

1 **Supplementary Material A: Literature overview of studies on global hydrological effects of climate change**

| <i>Study</i>          | <i>Climate model</i>   | <i>Scenario</i>          | <i>Runoff</i>  | <i>Method</i>                   | <i>Horizon</i>           | <i>Parameters</i>   | <i>Significance / Consistency</i>   | <i>Rivers / Regions</i>  | <i>Results</i>   |
|-----------------------|--|--------------------------|--|---------------------------------|--------------------------|---|---|--|--|
| Aerts (2006)          | ECBilt-CLIO-VECODE   | A2                       | Hydrological model: STREAM including simplified routing scheme | Direct use of GCM meteo data    | 21 <sup>st</sup> century | - Mean decadal change in discharge compared to discharge 1750-2000<br>- Inter decadal variability compared with natural variability | ~<br>Comparison to past variability                                       | Globe<br>Amazon, Congo, Danube, Ganges, Lena, Mekong, Mississippi, Murray, Nile, Odra, Rhine, Syr-Darya, Yukon, Volga, Volta | Discharge increase: Congo, Mekong, Ganges, Amazon, Rhine, Murray, Volga<br>Discharge decrease: Nile, Danube, Mississippi<br>Seasonal shift: Lena   |
| Alcamo (2002)         | ECHAM4<br>HadCM3   | A2, B2                   | Hydrological model: WaterGAP including routing                 | Change factor                   | 2020s, 2050s, 2080s      | - Annual withdrawal-to-availability ratio<br>- Consumption- to-Q90 ratio<br>- Per capita water availability                         | Overlap between three parameters selected as indicators of climate change | Globe  | Severe water stress: Southwestern USA, central Mexico, northeast Brazil, West Coast Latin America, northern and southern Africa, Middle East   |
| Arnell (1999b)        | HadCM2<br>HadCM3   | 1% per year CO2 increase | Hydrological model, no routing                                 | Change factor                   | 2020s, 2050s, 2080s      | - Average annual runoff<br>- Water Stress   | -   | Globe<br>42 rivers   | Change in high flow: North-America, east Asia, Ghana<br>Increasing water stress: Mediterranean region, Middle-East, South- Africa, parts of south Asia<br>Seasonal shift: Belarus          |
| Arnell (2003, 2004)   | HadCM3<br>CGCM2<br>CSIRO Mk2<br>ECHAM4<br>GFDL_R30_c<br>CCSR/NIES2 | A1, A2, B1, B2           | Hydrological model, routing with monthly output                | Change factor                   | 2020s, 2050s, 2080s      | Average annual runoff<br>Drought runoff<br>InterAnnual variability<br>Flood runoff<br>Annual cycle                                  | Consistency among scenarios, compared to consistency among models         | Globe  | Runoff increase: High latitudes, east Africa, south and east Asia<br>Runoff decrease: Southern and eastern Europe, western Russia, Middle East, Africa and much of North- and South-Africa |
| Arora and Boer (2001) | CGCM1  | GHG+A based on IS92a     | runoff from from GCM as input for routing model                | Direct use of GCM runoff fields | 2070-2100                | Mean discharge, amplitude and phase, flood discharge, annual max discharge and sdv, flow duration curve                             | -   | Globe<br>23 rivers   | Runoff decrease: Africa, Amazon, Yangtze, Mekong. Global decrease 14%<br>Seasonal shift, decrease in amplitude: Mid- and High latitude rivers  |

| <i>Study</i>  | <i>Climate model</i>                                      | <i>Scenario</i>        | <i>Runoff</i>                                      | <i>Method</i>                   | <i>Horizon</i> | <i>Parameters</i>   | <i>Significance / Consistency</i>   | <i>Rivers / Regions</i>   | <i>Results</i>   |
|---------------|---|------------------------|--|---------------------------------|----------------|---|---|---------------------------|--|
| Manabe (2004) | GFDL-GCM ensemble of 8 experiments                        | IS92a, CO2 quadrupling | Discharge derived from runoff fields GCMs          | Direct use of GCM runoff fields | 2050           | Change in annual mean runoff<br>Spatial pattern of change in seasonal soil moisture | -   | Globe<br>42 rivers        | Discharge increase: globally 7.3% by 2050, Arctic rivers, Brazil, Andes, northern India, Tibet, Indonesia, West-Africa, Amazon, Ganges, Brahmaputra<br>Discharge decrease: Nile, Mekong<br>Soil moisture decrease: North-America, Mediterranean Coast, northeast China, grasslands of Africa and southern and western regions of Australia |
| Milly (2005)  | 12 GCMs (best models from ensemble of 21 IPCC AR4 models) | A1B                    | Runoff fields from GCMs as input for routing model | Direct use of GCM runoff fields | 2041-2060      | Annual mean change compared with past trends  | Number of models showing positive changes minus number of models showing negative change                      | Globe                     | Runoff decrease: Southern Europe, Middle-East, mid-latitude western North-America, southern Africa<br>Runoff increase: High latitude North-America, Eurasia, South-America, eastern equatorial Africa  |
| Nohara (2006) | ensemble of 19 AOGCMs (part of IPCC AR4)                  | A1B                    | runoff fields from GCM as input for routing model  | Direct use of GCM runoff fields | 2100           | Mean annual change in runoff<br>River regimes                                       | - Ensemble change relative to intermodel variability of the change signal<br>- GCM deviations for control run | Globe,<br>24 major rivers | Discharge increase: Northern Hemisphere (Arctic rivers), southern to eastern Asia (Mekong, Ganges), central Africa<br>Discharge decrease: Central America, southern Africa, Mediterranean region, southern North-America, Rhine, Danube<br>Seasonal shift: Arctic and mid-latitude rivers  |

| Study             | Climate model  | Scenario     | Runoff   | Method  | Horizon      | Parameters   | Significance / Consistency  | Rivers / Regions   | Results  |
|-------------------|--|--------------|--|---|--------------|--|---|--|--|
| Nijssen (2001)    | HCCPR-CM2, HCCPR-MC3, ECHAM4, DOE-PCM3 (selected on resolution out of eight GCMs)  | IS92a, A,B,C | Hydrological model: VIC including routing        | Basin wide change factor  | 2025<br>2045 | Annual hydrological cycle<br>Change in water balance<br>Seasonal change<br>Moisture deficit periods<br>Basin sensitivity | -   | Amazon, Amur, MacKenzie, Xi, Mekong, Yellow River, Yenisei, Mississippi, Severnaya Dvina | Discharge increase: Arctic rivers<br>Discharge decrease: mid-latitude and tropic basins<br>Seasonal shift: Arctic rivers   |
| Vörösmarty (2000) | HadCM2, CGCM1  |              | Hydrological model: WBM including routing        | Direct use of GCM meteorological data<br>Change factor for discharge change | 2025         | Water stress<br>Annual runoff  | -   | Globe  | Decreased water availability: East Africa, southeast Asia, Mexico, Spain, parts of North- and South-America  |
| This study        | BCM2.0, CGCM3.1, CGCM2.3.2, CSIRO-Mk3.0, ECHAM5, ECHOG, GFDL-CM2.1, GISS_ER, HADGEM1, IPSL-CM4, MIROC3.2medres, NCAR-CCSM3 | A1B and A2   | Hydrological model: PCRGLOB-WB including routing | Direct use of meteorological data   | 2100         | Mean, max and minimum annual runoff, annual cycle, inter-annual variability  | - Significance compared to natural variability and ensemble uncertainty<br>- Consistency amongst GCMs | Globe, 20 major rivers   | Discharge increase: Arctic rivers<br>Discharge decrease: Southern Australia, southern Africa, Mediterranean region, southwest South-America<br>Seasonal shift: Sub-Arctic rivers |

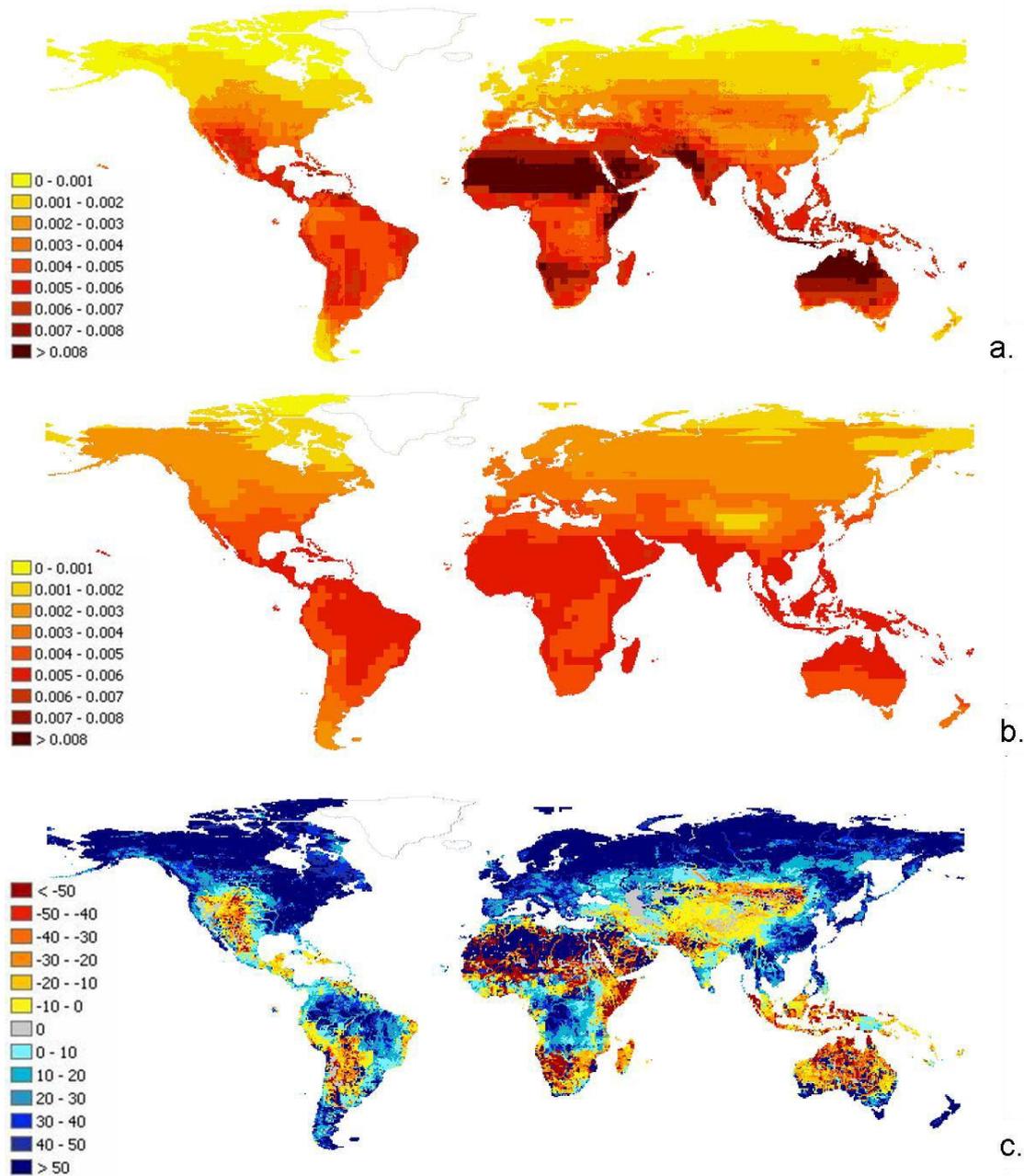
#### 4 **Supplementary Material B: Penman-Monteith vs Blaney-Criddle**

5 For most GCMs reference potential evaporation is calculated with a modification of the  
6 Penman-Monteith equation (Allen et al., 1998; Van Beek, 2008; Sperna Weiland et al.,  
7 2010). However, for some GCMs several input variables (e.g. wind speed, air pressure,  
8 radiation) required to calculate the Penman-Monteith evaporation were missing. For these  
9 GCMs the Blaney-Criddle equation is used (Brouwer and Heibloem, 1986; Oudin et al.,  
10 2005). The Blaney-Criddle equation is a simple temperature based potential evaporation  
11 estimator, whereas Penman-Monteith considers aerodynamics and radiation as well.

12 For several GCMs we compared potential evaporation calculated with the Penman-  
13 Monteith and the Blaney-Criddle equations and their resulting discharges. For brevity  
14 results are only shown for the CGCM2.3.2 model (Fig. 11a and b). For most GCMs  
15 potential evaporation calculated with the Penman-Monteith equation is high compared to  
16 Blaney-Criddle potential evaporation in Northern Australia, the Sahara, Southern Africa,  
17 the southwest US and Northern India and relatively low for Europe, the northern US,  
18 Canada, Russia, southeast Asia and the Amazon. However, only for specific periods, and  
19 in regions where evaporation limitation by soil moisture conditions is small, deviations in  
20 potential evaporation will introduce deviations in actual evaporation and runoff. Fig. 11c  
21 shows the percentage difference in discharge calculated using either the Blaney-Criddle  
22 or the Penman-Monteith potential evaporation. Deviations are large for the northern  
23 regions of the northern Hemisphere, the Amazon basin, Europe and parts of southeast  
24 Asia where discharge calculated with Penman-Monteith potential evaporation is  
25 relatively high. The Penman-Monteith based discharge is relatively low in arid regions,  
26 the Indus basin and Himalayas. Unfortunately hydrological studies are restricted to the  
27 available GCM datasets and, since not all required Penman-Monteith variables are  
28 reported for all GCMs, the use of a simple temperature based equation like Blaney-  
29 Criddle can not be avoided. Still, for those GCMs where all variables were available we  
30 preferred to use the FAO recommended Penman-Monteith equation (Allen et al., 1998).

31 In this study not the absolute discharge quantities, but the changes in average discharge  
32 and discharge extremes were of interest. Therefore we analyzed the influence of using  
33 either Blaney-Criddle or Penman-Monteith potential evaporation as input to the  
34 hydrological model on the resulting discharge changes. Hereto discharge changes derived

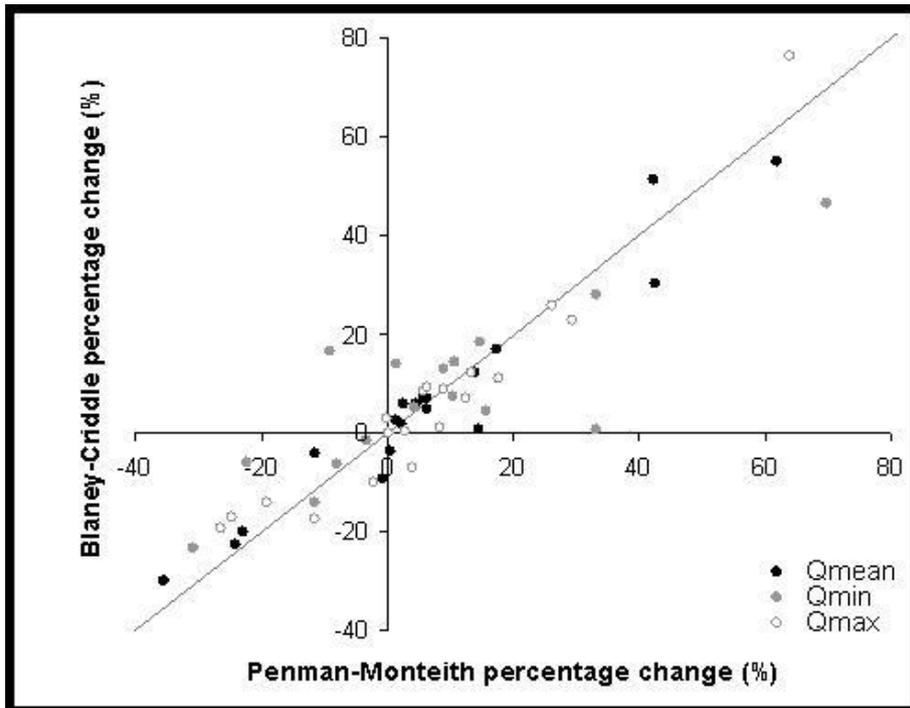
35 with the hydrological model forced with Blaney-Criddle potential evaporation are  
36 regressed on discharge changes derived with the hydrological model forced with Penman-  
37 Monteith potential evaporation (Fig. 12). For this analysis we used data from the first  
38 realization of CGCM2.3.2 for the 20C3M experiment and A1B scenario. Overall, for  
39 2100, the different potential evaporation equations result in similar directions of change.  
40 There are two exceptions. The first is the direction of change for maximum discharge in  
41 the MacKenzie, which is negative when using the Blaney-Criddle equation and positive  
42 for the Penman-Monteith equation. This is a result of the large differences in absolute  
43 discharge quantities for the MacKenzie which tend to be twice as high for the Penman-  
44 Monteith equation. The second exception is the Ganges where minimum discharge  
45 decreases with the Penman-Monteith method, while it increases for the Blaney-Criddle  
46 method. For the remaining catchments directions of change in minimum, maximum and  
47 mean discharge are the same when using either two equations. In general the projected  
48 changes follow the 1:1 slope, although differences in magnitude of projected change  
49 exist.



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51 Figure 11: Maps with reference potential evaporation (m/day) and resulting percentage discharge  
 52 difference (%). Fig. 11a. twenty year average reference potential evaporation (m/day) calculated  
 53 with the Penman-Monteith equation. Fig. 11b. twenty year average reference potential  
 54 evaporation (m/day) calculated with the Blaney-Criddle equation and Fig. 11c. percentage  
 55 difference (%) between twenty year average discharges calculated with PCR-GLOBWB from  
 56 either Penman-Monteith or Blaney-Criddle.

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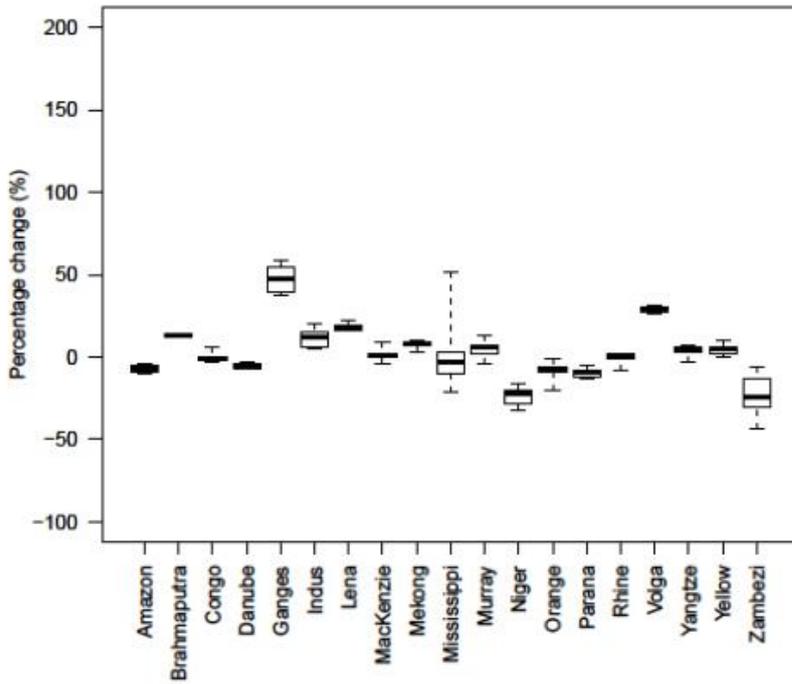
59 Figure 12: Percentage change in discharge calculated using potential evaporation derived with  
 60 Blaney-Criddle vs Penman-Monteith. Black dots represent projected change in average discharge,  
 61 grey dot represent changes in high flows (Qmax), white dots represent changes in low flows  
 62 (Qmin). The solid line represents the 1:1 slope.

63 **Supplementary Material C: Consistency of change for multiple realizations**  
64 **of one GCM**

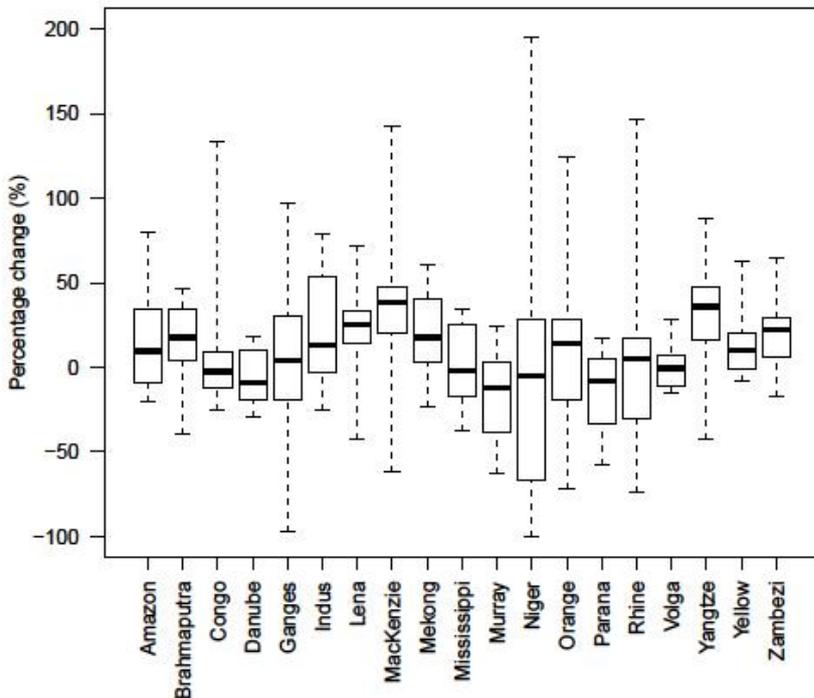
65 For CGCM2.3.2, the GCM with the highest number of realizations for both the 20C3M  
66 experiment and the A1B scenario, we calculated change in discharge by 2100 for all five  
67 available realizations. Boxplots of projected changes for the 19 catchments are shown in  
68 Fig. 13a. Fig. 13b shows boxplots of changes projected by the twelve individual GCMs  
69 included in our ensemble. The boxplots of the five realizations of CGCM2.3.2 cover  
70 much smaller ranges than the boxplots derived from the ensemble of GCMs.  
71 Furthermore, for 13 out of 19 catchments the direction of change is consistent for all five  
72 realizations of the CGCM2.3.2 model. In Fig. 14a a global map with the number of  
73 CGCM2.3.2 realizations projecting change in the dominant direction (the direction of  
74 change projected by the majority of GCMs) is shown. For 55% of the globe all five  
75 realizations agree on the projected direction of change, for 81% of the globe at least four  
76 realizations agree on the direction of change and for the remaining 19% only three  
77 realizations are consistent.

78 From these results two conclusions can be drawn. Firstly, Fig. 13 shows the inter-model  
79 uncertainty is much larger than the intra-model uncertainty, at least for the GCM data we  
80 have at our disposal. And secondly, for the majority of catchments the projected  
81 directions of change are consistent for the five realizations. This indicates that including  
82 different numbers of realizations for the individual GCMs in our ensemble would result  
83 in overweighting the direction of change projected by the GCMs with multiple  
84 realizations. Therefore we restricted ourselves to a single realization for each of the  
85 twelve GCMs included in the ensemble.

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109 Figure 13a. Boxplots of changes projected by the five available realizations of the GCM CGCM  
110 2.3.2 for the A1B scenario for 2100. Whiskers mark the maximum and minimum projected  
111 changes, boxes span the quartile range and horizontal dashes represent the median of projected  
112 changes. Figure 13b. same as Fig. 13a. but now for the changes projected by the ensemble of 12  
113 GCMs.



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115 Figure 14: Map showing the number of realizations of CGCM2.3.2 projecting mean change in the  
116 dominant direction. Black indicates regions where the five realizations project changes in the  
117 same direction, grey indicates regions where four realizations project similar directions of change  
118 and in the white regions only three realizations project the same direction of change. The  
119 dominant direction is the direction of change projected by the majority of models.