.]

286 Model Performance Evaluation

287 The study considers some of the most commonly used statistical parameters for the model's
288 performance evaluation. The hydrological model performance for discharge simulation was
289 evaluated using the Coefficient of Correlation (R), Nash Sutcliffe Efficiency (NSE), and Root
290 Mean Square Error (RMSE), and hydraulic model performance for the water level simulation was
291 evaluated using the Coefficient of Correlation (R), Relative Peak Error (RPE) and Volume Bias
292 (VB). Further detail on these statistics is reported in Shrestha and Lohpaisankrit 2017; Leta et al.
293 2018.

294 Results and Discussion

295

296 Calibration of 1D-hydrodynamic model

297 Discharge and water level in 1D-hydrodynamic model were calibrated and validated for the period
298 of 4 years (2006-2009) and 2 years (2011-2012), respectively. The discharge and water level
299 stations located along main river (Ping river) are represented in Figure 1. The calibration
300 parameters presented in Table 4 tuned to improve hydrological model performance, and global
301 and local manning's 'n' values were adjusted for improvement of model's hydraulic performance.
302 The global and local manning's 'n' values are differentiated based on their "effecting scale", the
303 adjustment of global value has uniform effect on all river network, however, the local manning's
304 'n' values are defined for adjustment of flow by changing the roughness of specific rivers or section
305 of river. The manning coefficient was adjusted between 0.29-0.50 during model calibration and
306 validation. Table 5 represents three statistical parameters (R, NSE, RMSE) for performance
307 evaluation of 1D hydrodynamic model for discharge simulation, and model' performance was
308 reasonably acceptable for calibration and validation at both stations, P.1 and P.67. The discharge
309 calibration and validation results are shown in Figures 3a and 3b, and 4a and 4b, respectively.
310 Generally, the simulated discharge hydrograph is well-matched with observations at both locations
311 (P.1 and P.67) for low and high periods during calibration, however, simulated discharge peak was
312 relatively underestimated for 2006 and 2009 years at both locations. Similar results pattern was
313 noticed at both locations during discharge validation. In case of water level calibration and
314 validation results, the model performance was relatively better at P.1 station compared to P.67, as
315 statistical parameters can be seen in Table 6. It might be because P.67 station is located at just
316 downstream of Mae Nam NGAT channel that starts at downstream of NGAT Dam, so the
317 simulated water level at P.67 station may not fully capture the controlled flow's pattern as it was
318 in real condition. However, the model performance is still quite acceptable and reliable for flood
319 inundation modeling at a larger scale considering limited data availability with coarser river
320 geometry. The water level calibration and validation results have been represented in Figures 3c
321 and 3d, and 4c, and 4d, respectively. Multiple sources of uncertainties that exist and affect the
322 accuracy of river modeling results have been previously discussed for particularly in this region
323 by the author in Tansar et al., 2020.

324

325

[Figure 3.]

326 [Figure 4.]

327 [Table 4.]

328 [Table 5.]

329 [Table 6.]

330 Calibration of 2D-hydrodynamic model

331 The 2D-hydrodynamic calibration was performed for the flood year 2011 by using MIKE Flood.

332 The 1D-2D hydrodynamic models were simulated for a period of 13 days (September 27, 2011 to

333 October 10, 2011) to reproduce flood inundation maps. The remotely sensed observed flood map

334 of 2011 year obtained from the Geo-Informatics and Space Technology Development Agency

335 (GISTDA), Thailand and compared with simulated inundation extent, as shown in Figure 5. The

336 area of observed and simulated flood inundation extent was 550.6 km² and 489.6 km², respectively.

337 Generally, the observed and simulated flood inundation extent reasonably matched well

338 throughout the inundated area. The flood inundations produced along the river network including

339 Ping river, Mae Nam NGAT, Mae Nam Taeng, Mae Wang, Mae Klang and Mae Nam Tha have

340 represented satisfactory better performance against the observed flood map. However, there are

341 some differences noticed between the observed and simulated extent around the lower section of

342 Mae Nam Kuang river and the observed flood extent relatively has a larger inundation area

343 compared to the simulated. It is reasonable to explain that the area along Mae Nam Kuang river in

344 Ban Thi and Mueang Lamphun districts also have a small network of small streams, located along

345 the Mae Nam Tha river that transferred the monsoon flow from the north-eastern part of this region

346 to the south-western areas. Occasionally, these streams also flooded during the monsoon period

347 and it might be possible that the simulated flood extent which does not cover by the overflow of

348 Mae Nam Tha river occurred because of these local streams. In addition, some cross-sections of

349 the rivers were taken from DEM (30m resolution) that could also lead to produce uncertainty in

350 flood inundation results. During this calibration process, the flood depth could not be calibrated in

351 the floodplain because of the unavailability of flood depth measurements. The average flood depth

352 of all floodplain was 1.86m.

353

354 [Figure 5.]

355 **Flood inundation mapping**

356 The flood inundation maps of return period 2, 5, 10, 25, 50 and 100 years are analyzed to compare
357 spatial distribution of flood extent and depth in different districts located in upper ping river basin,
358 as represented in Figure 6. The total flooded area against return period 2, 5, 10, 25, 50 and 100
359 years are 601.8 km², 743.0 km², 811.4 km², 878.3 km², 935.6 km² and 996.9 km² (3.94% of total
360 basin area), respectively, as represented in Table 7. Based on flood inundation results, some plain
361 areas are inundated along with the Ping, Mae Nam NGAT and Mae Nam Taeng rivers in upstream
362 districts (Chaing Dao, Phrao, Mae Taeng and Mae Rim) for the above-mentioned return periods
363 and the flood extent did not substantially expand during simulation of higher return periods (25,
364 50, 100 years), but the significant increase in flood depth was observed. It happened because the
365 topography of the neighboring areas around the river is plain and surrounded by mountains. The
366 major flood inundations took place in central part of upper ping river basin, and flood extent and
367 depth also significantly increased during higher return periods. The Ping and Mae Nam Kuang
368 rivers and some local streams located in central region overflowed and caused flooding in nine
369 districts (Mae Rim, San Sai, Hang Dong, San Pa Tong, Mueang Lamphun, Saraphi, Ban Thi, San
370 Kamphaeng), including in Chiang Mai district which is a major tourist place in northern part of
371 Thailand. The Ping river passes at the center of Chaing Mae district and the community and local
372 tourist spots are highly vulnerable to flood risk during peak monsoon period, especially, when the
373 upstream areas of upper ping basin receive continuous rainfall events followed by local rainfall
374 events in central regions. In such situation, the capacity of Ping river is unable to hold the flood
375 volume within the embankments and the neighboring areas are inundated. In case of lower part,
376 the substantial increase in flood extent and depth was observed from lower to higher return periods.
377 Approximately, the river Mai Li's was not flooded in lower return periods (2 & 5 years) in Ban
378 Hong and Li district and inundations made in neighboring areas with about 1m depth during higher
379 return periods. The famous bigger Doi Tao lake is located in Hot and Doi Tao districts, in the
380 lower region of upper ping river basin. The flood volume that passed from central region to lower
381 region stored in Doi Tao lake with higher flood extent and depth (more than 4.5m) and this area is
382 almost less populated and not as much of susceptible to economic losses because of flooding.

383 **[Figure 6.]**

384 Flood hazard assessment

385

386 The flood hazard categories are classified based on criteria of the critical depths, as introduced by
387 Tingsanchali and Karim 2010, and a detailed description of flood hazards has been described in
388 Table 3. The flooded area of 2, 5, 10, 25, 50 and 100 years return periods are categorized under
389 different flood hazard levels in Table 7. The classification of flood results represents that the
390 overall major flood extent occurred under the “high hazard” category during all return periods,
391 followed by “very high hazard” and “low hazard” categories, and the least flooded area was classed
392 under “medium hazard” category. Besides the “very high hazard” category, the flood extent of
393 other categories was not significantly changed during all return periods, however, the major
394 increase in the flood extent area of “very high hazard” category observed from lower return period
395 (2-year) towards higher return period (100-year). During a 100-year return period, about 1.74 %
396 (439.8 km²), 1.51% (382.9 km²), 0.54 % (137.6 km²) and 0.14 % (36.6 km²) of the total flooded
397 area are classified under very high hazard, high hazard, low hazard, and medium hazard category,
398 respectively. Based on the flood hazard description in Table 3 and flood hazard results, most of
399 the local community and property existed in flood inundated areas are highly vulnerable to flood
400 damages during all return periods, and following by the inundated areas that can have severe flood
401 damages during higher return periods floods, because the flooding depth would be greater than
402 3.5m, classified as “extreme danger zone”. In case of low and medium flood hazard levels, the
403 flooding water can enter surrounding buildings and can affect the daily routine activities of local
404 residents. In case of Thailand, the local administration departments, department of drainage and
405 sewerage, and department of disaster prevention and mitigation together with local residents take
406 some preliminary actions before monsoon season to avoid flood hazards and reduce the flood
407 damages to local infrastructure. However, the potential of such flood mitigation measures is
408 limited and can only help to reduce the flood impacts of low and medium hazard levels and need
409 to improve further for mitigation of higher hazard levels. The government needs to plan and
410 implement the state-of-the-art telemetry system in northern areas of Thailand that can help in flood
411 early warning system, real-time flood forecasting and decision support system for dissemination
412 of flood warning alarms to local community before flooding, as a precautionary step for flood
413 mitigations.

414 Figure 7 represents the flood inundation area against 2, 5, 10, 25, 50 and 100 years return periods.
415 The flood inundation area is linearly increased against return periods from lower to higher return
416 periods floods.

417 [Table 7.]

418 [Figure 7.]

419 **Flood hazard classification at sub-district scale**

420 The 159 sub-districts located in upper ping river basin of Chiang Mai, Lamphun and Tak provinces
421 are classified into different flood hazard categories based on average flood depth of each sub-
422 district. Figure 8 represents the flood hazard zoning at sub-district scale (the smallest
423 administration unit in Thailand) for 2, 5, 10, 25, 50 and 100 years return periods. Overall, the
424 maximum sub-districts lie under “high hazard”, followed by “very high hazard” and “medium
425 hazard” categories, and very few sub-districts classed under “low hazard” category during all
426 return periods. Generally, the assessment of classification concludes that most of the region in
427 upper ping river basin might be expected to face flood depth of 1-3.5m followed by more than
428 3.5m, and the flooding probability of less than 1m is may possibly be less during monsoon season.
429 Specifically, the 122 sub-districts classified under high hazard category for a 2-year return period
430 and the number is gradually decreased to 104 for 100-year return period. In contrast, the number
431 of subdistricts increased from 2 to 37, for 2- to 100-year return period. However, the classified
432 sub-districts for low and medium hazard levels remained almost constant during all return periods,
433 and subdistrict numbers varied from 5 to 9 in low hazard and 10 to 15 in medium hazard category.

434 All sub-districts located along ping river categorized under high hazard level during each return
435 period. In case of higher return periods, the flood depth in neighboring areas of ping river (upper,
436 central and lower section) increased and the classification of sub-districts changed from high
437 hazard to very high hazard category. Further, the sub-districts located along with Mae Nam Tha
438 and Mae Li’s rivers were less anticipated to flooding and categorized under low and medium
439 hazard levels. It is also reasonable to explain that there was no flooding observed at the
440 downstream side of upper ping river basin during lower return periods (2 & 5 years). Based on
441 these results, the community and property existed along the ping river, especially in upper, central,
442 and lower sections, are highly vulnerable to flood risk during the monsoon season. In particular,

443 Chiang Mai city and its surrounding areas are also classed under high hazard flood; as this city is
444 received a lot of tourists during the whole year and most of the economic activities in this region
445 are directly or indirectly interlinked with tourism. The fluvial and pluvial flooding has also been
446 reported in recent years in Chiang Mai city (Bicksler 2019). Comparatively, the probability of
447 flooding in distant areas is very less.

448 [Figure 8.]

449

450 **Capacity assessment of flood adaptation and mitigation measures**

451 This section evaluates the potential of local flood mitigation and adaptation measures at three
452 administration scales (1) National Government Level (2) Local Government Level (3) Community
453 Level (Table 8) and discusses various gaps that existed in flood adaptation measures at institutional
454 level and community scale. Also, future flood management recommendations are suggested for
455 the study area considering flood inundation and hazard results described in previous sections.

456 Multiple structural and non-structural flood mitigation measures are applied under each
457 administration level during flooding period to reduce the flood impacts in rural and urban areas,
458 however, the upper ping river basin is still severely facing flooding problems in monsoon period.
459 The maximum capacity of Ping river at P.1 station is 460 m³/s, and additional discharge at this
460 location in monsoon period leads to flooding in downstream region especially Chiang Mai city
461 and its neighboring areas. However, to mitigate this problem, the “Ping River expansion project”
462 has been recently completed with its total length of 6km in Chiang Mai city. Indeed, the city
463 municipality has only considered river’s length within their jurisdiction because of fund limitation
464 and technically the flood risk now has been redistributed to downstream areas (Jarungrattanapong
465 and Manasboonphempool 2011). Based on river capacity and floodplain topography assessment
466 in 1D-2D hydrodynamic model, it is suggested to further expand Ping river’s capacity of 16km
467 following previous improved river length to potentially reduce future flood risk in this region.

468 The flood prediction capacity of currently available forecasting system is “seven hours” before it
469 hits the Chiang Mai City, and Hydrology and Water Management Center (HWMC) disseminate
470 flood warnings to relevant government and communities (Jarungrattanapong and
471 Manasboonphempool 2011). However, the current system needs sufficient improvement with

472 state-of-the-art real-time flood early warning system with high prediction accuracy and larger
473 flood forecasting capability. Following this, the flood alerts system also needs to be modified for
474 faster dissemination of flood warnings to all relevant stakeholders and local residents.

475 [Table 8.]

476 **Flood adaptation gaps and future recommendations**

477 Multiple gaps found in previous flood mitigation planning and management procedures of
478 government and local departments, and local stakeholders. Each department or community
479 completed their assigned tasks on an individual basis, either it is short-term or long-term, however,
480 now there is need to re-think and establish an integrated flood management framework in which
481 combined potential from all level of resources should be systematically managed and applied.
482 Moreover, the local community involvement is highly recommended and would be beneficial in
483 flood mitigation and adaptations process as their previous real flood experiences can help to develop
484 technically sound, practically applicable, socially acceptable, and economically viable structural
485 and non-structural flood mitigation strategies.

486 As local community now is habitual of annual floods, so sometimes they do not take flood
487 warnings seriously and also do not prepare themselves for disaster, resultantly, more tangible and
488 intangible flood damages have been occurred in past flood events. Moreover, some local residents
489 do not participate in individual and collective adaptation activities as they think that it is the
490 government's responsibility to protect people from flooding; there is need to educate and train
491 local community through disaster awareness and preparedness workshops to involve full human
492 resources potential during flood mitigation periods. A detailed practical research study is
493 recommended in future for the development of an integrated flood management framework with
494 the consideration of availability and limitation of local resources, administrative priorities and
495 conflicts, and preferences of local community' concerns. The study could provide guidelines to
496 update current administrative flood adaption and management set-up into an improved version for
497 better performance in future flood risk reduction.

498

499 **Conclusion**

500 Flood inundation mapping and hazard assessment of the Upper Ping River basin performed for
501 different return periods with 1D-2D coupled hydrodynamic modelling approach to categorize
502 flooded areas into different hazard levels at the sub-district scale (the smallest administration unit
503 in Thailand). This study contributes to better utilization of flood adaption and mitigation measures
504 taken by multiple administration authorities before, during, and after floods for the protection of
505 local people, their properties, and other infrastructure.

506 The flood inundation and hazard analysis results represent that the sub-districts located around
507 Ping River are highly vulnerable to future floods, during low or high return periods in monsoon
508 season. The maximum flood inundation occurred at an area of 996.9 km² (3.94% of total basin
509 area) during the 100-year return period. The sub-districts are classified under “high hazard”,
510 followed by “very high hazard” and “medium hazard” categories, and very few sub-districts
511 classed under “low hazard” category during all return periods. The severity of flooding problems
512 reduced for those sub-districts located at distant places from the river and vice versa.

513 Considering the flood problems, topographical characteristics of the basin, and locally available
514 flood mitigation resources, the current national and local administration needs to find an integrated
515 flood management approach with the better utilization of local resources to improve the current
516 flood protection system. The potential of existing flood early warning and information
517 dissemination systems needs to upgrade with state-of-the-art real-time flood forecasting and the
518 latest dissemination system.

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Figures

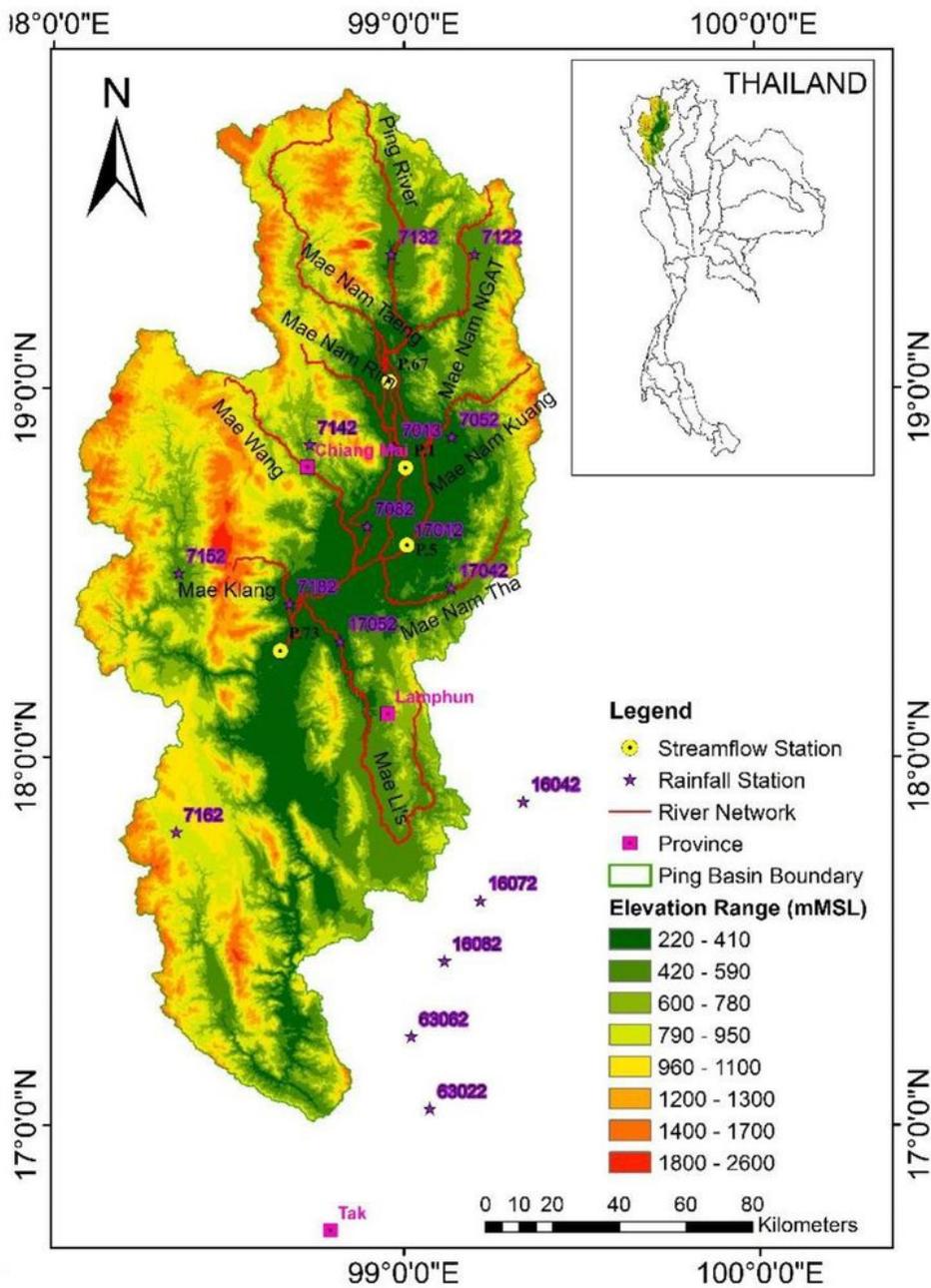


Figure 1

Figure 1

Location map of study area with hydro-meteorological stations in Upper Ping River Basin, Thailand Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country,

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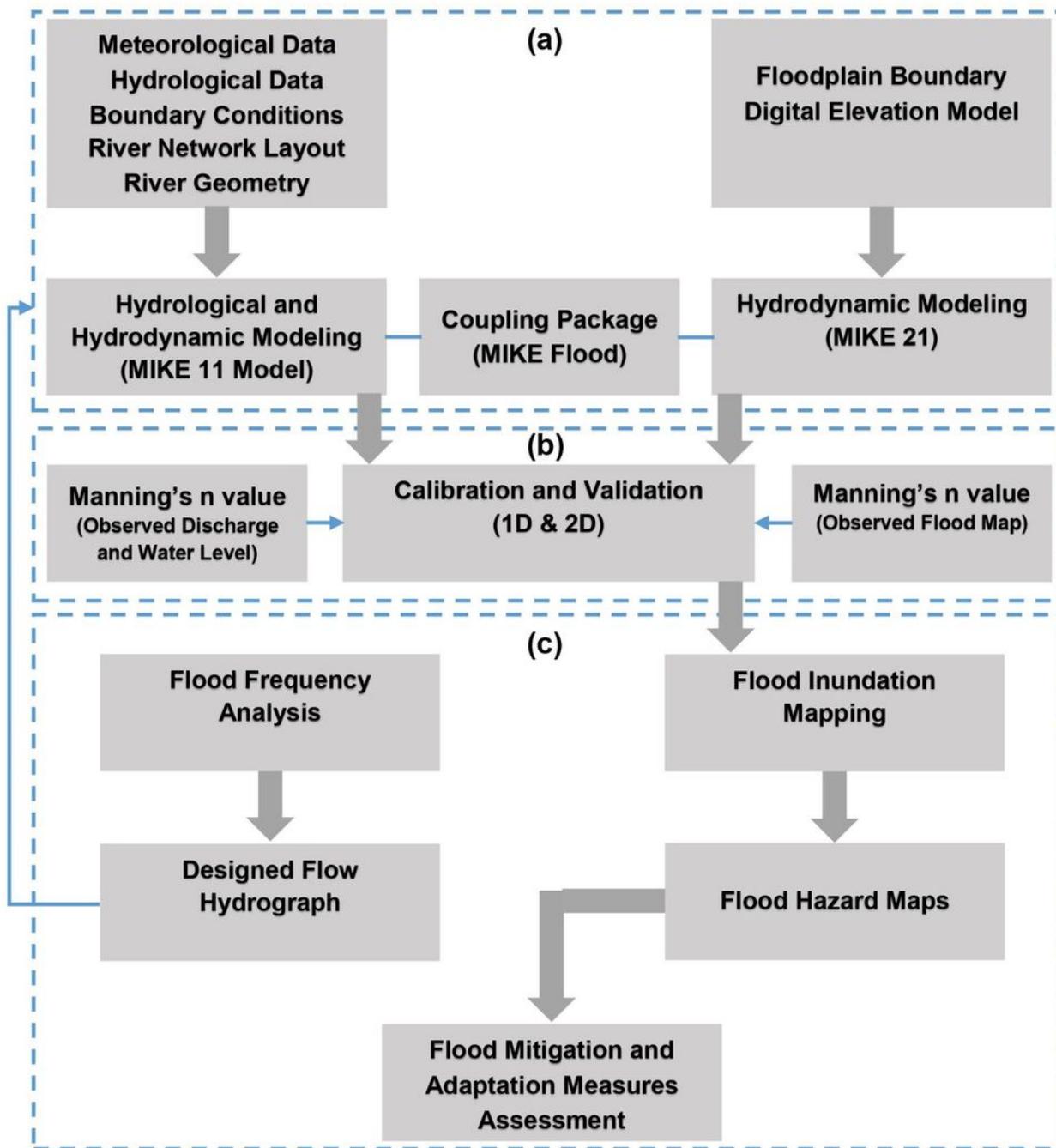


Figure 2

Figure 2

Methodological Framework for flood hazard assessment and adaptation measures

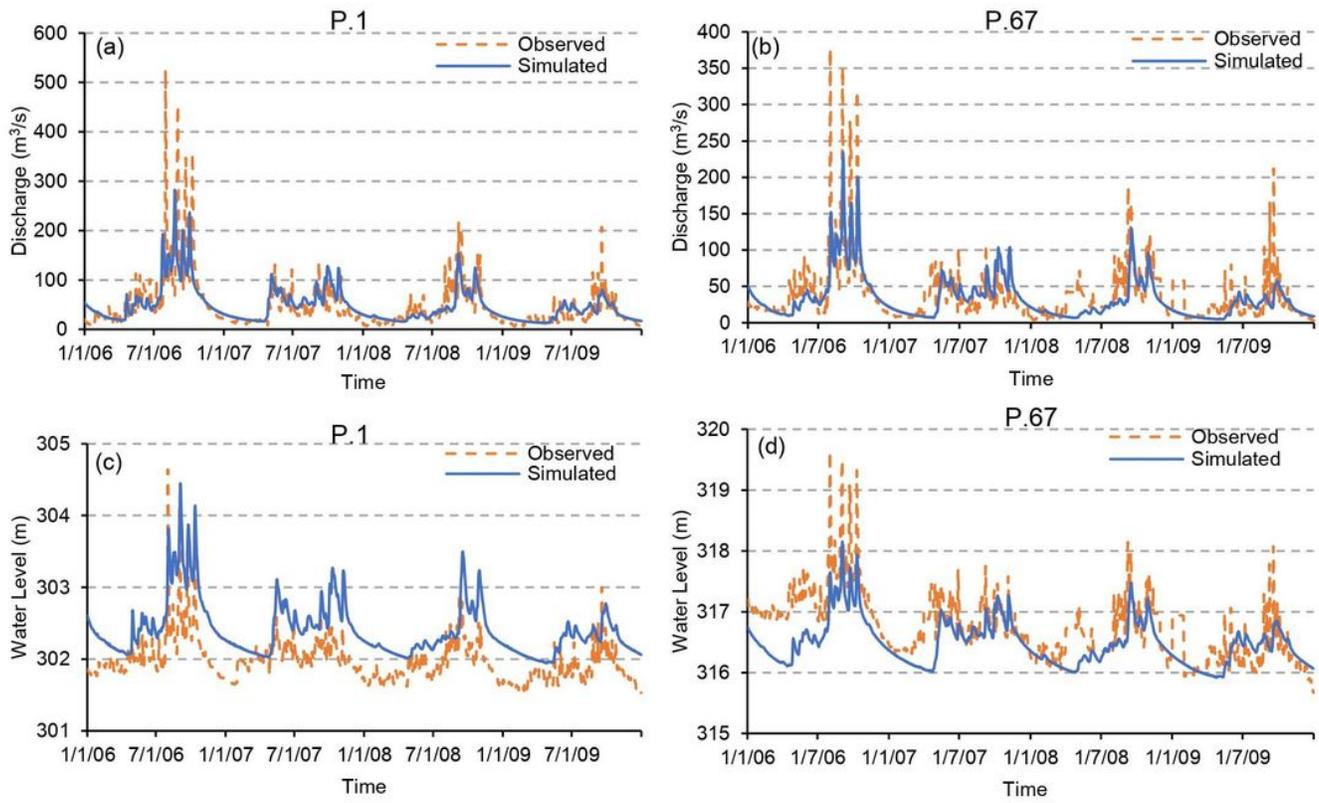


Figure 3

Figure 3

Calibration results of discharge at (a) P.1 and (b) P.67 and water level at (c) P.1 and (d) P.67

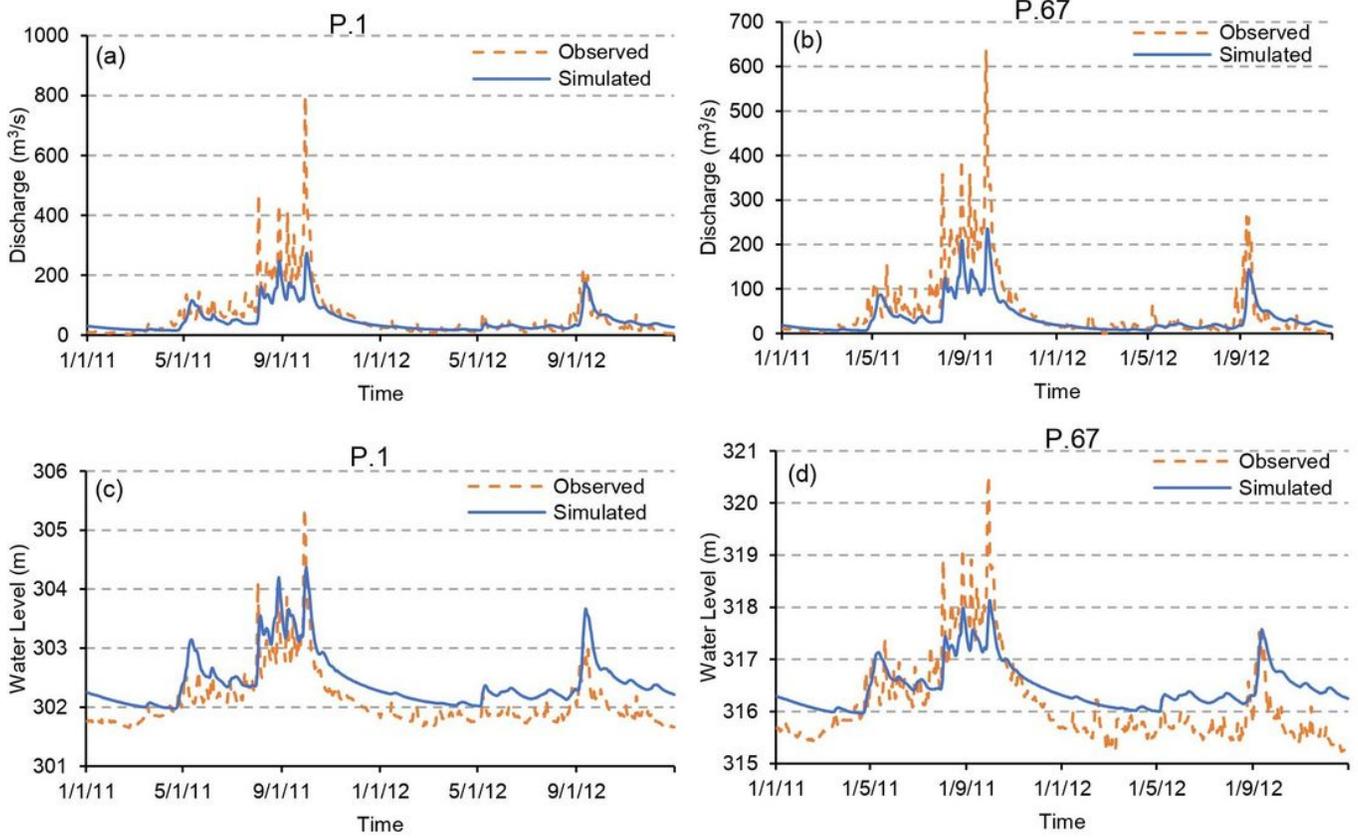


Figure 4

Figure 4

Validation results of discharge at (a) P.1 and (b) P.67 and water level at (c) P.1 and (d) P.67

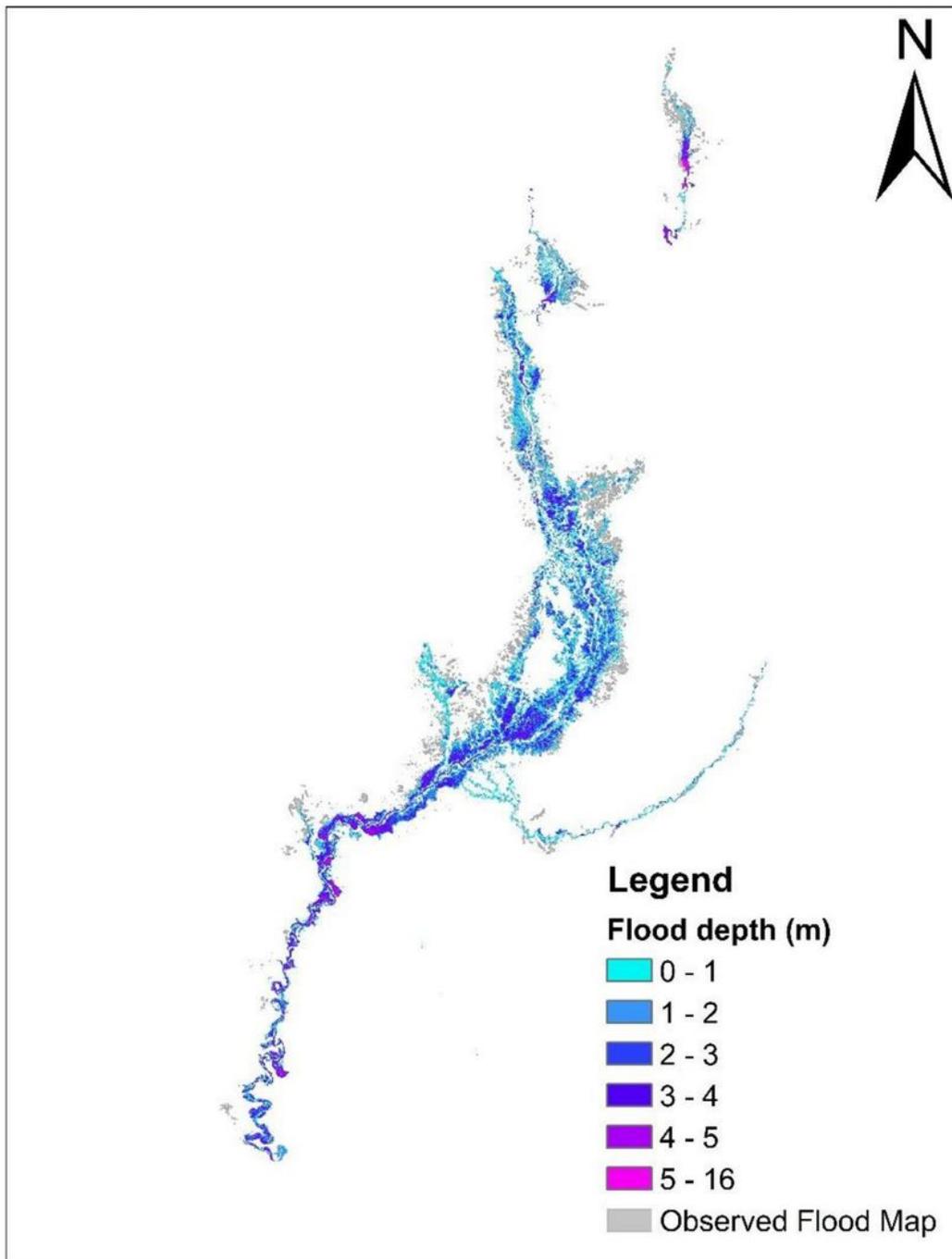


Figure 5

Figure 5

Calibration results of 2D-hydrodynamic model Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

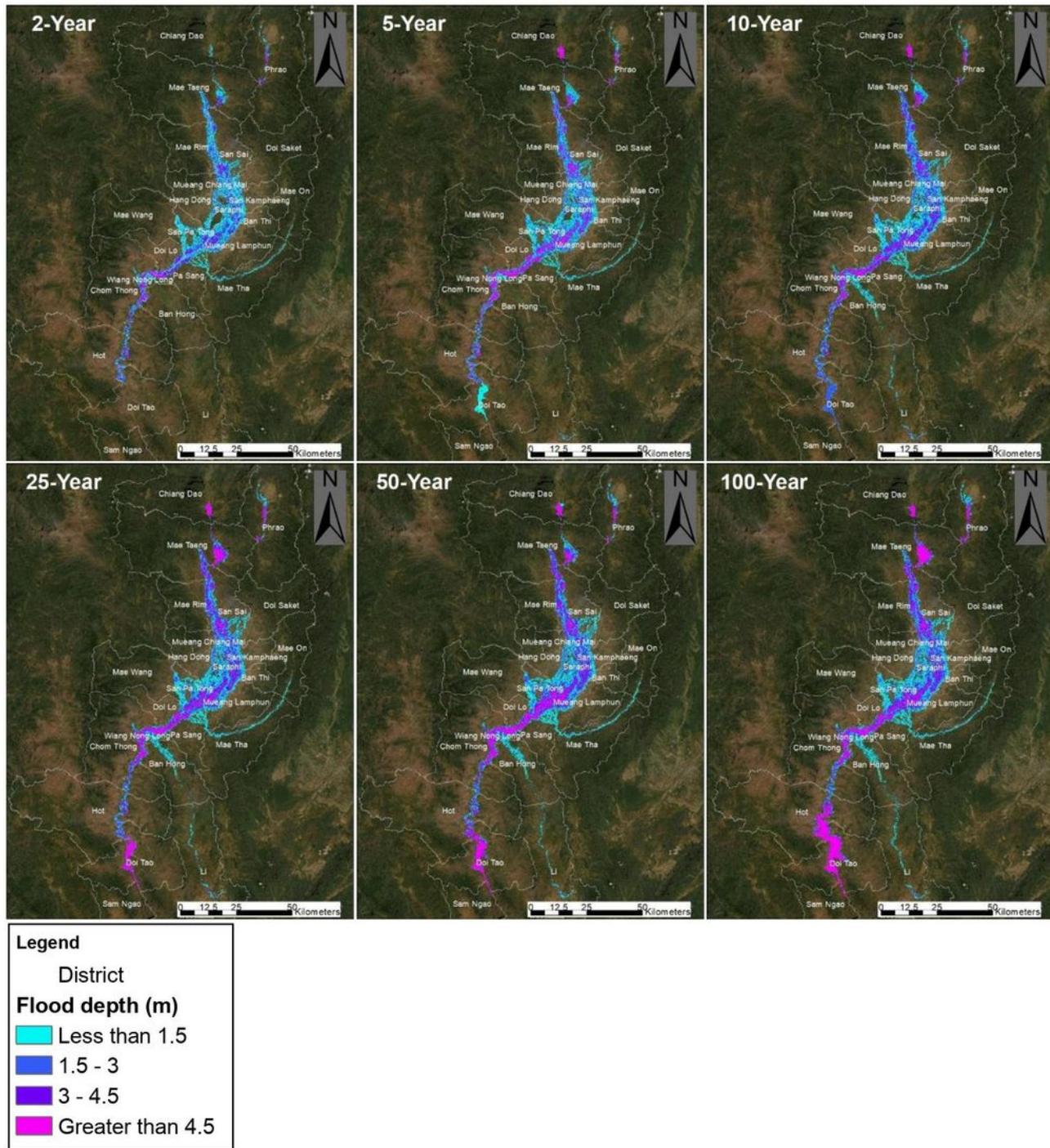


Figure 6

Figure 6

Flood inundation depth for different return periods Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

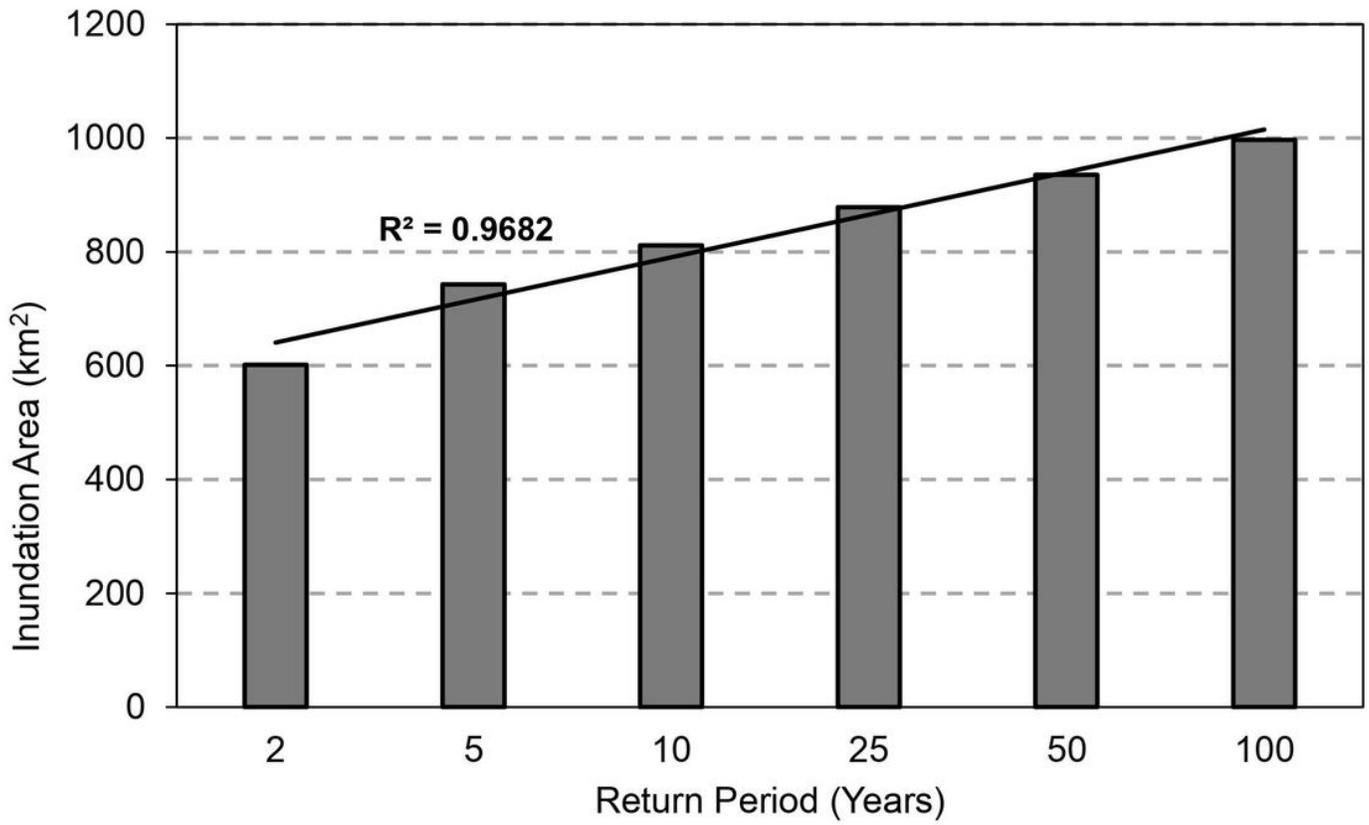


Figure 7

Figure 7

Comparison of inundation area against different return periods

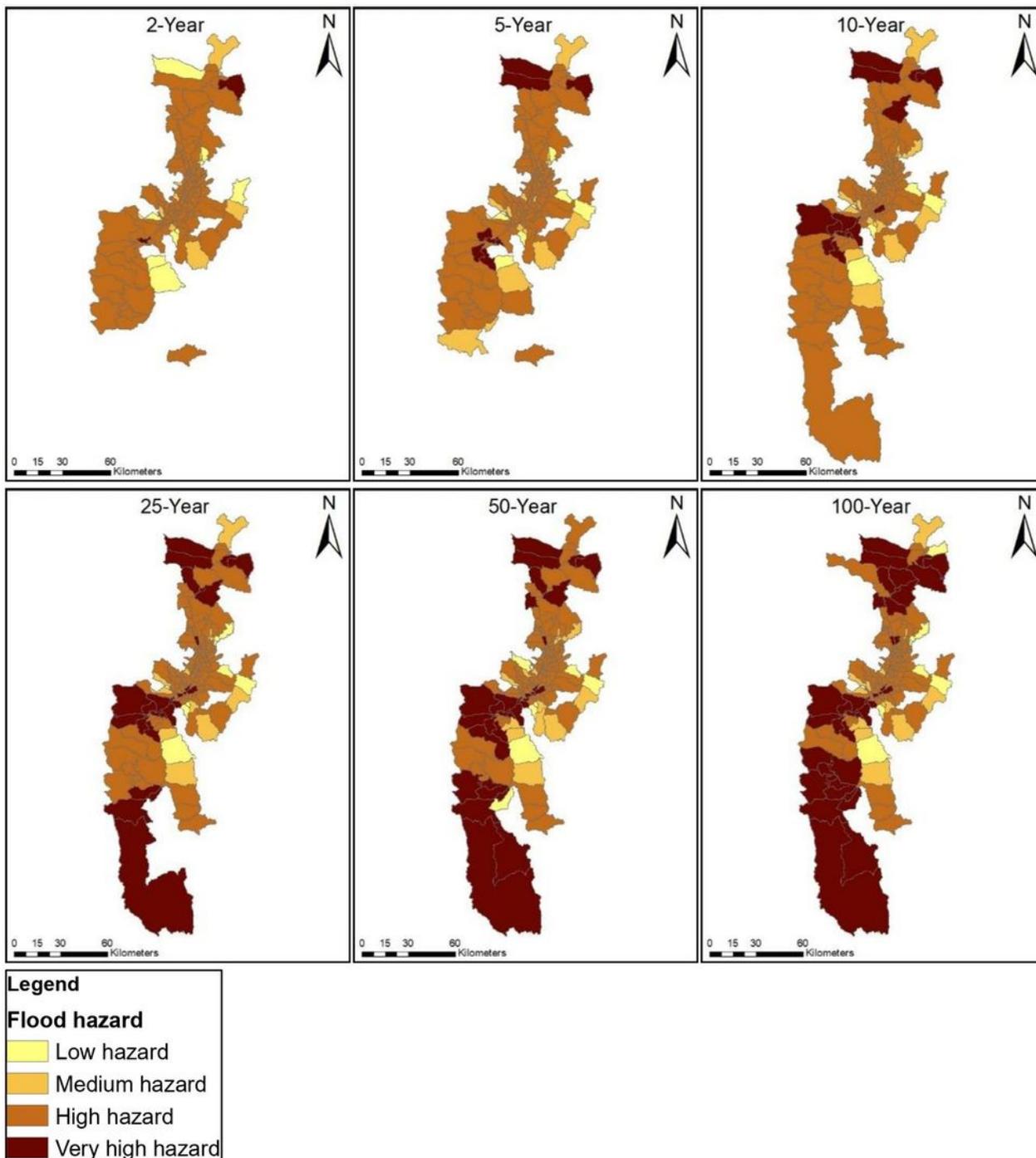


Figure 8

Figure 8

Flood hazard assessment of sub-districts for different return periods Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

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