

Sensitivity of the sorghum yield to individual changes in climate parameters: modelling based approach

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Abstract

Based on sensitivity analyses the effect of changing in climate on sorghum has been investigated. This has been achieved by conducting crop modeling experiments carried out with weather observations and output of global climate models. As can be anticipated results demonstrated that the sorghum yield is more sensitive to rainfall, solar radiation and CO₂ concentration. However, the increase in CO₂ concentration has a positive effect on yield in years that the crop is submitted to water stress. This effect is extremely reduced under well watered conditions and therefore is subject to the climate inter-annual variability. Accumulated solar radiation along the sorghum cycle below 1900 MJ m⁻² has also been shown to reduce the yield. Compared to other weather variables the sorghum yield is less sensitive to changes in relative humidity. In this sense, changes in stomatal closure and therefore photosynthesis is not highly dependent on water vapor pressure. These finds can be applied for both climate observations and global models output.

Key words: climate change, crop models, sensitivity analyses, CO₂ concentration, productivity.

1. INTRODUCTION

Water shortage is a limiting factor to profitable crop productivity in particular for maize, common beans and other subsistence crops (Teixeira & Bassoi, 2009). The productivity of sorghum, and of other crops may be altered as a result of global climate change. However, Conley et al. (2001) argued that increasing concentration of carbon dioxide (CO₂) in rainfed conditions, can increase the sorghum productivity in the future due to increased water use efficiency (WUE).

To analyze the crop response to future climate change scenarios, outputs of global circulation models (GCM) have been utilized as input data to crop models (Tubiello et al., 2000). Moreover, the use of calibrated and validated crop model can replace experiments traditionally performed in the field (Matthews et al., 2000). According to Saltelli et al. (2004), sensitivity analysis is the study of how the variation in the output of a model can be attributed qualitatively or quantitatively to variations of a model parameters or input data.

In semi-arid regions current weather conditions are in general adverse to crop production. Family farmers rely on rainfed agriculture and are therefore affected by changes in

climate/weather, and future climate changes can substantially reduce the productivity (Tubiello et al., 2000). Srivastava et al. (2010) indicated a decrease in productivity in India in two of the three regions studied. Chipanshi et al. (2003) has also found reduction of the sorghum productivity. However, Carbone et al. (2003) found an increase in productivity by up to 40% in the southeast region of the USA, by considering the CO₂ fertilization effect, adaptation strategies and shifting cultivars. Costa et al. (2009) argued that adverse future climate conditions on crops can be attenuated by the highly efficient CO₂ fertilization effect and suitable crop management.

It should be noted that few studies have been conducted for the tropical semi-arid regions focusing on modeling the sorghum productivity under current and future climate conditions (e.g. Grossi et al., 2013). Therefore, the present study aims to evaluating the productivity of sorghum, considering current and future climate conditions based on observations and on Global Climate Models (GCMs) output.

2. MATERIALS AND METHODS

Experimental setup and modeling design

Modeling experiments focusing on the productivity of sorghum were performed taking into account the conditions of soil and climate of the municipality of Janaúba (Lat. 15, 80°S, Long. 43.30°W, Elevation 516m). This city is located in the northern part of Minas Gerais State, Brazil, and is the second largest city of northern Minas Gerais. The annually averaged total rainfall is 774 mm based on climatology (1982-2009) obtained from the Brazilian Institute of Meteorology (INMET, Figure 1a). The highest average daily solar radiation of 24 MJm⁻²day⁻¹ occurs in February and the lowest of 17 MJm⁻²day⁻¹ is registered in June. Records of monthly average air temperatures (Figure 1b), show that in Janaúba maximum temperatures varies between 29° C in July and 33° C in October.

The modeling experiments have been conducted by using soil conditions as proposed by the Embrapa Maize and Sorghum experimental station located nearby Janaúba, which is classified as Haplustox. Some of its physical and chemical characteristics are presented in table 1.

In DSSAT the soil is represented by a one-dimensional profile, which is horizontally homogeneous, and has several vertical layers (Jones et al., 2003). It is considered a depth of 1.90m with the most roots concentrated in a depth of 1.40 as observed in the field. The crop root distribution is represented by the root growth factor that varies from 0 to 1. The root growth factor values are stored in the soil profile file. For the present study the depth of the sorghum rooting system was set to 1.40m. Details of the experimental design and field trials can be found in Grossi et al. (2013).

Description of the CSM-CERES-Sorghum model

The models of the CERES (Crop Estimation through Resource and Environment Synthesis) are employed to estimate the productivity of grains such as maize, rice, wheat, barley, millet and sorghum. The CSM-CERES-Sorghum is part of the modular system called Crop System Model (CSM) present in the Decision Support System for Agrotechnology Transfer (DSSAT) version 4.5 (Jones et al., 2003). It is based on the model CERES-Sorghum (Alagarwamy et al., 1988) that describes the production and partitioning of the plant biomass and its phenology. The model interconnects the processes of environmental variations and plant management (Carbone et al., 2003). Indeed, crop modeling allows for investigating the impact of individual weather variables on the growth and development of the plant, by performing sensitivity analysis.

Climate Timeseries

The DSSAT model has additional modules to include soil characteristics (e.g. carbon, nitrogen, water) and environmental conditions (e.g. weather, soil-plant-atmosphere). In the present study, the Penman-Monteith method (FAO-56) is used to calculate the potential evapotranspiration. Thus, meteorological variables used as daily input data are precipitation

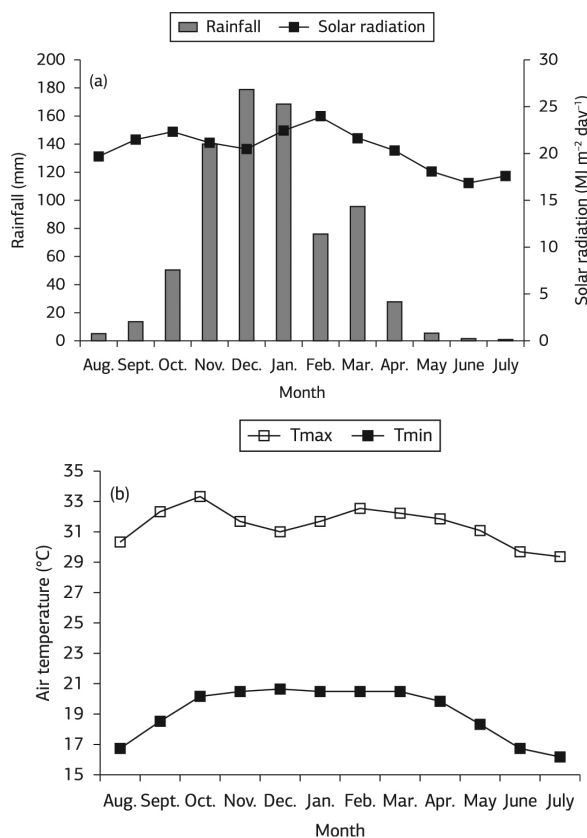


Figure 1. Monthly climatology of rainfall and maximum and minimum temperatures based on INMET dataset.

Table 1. Soil properties for the experiments conducted in Janaúba and Sete Lagoas, MG, Brazil. Average values of soil profiles. This table is based on Grossi et al. (2013)

Physical properties	Unit	Janaúba
Sand	(%)	42.43
Silt	(%)	22.71
Clay	(%)	34.86
Bulk density	(kg 10 ³ m ⁻³)	1.54
Upper limit, drained	(m ³ m ⁻³)	0.25
Lower limit	(m ³ m ⁻³)	0.12
Upper limit, saturated	(m ³ m ⁻³)	0.37
Chemical properties		Janaúba
pH in water	-	7.06
Organic matter	(dag Kg ⁻¹)	0.65
Nitrate	(mg Kg ⁻¹)	-
Ammonium	(mg Kg ⁻¹)	-
Total nitrogen	(%)	0.07
Textural Class		Medium

(mm day⁻¹), solar radiation (MJ m² day⁻¹), maximum and minimum air temperature (°C), relative humidity (%) and wind speed (m s⁻¹). For the INMET data, the values of global solar radiation were estimated from the number of hours of sunshine and solar radiation at the top of the atmosphere using the coefficients $a = 0.25$ and $b = 0.50$ recommended by Allen et al. (1998) in the Angstrom equation.

In addition to observed data from INMET, three databases obtained through global circulation climate models, namely ECHAM5/MPI-OM, CCCma CGCM3.1 (T63) and GFDL-CM2.1 have been used primarily to simulate the effect of future climate conditions on sorghum yield. The coupled model ECHAM5/MPI-OM used in this study has a spatial resolution of 1.9° latitude and longitude, with 31 vertical pressure levels. The Coupled Global Climate Model (CGCM) 3.1 is the third version generated by the Canadian Centre for Climate Modelling and Analysis (CCCma). It has a surface grid with spatial resolution of approximately 2.8° × 2.8° latitude and longitude and 31 vertical levels.

The GFDL coupled model run under a resolution of 2° latitude and 2.5° longitude and 24 vertical levels. Detailed information on the global circulation climate models were described elsewhere (Jungclaus et al., 2006; Flato & Hibler, 1992; Flato & Boer, 2001; Kim et al., 2002, 2003; Delworth et al., 2006).

Sorghum grain yields, on a dry matter basis, have been calculated for the period of 1982 to 1999 by using the INMET's weather time series. This approach is here in identified as present day (PD). The impact of climate change in the crop yields has been computed for the 2047 to 2064 interval, here in identified as Greenhouse Warming experiment (GW). For PD, the CO₂ forcing was based on concentrations of greenhouse gases and anthropogenic aerosols observed at Mauna Loa (Hawaii, USA).

For GW, the CO₂ concentrations followed the SRES A1B scenario (Special Report on Emissions Scenarios) described by Nakicenovic et al. (2000). CO₂ concentrations were used as input data to the model CSM-CERES-Sorghum, so that the sorghum crop response to CO₂ fertilization could

be simulated. In this simulation the CO₂ concentration is between 420 and 500 ppm.

Crop genetic coefficients and seasonal analyses

The calibration was performed from the average data collected in two field experiments conducted in Sete Lagoas (Minas Gerais). In table 2 are presented the averages of the repetitions of field experiments used to guide the process of calibration coefficients, as well as the standard deviation of the repetitions. The values of the variables related to the crop phenology are presented in days after sowing (DAS), while growth and yield variables refer to the dry weight (0% humidity) in the final harvest.

Table 2 are the values of percentage of deviance (PD), as well as the Willmott agreement index (d) and FSM. Positive values of PD indicate overestimate negative values indicate while under-estimating the model in relation to the observed data.

It is possible to notice that there was a good fit of the data. The model was able to simulate the growth and development of BRS 310 variety of exact and precise way. This can be proven through the high values of the coefficients d and FSM, equal to 0.99, in both plantations.

DSSAT crop models require parameterization of the genetic coefficients, which are specific for each cultivar to properly describe the processes related to growth, development and grain production. The coefficients used in this study have been previously adjusted and evaluated for the cultivar BRS 310, by using data from field trials carried out in 2 cities namely Sete Lagoas and Rio Verde in the central part of Brazil (Goiás State). These coefficients have been presented by Grossi et al. (2013) and for brevity they are not shown here.

The management practices considered in the simulations consisted of row spacing of 0.70 m and 18 plants m⁻². Nitrogen rates were 32 kg ha⁻¹ of nitrogen (N) per at sowing and 60 kg ha⁻¹ and 45 kg ha⁻¹, as urea, side-dressed 30 and 45 days after sowing, respectively. The most appropriated

Table 2. Observed and simulated development data used for calibration of cultivar -specific coefficients of the hybrid BRS 310

Experiment	Stage	Observed		Simulated	PDC (%)	dd	MEFe
		S.D. ^b	Mean				
SL0902	Emergence (DAS ^a)	0.0	7	6	-14.3	0.99	0.99
	Panicle initiation (DAS)	0.0	36	37	2.8		
	Anthesis (DAS)	1.2	71	70	-1.4		
	Physiological maturity (DAS)	1.9	105	109	3.8		
	Grain yield (kg ha ⁻¹)	406.1	6085	6093	0.1		
	Aboveground dry biomass at maturity (kg ha ⁻¹)	2309.8	16886	15947	-5.6		
SL1001	Emergence (DAS)	0.0	4	6	50.0	0.99	0.99
	Panicle initiation (DAS)	0.0	32	32	0.0		
	Anthesis (DAS)	1.0	67	64	-4.5		
	Physiological maturity (DAS)	1.5	113	112	-0.9		

^aDays After Sowing. ^bStandart Deviation (variable unit). ^cPercentage deviation (%). ^dAgreement index Willmott. ^eModelling Efficiency.

date for sowing sorghum in Janaúba is previously determined by Grossi et al. (2013), and are used in the present study (Table 2). Therefore, sensitivity analyses were performed solely for that maximum yield date which varies depending on climate observation and the GCMs datasets.

Sensitivity analysis

Sensitivity analyses were conducted to evaluate the CSM-CERES-Sorghum response to variation in meteorological elements and in distinct concentrations of CO₂. The analyses considering observed data were performed in four years that have shown different yield levels: (1) a year with low productivity (2) two years with intermediate productivity and, (3) a year with higher productivity.

To determine the interval of changes for each variable in the sensitive analyses, its maximum and minimum values within the four selected years have been taken. Thus, for relative humidity and solar radiation it was considered a reduction and increment of 20% in the daily values. For precipitation, the variation was $\pm 40\%$, and for the wind speed $\pm 50\%$. Exception to this applies for the CO₂ and temperature. For the former, we adopted a variation of $\pm 50\%$ with initial value of 380 ppm aiming to represent both concentrations observed in the past and those that are likely to be present in the future. It should be stressed that changes in CO₂ maintaining the temperature unchanged is questionable because in nature they change in phase. Nevertheless, this kind of experiment allows for the investigation of the individual fertilization effect on sorghum yield (Table 3).

To determine the interval of the climatic variables admitted that the distributions of variables per cycle are approximately normal. It is known that for a normal distribution, 99.73% of the occurrences are within 3 standard deviations from the average. Thus, the percentages of variation were determined

Table 3. Range of Study of variables

	Range of study			
	Rainfall (mm cycle ⁻¹)	Solar radiation (MJ m ⁻² cycle ⁻¹)	Relative Humidity (%)	Wind Speed (m s ⁻¹)
Variation	40%	20%	20%	50%
Minimum Value	245.82	1489.68	50.008	0.36
Maximum Value	1238.16	2740.92	88.452	1.65

Table 4. Meteorological variables values during the sorghum cycle based on INMET dataset

Yield (kg ha ⁻¹)	Year	Cycle length (day)	Accum. Precipitation (mm cycle ⁻¹)	Tmax (°C)	Tmin (°C)	Accum. Solar radiation (MJ m ⁻² ciclo ⁻¹)	Mean Wind speed (m s ⁻¹)	Mean relative humidity (%)
1758	2007	101	431	32	21	2251	0.7	63
3635	1993	99	410	32	22	1862	1.1	70
4679	2008	104	526	31	21	2182	0.9	67
6353	2001	105	955	31	21	2284	0.9	74

in such a way that the meteorological variables were studied in this range.

Sensitivity experiments including changes in temperature are implemented by modifications in the daily thermal amplitude, but keeping the daily mean temperature constant. It should be noted that the analyses must be done in simulations with the same cycle duration. Variations in productivity caused by changes in the input variables were calculated as follows:

$$Var = \left(\frac{Pr - Pr_i}{Pr_i} \right) \times 100 \quad (1)$$

Where

Var: is the variation in grain yield caused by the variation of daily rainfall, solar radiation, wind speed, relative humidity or CO₂, in%;

Pr_i: is the simulated yield obtained by using observed data in kg ha⁻¹.

Pr is the yield due to changes in the input variables, kg ha⁻¹.

3. RESULTS AND DISCUSSION

Variability of the simulated yields

The four values of grain yield chosen for the sensitivity analysis are in the range of 1,758 and 6,353 kg ha⁻¹ for years 1993, 2001, 2007 and 2008 (Table 4). One can note that the maximum and minimum temperature varied by only 1 °C among the analyzed years. For relative humidity, solar radiation, wind speed and precipitation, the differences between the 2007 season (minimum yield) and 2001 (maximum yield) were 17.5%, 22.6%, 57.0% and 133.0%, respectively. Since no changes occurred in the crop management, it is reasonable to assume that the weather conditions were crucial to variations in the sorghum yields.

In DSSAT, photosynthesis and dry matter production is affected by the plant water stress, which depends on soil-water and precipitation. The water stress index is determined by the actual and potential crop evapotranspiration. If this ratio is lower than 1.0, it indicates that the stomatal conductance should be reduced to prevent desiccation of the plant. This assumption is used to reduce photosynthesis in proportion to transpiration (Jones et al., 2003). Figure 2 shows that in

2007, water stress was observed in three of the five phases in which the crop cycle was divided. Despite having a total accumulated rainfall depth higher than that observed in 1993, sorghum yield was lower in 2007. The year 2001, which delivered the highest productivity, water stress is not observed throughout the crop cycle (Figure 2). In 2008 and 1993 water stress occurs during the grain filling stage a critical periods regarding the effect of water stress on sorghum productivity.

Results based on the sensitivity analyses (Figure 3a), demonstrated that variations in precipitation are in phase and positively correlated with variation of productivity. It is important to note that in 2007, whose precipitation was 431 mm, changes in yield due to changes in precipitation is higher as compared to the other years (Figure 3a). Increase or reduction of rainfall in the range of $\pm 40\%$ resulted in fluctuation in yield from -50% to $+54\%$. On the other hand, in 2001 whose accumulated precipitation is considerably higher and well distributed (Figure 2), the sorghum yield is hardly affected by changes in rainfall.

In 1993 (Figure 3b) and 2001 (Figure 3d) our results show that changes in solar radiation lead to the largest yield variation. A 20% increment in the solar radiation resulted in a 30% increase in yield. This effect is however no linear in the sense that 10% reduction of radiation is associated with a 35% drop in yield.

In 2007 (Figure 3a) the yield response curve to solar radiation shows a different shape. Reduction in radiation by up to 18% leads to increased productivity. However, additional reduction in solar radiation led to substantial yield drop. In 2008 (Figure 3c) the crop response to changes in radiation is weak as compared to the other years. This non-linear effect of radiation on the productivity indicates that the yield response is associated also with the interaction among the other variables. This turns out that evaluation for future crop response to climate changes is complicated, as discussed later.

The changes in productivity associated with variations in relativity humidity and wind are smaller as compared to changes in precipitation and radiation. This was not

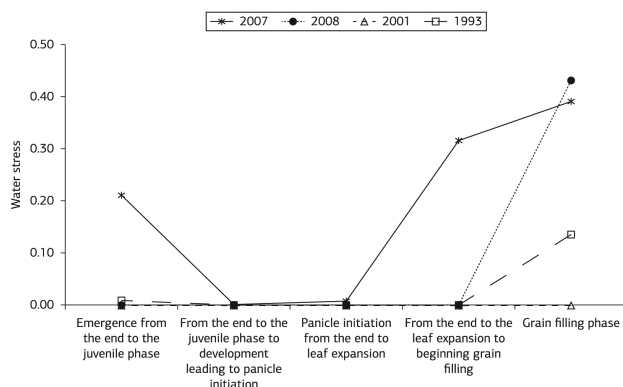


Figure 2. Simulated water stress in different phases of the sorghum cycle.

anticipated because the relativity humidity and surface wind play an important role on crop evapotranspiration. The wind speed had similar behavior in years 2007, 1993 and 2008 showing a negative relationship with productivity (e.g. higher winds/lower productivity). In 2001, when rainfall was high (955mm) the effect of wind speed (50% increase) was practically absent, showing a slight reduction in productivity due to small changes in the water balance. The effect of relative humidity on yield has demonstrated an opposite pattern as compared with that observed for wind speed.

It is interesting to note, the remarkable effect of CO_2 on sorghum yields especially when it faces water stress conditions. It was evident that the crop response to CO_2 is related to the amount of precipitation accumulated during the cycle (Figure 3). For instance, increasing CO_2 levels under mild water stress conditions does not result in increased productivity, such as in 2001 (Figure 3d). However, for dryer years the crop dependence to CO_2 is higher. The CO_2 effect on the crop yield is stronger for concentration below 270 ppm or less, a value close to the observed prior to the Industrial Revolution. For the year 2001, an increase in 50% of the atmospheric CO_2 concentration (ca. 570 ppm) is responsible for only 2% increase in productivity.

In general the results discussed above insofar the link between the CO_2 concentration and sorghum yield is concerned, agree with those obtained by Conley et al. (2001) in a FACE (Free-air CO_2 Enrichment) experiments. They found enhanced productivity of sorghum under water deficit conditions. One can argue, moreover, that when water conditions is potential (minor stress), the effect of adding CO_2 may be neglected, but under drought stress conditions it contributes to increase the sorghum yield.

Effect of temperature

The relationship between the daily thermal amplitude (difference between the daily maximum and minimum temperatures) and productivity is shown in figure 4. It should be noted that changes in the temperature amplitude was only due to reduction in the minimum temperature. It was observed that in 2001 changes in the temperature amplitude hardly affected sorghum productivity. For this year, changes in the yield occurred only when differences between the daily maximum and minimum temperatures exceeded 12 °C.

For the years 1993, 2007 and 2008 the initial pattern was similar, showing a downward trend in productivity as result of increased thermal amplitude. However, for amplitudes higher than 13 °C a positive trend is observed in the year 1993 (Figure 4).

The relationship between the amplitude changes and productivity is not straightforward because it depends on the crop water stress. This is exemplified by the crop

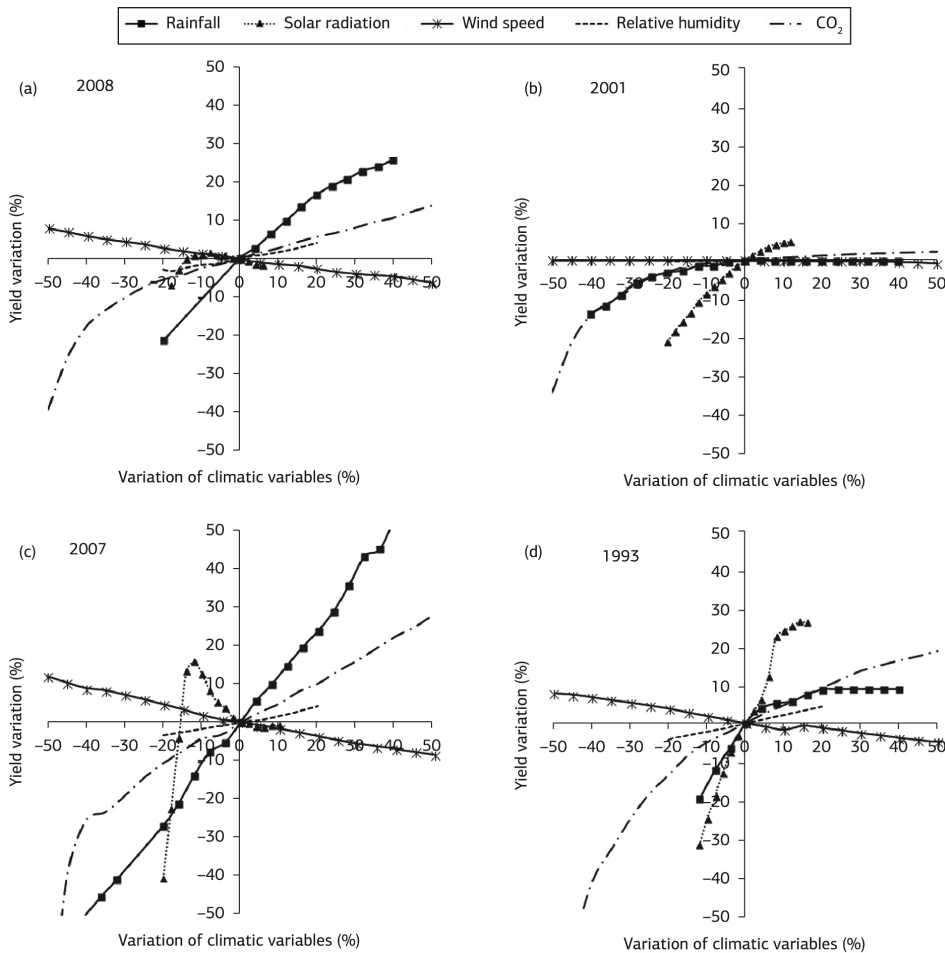


Figure 3. Sensitivity analyses of sorghum yields due to changes in meteorological variables. (a) Years 2007, (b) 1993, (c) 2008 and (d) 2001.

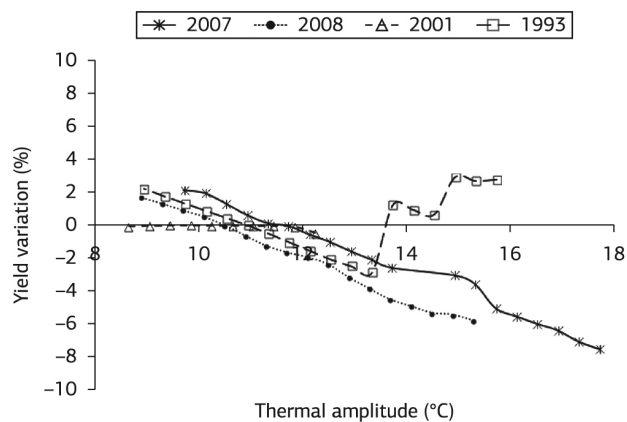


Figure 4. Response of the sorghum yield to changes in the thermal amplitude.

development in 2001 and 1993 presented here. However, it is beyond the scope of this paper to provide a detailed evaluation of the link between the thermal amplitude and the water stress.

Simulated yield variability for present and future conditions based on GCMs

The seasonal tool of DSSAT is also used to analyze the weather dependence of the sorghum yield. It has been chosen sowing dates in which the productivity of sorghum was maxima (Table 4). Based on observed data (INMET) the optimal sowing dated was November 14. When simulated based on GCMs the best sowing dates were shifted forward in the season. For the ECHAM, the best date occurs in December while for CCCma and GFDL the sowing windows move to January. According to Grossi et al. (2013) differences between the INMET and the GCMs are due to changes in the simulated rainfall. It has also been observed that with the exception of the GFDL, the date of maximum productivity occurs later in the future (Table 5).

Analyzing the distributions of productivities for the INMET and GCMs, considerable differences were observed (Figure 5). Indeed, under present day conditions yields simulated by DSSAT forced with GFDL and CCCma data were lower as compared to the INMET, while ECHAM yields are higher. Yields simulated with the CCCma data as compared to the INMET delivers differences in the median by up to 1500 kg ha⁻¹. The ECHAM, on the other

Table 5. Climate series and date of sowing

Timeseries	INMET	CCCma		ECHAM		GFDL	
		Pres.	Fut.	Pres.	Fut.	Pres.	Fut.
Period	1982-2009	1982-1999	2047-2064	1982-1999	2047-2064	1982-1999	2047-2064
Date of sowing	14 Nov	9 Jan	23 Jan	12 Dec	2 Jan	23 Jan	2 Jan

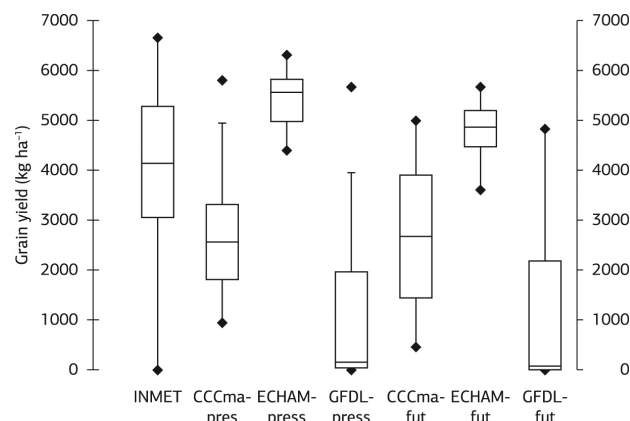


Figure 5. Simulated sorghum productivity based on INMET and GCMs for present (pres) and future (fut) intervals. In this Figure, and all the other box plots, exterior lines represent the maximum and minimum values (non-discrepant values); the outer points represent the discrepant maximum and minimum values; the upper and lower lines of the box represent, respectively, the upper quartile (75%) and the lower quartile (25%), while the center line represents the median (50%).

hand provided the highest yield with the lowest amplitude (Figure 5).

Comparison between the present and the future period, shows a slight decrease in productivity based on ECHAM model for the 2047 to 2062 interval (Figure 5). For the other GCMs, a small difference between the medians were observed but with an increased variability of the sorghum productivity for future period.

Simulated yields are affected by the accumulated rainfall as well as by the daily variability during the crop cycle (Figure 6a). For the 1982-1999 period, the ECHAM simulated 575 mm cycle⁻¹ of rainfall and productivity of 5561 kg ha⁻¹. The GFDL model, simulated precipitation by about 92mm cycle⁻¹ and the lowest yield 142 kg ha⁻¹. This is in agreement with the sensitivity analyses previously performed, showing that the water resource is the major constrain to overcome lower sorghum yield. Reduced productivity of sorghum for the future interval is also predicted to occur by the ECHAM model as compared to PD conditions, due to lower rainfall estimates. However, this was not observed with the CCCma, in which reduced precipitation did not lead to lower yields because other climate variables compensate reduced precipitation.

When mean conditions are evaluated, observed (INMET) values of radiation were similar to those simulated by the

GCMs (Figure 6b), but they exhibit distinct inter-annual variability. The GFDL model results demonstrated that for the present period in more than 25% of the years the accumulated radiation along the sorghum cycle was below the critical value of 1900 MJ m⁻². This very likely reduces productivity in some years (Figure 6b). Regarding to the future period, a sharp drop in radiation for all GCMs are observed that potentially may affect the productivity. It should be mentioned that reduced productivity was only observed if DSSAT is driven with the ECHAM model output.

In terms of temperature it was observed that all GCMs are colder as compared to the INMET observations. The observed averaged temperature was 26.1 °C, which was approximately 2 °C higher than the 24.4 °C, 23.8 °C and 24.4 °C simulated by the CCCma, ECHAM and GFDL, respectively (Figure 5c). The lowest average temperature predicted by the ECHAM is another factor that may be related to the higher sorghum yields, as compared to the other models and observations.

The implementation of future climate conditions leads to higher temperature with respect to PD, by 2.3 °C for CCCma and ECHAM and by 2.7 °C for the GFDL. This led to shorter sorghum cycle under future climate conditions (not shown). Moreover, reduced precipitation and radiation amounts, which were simulated by all models do not favor future sorghum yields. The temperature amplitude (bellow 8 °C) simulated by the ECHAM for both present and future periods, is substantially lower as compared with the other GCMs (Figure 6d). According to the sensitivity analyses (Figure 7), the individual effect of lower temperature amplitude resulted in higher simulated productivity.

In general the GCMs overestimated the wind speed as compared to observed data (INMET). As discussed in the sensitivity analyses, wind speed and relative humidity are variables that indirectly affect productivity through the soil-water availability/evapotranspiration. Figures 6e, f show that the sorghum yield benefited by these two variables, insofar the ECHAM input data is concerned with weaker wind speed and higher relative humidity in comparison with other models. As for the GFDL model, the lowest relative humidity values were simulated for both present and future periods that led to the lowest simulated yields. Sensitivity analysis to investigate the effect of CO₂ (not shown) revealed that under optimum water conditions (no stress), the effect of increased CO₂ is negligible whereas under

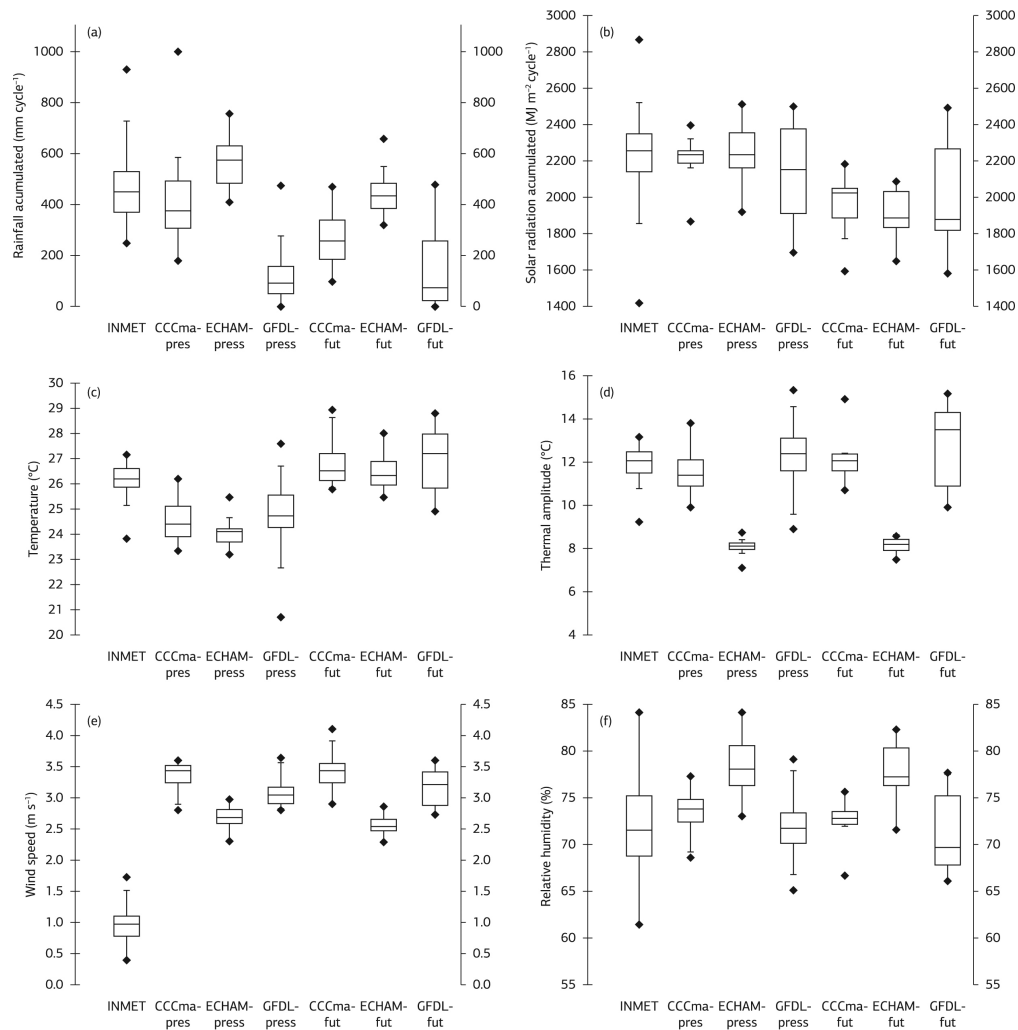


Figure 6. Simulated present day and future values of rainfall (a), solar radiation (b), temperature (c), thermal amplitude (d), wind speed (e) and relative humidity (f).

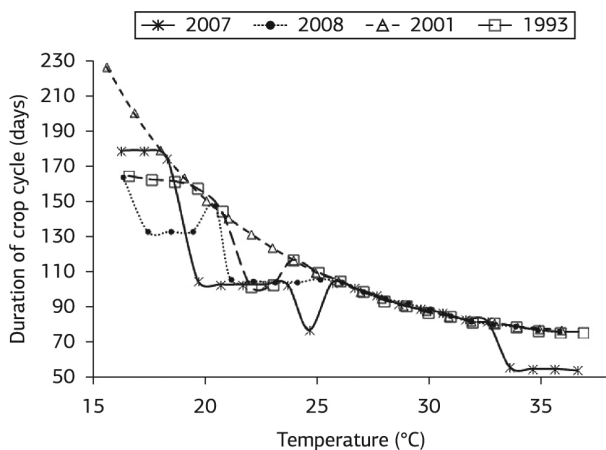


Figure 7. Response of the sorghum cycle length to changes in temperature.

water stress conditions it has a positive effect increasing the productivity.

This gives a broad range of possibilities in estimating the response of agriculture to climate variability and change.

This limitation can be partially alleviated by using outputs of different GCMs to figure out the most suitable climate model for a specific region.

CONCLUSION

Based on sensitivity analyses we demonstrated that:

The sensitivity of the sorghum productivity to climate variables varied from year to year and is tightly dependent on water availability.

The sorghum yield is more sensitive to rainfall, solar radiation and CO₂ concentration. The increase in CO₂ concentration has a positive (marginal) effect on yield in years that the crop has (has not) been submitted to water stress.

Accumulated solar radiation along the sorghum cycle below 1900 MJ m⁻² reduces the yield.

The sorghum yield is less sensitive to changes in relative humidity and wind speed as compared to other weather elements.

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