Responses to Anonymous Reviewer 1

General Comment:

The impact of lateral boundary forcing in the CORDEX-Africa ensemble over southern Africa by Karypidou et al.

This manuscript attempts to answer the challenging question regards the extent to which RCM simulations can reduce biases present in GCM simulations for the regional climate of southern Africa. The manuscript is well-written with clearly documented methods, a well-justified aim, and appropriate figures.

There are a few minor and major queries or difficulties I had in understanding these results, which I detail here.

RESPONSE: We would like to thank the Anonymous Reviewer #1 for the positive interpretation of the manuscript. Based on the suggestions and comments, we provide the following replies.
Major Comments

1st Comment:

Variance analysis presented in Fig.8 and 13.: These results are crucial to the stated aims of the manuscript. However, I am left wondering to what extent the physical interpretation made in this analysis is undermined by the low number of RCM members (three) along with a sensitivity to total rainfall in the following way:

1. Variance is constrained to be lower in October than January. Indeed, much of southern Africa only experiences full onset of rains by November. So, any intermodel variability is constrained by total rainfall. This seems reflected by Fig. 8 C and D, where GCM variance is higher in the respective region’s wetter months. Similarly, in October in the future [Fig. 13] all variance is very low, reflecting the lack of rain in this month.

2. The behaviour set out in a. would be seen in a well-sampled system (e.g. 12 GCMs) but with 3 RCMs there the is a high risk that either they all look the same and there is no variance or there is one that is very different and the variance is substantial. This happens here as CCLM4 is biased dry in October with the others two biased wet.

3. The behaviour in b. is strongly dependent on the behaviour of only one RCM. If that RCM comes in line with the others later in the season, as CCLM does with similar wet/dry bias patterns to the other RCMs in January, the variance is much lower. Indeed, in region B and D it is surprisingly almost zero in some months.

4. Part of this query may be unpicked if the rainfall is standardized per month in order to remove problem (a). I’m not sure if this will address b and c though.

RESPONSE: Thank you for all the points raised! We provide the answers below.

Following Vautard et al. (2021) we apply a common standardization of precipitation variance across all months and all subregions examined. We do that in order to place precipitation variances emanating from RCMs and GCMs on a scale that would range from 0 to 1, and thus provide easily comparable results for all subregions and all months. A schematic of the method we use for calculating inter-RCM and inter-GCM variances is displayed in Figure S1 of Supplementary material. We do agree that the fact that October is much drier than January and the fact that the common standardization is applied for all months, results in variance being constrained for October. However, our main aim through these plots (Figures 8 and 13, now Figures 9 and 14) is to distinguish between RCM or GCM dominated precipitation variances. Our primary goal is not to make a statement about the actual value of standardized precipitation variance, but rather to examine our hypothesis of RCMs dominating precipitation signal in the early rainy season (ON) and GCMs dominating precipitation in the core rainy season (DJF). The actual magnitude of precipitation variance will be affected by the way standardization is performed; however, it would be problematic and therefore challenge our initial hypothesis, had the standardization method affected the classification of each month to GCM or RCM dominated regimes. In fact, we have now performed a standardization in which minimum and maximum precipitation variance are performed for each specific month separately. The results are displayed below. The two main
conclusions are that October and November remain on the upper triangle (RCM dominated regime), while Dec-Mar remain on the lower triangle (GCM dominated regime), as in the original plot contained in the manuscript. Of course, the new standardized values of precipitation variance are changed, since standardization is performed for each month separately, and each month in concern displays the highest variance (1) in each respective triangle. Because standardization using October minimum and maximum values was very low, the rest of the months do not appear in the respective month’s panel. Same is the case for November and December.

We also note that standardization of precipitation variances was performed using the following formula:

\[
\text{Norm} = \frac{x - \min(x)}{\max(x) - \min(x)}
\]

\(x\): Precipitation variance for each month

\(\min(x)\): Minimum precipitation variance for all months for all subregions

\(\max(x)\): Maximum precipitation variance for all months for all subregions

In addition, and with regards to the fact that variance analysis is not well-sampled (only three RCMs participate in the CORDEX-Africa ensemble) we agree that this is far from optimal, however, it is a necessary compromise imposed by data availability. There are statistical workarounds in “filling-the-gaps” in the GCM-RCM simulation matrices, however, they also come with a set of considerable deficiencies (Christensen and Kjellström, 2022).

2. Comment:

Based on my understanding of recent literature for the region, I disagree with the interpretations about the regional climate drivers.

1. Munday & Washington 2018 demonstrated the heat low to tropical low switch of the Angola Low. Howard & Washington showed the tropical Angola low was in fact the monthly aggregate of frequent tropical depressions crossing southern African from Mozambique and stalling in Angola. This is at odds with the interpretation provided in line 143-145, where the Angola Low is viewed separately.

2. In the Heat Low phase, the Angola Low is not directly driving rainfall. It cannot because it can only develop under subsiding clear-sky conditions. Rainfall in the early season happens when the heat low is temporarily displaced/dissolved. The leading candidate for this displacement is large-scale synoptic westerly waves. See c.

3. Work from the early 1990s by D’Abreton and picked up by others, including recently Hart et al 2018, suggested southern African rainfall is controlled by mid-latitude westerly wave dynamics (large-scale) earlier in the season. This then gives way to more local processes later in the season as the moist thermodynamic environment becomes more tropical (less subtropical) by the height of summer. Your speculation (Line 25-28) counters this, which is fine, but see d. below.

4. Speculation about land-surface coupling seems key to your argument, but at least as far as I am aware this is not well-established for early season over southern Africa. Please include references which point to this if you have them available. I am not sure if the literature, as yet, has shown that for example the soil-moisture – rainfall coupling seen in Indian, and the Sahel does play a role in southern Africa. And it is an open question whether this is true in the real-world, let alone whether it is resolved in RCMs.

RESPONSE: Thank you for this comment and all the issues raised. We acknowledge their importance to the theoretical assumptions of our work and address it point-by-point below.

1. We agree with this statement. We also consider the Angola Low as a climatic feature that switches from the heat low phase at the early rainy season to the climatological aggregate of transient depressions during DJF. The phrasing over lines 143-145 (line numbering prior to revision) is now corrected to the following: “Since precipitation during Dec-Feb is caused by the tropical low phase of the Angola low pressure system, which is the monthly aggregate of frequent transient low pressure systems crossing southern African (Munday and Washington, 2017; Howard and Washington, 2018; Howard et al., 2019), we hypothesize that the impact of the driving GCM fields during Dec-Feb is enhanced”.

2. We do recognize the importance of large-scale synoptic westerlies; however, the CORDEX-Africa ensemble does not allow a detailed analysis of the properties of the upper-level westerly flow because of the geographical extent of the CORDEX-Africa domain
More specifically, the southernmost boundary of the CORDEX-Africa domain is placed at 44°S, which limits considerably the window over which upper-level westerlies can be analyzed. Such an effort has been made in Karypidou, (2022) (Figures 5.4.6 – 5.4.7), however, the “landscape” over which upper-level westerlies were analyzed was extremely limited. Moreover, the CORDEX-Africa ensemble simulations do not use an ocean model coupled to the atmospheric component of the RCM; prescribed SST’s only are used by RCMs. Considering that, it would be problematic to analyze westerly winds mainly blowing over the southern hemisphere oceans and sporadically crossing over land (such as southern Africa or South America, and Australia). Further explanations are provided in [3].


3. We understand this point and find it extremely interesting concerning the impact of the Tropical-Extratropical (TE) cloud bands on precipitation over southern Africa, as analyzed in Hart et al., (2018) (with background fundamental work being described in D’Abreton and Lindesay, (1993)). As stated in Hart et al. (2018) “The seasonality of TE cloud band likelihood emerges from the finely tuned interaction between the asynchronous seasonal cycles in subtropical upper-level westerlies and lower-tropospheric instability.” Also, as it is stated in Howard and Washington (2018), the Angola Low pressure system can be considered as a precursor to the Tropical Temperate Troughs (TTTs), as “The Angola low enables southward transport of atmospheric water vapor from the tropics, crucial to the development of TTTs”. We consider that the Angola low pressure system provides the necessary lower-tropospheric instability required for the formation of the TE cloud bands analyzed in Hart et al. (2018). However, since the upper-level westerly flow is available almost throughout the whole year (with variations in intensity and latitude of occurrence – Figure 5.4.5 in Karypidou, (2022)), the key agent for rainfall during the early rainy season is the occurrence of low-tropospheric instability (i.e. Angola low). Therefore, although the interplay between upper-level westerly flow and low-level atmospheric instability is crucial for the development of cloud bands, rainfall during the early rainy season can originate from alternative agents. By “alternative agents” we mean again the Angola low pressure system (at its heat low phase) and moisture supply being provided to the Angola region from low-level westerly winds. Considering the work by Howard and Washington (2019), the Congo Air Boundary is crucial in constraining moisture to the northwest part of southern Africa, necessary to fuel early season rainfall over the greater Angola region. The location of CAB significantly affects the amount of rainfall that the northwestern part of southern Africa will experience (“the CAB plays a primary control on the spring and early summer rainfall in southern Africa. Early in the season, more rainfall may occur in this region when the CAB is farther south”). In addition, considering all the technical constraints concerning the geographical extent of the CORDEX-Africa domain, the methodology employed in Hart et al. (2018) could not have been employed here. More specifically, Hart et al. (2018) analyze streamfunctions at 200 hPa and distinguish between eddy-driven jet axes and distinguishable jet axes over latitudes ranging up 70°S. Such analysis could not have been performed in CORDEX-Africa, considering that the southernmost latitude of the CORDEX-Africa domain is 44°S.

Concerning the land-atmosphere coupling over southern Africa, we make reference to the work by Careto et al., (2018), who investigated land-atmosphere coupling metrics within the CORDEX-Africa ensemble (hindcast simulations: ERA-Interim driven). More specifically, the Pearson correlation between the surface upward latent heat fluxes (hfls) and the sensible heat fluxes (hfss) were found to be strongly anticorrelated over the Sahel, southern Africa, and eastern Africa regions, especially during DJF, but also during SON. In Figure 6 of Careto et al., (2018), strong negative correlation values are indicative of strong land-atmosphere coupling. In addition, Careto et al., (2018) introduce the Latent Heat Flux-Temperature Coupling Magnitude (LETCM) metric, which for SON displays its highest values over southern Africa (Figure 7c) and co-occurs with the region of strong negative correlation between hfls ~ hfss (strong coupling situations). Since coupling as quantified using both metrics (correlation hfls ~ hfss and LETCM) is strongest over the Angola region during the early rainy season (SON), and since coupling in climate models is highly dependent on parameterization schemes and coupled model components simulating land processes (Wilhelm et al., 2014), we hypothesize that during the early rainy season RCMs will dominate precipitation signal over southern Africa.


Minor Comments

3rd Comment:

Line 197-198, (line 300 too). Excess surface heating is surely even greater in peak summer months? Furthermore, heating is insufficient for convection when surface environments are moisture limited as they are during October.

RESPONSE: October is characterized by a transitional weather regime namely between the dry and wet season. The sensible heat flux is surely consistently larger during summer peak months but here we point to the following:

(i) What happens in the upper levels of the atmosphere: October being a transitional month, we can expect some of the first intrusion of “colder” air in the upper pressure levels. Higher vertical gradient increases parcels buoyancy, i.e., larger CAPE and the general instability of the atmosphere which allows convective phenomena to occur also (especially) in moisture-limited contexts.

(ii) Soil moisture precipitation feedback: Several works stress how this feedback can be both positive (more precipitation initiation where is wet) or negative (more precipitation initiation where is dry). Moreover, another crucial aspect is the spatial distribution of the soil moisture in the study area but also in the surrounding areas, its “patchiness”, which is able to alter mesoscale circulation and precipitation pattern distribution (Seneviratne et al. 2010; Taylor et al. 2012; Graf et al. 2021). These works clearly show that moisture-limited environments not necessarily trigger negative soil moisture-precipitation feedback.

In synthesis, the upper pressure level circulation and the moisture spatial heterogeneity can trigger convective phenomena also in moisture-limited context.

4th Comment:

Line 208-209: I am not quite sure how to understand this statement about smooth topography. I read this to imply that the topography should be smooth, but is not the point of RCMs to include more detailed “jaggedness” that the real-world topography contains?

RESPONSE: Thank you for this comment! Indeed, the statement as it was initially framed is not accurate and may lead to specific misreading that is not valid for RCMs. RCMs, because of their higher horizontal resolution do represent surface characteristics such as elevation in a more accurate manner. In fact, the improvement of precipitation over the southern Africa region in the CORDEX-Africa ensemble relative to the CMIP5 GCMs was attributed to the fact that orography over the greater Tanzania region was more accurately represented in the CORDEX-Africa ensemble, blocking excess low-level moisture transport from the tropical Indian Ocean from entering mainland southern Africa (Karypidou et al., 2022), as it was the case for the CMIP5 GCMs (Munday and Washington, 2018). The comment in lines 208-209 was referring to RCA4.v1 and not all CORDEX-Africa RCMs used in the current analysis.

The following sentence has now been deleted: “This may be attributed to the fact that the topography is not smooth enough and leads to high precipitation values over grid boxes with high elevation (Van Vooren et al., 2019).”

Instead, the following sentence has now been placed in the text: “This attribute is indicative of specific structural model biases related to how high-resolution elevation affects precipitation in RCA4.v1 (Van Vooren et al., 2019).”
Line 358, 359 linked to line 388. The only truly unambiguous signal, already well-established in literature, is this early season drying for southern Africa. So, these statements about models struggling with transition (October) seem paradoxical with the clear drying signal. In the revision of the manuscript, hopefully this can be rethought and rewritten.

RESPONSE: Thank you for this comment! Indeed, that was a point that was not stated properly and is indeed paradoxical! In the revised manuscript it has now changed to the following (the line references have now changed; however, we refer to what was prior to revision lines 358, 359, and 388.

In lines prior numbered as 358, 359, the following sentence has been omitted: “November is the month during which there is a transition of the AL from a heat low phase to a tropical low system, and March indicates the end of the rainy season. Hence, precipitation during the transition months is challenging for both RCMs and GCMs.”

This part has now been added instead: “The Angola region, which encompasses the activity of the Angola Low pressure system, displays the highest wet biases with regards to mean monthly precipitation, among all subregions examined. The months with the largest wet biases (for the Angola region) is found to be November, while the month with the largest precipitation bias spread is found to be March. In all months except of October, the CMIP5 GCMs display biases that are approximately 1-1.5 mm/d wetter than the wets test CORDEX-Africa RCM ensemble members.”

Statement in line 388 has been left unchanged since it is not in contradiction to what has been stated in lines above.

The following figure has now been added to the Supplementary material as Figure S10. It displays monthly precipitation biases averaged over the whole southern Africa region (SAF-All) and the three subregions examined, namely the Angola region, East Coast regions, and the S Afr region.

The text in which reference is made to Figure S10 is the following: “Monthly precipitation biases averaged over southern Africa (SAF-All) and the three subregions examined are displayed in Fig. S10.”
Fig. S10 Spatial average of precipitation bias (mm/d) from RCMs and their driving GCMs over southern Africa and the three sub-regions examined.
6th Comment:

Line 383 talks about observation products being kept in sight. I suspect you mean, in mind. Taking this quite literally, Kendon et al 2019 made this observation uncertainty visible by display TRMM-CMORPH bias alongside other bias or change plots (panel d in Figures 2-6). Including such a figure in your manuscript would help stay true to your line 383 statement and give the reader and indicate of magnitude of changes relative to obs. uncertainty.

RESPONSE: Thank you for this comment! Yes, we mean “observation products being kept in mind”. We have now added an additional figure, in which precipitation climatology for the rainy season months (Oct-Mar) is shown for: ERA5, CHIRPS, CRU, and MSWEP. In addition, we make reference to Karypidou et al., (2022), in which precipitation uncertainty is investigated in detail among five gauge based products (datasets that are derived by spatial interpolation of rain gauges and station data: CRU.v4.01, UDEL.v7, PREC/L.v0.5, GPCC.v7, CPC-Global.v1), six satellite products and ERA5 (Figures 1, 2, 3 and Figures S10, S11).

The following has now been added in the Data section:

“A fact that is commonly obscured is that observational datasets are often considered as “ground truth” however, they also are subject to multiple sources of uncertainty, caused by the underlying station datasets used, the statistical algorithms employed in spatially interpolated methods or the algorithms employed in satellite rainfall products (Le Coz and van de Giesen, 2020). More specifically, over southern Africa, it was found that gauge-based products employing spatial interpolation methods displayed high uncertainty over regions where the underlying station network was scarce, mainly over the Angola region and the northern parts of SAF (Karypidou et al., 2022). In addition, it was found that this attribute was inherited by all rainfall satellite products that were using direct merging techniques with gauge-based datasets. Here, we display monthly precipitation during the historical period (1985-2005) across four observational datasets, given in Table 1. More specifically, we use the CRUv4.06 dataset (Harris et al., 2020), which is a purely gauge-based product (employing station data and a spatial interpolation algorithm to provide a spatially continuous gridded product), ERA5 (Hersbach et al., 2020), which is a reanalysis product, CHIRPS (Funk et al., 2015), which is a satellite rainfall product, and finally, MSWEP (Beck et al., 2017) which is a product merging station data, satellite data and dynamic model outputs. All datasets have been analyzed using monthly mean values. The results are displayed in Fig. 1. As shown, there is a substantial agreement among them both with regards to the spatial and temporal pattern of monthly precipitation over southern Africa.”
Table 1 Gauge-based, satellite, reanalysis and merged precipitation products analyzed over the study region using monthly mean precipitation for the period 1985-2005.

<table>
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<th>Dataset</th>
<th>Resolution</th>
<th>Frequency</th>
<th>Type</th>
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<td>Monthly total</td>
<td>Gauge-Based</td>
<td>1901-2021</td>
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<tr>
<td>MSWEP</td>
<td>0.1°</td>
<td>3-hourly</td>
<td>Merged product</td>
<td>1979-present</td>
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<td>CHIRPS.v2</td>
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<td>Daily totals</td>
<td>Satellite</td>
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<tr>
<td>ERA5</td>
<td>~0.25°</td>
<td>Hourly</td>
<td>Reanalysis</td>
<td>1979-present</td>
</tr>
</tbody>
</table>

Figure 1. Monthly mean precipitation climatology for the period 1985-2005.
7th Comment:

For clarity remove region letters in text and figures. Just go with SAF-All, Angola, East Coast, SAfr, or similar throughout to make for easier reading.

RESPONSE: Thank you! We have done so in all text and figures.
8th Comment:

Bias plots will be much clearer to interpret if expressed in % bias from climatology (with mask for negligible rainfall areas, e.g. <1mm/month) and if colours were white for low/no bias increasing to dark red (dry) or blue (wet).

RESPONSE: Thank you for this suggestion. Indeed % bias helps to clearer image of bias, however, percent bias is tricky with very small values. Reviewer suggests we mask areas with negligible rainfall to bypass this issue. However, very large areas during October experience rainfall <1 mm/d. During November and March there are also large areas with values <1 mm/d or marginally larger than this threshold, so eventually large regions are masked from the panel maps – and the masks are variable among the months of the rainy season. For this reason, we chose to display biases as the difference model-obs.
9th Comment:

A paper you may have missed by which is relevant to some of your analysis here is Munday, C., & Washington, R. (2019). Controls on the diversity in climate model projections of early summer drying over southern Africa. Journal of Climate, 32(12), 3707-3725.

RESPONSE: Thank you very much for mentioning this paper! We have now included it in our analysis. More specifically, the following portion has now been added in the 5th paragraph of the Discussion and conclusions section (new sentences are indicated in bold):

“Concerning the climate change signal, there is a strong agreement among all GCMs and RCMs that precipitation during October will decrease by (-0.1) – (-1) mm/d, a fact associated with a projected later onset of the rainy season, which is further linked with a northward shift of the tropical rain belt (Dunning et al., 2018; Lazenby et al., 2018). The topic of reduced early rainfall over southern Africa for the end of the 21st century under all emission scenarios/pathways has been examined extensively for the CMIP3 and CMIP5 GCM ensembles (Seth et al., 2011; Cook and Vizy, 2021; Lazenby et al., 2018; Howard and Washington, 2019b). A common observation in all CMIP5 GCMs for the early rainy season by the end of the 21st century is that instability over southern Africa reduces, surface temperature increases, and the heat low phase of the Angola Low pressure system is strengthened (Howard and Washington, 2019). However, rainfall decline in the CMIP5 ensemble over southern Africa should be additionally considered in the context of the systematic precipitation biases already diagnosed in the historical simulations (Munday and Washington, 2018; Howard and Washington, 2019). Considering that the systematic wet precipitation bias is significantly reduced in the CORDEX-Africa ensemble relative to their driving CMIP5 GCMs (Karypidou et al., 2022), we gain confidence that future precipitation projections according to the CORDEX-Africa ensemble provide a more plausible future scenario. For the rest of the months, the results are variable, indicating the need for a multi-model approach, when climate change impacts are assessed. A feature that is identified in some GCMs and is transferred to the downscaling RCMs, is a precipitation increase that extends from the central SAF region towards the southeast. This result is consistent with previous work that shows an increase in frequency of landfalling cyclones along the eastern seaboard of SAF (Muthige et al., 2018). Since tropical cyclones are a particular cause of severe flooding events over the region of Mozambique, there is an urgent need for planning and mitigation strategies over the region.”
References:

