Impacts of 1.5 versus 2.0°C on cereal yields in the West African Sudan Savanna

SI Materials

Table S1. Annual absolute change (average absolute change ± standard deviation) of daily maximum and minimum temperature (°C) and annual precipitation sum (mm) for two warming scenarios (1.5°C, 2°C) and three general circulation models (GCM) relative to the baseline (2006-2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.

Variable	Scenario	GCM					
		ECHAM6	MIROC5	NorESM1			
T _{max}	1.5°C	1.0 ± 0.3	1.0 ± 0.1	0.8 ± 0.2			
	2.0°C	1.4 ± 0.2	1.4 ± 0.1	1.3 ±0.2			
T _{min}	1.5°C	1.1 ± 0.1	0.9 ± 0.1	0.8 ± 0.1			
	2.0°C	1.7 ± 0.1	1.2 ± 0.1	1.4 ± 0.1			
Precipitation	1.5°C	-79 ± 65	-5 ± 30	-11 ± 53			
	2.0°C	-90 ± 55	-2 ± 42	6 ± 48			

Table S2: Absolute change (average absolute change \pm standard deviation) of daily maximum and minimum temperature (°C) and precipitation sum (mm) during the growing season for West Africa (assumed here as June to end of September) for two warming scenarios (1.5°C, 2°C) and three general circulation models relative to the baseline (2006-2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.

Variable	Scenario	GCM					
		ECHAM6	MIROC5	NorESM1			
T _{max}	1.5°C	0.9 ± 0.2	1.2 ± 0.1	0.8 ± 0.2			
	2.0°C	1.5 ± 0.2	1.5 ± 0.1	1.3 ±0.2			
T _{min}	1.5°C	0.9 ± 0.1	0.9 ± 0.1	0.7 ± 0.0			
	2.0°C	1.5 ± 0.1	1.2 ± 0.1	1.2 ± 0.1			
Precipitation	1.5°C	-35 ± 32	8 ± 24	4 ± 41			
	2.0°C	-28 ± 32	12 ± 31	21 ± 44			



Figure S1. Average absolute changes (Delta Tmax) in average daily maximum temperature (T_{max} , °C) for two warming scenarios (1.5°C, 2°C) and three general circulation models (GCM) relative to baseline climate (2006 to 2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.



Figure S2. Average absolute changes (Delta Tmin) in average daily minimum temperature (T_{min} , °C) for two warming scenarios (1.5°C, 2°C) and three general circulation models (GCM) relative to baseline conditions (2006 to 2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.



Figure S3. Average absolute changes (Delta Rain) in annual precipitation sum (P, mm) for two warming scenarios (1.5°C, 2°C) and three general circulation models (GCM) relative to baseline climate (2006 to 2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.



Figure S4. Average absolute changes (Delta Radiation) in annual solar radiation sum (R, MJ m²) for two warming scenarios (1.5°C, 2°C) and three general circulation models (GCM) relative to baseline climate (2006 to 2015). Summary statistics were calculated over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.



Figure S5. Variation of soil total available water (TAW, mm) for each simulation unit.



Figure S6. Difference between DSSAT and SIMPLACE simulated impact of 1.5°C (blue) and 2.0°C (red) of warming on maize, millet and sorghum yields for the West African Sudan Savanna region relative to the current baseline period (2006 – 2015). Impacts are shown for current fertilizer levels (top row) and for fully fertilized case (Intensified, bottom row) systems. Uncertainty captured in the figure covers three GCMs and three sowing windows. All simulations shown include the effects of elevated atmospheric [CO₂].



Figure S7. Simulated SIMPLACE-model impact of 1.5°C (blue) and 2.0°C (red) of warming on maize, millet and sorghum yields for the West African Sudan Savanna region relative to the current baseline period (2006 – 2015). Impacts are shown for systems with current ambient (amb) and scenario projected atmospheric CO₂ concentrations (elev) for current fertilizer levels (top row) and for fully fertilized case (Intensified, bottom row). Uncertainty captured in the figure covers three GCMs and three sowing dates. Simulations were performed with the SIMPLACE model.



Figure S8. Effect of intensification on simulated coefficient of variation (CV) for baseline (grey, 2006 – 2015), 1.5°C (blue) and 2.0°C (red) of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region. Impacts are shown for both systems with current fertilizer levels and for fully fertilized case (Intensified), considering $[CO_2]$ fertilization effects. Uncertainty captured in each boxplot is due to the three sowing windows, three GCMs and two crop models.



Figure S9. Effect of crop model on simulated coefficient of variation (CV) for baseline (grey, 2006 – 2015), 1.5°C (blue) and 2.0°C (red) of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region. Impacts are shown for DSSAT and SIMPLACE for both systems with current fertilizer levels (top row) and for fully fertilized case (Intensified, bottom row), considering $[CO_2]$ fertilization effects. Uncertainty captured in each boxplot is due to the three sowing windows and three GCMs.



Figure S10. Effect of ambient versus elevated CO₂ (with SIMPLACE-model) on simulated coefficient of variation (CV) for baseline (grey, 2006 – 2015), 1.5°C (blue) and 2.0°C (red) of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region. Impacts are shown for current fertilizer use (top row) and intensified, non-limiting fertilizer use (bottom row) for both without $[CO_2]$ fertilization effects (amb) and with consideration of $[CO_2]$ fertilization effects (elev). Uncertainty captured in the figure covers 20 instances of each of three GCMs.



Figure S11. Comparing total yield changes (dark green) with the change in growing season duration (green) for 1.5°C and 2.0°C of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region across all years (top row) and in years with yields in the lowest decile (bottom row) relative to baseline climate (2006 to 2015) for current fertilizer case. Relative changes were determined over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario. Results are shown for both crop models in adjacent panels for (a) maize, (b) millet and (c) sorghum. All simulations considered elevated [CO₂] and are averaged across sowing dates and GCMs.



Figure S12. Comparing total yield changes (dark green) with the change in growing season duration (green) for 1.5°C and 2.0°C of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region across all years (top row) and in years with yields in the lowest decile (bottom row) relative to the baseline climate (2006 to 2006) for intensified, non-limiting fertilizer case. Relative changes were determined over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario. Results are shown for both crop models in adjacent panels for (a) maize, (b) millet and (c) sorghum. All simulations considered elevated [CO₂] and are averaged across sowing dates and GCMs.



Figure S13. Simulated relative change in duration of the growing cycle for 1.5°C and 2.0°C of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region relative to the baseline climate (2006-2015). Change in cycle duration is shown early (green), medium (blue) and late (purple) sowing windows. The cycle is independent of nutrient or [CO₂] level. Uncertainty captured in the figure covers the three GCMs. Relative changes were determined over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.



Figure S14. Drivers of yield change for 1.5°C and 2.0°C of warming scenarios on maize, millet and sorghum yields for the West African Sudan Savanna region with current fertilizer use (top row) and intensified, non-limiting fertilizer case (bottom row). The total yield change and change in cycle duration are relative to values in baseline climate (2006 to 2015). Change due to $[CO_2]$ is in the climate of the scenario but relative to the case of no $[CO_2]$ fertilization effects. The heat losses are determined as the difference between yield changes with heat-waternutrient limitation and yield changes with water-nutrient limitation in a given warming scenario. Simulations are for the SIMPLACE model and considered elevated $[CO_2]$, except to determine the CO_2 effects which considered simulations both with and without $[CO_2]$ fertilization. Relative changes were determined over 200 years (e.g. pooling the 20 sets of 10-year climate series) for each GCM and scenario.





Crop model description

SIMPLACE is a framework that enables linking crop models that represent different aspects and processes in agricultural systems for a variety of scientific purposes (Gaiser et al., 2013). A combined model solution was SIMPLACE <Lintul5, SLIMWater, FAO ET, CanopyT, Heat Stress>, referred to as SIMPLACE< LINTUL5+>. LINTUL5 is a generic model, which can be used for different annual crop types growing under a large range of soil and weather conditions (Wolf, 2012). It simulates growth as a function of intercepted radiation and radiation use efficiency (RUE), which in turn is a function of daily mean temperature, water or nutrient limitation and atmospheric CO₂ concentration. Crop development is a function of daily accumulated thermal time above a base temperature and crop specific thermal time requirements from emergence to anthesis (TSUM1) and from anthesis to maturity (TSUM2). Slimwater is used to simulate crop water uptake. Crop water demand is calculated according the FAO Penman-Monteith method (Allen et al., 1998) with a reference crop and considering the dual crop coefficient method. Canopy temperature is estimated based on a solution of an hourly energy balance at the crop surface, correcting for atmospheric stability conditions using the Monin-Obukhov Similarity Theory (MOST) (Webber et al., 2016). A heat stress module (Gabaldón-Leal et al., 2016) reduces grain yield when the hourly temperature is above a critical threshold temperature around anthesis to simulate failure of flowering and grain abortion.

DSSAT is a process based, mechanistic and crop management oriented model comprising a suite of modules (Jones *et al.*, 2003). It utilizes information from soil profile, crop management, crop genetic coefficients, daily weather (maximum and minimum temperature, rainfall and solar radiation) to simulate soil and crop processes, thus predicting crop yield and other plant specific outputs. Optimal plant growth is a function of photosynthetic capacity, radiation capture, thermal time and photoperiod sensitivity whereas actual growth and development are constrained by nutrient and water stress as well as sub-optimal temperatures (Soler *et al.*, 2007). Growth is constrained by nutrient and water sub-modules through stress factors. Plant nitrogen availability is guided via fertilizer input and mineralization of soil organic carbon. The Century model (Parton *et al.*, 1988) embedded in DSSAT (Porter *et al.*, 2010) was used in simulating soil organic matter mineralization. The Ritchie (1998) cascading water balance approach describes movement of water between soil layers. Daily water balance is a function of precipitation, irrigation, transpiration, evaporation, drainage and runoff. DSSAT has been applied in the West African Sudan Savanna in a number of studies (Naab *et al.*, 2015; Akinseye *et al.*, 2017; Parkes *et al.*, 2017; Amouzou *et al.*, 2018).

Fertilizer rates

No fertilizer was applied for millet and sorghum while fertilization for maize was based on aggregated NPK rates for each sub-region. With the assuming that a compound NPK 15-15-15 is typically used in all sub-regions, the amount of N application was derived. The total amount of N was applied at the beginning of the growing season equal to 12 kg N ha⁻¹ in Ghana-North, 15 kg N ha⁻¹ in Benin-North, Burkina-Northeast, and Burkina South and 14 kg N ha⁻¹ in Mali-South. Other nutrients were assumed non-limiting. In the non-limiting fertilizer intensification case, the crop models were run without nutrient limitation.

Calibration procedure

All cultivars were already calibrated in DSSAT and published or under process for publication (Table S3). The same datasets were used for calibration in SIMPLACE as described in the following section. Tables S4-S6 give an overview of the experiments used for maize, millet and sorghum calibration, respectively.

Table S3. Calibrated cultivars in DSSAT by crop and by region and reference sources to the experiment description/model validation results.

	Benin-north	Burkina-south	Burkina-centre	Mali-south	Ghana-
					north
Sorghum	Kadaga	CSM335	CSM63E	CSM335	Kadaga
	(Naab 2016,	(Clerget et al	(Clerget et al	(Clerget et al	(Naab 2016,
	unpublished	unpublished)	unpublished)	unpublished)	unpublished
	data)				data)
Maize	EVDT-97 STR	Obatanpa (Naab 2004)	Obatanpa (Naab	Obatanpa	Obatanpa
	(Naab 2016,		2004)	(Naab 2004)	(Naab 2004)
	unpublished				
_	data)				
Millet	Bolga-local	CIVT	CIVT	CIVT	Bolga-local
	(Naab 2016,	(Akponikpè <i>et al.,</i>	(Akponikpè <i>et al.,</i>	(Akponikpè <i>et</i>	(Naab 2016,
	unpublished	2010)	2010)	al., 2010)	unpublished
	data)				data)

TableS4. Datasets used in SIMPLACE for maize calibration

Description	Cultivar	Site	Year	Fertilizer application:	Planting day	Crop cycle
				treatment (bold) and	of year	length in
				rate (not bold)	(DOY)	days
Data collected in North Ghana (Wa,	Obatanpa	Wa/Ghana	2004	1: NO PO	169	100-105
UWR) with 9 treatments				2: NO P60		
				3: N0 P90		
				4: N60 P0		
				5: N60 P60		
				6: N60 P90		
				7: N120 P0		
				8: N120 P60		
				9: N120 P90		
Data collected in Dassari (North	EVDT97-	Dassari /Benin	2015	192 : 30N13P26K	177	90-95
Benin) with two different NPK	SPR			227 : 30N		
fertilizer applications						

TableS5. Datasets used in SIMPLACE for millet calibration

Description	Cultivar	Site	Year	Fertilizer application	Planting day	Crop cycle
				DOY (bold) and rate	of year (DOY)	length in
				(not bold)		days
Data collected in Sadoré (Niger)	CIVT	Sadoré	2005	Sowing 1:	Sowing 1: 109	90-95
for four different sowing dates				109 : 10N 12P 9K	Sowing 2: 138	
				139 : 25N	Sowing 3: 169	
				160 : 25N	Sowing 4:200	
				Sowing 2:		
				138 : 10N 12P 9K		
				169 : 25N		
				189 : 25N		
				Sowing 3:		
				170 : 10N12P9K		
				200 : 25N		
				221 : 25N		
				Sowing 4:		
				200 : 10N12P9K		
				231 : 25N		
				252 : 25N		
Data collected in Vea (North	Bolga_local	Northern	2015	without fertilizer	178	90
Ghana) without fertilizer		Ghana		application		
application						
Data collected in Vea (North	Bolga_local	Northern	2016	188 : 20N	171	90
Ghana) with 20N fertilizer		Ghana				
application						

TableS6. Datasets used in SIMPLACE for sorghum calibration

Description	Cultivar	Site	Year	Fertilizer application	Planting day	Crop
				DOY (bold) and rate	of year	cycle
				(not bold)	(DOY)	length in
						days
Data collected in Vea (North Ghana) with fertilizer	Kadaga	North	2015	209 : 20N	178	90
application (N) in two different dates		Ghana		223 : 20N		
Data collected in Vea (North Ghana) with fertilizer	Kadaga	Nord	2016	188 : 20N	171	90
application (N) in two different date		Ghana		202 : 20N		
Data collected in Samanko (North Mali) with two	CSM335	Samako	2010	Sowing 1:	Sowing1-165	125
NPK fertilizer application levels for each of the				157 : 18N 20P	Sowing2-190	
three different sowing dates				203 : 23N	Sowing3-217	
				Sowing 2:		
				185 : 18N		
				233 : 23N		
				Sowing 3:		
				211 : 18N		
				245 : 23N		
Data collected in Samanko (North Mali) with two	CSM63E	Samako	2010	Sowing 1:	Sowing1-165	100
NPK fertilizer application levels for each of the				157 : 18N 20P	Sowing2-190	
three different sowing dates				203 : 23N	Sowing3-217	
				Sowing 2:		
				185 : 18N		
				233 : 23N		
				Sowing 3:		
				211 : 18N		
				245 : 23N		

SIMPLACE calibration results

The calibration aims to ensure that the model could reproduce the phenology observations and the yields. The phenology for all crops was accurately simulated on average (Table S7).

	А	nthesis	Maturity		
	Observed	Simulated Observed		Simulated	
Maize					
EVDT97	236	236	273	273	
Obatanpa	224	224	269	270	
Millet					
CIVT	222	220	250	249	
Bolga-local	261	261	279	280	
Sorghum					
Kadaga	247	247	266	266	
CSM335	279	274	307	304	
CSM63E	253	252	282	280	

 Table S7: Observed and simulated by SIMPLACE phenology data for three crops



Figure S16. Simulated by SIMPALCE versus observed biomass (bottom row) and grain yield (top row) for three sorghum cultivars (CSM335, CSM63E and Kadaga). The black line is the 1:1 line added for improving visualization of the goodness of fit, RMSE – root mean square error. Note, 1 g m^{-2} is equivalent 10 kg ha⁻¹.



Figure S17. Simulated by SIMPLACE versus observed biomass (bottom row) and grain (top row) yield for two millet cultivars (CIVT and Bolga local). The black line is the 1:1 line added for improving for visualization of the goodness of fit, RMSE – root mean square error. Note, 1 g m⁻² is equivalent 10 kg ha⁻¹.



Figure S18. Simulated by SIMPLACE versus observed biomass (bottom) and grain (top) yield for maize cultivar (Obatanpa). The black line is the 1:1 line added for improving for visualization of the goodness of fit, RMSE – root mean square error. Note, 1 g m^{-2} is equivalent 10 kg ha⁻¹.

Crop	Region	Early s	owing	Medium	n sowing	Late s	owing
		startdate	enddate	startdate	enddate	startdate	enddate
Sorghum	Benin-north	01.06	15.06	16.06	30.06	01.07	15.07
	Burkina-south	01.06	15.06	16.06	30.06	01.07	10.07
	Burkina-centre	10.06	20.06	21.06	30.06	01.07	10.07
	Ghana-north	10.06	20.06	21.06	05.07	06.07	15.07
	Mali-south	10.06	20.06	21.06	30.06	01.07	10.07
Maize	Benin-north	01.06	15.06	16.06	30.06	01.07	15.07
	Burkina-south	10.06	20.06	21.06	30.06	01.07	10.07
	Burkina-centre	10.06	20.06	21.06	30.06	01.07	10.07
	Ghana-north	01.06	15.06	16.06	30.06	01.07	01.07
	Mali-south	10.06	20.06	21.06	30.06	01.07	10.07
Millet	Benin-north	01.06	15.06	16.06	30.06	01.07	30.06
	Burkina-south	10.06	20.06	21.06	30.06	01.07	10.07
	Burkina-centre	10.06	20.06	21.06	30.06	01.07	10.07
	Ghana-north	01.06	15.06	16.06	30.06	01.07	01.07
	Mali-south	10.06	20.06	21.06	30.06	01.07	10.07

Table S8. Sowing windows by crop and by region

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