

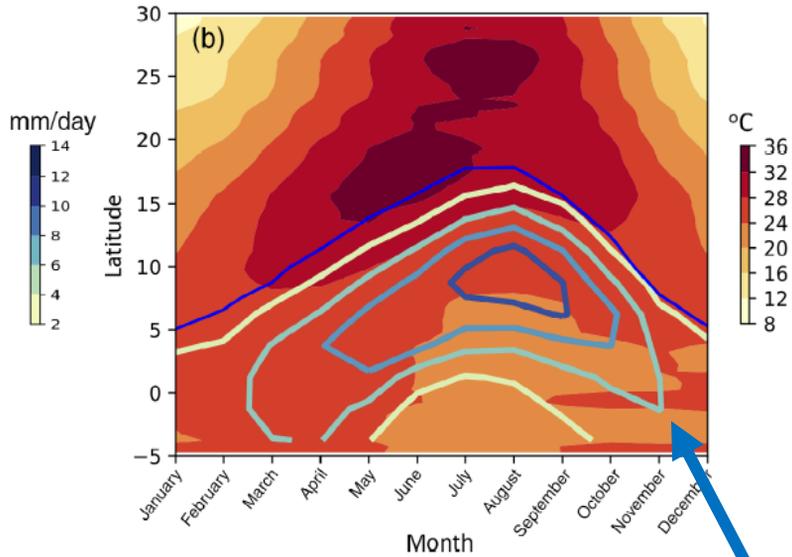
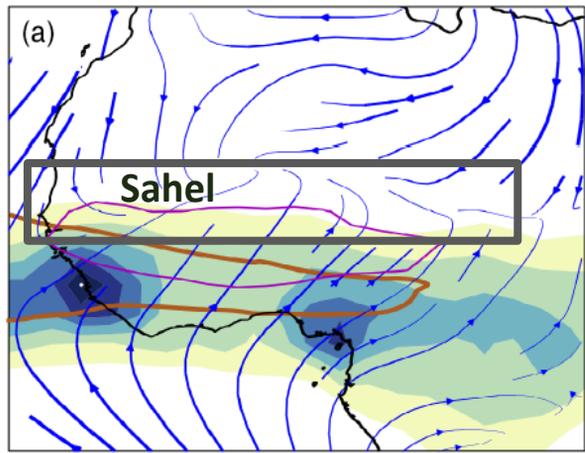
# Model uncertainty in future changes in West African precipitation: component replication

Paul-Arthur Monerie<sup>1</sup>, Ana Arama<sup>1</sup>, Julien Boé<sup>2</sup>

<sup>1</sup>University of Reading/NCAS, Reading, UK

<sup>2</sup>CERFACS/CNRS, Toulouse, France

# The West African Monsoon



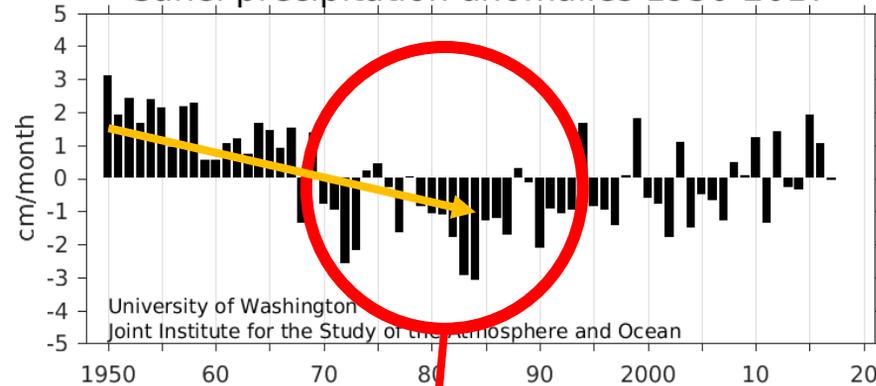
Precipitation

[Left] Precipitation (May-October colour), and 925 hPa wind (arrows).  
[Right] Precipitation (contours) and zonal mean temperature (colour) [20°W-30°E].

From Biasutti 2019

# Motivations

## Sahel precipitation anomalies 1950-2017



Strong variability in Sahel precipitation

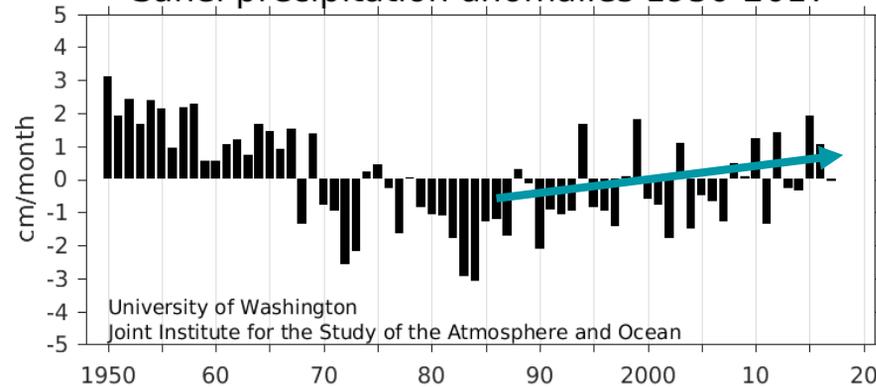
June through October averages over 20°-10°N, 20°W-10°E. 1950-2017 climatology  
Deutscher Wetterdienst Global Precipitation Climatology Centre data

1970's-1980's : The drought



# Motivations

Sahel precipitation anomalies 1950-2017



June through October averages over 20-10°N, 20°W-10°E. 1950-2017 climatology  
Deutscher Wetterdienst Global Precipitation Climatology Centre data

A recent positive trend, the  
“recovery” period

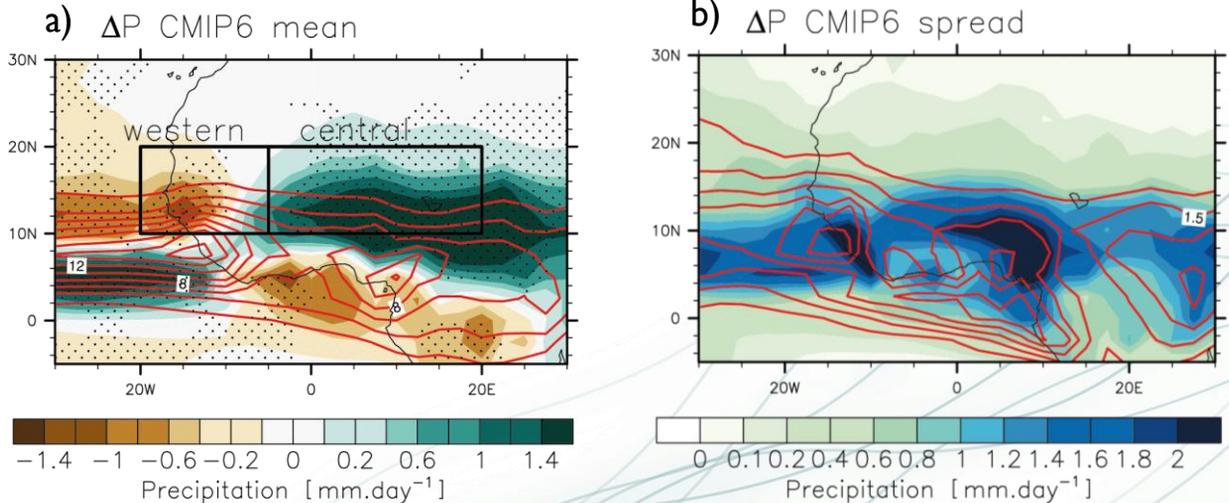
*Nicholson et al. 2013*

**Future?**

The **West African Monsoon** brings precipitation to around 80 million people from Senegal to Chad, in summer.

- **Human health** .e.g., Jankowska et al. 2012; Cissé 2019
- **Agriculture** .e.g., Sultan and Gaetani 2016
- **GDP** .e.g., Sainte Fare Garnot et al. 2018; Baarsch et al. 2020
- Among others

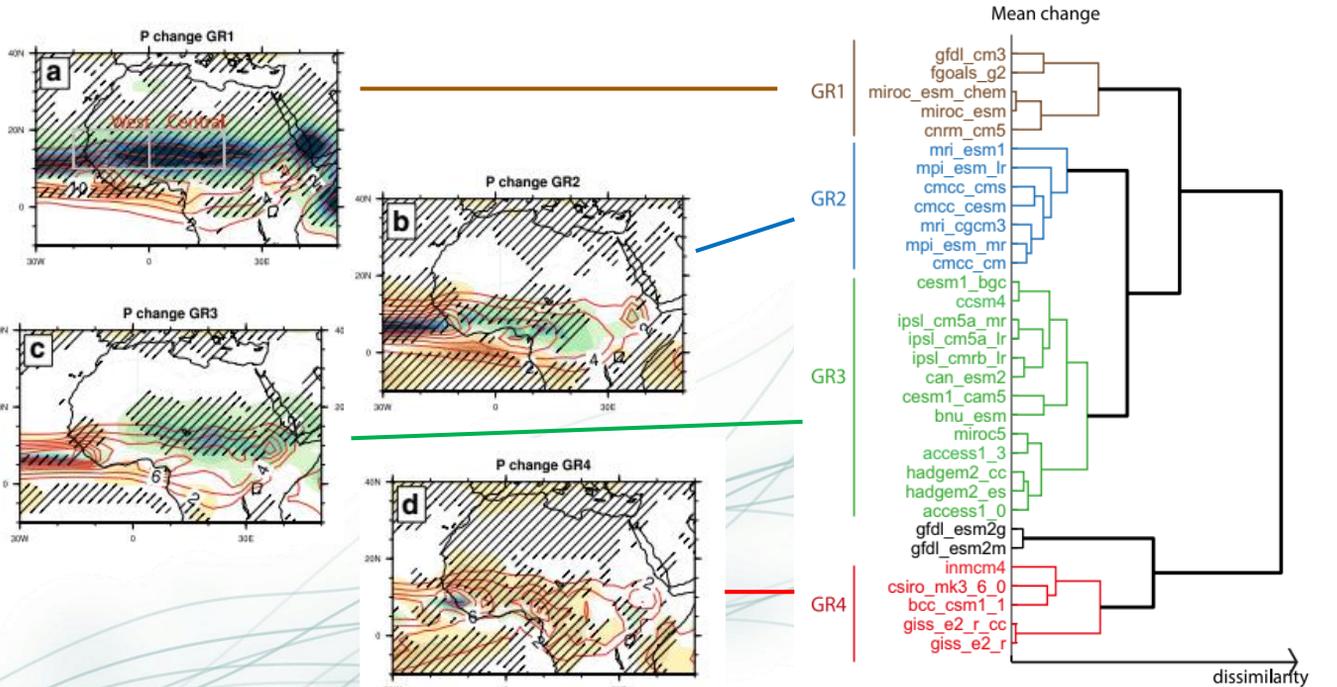
# Future changes in precipitation are uncertain



*Effect of climate change on precipitation [ $\text{mm}\cdot\text{d}^{-1}$ ] for an ensemble of CMIP6 simulations: (a) ensemble mean and (b) ensemble spread (uncertainty). The effect of climate change on precipitation is defined as the difference between a future period (2060-2099; SSP5-8.5) relative to an historical period (1960-1999) in summer (JAS). Monerie et al. 2020.*

A zonal contrast in precipitation change  
Changes are uncertain

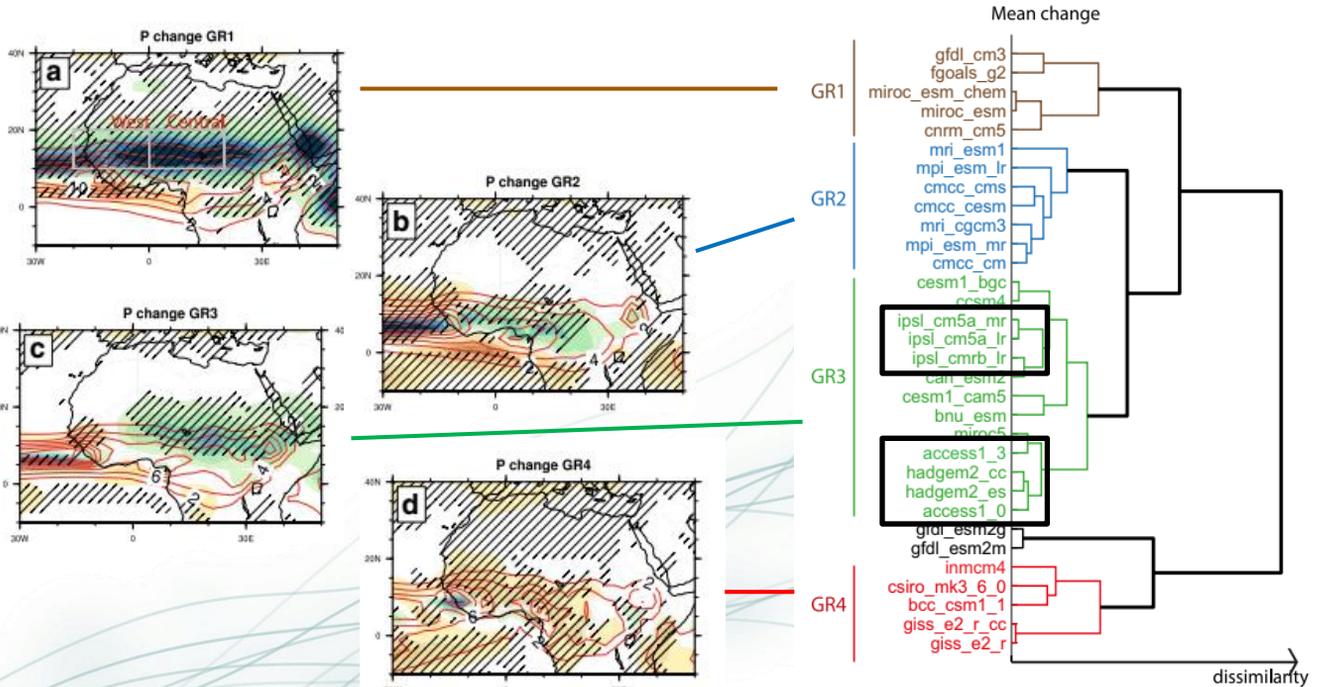
# Future changes in precipitation are uncertain



Monerie et al. 2017

The different trajectories in Sahel precipitation change for the end of the 21<sup>st</sup> century, emissions scenario RCP8.5

# Future changes in precipitation are uncertain



Monerie et al. 2017

The different trajectories in Sahel precipitation change for the end of the 21<sup>st</sup> century, emissions scenario RCP8.5

**Role of the structural uncertainty?  
(How do we account for model independence?)**

# Data

We use the data from **39 CMIP6 models**, under the SSP5-8.5 emissions scenario. One ensemble member for each model.

We further assess the effects of **internal climate variability** using outputs from four Single Model Initial-condition Large Ensembles (between 30 and 50 ensemble members).

## Similarity between models

We computed the Root Mean Square Error between simulations of different models

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (X_i - Y_i)^2}$$

For two models, X and Y, and each grid point i, over West Africa [20°W-20°E; 5°S-25°N]

# Components selection

Model	Atmos.	Ocean.	land.	Sea ice.
ACCESS-CM2	GA7	MOM5	CABLE2	CICE5
ACCESS-ESM1-5	GA7	MOM5	CABLE2	CICE4
HadGEM3-GC31-MM	GA7	NEMO3	JULES7	CICE4
INM-CM5-0	AM5	OM5	LNDI	ICEI

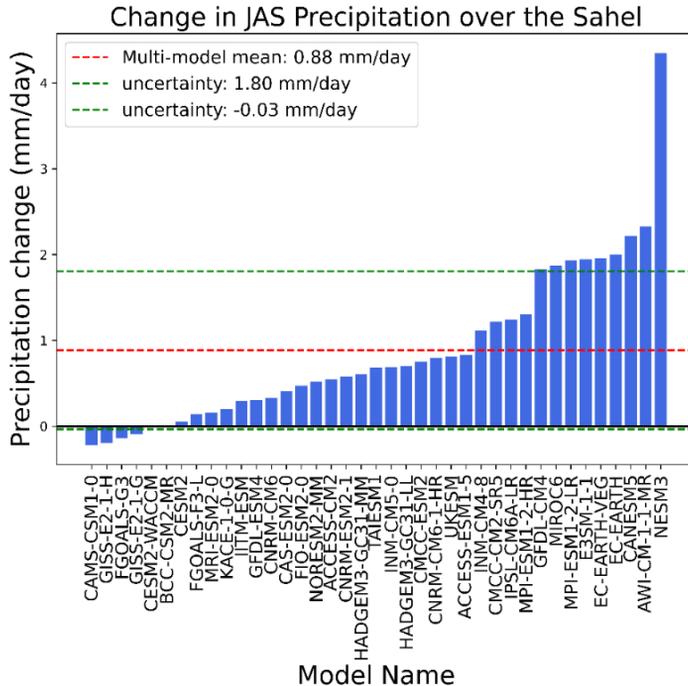
3  
components  
in common

1  
components  
in common

No  
component  
in common

Following Boé et al. 2018

# Results: changes in Sahel precipitation



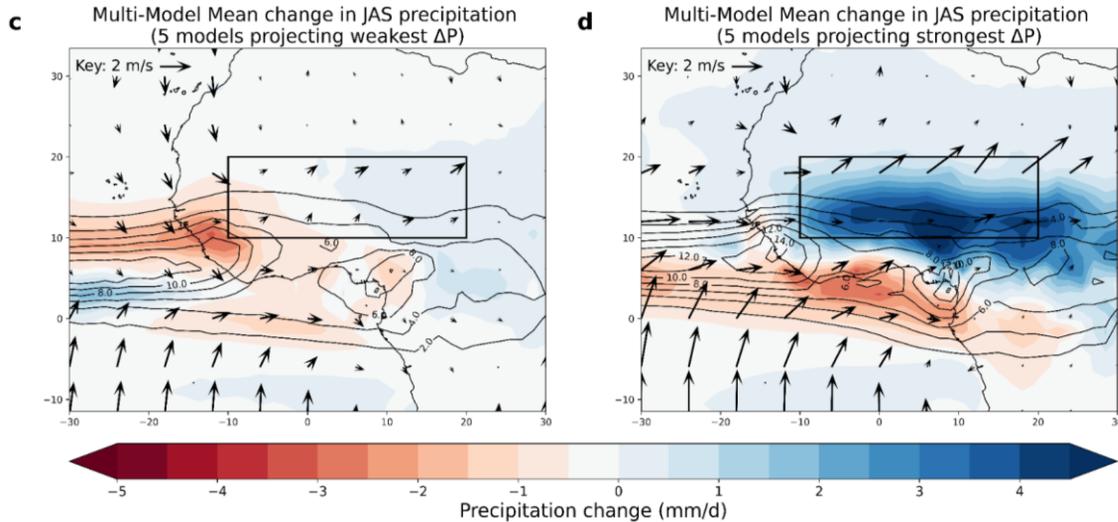
Increase in precipitation over the Sahel, but results are model-dependent

Multi model mean

Uncertainty (spread)

*Effect of climate change on precipitation [mm.d<sup>-1</sup>] for an ensemble of CMIP6 simulations (2060-2099; SSP5-8.5) relative to an historical period (1960-1999) in summer (JAS).*

# Results: changes in Sahel precipitation

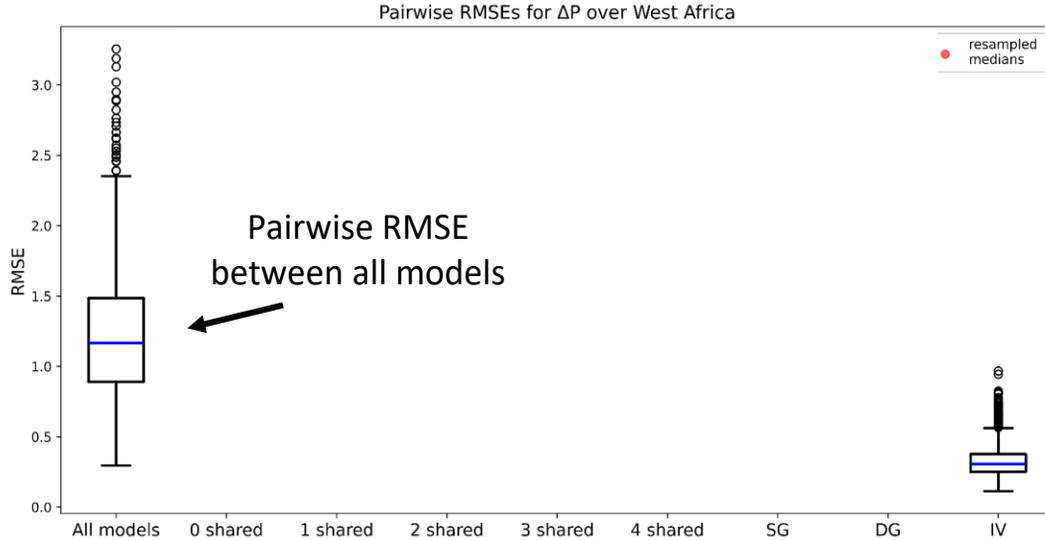


*Effect of climate change on precipitation [ $\text{mm}\cdot\text{d}^{-1}$ ] for an ensemble of CMIP6 simulations (2060-2099; SSP5-8.5) relative to an historical period (1960-1999) in summer (JAS), using (left) the five models that show the more negative/smallest change in precipitation and (right) the five models which show the strongest increase in precipitation. The arrows show the 850 hPa wind. The contours show the climatology.*

Two different futures

# Results: the component replication

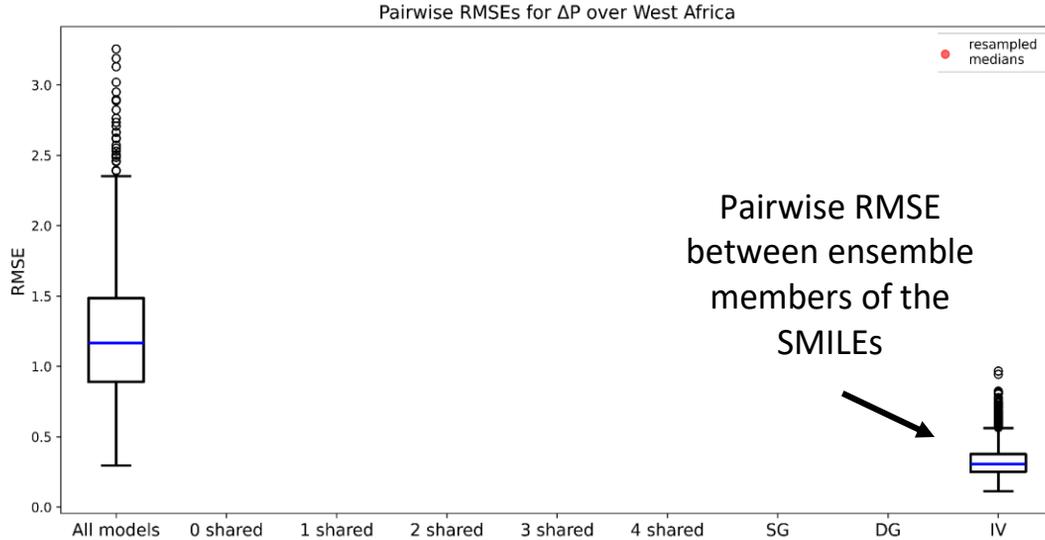
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: the component replication

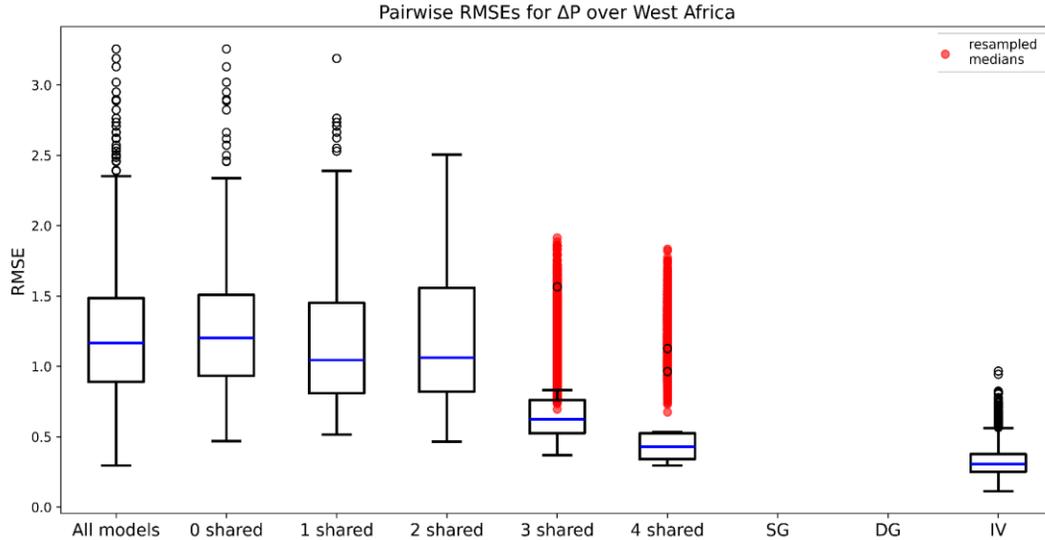
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: the component replication

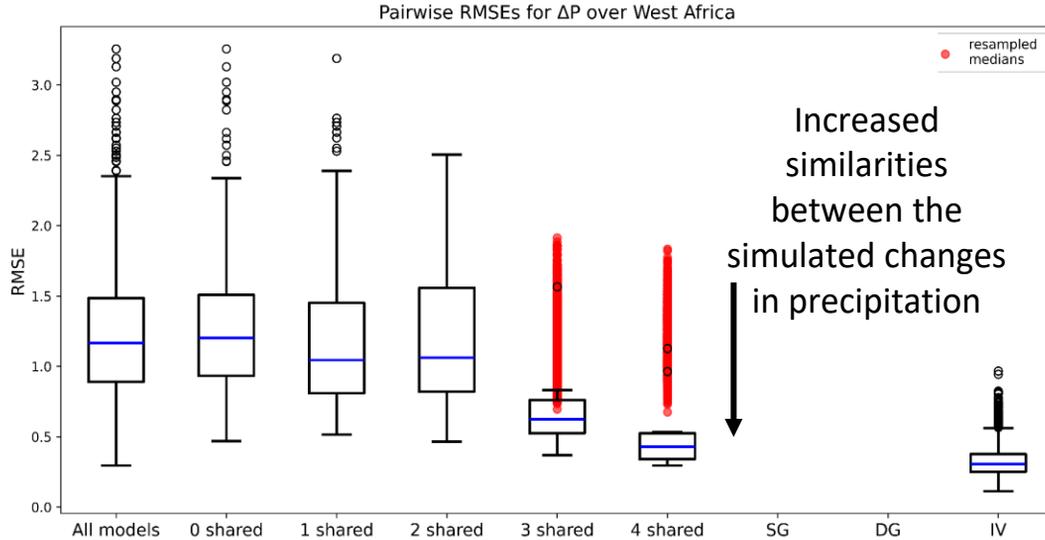
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: the component replication

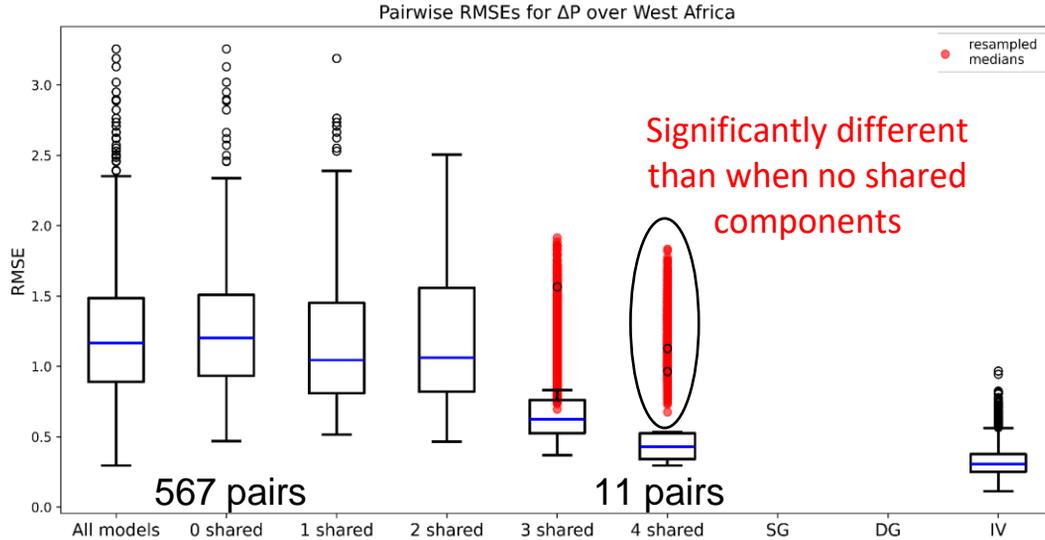
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: the component replication

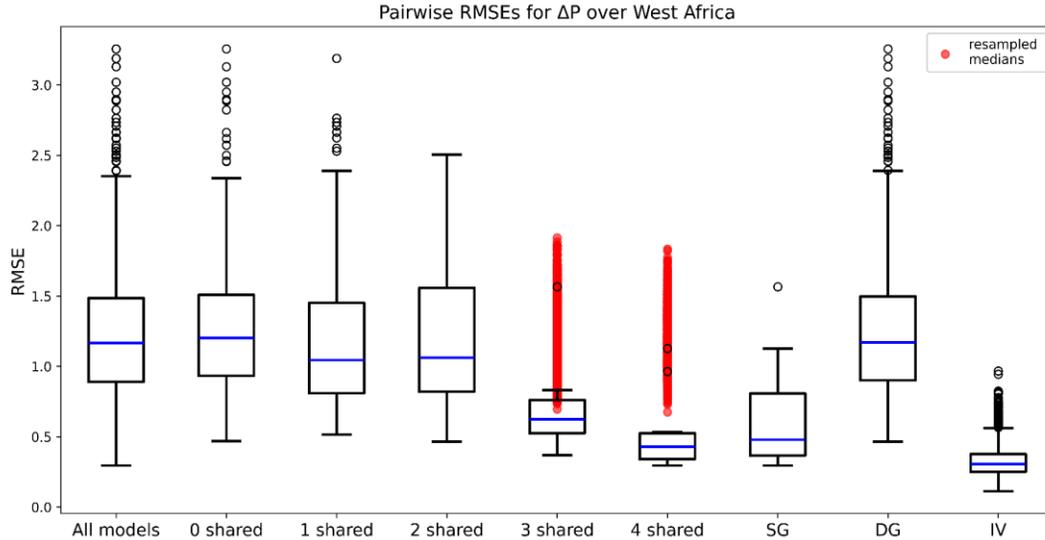
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: the component replication

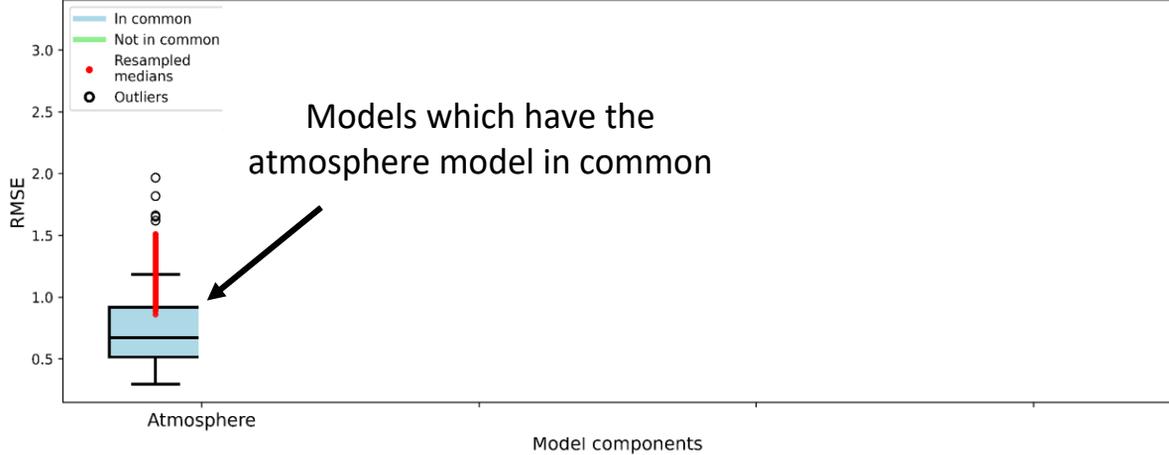
a



Pairwise RMSEs for (a) precipitation change and (b) model bias over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. In both (a) and (b), SG and DG show the pairwise RMSEs when model pairs are in the same modelling group and different modelling groups respectively. The boxplot denoted 'IV' represents the pairwise RMSEs between multiple ensemble members from four different models. This represents internal variability. The red dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the dark blue line.

# Results: by components

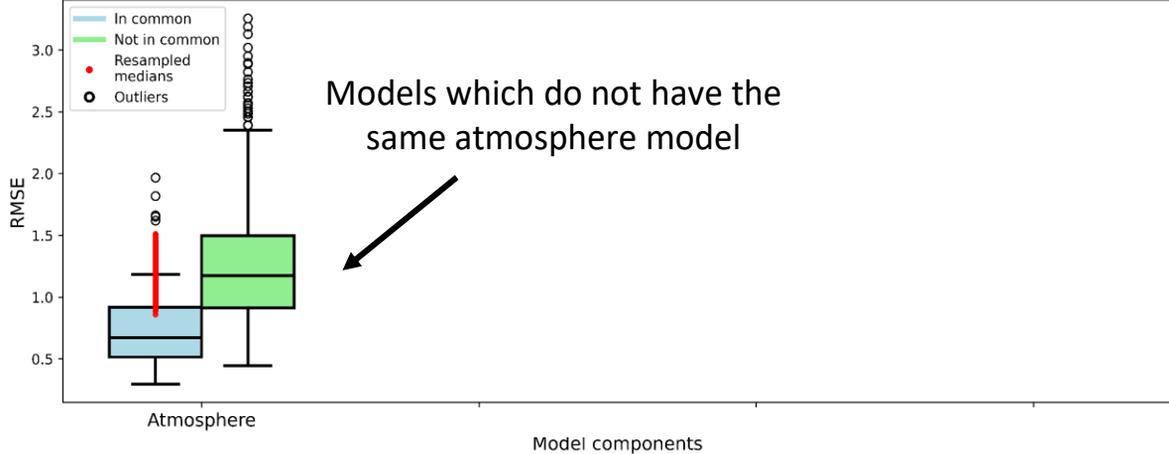
Pairwise RMSEs for JAS  $\Delta P$  over West Africa when models do and do not have specific model components in common



Pairwise RMSEs for change in land precipitation over West Africa when models do and do not have atmosphere/ocean/land/sea ice model components in common. The light blue boxes indicate the samples when a specific model component was in common whereas the light green boxes indicate samples when a specific model component was not in common. The red dots on both plots show the result of random resampling outlined in data/methods section.

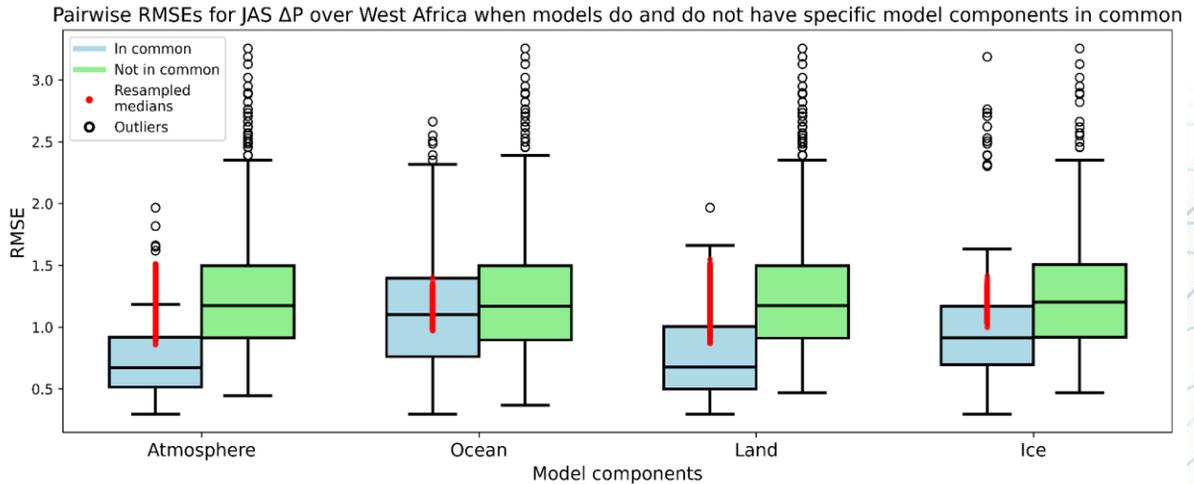
# Results: by components

Pairwise RMSEs for JAS  $\Delta P$  over West Africa when models do and do not have specific model components in common



Pairwise RMSEs for change in land precipitation over West Africa when models do and do not have atmosphere/ocean/land/sea ice model components in common. The light blue boxes indicate the samples when a specific model component was in common whereas the light green boxes indicate samples when a specific model component was not in common. The red dots on both plots show the result of random resampling outlined in data/methods section.

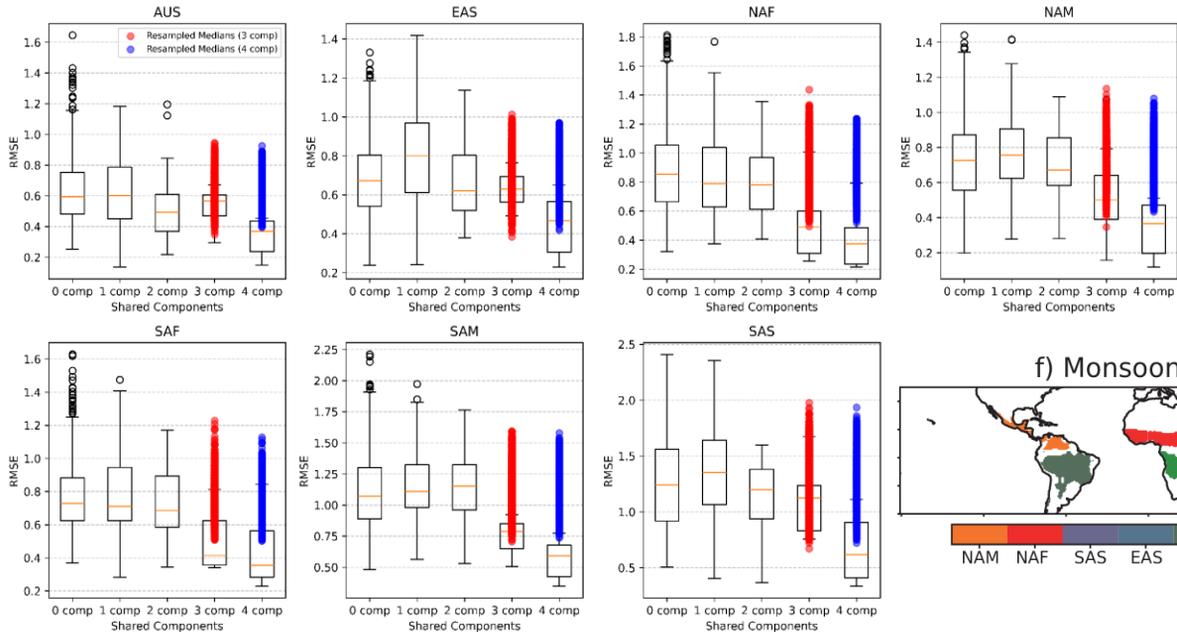
# Results: by components



Pairwise RMSEs for change in land precipitation over West Africa when models do and do not have atmosphere/ocean/land/sea ice model components in common. The light blue boxes indicate the samples when a specific model component was in common whereas the light green boxes indicate samples when a specific model component was not in common. The red dots on both plots show the result of random resampling outlined in data/methods section.

**Strong effect of the atmosphere and land models**

# Results: other monsoon domains



*Pairwise RMSEs for (a) precipitation change over West Africa [20°W-20°E; 5°S-25°N] based on the number of shared model components. The red/blue dots on both plots show the result of random resampling outlined in data/methods section. The median for each boxplot is shown by the orange line.*

Results are monsoon-domain dependent

# Take home message

The future change in Sahel precipitation is **uncertain**.

- direction/magnitude
- strong model uncertainty

# Take home message

The future change in Sahel precipitation is **uncertain**.

- direction/magnitude
- strong model uncertainty

We show that similarities between **model components** matter

- having at least 3 components in common increases similarity
- the use of a same atmosphere and land models increases similarities between projections
- the effect is primarily associated with the change of the atmospheric circulation

# Take home message

The future change in Sahel precipitation is **uncertain**.

- direction/magnitude
- strong model uncertainty

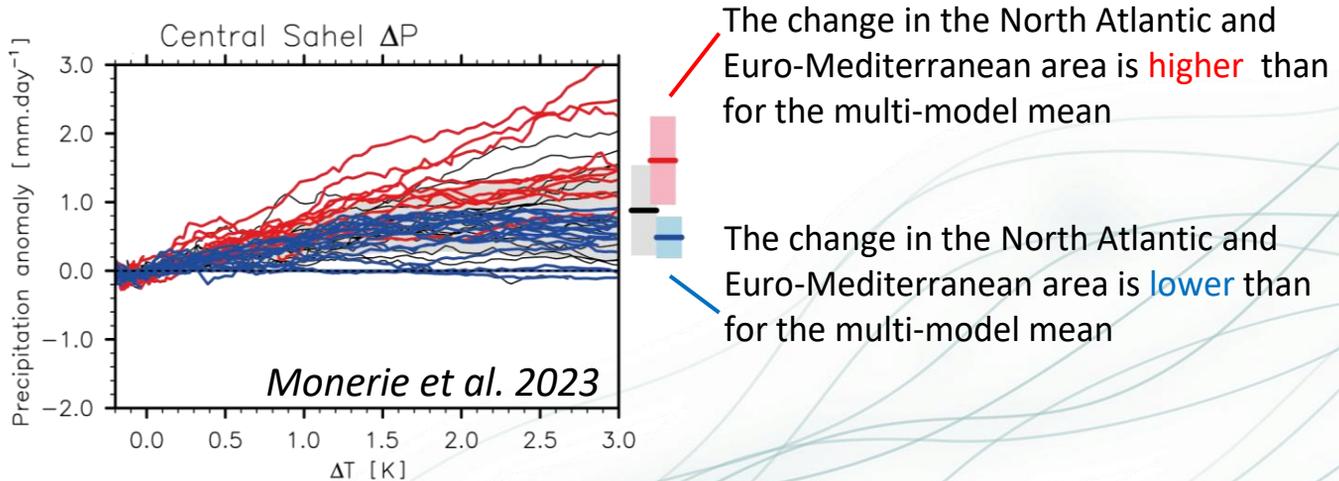
We show that similarities between **model components** matter

- having at least 3 components in common increases similarity
- the use of a same atmosphere and land models increases similarities between projections
- the effect is primarily associated with the change of the atmospheric circulation

Our results have implications for accounting for **model independence** (e.g. to perform the ensemble mean) and **model selection** (e.g. impact studies).

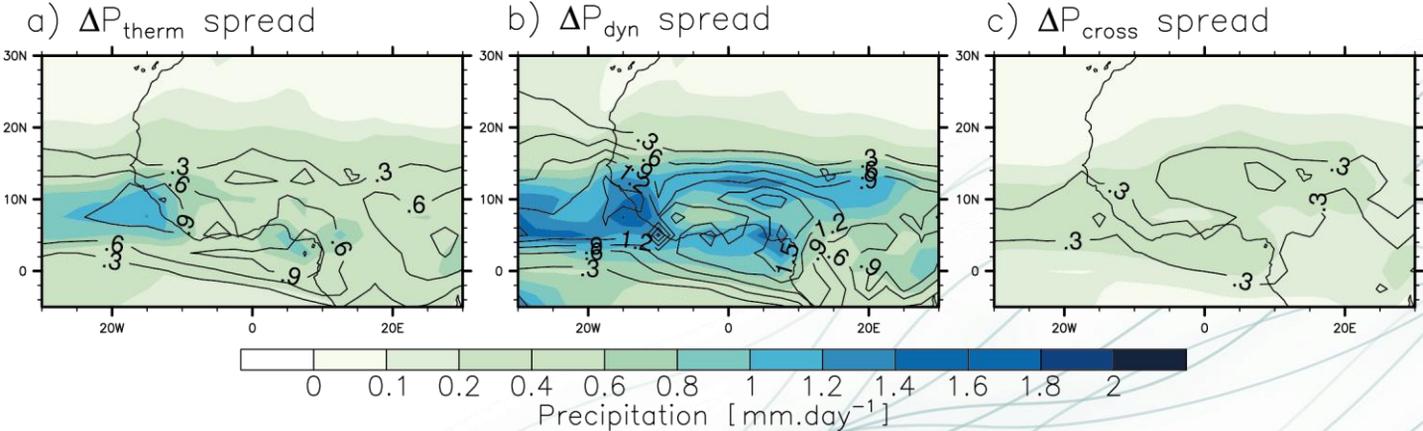


# Future changes in precipitation are uncertain



Changes in central Sahel precipitation [ $\text{mm}\cdot\text{day}^{-1}$ ] as a function of global-mean warming ( $\Delta T$ ) for all models (black lines), the CMIP6 envelope (gray shading, spanning 2 times the CMIP6 standard deviation), the A+M+ models (red), and the A-M- models (blue). On the right-hand side of each panel, the horizontal thick lines indicate the ensemble mean, and the bars indicate the ensemble envelope (spanning 2 times the ensemble standard deviation), computed across the CMIP6 models (black), the A+M+ models (red) and the A-M- models (blue), for a warming of 2.5–3°C.

# Future changes in precipitation are uncertain



*Ensemble spread (uncertainty) in the (a) thermodynamic, (b) dynamic, and (c) cross non-linear change in precipitation.*

*Monerie et al. 2020*

The uncertainty mainly arises from differences between models in simulating future changes in atmospheric circulation.

# Partitioning precipitation

$$P = M^*q \quad (\text{and } M^*=P/q)$$

where  $P$  is the precipitation

$q$  is the near surface specific humidity

$M^*$  is a proxy for convective mass flux from the boundary layer to the free troposphere

*Held & Soden (2006)*

We make the assumption that precipitation can be approximates from the circulation and the near surface humidity

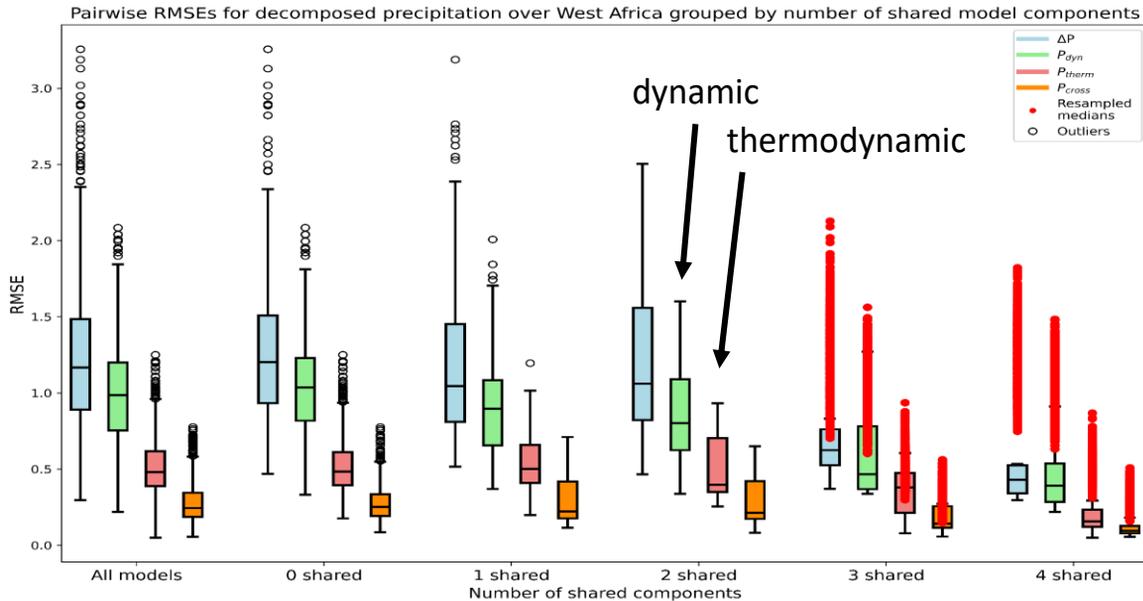
$$\Delta P = M^*\Delta q + q\Delta M^* + \Delta q \Delta M^*$$

*Kent et al. (2015); Chadwick et al. (2016); Rowell and Chadwick (2018)*

$$\Delta P = \Delta P_{therm} + \Delta P_{dyn} + \Delta P_{cross}$$

thermodynamic                      dynamic                      cross non-linear

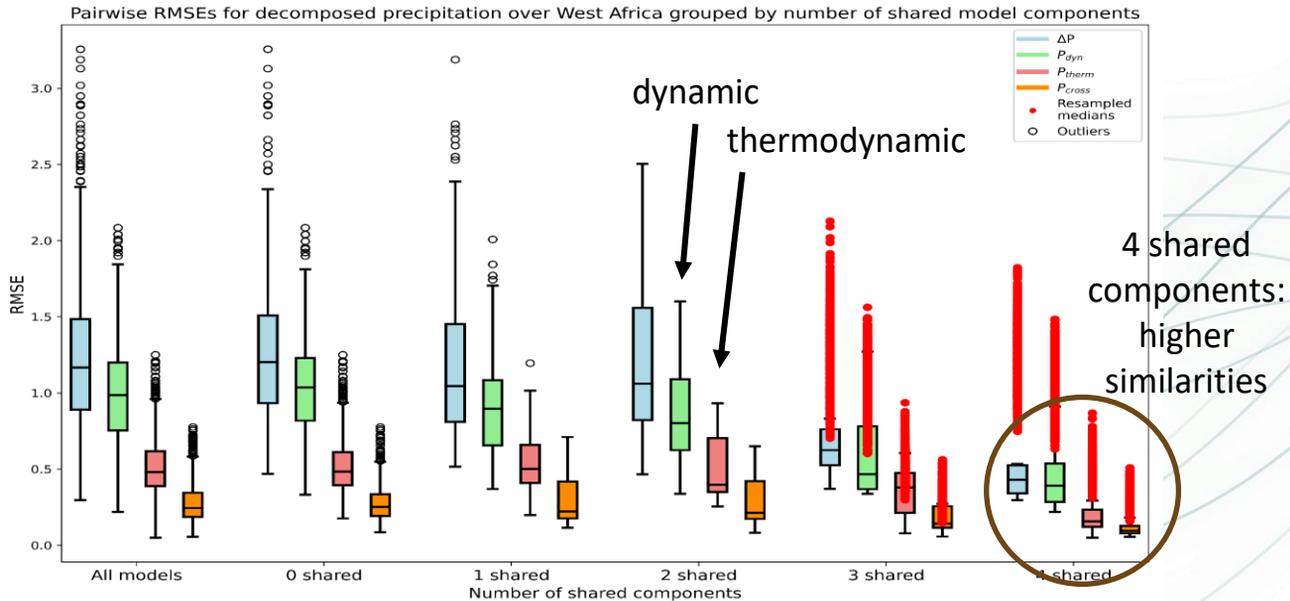
# Results: Mechanisms



Pairwise RMSEs grouped by number of shared components for precipitation change decomposed into dynamic, thermodynamic and cross components over West Africa. The light blue, light green, pink and orange boxplots show the pairwise RMSEs for change in precipitation over West Africa, the dynamic component of precipitation, the thermodynamic component of precipitation and the cross term of precipitation respectively. The red dots on both plots show the result of random resampling outlined in data/methods section.

Differences mainly come from the dynamic change of the atmospheric circulation

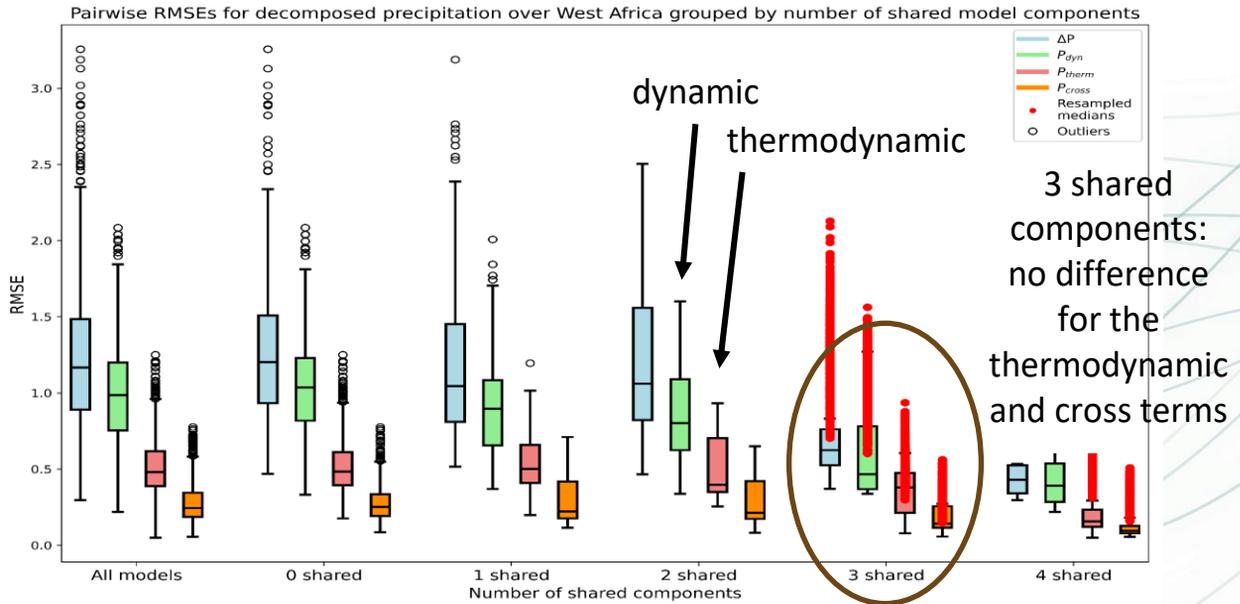
# Results: Mechanisms



Pairwise RMSEs grouped by number of shared components for precipitation change decomposed into dynamic, thermodynamic and cross components over West Africa. The light blue, light green, pink and orange boxplots show the pairwise RMSEs for change in precipitation over West Africa, the dynamic component of precipitation, the thermodynamic component of precipitation and the cross term of precipitation respectively. The red dots on both plots show the result of random resampling outlined in data/methods section.

Differences mainly come from the dynamic change of the atmospheric circulation

# Results: Mechanisms



Pairwise RMSEs grouped by number of shared components for precipitation change decomposed into dynamic, thermodynamic and cross components over West Africa. The light blue, light green, pink and orange boxplots show the pairwise RMSEs for change in precipitation over West Africa, the dynamic component of precipitation, the thermodynamic component of precipitation and the cross term of precipitation respectively. The red dots on both plots show the result of random resampling outlined in data/methods section.

Differences mainly come from the dynamic change of the atmospheric circulation