



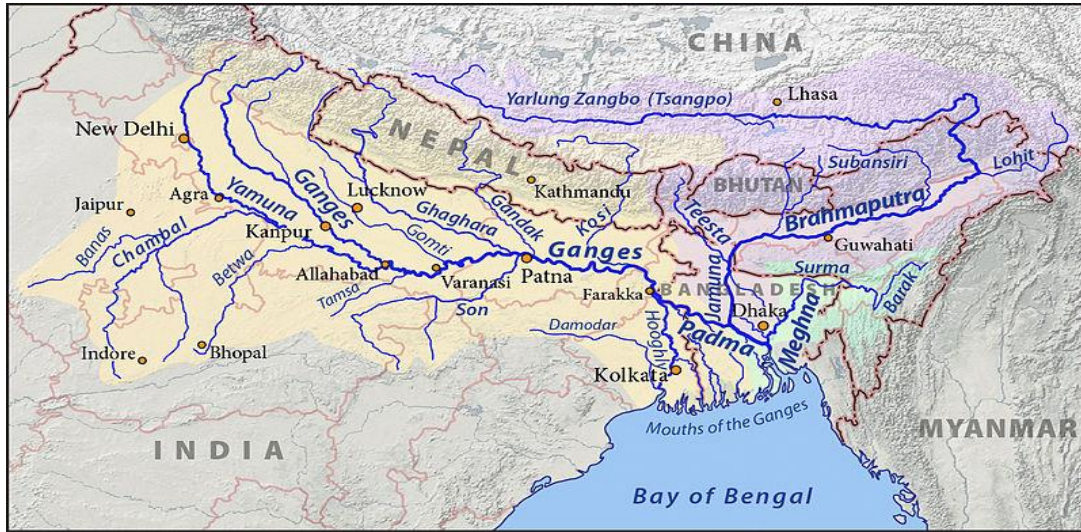
## 37 INTRODUCTION

38 The delta regions of Bangladesh and India are particularly vulnerable to flooding, with large tracts of land  
39 at low elevation making the deltas vulnerable to sea-level rise. Moreover, there is the potential for  
40 extensive flooding due to enhanced cyclone activity and increased river flows or extended droughts with  
41 changes in monsoon rainfall. Deltas have some of the highest population densities in the world with often  
42 poor and vulnerable residents (Nicholls et al., 2015, 2016, Hill et al., 2018, this volume). The adaptive  
43 strategies available to delta residents (e.g. disaster risk reduction, land use management or polders)  
44 may not be adequate to cope with pervasive, systematic, or surprise changes associated with climate  
45 change. Hence large movements of deltaic people are often projected under climate change (Nicholls et  
46 al., 2018, this volume). In addition, socio-economic change such as population increases are placing  
47 increasing pressure on resources making issues of water scarcity and food production crucial  
48 components of government planning and the focus of international interest. Climate change combined  
49 with socio-economic considerations are key strategic concerns of the Intergovernmental Panel on  
50 Climate Change (IPCC) (Fifth Assessment Report IPCC, 2014). The IPCC report highlights the likely  
51 impacts of climate change and proposes a strategy for assessing future Shared Socio-economic  
52 Pathways (SSPs) and how these might interact with climate change to generate a combined effect on  
53 catchments, people and livelihoods. This strategy has already been evaluated in a global rivers study  
54 with respect to flows (Arnell et al., 2013), and has been considered in relation to both flow and water  
55 quality (Whitehead et al., 2015 a, b, Jin et al., 2015). Other studies such as by Shi et al. (2011) have  
56 examined the impacts of climate change on agricultural aspects and how they affect water quality.

57 The DECCMA project is concerned with the impacts of climate change and other environmental drivers  
58 across contrasting deltas in Africa and Asia. Processes of migration are analysed using survey,  
59 participatory research and economic methods. Potential migration of people is contrasted with other  
60 adaptation approaches using a stakeholder-driven and co-produced integrated assessment approach.  
61 The project study sites are the Ganga-Brahmaputra-Meghna Delta (Bangladesh and India), the  
62 Mahanadi Delta (India) and the Volta Delta (Ghana). The Ganga-Brahmaputra-Meghna (GBM) River  
63 System and the River Hooghly constitute one of the largest river basins in the world serving a catchment  
64 population of over 780 million, and is of vital concern to India and Bangladesh as it provides fresh water  
65 for people, agriculture, industry, conservation and for the Delta System downstream. In the DECCMA  
66 programme a set of physical, geographical and chemical models have been used to simulate the  
67 catchments, the river systems, the delta estuary system and the coastal ecosystems in order to gain an  
68 understanding of the complex interactions and to project future change (Nicholls et al., 2018, this  
69 volume). Given the complex flow dynamics, diversified land uses, highly variable rainfall and temperature  
70 patterns, modelling the Indian and Bangladesh River Systems is a complex task. However, there have  
71 been several modelling studies of these rivers, with a greater focus on the Ganga River System. Many of  
72 these have been funded by Government departments or international organisations, such as the World  
73 Bank, with summary papers that capture the major findings and large scale macroeconomic aspects  
74 (Sadoff et al., 2013). Also, there have been several water quality modelling studies of other Asian River  
75 Systems such as nitrogen dynamics in small Himalayan catchments (Collins et al., 1999), pollution in the  
76 Ganga and Ramganga River Systems (Whitehead et al., 2015., Jin et al., 2015, Pathak et al., 2018, this  
77 volume) and sediment fluxes and morphology of rivers (Sinha et al., 2005, Roy and Sinha, 2014). Most  
78 previous climate modelling studies over the region have used data either directly from Global Climate  
79 Models (GCMs) or data downscaled from them from a finer resolution Regional Climate Model (RCM) of  
80 approximately 50 km. In this study, we have used data downscaled to 25 km from three different GCMs.

81 The main advantage of the finer grid RCM is that it is better able to represent local-scale climate  
82 processes than coarser resolution models. The downscaled climate data from the RCM is used to drive  
83 catchment models to assess impacts on flows and water quality. Then socio-economic changes are  
84 considered and 3 projected future strategies for development are considered. The impact on flow and  
85 water quality can be modelled using the INCA models, and a combined assessment of both climate  
86 change and socio-economic change evaluated.

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Figure 1 Map of the GBM and Hooghly Catchments Draining into the Bay of Bengal

91 **THE GANGA, BRAHMAPUTRA, MEGHNA (GBM), HOOGHLY AND MAHANADI RIVER**  
92 **SYSTEMS**

93 The GBM and associated Hooghly River System extends between the latitude of 22° 10' N  
94 to 31° 30'N and longitude of 78°0'E to 92° 0' E in the countries of India, Nepal, China,  
95 Bhutan, and Bangladesh (Figure 1), with a total catchment area of 1,612,000 km<sup>2</sup>. The  
96 GBMH River System is considered to be one large trans-boundary river basin, even though  
97 the rivers of this system have distinct characteristics and flow through very different  
98 geographical regions for most of their lengths. The Ganga River originates from the Gangotri  
99 glacier in the Himalayas at an elevation of nearly 7010 m and traverses a length of about  
100 2550 km (measured along the Bhagirathi and the Hooghly) before it flows southeast into the  
101 Bay of Bengal (see Figure 1). Along its way, the Ganga is joined by a number of tributaries  
102 to form the large fertile alluvial plain in North India (Figure 2). At Farakka Barrage, a major  
103 diversion delivers water from the Ganga into the Hooghly River, which then flows south into  
104 the Bay of Bengal on the Indian side. Approximately 50% of flows are diverted except during  
105 high flows (> 70,000 m<sup>3</sup>/s), with the exact diversions varying depending on inflows and  
106 season. The Farakka treaty signed between India and Bangladesh in 1996 was a significant  
107 agreement between the two countries and provides an agreed mechanism for sharing the  
108 available water. After the Farakka Barrage, the remaining flow of the Ganga plus the  
109 Brahmaputra and Meghna Rivers join and flow into the Bay of Bengal on the Bangladesh  
110 side of the delta, whilst the Hooghly flows into the Bay of Bengal on the Indian side of the  
111 Delta.

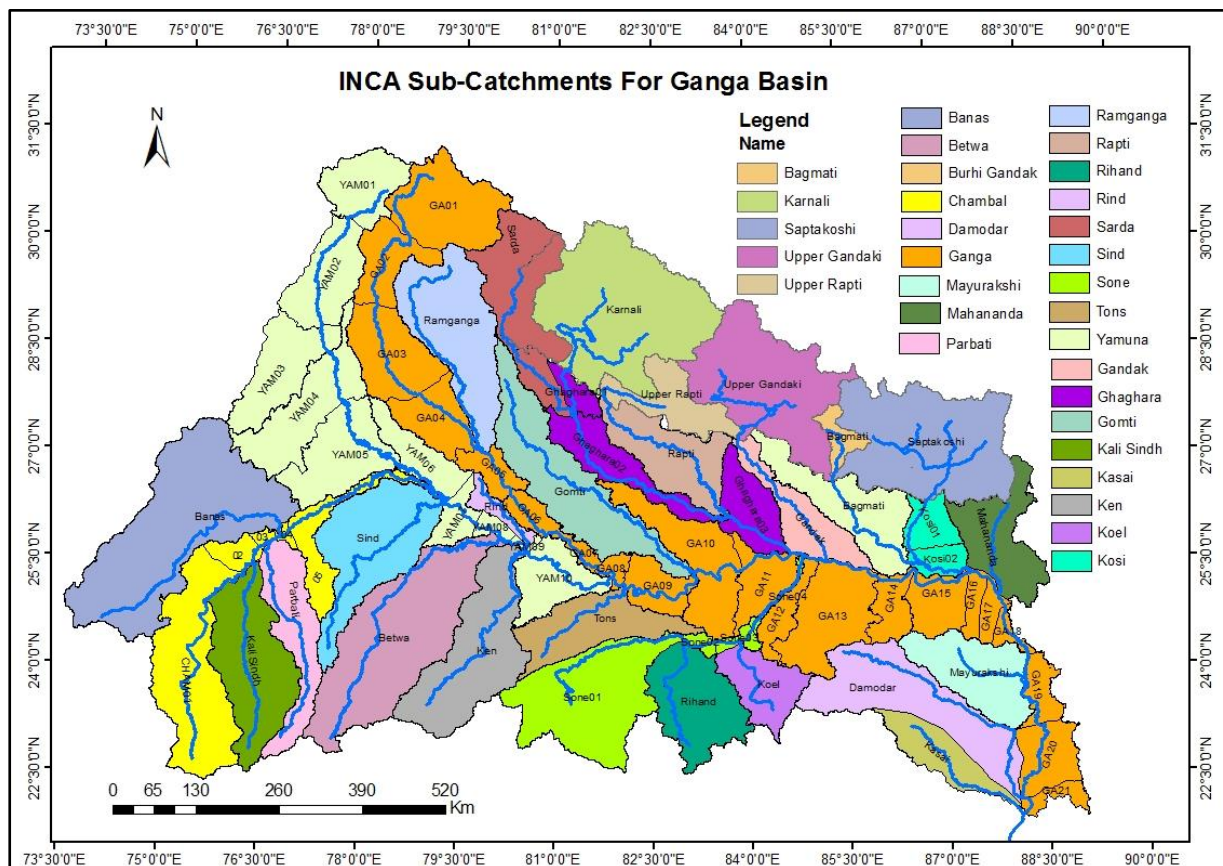
112 The Brahmaputra River originates on the northern slope of the Himalayas in China (Figure  
113 3), where it is called Yalung Zangbo. It flows eastwards for about 1130 km, then turns  
114 southwards and enters Arunachal Pradesh (India) at its northern-most point and flows for  
115 about 480 km. Then it turns westwards and flows through Arunachal Pradesh, Assam and  
116 Meghalaya for another 650 km and then enters Bangladesh, where it is also called Jamuna,  
117 before merging with the Ganga and Meghna rivers. The tributaries of the Meghna River  
118 originate in the mountains of eastern India and flow southwest to join the Ganga and  
119 Brahmaputra rivers before flowing into the Bay of Bengal (Figure 1).

120 The Mahanadi (Figure 4) is a major east-flowing peninsular river in eastern-central India.  
121 Extending between the longitudes of 80°28'E to 86°43' E and latitudes of 19°8'N to 23°32' N,  
122 the Mahanadi River System has a coverage area of 141,589 km<sup>2</sup>. The Mahanadi, which is  
123 851 km in length, starts from the Dhamtari District of Chhattisgarh and drains into a delta on  
124 the east coast before flowing into the Bay of Bengal. Although the major part of the basin  
125 covers the state of Orissa and Chhattisgarh, a smaller part of the catchment lies in the states  
126 of Jharkhand, Maharashtra and Madhya Pradesh. The Mahanadi is a great source of water  
127 for irrigation, industry, domestic utilities and for producing hydroelectricity.

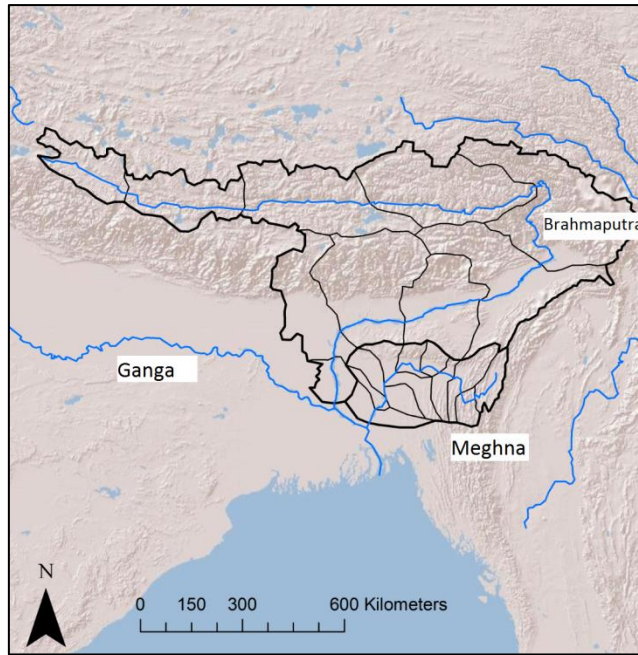
128 Bangladesh and Eastern part of West Bengal in India constitute the greatest deltaic plain in  
129 the world at the confluence of the Ganga, Brahmaputra and Meghna rivers and their  
130 tributaries. About 80 % of Bangladesh is made up of fertile alluvial lowland that becomes  
131 part of the Greater Bengal Plain. The country is flat with some hills in the northeast and  
132 southeast. About 7 % of the total area of Bangladesh is covered with rivers and inland water  
133 bodies and the surrounding areas are routinely flooded during the monsoon. Monsoon  
134 precipitation in the Ganga river basin lasts from July to October with only a small amount of  
135 rainfall occurring in December and January. The delta region experiences strong cyclonic

136 storms, both before the commencement of the monsoon season, from March to May, and at  
 137 the end of the monsoon from September to October. Some of these storms result in  
 138 significant life and the destruction of homes, crops and livestock, most recently in Cyclone  
 139 Sidr in 2007.

140 In Indian and Bangladesh rivers there tends to be three main sources of pollution which  
 141 include household and municipal untreated sewage disposal, effluents from commercial  
 142 activity or industrial sites and agricultural runoff. There is also atmospheric pollution which  
 143 can be significant with Nitrogen deposition being a large factor in the nitrogen budget, and  
 144 this needs to be incorporated into the modelling study.

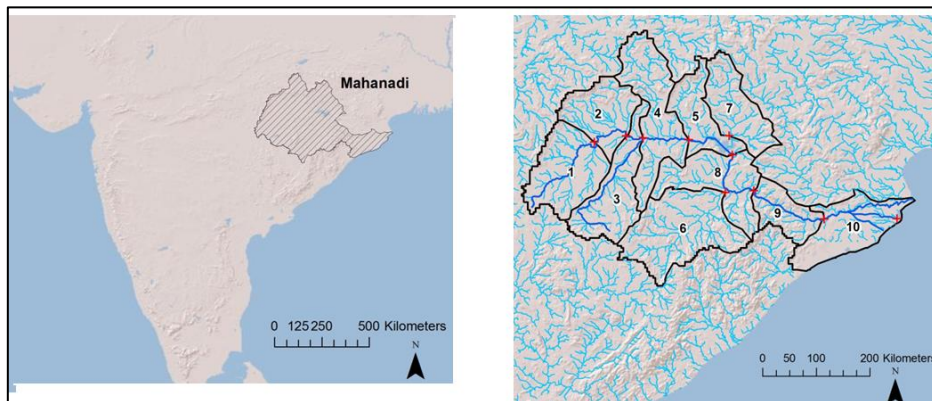


145  
 146 Figure 2 Map showing the multi-branch Ganga and Hooghly River System (Lower Reaches  
 147 GA17-21) and sub-catchments



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149 Figure 3 Map showing the multi-branch Brahmaputra and Meghna River Systems with the  
 150 sub-catchment areas



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152 Figure 4 The Mahanadi River Basin with INCA reaches and associated sub-catchment.

153 **THE MODELLING METHODOLOGY**

154 Modelling complex river systems such as the Ganga, Brahmaputra, Meghna, Hooghly and  
 155 Mahanadi (GBMHM) requires a semi-distributed model that can account for the spatial  
 156 variability of land use and topography across the catchment. The Integrated catchment  
 157 model INCA is one such model that has been applied extensively to heterogeneous  
 158 catchments and has the advantage that it is dynamic, process-based and integrates  
 159 hydrology and water quality (Whitehead et al., 1998 a, b, 2015, Wade et al., 2002). The  
 160 INCA N and INCA P models have been developed over many years as part of UK Research  
 161 Council (NERC) and EU funded projects and simulate hydrological flow pathways in the  
 162 surface and groundwater systems and tracks fluxes of solutes/pollutants on a daily time step  
 163 in both the terrestrial and aquatic portions of catchments. The model allows the user to  
 164 specify the spatial nature of a river basin or catchment, to alter reach lengths, rate  
 165 coefficients, land use, velocity-flow relationships and to vary input pollutant deposition loads  
 166 from point sources, diffuse land sources and diffuse atmospheric sources. INCA originally

167 allowed simulation of a single stem of a river in a semi-distributed manner, with tributaries  
 168 treated as aggregated inputs. The revised version now simulates nutrient dynamics in  
 169 dendritic stream networks as in the case of the GBM system with many tributaries. The  
 170 model is based on a series of interconnected differential equations that are solved using  
 171 numerical integration method based on the fourth-order Runge-Kutta technique. The  
 172 advantage of this technique is that it allows all equations to be solved simultaneously. The  
 173 INCA N and P model is described in detail (Whitehead et al., 1998, Wade et al., 2002, a, b)  
 174 and a detailed application to the Ganga River is given by Whitehead et al. (2015) and by Jin  
 175 et al. (2015).

176 **INCA N and INCA P SET UP FOR THE GBMHM RIVERS**

177 The INCA N and P models have been set up for the Ganga and Hooghly as a multi-reach  
 178 model with all the tributaries and sub-catchments, as shown in Figure 2. Reach boundaries  
 179 have been selected based on a number of factors such as a confluence point with a  
 180 tributary, a sampling or monitoring point or an effluent input or an abstraction point  
 181 associated with a major irrigation scheme or a large city. Digital terrain maps (DTMs) have  
 182 been used to establish the sub-catchment boundaries. For the Brahmaputra, the Meghna, a  
 183 similar multi-reach model set up has been established, as illustrated in Figure 3, which  
 184 shows all the reach boundaries and sub catchments. The land use data have been derived  
 185 using a 1 km grid resolution DTM with land cover data generated from the MODIS satellite  
 186 and direct discharges of effluents are incorporated into the INCA model set up. For the  
 187 Mahanadi the reach structure is shown in Figure 4. Further details of the Mahanadi setup  
 188 and flow and water quality simulations are given by Jin et al. (2018, this volume). Input data  
 189 for the rivers modelled has been obtained as shown in Table 1.

190 **Table 1 Data Sources for the River Modelling Study.**

| <b>Data Required</b>                          | <b>Data Source</b>  |
|---|---|
| Digital Terrain Model (DTM) of the Study Area | SRTM 90 x 90 m resolution raster data.  |
| Landuse and Land cover for Ganga basin        | National Remote Sensing Centre (NRSC) 56 x 56 Resolution Grid raster data.  |
| Sewage Treatment Plant                        | Design capacity, nitrate and ammonium concentration of outlet of STPs from the reports of Central Pollution Control Board.  |
| Crop growth data                              | FAO and Ministry of Agriculture Reports- Kharif crops - April-September, Rabi Crops- October to March, Double/Triple Crops & Plantation throughout the year.  |
| Fertilizer input data                         | Fertilizer used for the different crops available from FAO and Department of Fertilizers, Ministry of Chemicals and Fertilizers, Govt. of India.  |
| Discharge                                     | Observed discharge value (1979-2000) at the Hardinge Bridge in Bangladesh is available from CEGIS in Bangladesh, Observed Data of Mean Annual Discharge (1968-2000) is available at 5 Ganga river stations.   |
| Water quality Data                            | Data available from the Indian Central Pollution Control Board. Annual maximum, minimum and mean NO <sub>3</sub> -N (2003-2011) at various monitoring stations along Ganga river. Annual maximum, minimum and mean NH <sub>3</sub> -N (2003-2007) at various monitoring stations along Ganga River. Monthly NO <sub>3</sub> -N and NH <sub>3</sub> -N data (2010-2013) for few stations in West Bengal. |
| Meteorological                                | Daily Soil Moisture Deficit (SMD, mm), Hydrologically Effective Rainfall (HER, mm), 1.5m Mean Air Temperature (°C) and  |

|   |
|---|
| Actual Precipitation (mm) data were obtained or derived from three RCM simulations of the period 1971-2099. |
|---|

191

## 192 CLIMATE DATA

193 To give information about future hydrological conditions, INCA N requires as input a daily  
194 time series of catchment-average data describing relevant aspects of the future climate,  
195 namely precipitation, HER, temperature and soil moisture deficit (SMD). The model uses  
196 these data in smd

197 gical routines that calculate the sub-catchment river flows. The IPCC Fifth Assessment  
198 Report used the latest generation of GCMs to provide future projections of precipitation and  
199 temperature for all regions of the world, including over the GBMHM catchments (IPCC,  
200 2014). However, GCMs typically have coarse spatial resolutions with horizontal grid boxes of  
201 a few hundred kilometres in size, and cannot provide the high-resolution climate information  
202 that is often required for climate impact and adaptation studies. The use of an RCM, which  
203 dynamically downscales the GCM simulations through being driven using boundary  
204 conditions from GCMs, can provide higher resolution grids (typically 50km or finer) and is  
205 better able to represent features such as local topography and coast lines and their effects  
206 on the regional climate. There have been relatively few climate impact studies focused upon  
207 the Ganga River linked to the Bangladesh region that have used RCM output. Whitehead et  
208 al. (2015) used the 25 km resolution data over south Asia for the period 1971-2099  
209 downscaled by the Met Office using the PRECIS RCM system. The RCM is based on the  
210 atmospheric component of the HadCM3 GCM (Gordon et al., 2000) with substantial  
211 modifications to the model physics (see Jones et al. 2004 for details). The RCM was  
212 validated by comparing model output temperature and precipitation during the summer  
213 monsoon season with observational datasets, as described by Caesar et al. (2015).

214 For the DECCMA project, PRECIS has again been used to downscale GCM simulations to a  
215 resolution of 25km. In this study, three different GCMs have been downscaled to span the  
216 uncertainty in GCM-simulated future climate changes for the region, namely CNRM-CM5,  
217 GFDL-CM3 and HadGEM2-ES (Janes et al., 2018, this volume). Janes et al. (2018, this  
218 volume) have validated the three RCM simulations against precipitation and temperature  
219 observations covering northern India and Bangladesh and have found that all three RCMs  
220 reproduce the timing of the wet/dry and warm/cool seasons in the region, except for a delay  
221 in the wet season in the simulation forced with GFDL-CM3. Differences between the  
222 simulation outputs and observations differ in their detail, but all three simulations are slightly  
223 too dry during the monsoon season in many areas and slightly too cold throughout the year.  
224 In the Himalayas, in common with other RCM simulations performed for this region, all three  
225 simulations are wetter and colder than the observed climate, though deficiencies in  
226 observational datasets may contribute to these apparent biases.

227 The GCM simulations downscaled are of Representative Concentration Pathway 8.5 (RCP  
228 8.5), which has been selected as the main focus of the DECCMA project in order to consider  
229 the strongest climate change signal (Kebede et al. 2018, this volume). RCP 8.5 is consistent  
230 with greenhouse gas emissions continuing to rise throughout the 21st century and  
231 represents a relatively challenging situation for climate change adaptation, but one that does  
232 not appear unrealistic given recent changes in the Paris Climate Accord.



## 233 **SOCIOECONOMICS**

234 In addition to the climate impacts we need to consider the effects of changing socio-  
235 economics. Population change, industrial development, agriculture and land use change will  
236 all affect flows and water quality in River Systems and these changes will eventually also  
237 impact coastal systems. In terms of the socio-economic scenarios, three narratives have  
238 been defined in DECCMA based on the IPCC Shared Socio-economic Pathways or SSPs  
239 (IPCC, 2014). The three SSP scenarios have been selected based on medium economic  
240 growth (SSP2), medium plus with some higher economic growth (SSP5), and medium  
241 growth minus with a lower economic growth (~SSP3), all up to the 2050s. Beyond 2050,  
242 SSP5 is considered as the most likely scenario consistent with RCP 8.5. In the Indian and  
243 Bangladesh catchments there are many factors that affect the socio-economic conditions  
244 and potential futures from a flow and a water quantity perspective. These include population  
245 change and public water use, effluent discharge, water demand for irrigation and public  
246 supply, land use change, atmospheric deposition driven by industrial development or GDP  
247 and water transfer plans. Each of these aspects of socio-economic change need to be  
248 considered for the GBMHM catchments and a comprehensive description of the socio-  
249 economic assumptions are given elsewhere (Kebede et al., 2018, this volume). The  
250 following sub-sections present brief description and summary details of the catchment-  
251 specific scenarios and data considered.

### 252 **Population**

253 Population forecasts for Indian States vary widely depending on assumptions about fertility  
254 rate and economic wellbeing. UNDP population projections for 2041-2060 and 2080-2099  
255 and other socio-economic factors indicate a wide range of population growth (Kebede et al.,  
256 2018) and these are indicated in Table 2, showing a large increase under the low economic  
257 growth scenario and a much reduced rate under better economic conditions. Population  
258 increase also drives domestic effluent discharge, although the percentage of people moving  
259 to urban areas, where there are constructed sewerage systems, is an important factor. In  
260 general it is thought that the trends towards urban living will continue, with over 50% of the  
261 population living in urban areas in the future. The upgrading of Sewage Treatment Works  
262 (STWs) is also an important factor affecting water quality, with tertiary treatment being  
263 progressively installed in modern STWs, thereby reducing the phosphorus loads into rivers  
264 from these works by 86%. The Ganga Cleanup and Management Plan aims to considerably  
265 upgrade the sewerage and treatment processes. Assuming the secondary treatment  
266 processes are introduced, average ammonia discharge concentrations should fall from 19  
267 mg/l to 5 mg/l. Nitrate is likely to stay much the same unless N tertiary treatment is  
268 implemented, which is very expensive. STWs can also be designed to remove phosphorus  
269 and a reduction from 5 mg/l to 1 mg/l is highly likely as part of the cleanup process.

### 270 **Water demand for irrigation and public supply**

271 The demand for public water supply will increase with population growth, although much of  
272 the supply in rural areas is from groundwater. Changes in irrigation water demand reflect  
273 changes in agriculture and land use. However, agricultural changes in India are difficult to  
274 predict as any changes will depend on factors such as world food prices, which are driven by  
275 increasing global population, potential food scarcity and how farmers react to changing crop  
276 prices. Other key influential factors include technological developments, such as the

277 introduction of new crop varieties adapted to changing local environmental conditions. The  
 278 Food and Agriculture Organization of the United Nations (FAO) estimates a 22% rise in  
 279 Kcal/person/day in food production in India by 2050 (FAO, 2013) with much of this from  
 280 increased production of dairy and meat, as well as additional crops producing vegetable oils  
 281 and sugar. Agricultural expansion and intensification will be required to feed a growing  
 282 population. It is assumed in the medium scenario that new improved crop yields and more  
 283 efficient farming will occur, and irrigation abstraction from the rivers and groundwater will  
 284 increase. For the purposes of this study we have assumed that the abstraction from the  
 285 Ganga River will increase by 22 % on average but will vary slightly between scenarios.

286 **Atmospheric Nitrogen Deposition**

287 Atmospheric nitrogen pollution has become an increasing problem around the world, as  
 288 industrial development, power generation and ammonia release from intensive agriculture  
 289 has expanded. For example, across Europe, a set of Nitrogen Protocols have been  
 290 established by the UN/ECE Commission of Transboundary Pollution and these protocol  
 291 have been agreed and implemented by all EU countries. Deposition can be high with 15kg  
 292 N per hectare per year being deposited in certain parts of Europe such as the UK. The effect  
 293 of high atmospheric N is to alter the terrestrial ecology of plants and natural vegetation, and  
 294 provide a baseline source of N to groundwaters and streams, which can then affect aquatic  
 295 ecology. Research in the Himalayas, in which INCA N was applied to a range of basins,  
 296 suggests generally low concentrations of atmospheric N, but across India, levels are likely to  
 297 be much higher, with greater urban and industrial sources of atmospheric N (Whitehead et  
 298 al., 2015). In the future, increased industrial development and more intensive farming  
 299 methods will cause atmospheric N concentrations to increase. INCA N can incorporate these  
 300 effects as deposition loads to the sub-basins, and thus N levels have been altered to reflect  
 301 the different socio-economic scenarios into the future. It has been assumed that N  
 302 deposition rates are 8, 10, and 6 kg/ha/year for the three scenarios with medium growth,  
 303 plus and minus, respectively.

304 **Land use change**

305 Kathpalia and Kapoor (2010) and the FAO (World Agriculture Report 2013) reviewed  
 306 projected changes in agriculture in India. Their predicted changes in agriculture translate into  
 307 crop production and land use change across the basins. In general, they predict modest  
 308 changes in land use reflecting the fact that land in India is already used intensively for  
 309 growing a wide range of crops. They predict modest reductions in forest cover but an  
 310 increased area of double/triple crops to meet enhanced food demands, as indicated in Table  
 311 2.

312 Table 2 Summaries of three socio-economic scenarios for the catchments, under medium,  
 313 medium plus and medium minus development for the 2050s and the 2090s.

|                   | Medium |       | Medium + |       | Medium - |       |
|-------------------|--------|-------|----------|-------|----------|-------|
|                   | 2050s  | 2090s | 2050s    | 2090s | 2050s    | 2090s |
| Population change | 33%    | 29%   | 58%      | 108%  | 16%      | -8.4% |

|  |                             |                             |                                      |                                       |                             |                             |
|--|-----------------------------|-----------------------------|--------------------------------------|---------------------------------------|-----------------------------|-----------------------------|
| STW flow and design for water quality control (given urban % change) | flow increase by 33%        | flow increase by 29%        | flow increase by 58% and P at 1 mg/l | flow increase by 108% and P at 1 mg/l | flow increase by 16%        | flow decrease by 8.4%       |
| Water demand for irrigation and public supply                        | abstraction increase by 22% | abstraction increase by 22% | abstraction increase by 25%          | abstraction increase by 30%           | abstraction increase by 18% | abstraction increase by 18% |
| Atmospheric deposition of N  | 8 kg /ha/year               | 12 kg /ha/year              | 10 kg /ha/year                       | 15 kg /ha/year                        | 6 kg /ha/year               | 9 kg /ha/year               |
| Int. Agric. Land Use Change  | 5% increase in agriculture  | 7% increase in agriculture  | 7% Increase in agriculture           | 10% increase in agriculture           | 4% increase in agriculture  | 6% increase in agriculture  |

314

### 315 **MODELLING HYDROLOGY AND WATER QUALITY**

316 The INCA N and P models have already been set up for the Ganga River as part of a  
317 separate study (Whitehead et al., 2015, Jin et al., 2015) and used to model the hydrology,  
318 nitrate and ammonia, phosphate in all the tributaries and the main river systems. This study  
319 required setting up the INCA N and P model for the Ganga, Brahmaputra, Meghna, Hooghly  
320 and Mahanadi using a complex reach structure, as shown in Figures 2, 3 and 4. Details of  
321 the Mahanadi application are given by Jin et al. (2018, this volume). The daily precipitation  
322 and temperature data from the RCMs have been averaged across the study catchments and  
323 these data then used to calculate evapotranspiration rates, hydrologically effective rainfall  
324 (HER) and soil moisture deficit (SMD) using the PERSIST model (Futter et al., 2015). The  
325 PERSIST model is a daily hydrological model that can be calibrated against observed flows  
326 and is driven by daily precipitation and temperature. Key outputs of the model are the daily  
327 Hydrologically Effective Rainfall (HER) and the daily SMD and these data sets can then be  
328 used to drive the daily INCA model hydrology. This analysis process has been applied to all  
329 the GBMHM catchments and the daily HER, SMD and temperature data then used drive  
330 INCA N and INCA P. The simulations from the catchment models have then been provided  
331 for the downstream coastal modellers in order to assess impacts of climate change on the  
332 coastal systems. Details of the model calibration and validation for both flow and water  
333 quality are given by Whitehead et al. (2015) and Jin et al. (2015, 2018, this volume). The  
334 observed flow and quality data is sparse on the Indian and Bangladesh River systems,  
335 although there is a flow gauge on the Brahmaputra at Bahadurabad. Table 3 shows  
336 calibration and validation statistics plus N-S statistics (Nash and Sutcliffe, 1970) fits to the  
337 observed data for the Ganga River system at 4 locations

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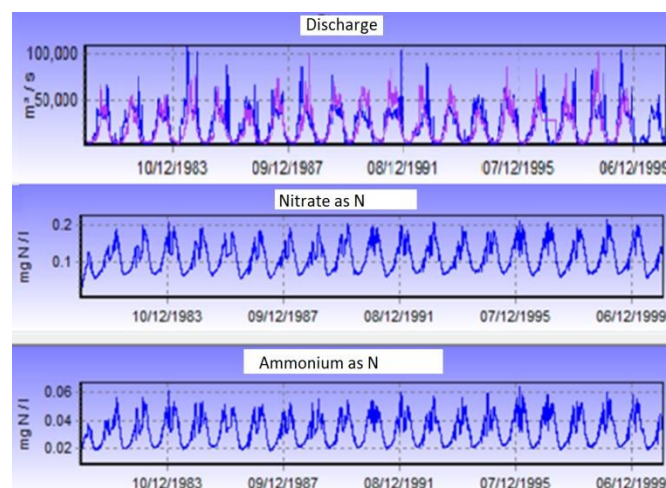
Table 3 Statistics of Model Fit for Flow on the Ganga at 4 locations

| Ganga River Locations | R <sup>2</sup> | N-S  | Flow calibration R <sup>2</sup> | Flow validation R <sup>2</sup> |
|-----------------------|----------------|------|---------------------------------|--------------------------------|
| Kachlabridge          | 0.6            | 0.49 | 0.6                             | 0.54                           |
| Ankinghat             | 0.73           | 0.66 | 0.56                            | 0.51                           |
| Kanpur                | 0.56           | 0.45 | 0.49                            | 0.49                           |
| Hardinge bridge       | 0.73           | 0.49 | 0.73                            | 0.7                            |

340 Figures 5 and 6 show typical results for flows and nitrate simulations and fits to the observed  
 341 data. The N load calculations in Figure 6 are based on simulated and observed Nitrate-N  
 342 concentrations and flows. In this study, three 20 year time slices have been evaluated using  
 343 1981-2000 as a baseline period, with 2041-2060 as a future mid-term time slice and the  
 344 period 2079-2098 as a far future time slice.

### 345 **Model uncertainty analysis and limitations**

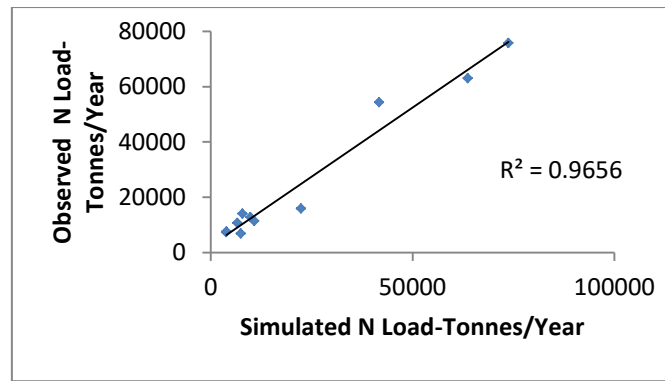
346 There are always issues concerned with model uncertainty and model limitations in any flow  
 347 and water quality modelling study. Models are a simplification of reality and allow an  
 348 approximation to any system to be evaluated (Whitehead et al., 2016). The fact that the  
 349 system is dynamic and that chemical processes are kinetic in nature and interact with the  
 350 hydrology makes modelling a very complex process. There have been extensive studies of  
 351 uncertainty using the INCA models with techniques such as generalized sensitivity analysis  
 352 (Spear and Hornberger, 1980, Hornberger and Spear, 1980) used extensively to test INCA  
 353 model uncertainty (Rankinen et al. 2006, Futter et al. 2007, Wade et al. 2002 a, b, c) to  
 354 evaluate both parametric uncertainty in the INCA Model and uncertainty in the data used to  
 355 drive the model. For example, Wade et al. (2002a) found that the INCA N model  
 356 performance was sensitive to the parameters defining nitrogen dynamics such as  
 357 denitrification processes as well as key hydrological factors such as base flow index. There  
 358 are limitations in any water quality modelling study because of all the complexities. In this  
 359 study a key limitation is the lack of frequent water quality data to allow a full calibration and  
 360 validation of the model and even the flow data is fairly limited. So there will be uncertainty in  
 361 any results from such a study.



362

363 Figure 5 Simulated (blue line) and Observed (purple line) daily flows at the Flow Gauge on  
 364 the Brahmaputra River System at Bahadurabad for 1981-2000 together with simulated  
 365 Nitrate and Ammonium

366



367

368

Figure 6 Simulated and Observed Loads in the Ganga River at Kanpur

369

### EFFECTS OF CLIMATE AND SOCIO-ECONOMIC CHANGES ON RIVER FLOW

370

With 5 rivers to compare and 3 Climate Scenarios plus 3 socio-economic scenarios there are many combinations of results to discuss for flow and water quality. Here we discuss a limited set but they capture the likely changes as suggested by the modelling analysis. Figure 7 shows the impacts of climate change on flow in the 5 River Systems using the climate data downscaled from HadGEM2-ES. The Brahmaputra simulated monthly mean flows in Figure 7 show relatively little change into the future for the 2050s and into the 2090s. However, the pattern of monthly flows is quite different for the Ganga, the Hooghly and the Mahanadi which all show limited changes by the 2050s, but significant increases in monsoon flows in the 2090s, reflecting higher precipitation in the monsoon period (Janes et al., 2018, this volume).

379

380

Table 3 shows changes in extreme flow for the downscaled GFDL-CM3 GCM. It compares the extremes of behaviour with the statistic Q95 representing the low flow or drought conditions and Q5 representing the high flow or flooding conditions. The simulated flow values are shown in the Table 3 together with the percentage change in Q95 and Q5 by the 2050s and the 2090s. The high flow conditions (Q5) show a modest increase in the 2050s in all rivers and a very large change in the 2090s, suggesting that flooding will significantly increase. With regard to low flows, there are differing patterns of behaviour, with the Brahmaputra, Meghna and the Mahanadi showing significant increases by the 2050s and all rivers showing increases by the 2090s. This initially suggests an easing of water security issues, but these results do not take into account the socio-economic changes where increased water use for irrigation and public use will lower baseflow conditions. The socio-economic effects are considered below in terms of flow and water quality.

391

392

An interesting result is the comparison of the 3 downscaled GCMs (CNRM-CM5, GFDL-CM3 and HadGEM2-ES). Each of these generates different time series of precipitation, temperature and evaporation (Janes et al., 2018, this volume), and these also vary from catchment to catchment. Figure 8 shows the variation between the regional climate model output in terms of flow in the 5 different rivers. Results for the downscaled CNRM-CM5 and HadGEM2-ES GCMs are remarkably similar in terms of flow changes by the 2050s. However, results for the downscaled GFDL-CM3 GCM show lower flows in the Ganga and Hooghly Rivers.

399

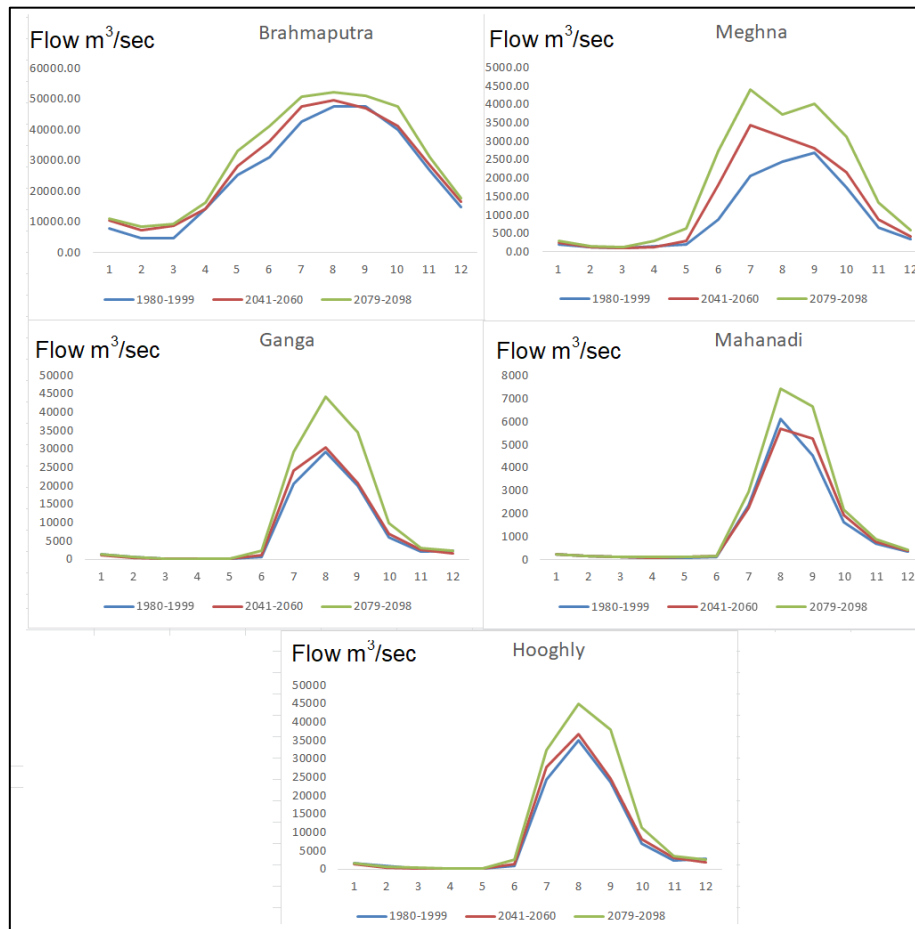
400

The effects of socio-economic changes on river flows is illustrated in Figure 9, which shows the 2050s total flows for the combined Ganga, Brahmaputra and Meghna in Bangladesh.

401

402 The effects of the socio-economic factors are fairly minimal by the 2050s and are  
 403 outweighed by the increased flows due to climate change. However, if the effects of major  
 404 water transfer schemes are considered, such as a 30% diversion of the Brahmaputra flows,  
 405 then then there is a significant reduction in flows in Bangladesh. This could be quite serious  
 406 in Bangladesh and especially in the low flow period when the water is used in Bangladesh  
 407 for irrigation and is a key freshwater driver for fisheries in the rivers and estuary systems.  
 408 Note in Figure 9 the 3 socioeconomic scenarios are indistinguishable, apart from the  
 409 Brahmaputra water diversion flow scenario.

410



411

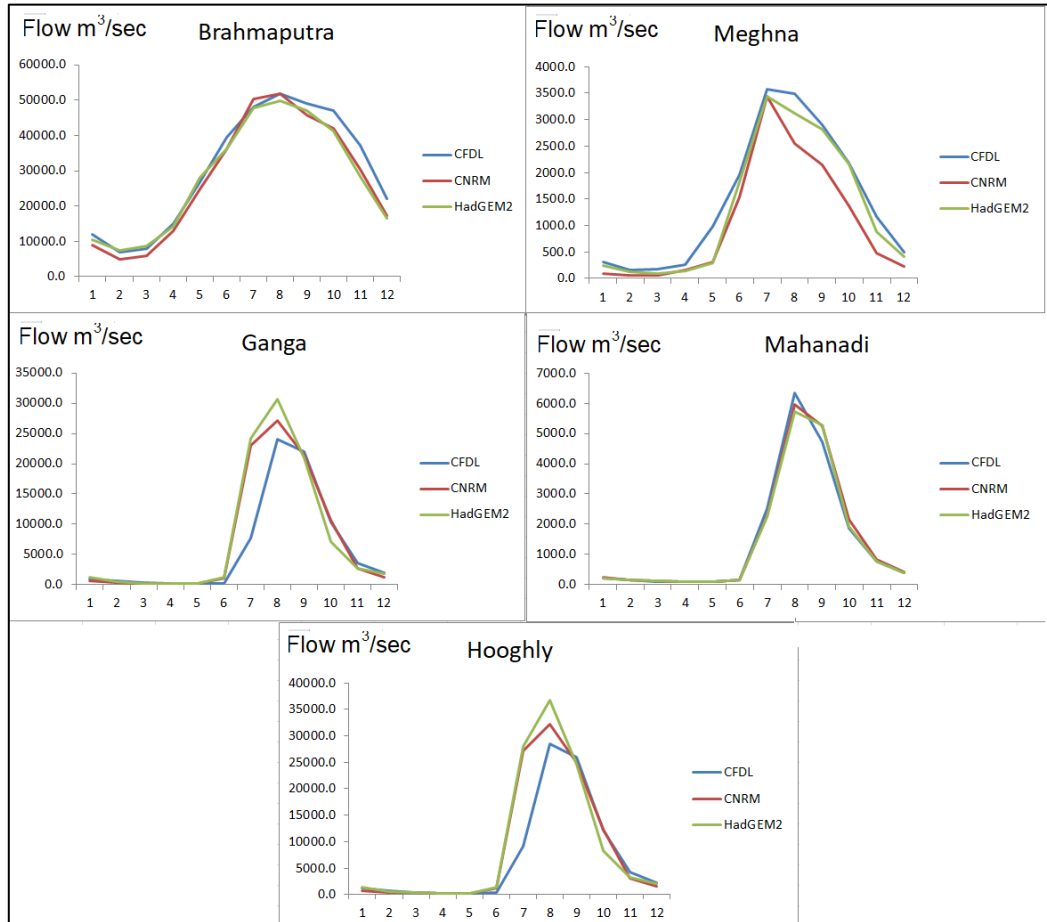
412 Figure 7 Impacts of Future Climate Change on Monthly Mean Flows in the last reaches of 5  
 413 Rivers for the downscaled HadGEM2-ES climate data.

414 Table 3 Simulated Current Flow Statistics for flood and low flow conditions and percentage  
 415 changes into the future for the 5 River Systems for climate data downscaled from the GFDL-  
 416 CM3 GCM.

|  | 1990s Flow          |     | 2050s Flow |     | 2090s Flow |     |
|--|---------------------|-----|------------|-----|------------|-----|
|  | m <sup>3</sup> /sec |     | % change   |     | % change   |     |
|  | Q5                  | Q95 | Q5         | Q95 | Q5         | Q95 |
|  |                     |     |            |     |            |     |

|             |         |        |      |      |      |       |
|-------------|---------|--------|------|------|------|-------|
| Brahmaputra | 50561.4 | 3529.6 | 6.8  | 55.6 | 13.7 | 55.6  |
| Meghna      | 3277.4  | 30.0   | 35.2 | 15.1 | 86.9 | 98.1  |
| Ganga       | 31116.3 | 127.2  | 5.9  | 3.5  | 56.7 | 37.0  |
| Hooghly     | 36600.2 | 148.4  | 9.6  | 3.5  | 29.5 | 37.0  |
| Mahanadi    | 6064.2  | 90.1   | 17.6 | -0.1 | 97.7 | 12.34 |

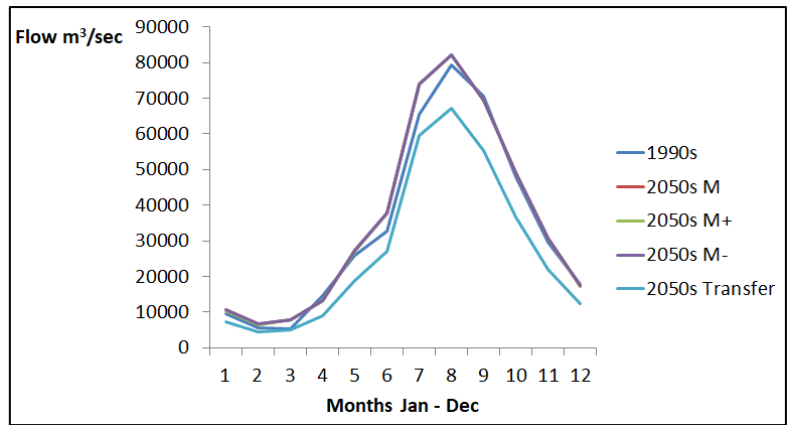
417



418

419 Figure 8 Impacts of Climate Change on 2050s Monthly Mean Flows in 5 Rivers for  
 420 climate data downscaled from the 3 GCMs

421



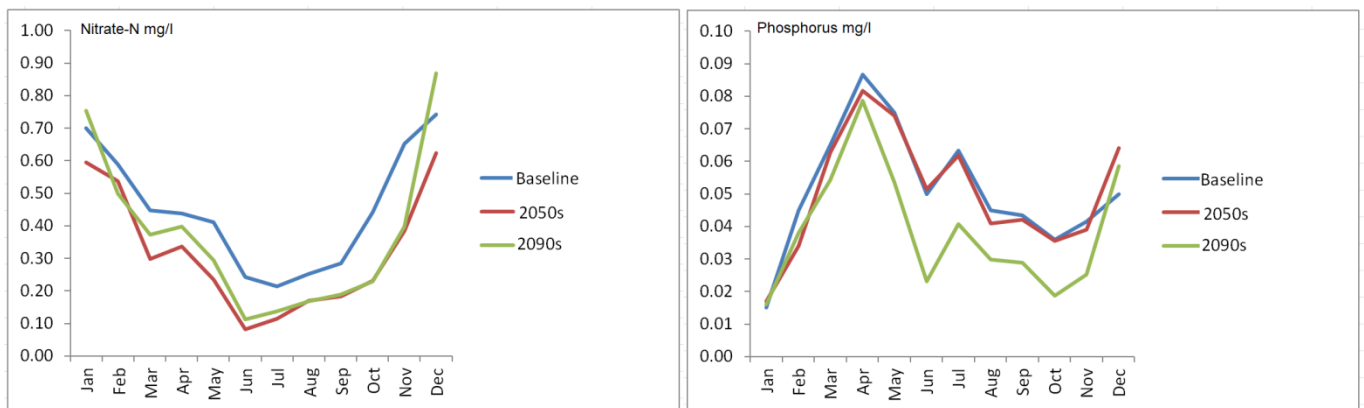
422

423 Figure 9 Impacts of Socio-Economic Change and also a Water Transfer Scheme in the  
424 2050s

425 **EFFECTS OF CLIMATE AND SOCIO-ECONOMIC CHANGES ON WATER QUALITY**

426 In terms of water quality, the climate and socio-economic changes will impact Nitrate and  
427 Phosphorus in different ways. For example, Figure 10 shows the impacts of climate change  
428 on Nitrate and Phosphorus in the 2050s and 2090s for the Ganga River using climate data  
429 downscaled from the climate model combination GFDL-CM3. The higher monsoon flows in  
430 the 2050s and 2090s dilute the sources of N and P in the catchment and hence generate  
431 lower N and P in the high flow periods. However in the low flow periods P increases whereas  
432 N decreases. This is due to the different process affecting N and P, where N undergoes  
433 extensive denitrification under low flows, due to the higher temperatures and the increased  
434 residence time of the river water. Whereas P concentrations rise in the low flow periods as  
435 there is less dilution of effluent discharges and agricultural runoff.

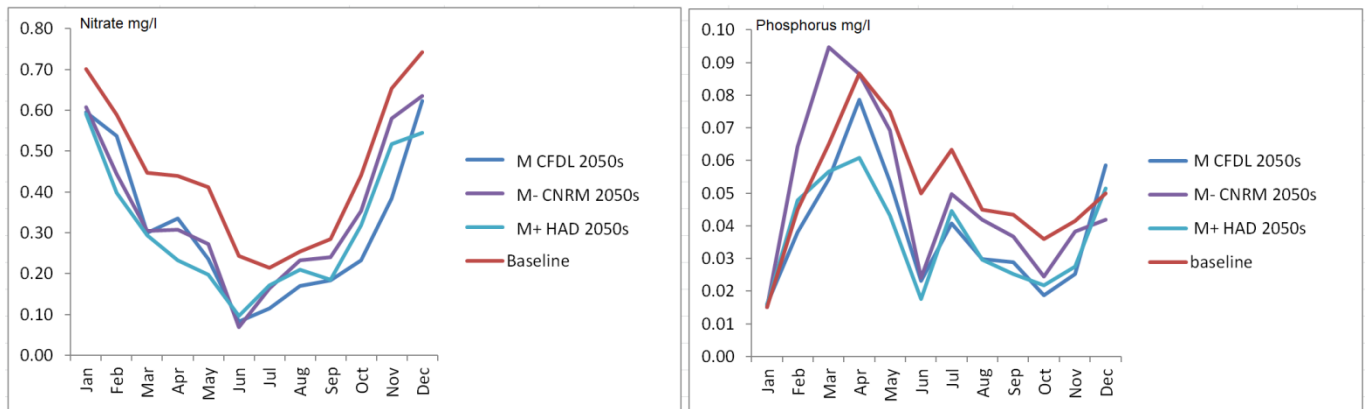
436 Figure 11 illustrates the impacts on river nutrients of socio-economic changes combined with  
437 climate changes. The results suggest that socio-economics will have a fairly limited impact  
438 on water quality except in the case of the medium minus economy as it affects P in the dry  
439 season. This is because there is the assumption that STWs will not be upgraded and with a  
440 decline in low flows plus there is less dilution of pollutants entering the River Systems. The  
441 resulting increasing P will enhance eutrophication and lead to enhanced algal blooms, some  
442 of which are toxic, such as cyanobacteria (Bussi et al., 2016)



443

444 Figure 10 Impacts of Climate Change on Nitrate and Phosphorus in the 2050s and 2090s for  
445 the Ganga using climate data downscaled from the GFDL-CM3 GCM





447

448 Figure 11 Impacts of Climate Change on Nitrate and Phosphorus concentrations in the  
 449 2050s for 3 climate scenarios (abbreviated to “GFDL”, “CNRM” and “HAD”) and 3 socio-  
 450 economic scenarios (abbreviated to “M”, “M-“ and “M+”)

## 451 DISCUSSION AND CONCLUSIONS

452 The GBMHM Rivers are of crucial importance providing water for public supply, hydropower,  
 453 irrigation water for agriculture, and are also of great cultural significance. In this study, INCA  
 454 N and P have been applied the five rivers to assess the likely future impacts of climate  
 455 change and socio-economic changes. It is recognised that there are a number of  
 456 uncertainties within the model, input data and parameters. The lack of adequate flow and  
 457 water quality data limits the ability to fully evaluate the model’s performance. However,  
 458 comparison with the data that is available demonstrates reasonable replication of the overall  
 459 magnitude and pattern of flows and water quality.

460 The model results suggest that there is a significant increase in flows projected under a  
 461 future climate change during the monsoon season. This is due to future increases in  
 462 monsoon rainfall in the climate data downscaled from all three of GCMs that we have  
 463 considered (Janes et al., 2018, this volume). Continuing emissions of greenhouse gases  
 464 over the 21<sup>st</sup> century are consistent with the RCP 8.5 scenario and result in increases in  
 465 monsoon rainfall over the century in most of the current generation of GCMs. Hence, this  
 466 result is likely to be robust to a different choice of GCMs for downscaling. The increased  
 467 flows in the monsoon suggest there will be increased flooding into the future, which could  
 468 have significant consequences for India and Bangladesh. Such changes in flow on a  
 469 seasonal basis will also affect nutrients with N and P being diluted under the higher flows but  
 470 phosphorus increasing in low flow conditions as dilution is reduced.

471 Changes to low flows are also likely to occur given projected increases in variability. Here,  
 472 the model results suggest that drought duration may become more frequent, whilst extreme  
 473 low flows may actually increase in extent. However, changes to low flows are likely to be  
 474 more sensitive to uncertainties in climate projections and assumptions regarding land-  
 475 surface runoff and river channel transport. In general, the socio-economic changes  
 476 considered had minimal impact on flows. However, the magnitude of these changes is also  
 477 uncertain and large scale water transfers will significantly alter flows (Whitehead et al., 2015)  
 478 The socio-economic scenarios mostly affect the nutrient balance with increasing

479 concentration of N and P under the medium minus scenario. However, these socio-  
480 economic changes are offset by the changes in climate.

481 The development of models for such large and complex river systems provide an important  
482 planning tool for assisting in exploring future scenarios, engaging stakeholders in dialogue  
483 on water resources management, and identifying gaps in knowledge and data. Considering  
484 both flows and water quality for a range of climate and socio-economic scenarios can assist  
485 in a more holistic management approach.

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#### 492 **REFERENCES**

493 Arnell N W and Lloyd-Hughes B, 2013, The global-scale impacts of climate change on water  
494 resources and flooding under new climate and SSPs, *Climatic Change* (2014) 122:127–140 ,  
495 DOI 10.1007/s10584-013-0948-4

496 Caesar, J., Janes, T., and Lindsay, A. 2015 Climate projections over Bangladesh and the  
497 upstream Ganges-Brahmaputra-Meghna system, [Environ Sci Process Impacts](#), 2015  
498 17(6):1047-56. doi: 10.1039/c4em00650j.

499 Collins, R., Whitehead, P.G. And Butterfield, D. (1999) Nitrogen Leaching from Catchments  
500 in the Middle Hills of Nepal; an application of the INCA model, *Science of the Total  
501 Environment*, 228, 259-274.

502 Futter, M.N., Butterfield, D., Cosby, B.J., Dillon, P.J., Wade, A.J., and Whitehead, P.G.,  
503 2007, Modeling the mechanisms that control in-stream dissolved organic carbon dynamics in  
504 upland and forested catchments: *Water Resources Research*, v. 43, p. W02424,  
505 doi:10.1029/2006WR004960.

506 Futter, M. N., Whitehead, P.G., S. Sarkar, H. Rodda and J. Crossman, 2015, Rainfall Runoff  
507 Modelling Of The Upper Ganga And Brahmaputra Basins Using Persist, *Environ. Sci.:  
508 Processes Impacts*, 2015,17, 1070-1081 DOI: 10.1039/C4EM00613E

509 Gordon, C., C. Cooper, C. A. Senior, H. Banks, J. M. Gregory, T. C. Johns, J. F. B. Mitchell  
510 and R. A. Wood, 2000: The simulation of SST, sea ice extents and ocean heat transports in  
511 a version of the Hadley Centre coupled model without flux adjustments. *Clim. Dyn.*,16:147-  
512 168

513 Hill, C., Nicholls, R J, Whitehead, P G, Dunn F, Haque, A., Appeaning-Addo, K, and Raju P  
514 V, (2018), Delineating Climate Change Impacts on Biophysical Conditions in Populous  
515 Deltas , Editorial, A special Issue of *Science of the Total Environment*, this volume

516 Hornberger, G. M. and Spear, R. C., 1980 Eutrophication in peel inlet: The problem-defining  
517 behaviour and a mathematical model for the phosphorus scenario, In *Water Research*,

518 Volume 14, Issue 1, 1980, Pages 29-42, ISSN 0043-1354, <https://doi.org/10.1016/0043->  
519 1354(80)90039-1.

520 IPCC, 2014: Summary for Policymakers, In: Climate Change 2014, Mitigation of Climate  
521 Change. Contribution of Working Group III to the Fifth Assessment Report of the  
522 Intergovernmental Panel on Climate Change [Edenhofer, O., R. Pichs-Madruga, Y. Sokona,  
523 E. Farahani, S. Kadner, K. Seyboth, A. Adler, I. Baum, S. Brunner, P. Eickemeier, B.  
524 Kriemann, J. Savolainen, S. Schlömer, C. von Stechow, T. Zwickel and J.C. Minx (eds.)].  
525 Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA.

526 Janes, T., McGrath, F., Macadam, I., and Jones, R.G., 2018, High-resolution climate  
527 projections for South Asia to inform climate impacts and adaptation studies in the Ganges-  
528 Brahmaputra-Meghna and Mahanadi deltas. *Science of the Total Environment*, This Volume

529 Jin, L., Whitehead, P G, Sarkar, S., Sinha, R., Futter, M N, Butterfield, D, Caesar J., and  
530 Crossman J. , 2015, Assessing The Impacts Of Climate Change And Socio-Economic  
531 Changes On Flow And Phosphorus Flux In The Ganga River System, *Environ. Sci.:*  
532 *Processes Impacts*, 17, 1098-1110 , DOI: 10.1039/C5EM00092K

533 Jin I., Whitehead P G, Rodda H, Macadam I, Sarkar S. 2018 RCP Climate Change Impacts  
534 of Flows and Water Quality in the Mahanadi River System, India, *Science of the Total*  
535 *Environment*, This Volume

536 Jones, R.G., Noguera, M., Hassell, D.C., Hudson, D., Wilson, S.S., Jenkins, G.J. and Mitchell,  
537 J.F.B. (2004) Generating high resolution climate change scenarios using PRECIS, Met  
538 Office Hadley Centre, Exeter, UK, 40pp April 2004

539 Kathpalia, G.N., and Kapoor, R., 2010, Management of Land and other Resources for  
540 Inclusive Growth: India 2050: Alternative Futures, Dehli, p. 32.

541 Kebede, A.S., Nicholls, R.J., Allan, A., Arto, I., Cazcarro, I., Fernandes, J.A., Hill, C.T.,  
542 Hutton, C.W., Kay, S., Lawn, J., Lazar, A.N., Macadam, I., Palmer, M., Suckall, N.,  
543 Tompkins, E.L., Vincent, K., and Whitehead, P.G., 2018, Integrated Scenario Framework to  
544 Explore Migration and Adaptation in Deltas: The DECCMA Approach: *Science of the Total*  
545 *Environment*, This Volume.

546 Nash, J. E. and Sutcliffe, J. V. 1970. River flow forecasting through conceptual models part I-  
547 -A discussion of principles. *Journal of Hydrology* **10** (3): 282–290. doi:10.1016/0022-  
548 1694(70)90255-6

549 Nicholls, R.J., C.W. Hutton, A.N. Lázár, A. Allan, W.N. Adger, H. Adams, J. Wolf, M.  
550 Rahman, and M. Salehin. 2016. Integrated assessment of social and environmental  
551 sustainability dynamics in the Ganga-Brahmaputra-Meghna delta, Bangladesh. *Estuarine,*  
552 *Coastal and Shelf Science* 183, 370-381. <http://dx.doi.org/10.1016/j.ecss.2016.08.017>

553 Nicholls, R J, Wolf, J, Whitehead P G, Rahman M , Salehin, M, Hutton C, 2015 A Synthesis  
554 of Environmental Change impacts on the ESPA DELTA region (this volume|)

555 Nicholls, R. J., Whitehead, P., Wolf, J., Rahman, M., & Salehin, M. (2015). The Ganga–  
556 Brahmaputra–Meghna delta system: biophysical models to support analysis of ecosystem

557 services and poverty alleviation. *Environmental Science: Processes & Impacts*, 17(6), 1016-  
558 1017. DOI: 10.1039/C5EM90022K

559 Pathak D., Whitehead P G , Futter M N and Sinha R, 2018 Water Quality Assessment And  
560 Catchment-Scale Nutrient Flux Modeling In The Ramganga River Basin In North India: An  
561 Application Of The INCA Model *Science of the Total Environment* (this Volume)

562 Rankinen, K., Karvonen, T., Butterfield, D., 2006. An application of the GLUE methodology  
563 for estimating the parameters of the INCA N model. *Science of the total environment*, 365(1):  
564 123-139.

565 Roy, N.G. and Sinha, R. (2014). Effective discharge for suspended sediment transport of the  
566 Ganges river and its geomorphic implications. *Geomorphology*, 227, 18-30.

567 Sadoff, C 2013 Ten fundamental questions for water resources development in the Ganga:  
568 myths and realities, *Water policy* 15 (2013): 147-164.

569 Shi, Y., Gao, X., Zhang, D., and Giorgi, F.: Climate change over the Yarlung Zangbo–  
570 Brahmaputra River Basin in the 21st century as simulated by a high resolution regional  
571 climate model, *Quaternary international*, 244, 159-168, 2011.

572 Sinha, R., Jain, V., Babu, G.P., Ghosh, S., 2005. Geomorphic characterization and diversity  
573 of the fluvial systems of the Gangetic Plains. *Geomorphology*, 70(3): 207-225.

574 Spear RC and Hornberger GM, 1980 Eutrophication in the Peel Inlet II. Identification of  
575 critical uncertainties via generalised sensitivity analysis. *Water Research* 1980;14:43 - 49.

576 Wade, A .J., Durand, P., Beaujouan, V., Wessel, W. W., Raat, K. J., Whitehead, P. G.,  
577 Butterfield, D., Rankinen, K. And Lepisto, A., 2002a Towards a generic nitrogen model of  
578 European ecosystems: INCA, new model structure and equations. *Hydrol. Earth Syst. Sci.*,  
579 6, 559-582.

580 Wade, A.J., Whitehead, P.G., Butterfield, D., 2002b. The Integrated Catchments model of  
581 Phosphorus dynamics (INCA P), a new approach for multiple source assessment in  
582 heterogeneous river systems: model structure and equations. *Hydrology and Earth System  
583 Sciences*, 6(3): 583-606.

584 Wade, A., Whitehead, P., Hornberger, G., Jarvie, H., Flynn, N., 2002c. On modelling the  
585 impacts of phosphorus stripping at sewage works on in-stream phosphorus and  
586 macrophyte/epiphyte dynamics: a case study for the River Kennet. *Science of the total  
587 environment*, 282: 395-415.

588 Whitehead, P. G. Sarkar, S. and Jin, L. and Futter, M. N. and Caesar, J. and Barbour, E. and  
589 Butterfield, D. and Sinha, R. and Nicholls, R. and Hutton, C. and Leckie, H. D., 2015,  
590 Dynamic Modelling Of The Ganga River System: Impacts Of Future Climate And Socio-  
591 Economic Change On Flows And Nitrogen Fluxes In India And Bangladesh, *Environ. Sci.:*  
592 *Processes Impacts*, 17,6 1082-1097, doi 10.1039/C4EM00616J

593 Whitehead, P.G., E. Barbour, M. N. Futter, S. Sarkar, H. Rodda, J. Caesar, D. Butterfield, L.  
594 Jin, R. Sinha, R. Nicholls and M. Salehin , 2015, Impacts Of Climate Change And Socio-  
595 Economic Scenarios On Flow And Water Quality Of The Ganges, Brahmaputra And Meghna

596 (GBM) River Systems: Low Flow And Flood Statistics, Environ. Sci.: Processes Impacts,  
597 2015,17, 1057-1069, DOI: 10.1039/C4EM00619D

598 Whitehead P.G., Wilson E.J., Butterfield D. 1998a A semi-distributed Integrated Nitrogen  
599 model for multiple source assessment in Catchments (INCA): Part I - model structure and  
600 process equations. Science of the Total Environment, 210/211, 547-558.

601 Whitehead P.G., Wilson E.J., Butterfield D., Seed K. 1998b A semi-distributed Integrated  
602 flow and Nitrogen model for multiple source assessment in Catchments (INCA) : Part II -  
603 application to large river basins in south Wales and eastern England. Science of the Total  
604 Environment, 210/211, 559-583.

605 Whitehead P G, Water Quality Modelling, (2016) Wiley Statistics John Wiley & Sons, Ltd.  
606 DOI:10.1002/9781118445112.stat07793.pub2

607