



Supplement of

The relevance of uncertainty in future crop production for mitigation strategy planning

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3 Uncertainty in future crop production and its role in climate

- 4 mitigation strategies

1 Short descriptions of the Global Gridded Crop Models (GGCM)

model	CO2	Fertilizer use	Adaptation	starting conditions
	fertili			(representation of
	zation			present day yields
				or potential
				yields?)
EPIC (1, 2)	RUE,	flexible N	annual adjustment	present day
	IE	application rates (N-	of planting dates;	potential yields
Environmental		stress free days in	total heat units to	
Policy		90% of crop growing	reach maturity	
Integrated		period to an upper	remain constant	
Climate		application limit of		
		200 kg ha-1)	no adjustment of	
		Constant P	cultivars	
		application rates.		
GEPIC (3, 4)	RUE,	TIEXIBLE N	decadal	present day yields
	IE	application based on	adjustment of	
GIS-based		N stress >10%	planting dates,	
agroecosystem		(limitation of	total heat units to	
model		potential biomass	reach maturity	
integrating a		increase due to N	remain constant	
bio-physical		stress) up to an		
EPIC model		upper national	Adjustment of	
(Environmental		application limit	winter and spring	
Policy		according to	wheat sowing	
Integrated		FertiStat, fixed	areas based on	
Climate) with a		present day P	temperature	
Geographic		application rates		
Information		TOILOWING FAU		
System (GIS)		rertiStat database		
		(2010)(5)		
GAEZ-IMAGE (6)	LLP	Soil nutrient limiting	Adjustment of	present day yields
(called IMAGE in		factors are not	planting dates,	

the main text)		accounted for	total heat units to	
			reach maturity	
Integrated			remain constant	
Model to Assess				
the Global			Adjustment of	
Environment			summer and	
			winter varieties in	
			case of wheat and	
			maize	
LPJ-GUESS	LLP,	Soil nutrient limiting	Adjustment of	potential yields
	CC	factors are not	planting dates and	
		accounted for	total heat units to	
			reach maturity	
			Deserved	
			Decadal	
			adjustment of crop	
			cultivar to give	
			appropriate	
			maturity dates (7)	
I PImL (8)	LI P.	Soil nutrient limiting	Fixed sowing dates	present day vields
		factors are not	(9) total heat units	(10)
		accounted for	to reach maturity	(10)
			remain constant	
PEGASUS (11)	RUE,	Fixed N, P, K	adjustment of	present day yields
	TE	application rates	planting dates,	
Predicting		(IFA national	variable heat units	
Ecosystem		statistics)	to reach maturity	
Goods And		,		
Services Using				
Scenarios				
pDSSAT	RUE,	Fixed N present day	No adjustment of	present day yields
	LLP,	application rates	planting dates,	
	CC		total heat units to	
			reach maturity	
			remain constant	

2 Table S1: Basic crop model characteristics with respect to the implementation of CO2

3 fertilization effect (as affecting 1) radiation use efficiency (RUE), 2) transpiration efficiency (TE),

4 3) leaf level photosynthesis (LLP), 4) canopy conductance (CC)), the accounting for nutrient

5 constraints with respect to the CO2 fertilization effect, the assumption with respect to fertilizer

- 1 application (N = nitrogen, P = Phosphorus, K = Potassium), implemented adaptation measures,
- 2 and starting conditions.
- 3

4 2 Relative change in global production per global warming

5 2.1 Scenario Independence

6 The following plots show the global productions values for the four high priority crops under 7 different management assumptions (unlimited irrigation, no irrigation and current irrigation) 8 and CO2 concentrations ("noCO2" = based on fixed historical CO2 levels not accounting for the 9 positive effect on crop production based on increasing CO2 (see caption of Figure S6 for the crop model specific levels of CO2); "CO2" = CO2 concentrations follow the historical evolution 10 and the Representative Concentration Pathways (RCPs) afterwards.) The Figures show a relative 11 close relationship between global productions and global mean warming. The relationship is 12 only weakly dependent of the emissions scenarios. 13



2 Figure S1: Relative changes in global wheat production with respect to the 1980-2010 reference period. Reference values are based on the "current irrigation + CO2" runs and identical for all 3 4 considered management and CO2 scenarios. Color coding: Climate scenarios (red = RCP8.5, orange = RCP6.0, light blue = RCP4.5, dark blue = RCP2.6). Symbols: irrigation scenarios (dots = 5 6 unlimited irrigation on the present day harvested area for wheat (MIRCA), squares = unlimited 7 irrigation on the currently irrigated wheat area (MIRCA), triangles = global productions 8 assuming no irrigation on the present day area used for wheat). Dashed lines represent the noCO2 runs while solid lines are based on the associated RCP-CO2 data. 9



2 Figure S2: Relative changes in global maize production with respect to the 1980-2010 reference period. Reference values are based on the "current irrigation + CO2" runs and identical for all 3 4 considered management and CO2 scenarios. Color coding: Climate scenarios (red = RCP8.5, 5 orange = RCP6.0, light blue = RCP4.5, dark blue = RCP2.6). Symbols: irrigation scenarios (dots = 6 unlimited irrigation on the present day harvested area for maize (MIRCA), squares = unlimited 7 irrigation on the currently irrigated maize area (MIRCA), triangles = global productions assuming 8 no irrigation on the present day area used for maize). Dashed lines represent the noCO2 runs while solid lines are based on the associated RCP-CO2 data. 9



2 Figure S3: Relative changes in global rice production with respect to the 1980-2010 reference 3 period. Reference values are based on the "current irrigation + CO2" runs and identical for all considered management and CO2 scenarios. Color coding: Climate scenarios (red = RCP8.5, 4 5 orange = RCP6.0, light blue = RCP4.5, dark blue = RCP2.6). Symbols: irrigation scenarios (dots = 6 unlimited irrigation on the present day area used for rice (MIRCA), squares = unlimited 7 irrigation on the currently irrigated rice area (MIRCA), triangles = global productions assuming 8 no irrigation on the present day area used for rice). Dashed lines represent the noCO2 runs 9 while solid lines are based on the associated RCP-CO2 data.



2 Figure S4: Relative changes in global soy production with respect to the 1980-2010 reference 3 period. Reference values are based on the "current irrigation + CO2" runs and identical for all 4 considered management and CO2 scenarios. Color coding: Climate scenarios (red = RCP8.5, orange = RCP6.0, light blue = RCP4.5, dark blue = RCP2.6). Symbols: irrigation scenarios (dots = 5 6 unlimited irrigation on the present day area used for soy (MIRCA), squares = unlimited irrigation 7 on the currently irrigated soy area (MIRCA), triangles = global productions assuming no irrigation on the present day area used for soy). Dashed lines represent the noCO2 runs while 8 9 solid lines are based on the associated RCP-CO2 data.

10



1 **2.2** Trend in global production in terms of global mean warming

2

Figure S5: Individual bars and dots are identical to the elements of the first column of each global mean warming bin shown in Figure 1 of the main text except for the linear trend lines included here. They are based on a linear regression based on the four crop model specific mean values (colored dots at each level of global warming) and forced through zero. The associated slopes are reported in Table S2.



1

Figure S6: Analogous to Figure S5 but based on the available " constant CO2" runs for the HadGEM2, IPSL, and MIROC model reaching 4 K of warming in comparison to the 1980-2010 reference value. Fixed CO2 levels for constant CO2 runs: 380 ppm (year 2005) for EPIC, 364 ppm (year 2000) for GEPIC, 369 ppm (year 2000) for IMAGE; 379 ppm (year 2005) for LPJ-GUESS; 370 ppm (year 2000) for LPJmL; 369 ppm (year 2000) for PEGASUS; 330 ppm (year 1975) for pDSSAT. The slopes of the regression lines are reported in Table S3.

- 8
- 9
- 5
- 10

	Wheat ($\Delta P / \Delta T$)		Maize ($\Delta P / \Delta T$)		Rice ($\Delta P / \Delta T$)		Soy ($\Delta P / \Delta T$)	
	[%/K]		[%/K]		[%/K]		[%/K]	
	\uparrow	\rightarrow	\uparrow	\rightarrow	\uparrow	\rightarrow	\uparrow	\rightarrow
EPIC	-0.19	-5.15	-5.32	-7.82	-3.73	-8.60	-3.85	-8.98
GEPIC	1.31	-1.94	-1.71	-5.41	-2.05	-6.69	-0.23	-7.03
IMAGE	-1.57	-5.33	1.22	-0.65	3.34	-1.04	6.27	3.26
LPJ-GUESS	7.48	-4.54	5.90	3.77	19.35	-3.56	12.31	-4.97
LPJmL	0.96	-5.18	-0.59	-4.05	4.45	-6.55	7.44	-8.96
pDSSAT	0.56	-6.30	-1.68	-5.36	0.22	-4.72	-0.66	-10.21
PEGASUS	-3.05	-10.16	-8.00	-9.23			-8.04	-14.30
median	0.56	-5.18	-1.68	-5.36	1.78	-5.64	-0.23	-8.96
mean	0.78	-5.51	-1.47	-4.11	3.60	-5.19	1.89	-7.13
stdev	3.33	2.45	4.47	4.42	8.32	2.68	7.08	5.48
stdev	2.02		0.08		7 26		E E0	
CO2 effect	2.92		0.98		7.30		5.50	

Table S2: Relative change in global crop production [%/K] assuming fixed present day harvested areas and unlimited irrigation on the reported irrigated land (12). Results are listed for the model runs accounting for the CO2 fertilization effect under changing CO2 concentration corresponding to the RCP scenarios (\uparrow) and fixed CO2 concentrations (\rightarrow).

6

7 3 Calculation global crop production under additional irrigation accounting for

8 water constraints

9 **3.1** Available water for irrigation

10 Approximation of the overall amount of water available in a Food Production Unit (FPU)

11 To estimate the amount of water available for irrigation within each FPU we use the sum of surface and sub-surface runoff provided by the different ISI-MIP water models (see Table S3 for 12 their basic characteristics). This means that the "available irrigation water" used within our 13 study only includes renewable surface and groundwater but not groundwater that could be 14 15 withdrawn in excess of groundwater recharge (thus leading to groundwater depletion). Part of the irrigation water available within an FPU might not be generated within the same FPU but 16 17 transported from an upstream FPU. The spatially aggregated runoff only represents a proxy for the irrigation water "generated" in the associated FPU. To also account for the transported 18 19 water we proceed in the following way:

The overall amount of water generated within a river basin (aggregated runoff) is re distributed within the same river basin according to the discharge values. In this way the overall
 amount of water does not change but is "concentrated" in the rivers.

4 2. The re-distributed runoff is afterwards aggregated over the FPUs that might only cover part5 of the river basins.

6 Runoff (qtot [kg/m2/s]) and discharge (dis [kg/m2/s]) are provided on a monthly or daily 7 resolution by all participating hydrological models. They are annually averaged before re-8 distribution as the irrigation water requirements (pirrww) estimated by the crop models are 9 also only reported on an annual basis.

The "irrigation water" required for full irrigation is usually provided by the crop models as the 10 overall amount of irrigation water applied to the harvested crops and reported for the 11 harvesting year. It might be partly used in the previous year if the planting day falls in that year. 12 13 Therefore we used a two year running mean of the "available water" (calculated as described 14 above) to estimated the available water relevant for the reported yields. For all crop models 15 reporting pirrww in the harvesting year the two year average is taken over the actual and the previous year. For pDSSAT we use the average over the actual and the following year as pirrww 16 17 is reported in the planting year.

18 Calculation of the crop specific amount of available irrigation water

To calculate the crop specific available irrigation water we have to take into account that not all of the "available water" can be used for irrigation and that part of the water spent for irrigation will be used for other crops not covered by the four main ISI-MIP crops. To account for both

22 effects we proceed in the following way:

The maximum available irrigation water for one specific high priority crop is assumed to be a fraction of the "available" water taking into account the area used for the specific crop and the amount of water needed for its full irrigation. The division is done according to the following assumptions:

1) 40% of the overall runoff can be used for irrigation or other purposes. The 40% availability constraint follows (13) and attempts to account for the various limitations that reduce utilization of water resources for irrigation: spatial and temporal disagreement between water requirements for crops and water availability in rivers, lakes, and aquifers as well as losses due to transport, storage or inefficient application.

2) To estimate the amount that is actually available for irrigation the amount of water used for
 manufactoring and electricity productions (amanww) as well as domestic withdrawal

1 (adomww) provided by H08 is subtracted from the "40% budget". Irrigation water withdrawals

2 from currently irrigated lands (airrww) are not subtracted as the starting point in this analysis is

3 the rainfed-only situation.

4 3) 60% of the remaining amount is assumed to be directly available to the plants¹. It is split up

between the high priority crops and the other crops planted within the considered FPU basedon LPJmL simulations:

7 Assuming full irrigation on current harvesting areas (12), crop specific pirrww data were 8 spatially aggregated within each FPU and averaged over the reference period (1980-2010). For the crops included in the MICRA data set but not covered by LPJmL simulations we used a) 9 10 median, b) maximum and c) minimum irrigation values calculated at each grid point over all crops and managed grass land simulated by LPJmL. In the case of the MIRCA data set the 11 12 simulated crops are wheat, rice, maize, soybeans, cassava, millet, peanuts, pulses, rapeseed, 13 sugar beet, sugar cane and sunflowers. Not covered crops are barley, rye, sorghum, potatoes, 14 oil palm, citrus, date palm, vine, cotton, cocoa, coffee, "other perennial" and "other annual 15 crops".

- 16 In the minimal and median irrigation setting we assume that managed grass land is not irrigated 17 at all (which means that more water is available for the four crops considered in this paper
- 18 following the reasoning of an optimistic scenario).

The LPJmL specific fraction of the aggregated pirrww data associated with the high priority crops is then assumed for all crop models providing simulations of the four high priority crops.

21 PEGASUS is excluded from this experiment as it does not provide rice data.

3) The split-up of irrigation water within the group of the high priority crops is determined in a
similar way as described above. However in the case we used crop model specific values based
on the aggregated pirrww data of the full irrigation runs (assuming MIRCA current land use) of
the high priority crops.

26 **3.2 Optimal Distribution of the "available water"**

Within the paper "optimal distribution of water" within one FPU means a distribution according to the highest water use efficiency defined at each grid cell as yield increase per applied irrigation water. Starting from the grid cell with the highest water use efficiency the water

- 30 assigned for irrigation of each crop is used to fully irrigate as much area of the respective crop
- 31 in the considered FPU as possible. If the water is not sufficient to irrigate the whole crop area in

¹ The irrigation water demand reported as "pirrww" by the crop models partly includes assumptions about the efficiency of irrigation, i.e. pirrww is higher than the water assumed to be directly available to the plants. Here, the associated pirrww values are harmonized assuming a homogeneous project efficiency of 60%.

- 1 a cell the irrigated area is reduced accordingly and the remaining area in that cell and in all cells
- 2 with lower water productivity is assumed to be rainfed.

3 Calculation of "water use efficiency"

Yields and the amount of irrigation water applied are usually reported for the harvesting year 4 5 except for pDSSAT where the information is written to the associated planting year. In most of 6 the cases there is a one to one relationship between yield increases by optimal irrigation and 7 the applied irrigation water as pirrww is reported as the overall amount of water required to 8 generate the "full irrigation yields" no matter if part of the growing season lies in the previous 9 calendar year. In these cases "water use efficiency" can be estimated on an annual basis by 10 subtracting the no-irrigation yields from the full irrigation yields and dividing the difference by the pirrww value of the same year. That is different for LPJmL where the reported irrigation 11 12 water is not directly associated with the generated yields but aggregated over the individual 13 calendar years. Therefore the water actually needed to generate the full irrigation yields might 14 be a mixture of the pirrww data of the harvesting year and the previous year. As a proxy for the "required irrigation water" we used the mean pirrww value of both years to calculate annual 15 16 data of water use efficiency in case of LPJmL.

- To account for the fact that irrigation measures have to be installed and will not change from year to year we use ten year averages (covering the actual and the 9 previous years) of the water use efficiency to distribute the water.
- 20 The scheme assumes that irrigation is always 'all or nothing' while realistically crop production
- could well be increased in water-limited FPUs by "deficit irrigation" in favor of larger irrigated
- areas, as the irrigation water use efficiency typically declines at higher irrigation levels (14).

3.3 Irrigation scheme for the LU change experiment

In this experiment we apply a land use pattern provided by the LPJmL + MAgPIE agro-economic simulation for the HadGEM2, RCP8.5 climate scenario and the SSP2 assumptions with respect to population and GDP. The agro-economic land use model MAgPIE is designed to provide demand fulfilling land use patterns in this case based on the biophysical input data delivered by LPJmL. The assumed demand of rice, wheat, maize and soy is identical to the demand data used in the main text.

- 30 While within MAgPIE it is assumed that there is technological progress increasing crop yields we
- 31 only use the land use information and do not assume changes in other management options.
- 32 The irrigations scheme follows the above procedure based on the optimal distribution of the
- 33 crop specific available irrigation water.

In this exercise we estimate the additional effect of land use changes and estimate the
 remaining adaptive pressure that might be met by other management options not covered by
 the ISI-MIP simulations.

The calculation of the crop specific amount of available irrigation water is analog to the procedure described for the MIRCA data set. However, for MAgPIE the simulated crops are cassava, managed grass, peanuts, maize, pulses, rapeseed, rice, soybeans, sugar beet, sugar cane, sunflowers, millet and wheat. The following crops are considered as "not simulated": bioenergy grass, bioenergy tree, cotton, oil palm, potatoes, and "others".

9

3.4 Calculation of global production under additional irrigation for crop models

11 that do not provide the full information to apply the above water distribution

12 scheme

PEGASUS and GAEZ-IMAGE provide full irrigation and no irrigation yields but not the complete information necessary to apply the above water distribution scheme. In order to not lose the available information about $P_{unlimited}$ (global production under unlimited irrigation) and P_R (rainfed global production) the production under water constraints ($P_{limited}$)_{ij} is calculated as ($P_{limited}$)_{ij} = - (P_R)_i + α_j * (($P_{unlimit}$)_i - (PR)_i), where α_j is the water model specific mean over all production fractions α_{kj} calculated for each crop model k where the full information is available.

19

20 3.5 Participating hydrological models

Model name	Energy balance	Evaporation scheme	Runoff scheme	Snow scheme	Vegetation dynamics	CO ₂ effect
DBH (15 <i>,</i>	Yes	Energy	Infiltration	Energy	No	Constant
16)		balance	excess	balance		
H08 (17,	Yes	Bulk formula	Saturation	Energy	No	No
18)			excess, non-	balance		
			linear			
JULES (19,	Yes	Penman-	Infiltration	Energy	Yes	Yes
20)		Monteith	excess,	balance		
			saturation			
			excess,			
			groundwater.			
LPJmL (8,	No	Priestley-	Saturation	Degree-day	Yes	yes
21)		Taylor	excess			

Mac- PDM.09 (22, 23)	No	Penman- Monteith	Saturation excess, non- linear	Degree-day	No	No
MATSIRO (24, 25)	Yes	Bulk formula	Infiltration excess, saturation excess, groundwater.	Energy balance	No	Constant
MPI-HM (26, 27)	No	Penman- Monteith	Saturation excess, non- linear	Degree-day	No	No
PCR- GLOBWB (28, 29)	No	Hamon	Saturation Excess Beta Function	Degree Day	No	No
VIC (30, 31)	Only for snow.	Penman- Monteith	Saturation excess, non- linear	Energy balance.	No	No
WaterGAP (32–34)	No	Priestley- Taylor	Beta function	Degree day	No	No
WBM (35, 36)	No	Hamon	Saturation Excess	Empirical temp and precip based formula	No	No

Table S3: Basic water model characteristics and references. In case of the evapotranspiration scheme "Bulk formula" means that bulk transfer coefficients are used when calculating the turbulent heat fluxes. In case of the runoff scheme "beta function" means that runoff is a nonlinear function of soil moisture.

4 Production Fractions

2 To assess whether the increase in production under "additional irrigation" is mainly 3 biophysically limited by potential yields under unlimited irrigation or by water availability we introduce the production fractions $\alpha_{ij} = ((P_{limited})_{ij} - (P_R)_i) / ((P_{unlimit})_i - (PR)_i)$, where $(P_{limited})_{ij}$ is the 4 potential production accounting for water constraints (scenario discussed above), $(P_R)_i$ denotes 5 rain-fed production, and (Punlimit) is the potential production under full irrigation for each GGCM 6 (i) and hydrological model (j). Cumulative distributions of α_{ij} for different levels of global 7 warming (see Fig. 3) show that the uncertainty budget of the production fractions α differs 8 9 from the one of projected production changes. The crop-model-induced spread of α (spread of 10 light blue squares in Fig. S8 below) is comparable or smaller than the spread introduced by the different hydrological models (spread of dark blue triangles in Fig. S8). The GGCM contribution 11 to the spread of α is caused by 1) differences in the absolute amount of water required to reach 12 full irrigation, 2) differences in the determination of the crop-specific fractions of the overall 13 14 amount of irrigation water (see SI), and 3) the production increase due to irrigation. The median α , based on the "additional irrigation" scenario, is about 60% in the case of wheat and about 15 80% for maize. In comparison, assuming unlimited irrigation on currently irrigated land the 16 median production fraction for wheat over all GGCMs already reaches 50% at 1°C, 50% at 2°C, 17 40% at 3°C, and 30% at 4°C. Thus the benefit of additional irrigation is relatively small. In the 18 case of maize the corresponding values are 30% at 1, 2, and 3°C and 40% at 4°C. Assuming 19 unlimited irrigation on currently irrigated land the median production fraction for rice over all 20 crop models already reaches about 75% at 1, 2, 3, and 4°C. For soy the value only reaches about 21 22 10% at all levels of global warming. In both cases median values under the "additional irrigation" scenario also reach about 80%. In the LU change scenario the production fraction 23 24 reaches about 90% for all crops. Thus, the projected change in crop production is close to the 25 maximum change assuming unlimited irrigation.

26 The distribution does not show a clear dependence on global warming. This could be due to 1)

the high level of aggregation (though runoff decreases in many regions, it increases in others,

Fig. 1 of (14)), 2) reduced water demand by reduced growing seasons lengths, and 3) decreasing

29 water demand under CO₂ fertilization (20).



3 Figure S7. Limits of irrigation according to hydrological models (listed in darkblue) and GGCMs (listed in light blue). The colored lines represent the cumulative probability distribution of the 4 5 production fraction α at different levels of global warming (green = 1°C, darkblue = 2°C, orange = 3°C, red = 4°C). While the inner lines are based on the "median" assumption regarding water 6 7 spent on crops not simulated by LPJmL (Section 3.1 of the SI), the outer lines surrounding the 8 grey shaded area represent results for the "min" and "max" setting at 2°C. All results are based 9 on the HadGEM2-ES, RCP8.5 climate input. Lightblue squares: GGCM specific means of α . Darkblue triangles: water model specific means of α . In case of wheat, rice, and soy the GGCM 10 induced spread is comparable to the spread introduced by the hydrological models while for 11 maize the GGCM induced spread is minor. Colored lines at the upper end of the plot represent 12 the inner 66% of the distribution at different levels of global warming. The median is indicated 13 14 by colored dots.

1 The production fraction increases non-linearly with the project efficiency. This is due to the 2 optimal distribution of the available water starting from the grid cells where it leads to the 3 highest yield increases per applied amount of water (see Figure S8 below).



Figure S8. Median production fractions for different project efficiencies (20%, 50%, 60%, 70%, and 100%). Color coding indicates different levels of global warming (green = 1°C, darkblue = 2°C, orange = 3°C, red = 4°C). At a project efficiency of 60% the plotted production fractions corresponds to the median of the cumulative distributions shown above (colored dot in the upper part of each panel).

- 4 F

1 5. LU changes

5.1 Assumed growth rates within MAgPIE

The annual growth rates due to assumed within MAgPIE to generate the land use patterns used in the paper are time dependent and vary from region to region. The table contains the annual average growth rates for the period 2005-2045 (considered within (37)) and the longer period until 2085 used in this paper. The growth rates are only relevant for the generation of the LU pattern. Crop production changes considered in this paper do not account for these growth rates.

Region	2005-2045	2005-2085
AFR (Sub-Saharan Africa)	4,8%	9,5%
CPA (Central Asia, mainly China)	1,4%	0,7%
EUR (Europe)	0,7%	0,4%
FSU (Former Soviet Union)	0,4%	0,3%
LAM (Latin America)	0,7%	0,9%
MEA (Middle East and North Africa)	3,0%	2,8%
NAM (North America)	1,1%	0,7%
PAO (Pacific OECD countries, Australia, New		
Zealand, Japan)	1,3%	1,2%
PAS (Pacific Asia)	1,1%	1,0%
SAS (South Asia)	2,2%	2,0%

Figure S4. Average annual growth rates of yields used for the generation of the LU patterns by
 MAgPIE.

5.2 Comparison of MIRCA present day LU patterns to MAgPIE LU future patterns











3 data set and the future (2085) physical land area where maize is grown (which is assumed to be

4 identical to the harvested area in our analysis) as described by the MAgPIE land use model.

5





3 data set and the future (2085) physical land area where rice is grown (which is assumed to be

4 identical to the harvested area in our analysis) as described by the MAgPIE land use model.

5





1 5.3 Loss of natural vegetation



Figure S13: Area of natural vegetation as provided MAgPIE for the reference year 1995.



Figure S14: Reduction of the area of natural vegetation as provided by MAGPIE. The map shows
the differences between the 2085 fractions of natural vegetation per grid cell and the 1995
fractions (shown in Figure S10 above). Blue areas indicate no changes with respect to the 1995

10 pattern and yellow to red areas mean reductions with respect to the 1995 reference pattern.

1 6 Participating bio-geochemical models

Model abbreviation	Representation of	Represented bio-geochemical cycles
	dynamic vegetation	
LPJmL (38, 39)	yes	Representation of a fully coupled
		water and carbon cycle (assuming
Lund-Potsdam-Jena		optimal leaf nitrogen allocation, but
managed Land Dynamic		no limitation of CO2 fertilization by
Global Vegetation and		nutrient supply)
Water Balance Model		
JULES (19, 20)	yes	Representation of C cycle (no
		limitation of CO2 fertilization by
Joint UK Land		nutrient supply e.g. N or P)
Environment Simulator		
JeDI (40)	yes	representation of water and C cycle,
		no limitation of CO2 fertilization by
Jena Diversity Model		nutrient supply
SDGVM (41, 42)	no	fully coupled water and carbon
		cycle, below
Sheffield Dynamic		ground nitrogen cycle
Vegetation Model		
VISIT (43, 44)	no	Representation of C and N cycle (but
		no limitation of CO2 fertilization by
Vegetation Integrative		N supply in this simulation).
Simulation for Trace gases		
		For vegetation processes, single
		vegetation-layer carbon cycle model.
Hybrid (45)	yes	Representation of C and N cycles (N
		provides constraints on
		photosynthesis, growth, and affects
		allocation of C to leaf area)
ORCHIDEE (46, 47)	Not in the configuration	Representation of C cycle (no
	used for ISI-MIP	limitation of CO2 fertilization by
		nutrient supply e.g. N or P)
		Land surface model – calculates
		energy fluxes and surface
		temperature on a 30 min time step

- 1 Table S5: Basic characteristics and references for the bio-geochemical models participating in
- 2 ISI-MIP and used in this paper to calculate the loss of carbon sinks, reduction of the vegetation
- 3 carbon stock and the area under risk of severe ecosystem changes.

5 7 Changes in Ecosystem Atmosphere C Fluxes under the fixed 1995 LU pattern



Figure S15: Changes in carbon sinks (ecosystem-atmosphere C flux) integrated over the fixed
1995 area of natural vegetation representing the reference situation of the MAgPIE LU
patterns. Dashed lines indicate Biomes models not allowing for dynamical vegetation changes.
Simulations are based on HadGEM2-ES, hist+ RCP8.5 climate input. Dashed vertical lines
indicate the year where the global mean warming with respect to the reference period 19802010 reaches the 1, 2, 3, and 4°C level.



1 8 Loss of carbon sinks and stocks under constant CO2 conditions

2

Figure S16: Difference in (a) ecosystem-atmosphere C fluxes (b) the vegetation carbon stock 3 4 between the scenario where the area of natural vegetation is reduced according to the MAgPIE projections and the reference scenario where the area of natural vegetation is assumed to 5 6 remain constant (1995 pattern). For reference panel (c) shows the Ecosystem-Atmosphere C fluxes for the fixed 1995 pattern of natural vegetation. All results are based on the simulations 7 8 under fixed present day CO2 concentrations. The Figure is analogous to Panel a and b of Figure 9 4 of the main text and based on HadGEM2-ES, hist+ RCP8.5 climate input. Dashed vertical lines 10 indicate the year where the global mean warming with respect to the reference period 1980-11 2010 reaches the 1, 2, 3, and 4°C level.

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9 References for the applied climate models

Institute	Abbreviation	used model version
Met Office Hadley Centre (additional HadGEM2-ES realizations contributed by Instituto Nacional de Pesquisas Espaciais)	MOHC (additional realizations by INPE)	HadGEM2-ES
Institut Pierre-Simon Laplace	IPSL	IPSL-CM5A-LR
Japan Agency for Marine- Earth Science and Technology, Atmosphere and Ocean Research Institute (The University of Tokyo), and National Institute for Environmental Studies	MIROC	MIROC-ESM-CHEM

Table S6: All impact simulations are based on the climate projections based on the above 3 models. Daily data as provided via the CMIP5 archive were bias corrected and provided to all modeling groups participating in ISI-MIP. All simulations are based on the available simulations provided for the historical period and the Representative Concentrations Pathways (RCP2.6, RCP4.5, RCP6.0, and RCP8.5 (48)).

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