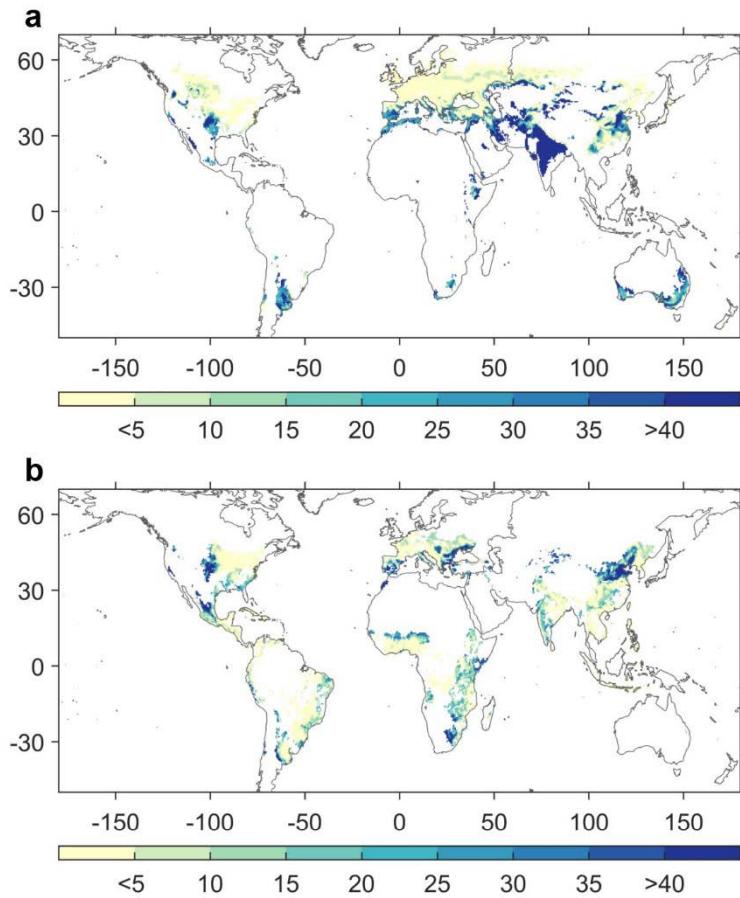
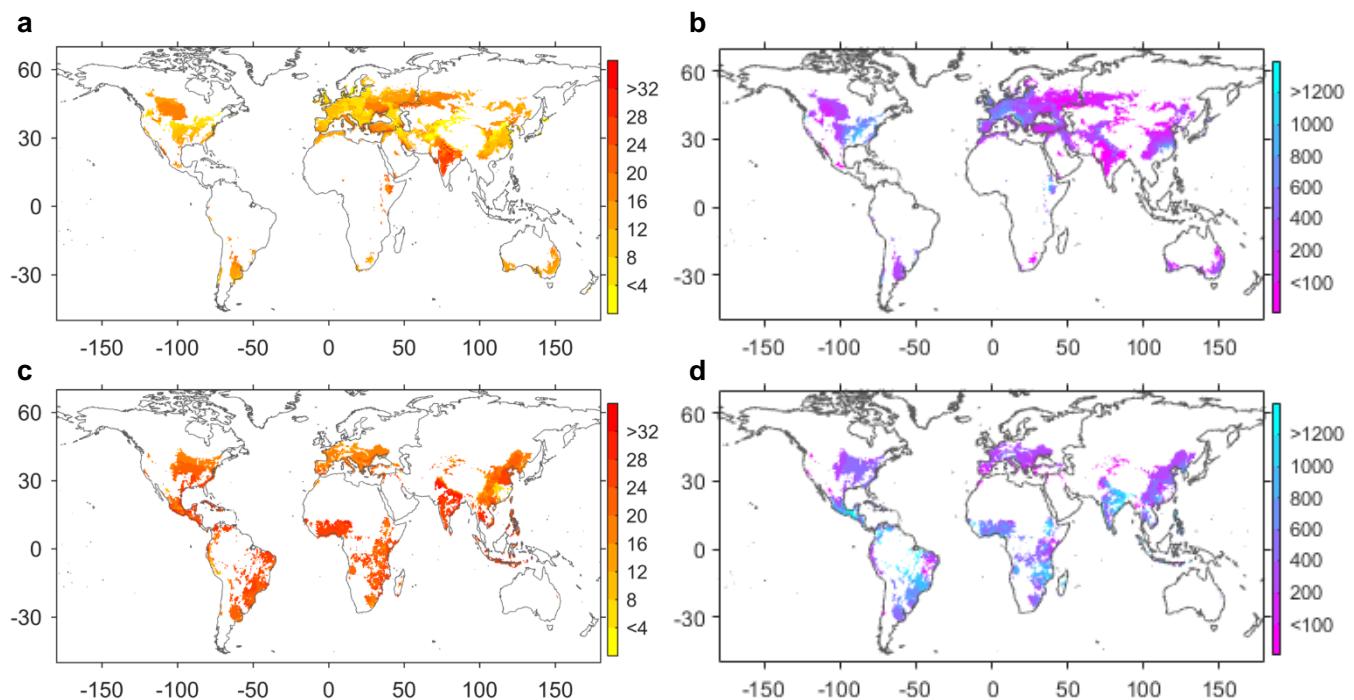


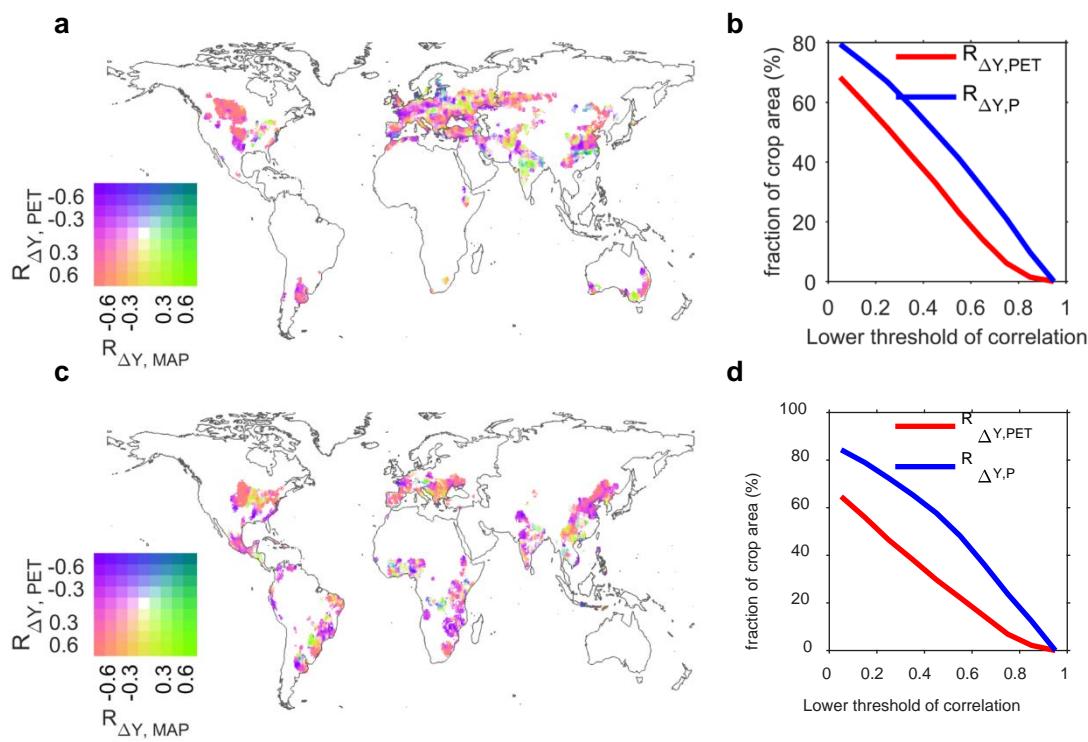
**Supplementary Figure 1.** Spatial distribution of  $\Delta Y$  from climate analogue approach for **a** wheat and **b** maize over contemporary growing area.



**Supplementary Figure 2.** Spatial distribution of mean growing season temperature and precipitation for wheat (top panels) and for maize (bottom panels) during 1980-2010. **a** mean growing season temperature ( $^{\circ}\text{C}$ ) for wheat; **b** mean growing season precipitation (mm) for wheat; **c** mean growing season temperature ( $^{\circ}\text{C}$ ) for maize; **d** mean growing season precipitation for maize (mm).

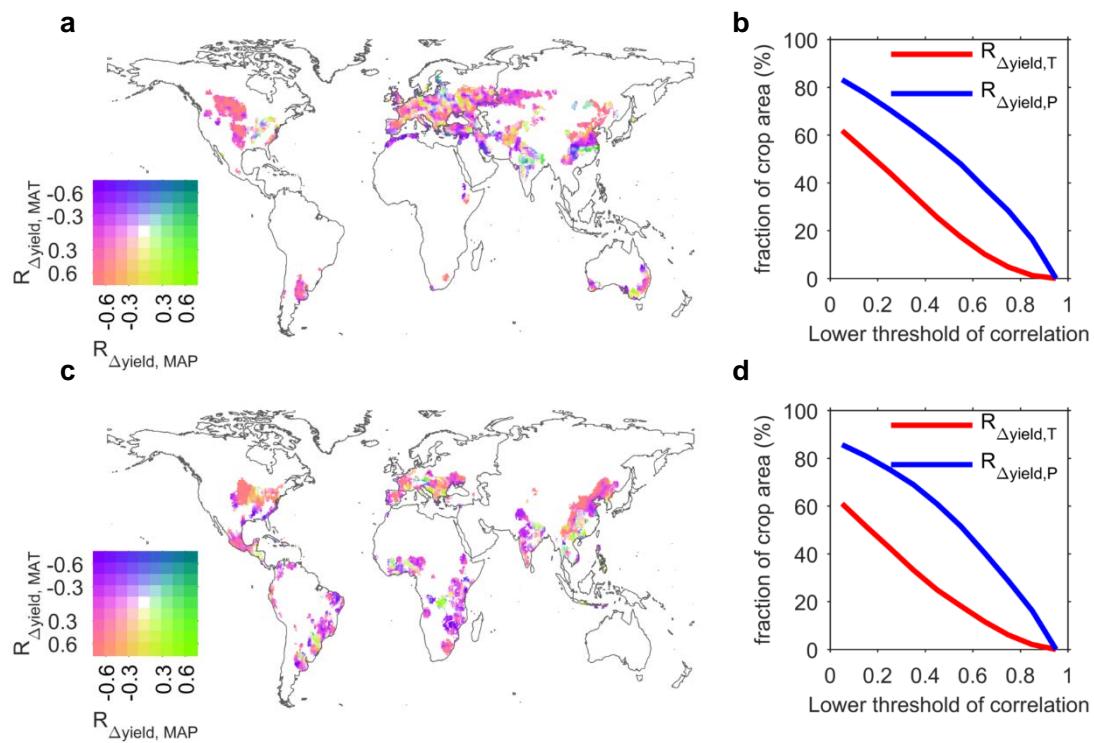


**Supplementary Figure 3.** Partial correlation in the spatial domain between  $\Delta Y$  and climatic variables (potential evapotranspiration (PET) and mean annual precipitation (MAP)) for wheat (top panel) and for maize (bottom panel). **a,c** bivariate mapping for spatial distribution of the partial correlation coefficient between  $\Delta Y$  and PET ( $R_{\Delta Y, PET}$ ) and that between  $\Delta Y$  and MAP ( $R_{\Delta Y, MAP}$ ). **b,d** Percentage of cropland area where  $\Delta Y$  is controlled by PET or precipitation depending on the chosen threshold (x-axis) for the partial correlation coefficients. PET was calculated following Penman-Monteith equations provided by Harris et al.<sup>1</sup>.

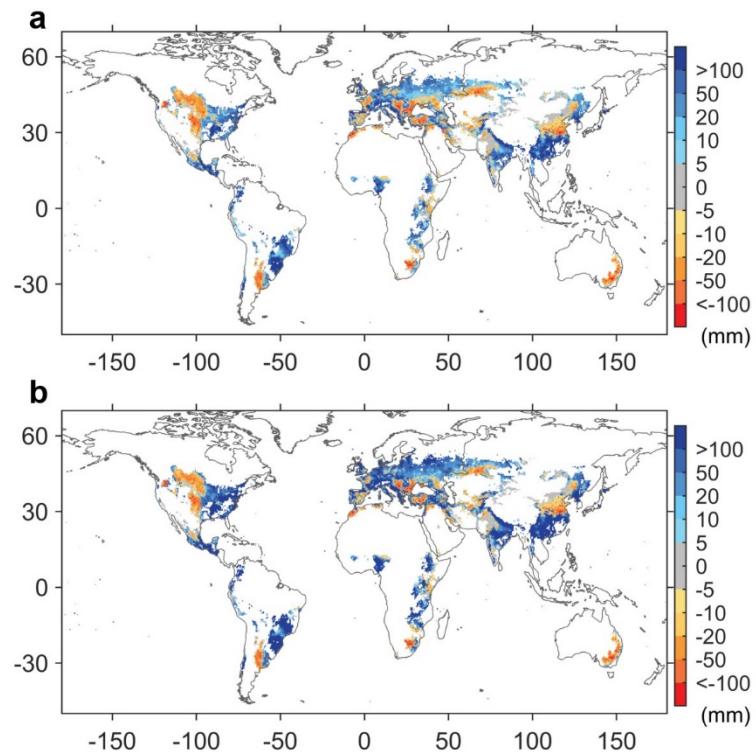


**Supplementary Figure 4.** Partial correlation in the spatial domain ( $3.5^\circ \times 3.5^\circ$  moving windows)

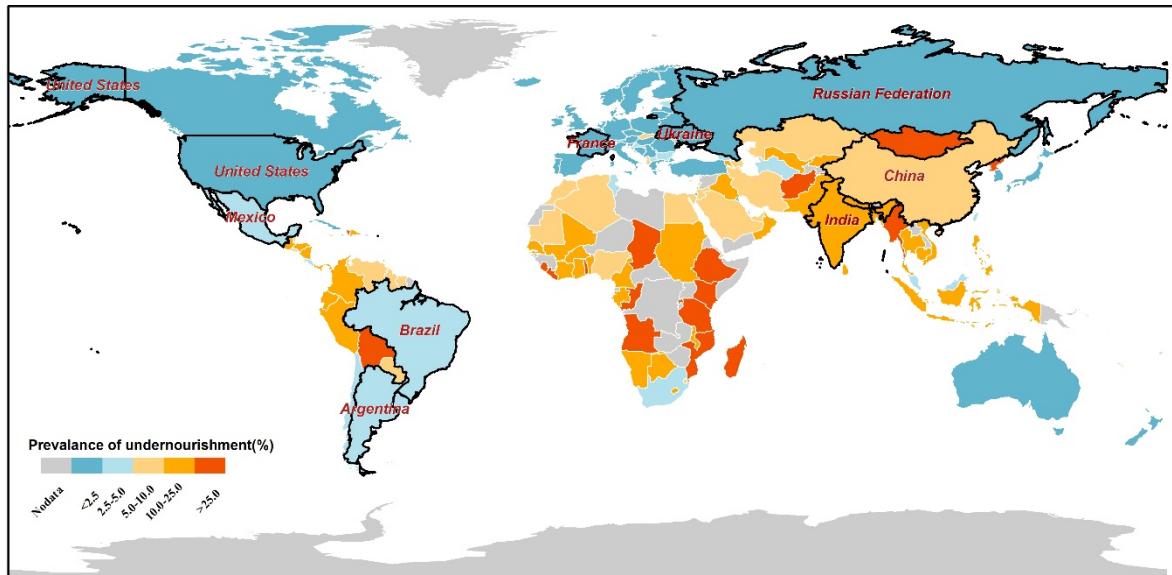
between  $\Delta Y$  and climatic variables (mean annual temperature (MAT) and mean annual precipitation (MAP)) for wheat (top panel) and for maize (bottom panel). **a,c** bivariate mapping for spatial distribution of the partial correlation coefficient between  $\Delta Y$  and MAT ( $R_{\Delta Y, \text{MAT}}$ ) and that between  $\Delta Y$  and MAP ( $R_{\Delta Y, \text{MAP}}$ ). **b,d** Percentage of cropland area where  $\Delta Y$  is controlled by temperature or precipitation depending on the chosen threshold (x-axis) for the partial correlation coefficients. Same to Figure 3 but using MAT to replace PET.



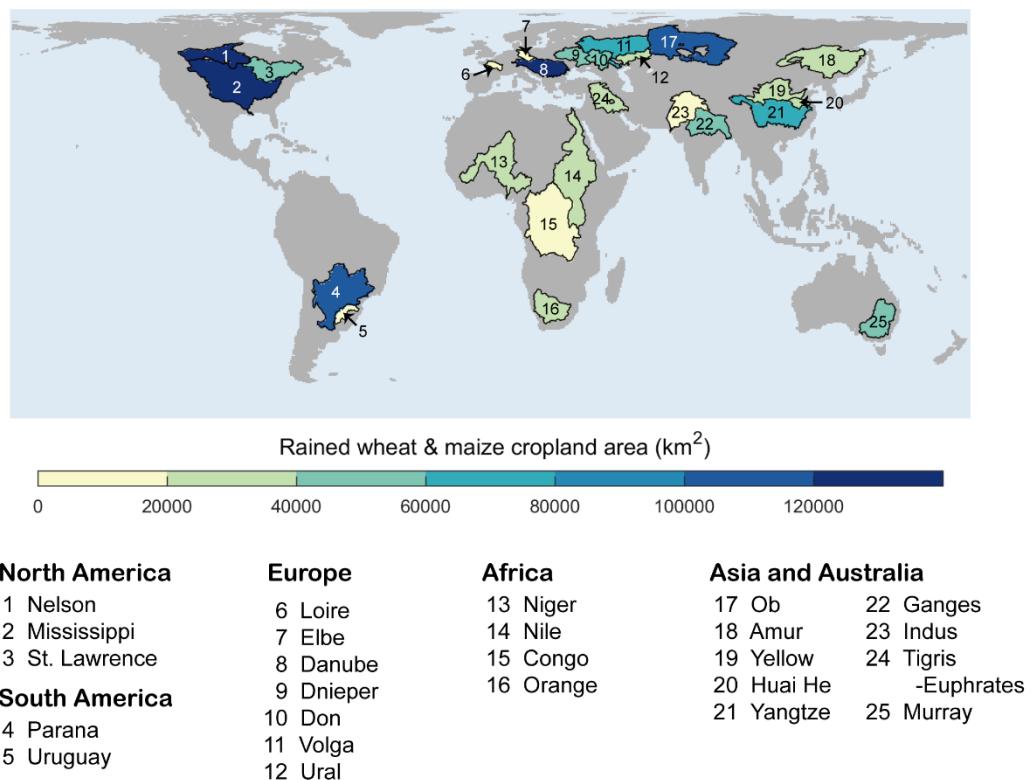
**Supplementary Figure 5.** The spatial distribution of the difference between irrigation demand and available runoff resources determined with maximum runoff usage threshold of **a** 20% and **b** 40%.



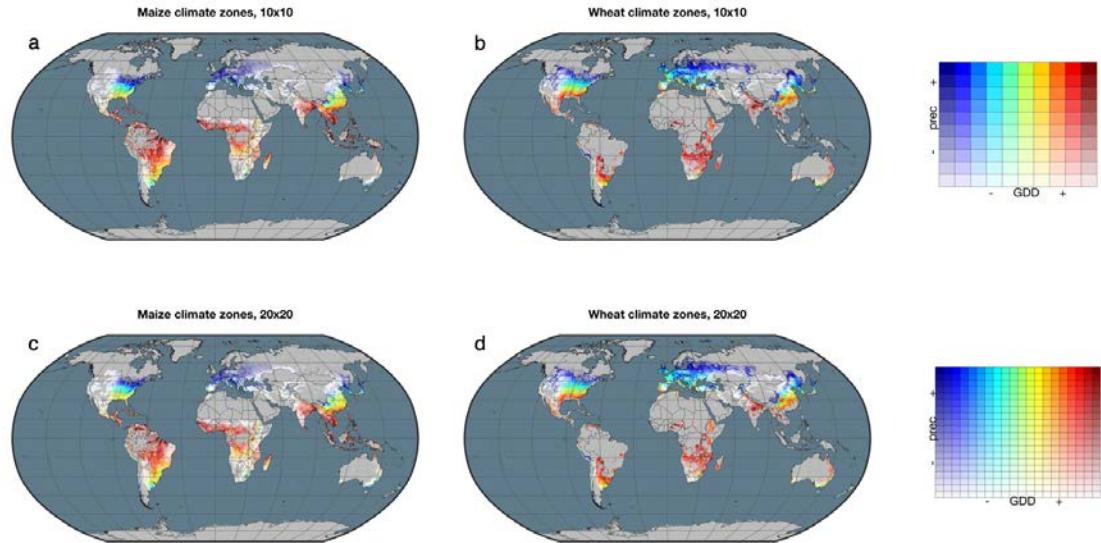
**Supplementary Figure 6.** Spatial distribution of prevalence of undernourishment (%) during 2000-2010 according to Food and Agriculture Organization of the United Nations (<https://unstats.un.org/sdgs/indicators/database/?indicator=2.1.1>). Solid black lines delineate major producers of wheat and maize and their names.



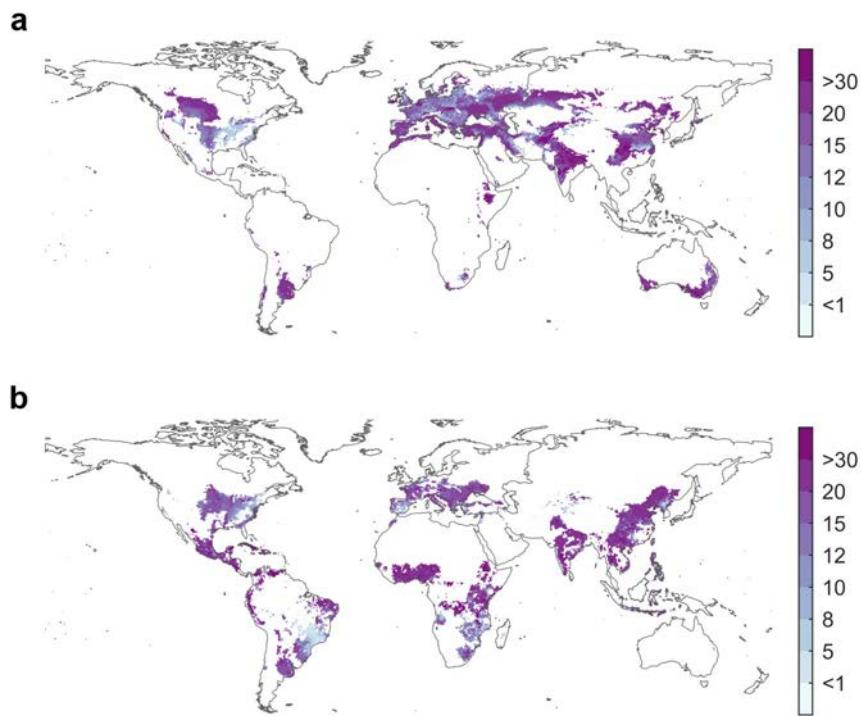
**Supplementary Figure 7.** Spatial distribution of top 25 river basins having largest rainfed wheat and maize croplands. The color depicts the area of rainfed wheat and maize croplands within the basin according to Portmann et al.<sup>2</sup>.



**Supplementary Figure 8.** Maps of the 10x10 and 20x20 climate zones (bounding the range we utilize) for wheat and maize used by the climate analogue approach.



**Supplementary Figure 9.** The spatial distribution of the standard deviation of  $\Delta Y$  (%) by GGCMs. **a** wheat, **b** maize.



**Supplementary Table 1.** Balance of river discharge and irrigation demand of contemporary rainfed wheat and maize croplands for 25 river basins with largest rainfed area of wheat and maize. River discharge is the mean annual discharge of the gauging station nearest to the mouth that is represented in GRDC database ([https://www.bafg.de/GRDC/EN/01\\_GRDC/13\\_dtbse/database\\_node.html](https://www.bafg.de/GRDC/EN/01_GRDC/13_dtbse/database_node.html)). Irrigation demand is estimated by reanalyzed irrigation demand by GGCMs (see Methods). Rainfed crop area is derived from Portmann et al.<sup>2</sup>.

Basin Name	Rainfed area (10 <sup>3</sup> km <sup>2</sup> )	Discharge (10 <sup>3</sup> m <sup>3</sup> )	Demand (10 <sup>3</sup> m <sup>3</sup> )	Demand to supply ratio (%)
Mississippi	367.2	535.0	60.0	11.2
Danube	126.0	202.3	25.0	12.3
Nelson river	120.4	95.5	14.7	15.4
Ob	110.9	400.4	17.2	4.3
Parana	106.5	476.3	10.6	2.2
Volga	64.7	256.7	2.7	1.1
Yangtze river	60.2	899.4	3.7	0.4
Don	56.5	25.5	7.7	30.0
Murray	51.8	6.7	24.9	372.0
Ganges	47.0	379.6	6.4	1.7
Dniepr	46.7	47.1	3.6	7.7
St.lawrence	45.6	268.2	2.0	0.8
Amur	39.4	314.7	4.2	1.3
Huai River	30.4	27.9	7.8	28.0
Tigris & euphrates	29.7	37.6	15.6	41.6
Yellow river	27.5	45.0	9.3	20.8
Orange	27.2	9.0	13.7	152.4

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Niger	25.7	159.5	0.3	0.2
Ural	25.3	9.4	5.0	53.4
Nile	21.2	39.5	1.8	4.5
Congo	19.5	1269.3	0.1	0.0
Indus	19.2	91.6	3.6	3.9
Uruguay	19.0	170.5	0.3	0.2
Elbe river	17.8	22.4	1.6	7.1
Loire	17.3	26.4	4.3	16.1

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**Supplementary Table 2.** Characteristics of used crop models

Model	Type <sup>1</sup>	CO2 effects <sup>2</sup>	Stress es <sup>3</sup>	Fertilizer application <sup>4</sup>	Calibration <sup>5</sup>	Calibrated parameters	Reference <sup>6</sup>
<b>EPIC-BOKU</b>	Site-based	RUE, TE	W, T, A, N, P, BD, AL	automatic N input (max 200 kg Ha-1 yr-1) PK (national stat. IFA) dynamic application	Site-specific (EPIC 0810)	NA	Ref. 3
<b>EPIC-IIASA</b>	Site-based	RUE, TE	W, T, A, N, P, BD, AL	NP (sub-national stat by Balkovič et al. (2013); Mueller et al. (2012)); P timing: rigid; N timing: automatic (based on N stress)	Site-specific and global	F, HIpot (ric, mai) F (others )	Ref. 3
<b>EPIC-TAMU</b>	Site-based	RUE, TE	W, T, H, A, N, P, BD, AL	NPK at planting	Site-specific and global	HIpot (maize)	Ref. 4
<b>GEPIC</b>	Site-based	RUE, TE	W, T, A, N, P, BD, AL	NP (national stat: FertiSTAT), dynamic application of N, rigid application of P	Site-specific and global	F HIpot (for maize and rice)	Ref. 5
<b>LPJ-GUESS</b>	Ecosystem	LF, SC	W, T	NA	Uncalibrated	NA	Ref. 6
<b>LPJmL</b>	Ecosystem	LF, SC	W, T	NA	National	LAImax HI oa	Ref. 7
<b>ORCHID EE-crop</b>	Ecosystem	LF, SC	W,T,N	Automatic N input (IFA)	Uncalibrated	-	Ref. 8
<b>pAPSIM</b>	Site-based	RUE	W, T, H, A, N	SPAM by You et al. (2014), (1/2 at planting, 1/2 at day 45)	Site-specific (APSIM)	NA	Ref. 9
<b>pDSSAT</b>	Site-based	RUE (for wheat, rice,	W, T, H, A, N	SPAM by You et al. (2014), (1/2 at planting, 1/2 at	Site-specific (DSSAT)	NA	Ref. 9

		maize) and LF (for soybean)		day 45)			
<b>PEGASU S</b>	Ecosyst em	RUE, TE	W, T, H, N, P, K	NPK (national stat. IFA), annual application	Global	$\beta$	Ref. 10

Notes: (NA where not applicable)

<sup>1</sup> Site-based: site-base crop model; Ecosystem: global ecosystem model

<sup>2</sup> Elevated CO<sub>2</sub> effects: LF: Leaf-level photosynthesis (via rubisco or quantum-efficiency and leaf-photosynthesis saturation; RUE: Radiation use efficiency; TE: Transpiration efficiency; SC: stomatal conductance

<sup>3</sup> W: water stress; T: temperature stress; H: specific-heat stress; A: oxygen stress; N: nitrogen stress; P: phosphorus stress; K: potassium stress; BD: bulk density; AL: aluminum stress (based on pH and base saturation)

<sup>4</sup> Fertilizer application, timing of application; NPK annual application of total NPK (nutrient-stress factor); source of fertilizer application data; timing: annual or dynamic

<sup>5</sup> F: fertilizer application rate; HI<sub>pot</sub>: Potential harvest index; LAI<sub>max</sub>: maximum LAI under unstressed conditions; HI: harvest index;  $\alpha_a$ : factor for scaling leaf-level photosynthesis to stand level;  $\beta$ : radiation-use efficiency factor; TH: Total Heat unit required for the maturity; TC: Technological coefficient; TS: Temperature sensitivity of photosynthesis; LR: ratio of leaf to above ground biomass.

<sup>6</sup> See Supplementary Reference.

**Supplementary Table 3.** irrigation contribution to yield ( $\Delta Y$ ) (%) for major wheat and maize producers

	Wheat		Maize
China	42.2	USA	24.9
India	53.5	China	22.6
Russia	15.7	Brazil	22.2
USA	31.9	France	24.4

**Supplementary Table 4.** Web links to GGCMI model output

Model	Wheat	Maize
<b>EPIC-BOKU</b>	<a href="http://dx.doi.org/10.5281/zenodo.1404761">http://dx.doi.org/10.5281/zenodo.1404761</a>	<a href="http://dx.doi.org/10.5281/zenodo.1404767">http://dx.doi.org/10.5281/zenodo.1404767</a>
<b>EPIC-IIASA</b>	<a href="http://dx.doi.org/10.5281/zenodo.1403195">http://dx.doi.org/10.5281/zenodo.1403195</a>	<a href="http://dx.doi.org/10.5281/zenodo.1403203">http://dx.doi.org/10.5281/zenodo.1403203</a>
<b>EPIC-TAMU</b>	<a href="http://dx.doi.org/10.5281/zenodo.1409013">http://dx.doi.org/10.5281/zenodo.1409013</a>	<a href="http://dx.doi.org/10.5281/zenodo.1409009">http://dx.doi.org/10.5281/zenodo.1409009</a>
<b>GEPIC</b>	<a href="http://dx.doi.org/10.5281/zenodo.1408571">http://dx.doi.org/10.5281/zenodo.1408571</a>	<a href="http://dx.doi.org/10.5281/zenodo.1408577">http://dx.doi.org/10.5281/zenodo.1408577</a>
<b>LPJ-GUESS</b>	<a href="http://dx.doi.org/10.5281/zenodo.1408623">http://dx.doi.org/10.5281/zenodo.1408623</a>	<a href="http://dx.doi.org/10.5281/zenodo.1408647">http://dx.doi.org/10.5281/zenodo.1408647</a>
<b>LPJmL</b>	<a href="http://dx.doi.org/10.5281/zenodo.1403013">http://dx.doi.org/10.5281/zenodo.1403013</a>	<a href="http://dx.doi.org/10.5281/zenodo.1403073">http://dx.doi.org/10.5281/zenodo.1403073</a>
<b>ORCHIDEE-crop</b>	<a href="http://dx.doi.org/10.5281/zenodo.1408191">http://dx.doi.org/10.5281/zenodo.1408191</a>	<a href="http://dx.doi.org/10.5281/zenodo.1408199">http://dx.doi.org/10.5281/zenodo.1408199</a>
<b>pAPSIM</b>	<a href="http://dx.doi.org/10.5281/zenodo.1403183">http://dx.doi.org/10.5281/zenodo.1403183</a>	<a href="http://dx.doi.org/10.5281/zenodo.1403189">http://dx.doi.org/10.5281/zenodo.1403189</a>
<b>PDSSAT</b>	<a href="http://dx.doi.org/10.5281/zenodo.1403171">http://dx.doi.org/10.5281/zenodo.1403171</a>	<a href="http://dx.doi.org/10.5281/zenodo.1403181">http://dx.doi.org/10.5281/zenodo.1403181</a>
<b>PEGASUS</b>	<a href="http://dx.doi.org/10.5281/zenodo.1409546">http://dx.doi.org/10.5281/zenodo.1409546</a>	<a href="http://dx.doi.org/10.5281/zenodo.1409550">http://dx.doi.org/10.5281/zenodo.1409550</a>

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