



Climate change impacts on potential maize yields in Gambella Region, Ethiopia

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Abstract

Changing climate conditions are supposed to have particularly strong impacts on agricultural production in the tropics with strong implications on food security. Ethiopia's economy is profoundly dominated by agriculture, contributing to around 40% of the gross domestic product. Thereby, Ethiopia is one of the most vulnerable countries to the impact of climate change and has a wide gap in regional climate change impact studies. In this study, we systematically investigate climate change impacts on yields for the Gambella region in Ethiopia, exemplarily for maize. Here, we show how yields change until 2100 for RCPs 2.6, 4.5, and 8.5 from a climate model ensemble under rainfed and irrigated conditions. While rainfed yields decrease by 15% and 14% respectively for RCPs 2.6 and 4.5, yields decrease by up to 32% under RCP 8.5. Except for RCP 8.5, yields are not further decreasing after 2040–2069. We found that temperature increase, changing soil water availability, and atmospheric CO₂ concentration have different effects on the simulated yield potential. Our results demonstrate the dominance of heat response under future climate conditions in the tropical Gambella region, contributing to 85% of total yield changes. Accordingly, irrigation will lose effectiveness for increasing yield when temperature becomes the limiting factor. CO₂, on the other hand, contributes positively to yield changes by 8.9% for RCP 8.5. For all scenarios, the growing period is shorted due to increasing temperature by up to 29 days for RCP 8.5. Our results suggest that new varieties with higher growing degree days are primarily required to the region for adapting to future climate conditions.

Keywords Climate change · Agriculture · Regional study · Crop model

Introduction

Many studies agree that climate change is real and that the poorest and most vulnerable people will be the most affected by reduced availability, lower quality, and higher prices of food (Barbier and Hochard 2018; Diffenbaugh and Burke 2019; Hallegatte et al. 2018). The last three decades were each warmer than every previous decade since temperature records

began in 1850 (IPCC 2014). However, there are significant regional differences in temperature and precipitation changes (IPCC 2013). Many studies have considered the impacts of future climate changes on food production at the global or very large scale (Iizumi et al. 2017; Lobell et al. 2008; Najafi et al. 2018; Parry et al. 2004).

However, there is a gap in regional impact studies, especially for Ethiopia, which is vulnerable to climate change, since it is not a food secure country (Alemu and Mengistu 2019; Cochrane 2018; Endalew et al. 2015; Mekonnen and Gerber 2017). Earlier studies on climate change impacts on crop production in Ethiopia were either at the national (Bryan et al. 2009; Kassie et al. 2014; Wubie 2015; Yalew et al. 2018) or larger scale such as the East African regional levels (Abera et al. 2018; Leal Filho et al. 2017; Niang et al. 2014). There are only a few studies at subnational levels within Ethiopia (Abera et al. 2018).

Ethiopia's economy is highly dominated by agriculture (Alemu and Mengistu 2019; Alemu et al. 2003; Yalew et al. 2018), contributing about 40% of the GDP (Shiferaw 2017).

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The World Bank's analysis predicts that climate change will lower Ethiopia's gross domestic product (GDP) growth by 0.5–2.5 % per annum (Gebreegziabher et al. 2016). In Ethiopia, effects of an enduring drought for continuous years from 1983 to 1985 reduced agricultural production in the county (Aragie 2013; Kassie et al. 2014). Food supply is one of society's key sensitivities to climate in Ethiopia (Porter et al. 2014).

The IPCC's fifth assessment report indicates that future climate change will lead to an increase in climate variability and in frequency and intensity of extreme events in Ethiopia (Niang et al. 2014). Over the past four decades, changing precipitation patterns (both dry and wet periods) have been observed in many parts of Ethiopia (Aragie 2013; Seleshi and Zanke 2004). Thereby, precipitation has shown a general decreasing trend since the 1990s (Leal Filho et al. 2017). This decrease already manifests in multiple effects on agricultural production in Ethiopia (Aragie 2013; Seleshi and Zanke 2004). In the last three decades, the Gambella region in the western part of Ethiopia was faced with frequent climatic variability and agro-ecological change (Gummadi et al. 2018). The changing rainfall pattern in combination with warming trends could make rainfed agriculture more risky and barrier food production in Ethiopia (Gebreegziabher et al. 2016) and in the region. For future conditions, a temperature increase of 2 to 2.5 °C is supposed to have significant implications on agriculture in parts of Africa (Belloumi 2014) and Ethiopian (Leal Filho et al. 2017). In general, climate change induced increases in temperatures, and rainfall variation and the frequency and intensity of extreme weather events are adding pressures on Ethiopian agricultural production (Adhikari et al. 2015; Gbegbelegbe et al. 2014; Gebreegziabher et al. 2016). This creates challenges for the agricultural sector for possible future adaptation (FAO 2017).

For proper planning, it is required to investigate impacts of climate change on crop yields at regional scales. This is the first regional study that systematically investigates climate change impacts on agricultural production for Gambella in Ethiopia, exemplarily for maize. Over the last few decades, the maize yield in Ethiopia has shown an unprecedented transformation. Maize yields have doubled from about 1.6 t ha⁻¹ during the early 1990s to 3.7 t ha⁻¹ in recent years (Van Dijk et al. 2020). In 2016/2017, maize crop accounts for 57% of Gambella region's crop production by small-scale farmers and 56% of the region's harvested area (Degife et al. 2019). The following study comprises a representative selection of climate change impact scenarios to show the range of possible futures for different representative concentration pathways (RCPs). The approach is based on climate model results, which feed a process-based biophysical crop model that simulates crop yields under specific climate conditions.

Data and methods

Study area

Gambella is one of nine administrative regions in the southwestern part of Ethiopia. The region covers a total area of 25,521 km² (Tadesse 2007). It shares borders with South Sudan and two other Ethiopian regions: Oromia to the north and east and the Southern Nations, Nationalities and Peoples' Regional State (SNNPRS) to the south (Fig. 1). Gambella region altitude progressively declines from east to west, ranging from 2200–1000 m a.s.l. in the east to 500–900 m in the center and 300–500 m in the west (Woube 1999). The mean annual temperature of the region varies from 17.3 to 28.3 °C and the annual precipitation of the region varies from 800 to 1200 mm (Degife et al. 2019). The region is highly suitable for agriculture (Zabel et al. 2014).

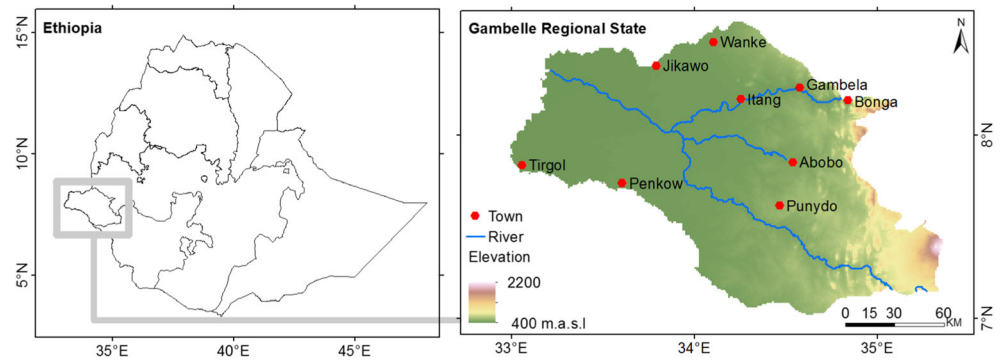
Climate data

Dynamically downscaled general circulation models (GCM) projections have proven to be suitable tools for providing high-resolution climate drivers for regional and local assessments of climate change impacts and extremes (Shiferaw et al. 2018). We used climate data (near-surface air temperature, precipitation, surface downwelling longwave radiation, surface downwelling shortwave radiation, total cloud fraction, near-surface wind speed, sea level pressure, near-surface relative humidity) from the regional climate model Rossby Centre Regional Atmospheric Climate Model (RCA4) with large-scale forcing from five different GCMs (Table 1). They originated from the coupled model intercomparison project phase 5 (CMIP5). To capture the range of existing impact scenarios, we used three representative concentration pathways: from a very high baseline scenario (RCP 8.5), medium stabilization scenario (RCP 4.5), and a very low forcing level scenario (RCP 2.6). The authors do not consider medium stabilization scenarios (RCP 6.0) to prevent data replication (IPCC 2014; Van Vuuren et al. 2011).

RCA4 data was acquired over the African domain from the coordinated regional climate downscaling experiment (CORDEX) (Dosio and Panitz 2016) with horizontal grid spacing of 0.44° (≈ 50 km) and 3-h temporal resolution. The applied data includes the historical period (1970–2005) and the RCPs representing climate change scenarios from 2020 until 2099. Both for the past and the future, each RCA4 run representing an RCP consists of a five-member ensemble from five different driving GCMs.

To use the climate data in the PROMET model (see the "PROMET model" section), we applied a statistical downscaling to 30 arc seconds and a temporal interpolation to 1 h. Additionally, a bias correction was applied to temperature and precipitation after the downscaling and temporal interpolation

Fig. 1 Gambella locational map: Federal Democratic Republic of Ethiopia (left) and Gambella regional state (right). Map uses UTM Zone 36N projection



(Marke et al. 2014). For this, we used global monthly observation climatology from the WorldClim database v2 at 30 arc second resolution (Fick and Hijmans 2017). It is representative for the time period 1970–2000. The bias correction ensures that the average 1970–2000 monthly bias corrected and interpolated set of 5 climate model results of air temperature and precipitation at each of the 30 arc second grid cells that encompass Gambella region closely resemble the monthly values in the WorldClim dataset.

PROMET model

The PROMET model was applied to simulate crop growth of maize in the Gambella region (Delzeit et al. 2017; Hank et al. 2015; Mauser et al. 2015; Minoli et al. 2019; Zabel et al. 2019). The PROMET model is an agro-hydrological land surface process model, which contains mechanistic and biophysical parameters that operate at different spatial scales — from field scale to global scale. It was used because it is well parameterized and validated for maize in various regions (Müller et al. 2017; Franke et al. 2020). It uses first-order physical and physiological principles to determine net primary production and respiration based on approaches from Ball et al. (1987), Farquhar et al. (1980), and Xinyou and Van Laar (2005) combined with a phenology and a two-layer canopy architecture component. PROMET takes into account the dependency of net primary production and phenology on environmental conditions including meteorology, CO₂ concentration for C3 and

C4 pathways, and water and temperature stress. The mass and energy balance of the canopy and underlying soil surface are iteratively closed for each simulation time step. The canopy and phenology component allocates assimilates into the different plant organs of the canopy depending on the phenological stage of development. The phenology model is described in detail in Minoli et al. (2019). Assimilates that are accumulated within the fruit fraction during the growing period determine the dry biomass available for yield formation. The simulation is performed on an hourly time step to account for non-linear reactions of crop growth to abiotic conditions like water and temperature stresses. Depending on the reaction of the considered crop to meteorological and soil-specific conditions, the crop either may die due to water, heat, or cold stress before being harvested or may not reach maturity. In both cases, this results in total yield loss.

In the context of this study, we simulate maize potential yields, assuming a perfect crop management. This means that crop is sowed at the appropriate date, perfectly supplied with nutrients at any time and pests and diseases are assumed to be controlled. These assumptions are chosen to isolate climate change impacts on yields without additional constraints. We simulated both rainfed and irrigated practices, in order to quantify the effect of irrigation for possible future adaptation. In the case of irrigation, we assume that no water stress occurs at any time.

Sowing dates are taken from available regional statistics for Gambella, stating that growing period starts in the middle of April. Sowing dates are kept constant and not shifted in this

Table 1 CMIP5 GCMs used in this study

Model	Description	Modeling center
EC-EARTH	A European community Earth-System Model	EC-Earth consortium
HadGEM2-ES	Hadley Global Environmental Earth System Model 2	Met Office Hadley Centre (MOHC)
MIROC5	Model for Interdisciplinary Research on Climate	International Centre for Earth Simulation
MPI-M-MPI-ESM-LR	MPI Earth System Model running on low resolution grid	Max Planck Institute for Meteorology (MPI-M)
NCC-NorESM1-M	The Norwegian Earth System Model	Norwegian Climate Centre (NCC)

study in order to investigate the climate effect only. Maturity is simulated internally by PROMET according to the crop phenological progress.

We applied a settling time of 5 years (1970–1974) to all PROMET simulations for initializing soil moisture and other hydrological parameters, so that we use the 30-year baseline period from 1975 to 2004.

Decomposition analysis

In order to separately quantify the different effects of increasing temperature, available soil water, and atmospheric CO₂ concentration on the simulated potential yield, we applied a decomposition analysis for the 2070–2099 period. Therefore, we conducted all simulations for both rainfed and irrigated conditions. Additionally, we conducted a control run with fixed CO₂ concentration at 360 ppm until 2099 to isolate the beneficial effects of CO₂.

First, to obtain the proportion of yield change due to temperature increase, we subtract 1975–2004 yields from yields in 2070–2099 from the irrigated fixed CO₂ run, where CO₂ and water have no effect. Second, we subtract the irrigated fixed CO₂ 2070–2099 yields from the irrigated 2070–2099 yields with increased CO₂ concentration to obtain the isolated benefit from CO₂ fertilization. Third, we subtract the irrigated 2070–2099 yields from the rainfed yields to get the isolated effect of water stress on yields.

By comparing historical and future yields for the fixed irrigated CO₂ yields, besides temperature, also other effects are cumulated, such as changes of wind speed and radiation, that impact on yields. Since PROMET considers the available soil water for the simulation of crop water stress, including a detailed description of relevant hydrologic processes in the soil-plant-atmosphere continuum, we refer to the water stress effect rather than precipitation in our analysis.

Land-use scenario

In this study, we assume that agricultural land in Gambella will expand into the not-yet-used legally available land for agricultural expansion. Therefore, all our results refer to all current cropland area in addition to the Top 50 expansion scenario according to Degife et al. (2019) and Zabel et al. (2019). The scenario assumes the use of today's cropland in addition to the best (in terms of highest potential yields) 50% expansion area. The used land is assumed to be constant over time in order to not affect the results.

Results

In average for 2070–2099, temperature increases by 1.6, 2.6, and 5.3 K under RCP 2.6, RCP 4.5, and RCP 8.5 scenarios for

the Gambella region from April to September (Fig. 2), referring to the growing period of maize (Central Statistical Agency 2017). The model median of the average temperature in the region for the historical 1975–2004 period is 26.5 °C.

Precipitation changes on average for 2070–2099 by –11.9, –10.9, and –22.6% under RCP 2.6, RCP 4.5, and RCP 8.5 scenarios for the Gambella region over the fixed statistical growing period of maize (Fig. 2). The model median of the absolute amount of precipitation in the region for the historical 1975–2004 period is 611 mm. While the models agree quite well for temperature and precipitation change as can be seen by the model spread in Fig. 2, absolute precipitation values vary largely over historical periods (e.g., mean 1975–2004 values range between 888 mm for EC-EARTH and 530 mm for HadGEM2-ES). Overall, the tropical climate is characterized by hot temperatures and high precipitation.

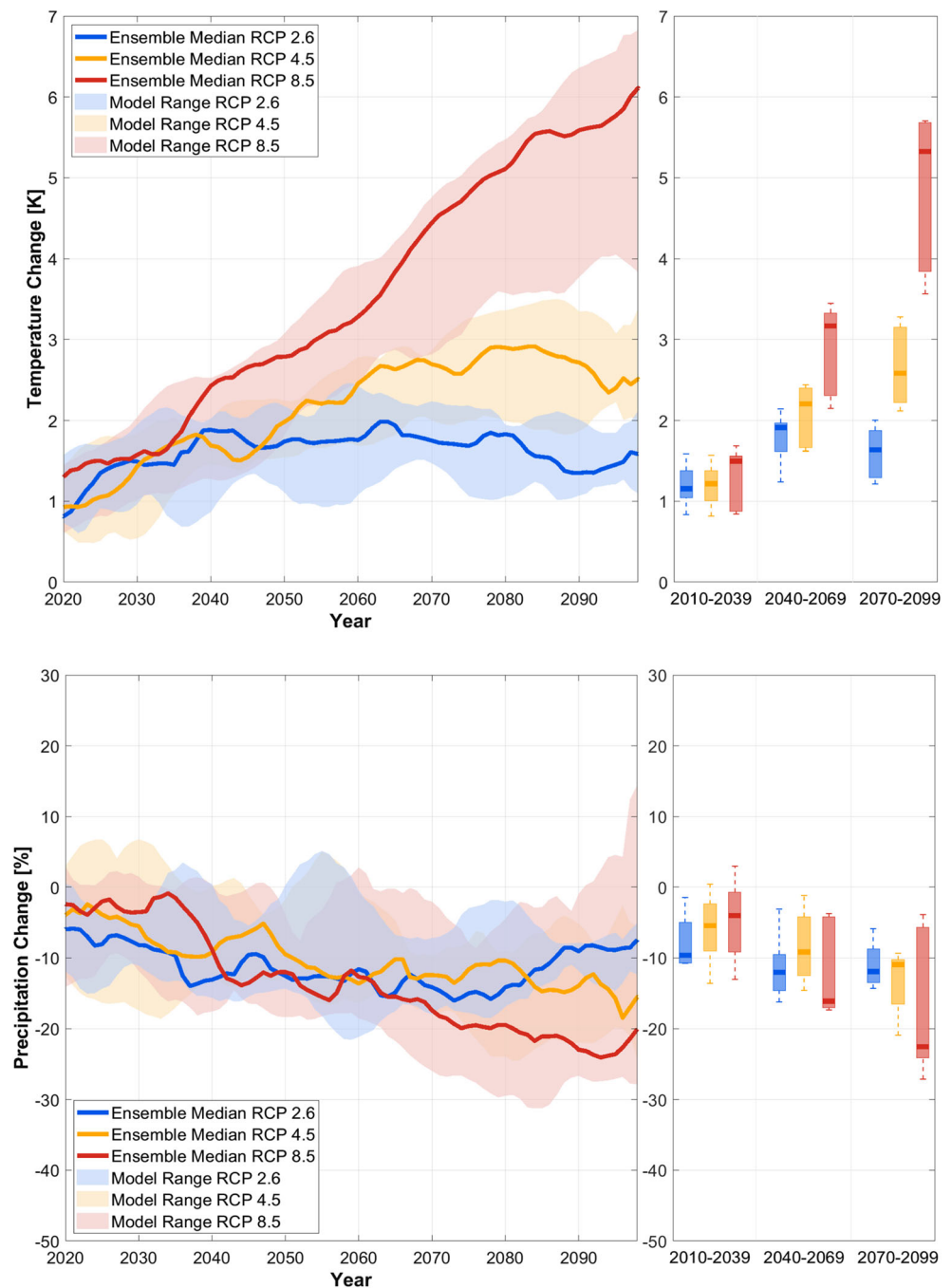
The changing climate conditions have several impacts on maize yield that are described in the following, first showing impacts on rainfed maize, followed by showing impacts on irrigated maize. Subsequently, a decomposition analysis shows the different impacts of each, temperature, water availability, and CO₂ on maize yield separately.

Climate change impact on rainfed yields

Figure 3 shows the temporal course of rainfed potential maize as median and min-max range of all five driving GCMs (Table 1) from 2020 to 2099 for RCP 2.6, RCP 4.5, and RCP 8.5. Thereby, potential maize yield decreases by 1.1%, 9.0%, and 26.2% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively when applying a linear regression from 2020 to 2099. The model spread is decreasing with higher RCP. The range of models is between –17.4% (MIROC5) and 17.2% (MPI-M-MPI-ESM-LR) for RCP 2.6, –29.4% (HadGEM2-ES) and –0.2% (MIROC5) for RCP 4.5, and between –37.6% (MPI-ESM) and –20.5% (NCC-NorESM1-M) for RCP 8.5.

Slicing the time frame into 30-year climate normals (2010–2039, 2040–2069, and 2070–2099) and comparing each climate normal with the reference period (1975–2004), the right side of Fig. 3 shows the percentage change of potential rainfed maize yield under the three RCPs. Maize yield declines by 12.8%, 4.9%, and 11.2% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2010–2039 and by 17.8%, 14.9%, and 22.1% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2040–2069 and by 14.7%, 14.1%, and 32.4% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2070–2099. Thus, yields for RCP 2.6 and RCP 4.5 decrease until the middle of the century and do not further decrease until 2100, while the decrease in RCP 8.5 continuously gets stronger. This effect is even stronger for irrigated conditions (compare the “Climate change impact on irrigated yields” section). Compared with Fig. 2,

Fig. 2 Projected absolute change of near surface air temperature in kelvin (upper) and precipitation (lower) over the statistical growing period of maize (April–September) from 2020 to 2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red) compared to the reference period (1975–2004). The dark colored line shows the model median, while the light color surface shows the range of the five different driving GCMs between minimum and maximum ensemble member for each RCP. A 5a moving average is applied. The right side shows the boxplots of each of the 30-year averaged values over the range of all models, illustrating the median and the interquartile range (25 to 75 percentile), while the whiskers show the highest (max) and lowest (min) 30-year average value of all models



also temperature does not further increase after 2060 in RCPs 2.6 and 4.5.

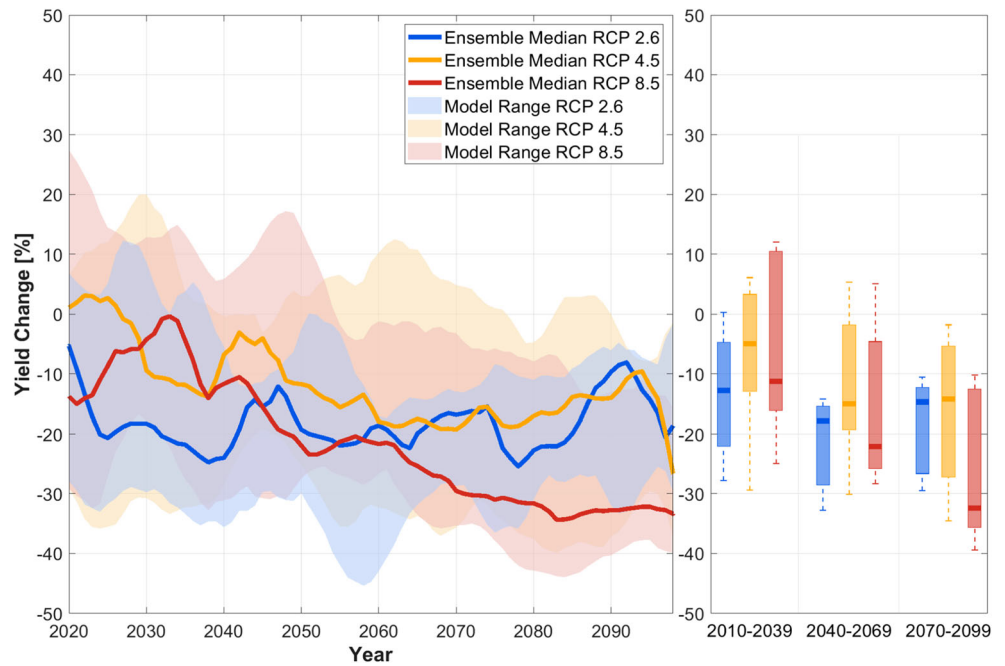
Climate change impact on irrigated yields

Figure 4 shows the temporal course of irrigated potential maize yield from 2020 to 2099 for RCP 2.6, RCP 4.5, and RCP 8.5 in comparison to the rainfed reference period (1975–2004). The potential irrigated maize yield decreases by 4.3%, 23.0%, and 44.5% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively, but still is higher than the rainfed yield. The

range of model results is between -12.9% (MIROC5) and 7.2% (MPI-M-MPI-ESM-LR) for RCP 2.6, between -37.1% (HadGEM2-ES) and -17.8% (EC-EARTH) for RCP 4.5, and between -52.4% (HadGEM2-ES) and -40.4% (NCC-NorESM1-M) for RCP 8.5.

Slicing the time frame into 30-year climate normals (2010–2039, 2040–2069, and 2070–2099) and comparing each climate normal with the reference period (1975–2004). Figure 4 shows maize irrigated yield changes by $+18.5\%$, $+27.3\%$, and $+18.5\%$ under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2010–2039 and by 8.7% , 4.9% , and -9.4%

Fig. 3 Projected percentage change of rainfed maize potential yield from 2020 to 2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red) compared to the rainfed maize reference (1975–2004). The dark colored line shows the model median, while the light color surface shows the range of resulting yield from the five different driving GCMs between minimum and maximum ensemble member for each RCP. A 5a moving average is applied. The right side shows the boxplots of each of the 30-year averaged yield over the range of all models, illustrating the median and the interquartile range (25 to 75 percentile), while the whiskers show the highest (max) and lowest (min) 30-year average value of all models



under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2040–2069 and by 17.7%, 1.4%, and – 27.1% under RCP 2.6, RCP 4.5, and RCP 8.5 respectively for the time period 2070–2099.

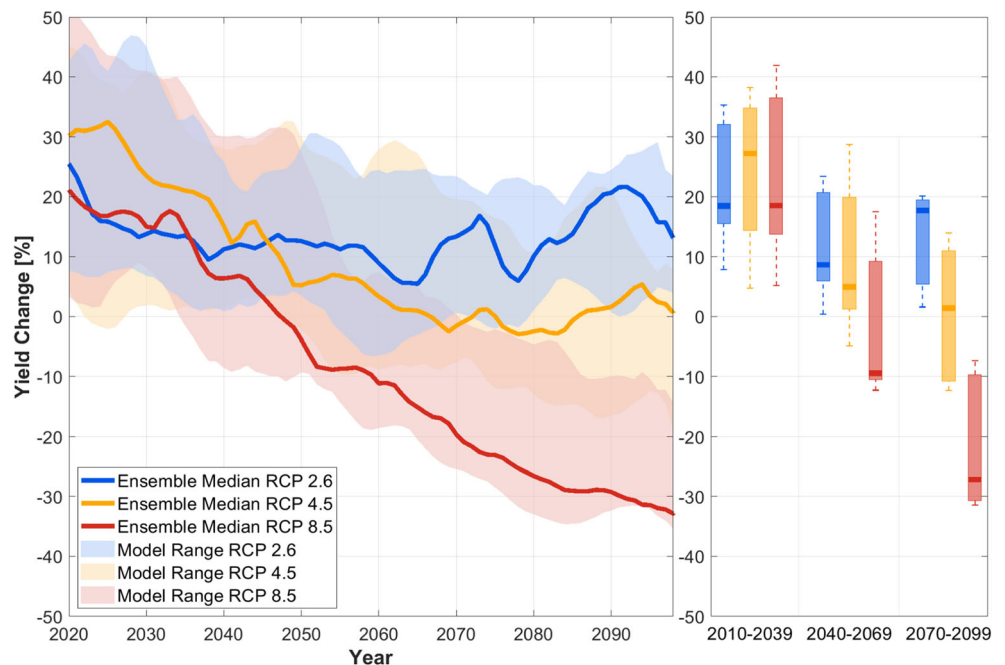
Comparing Fig. 3 with Fig. 4, both relative to the rainfed reference, irrigated maize in percentage change decreases more than rainfed maize. However, irrigation still increases maize yield compared to rainfed until 2100, at least for RCP 2.6 and RCP 4.5, while for RCP 8.5 applying irrigation does not contribute to higher maize yield. Around the year 2050,

irrigated maize yield will have the same magnitude than the reference rainfed yield for RCP 8.5.

Decomposition analysis

Figure 5 shows the contributions of temperature, water stress, and CO₂ on maize yield changes for all RCPs until 2070–2099. Temperature contributes to total maize yield losses by up to 85% for RCP 8.5, 67% for RCP 4.5, and 52% for RCP 2.6. This demonstrates the strong dominance of temperature

Fig. 4 Projected percentage change of irrigated maize potential yield from 2020 to 2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red) compared to rainfed reference maize yield (1975–2004). The dark colored line shows the model median, while the light color surface shows the range of resulting yield from the five different driving GCMs between minimum and maximum ensemble member for each RCP. A 5a moving average is applied. The right side shows the boxplots of each of the 30-year averaged yield over the range of all models, illustrating the median and the interquartile range (25 to 75 percentile), while the whiskers show the highest (max) and lowest (min) 30 year average value of all models



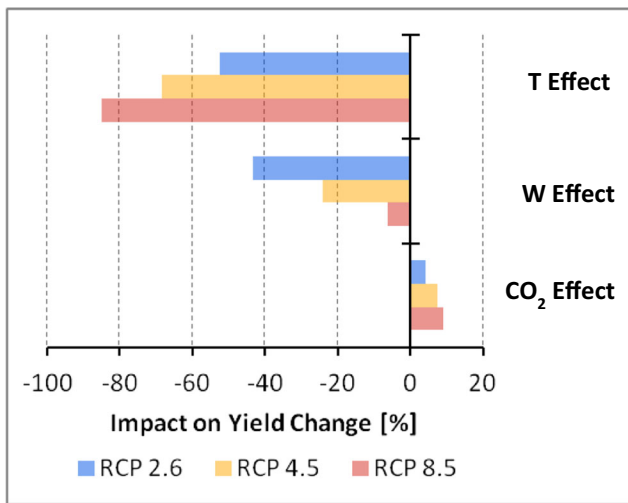


Fig. 5 Decomposition of temperature (T), water stress (W), CO₂ impacts on yield. Each element shows the proportion on median maize potential yield change from the five different driving GCMs for the 30-year average period 2070–2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red)

stress for all RCPs in the Gambella region until the end of the century. It is the by far dominant effect responsible for yield reduction in the region, mainly in RCP 8.5. Temperature is becoming the limiting factor since crops reach its maximum temperature thresholds. On the other hand, the effect of water stress on maize yield is decreasing with higher RCPs. Water stress reduces maize yield by 43.4% for RCP 2.6, 24.1% for RCP 4.5, and 6.2 for RCP 8.5. The positive CO₂ effect on maize yield was expected to be low, since maize is C4 crops. The effect is highest for RCP 8.5 (increases maize yield by 8.9%), because in this scenario also the highest CO₂ concentrations occur.

In order to get deeper understanding on how temperature increase impacts on yields, we analyze the growing period length (number of days from sowing to maturity) in the Gambella region. Temperature affects crop yields not only

by temperature limitations of photosynthesis but also by determining the phenological progress of crops. Thereby, crop varieties require different temperatures to reach maturity. Generally, higher temperature is associated with faster maturity that goes along with a reduced length of the growing period, which usually results in lower yields (Minoli et al. 2019). Besides temperature increase, water stress is another abiotic factor that can accelerate phenology during the reproductive stage and thus also leads to faster maturity.

Figure 6 shows a clear correlation (R^2 0.81) between temperature increase and decreasing growing period length for irrigated maize under RCP 8.5, without the occurrence of water stress. The shortened growing period is strongly associated with yield reductions, shown by the strong correlation (R^2 0.9) between growing period length and yield, which suggests that the impact of temperature increase on the growing period has an even larger effect on yields than temperature limitations on photosynthesis.

For the reference period (1975–2004), the growing period was 125 days for rainfed and 132 for irrigated. As expected, the growing period is shorter when considering water stress in rainfed case than for irrigated case. Figure 7 shows for all RCP scenarios the projected length of the growing period from 2020 to 2100 under rainfed and irrigated conditions. For all scenarios, the average growing period length becomes smaller, which means earlier mature harvest. While the growing period length is reduced by 5 days for RCP 2.6 both for irrigated and rainfed conditions between 2070 and 2099 in comparison with 2010–2039, it is reduced by 15 days for RCP 8.5 in case of irrigation and 11 days in case of rainfed. Compared to the reference, the growing period length decreases by 29 days in case of irrigation and by 23 days in case of rainfed until 2070–2099. Accordingly, the growing period is shortened more in the irrigated case than in the rainfed case, indicating that the effect of water stress on the growing period length is reduced. As visible in Fig. 7, the growing periods of irrigated and rainfed maize are getting more similar until the end of the century, since temperature increase alone (without

Fig. 6 Correlation between increasing near surface air temperature and the growing period length for irrigated maize (left) and between potential irrigated maize yield and the growing period length (right) for RCP 8.5 (2020–2099) in the Gambella region

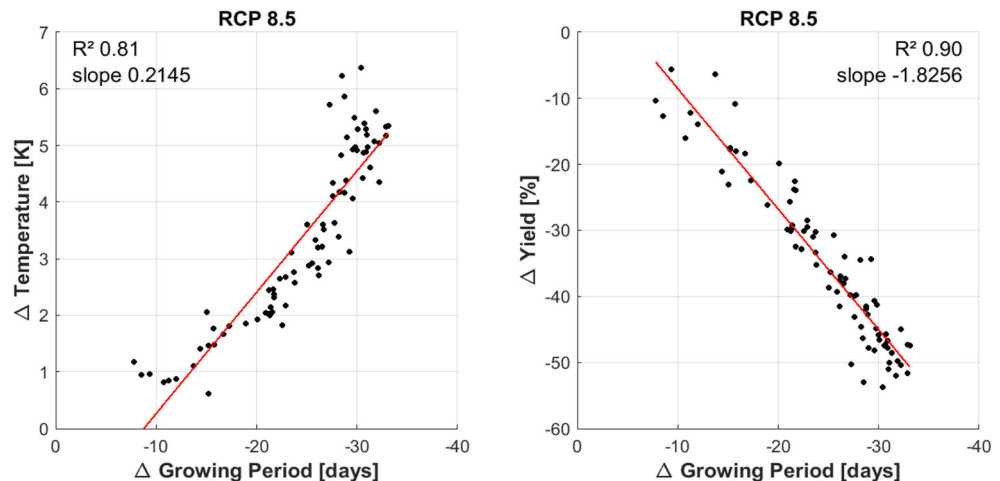
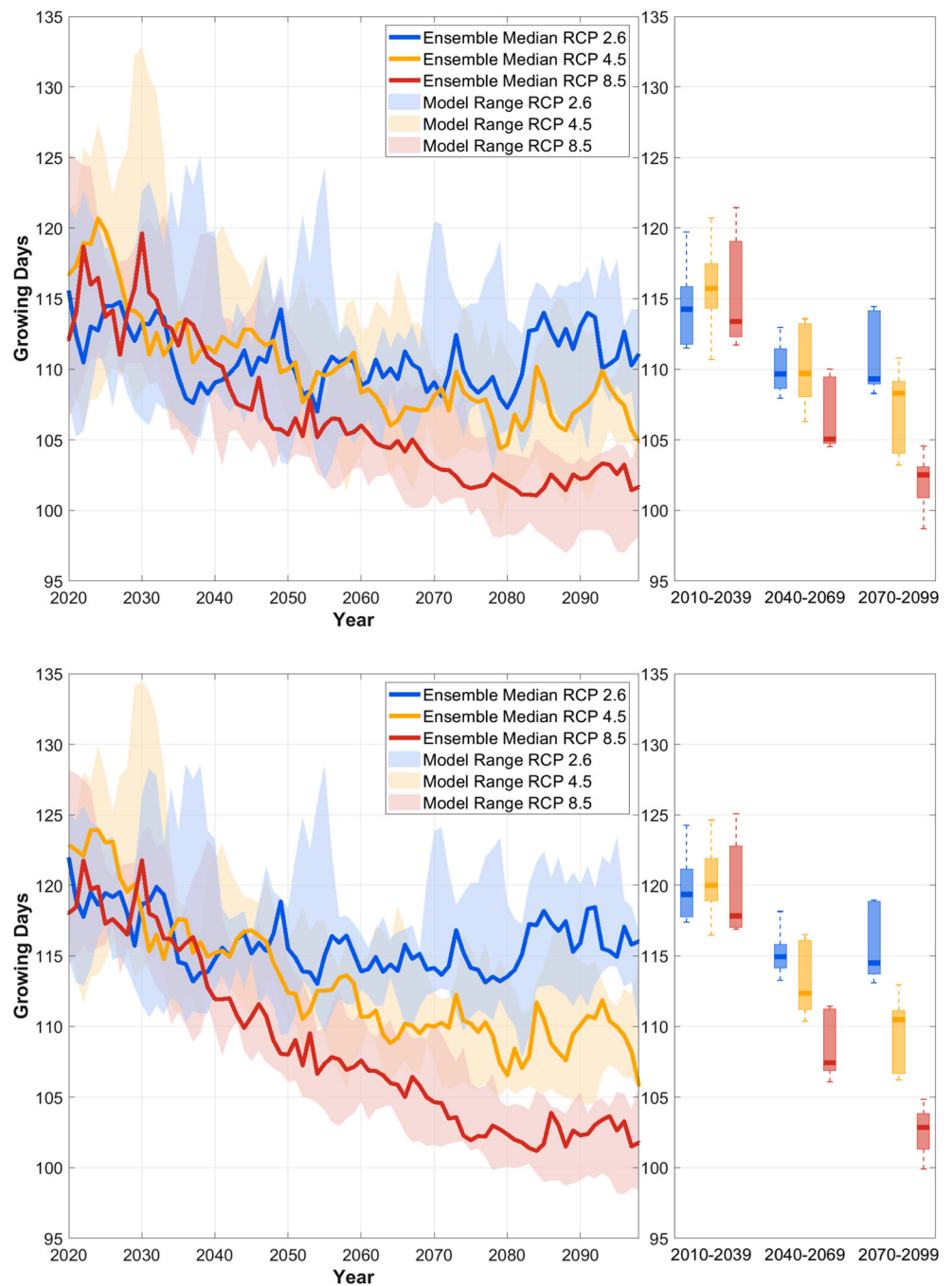


Fig. 7 Projected growing period length of rainfed (upper) and irrigated (lower) maize from 2020 to 2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red). The dark colored line shows the model median, while the light color surface shows the range of resulting maize yield from the five different driving GCMs between minimum and maximum ensemble member for each RCP. A 5a moving average is applied. The right side shows the boxplots of each of the 30-year averaged growing periods over the range of all models, illustrating the median and the interquartile range (25 to 75 percentile), while the whiskers show the highest (max) and lowest (min) 30-year average value of all models



water stress) shortens the growing period in such a strong magnitude, that water stress occurs less frequently during the reproductive stage (Fig. 8).

Discussion

Agriculture is a sector that is closely linked to climate and that is thereby naturally prone to impacts of climate change (Bedeke et al. 2019; FAO 2016; Lobell et al. 2008). Therefore, the drivers of these changes must be understood

to be able to propose more effective strategies for future food security (Alemu and Mengistu 2019; Najafi et al. 2018). We show in this paper the impact of climate change for different RCPs on potential maize yield in the Gambella region of Ethiopia, assuming the use of today's cropland in addition to the best 50% expansion area (Degife et al. 2019). Thereby, the RCPs describe ranges of possible future development and are selected to represent a low, medium, and strong climate change signal. The results demonstrate that maize potential yield is supposed to decrease under all selected RCPs, with highest yield reductions under RCP 8.5.

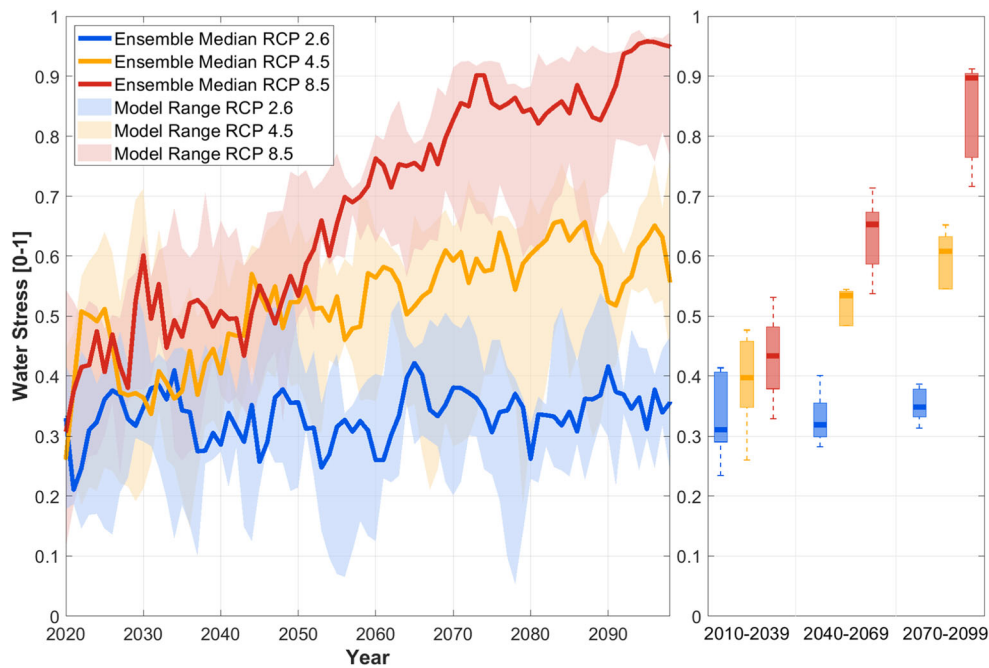


Fig. 8 Projected water stress for rainfed maize from 2020 to 2099 under RCP 2.6 (blue), RCP 4.5 (orange), and RCP 8.5 (red). Water stress is shown as a limiting factor between 0 and 1, where 1 indicates no water stress, while 0 means a full limitation of available soil water. The dark colored line shows the model median, while the light color surface shows the range of resulting maize yield from the five different driving GCMs

between minimum and maximum ensemble member for each RCP. A 5a moving average is applied. The right side shows the boxplots of each of the 30-year averaged growing periods over the range of all models, illustrating the median and the interquartile range (25 to 75 percentile), while the whiskers show the highest (max) and lowest (min) 30-year average value of all models

Although this RCP has the highest CO_2 increase, the physiological effect of temperature increase is most dominant in this scenario, contributing to 85% of the reduction for rainfed maize yield. As we show, increasing temperature is the most contributing factor which reduces maize potential yield in the Gambella region with a prevailing tropical climate meaning already high temperature levels but also a high precipitation amounts. Thereby, increasing temperatures not only reduces the efficiency of photosynthesis but also results in a faster development of maize crop which leads to a shorter life cycle resulting in smaller plants, to a shorter reproductive duration (grain filling period), and finally to lower yields. These effects in summary result in less water demand of the crop and thus reduced water stress, although precipitation declines between 9.3 and 15.8%. Given a temperature increase in Gambella by up to 6 K that results in maize yield decreases by approx. 30%, our results support findings from other studies that show a reduction of yields by 5% per degree warming (Challinor et al. 2014; Hatfield et al. 2011). Our results for CO_2 responses are also in line with measurements from the Free Air CO_2 Enrichment (FACE) experiments that show positive CO_2 responses for maize by approximately 8% for a 550-ppm CO_2 enrichment and irrigated conditions (Deryng et al. 2016) corresponding approximately to the atmospheric CO_2 concentration under RCP 4.5, where our results show a positive CO_2 response by 7.4%.

Although temperature increase and shortening growing periods are supposed to affect increasing incidence of diseases, pests, and weeds outbreaks in Africa (Adhikari et al. 2015), it is important to note that the simulated yield decrease rates refer to potential yield, assuming perfect crop management. Despite the expected reductions of maize potential yield, intensification potentials still remain and are a major strategy for possible future production increase in the region, due to high yield gaps (Degife et al. 2019). However, our results also show that in the course of intensification, irrigation as a strategy for intensification will lose effectiveness, due to the increasing dominance of temperature stress.

Generally, our results refer to maize that provides a substantial portion of daily caloric intake in the Gambella region. Other crops may be affected differently by climate change impacts, depending on different crop physiologies and different crop management.

Another expected impact of future climate change is an increased occurrence of extreme weather events (Gezie 2019). While other studies showed that mean climate change may lead to asymmetrical responses in the frequency and intensity of severe weather events that can cause large-scale droughts, flooding, or severe reduction of crop yields in the study area (Milman and Arsano 2014; Regan et al. 2019; Wakuma Abaya et al. 2009), we did not consider single weather events in this study. As rainfall becomes more

variable, farmers may no longer be able to rely on their knowledge of the seasonality of climatic variables. Shifting planting seasons and weather patterns will make it harder for farmers to maintain trust in planning and managing yield production (Akpodioyaga-a and Odjugo 2010; Lipper et al. 2014).

Adaptation as possible strategy to reduce climate change impacts is not considered in this study. An often discussed option is changing land-use patterns suggesting, e.g., relocating cropland to higher altitudes if available. However, relocation of cropland often is associated with negative impacts on ecosystems (Zabel et al. 2019) and higher altitudes are often not well suitable for agriculture (e.g., due to high slope). Shifting sowing dates and using adapted crop varieties or different cultivars are other possible measures for adaptation. Thereby, adapted crop varieties could prolong the shortened growing periods by increasing the heat units required to reach maturity. As shown by Minoli et al. (2019), the use of such adapted varieties potentially has a high impact on crop yields and can globally compensate warming up to 2 K. By doing so, irrigation and available water from precipitation could again become an important factor for intensification and possible adaptation in the Gambella region, since it is likely that adapted varieties require more water when they grow longer. For further studies, the effect of adaptive growing seasons by assuming adapted varieties in addition to shifting sowing dates should be investigated both for rainfed and irrigated conditions.

The main challenge may be on how to incorporate available knowledge and technology for possible adaptation in the Gambella region in a process in which small-scale farmers are involved.

Conclusion

This study analyzes climate change impacts on the local scale and therefore is beneficial for local policy and decision-makers and therefore allows for developing strategies for a sustainable agricultural development within the Gambella region under a range of possible future climate conditions. Maize is predominantly grown by smallholder farmers in the region, who mostly cultivate small parcels of land, which are often degraded. Climate change adds further challenges to the existing problems and undermines efforts that are being made to enhance food security in the region.

The strong heat response in the Gambella regions is mainly responsible for yield losses by more than 30% in case of rainfed and up to 50% in case of irrigation until the end of the century. Thereby, temperature increase is becoming the dominating effect over time and with higher RCP scenario, resulting in a decreasing role of water stress in context of yield reduction. Consequently, irrigation will lose effectiveness for increasing yields when temperature becomes the limiting

factor. Providing new varieties from breeding could be a great benefit for the region. Thereby, our results suggest that adjusting the sowing date to minimize the impact of heat stress, as well as using late-maturing cultivars and new varieties are more effective and primarily required under future climate conditions for the Gambella region.

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