Dear Prof. Dr. Claussen, dear reviewers,

Thank you very much for your assessment and the constructive comments to improve our manuscript. Our response is detailed below, as well as being reflected in the revised manuscript. We hope that we have addressed all concerns and questions that were raised.

With best regards,

Minchao Wu (on behalf of all authors)

Comments from reviewer #1:

This paper presents analysis of simulations with a regional climate model constrained by an earth system model (ESM) and coupled with a dynamic vegetation model (DVM). In the 21st c. simulation forced with the RCP8.5 scenario, including vegetation feedback led to drying in central Africa.

(1) A clarification on the model setup is needed. The authors mention (lines 478-481) that "SSTs were prescribed from CanESM2, therefore the land-ocean thermal contrast … originated solely from the changes … induced by vegetation dynamics". As far as I can see, this is the last mention of the prescribed SSTs in the paper, but exactly what are these prescribed SSTs? Climatology? For what period? Given the known sensitive of the West and central African climate to SSTs, this needs to be explained carefully and the SSTs prescribed need to be evaluated. A coupled model such as the CanESM2 is not necessarily producing correct SSTs for the observational period. Of particular concern for the region of the analysis is the seasonal formation of the Atlantic cold tongue which, I believe, generally fails to form in coupled GCMs.

Response: The SSTs (section 2.2) are from the CanESM2 simulations and they are applied in the same manner as other boundary conditions, i.e., from the time-evolving GCM simulation. They are not a climatology (abbreviation "SST" has been added in lines 172 and 176 in the revised manuscript). Both the FB and the NFB simulation use the same set of SST forcing.

The evaluation of CanESM2 SSTs in the oceans around Africa has been done in previous studies. Rowell (2013) indicated an acceptable agreement of CanESM2 SSTs with observations for their African teleconnection study; Xu et al. (2014) suggested relatively small SSTs biases from CanESM2 among CMIP5 models over the southeastern tropical Atlantic; LaRow et al. (2014) showed that SSTs of CanESM2 over the tropical oceans agree well with the reconstructed SSTs (derived from surface marine observational records).

We has summarised the findings from these studies in the section 3.1 of the revised manuscript (lines 265-267), and relevant references has been added, the revised texts are as "The SST forcing is also important for the African climate, and the CanESM2 SSTs have been validated and shown to be accurate in previous studies (e.g. Rowell, 2013;LaRow et al., 2014;Xu et al., 2014)".

(2)The biases in the regional model are significant (Fig. 1). The dry bias in the Congo Basin in the regional model (Fig. 1. b2), while common in models, seems extreme but it is similar to the dry bias in the ESM (Fig. 1.b3). It is important to consider how these biases influence the results, especially since one of the big results is additional drying in central Africa.

Response: The figure title "Model (CanESM2)" in the original version of Figure 1, referred to by the reviewer, refers to the NFB simulation with the RCM forced by CanESM2 boundary conditions, and not to the global CanESM2 model. We have clarified this by changing the title and figure text in Fig. 1.

On the other hand, there may be an issue for CRU dataset conversion applied in our post-processing for model evaluation. We have now used the consistent conversion method and found that this can reduce model bias for Sahel and central Africa, and reduce observation uncertainty, more information can be found in comments #4. The dry bias is similar to other RCMs (Nikulin et al., 2012;Kim et al., 2014), and very likely can be traced back to the convective scheme, rather than the circulation simulated by the physical sub-model. Following the reviewer's suggestion, we have also evaluated the low-level circulation and humidity in the CanESM-forced RCM simulation (the new figure Fig. A1), and we found that the dry bias over central Africa and wet bias over Sahelian savannah are not primarily related to the bias in circulation. In contrast, the model has done a relatively good job in reproducing the overall circulation patterns, including the southern and northern trade wind over the Atlantic ocean (the new figure Fig. A1) and the Walker circulation (Fig. 6), which is important for this study.

We have given further explanations of this issue in section 3.1 & 4.3 (lines 247 – 277, lines 512 – 523 in the revised manuscript), and relevant references has been added accordingly.

(3)The wet bias in the Sahel in the regional model is unusual – many models fail to bring rainfall into the Sahel, as is the case for the ESM that is providing boundary conditions for the regional simulation (Fig. 1.b3). Is it relevant to the results that the regional model over-produces rainfall primarily in the spring?

Response: Yes, as shown by Fig.1b3, this could relate to the early onset of the rainy season. The issue is not unique to RCA and is common among RCMs (Kim et al., 2014). This can relate to the bias of the simulated West Africa Monsoon (WAM) dynamics, one possible explanation can be the biased interaction between deep convection and the Africa Easterly Waves (AEW). The propagation of AEW, which brings moisture to the Sahel regions, is dependent on the strength of deep convection: a strong deep convection can usually spread moisture at higher vertical atmospheric level, and cause rainfall over a wider latitudinal band along the ITCZ, whereas a weaker deep convection can result in a narrower but more concentrated precipitation band (Sylla et al., 2011). RCMs' sensitivity to the intensity of WAM can explain their different precipitation pattern over Sahel (Gbobaniyi et al., 2014). For the precipitation over central Africa, however, precipitation is primarily driven by orographic uplifting and low-level convergence, and it is maintained by low-level mass convergence over the ITCZ (Sylla et al., 2011) and the

Walker circulation (Nicholson and Grist, 2003;Cook and Vizy, 2015). Therefore, the influences from such bias on the dynamics in our study should be limited.

We have given further explanations of this issue to section 3.1 (lines 247-253 in the revised manuscript) as: “the dry bias for annual mean precipitation over central Africa may be partly due to the underestimated daily precipitation during the late afternoon and night in addition to observational uncertainties (Nikulin et al., 2012). The wet bias over the northern savannah is mainly caused by a too early onset of the rainy season (b1, Fig. 2), which is possibly caused by the interactions between the simulated deep convection and the Africa Easterly Waves (Sylla et al., 2011)”.

(4) I am puzzled by the large differences between the GPCP and CRU precipitation observations shown in the Sahel (“northern savanna”) Fig. 2.b1. I think this is more related to the choice of averaging region than to a disparity in the observations, given the difference in the resolution of these 2 data sets. Please check this.

Response: Thanks for spotting this for us, this may relate to the inconsistent data conversion methods applied in our data post-processing for CRU dataset and GPCP in this study, have now corrected this with the latest CRU dataset TS3.23, we found that model bias and observation uncertainty have reduced. This has been updated accordingly in section 3.1 (line 220, lines 244-245 in the revised manuscript, Fig 1&2 are now updated with the new CRU dataset).

(5)References to the Charney (1975) and related studies are problematic since the idea that vegetation changes (i.e., “over-grazing”) caused the precipitation decline in West Africa during the 1960’s and 70’s has been thoroughly refuted in the mode modern literature. It’s SSTs forcing, of course.

Response: We agree that the role of SSTs is central, but also want to acknowledge Charney’s paper which was seminal in hypothesising the potential impacts of vegetation changes on the monsoon circulation, which has formed the basis for many other vegetation change-related studies, not least those related to long-term vegetation changes in this region. Our study investigates how vegetation changes can lead to feedback in this region, changes in albedo, land surface temperature, thus land ocean contrast is relevant to this hypothesis.

Here we would like to rephrase in lines 103-106 in the revised manuscript to “Hypothesised mechanisms of vegetation-atmosphere coupling include modulations of the surface albedo (Charney, 1975), changes in the North-African monsoon system (Claussen, 1997) and internal climate variability (Zeng et al., 1999)”.

(6)The authors note (lines 240-242) that “The simulated patterns and magnitude of precipitation for this area are similar to a previous study using an earlier version of RCA, RCA3.5, without dynamic vegetation”. So doesn’t that mean that dynamic vegetation is not influential, in contrast to the findings of this paper?

Response: The comparison to RCA3.5 in previous study refers to the simulated present-day climate. For the present-day period, influences from vegetation dynamics are limited as the present-day land cover types in terms of forest cover and open land are able to reproduce though given the bias in LAI, and large-scale vegetation changes rarely happen over the short period of comparison. However, for the century-long transition period under climate change considered in this paper, changes in climate and CO_2 forcing are strong enough, and lag effects of vegetation response short enough, to induce large-scale and long-term vegetation change and its feedback effect on climate is found to be much stronger than that seen during the present-day period.

(7)I would appreciate seeing an evaluation (e.g., a comparison with the ERAI reanalysis) of the circulation and specific humidity at 850 hPa wind and specific humidity from the present day, NFB simulation since the authors are pointing to changes in the circulation/moisture advection as relevant. This seems more crucial than evaluating LAI, for example.

Response: Thank you for this suggestion. A new figure has been added (as Fig. A1 in the revised manuscript) in the Appendix and new text has been added in the section 3.1 (lines 263-277).

(8)There's not a lot of literature on the dynamics of the Walker circulation in this region and its sensitivity to SSTs (and/or land/sea contrast), but these recent papers will help:
Pokam WM, Djiotang LAT, Mkankam FK, 2012: Atmospheric water vapor transport and recycling in equatorial central Africa through NCEP/NCAR reanalysis data. Climate Dyn. 38, 1715-1729.
Pokam MW, Bain CL, Chadwick RS, Graham R, Sonwa DJ, Kamga FM, 2014: Identification of processes driving low-level westerlies in West Equatorial Africa. J. Climate 27, 4245-4262.
Cook, K. H., and E. K. Vizy, 2015: The Congo Basin Walker Circulation: Dynamics and Connections to Precipitation, Climate Dynamics, DOI 10,1007/s00382-015-2864-y.

Response: Thank you for pointing us to these studies, new references has been added to the section 3.3 (line 349) and section 3.4 (line 378).

(9)A couple of minor points:
Please note that "Savannah" is the city in Georgia, U.S., while "savanna" is the grassland.
Response: The Oxford English Dictionary gives "savannah" as the preferred spelling.

(10)Figure A1 caption needs to be improved to provide more detail about what is plotted.
Response: Agreed. The revised caption now appears as "Fig. A2. Simulated seasonal surface temperature for present day (a-d), for changes in future in the NFB experiment (e-h, future minus present day), and for changes from vegetation feedback in future (i-l, FB minus NFB for future). Definitions for calculation period, climate change signal and vegetation feedbacks are given in Sect. 2.2."

Comments from reviewer #2:

This paper presents a future prediction study on climate-vegetation interactions in Africa. While the concept is not new, it does add to an emerging body of literature on interactive vegetation-climate predictions and will be of interest to many readers of ESD. The paper potentially merits publication, but quite a few major issues need to be addressed:

(1) Introduction: The flow of thought is very hard to follow. Part of the reason has to do with a rather liberal use of terminology. Probably a more strict use of the words "change" "variability" "pattern" "feedback" will help. The way it is now, many sentences are either vague or not accurate, which does not serve the readers well. Needs a better organization.

Response: We have revised the introduction and have attempted to use a more strict terminology throughout the article.

(2) Introduction: An important body of literature (e.g., Claussen 1997 climate dynamics, Claussen 1998 global change biology; Zeng et al., 1999 science; Alo & Wang, 2010 climate dyamics; Yu et al., 2015 climate dynamics) on vegetation-climate interactions is missing, although some of them are later mentioned in the Discussion section. The introduction part of a paper should be the place where the status of science is conveyed and gaps identified. Otherwise it will be misleading for readers who are new to the topic.

Response: We have added new texts to the introduction (lines 101-119) where we have cited most of these highly relevant papers and highlight the specific gaps that our paper is attempting to address.

(3) Partly related to (2), the statement in lines 111-112 is misleading. The first several sentences in section 4.1 should be moved here to provide readers an accurate description of the status of science, and the authors need to further elaborate to explain why this study adds values to existing literature.

Response: This has been addressed, see response to previous comment. The revised statements are as : "Recent studies have used a regional climate model to investigate the impact of climate-vegetation interaction for West Africa, identifying significant vegetation feedback in modulating local hydrological cycling (e.g. Alo and Wang, 2010;Wang and Alo, 2012;Yu et al., 2015). Additionally, a number of GCM-based studies have investigated the climate effects of anthropogenic perturbations, such as deforestation or afforestation (e.g. Lawrence and Vandecar, 2015). Such studies point to potentially significant forcing of regional climate dynamics, particularly rainfall patterns, as a result of changes in land cover. No study to date has, however, characterised the coupled dynamics of vegetation and climate under future radiative forcing for the entire African domain at a grid resolution high enough to capture regional features and forcings."

(4) Fig.1: The color scale is very difficult to read if one were to try to figure out the actual magnitude of the model biases. Should use more distinguishable color scales/ use stronger contrast between the colors.

Response: This has been addressed, thanks for the suggestion. Fig 1 is updated accordingly.

(5) Fig.1 and 2 showed severe bias of the model in capturing the spatial pattern of precipitation distribution and vegetation distribution. Essentially, LAI has negligible difference between the Sahelian savannan and the central Africa forest. The discussion and statement about model performance in Section 3.1 significantly downplayed the severity of this model biases.

Response: We agree that these biases are significant. First of all, we have corrected a mistake in data conversion for our post-processing for CRU dataset, in response to comments #4 of reviewer 1. Fig 1&2 are now updated accordingly. We found that model bias as well as observation uncertainty have reduced.

We have added a substantial discussion of the bias in precipitation pattern and LAI in Section 3.1, including acknowledgement for the LAI bias in lines 278-281 "A systematic overestimation is apparent for savannahs, and a significant underestimation for the central Africa rainforest area. These biases in LAI predominantly reflect the corresponding biases in precipitation (**Error! Reference source not found.** b1-b3 and 2c1-c3)" and more explanation for the precipitation bias in lines 247 – 253 "In RCA, the dry bias for annual mean precipitation over central Africa may be partly due to the underestimated daily precipitation during the late afternoon and night in addition to observational uncertainties (Nikulin et al., 2012). The wet bias over the northern savannah is mainly caused by a too early onset of the rainy season (b1, Fig. 2), which is possibly caused by the interactions between the simulated deep convection and the Africa Easterly Waves (Sylla et al., 2011)"

Additionally, in response to Comment (7) from Reviewer #1 which is also relevant to this comments, we have added a new figure (Fig A1) and related discussion evaluating low-level circulation and humidity. We found that the dry bias over central Africa and wet bias over Sahelian savannah are not primarily related to the biases in the circulation, but are more likely to be related to problems with the convection scheme in the regional model (Nikulin et al. 2012). In contrast, the model has done a relatively good job in reproducing overall circulation patterns, including the southern and northern trade wind over Atlantic oceans (Fig. A1), Walker circulation (Fig. 6), which are important for this study. This has been included in section 3.1 (lines 263 - 277).

(6) The model biases in precipitation and more importantly in vegetation could significantly influence the location and magnitude of the difference between FB and NFB, and need to be discussed explicitly.

Response: For the vegetation dynamics, the bias in simulated present-day vegetation is largely related to a bias in precipitation. We agree that this potentially influences the simulated difference between FB and NFB, as the latter uses the resultant, biased vegetation as forcing. This may lead to an offset in the locations of the strongest impact of vegetation feedbacks in the model, but we assume that this bias does not critically affect qualitative aspects of the feedbacks that we find. One reason for such confidence is that the bias is small in magnitude compared to the size of the simulated future changes in LAI and precipitation. We have add more discussion in section 4.2.

The influence of the bias on model's dynamics is explained in point (8) further below.

(7) Lines 315-320: The albedo difference is negligible? One would think that albedo changes can be significant in areas with increase of vegetation cover.

Response: We agree that albedo changes play a role for surface temperature changes, this was ill-phrased in the original manuscript. We have identified warming effects from albedo change, which gives an overall warming effect in northern hemisphere winter on the edge of the area of forest expansion in the northern savannah region (Fig. A2). An increase in vegetation (forest) cover gives both an albedo (warming) effect and an evaporative (cooling) effect, with the combined effect depending on their seasonal balance. In general, modelling studies tend to show that evaporative cooling effects are more dominant in the tropics while albedo warming effects are more dominant over high-latitude regions (e.g. Claussen et al., 2001;Bala et al., 2007).

The missing explanation for the albedo effect has been added to section 3.3 as "Overall, the turbulent heat fluxes increase, which tends to cool the surface and the lower atmosphere, exceeding the opposing (warming) effects of increased vegetation cover on albedo, thus resulting in an overall cooling effect. Similar behaviour was seen in southern Europe in a previous study with RCA-GUESS (Wramneby et al., 2010).".

(8) Lines 448-453: This is not true. The state of the vegetation is very important in determining the interannual variability of vegetation and the vegetation feedback effects. This is why the issue of severe model bias needs to be acknowledged and its implication explicitly discussed, as suggested in comment 6).

Response: We agree with the reviewer on this point, we did not express our point well in the original manuscript. We intended to point out that bias in LAI within a given land cover type (forest, savannah or grassland) is likely to have a smaller impact on the simulated climate than an inaccurate distribution of land cover types. Although our simulations have evident bias in LAI and precipitation, overall patterns of vegetation distribution across Africa are comparable to observations.

This has been further explained in section 4.3 (lines 512-523) as "Despite biases in the initial precipitation and vegetation state (LAI) for some regions, our model was able to reproduce the present-day land cover type, and the simulated present-date climate is close to previous study (Nikulin et al., 2012) using the same physical sub-model with observed land cover type. Under future climate change, vegetation-induced changes in circulation, thus a substantial change in moisture transport and precipitation, are mainly triggered by changes in land cover type (Fig. 4a), therefore, we argue that the influences from biases in initial conditions on such mechanism found in this study should be limited. Our study used prescribed SST forcing from a GCM and could thus not account for additional or opposing feedbacks mediated by ocean dynamics. However, as the ocean heat capacity is relatively large and variation in land-ocean thermal contrast can be greatly buffered by ocean heat uptake (Lambert and Chiang, 2007), we suggest that results should not change fundamentally if a dynamic ocean component was introduced to the model."

Minor comments: Lines 92-94: "... are important to … " is rather awkward. You mean "… are important determining factors for …" ? Fig. A4: "temperature gradient" should be changed to "temperature contrast" as y-label.

Response: changed as suggested, thank you for the suggestions.

Reference:

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Vegetation-climate feedbacks modulate rainfall patterns in Africa under future climate change

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Abstract

Africa has been undergoing significant changes in climate patterns and vegetation in recent decades, and continued changes may be expected over this century. Vegetation cover and composition impose important influences on the regional climate in Africa. Climate change-driven changes in regional vegetation patternsstructure and the distribution of forests versus savannah and grassland may feed back to climate via shifts in the surface energy balance, hydrological cycle and resultant effects on surface pressure patterns and larger-scale atmospheric circulation. We used a regional Earth system model incorporating interactive vegetation-atmosphere coupling to investigate the potential role of vegetation-mediated biophysical feedbacks on climate dynamics in Africa in an RCP8.5-based future climate scenario. The model was applied at high resolution (0.44 x 0.44 degrees) for the CORDEX-Africa domain with boundary conditions from the CanESM2 GCM. We found that changes in vegetation patternsincreased tree cover and leaf-area index (LAI) associated with a CO_2 and climate-driven increase in net primary productivity, particularly over sub-tropical savannah areas, not only imposed important local effect on the regional climate by altering surface energy fluxes, but also resulted in remote effects over central Africa by modulating the land-ocean temperature contrast, Atlantic Walker circulation and moisture inflow feeding the central African tropical rainforest region with precipitation. The vegetation-mediated feedbacks were in general negative with respect to temperature, dampening the warming trend simulated in the absence of feedbacks, and positive with respect to precipitation, enhancing rainfall reduction over the rainforest areas. Our results highlight the importance of accounting for vegetation-atmosphere interactions in climate projections for tropical and sub-tropical Africa.

Keywords: RCA-GUESS, Vegetation Dynamics, Biophysical feedback, Precipitation, Walker Circulation, Land-ocean Contrast, Regional Climate Model

1. Introduction

The Sahel greening and Congo rainforest browning observed since the 1980s suggest that Africa has been undergoing significant ~~vegetation~~ changes in the structure ~~and~~, composition and distribution of vegetation during ~~the~~ recent decades (~~Zhou et al., 2014;~~Eklundh and Olsson, 2003;Olsson et al., 2005;Jamali et al., 2014;Zhou et al., 2014). In addition to influences from anthropogenic activity (e.g. changes in land use), vegetation ~~shifts~~changes in the region have been linked to changes in recorded climatic conditions, including the trend and interannual variability of precipitation (Herrmann et al., 2005;Hickler et al., 2005;Olsson et al., 2005;Zhou et al., 2014~~;Hickler et al., 2005~~), which in turn have been related to decadal-scale changes in regional circulation (Camberlin et al., 2001;Giannini et al., 2003). On longer timescales, anthropogenic climate change has the potential to cause profound structural and compositional changes in vegetation over Africa (Sitch et al., 2008;Scheiter and Higgins, 2009).

Shifts in vegetation cover and composition~~, cover and seasonality~~ in terms of the distribution of trees and grasses and their seasonal changes (phenology) can ~~in turn~~ impose significant ~~feedbacks~~forcings on the physical climate system by ~~altering~~modulating surface-atmosphere energy exchange and hydrological cycling ~~(~~, resulting in biophysical feedbacks~~) as well as greenhouse gas concentrations and aerosol loads in the atmosphere (biogeochemical feedbacks).~~ along with the climate forcings. The type ~~and coverage~~ of vegetation alongside productivity-related structural aspects such as tree density and leaf area index (LAI) are important ~~to~~determinants for surface albedo, roughness length and evapotranspiration, affecting surface energy fluxes that in turn control lower boundary layer thermodynamics (Bonan, 2008Eltahir, 1996;Brovkin et al., 2006;Eltahir, 1996Bonan, 2008). Biophysical feedbacks operate locally and may also generate teleconnections via heat and moisture advection, leading to altered circulation patterns (e.g. Avissar and Werth, 2005;Nogherotto et al., 2013atmospheric circulation (e.g. Avissar and Werth, 2005;Nogherotto et al., 2013). Previous studies have shown the importance of vegetation-mediated biophysical feedbacks for the past (e.g. Claussen and Gayler, 1997;Texier et al., 1997), and present (e.g. Eltahir, 1996;Claussen, 1998;Wang and Eltahir, 2000) climate over Africa. Hypothesised mechanisms of vegetation-atmosphere coupling include modulations of the surface albedo (Charney, 1975), changes in the North-African monsoon system (Claussen, 1997) and internal climate variability (Zeng et al., 1999).

There is an increasing awareness of the significance of biophysical vegetation-climate interaction for Africa. Vegetation changes in the Sahel can alter local decadal-scale precipitation variability through changes in local energy and water fluxes and even through changes in atmospheric circulation (Charney, 1975;Xue and Shukla, 1996;Wang and Eltahir, 2000), while deforestation in the Congo basin increases surface albedo and weakens local moisture convection, resulting in decreased precipitation (Eltahir, 1996;Xue and Shukla, 1993;Bell et al., 2015;Nogherotto et al., 2013).

Anthropogenic climate change may lead to profound structural and compositional changes in the natural vegetation over Africa, especially for savannah areas where seasonal fluctuations in water availability and climate-mediated disturbance regimes (fires and grazing) serve to facilitate coexistence of trees and grasses in a fine competitive balance (Moncrieff et al., 2014;Sankaran et al., 2005;Doherty et al., 2010;Ahlström et al., 2015). Changed vegetation patterns may be expected to modulate the regional climate development. However, high-resolution studies of future vegetation-atmosphere coupling have not been performed earlier for Africa with a comprehensive approach.

Feedbacks mediated by shifts in vegetation structure and distribution can likewise play a role for the future regional climate. General circulation models (GCMs) have been applied at relatively coarse lateral grid resolutions to capture these dynamics (e.g. Kucharski et al., 2013). Recent studies have used a regional climate model to investigate the impact of climate-vegetation interaction for West Africa, identifying significant vegetation feedback in modulating local hydrological cycling (e.g. Alo and Wang, 2010;Wang and Alo, 2012;Yu et al., 2015). Additionally, a number of GCM-based studies have investigated the climate effects of anthropogenic perturbations, such as deforestation or afforestation (e.g. Lawrence and Vandecar, 2015). Such studies point to potentially significant forcing of regional climate dynamics, particularly rainfall patterns, as a result of changes in land cover. No study to date has, however, characterised the coupled dynamics of vegetation and climate under future radiative forcing for the entire African domain at a grid resolution high enough to capture regional features and forcings.

In this study, we employ a regional Earth system model (ESM) that couples the physical component of a regional climate model (RCM) with a detailed, individual-based dynamic vegetation model (DVM). This tool enables dynamic representation of biophysical interactions between the vegetated land surface and the atmosphere and their effects on the evolution of climate and land surface biophysical properties to be analysed in an explicit way. We perform simulations under the Representative Concentration Pathway (RCP) 8.5 ~~future~~ radiative forcing scenario (Moss et al., 2010) with and without vegetation feedbacks enabled, and investigate the potential coupled evolution of climate and vegetation ~~patterns~~ for the ~~CORDEX-Africa domain~~African continent over the 21st century. Our focus is especially on the central African rainforest areas and the surrounding savannah vegetation belt.

2. *Data and Method*

2.1 Model description

RCA-GUESS (Smith et al., 2011) is a regional ESM based on the Rossby Centre regional climate model RCA4 (Kjellström et al., 2005;Samuelsson et al., 2011) coupled with vegetation dynamics from the LPJ-GUESS DVM to account for land-atmosphere biophysical coupling (Smith et al., 2001;Smith et al., 2014).

The RCA4-based physical component of ~~the model~~RCA-GUESS incorporates advanced regional surface heterogeneity, such as complex topography and multi-level ~~presentations~~representations for forests and lakes, which are significant in controlling the development of weather events from the local- to meso-scale (Samuelsson et al., 2011). RCA4 has been ~~successfully~~ applied in a range of climate studies worldwide (e.g., Döscher et al., 2010;Kjellström et al., 2011;Sörensson and Menéndez, 2011~~;Kjellström et al., 2011;Döscher et al., 2010~~). The land surface scheme (LSS, Samuelsson et al., 2006) adopts a tile approach and characterizes land surface with open land and forest tiles with ~~separated~~separate energy ~~balance~~balances. The open land tile is divided into fractions for vegetation (herbaceous vegetation) and bare soil. The forest tile is vertically divided into three sub-levels (canopy, forest floor and soil). ~~Snow can exist in open land and/or forest tile as fractional cover.~~ Surface properties such as surface temperature, humidity and turbulent heat fluxes (latent and sensible heat fluxes) for different tiles in a grid box are weighted together to provide grid-averaged ~~surface boundary conditions.~~values. A detailed description is given by Samuelsson et al. (2006).

The vegetation dynamics component of RCA-GUESS employs a plant individual and patch-based representation of the vegetated landscape, optimized for studies at regional and global scale. Heterogeneities of vegetation structure and their effects on ecosystem functions such as carbon and water vapour exchange with the atmosphere are represented dynamically, affected by allometric growth of age-size classes of woody plant individuals, along with a grass understorey, and their interactions in competition for ~~limited~~ light and soil resources. Plant functional types (PFTs) encapsulate the differential functional responses of potentially-occurring species in terms of growth form, bioclimatic distribution, phenology, physiology and life-history characteristics. Multiple patches in each vegetated tile account for the effects of stochastic disturbances, establishment and mortality on local stand history (Smith et al., 2001). This explicit, dynamic representation of vertical structure and landscape heterogeneity of vegetation has been shown to result in realistic simulated vegetation dynamics in numerous studies using the offline LPJ-GUESS model (Smith et al., 2001;Weber et al., 2009;Hickler et al., 2012;Smith et al., 2014;Wårlind et al., 2014;Wu et al., 2015;Weber et al., 2009). Biophysical feedbacks have previously been studied in applications of RCA-GUESS to Europe and the Arctic (Wramneby et al., 2010;Smith et al., 2011;Zhang et al., 2014). A general description forof the coupling between the vegetation dynamics component LPJ-GUESS and the physical component RCA is provided in the Appendix. A more detailed description is given by Smith et al. (2011).

2.2 Model setup, experiments and analysis approach

The simulations were applied over the African domain of the Coordinated Regional Climate Downscaling Experiment (CORDEX-Africa, Giorgi et al., 2009;Jones et al., 2011) on a horizontal grid with a resolution of 0.44° × 0.44°. The period studied was 1961 to 2100. Forcing fields in 6-hour time intervals (atmospheric fields and sea-surface temperature (SST) as lateral and lower boundary conditions) followed, respectively) were derived from the historical and RCP8.5 simulations with the CanESM2 general circulation model (GCM) (Arora et al., 2011) in the Coupled Model Intercomparison Project Phase 5 (CMIP5, Taylor et al., 2012).2012). Time-evolving forcing fields from the GCM were prescribed for all variables, including SSTs.

The vegetation sub-model LPJ-GUESS was set up with eight global PFTs which represent the major groupselements of natural vegetation across Africa, including the tropical and warm-temperate forests and savannahs and C_3 and C_4 grassgrasslands. The characteristics for the PFTs are based on the study byPFT parameter settings follow Morales et al. (2007). They) and are summarised in Table A1.

PFTs inof the forest tile were simulated with 30 replicate patches. Average values of state variables across the replicate patches were used to determine biophysical parameters, i.e. forest fraction and leaf area index (LAI) for trees versus grasses, provided as forcing to the physical part of the sub-model. For the open land tile with herbaceous species, C_3 and C_4 grass were simulated deterministically and aggregated to characterise open land vegetation. Fire disturbance in response to climate and simulated fuel load (Thonicke et al., 2001) was included.

Following the approach of Wramneby et al. (2010) and Smith et al. (2011), RCA-GUESS was initialized with a spin-up in two stages to achieve a quasi-steady state representative for mid-1900's conditions. After the spin-up, the model was run in coupled mode from 1961 onwards, with simulated dynamic meteorological conditionsforcing from the physical sub-model affecting vegetation phenology and structural dynamics, and biophysical land surface properties being adjusted to reflect the changes in vegetation, thereby affecting the physical climate dynamics. For comparison, a recent past experiment (RP, Table 1) with the same vegetation spin-up but thenthereafter driven by the boundary conditionconditions derived from ECMWF re-analysis (ERA-Interim) project (Berrisford et al., 2009), was conducted for the period 1979-2011.

AThe simulation protocol was designed for inferringto enable biophysical feedbacks of vegetation changes to the evolving 21st century climate to be inferred. Three simulations were performed to investigate vegetation-climate feedbacks under future climate change (Table 1). The first simulation included the vegetation feedback (FB). It was run for 1961-2100 in coupled mode, allowing the effects of climate and atmospheric CO_2 concentration (the latter was taken directly from the RCP 8.5 data set) on vegetation state to feed back to the evolving climate. The second simulation was run without thewith vegetation feedback (“switched off” (non-feedback run, NFB). It started with the state of FB simulation at 1991 and used a prescribed climatology of daily vegetation for 1961-1990 from the coupled simulation, but without allowingtransferring the simulated changes in vegetation in LPJ-GUESS to feed back to the simulated climatethe land surface configuration, and associated biophysical surface properties, in RCA. To investigate the importanceLSS of RCA. To attribute the CO_2 component of the simulated vegetation changes resulting from physiological effects on vegetation changes under future climate changeof rising CO_2 concentrations of plant productivity and water-use efficiency, we performed a third simulation (FB_CC), which was similar to FB, but startingstarted from the state of the FB simulation of 1991 and usingused historical atmospheric CO_2 forcingconcentrations until 2005 and, held constant afterward forthereafter, to force the vegetation sub-model only.

In theOur analysis, we focus focuses on the future period 2081-2100 and compare, comparing this with the present-day (1991-2010). The climate change signal is inferred from the difference between the future mean and the present-day mean in the NFB run.

Vegetation feedbacks are calculated as the difference between the future means of the FB and NFB runs. These approaches were applied for the entire study, unless specified otherwise.

2.3 Methods to evaluate model performance

Simulated near-surface atmospheric temperature over open land, precipitation, and vegetation variable leaf area index (LAI)LAI were compared against observations within the common available time period 1997-2010. Temperature and precipitation were compared with gridded observations from the CRU TS3.2123 (Harris et al., 2014) dataset, focusing on the annual mean and seasonality. For precipitation we also employed the GPCP (Huffman et al., 2001, version 1.2 of One-Degree Daily product for 1996/10-2011/6) which uses satellite data to upscale rain gauge measurements and has been extensively used for African precipitation studies (e.g., Nikulin et al., 2012). For the LAI evaluation we used the GIMMS-AVHRR and MODIS-based LAI3g product (Zhu et al., 2013) which has been previously applied forto the evaluation of vegetation dynamics in ESMs (e.g., Anav et al., 2013).

To identify biases propagating from the model physics per se and from the GCM-derived boundary forcing data, we compared the reanalysis-driven RP simulation against observation and against the GCM-driven (CanESM2) FB simulation for the same period.

3. Results

3.1 Model evaluation

To evaluate the model's performance for the present day, the simulated annual mean and seasonality of 2-meter air temperature, precipitation and LAI are compared against the observations (Fig. 1 and Fig. 2). The simulated annual mean temperature (Fig. 1a1) is generally higher in northern-hemisphere (hereinafter "northern") Africa than in ~~the south~~southern-hemisphere (hereinafter "southern") Africa. The model generally shows a cold bias in the order of 1°C for northern and southern savannah (Fig. 1a2), dominated by the ~~boreal~~northern hemisphere summer ~~(~~(JJA, Fig. 2a1,2a3). Warm biases up to around 3°C occur ~~over~~in northern Africa, and warm biases up to around ~~3~~1°C~~, as well as~~ in central Africa ~~(around 2°C)~~ where the warm bias originates mainly from summer (Fig. 2a2).

The simulated precipitation is largest over western and central Africa up to 1600 mm year^{-1} within the simulated rainbelt between 25°N and 25°S, where the Atlantic moisture inflow (monsoon and equatorial westerlies) plays an important role (Fig. 1b1). Comparison with CRU reveals a considerable dry bias (-~~600~~500 mm year^{-1}) for the central African rainforest area and a wet bias (+~~300~~250 mm year^{-1}) for the northern savannah. ~~A comparison of the FB (CanESM2-driven) and the RP (ERA-Interim-driven) simulations (b3) indicates that apart from the uncertainty from RCM, the bias in simulated precipitation can be partly explained by the uncertainty from the boundary conditions.~~ The simulated patterns and magnitude of precipitation for this area are similar to a previous study using an earlier version of RCA, RCA3.5, without dynamic vegetation (Nikulin et al., 2012). ~~RCA3.5 was able to capture the main features of the seasonal mean rainfall distribution and its annual cycle, and the model biases were of~~

similar magnitude to the differences between observational datasets (Nikulin et al., 2012).

In RCA, the dry bias for annual mean precipitation over central Africa may be partly due to the underestimated daily precipitation during the late afternoon and night in addition to observational uncertainties (Nikulin et al., 2012). The wet bias over the northern savannah is mainly caused by a too early onset of the rainy season (b1, Fig. 2), which is possibly caused by the interactions between the simulated deep convection and the Africa Easterly Waves (Sylla et al., 2011). The biases in simulated precipitation for the savannah regions and the central African rainforest area mirror the temperature biases: warm biases coincide with dry biases in central Africa, and cold biases coincide with wet biases in savannah regions. A comparison of the CanESM2-driven (FB run) and the ERA-Interim-driven (RP run) simulations (Fig. 1Apart from the model uncertainty, observation uncertainty may contribute to the biases, which can be seen when compared with GPCP (b1, 2b2): For the northern savannah, CRU tends to present lower precipitation than GPCP and the modelled throughout the year. For the central African rainforest area, precipitation from CRU is considerably higher in the mid-year dry season, but lower for the rest of the year with much more moderate monthly precipitation variability than in GPCP and the modelled. In general, the simulated precipitation presents a better consistency with GPCP than with CRU, although it is difficult to evaluate the uncertainties between these two observational datasets.b3) indicates that the bias in simulated precipitation has contributions both from the RCM itself and from the GCM-generated boundary conditions. Nevertheless, Nikulin et al. (2012) have previously shown for Africa that the model is able to capture the ITCZ position and the main features of the seasonal mean rainfall distribution and its annual cycle, and the model biases in precipitation were of similar magnitude to the differences between observational datasets.

To further diagnose the effect of model dynamics on the precipitation bias, we evaluated the low-level circulation and humidity, which play an important role in the moisture transport between ocean and land(Nicholson and Grist, 2003). The SST forcing is also important for the African climate, and the CanESM2 SSTs have been validated and shown to be accurate in previous studies (e.g. Rowell, 2013;LaRow et al., 2014;Xu et al., 2014). We compare the simulated circulation and specific humidity at 850 hPa from the NFB run with the regional model against ERA-Interim reanalysis for 1997-2010 (Fig. A1). The simulated patterns of circulation and specific humidity at 850 hPa agree well with the reanalysis: the trade winds over both northern and southern Atlantic, West African monsoon as well as the Somali Jet (eastern Africa) are reproduced well by the model. However, there are small biases in wind speed at 850 hPa which generally appear in areas close to the domain boundary and around the African coastal regions. In the case of specific humidity, there are dry biases over the continent. These may be traceable to the different convective schemes used in RCA and ERA-Interim, exhibiting different diurnal cycle of precipitation over Africa (Nikulin et al., 2012).

The simulated seasonality of LAI generally reflects the simulated seasonality of precipitation. A systematic overestimation is apparent for savannahs, and ~~an~~a significant underestimation for the central Africa rainforest area. These biases in LAI predominantly reflect the corresponding biases in precipitation (Fig. 2 b1-b3 and 2c1-c3). A stronger LAI bias in the savannah ~~can be explained by~~is due to the presence of grasses, which are more sensitive to precipitation changes in the model compared to trees.

With present-day forcing, the simulated climate and vegetation patterns and phenology are generally consistent with ~~observed patterns.~~observations. Some of the biases in the simulated climate are common to ~~most~~many RCMs (Nikulin et al., 2012) and they are apparent for some sub-regions and seasons in our model. ~~However, we consider~~We conclude that the performance is adequate to capture the main details of the African climatology, ~~which provides~~providing sufficient confidence for the subsequent analysis of regional vegetation-climate interactions under future climate change.

3.2 Future climate and vegetation change

In the NFB simulation~~ without vegetation feedbacks (NFB) under the RCP8.5 scenario and CanESM2 boundary conditions~~, most of the African continent is simulated to be 4-6°C warmer by the end of the 21st century compared with present day (Fig. 3a). The subtropics exhibit a slightly stronger warming than the tropics, and land warming is slightly larger ~~than~~compared to warming of the surrounding ocean surface ~~(note that ~~as simulated by the ~~SSTs were~~CanESM2 GCM and represented in the SST forcing fields prescribed from ~~the GCM).~~that model. These changes are fairly similar throughout the year, except in Northern Africa and the Sahara, where the temperature increase is particularly pronounced in the local dry season (Fig. A2.e-h). Precipitation is projected to increase in most parts of the African monsoon area, western equatorial coastal area and the eastern African horn (Fig. A3.e-h). A slight decrease is projected in the Congo basin and for the southern part of the continent (Fig. 3c). For areas with a precipitation increase, the increase is mainly confined to the local wet season. The precipitation decrease over central and southern Africa is apparent throughout the year (Fig. A3.e-h).

Vegetation feedbacks (FB run) modify significantly the pattern and magnitude of simulated climate change. The effects are largest in low-latitude areas where the surface temperature increase is generally dampened (negative feedback), most notably in savannah areas and to a lesser extent in the equatorial rainforest area (Fig. 3b). The precipitation decrease is enhanced (positive feedback), most notably over the rainforest area (Fig. 3d).

With the effects of climate change and CO_2 fertilization, future vegetation growth depicts an enhancement not only of vegetation productivity in general, but also of tree cover in subtropical savannah areas (Fig. 4a), displacing grasses and ~~resulting in~~reflecting an increase in tree LAI of 0.5-2.4 during the growing season (Fig. 4b). This increase in tree cover reflects a general rise in vegetation productivity driven by rising atmospheric CO_2 ~~concentration~~concentrations on photosynthesis and water-use efficiency (Long, 1991;Hickler et al., 2008;Keenan et al., 2013). Results from the FB_CC experiment in which CO_2 fertilisation ~~is~~was disabled reveal that changes in climate drivers alone are simulated to have minor or opposing effects on tree productivity and LAI due to reduced water availability (Fig. A4.), and that the changes ~~above~~seen in tree cover and LAI in the

FB run hence originate primarily from CO_2 fertilization.

Temperature feedbacks tend to be strong in ~~the newly afforested~~ areas of increased tree cover (Fig. 3b, Fig. 4a). The cooling effects from vegetation feedbacks are strong (approximately -2°C) throughout the year, with the most pronounced cooling occurring in the local dry season (Fig. A2.i-l), when the newly established ~~forest~~tree (with larger root depth than grass) transpires water that is taken up from the deeper soil layer. Transpiration from present-day grass is constrained by the low moisture levels in the top soil layer. As a result, the evaporative cooling effect becomes stronger when forest replaces open land. In the central African rainforest area ~~with~~, where an increase ~~of~~in LAI ~~by~~of about 0.5-1 is simulated in FB run compared with the NFB run, vegetation feedbacks on temperature are much smaller in the rainy season, but cause ~~considerable~~ cooling ~~effect for~~in the dry season.

Vegetation feedbacks on precipitation are also pronounced. For the southern hemisphere savannah area, a slight increase in precipitation (approximately 10%, Fig. 3d) was simulated, which is caused by strengthened convective activity (which coincides with enhanced radiation and latent heat fluxes) in the rainy season (DFJ, Fig. A3.). This can be considered as a local effect ~~from forest expansion.~~of tree LAI increase. However, changes in precipitation are not restricted only to the areas where ~~forests expand~~tree cover increases (Fig. 3d, Fig. 4a), which is suggestive of remote effects ~~for~~on tropical precipitation. This is further investigated in the sections below.

3.3 *Vegetation feedback effects on circulation and precipitation*

Vegetation feedbacks on temperature in our simulations operate mainly via an increased surface area for evaporation and a stronger coupling to the atmosphere as tree cover, root depth and LAI increase relative to grasses, most notably in savannah areas, resulting in a shift of the evaporative fraction (ratio of latent heat flux to turbulent heat fluxes) and an increase in surface roughness length. Overall, the turbulent heat fluxes increase, which tends to cool the surface and the lower atmosphere., exceeding the opposing (warming) effects of increased vegetation cover on albedo, thus resulting in an overall cooling effect. Similar behaviour was seen in southern Europe in a feedbackprevious study with RCA-GUESS (Wramneby et al., 2010).

The variability of precipitation over Africa is greatly influenced by the moisture advection from the ocean to land. Previous studies have noted on the influence of Atlantic Walker circulation on central African precipitation, as well as the role of the west African monsoon (WAM) for the precipitation forover western Africa (e.g. Nicholson and Grist, 2003;Dezfuli and Nicholson and Dezfuli, 2013;Dezfuli and Nicholson, 2013Pokam et al., 2014). These circulation systems are associated with thermal contrasts between ocean and land, creating a pressure contrast that tends to promote the movement of moist surface air from the Atlantic over land. We examined the land-ocean thermal contrast (∇T) and geopotential contrast (∇ϕ) between the equatorial Atlantic and the near-coast African continent for three pressure levels between 850 hPa and 975 hPa, to

~~investigate~~characterise the circulation in the lower troposphere. We found that changes in ∇T and $\nabla\phi$ are highly inter-annually anti-correlated for the rainy seasons MAM and SON (r=-0.82 and -0.64, respectively, Fig. 5; Fig. A5). The sensitivity of $\nabla\phi$ to ∇T, depicted as the slope in Fig. 5, is generally maintained in the future, with a slight decrease in the sensitivity for DJF and a slight increase for MAM.

Under the NFB future simulation, ocean-land contrast becomes larger (the absolute value of ∇T increases by about 0.5-1°C, Table A2) as land temperature increases more than the GCM-simulated increase in SSTs provided as forcing to the regional model (Fig. A2.~~ocean surface temperature (), due to~~). Differential changes in features of the surface and lower atmosphere, such as changes in land-ocean contrasts in boundary layer lapse rate (Joshi et al., 2008) and changes in Bowen ratio over land (Sutton et al., 2007~~).~~) explain such divergence in temperatures between ocean and land. As a result, except for SON, $\nabla\phi$ is generally simulated to increase in the course of the simulation (Fig. A5), with the largest shift occurring in MAM (11.96 $m^2 s^{-2}$ by the end of 21st century, Table A2). For SON, ∇T increases but $\nabla\phi$ does not, suggesting that the trend of $\nabla\phi$ under climate change is associated with the GCM-derived boundary conditions, despite the strong regional coupling with ∇T in terms of variability (Fig. A5).

In contrast, the increase in the ∇T is dampened considerably when incorporating interactive vegetation. The resulting reduction in ∇T offsets $\nabla\phi$ uniformly and statistically significantly for all seasons, generally counteracting the climate change effect on $\nabla\phi$ (Fig. 5, Table A2).

3.4 Effects on Walker circulation and low-latitude precipitation

The low-level equatorial westerlies are important to the central African rainfall. They are associated with the lower branch of the Walker cell located near the western equatorial coast of Africa, and they transfer moisture from the adjacent Atlantic to the eastern equatorial coast and the Congo basin (e.g. ~~Schefuß et al., 2005;~~Nicholson and Grist, 2003;Schefuß et al., 2005;Cook and Vizy, 2015). These westerlies occur from March to October, being best developed in JJA. They shift northward with the excursion of the Inter Tropical Convergence Zone (ITCZ) and under the strong influence of the South Atlantic high pressure cell (Nicholson and Grist, 2003). This pattern is simulated by RCA-GUESS for the present-day climate (Fig. 6). Via this circulation system, moisture can reach far over the African landmass at around 28°E, upwell and integrate into the mid-level African Easterly Jet (AEJ) (Camberlin et al., 2001;Nicholson and Grist, 2003). RCA-GUESS reproduces this pattern with a realistic magnitude (Fig. 6, Fig. 7, Fig. 8, Fig. 9) when compared with previous studies based on reanalysis data (Camberlin et al., 2001;Nicholson and Grist, 2003).

In the NFB future simulation, equatorial westerlies are strengthened throughout the year both over ocean (Fig. 6) and over land (Fig. 7). Changes in wind speed (Δu) can be explained by changes in the low-level pressure contrast between land and ocean (sect. 3.3), where strengthened $\nabla\phi$ leads to enhanced u, especially for MAM when the zonal pressure contrast prevails (Table A2). Atmospheric specific humidity in the lower troposphere near the equator also increases by around 10%-20% for MAM and SON, extending from the ocean to inland along the equator (Fig. 8cd; Fig. 9cd). Meanwhile, changes in future rainfall are apparent along the equator, with increases over the equatorial coastal or inland areas (Fig. A3.), concurrent with stronger moisture inflow to land in the low-level troposphere (Fig. 8cd; Fig. 9cd).

Vegetation feedbacks are simulated to weaken the climate change enhancement of the Walker circulation, resulting in a weakening of the equatorial westerlies and counteracting the effects of climate change alone (Fig. 6i-l and Fig. 7i-l; Fig. 8ef and Fig. 9ef). These changes correspond well to changes in low-level ocean-land geopotential contrast $\Delta\nabla\phi$ with the biggest impact for MAM and SON (Table A2). The weakened Walker circulation is also represented as suppressed vertical uplifting motions over central Africa (Fig. 8f and Fig. 9f). Atmospheric specific humidity at 850 hPa is reduced by approximately 7% due to vegetation feedbacks which are comparable to the contribution of climate change (Fig. 8ef vs. Fig. 8cd; Fig. 9ef vs. Fig. 9cd).

Analysis of the moisture flux convergence also confirms the impacts of a weakened Walker circulation (Fig. 10) on the hydrological cycle caused by vegetation feedback. Moisture fluxes for most parts of the African continent diverge toward the ocean near the equatorial regions. This divergence is similar for both MAM and SON but the effect is slightly stronger for SON, which also corresponds to reduced humidity for these areas (Fig. 8e-f; Fig. 9e-f).

The changes in precipitation show a distinct spatial and temporal pattern with changes in the rainbelt area (defined as 2mm day^{-1} contour with 10-days smoothing, Fig. 11). Under future conditions, the rainbelt, which follows the ITCZ excursion, shifts around 3° northward during JAS (Fig. 11a). As a result, rainfall intensity increases from May to October, with the most pronounced increase by more than 30% relative to present-day levels of around 2 mm day^{-1} on the margins of the rainbelt. The rainy season becomes longer for Sahel (+9 days) as well as for central Africa (+1 day). The location of the rainbelt for the rest of the year remains unchanged, but there is a pronounced increase in rainfall intensity for the southern African rainy season (about 10%) and a decrease (about -10%) for the central African rainy seasons.

On top of the non-feedback climate change effect, vegetation feedbacks tend to cause a slight contraction of the rainbelt around the equator, and they impose a primarily counteractive effect on rainfall intensity compared to the climate change alone simulation (NFB). For central Africa, the considerable decrease in rainfall intensity in the dry season leads to a slight equatorward shrinking of the rainbelt (approximately 2°) and a shorter rainy season (on average 10 days, represented as a 4-day postponed onset and a 6-day earlier end). For southern Africa, strengthened convective precipitation results in a longer rainy season by on average 6 days. There is no pronounced effect for the Sahel regions except for some sparse changes over time and in some areas. To investigate the effects on ITCZ location, we analysed the position of the intertropical front (ITF) with a meridional wind criterion (Sultan and Janicot, 2003) by examining the location of maximum vertical uplifting wind speed at 850 hPa over Sahel in July and over southern Africa in January. However, we did not find pronounced effects for ITF (not shown) suggesting that changes in the rainbelt location for central Africa are mainly caused by changes in precipitation intensity rather than by changes in meridional circulation.

4. Discussion

4.1 Related tenets of Regional Earth System Modelling

Previous studies on land-climate interactions for Africa were carried out either over some African sub-region (e.g. Wang and Alo, 2012;Yu et al., 2015), or at a relatively coarse resolution within the implementation of GCMs (e.g. Kucharski et al., 2013), or without considering feedback effects from natural vegetation dynamics and only investigating anthropogenic impacts such as deforestation or afforestation (Lawrence and Vandecar, 2015). In this study, We investigated the coupled dynamics of climate and vegetation patterns over Africa under a future climate change scenario, applying a regional-scale ESM. The development that dynamically couples a dynamic representation of vegetation structure, composition and distribution to a physical climate model at a comparatively high grid resolution. Uniquely among existing studies of regional-scale coupling, including vegetation dynamics in a coherent way, enables the quantification of vegetation-change-induced feedbacks in climate simulations (Rummukainen, 2010;Smith et al., 2011;Giorgi, 1995;Zhang et al., 2014;Wramneby et al., 2010). In this way, it is abledynamics for Africa, this enabled us to isolate the regional biophysical feedbacks, which are usually not easy to disentangle in a global application in which the effects of changes in carbon-cycle and large-scale circulation tend to combine withcompound the biophysical effects.

In comparison with global ESMs, the added value from the regional ESMs lies in the enhanced resolution obtained in a regional setup as presented in this study, allowing for a more detailed representation of small-scalelocal surface features such as topography, land use, vegetation change, and consequently possible related feedbacks, and also enhancing the model's ability to capture climatic variability and extreme climatic events (Giorgi, 1995;Rummukainen, 2010). In addition,, 2016). Improvements in the representation of local processes (e.g. those that determine surface temperature) maymay be expected to result in improved larger scale features (e.g. sea level pressure) (Feser, 2006;, circulation patterns) (Diffenbaugh et al., 2005). As seen also in a previous evaluation study for the atmosphere-only version of RCA for Europe, a ;Feser, 2006). For example, Kjellström et al. (2005) found that reduced bias in surface air temperature results – largely determined by local energy balance – resulted in a better representation of interannual variability of mean sea level pressure and circulation patterns, and improvesimproved the simulation of precipitation (Kjellström et al., 2005)..

4.2 Vegetation dynamics over Africa for present and future

4.2 African vegetation patterns and change

Vegetation dynamics are critically important in modulating the evolution of the 21st century climate in our study. Land use and grazing (Sankaran et al., 2005;Bondeau et al., 2007;Lindeskog et al., 2013;Bondeau et al., 2007;Sankaran et al., 2005), which were not included in our study, represent additional potentially important drivers of land surface changes. The historical vegetation state is also relevant for future simulations, due to legacy effects lasting decades or even centuries (Wang et al., 2004;Moncrieff et al., 2014). Apart from the biases in climate (model) forcing, biases in simulated vegetation may come from the absence of consideration of these aspects and result in over- or under-estimation of the vegetation state. Nevertheless, the vegetation-feedback effects associated with the vegetation sub-model are still likely to be captured here, as vegetation dynamics in terms of forest cover changes and interannual and inter-seasonal variability of vegetation productivity are more important than the absolute vegetation state when considering vegetation feedbacks. Previous studies in the offline vegetation model LPJ-GUESS suggests that the vegetation dynamics for savannah and tropical forest vegetation are robust (Weber et al., 2009;Ahlström et al., 2015), providing additional confidence for the examination of the vegetation-climate interaction in our study.Moncrieff et al., 2014) and their influences on climate-vegetation equilibria (Claussen, 1998;Wang and Eltahir, 2000). While our model exhibited a degree of bias in simulated vegetation under the present climate, the overall distribution of the major vegetation types of the continent (forest, savannah and grassland) was broadly correct. Arguably, vegetation type is a more important determinant of climate-vegetation equilibrium than structural parameters of a given type, such as LAI (Claussen, 1994;Wang and Eltahir, 2000).

Under future climate change, the vegetation response to environmental changes will differ. As revealed by Previous experimental (Kgope et al., 2010) and modelling (Sitch et al., 2008;Moncrieff et al., 2014) studies, vegetation may be expected to become less sensitive to climate conditions when highlight the potential importance of physiological effects of atmospheric CO_2 concentration increases. This is because CO_2 fertilization of photosynthesisconcentrations on the productivity and enhanced water use efficiency linked to a reduction of stomatal conductance, which causes a shift towards higher woody-plant dominance, resultingof vegetation, particularly in densification of treelow latitude and shrub cover relative to grasses in savannahs, or replacement of savannah with forestwater-limited ecosystem types. Shrub encroachment and woody thickening has been observed in water-limited areas including Sahel in recent decades, coinciding with rising CO_2 concentrations (e.g. Liu et al., 2015). In our results, the simulated vegetation dynamics are consistent with these trends, presenting a similar trajectory of vegetation changes increased woody plant dominance (not shown), and a similar future vegetation pattern ()(Fig. 4) as in previous modelling studies (e.g., Sitch et al., 2008;Moncrieff et al., 2014). The vegetation changes simulated by our model under future climate forcing, are large relative to the bias noted in the representation of present-day vegetation state. This provides some confidence that the simulated future vegetation is not critically dependent on these biases and, in turn, that the emergent mechanisms of vegetation-climate interaction and their consequences for circulation and precipitation trends suggested by our study are robust.

4.3 Vegetation feedbacks and land-ocean temperature contrasts

The land-ocean contrast is an important driver of continental precipitation, as it determines the transport of moisture from ocean to land (e.g. LambertGiannini et al., 2011;Fasullo, 20102003;Giannini et al., 2005;Fasullo, 2010;Boer, 2011;GianniniLambert et al., 20032011). The recent changepositive trend in Sahel rainfall over recent decades is a good example of linking moisture transport to land-ocean contrast, where changes in SSTs over adjacent tropical oceans around Africa are key to the fragile balance that defines the regional circulation system (Camberlin et al., 2001;Rowell, 2001;Giannini et al., 2003;Camberlin et al., 2001). Land-surface feedback is found to modify the interannual to interdecadal climate variability in this region by vegetation-induced albedo or evapotranspiration effects (Charney, 1975;Zeng et al., 1999;Wang et al., 2004). In our study, the SSTs were prescribed from CanESM2GCM-generated data, therefore the altered land-ocean thermal contrast between simulations with and without feedback originated solely from the changes in land surface temperature induced by, in turn attributable to vegetation dynamics. Although this represents a land-forced mechanism in contrast to an ocean-forced one inferred in other studies (e.g. Giannini et al., 2003;Tokinaga et al., 2012), we assume that the mechanisms are similar given the similarity in the magnitude of simulated circulation changes.. Wind speed and land-ocean temperature contrast changeare reduced by approximately by 0.2 m s^{-1} and 0.2°C, respectively, between FB and NFBwhen vegetation feedbacks are enabled in our study (Fig. 5 and Table A2); these are comparable to the changes simulated in other studies for the Sahel (approximately 0.2-0.5 m s^{-1} per 0.2°C (Giannini et al., 2005)) and for the Pacific Oceans (approximately 0.3 m s^{-1} per 0.3°C (Tokinaga et al., 2012)). However, the relative importance of such changes may differ for local climate systems: the lower branch of the Walker cell over the eastern tropical Atlantic Ocean, which we have focused on in this study, may be in a fragile balance and is more vulnerable to changes in thermal contrasts (equatorial westerlies slowed down by approximately 0.2 m s^{-1} from less than 2 m s^{-1} of the present-day wind speed in rainy seasons, Table A2) compared to the stronger monsoonal circulation for Sahel and the Walker cell over the equatorial Pacific Ocean (> 5 m per second wind speed in their peak months, Young, 1999). Our results indicate that even a small disturbance of the eastern Tropical Atlantic circulation cell may produce profound impacts (larger relative reduction in precipitation compared with the studies by Giannini et al. (2005) and Tokinaga et al. (2012)).

~~Moreover, we assume that a study with a dynamic ocean component would result in a similar outcome~~

Despite biases in the initial precipitation and vegetation state (LAI) for some regions, our model was able to reproduce the present-day land cover type, and the simulated present-date climate is close to previous study (Nikulin et al., 2012) using the same physical sub-model with observed land cover type. Under future climate change, vegetation-induced changes in circulation, thus a substantial change in moisture transport and precipitation, are mainly triggered by changes in land cover type (Fig. 4a), therefore, we argue that the influences from biases in initial conditions on such mechanism found in this study should be limited. Our study used prescribed SST forcing from a GCM and could thus not account for additional or opposing feedbacks mediated by ocean dynamics. However, as the ocean heat capacity is relatively large and variation in land-ocean thermal contrast can be greatly buffered by ocean heat uptake (Lambert and Chiang, 2007~~).~~), we suggest that results should not change fundamentally if a dynamic ocean component was introduced to the model.

5. Conclusion and outlook

~~In this study~~ We investigated the potential role of vegetation-mediated biophysical feedbacks on climate change projections for Africa in the 21st century. In current savannah regions, enhanced forest growth results in a strong evaporative cooling effect. We also identify alterations in the large-scale circulation induced by savannah vegetation change, resulting in remote effects and modulation of tropical rainfall patterns ~~in~~over Africa~~.~~, favouring savannah ecosystems at the expense of equatorial rainforest. Our results ~~emphasize~~point to the potential importance of ~~accounting for~~ vegetation-atmosphere interactions ~~in~~for regional climate ~~projections for tropical~~dynamics and ~~sub-tropical Africa~~trends, and ~~stress~~motivate the ~~necessity to consider~~incorporation of vegetation ~~feedbacks for more reliable estimates of~~ dynamics and land-atmosphere biophysical coupling in regional ~~future~~models. This has become the standard in global climate ~~change projections~~modelling, but remains rare in regional climate modelling.

Future work can include detailed studies on the role of vegetation feedbacks in the regional climate projections with respect to shorter-term dynamics such as climate variability and extreme events, which may have crucial implications for ~~land surface~~landscape processes such as wildfire. ~~On the other hand,~~ Regional and global biogeochemical feedbacks on future climate change may be triggered by regional biophysical feedbacks, which can impose important influences to with implications for regional climatic trendtrends, variability and seasonality in conjunction with under future greenhouse forcing. Such changes may impose important influences (Zhang et al., 2014). Impacts on the African carbon balance, especially for the of semi-arid ecosystem like savannahs whose carbon balance is strongly associated with changes in climatic conditions, known to respond sensitively to variations in rainfall (Ahlström et al., 2015). Further examination of vegetation feedback effect in regional ESMs may have its distinct values especially when regional processes are concerned. Future work should also include the ongoing) may be particularly relevant to address for Africa. The development of regional ESMs includingto account for the improvementimpacts of the system's ability to represent land surface properties by incorporating important land surface processesland use interventions such as changes in land use (e.g.afforestation and reforestation, as well as forest clearing/, grazing) and landfire management (e.g. irrigation), for use in versatile land-atmosphericmay be a valuable next step, enabling land surface-atmosphere interaction studies linked to socioeconomic scenarios and climate change mitigation strategies.

Appendix A: Description ~~for~~of the coupling between RCA and LPJ-GUESS

In RCA-GUESS, the LSS in RCA is coupled with LPJ-GUESS, which feeds back vegetation properties to RCA. RCA provides net downward shortwave radiation, air temperature, precipitation to LPJ-GUESS. In return, LPJ-GUESS provides daily updated LAI and the annually updated tile sizes (determined from the simulated maximum growing season LAI summed across tree and herbaceous PFTs in the previous year (Smith et al., 2011)). In the forest tile in RCA, vegetation cover in this tile is estimated as the foliar projective cover (FPC) using Beer's law:

$$\mathrm{A}_{tree} = 1.0 - \exp(-0.5 \cdot LAI_{tree}), \tag{1}$$

where LAI_{tree} is the aggregated LAI of woody species, simulated by LPJ-GUESS in its forest tile in which vegetation is assumed to comprise trees and understory herbaceous vegetation. The natural vegetated faction of the open land tile was calculated similarly:

$$\mathrm{A}_{grass} = 1.0 - \exp(-0.5 \cdot LAI_{grass}), \tag{2}$$

where LAI_{grass} is the summed LAI of the simulated herbaceous PFTs from the herbaceous tile of LPJ-GUESS in which only herbaceous vegetation is allowed to grow. The relative covers of the forest and open land tiles affect surface albedo, which is a weighted average of prescribed albedo constants for forest, open land and bare soil and controls the absorption of surface incoming solar radiation, and therefore influences surface energy balance and temperature.

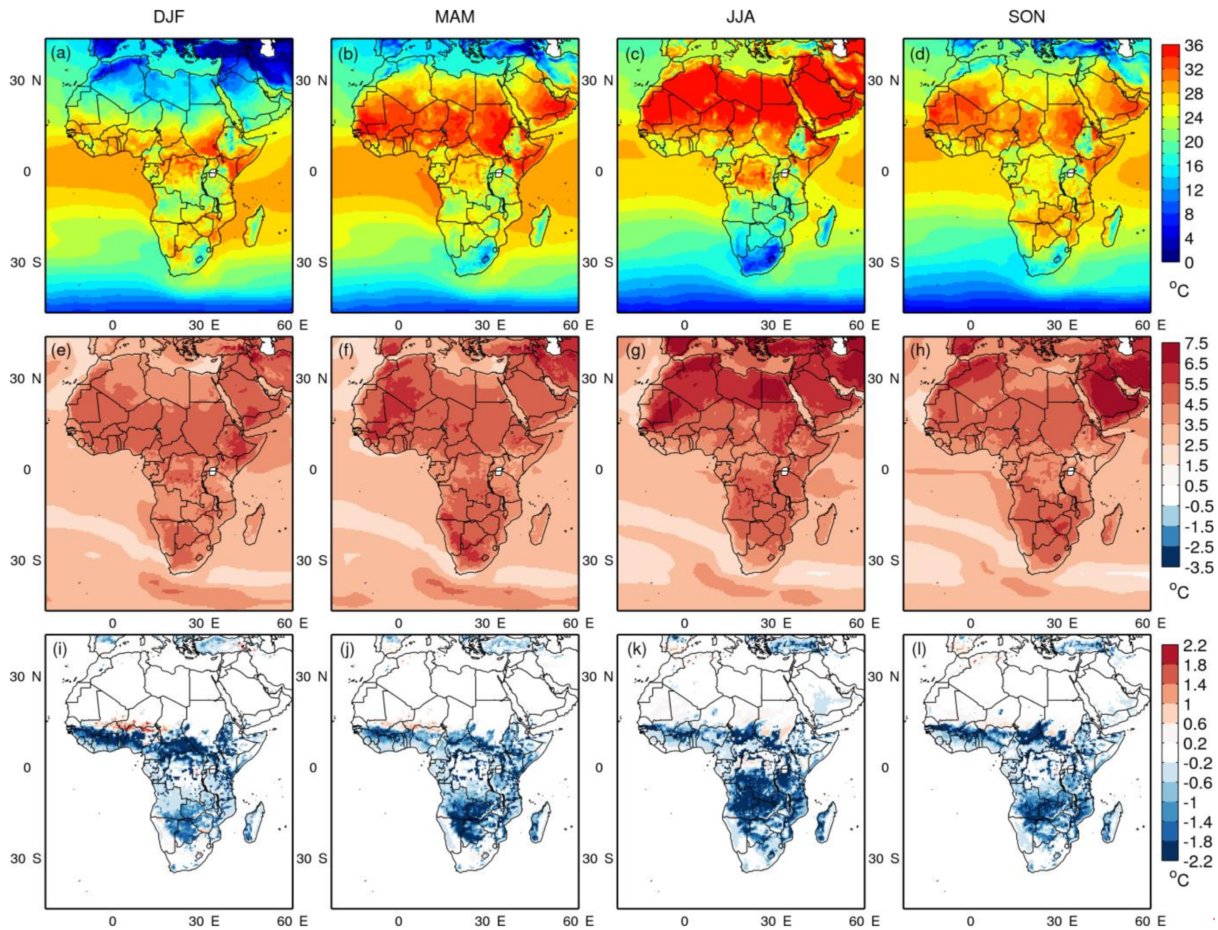
The turbulent heat fluxes are influenced by the properties of each tile, such as surface roughness and surface resistance, which partly depend on vegetation properties provided by LPJ-GUESS. The vegetation surface resistance controls vegetation transpiration and bare soil evaporation for latent heat flux calculation. It scales with LAI and varies between the different types of vegetation and affected by the incoming photosynthetically active radiation, soil-water stress, vapour pressure deficit, air temperature and soil temperature. The aerodynamic resistance controls both latent heat flux and sensible heat flux and is influenced by surface roughness length distinguished from open land and forest. The total heat fluxes and heat transfer determine the time evolution of the surface temperature and thus the thermodynamics in the lower boundary layer. More details about the LSS are given in Samuelsson et al. (2006), and the description of its coupling to the vegetation sub-model is provided by Smith et al. (2011).

Table A1. Characteristics of the plant functional types (PFTs) used in the vegetation sub-model LPJ-GUESS.

Characteristics	NE	BE	TrBE	TrBR	TBS	IBS	C3G	C4G
Leaf phenology[a]	E	E	E	D	D	D	R	R
Drought tolerance	low	low	low	low	low	low	very low	very low
Shade tolerance	high	high	high	low	high	low	low	Low
Optimal temperature range for photosynthesis (°C)	10-25	15-35	25-30	25-30	15-25	10-25	10-30	20-45

Min T_c for survival (°C)[b]	-	1.7	15.5	15.5	-18	-	-	15.5

Notes: NE, needleleaved evergreen tree; BE, broadleaved evergreen tree; TrBE, tropical broadleaved evergreen tree; TrBR, tropical broadleaved raingreen tree; TBS, shade-tolerant broadleaved summergreen tree; IBS, shade-intolerant broadleaved summergreen tree; C3G, C3 grass or herb; C4G, C4 grass or herb;
[a]E, evergreen; D, deciduous; R, raingreen.
[b]T_c = mean temperature (°C) of coldest month of year.



DJF
MAM
JJA
SON
(a)
(b)
(c)
(d)
(e)
(f)
(g)
(h)
(i)
(j)
(k)
(l)
30 N
0
30 S
0
30 E
60 E
36
32
28
24
20
16
12
8
4
0
°C
7.5
6.5
5.5
4.5
3.5
2.5
1.5
0.5
-0.5
-1.5
-2.5
-3.5
°C
2.2
1.8
1.4
1
0.6
0.2
-0.2
-0.6
-1
-1.4
-1.8
-2.2
°C

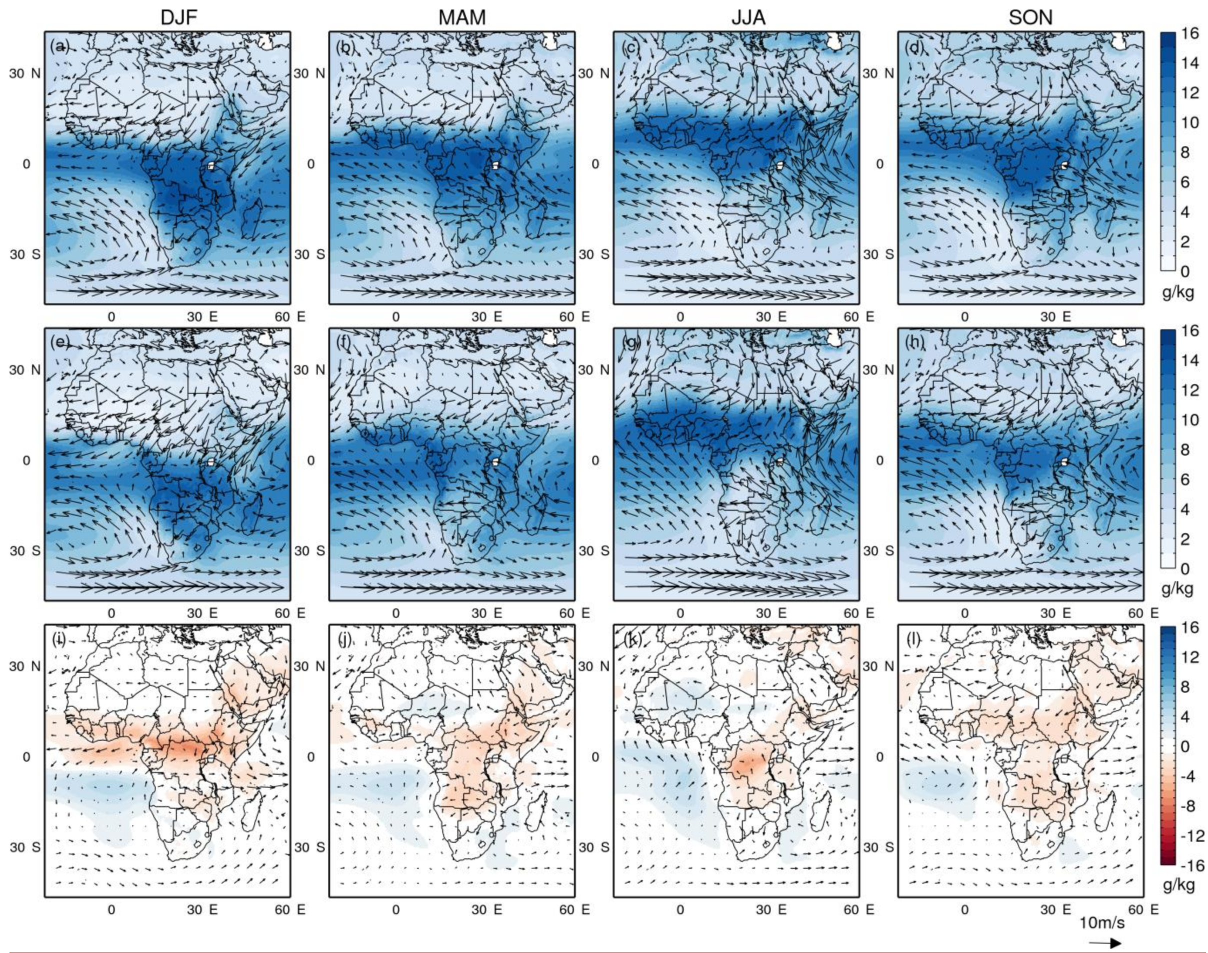


Fig. A1. Seasonal atmospheric circulation (arrows, m s^{-1}) and specific humidity (colour contours, g kg^{-1}) at 850 hPa pressure level from ERA-Interim (1st row), NFB run (2nd row), as well as their differences (3rd row, NFB minus ERA-Interim), for the period 1997-2010.

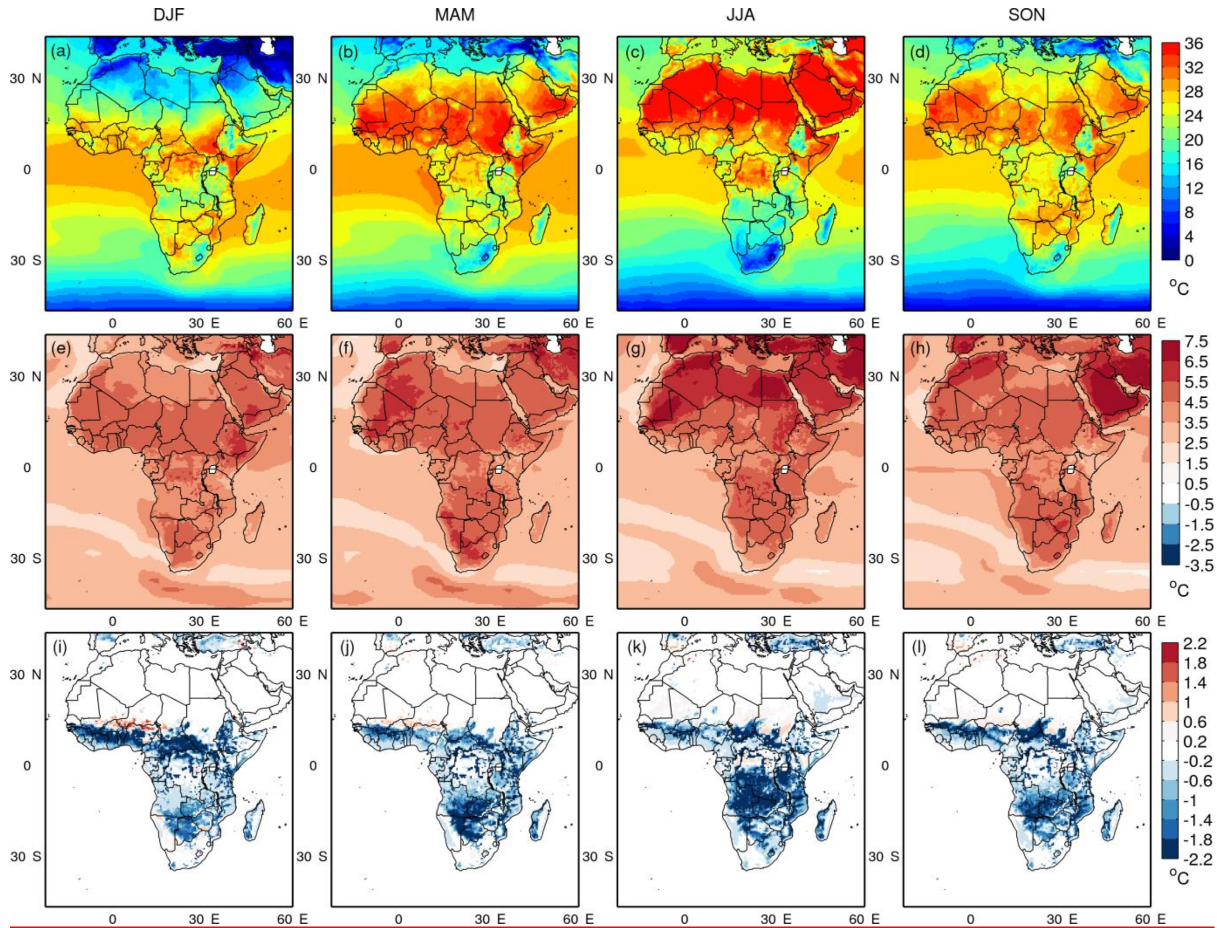


Fig. A2. Simulated seasonal surface temperature~~: 1st panel,~~ for present day~~; 2nd panel,~~ (a-d), for changes in future in the NFB experiment~~; 3rd panel,~~ (e-h, future minus present day), and for changes from vegetation feedback~~, represented as~~ in future (i-l, FB minus NFB for future). Definitions for calculation period, climate change signal and vegetation feedbacks are given in Sect. 2.2.

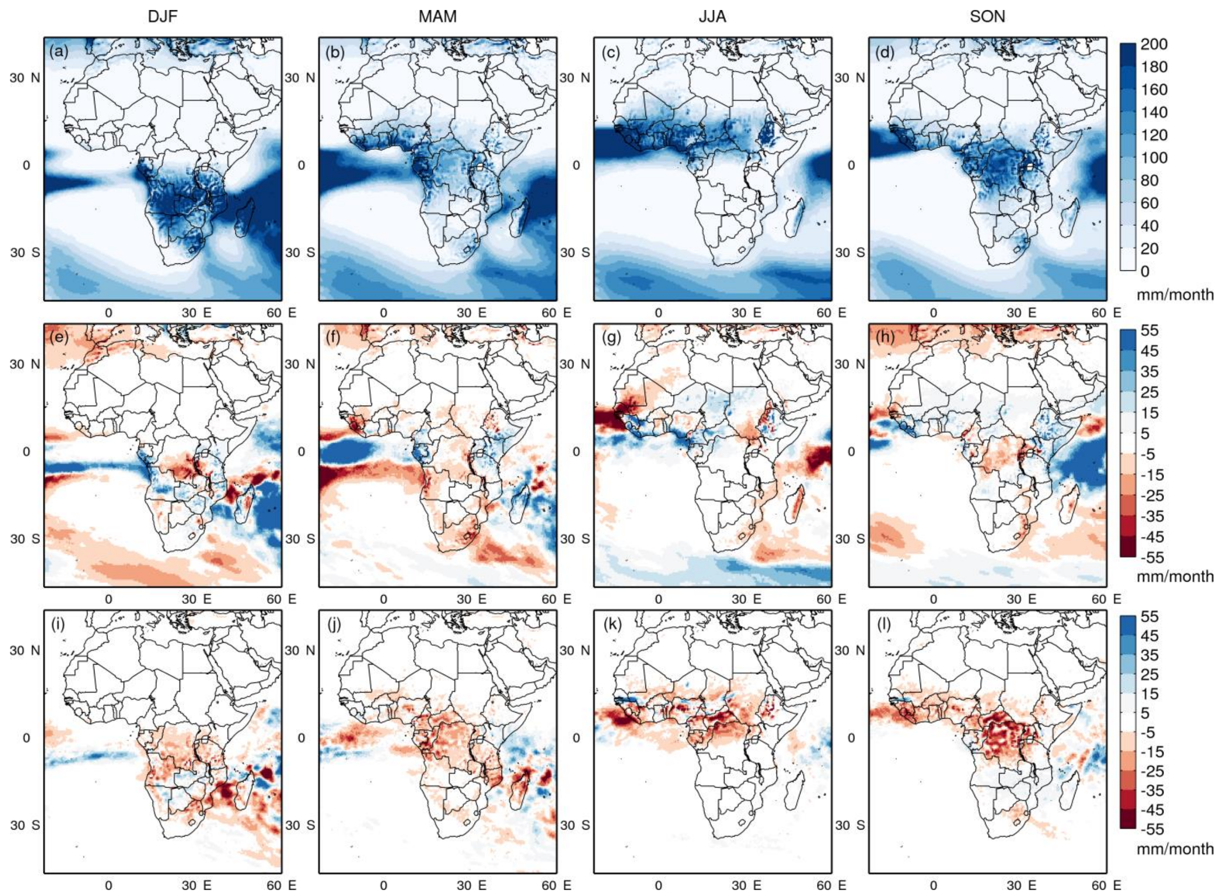


Fig. ~~A.~~A3. Similar to Fig. A2., but for precipitation.

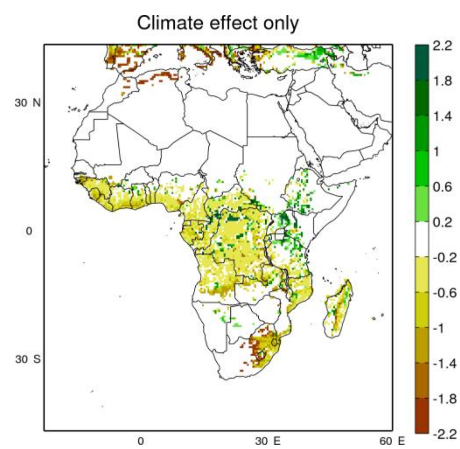


Fig. ~~A.~~A4. Changes in forest tile LAI from the period 1991-2010 to the period 2081-2100 in FB_CC experiment.

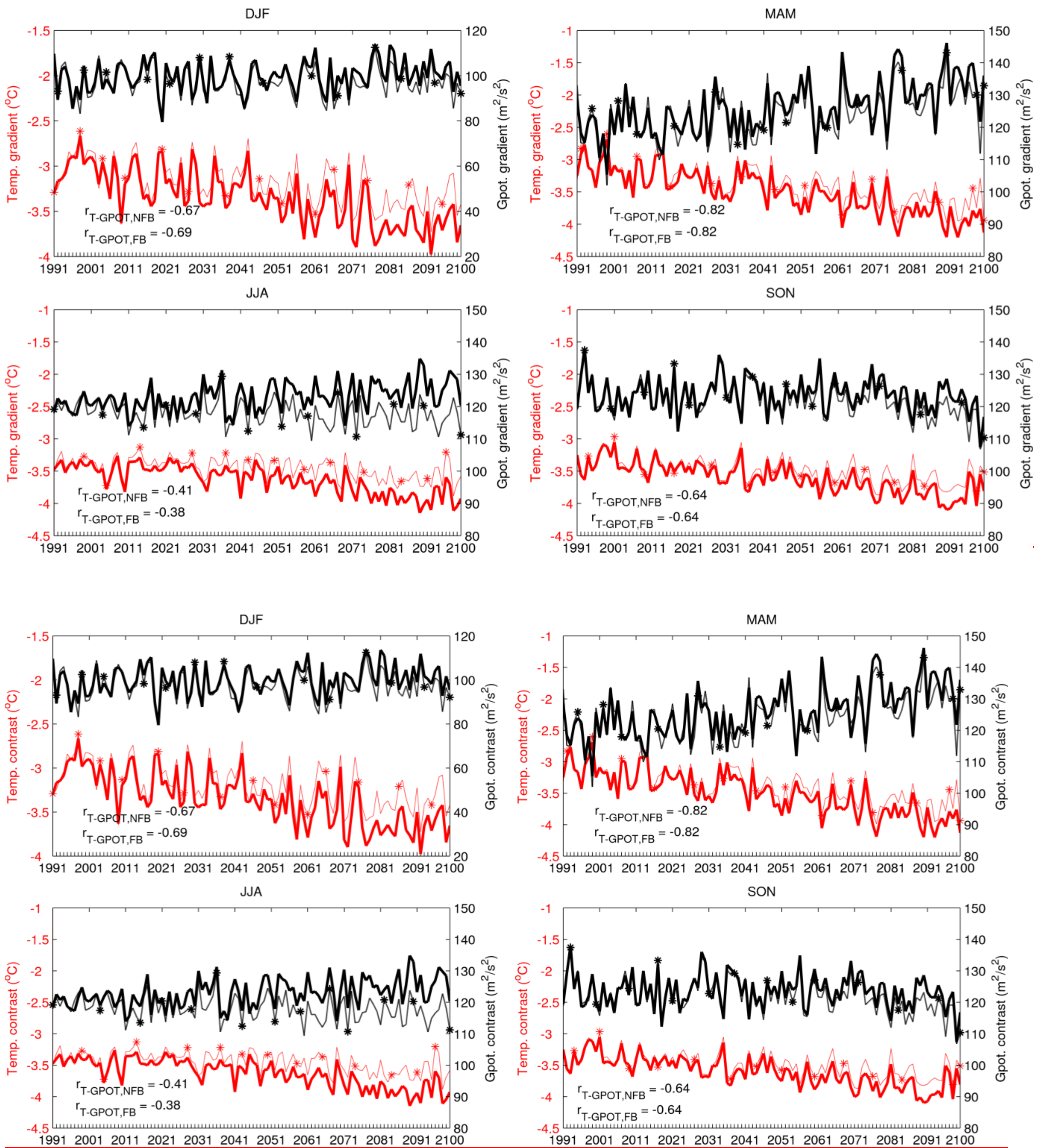


Fig. A.A5. Annual changes in atmospheric ocean-land temperature contrast (∇T) and geopotential contrast ($\nabla \phi$) in time series for four seasons, represented by the mean contrast at the three pressure levels 850, 925 and 975 hPa (ocean minus land) within the domain 15°N-15°S, 24°W-20°E (see the inset in the panel for JJA in Fig. 5). Correlation coefficient (r) between atmospheric temperature contrast (∇T) and geopotential contrast ($\nabla \phi$) are computed based on the de-trended annual time-series values for both FB (thick lines) and NFB (thin lines with asterisks) simulations. Changes between FB and NFB are significant at 95% confidence level for the whole time period. Note the different y-axis for DJF.

Table A2. Atmospheric temperature contrast, geopotential contrast and westerlies wind speed for the present-day state and contributions from climate change (CC subscript) and vegetation feedbacks (FB subscript), standard deviation is in parenthesis.

	DJF	**MAM**	**JJA**	**SON**
$\nabla T_{present\text{-}day}$ (°C) [a]	-3.06 (0.30)	-3.15 (0.34)	-3.47 (0.22)	-3.37 (0.24)
$\Delta\nabla T_{CC}$ (°C) [a]	-0.59*	-0.73*	-0.45*	-0.47*
$\Delta\nabla T_{FB}$ (°C) [a]	0.29*	0.23*	0.31*	0.22*
$\nabla\phi_{present\text{-}day}$ ($m^2 s^{-2}$) [a]	98.14 (5.92)	120.86 (7.03)	120.94 (3.83)	124.08 (4.58)
$\Delta\nabla\phi_{CC}$ ($m^2 s^{-2}$) [a]	3.94	11.96*	4.73*	-3.32
$\Delta\nabla\phi_{FB}$ ($m^2 s^{-2}$) [a]	-4.93*	-3.86*	-8.96*	-3.92*
$u_{zonal,present\text{-}day}$ ($m\ s^{-1}$) [b]	0.01 (0.27)	1.47 (0.32)	0.87 (0.37)	1.22 (0.31)
$\Delta u_{zonal,CC}$ ($m\ s^{-1}$) [b]	0.35*	0.32*	0.68*	0.17*
$\Delta u_{zonal,FB}$ ($m\ s^{-1}$) [b]	-0.00	-0.21*	-0.28*	-0.16*

Note: [a]: Calculations are same as Fig. 5.

[b]: u_{zonal} is the averaged zonal wind speed for the pressure levels 850, 925 and 975 hPa between 3.5°N-6.5°N and 0-10°E; The positive represents westerly and the negative represents easterly.

*: Changes are significant at 95% confidence level using Mann-Whitney U-test (Hollander and Wolfe, 1999).

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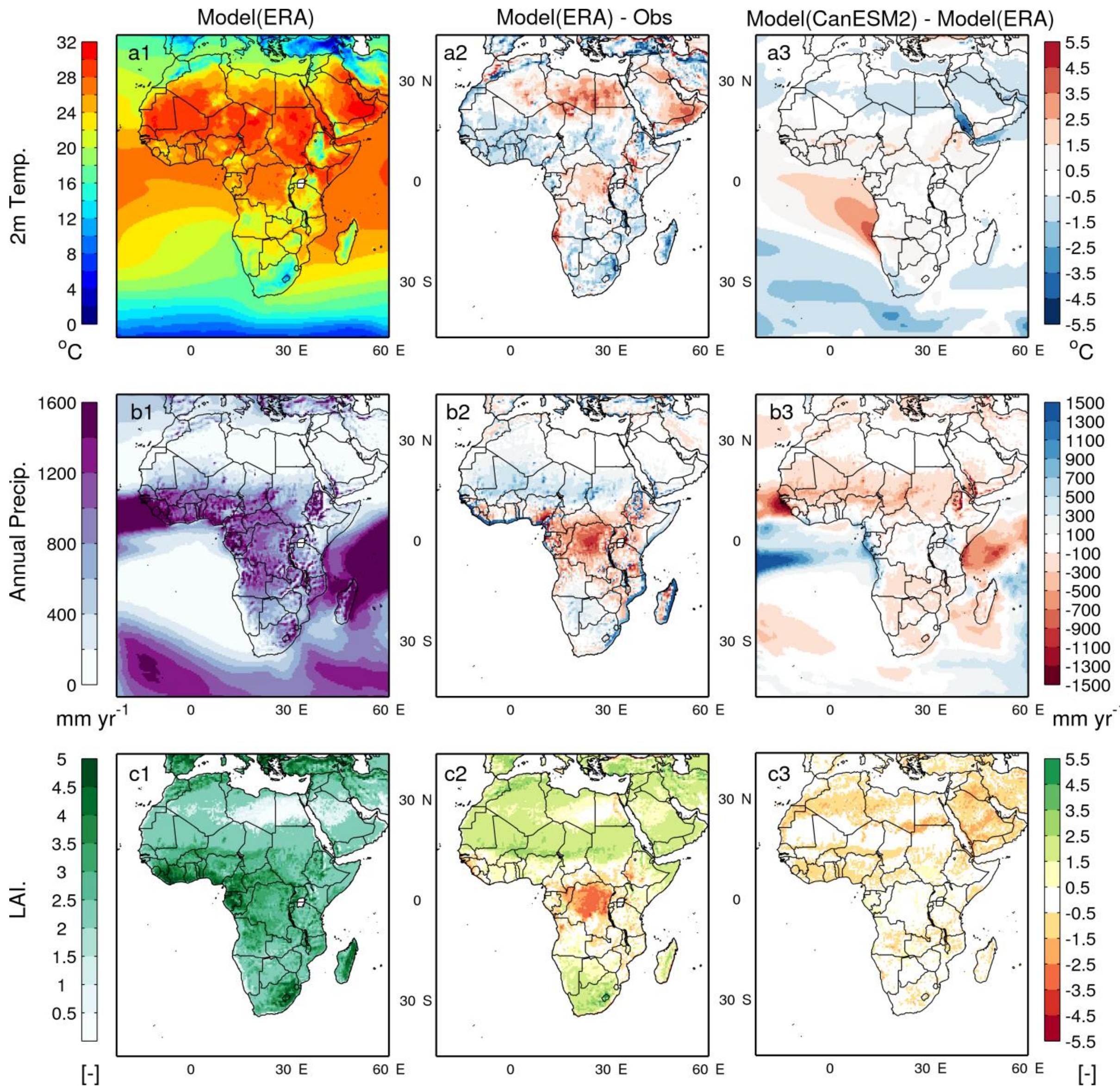
Figures and Tables



Model(ERA)
Model(ERA) - Obs
Model(CanESM2) - Model(ERA)
2m Temp.
Annual Precip.
LAI.
a1
a2
a3
b1
b2
b3
c1
c2
c3
°C
mm yr-1
[-]
30 N
0
30 S
0
30 E
60 E

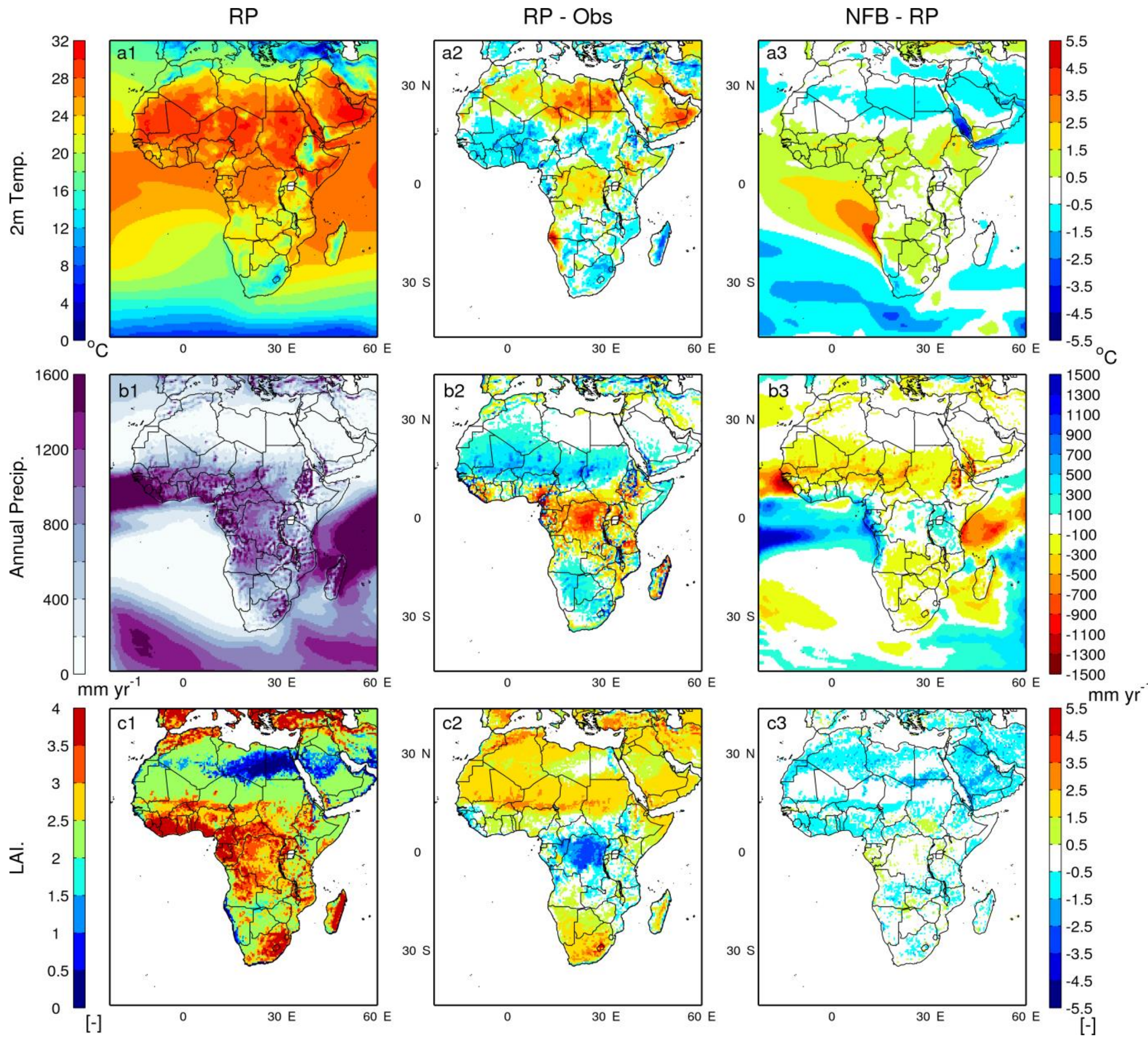
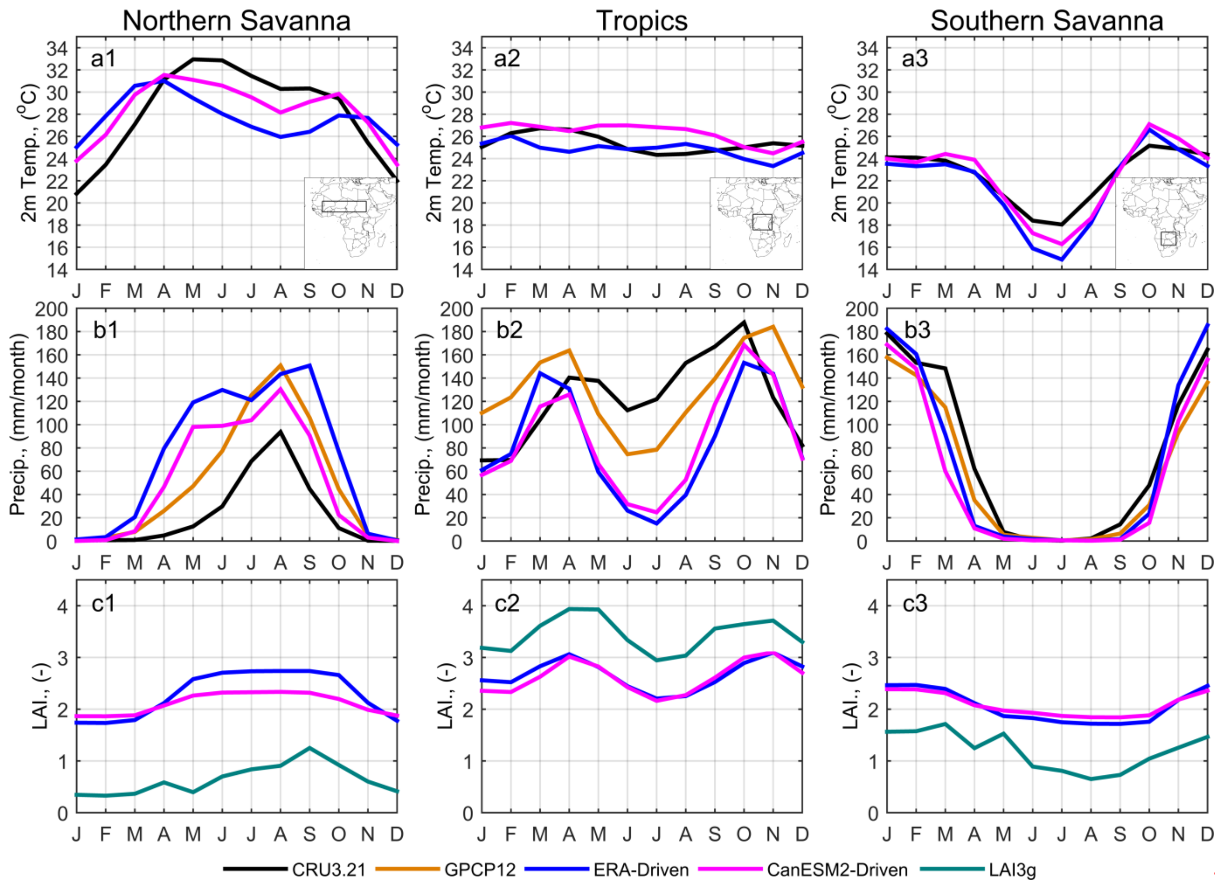


Fig. 1. Comparison between simulated and observed (a) annual mean near-surface air temperature, (b) annual precipitation and (c) annual maximum LAI for the period 1997-2010. Variables from the RP experiment (a1-c1) are compared with observations (a2-c2) and with those from the FB experiment (a3-c3), using RP minus observation and FB minus RP. For the comparison with observations (a2-c2), we used CRU temperature (a2) and precipitation (b2), as well as LAI3g (Zhu et al., 2013)(c2).


Northern Savanna
Tropics
Southern Savanna
a1
a2
a3
2m Temp., (°C)
b1
b2
b3
Precip., (mm/month)
c1
c2
c3
LAI., (-)
J F M A M J J A S O N D
CRU3.21
GPCP12
ERA-Driven
CanESM2-Driven
LAI3g

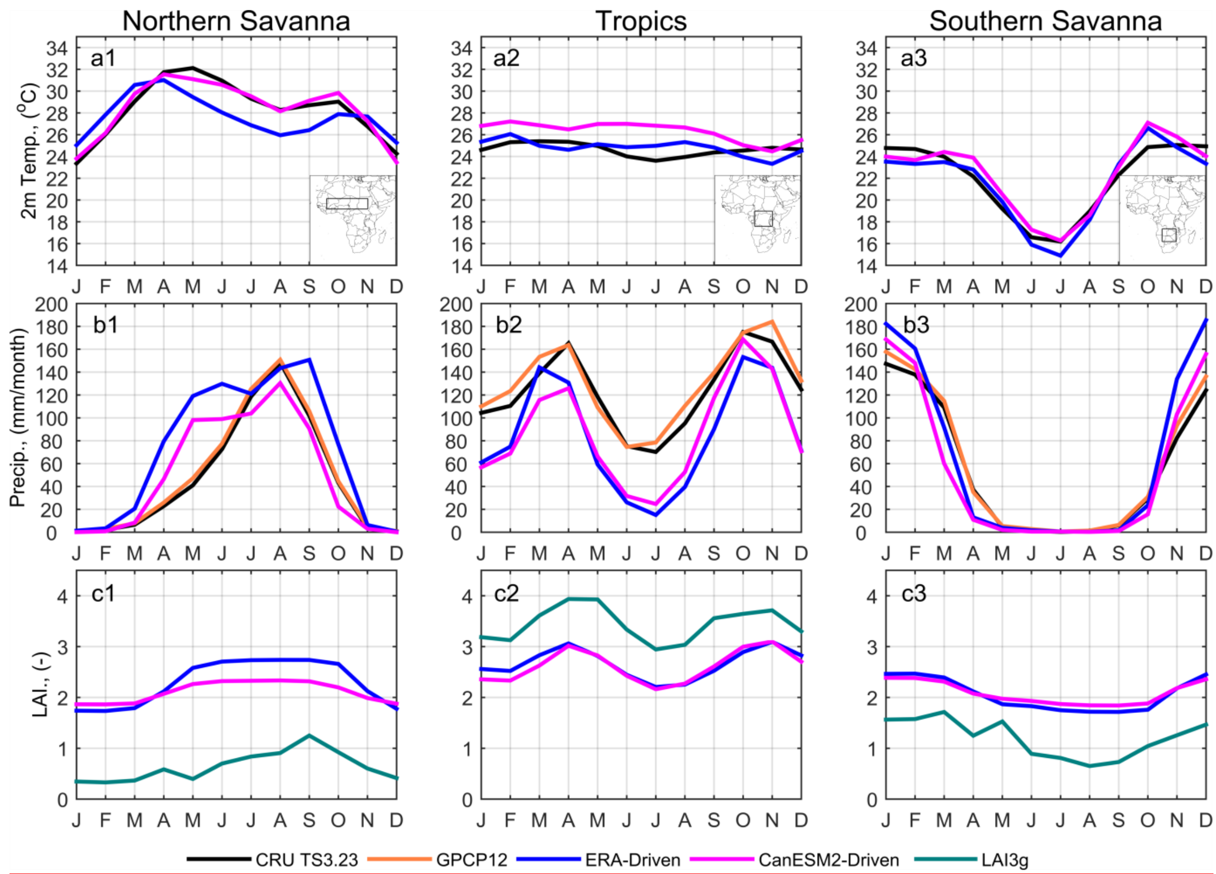


Fig. 2. Simulated seasonal cycle and observations for northern savannah (inset in a1), central Africa (inset in a2) and southern savannah (inset in a3) for the period 1997-2010. 2m temperature (a1-a3) and precipitation (b1-b3) are as Fig. 1. For LAI (c1-c3) monthly mean tile-weighted simulated LAI over the averaging period are used to compare with the observation.

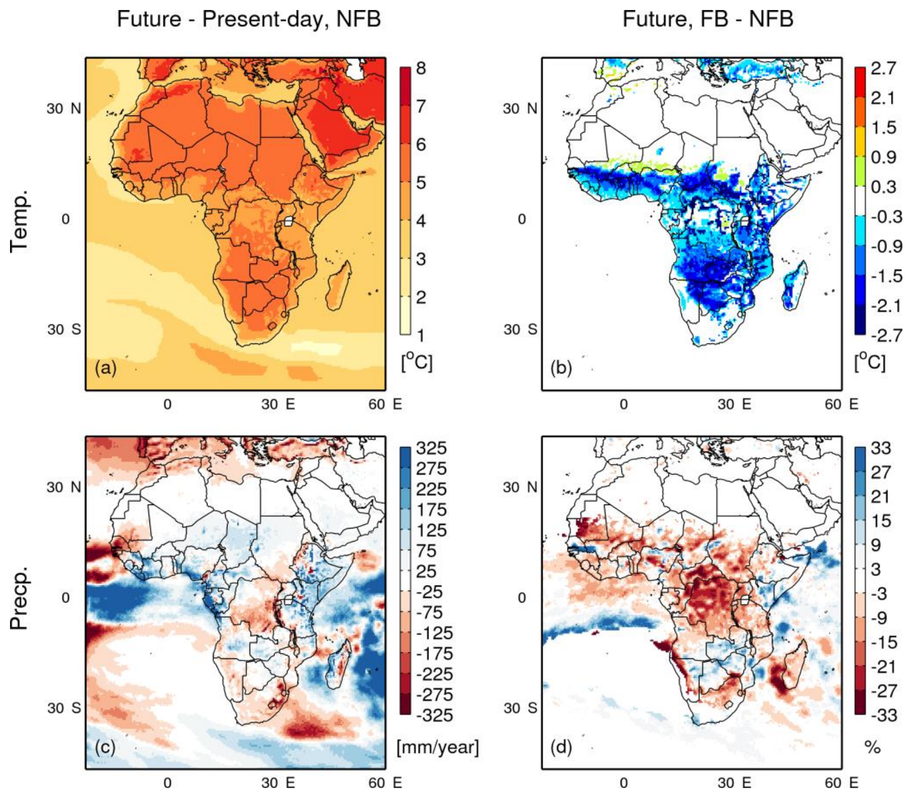


Fig. 3. Changes in surface temperature and precipitation due to climate change and vegetation feedback. The calculation of climate change signal and vegetation feedbacks, present-day and future periods are defined in Sect. 2.2. For (d), the percentage is calculated as the difference between FB and NFB (vegetation feedback) divided by the present-day level and multiplied by 100. Grid points with annual mean precipitation <20 mm year^{-1} are skipped.

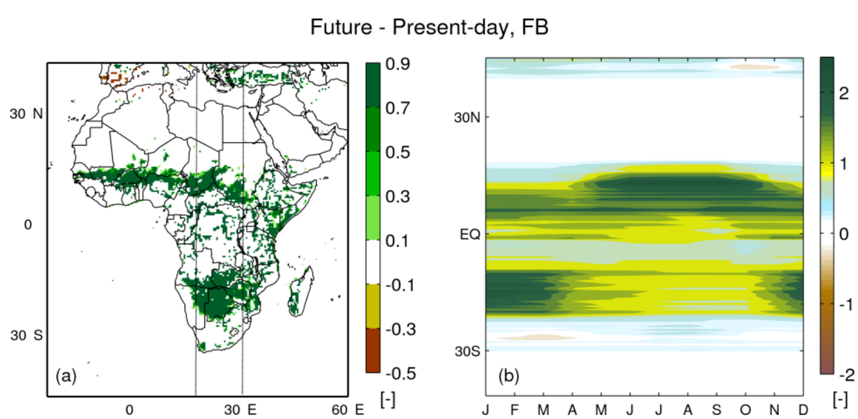


Fig. 4. (a) Change in forest fraction and (b) seasonal change in zonal mean forest LAI in the longitude band between 18°E and 30°E (lines in a), calculated as future minus present-day in FB experiment. Present-day and future periods are defined in Sect. 2.2.

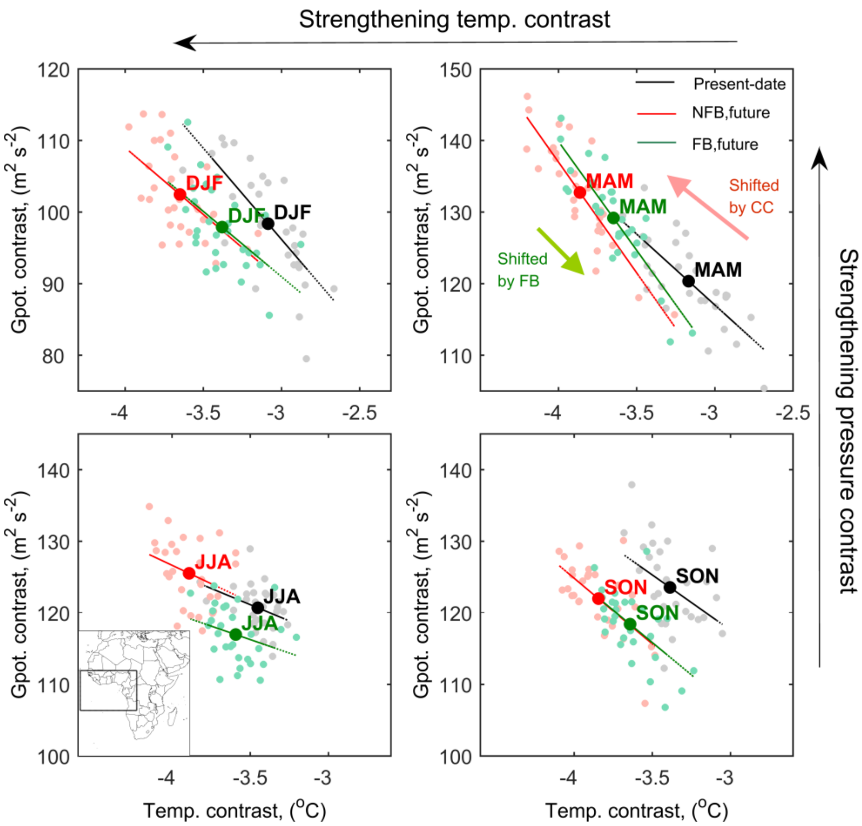


Fig. 5. Changes in atmospheric ocean-land temperature contrast (∇T) and geopotential contrast (∇ϕ), represented by the mean contrast at the three pressure levels 850, 925 and 975 hPa (ocean minus land) within the domain 15°N-15°S, 24°W-20°E (see the inset in the panel for JJA), for the NFB and FB simulation in the present-day and the future period (as defined in Sect. 2.2). Each scatter point represents the relation between ∇ϕ and ∇T for the correspondent season of one year, and the slopes represent its sensitivity during the selected periods.

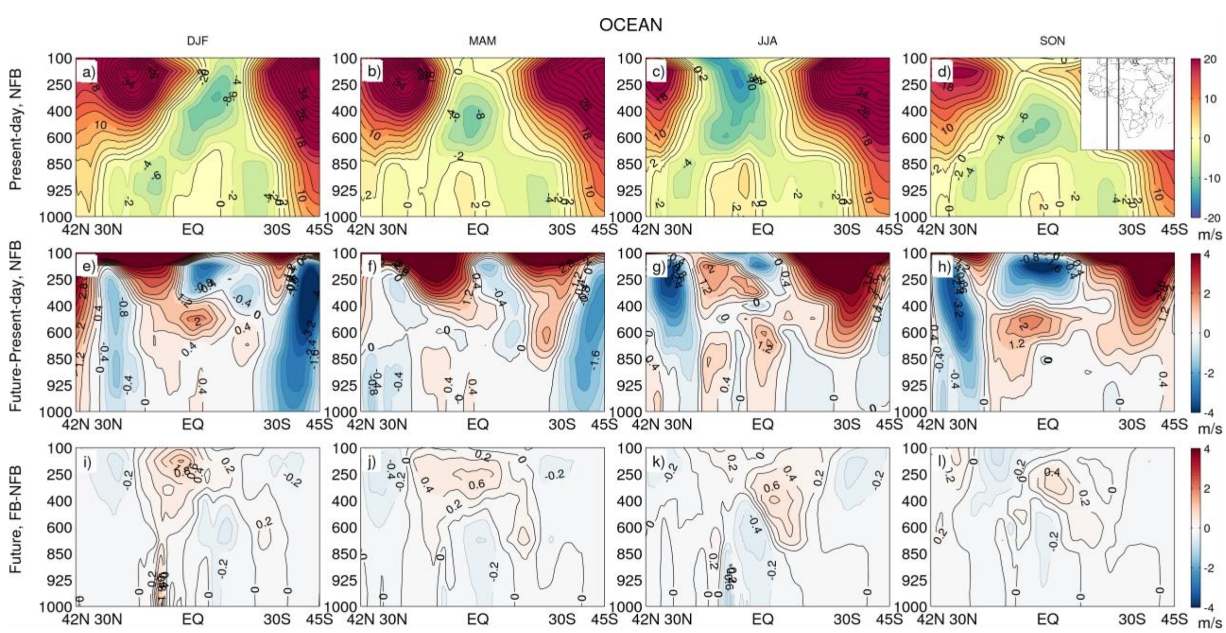


Fig. 6. Seasonal mean zonal wind speed in a cross section over adjacent Atlantic ocean (0-10°E, see the inset in d), for present-day (1st row), changes in future (future minus present-day, 2nd row) and the differences between FB and NFB runs in future (FB minus NFB, 3rd row). Unit is m s^{-1}, positive values represent westerlies and negative values represent easterlies. Present-day and future periods are defined in Sect. 2.2 Contour intervals from top row to bottom row are 2m s^{-1}, 0.4m s^{-1} and 0.2m s^{-1}, respectively.

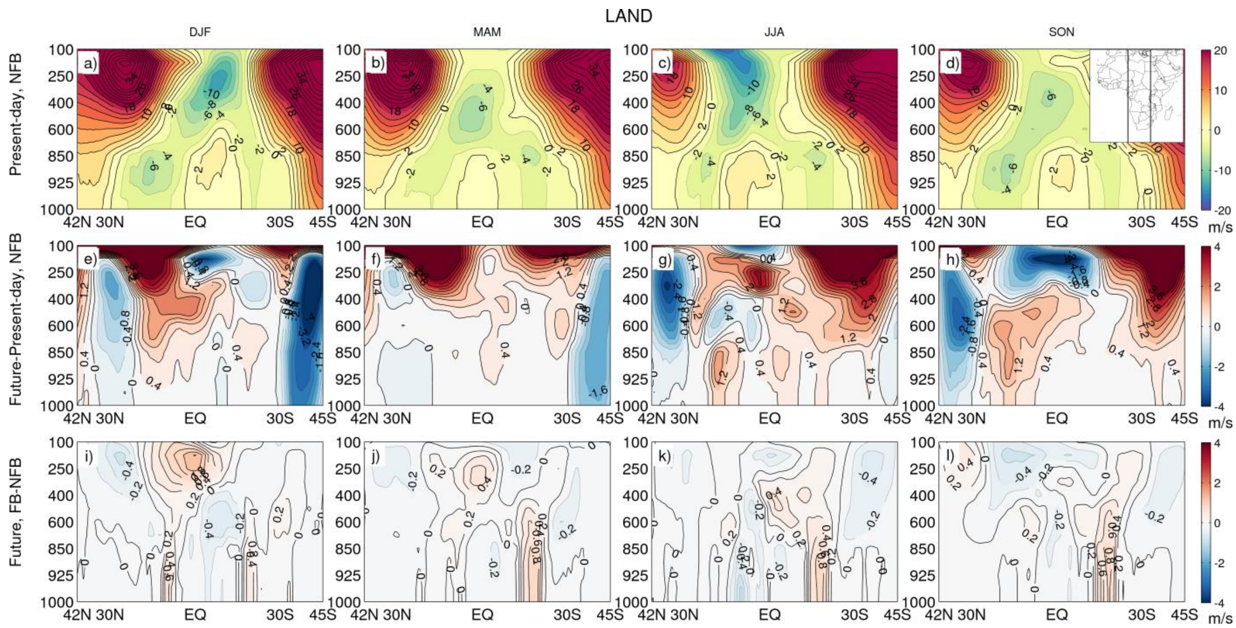


Fig. 7. As Fig. 6 but for longitudinal band over land (10°E-30°E, see the inset in d).

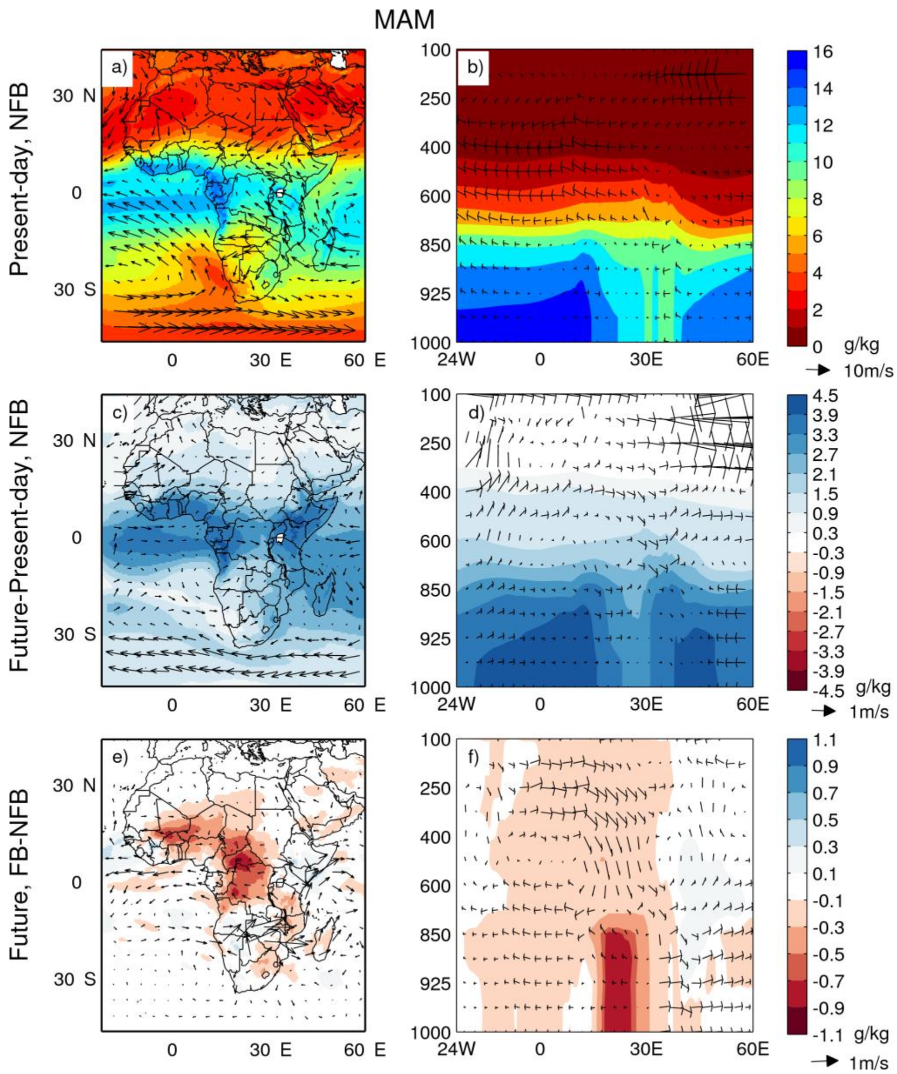


Fig. 8. Atmospheric circulation (arrows, m s^{-1}) and specific humidity (colour contours, g kg^{-1}) at 850 hPa pressure level for MAM, displayed as (a, c, e) for the entire domain, and (b, d, f) as a cross section for a latitude band between 2.5°S and 2.5°N, for present day (top), climate change impacts (middle) and the vegetation feedback (bottom). Definitions for calculation period, climate change signal and vegetation feedbacks are given in Sect. 2.2.

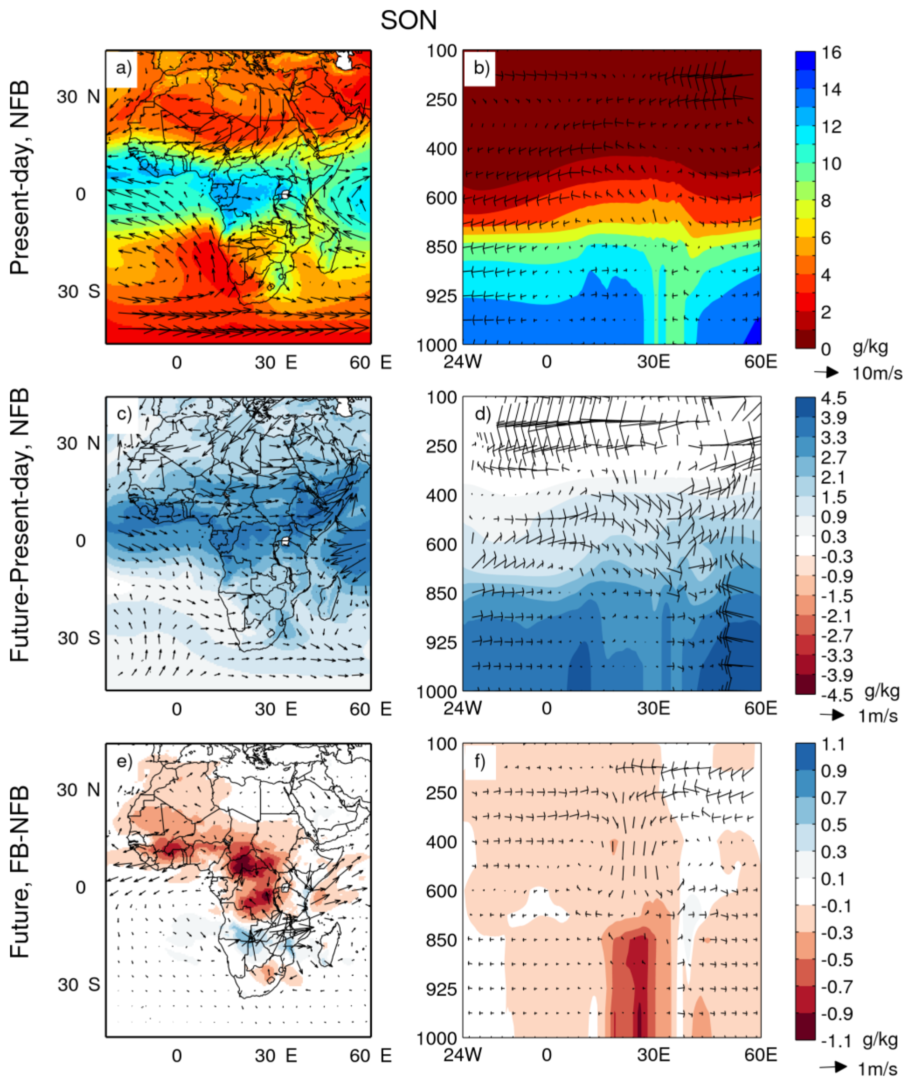


Fig. 9. As Fig. 8 but for SON.

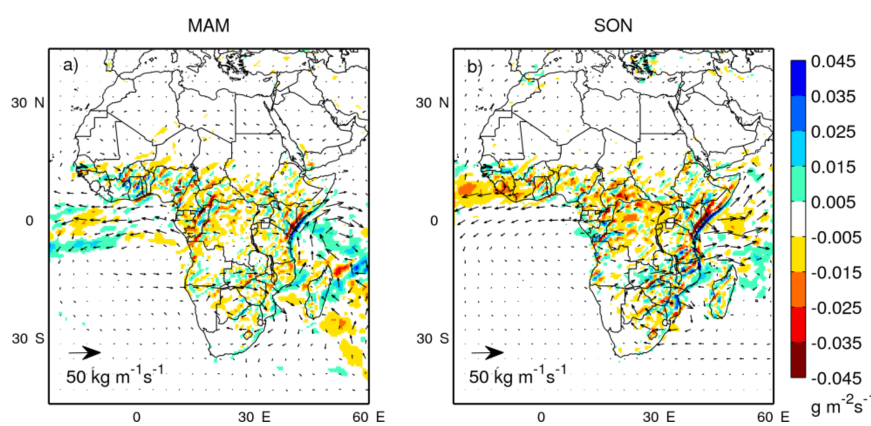


Fig. 10. Changes in vertically integrated moisture flux (arrows, kg $m^{-1}s^{-1}$) and moisture flux convergence (colour contours, g $m^{-2}s^{-1}$) caused by vegetation feedback, averaged over the future period (as defined in Sect. 2.2) for (a) MAM and (b) SON.

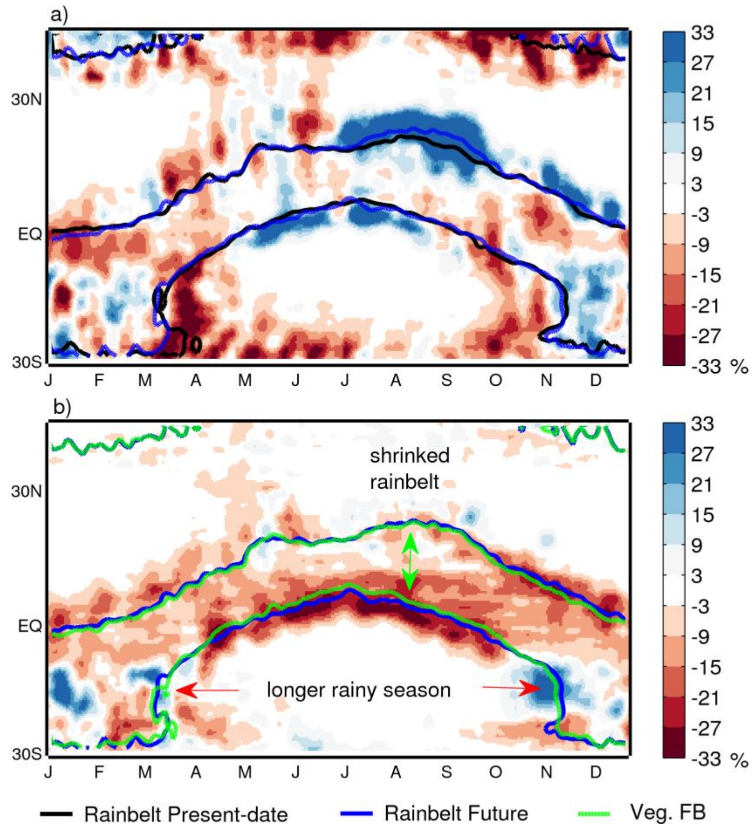


Fig. 11. Daily changes in precipitation averaged over the longitude band 18°E-30°E, represented as relative changes in daily precipitation intensity (shading, %) and rainbelt location (contour) due to (a) climate change and (b) vegetation feedback for future. The rainbelt location is defined as 2mm day^{-1} contour. 10-day running mean is applied for daily values.

Table 1. Experimental design for the investigation of the vegetation-climate feedbacks in this study.

Runs	Vegetation Feedbacks	Radiative forcing[a]	CO_2 forcing[b] for vegetation sub-model	Simulated period	Boundary condition
RP	Dynamic	Historical	Historical	1979-2011	ERA-Interim
FB	Dynamic	Transient under RCP8.5	Transient under RCP8.5	1961-2100	CanESM2
NFB	Prescribed vegetation simulated from	Transient under RCP8.5	Transient under RCP8.5	1991-2100	CanESM2

	1961 to 1990				
FB_CC	Dynamic	Transient under RCP8.5	Historical until 2005 and constant afterward	1991-2100	CanESM2

Notes: a, using equivalent atmospheric CO_2 concentration; b, using actual atmospheric CO_2 concentration.