

# Evaluating the Performance of VR-CESM for Modeling Precipitation in Southeast Asia

**Yi (Harry) Li<sup>1</sup>**, Dylan B. A. Jones<sup>1</sup>, and Colin M. Zarzycki<sup>2</sup>

<sup>1</sup>Department of Physics, University of Toronto, Toronto, Ontario, Canada

<sup>2</sup>Department of Meteorology and Atmospheric Science, Penn State University, PA, USA

# Southeast Asia

- Southeast Asia (SEA) extends from about 12°S-22°N and 94°E-140°E consists of 11 countries including Cambodia, Laos, Myanmar, Thailand, Vietnam, Laos, Brunei, East Timor, Indonesia, Malaysia, Philippines and Singapore
- The region has a total population of more than six hundred million (more than 8% of the global population).
- We focus on mainland southeast Asia, which is the region from 10°N-25°N and 100°E-110°E, also called the Indochinese Peninsula.



# Motivation

- Agriculture and industries in Southeast Asia are particularly vulnerable to changes in the hydrological cycle.
- The global economy is sensitive to weather-related impacts on the region's economy due to the supply chains linking the region to the global economy.
  - ➔ Widespread flooding in Thailand in 2011 resulted in \$46.5 billion in damages to Thailand's economy and contributed to a 2.5% decrease in global industrial production (*Haraguchi and Lall, 2015*).
- Simulating weather extremes in the region requires high-resolution modeling to better capture the complex topography in the region.

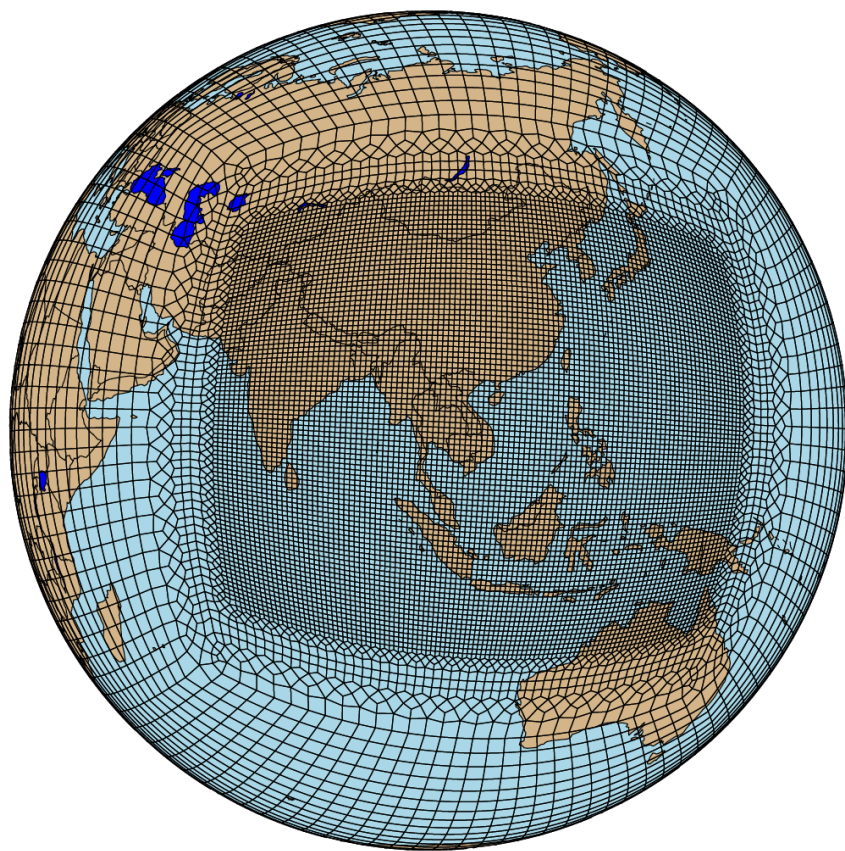


BANGKOK (Oct. 22, 2011) (U.S. Navy photo by Mass Communication Specialist 1st Class Jennifer A. Villalovos/Released)



# Model and Experiments

- We use CESM1.2 with variable resolution (VR-CESM, CAM5.3), which has a fine resolution of  $0.25^\circ$  over Southeast Asia and a coarser resolution of  $1^\circ$  globally.
- Conducted 27 year simulations with an AMIP configuration from 1979 to 2005.
- Compared the VR-CESM simulation with the three coarse resolution CESM simulations listed in the table below.



VRseasia refined grid

CESM version	Description
VRseasia	Spectral element $0.25^\circ \times 0.25^\circ$ in SEA $\sim 1^\circ \times 1^\circ$ globally
ne30	Spectral element $\sim 1^\circ \times 1^\circ$ globally
fv0.9x1.25	Finite volume $\sim 1^\circ \times 1^\circ$ globally
fv1.9x2.5	Finite volume $\sim 2^\circ \times 2^\circ$ globally



# Observations for Model Evaluation

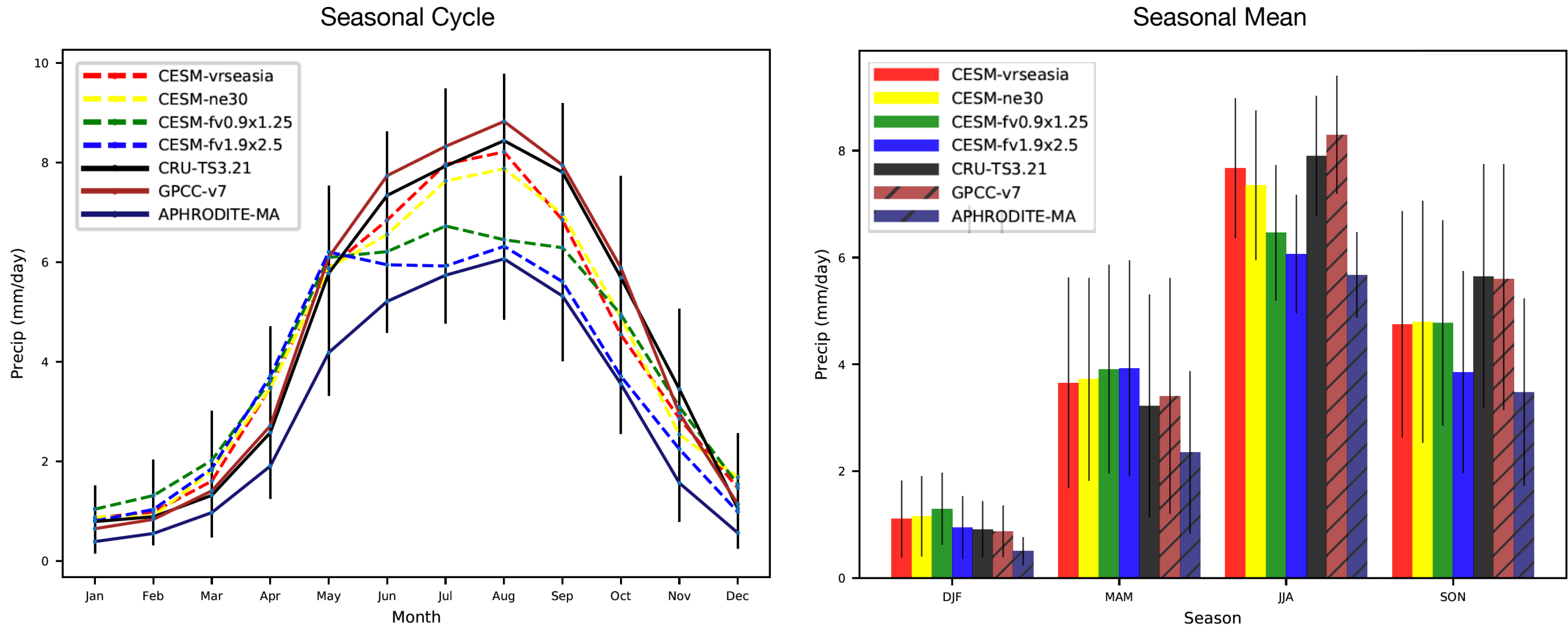
Observations	Period	Frequency	Coverage	References
CRU-TS3.21	1901-2012	monthly	Gauge, gridded	Harris, et al., 2014
GPCC-v7	1901-2010	monthly	Gauge, gridded	Becker, et al., 2011
APHRODITE- MA	1951-2007	daily	Gauge, gridded	Yatagai, et al., 2012
SA-OBS	Depend on stations	daily	stations	Van den Besselaar, et al., 2017

## Other Model Simulations for Comparison: CORDEX-SEA

- The Southeast Asia Regional Climate Downscaling (SEACLID) / Coordinated Regional climate Downscaling Experiment (CORDEX) Southeast Asia Project published high-resolution climate change scenarios (25 km x 25 km) for the Southeast Asia region.

Experiments	GCMs	RCMs	Resolution
1970–2005 historical	ICHEC–EC–EARTH	ICTP–RegCM4.3	25km x 25km
1970–2005 historical	IPSL–IPSL–CM5A–LR	ICTP–RegCM4.3	25km x 25km
1970–2005 historical	MPI–M–MPI–ESM–MR	ICTP–RegCM4.3	25km x 25km
1951–2005 historical	MOHC–HadGEM2–ES	SMHI–RCA4	25km x 25km

# Seasonal Variation in Precipitation

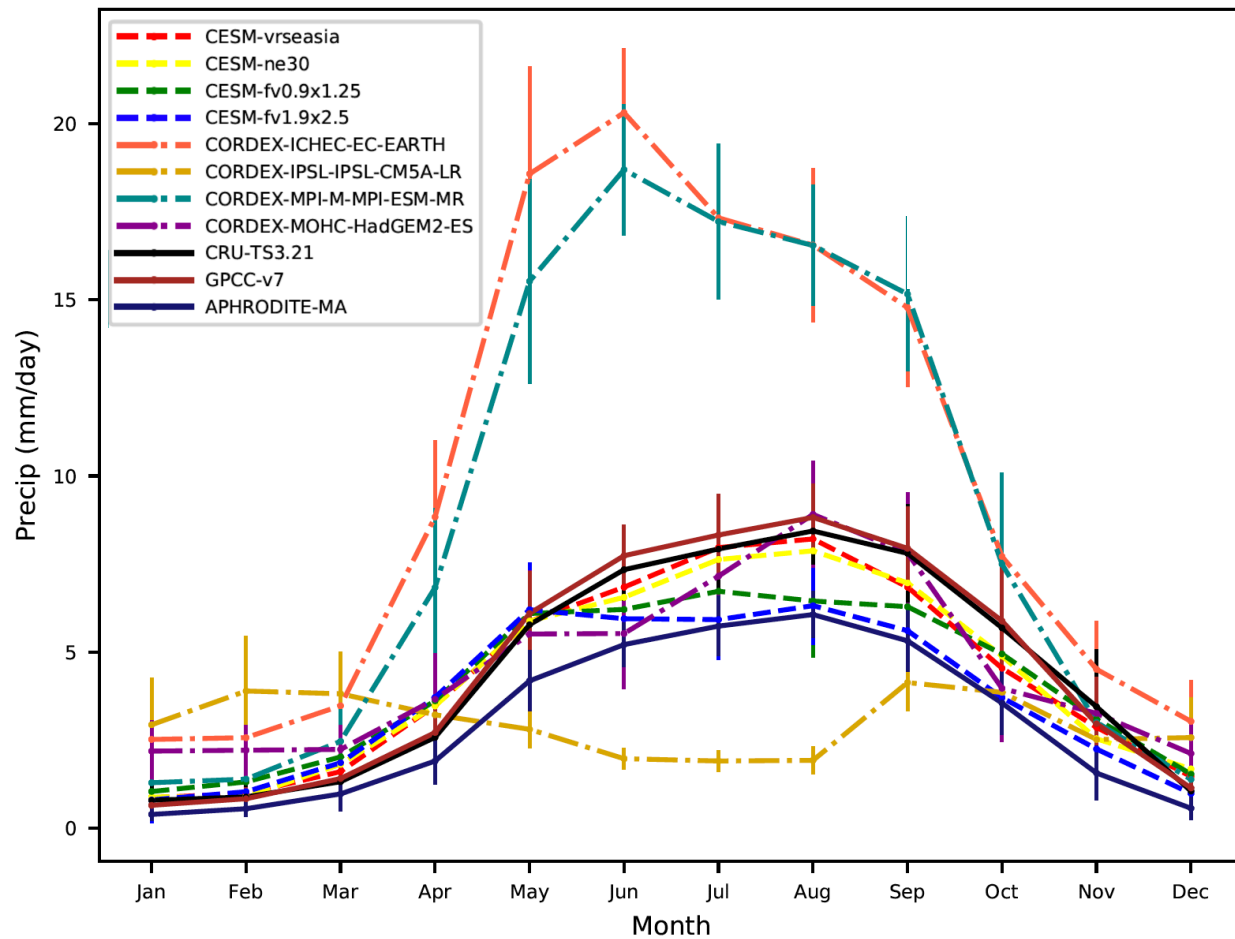


- The spectral element models (VRseasia and ne30) best simulate the seasonal cycle in the CRU and GPCC data.
- Overall, VR-CESM produced the best agreement with CRU and GPCC data, with the smallest bias (especially in summer).
- The APHRODITE-MA data suggest much lower precipitation rates than the CRU and GPCC data.

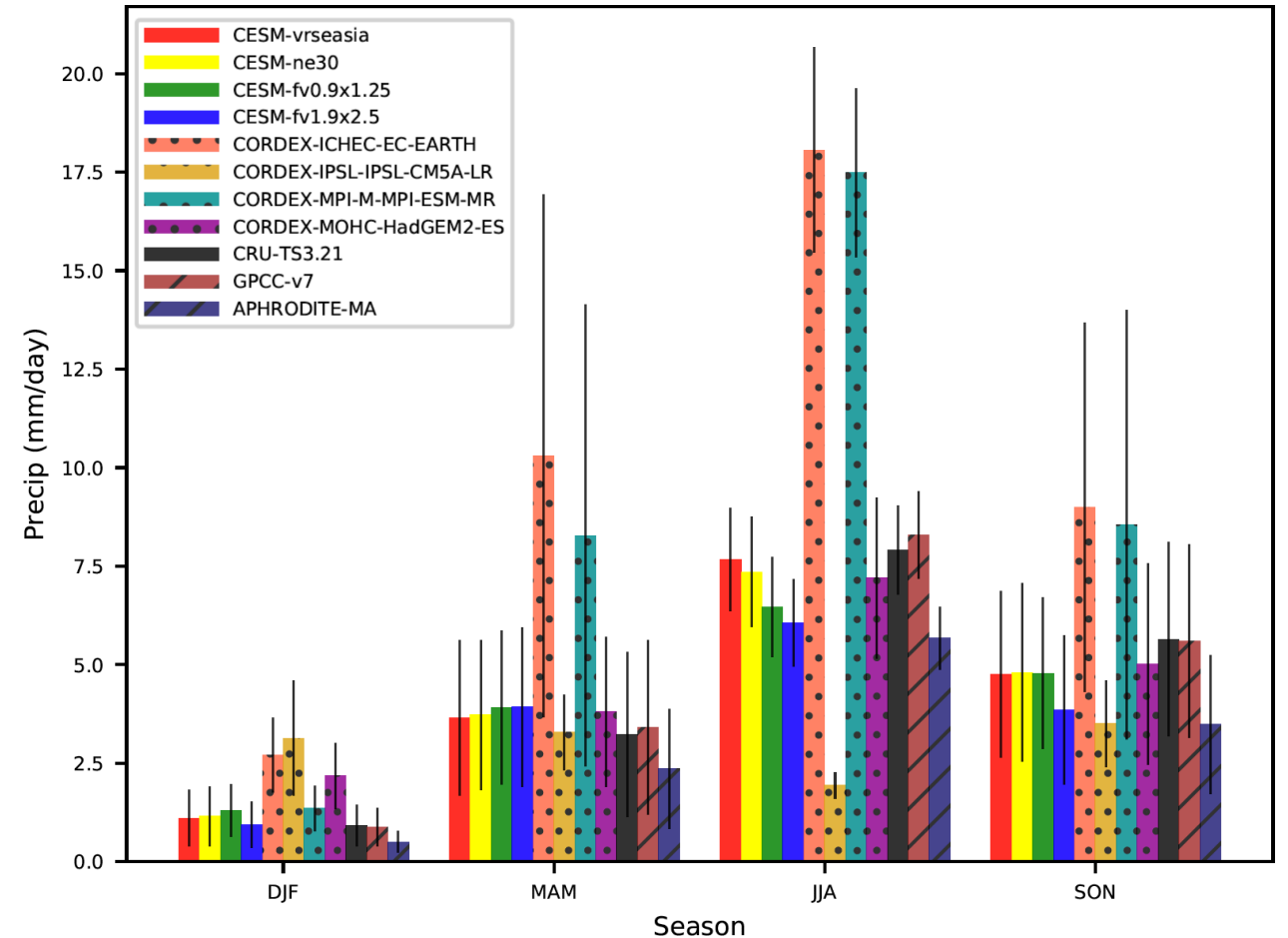


# Seasonal Variation in Precipitation

Seasonal Cycle

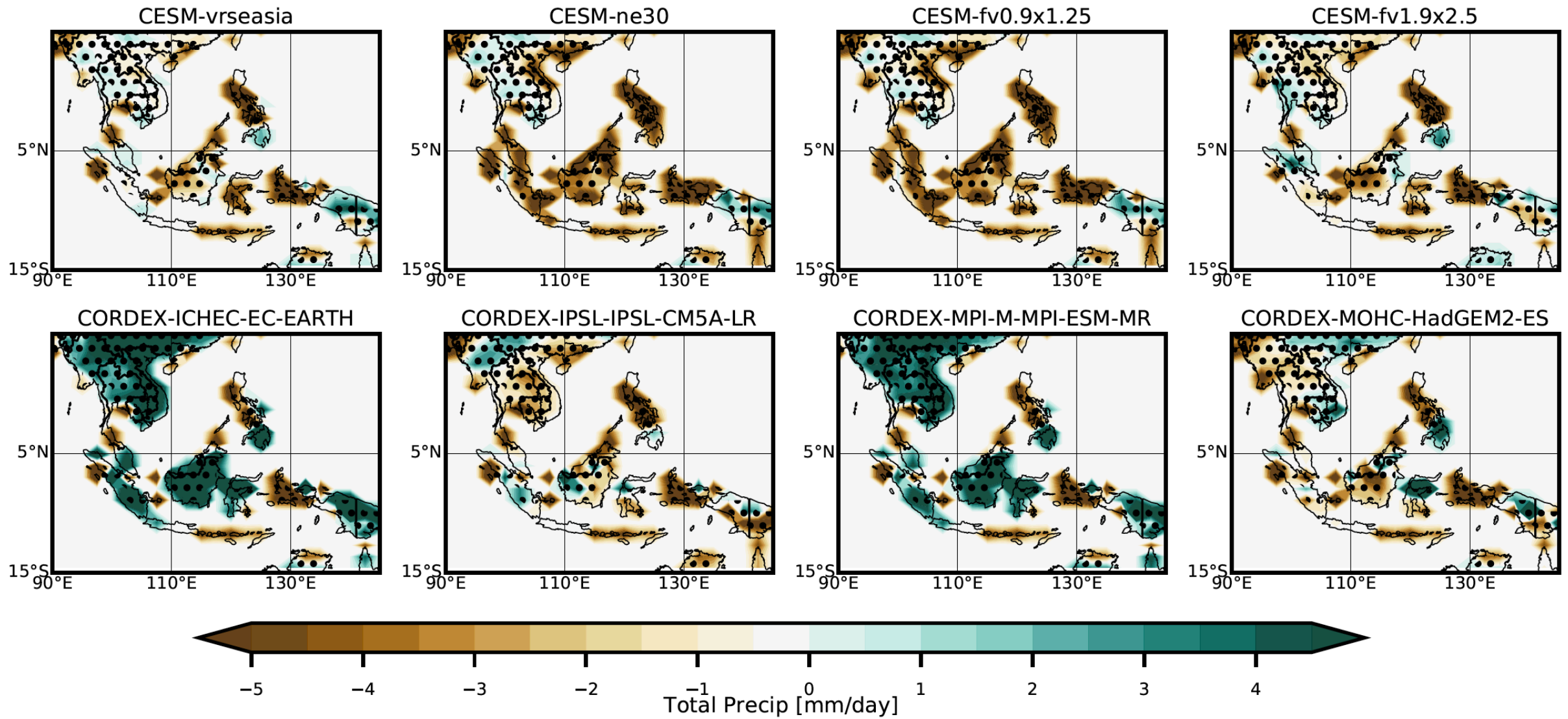


Seasonal Mean



- Two CORDEX-SEA models significantly overestimate the summer precipitation, whereas a third one fails to capture the seasonal cycle in precipitation in SEA.
- The HadGEM2 with RCA4 CORDEX simulation was most consistent with the CESM simulations.

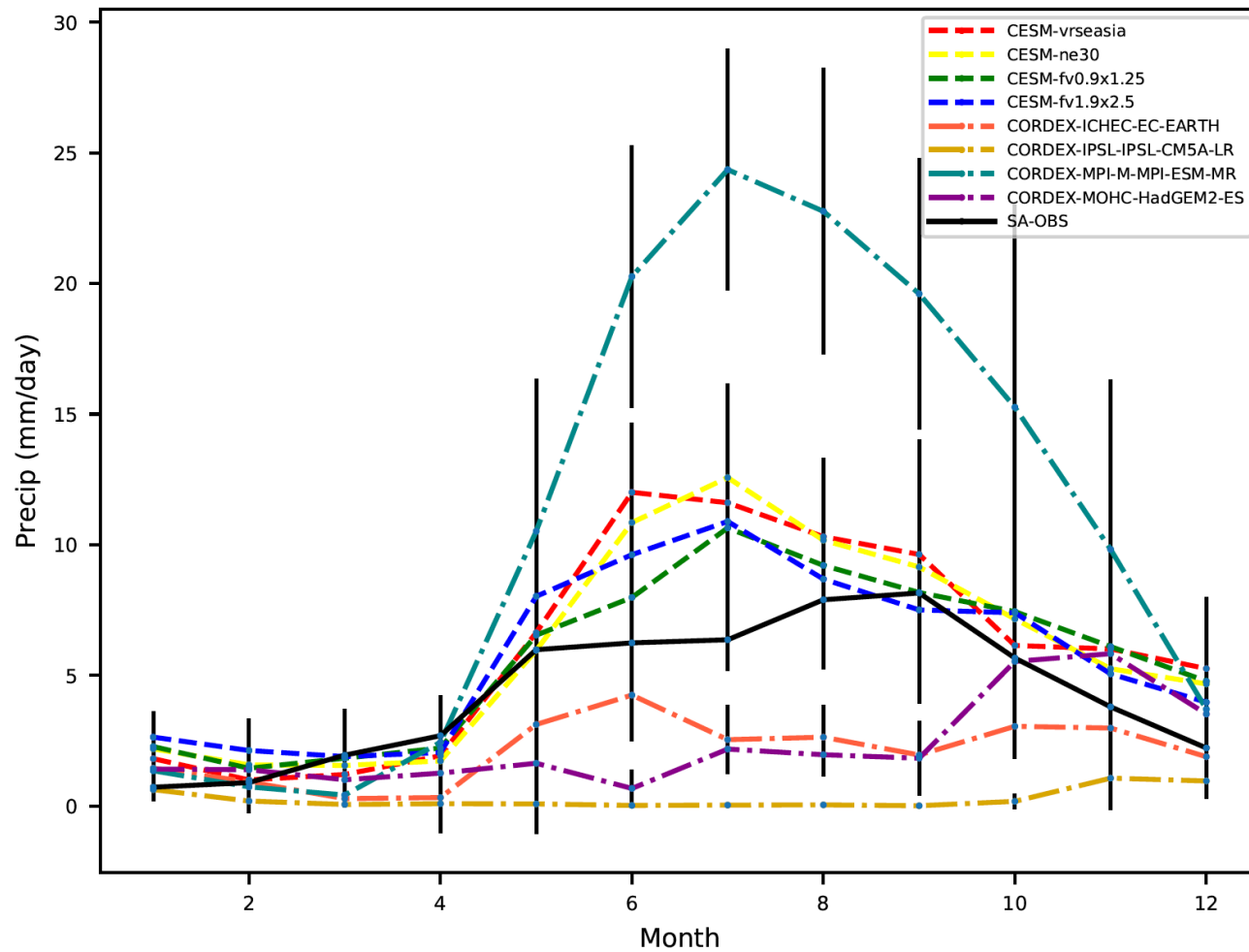
# Annual Mean Precipitation Bias (GPCC)



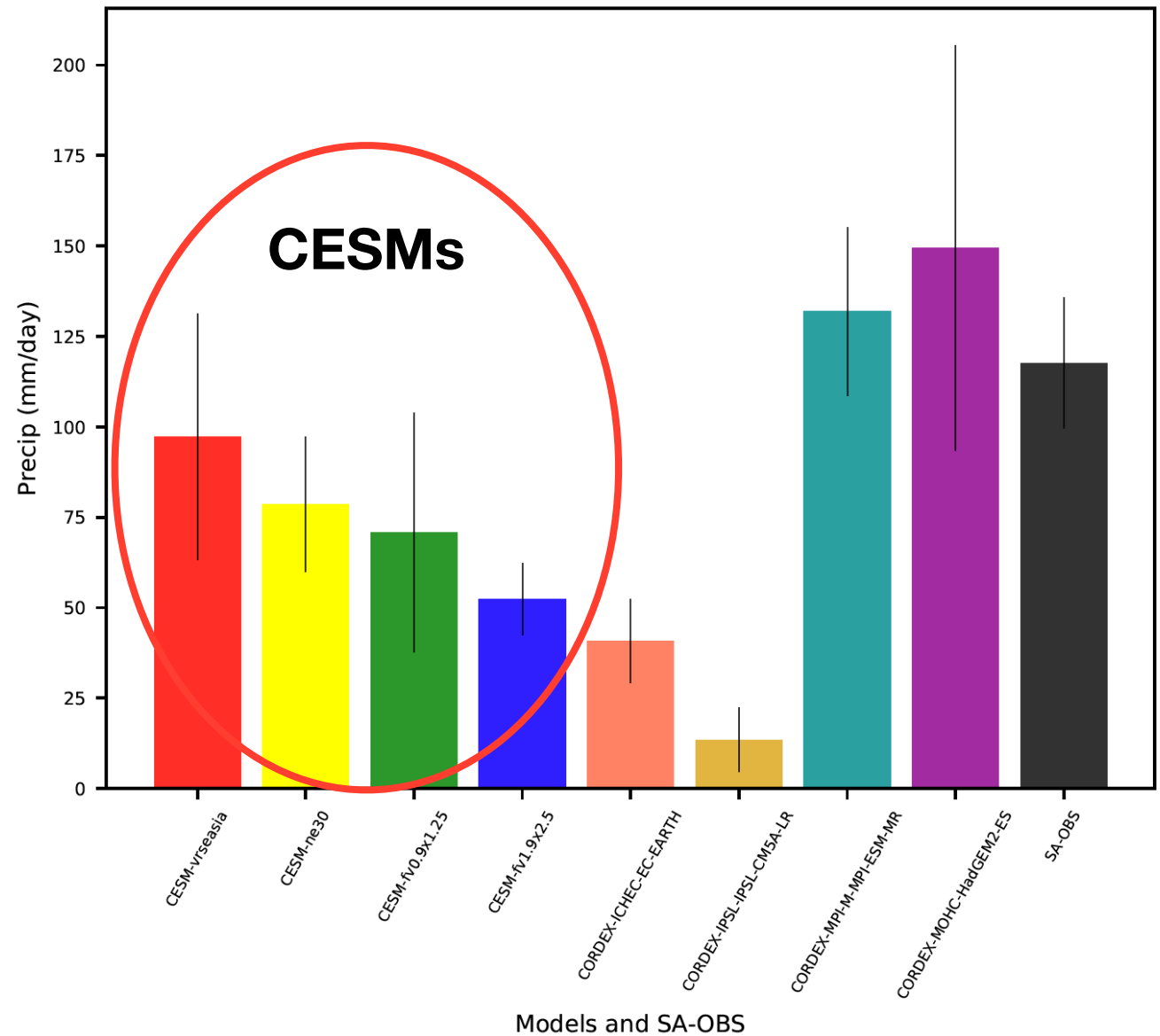
- Although the CESM simulations are drier than the observations, the models have much smaller biases than the CORDEX models.
- VR-CESM has the smallest biases across all the CESM models, especially in northern Vietnam and across maritime Southeast Asia.

# Comparison with Station Observations (SA-OBS)

Seasonal Cycle



Annual Maximum Precipitation

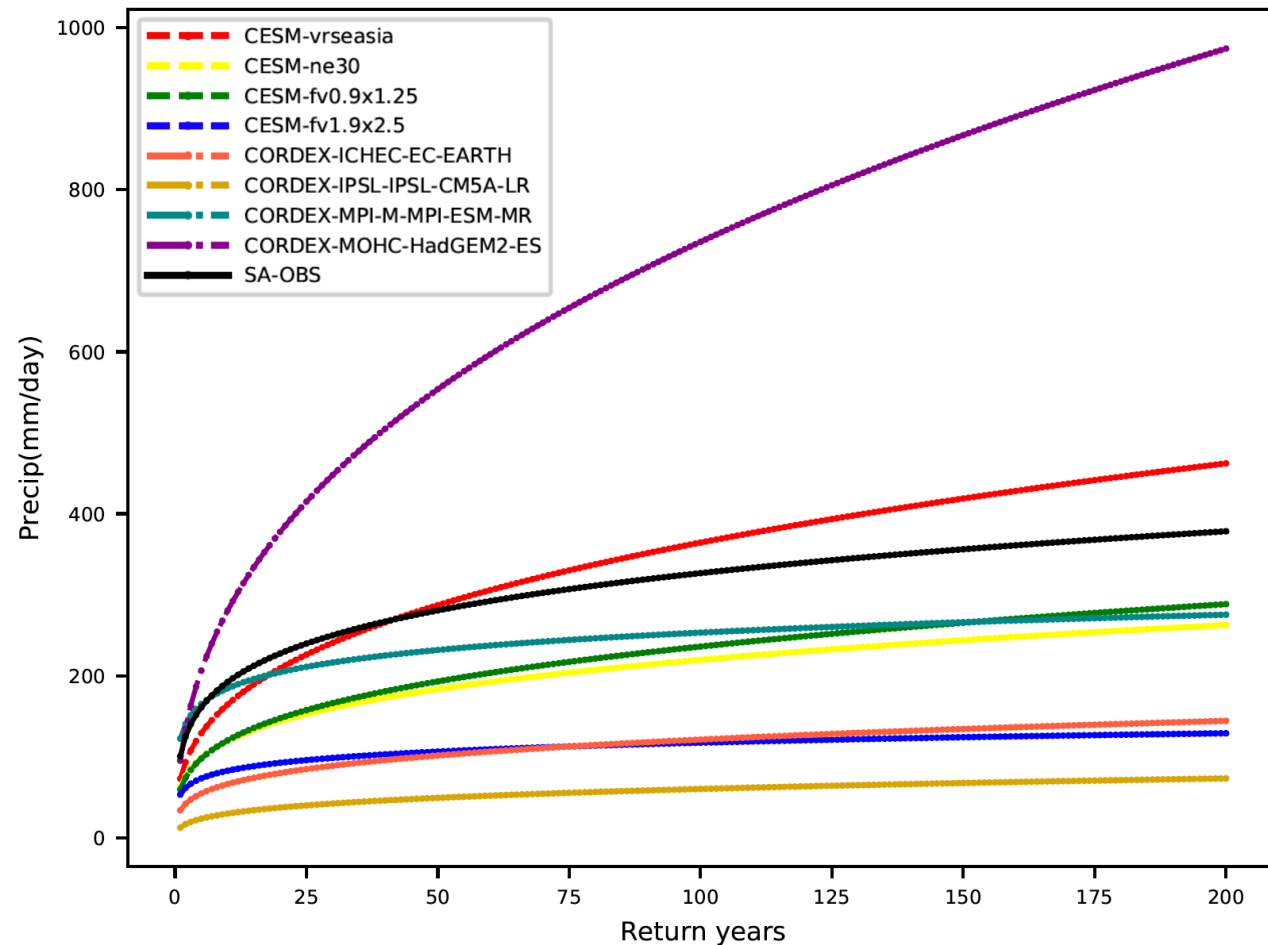


- All CESM simulations are bias high in summer, resulting in greater annual mean precipitation (not shown).
- Of all the CESM models, VR-CESM simulated the annual maximum precipitation the best.
- Two of the CORDEX models simulate the maximum precipitation better, but produce a much more biased seasonal cycle.

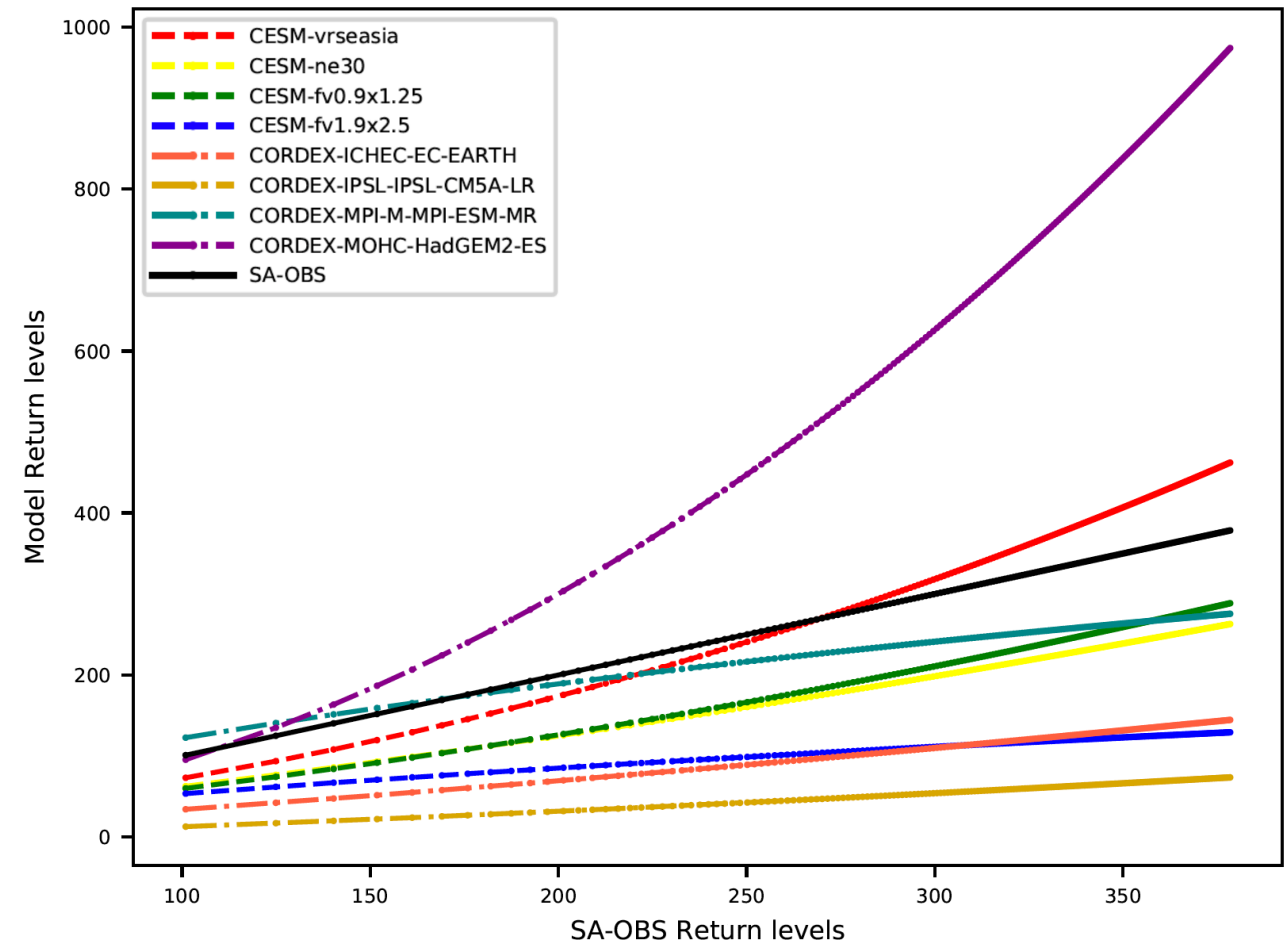


# Extreme Analysis Across all Stations

N-year Returns for 97th Precipitation



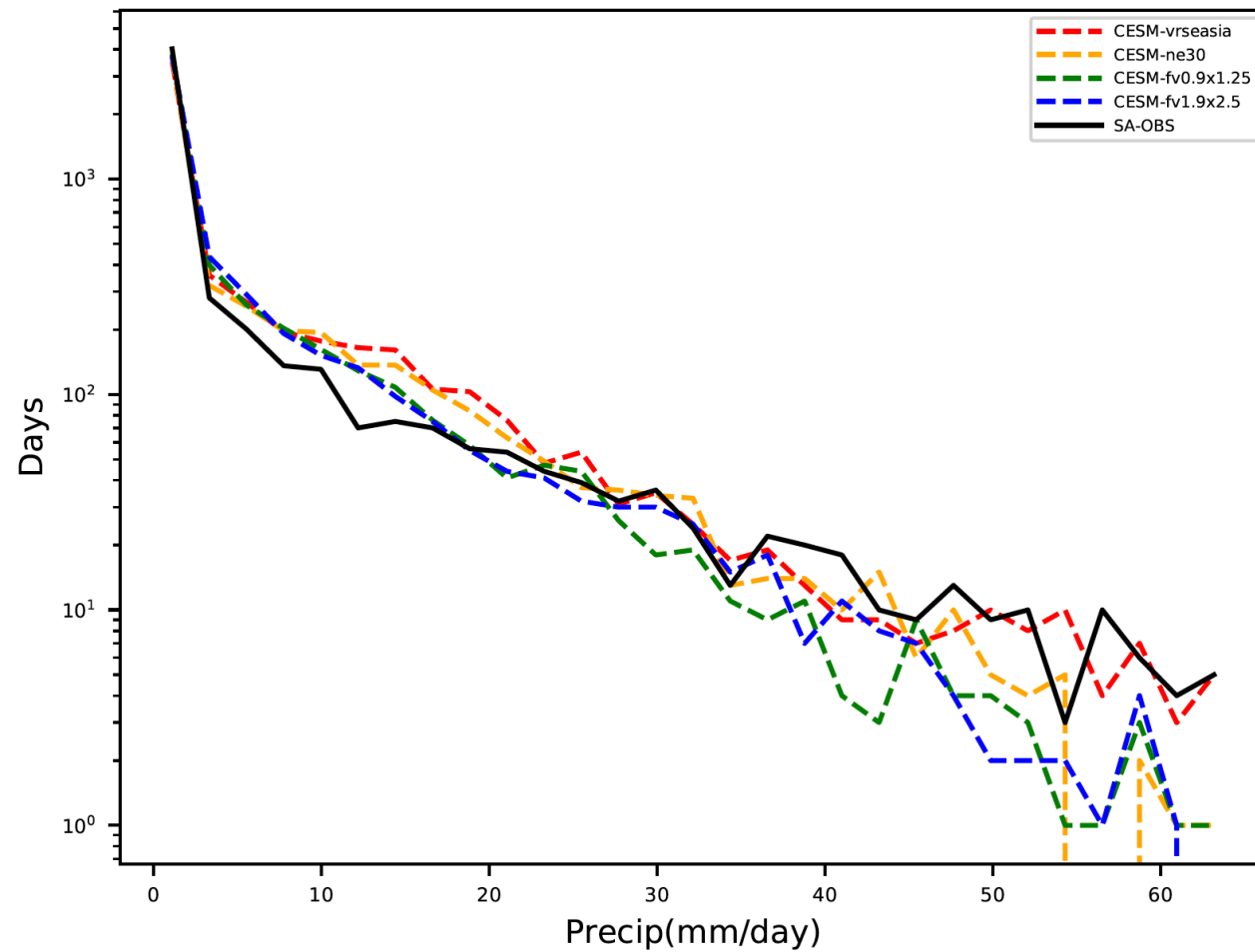
Return levels for 97th Precipitation



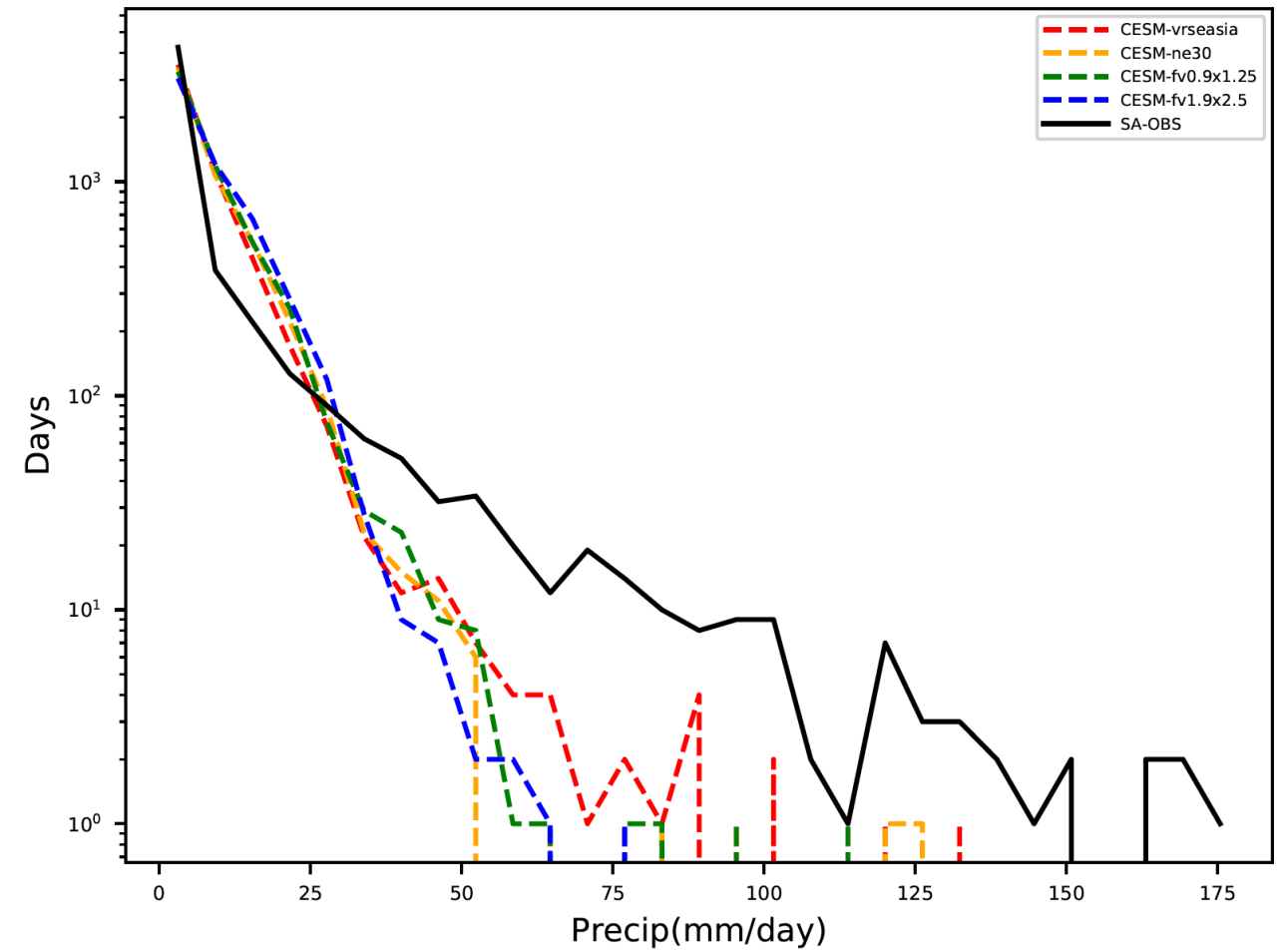
- As expected, increasing the resolution in CESM has a great impact on simulating the distribution of precipitation extremes;
  - ➔ VR-CESM best captures the observed N-year precipitation returns.
- Although HadGem2 with RCA4 was consistent with the CESM simulations in reproducing the seasonal cycle in the CRU and GPCPC data, the N-year returns in HadGem2 is much higher than in the all station observations.

# Heterogeneity in the Station Data: Precipitation Distribution

Station 2: CHIANG RAI in Thailand



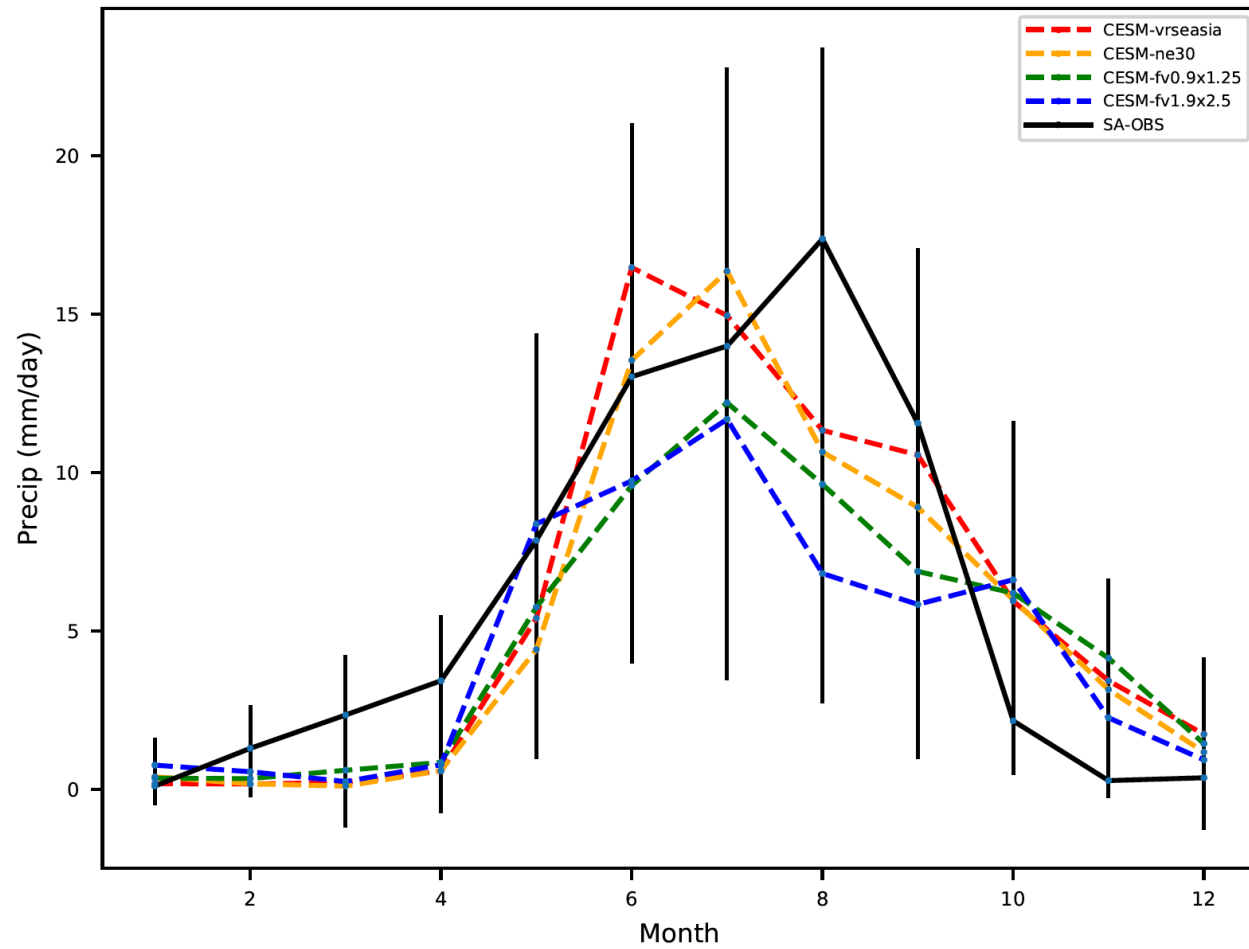
Station 19: NAKHON SI THAMMARAT in Thailand



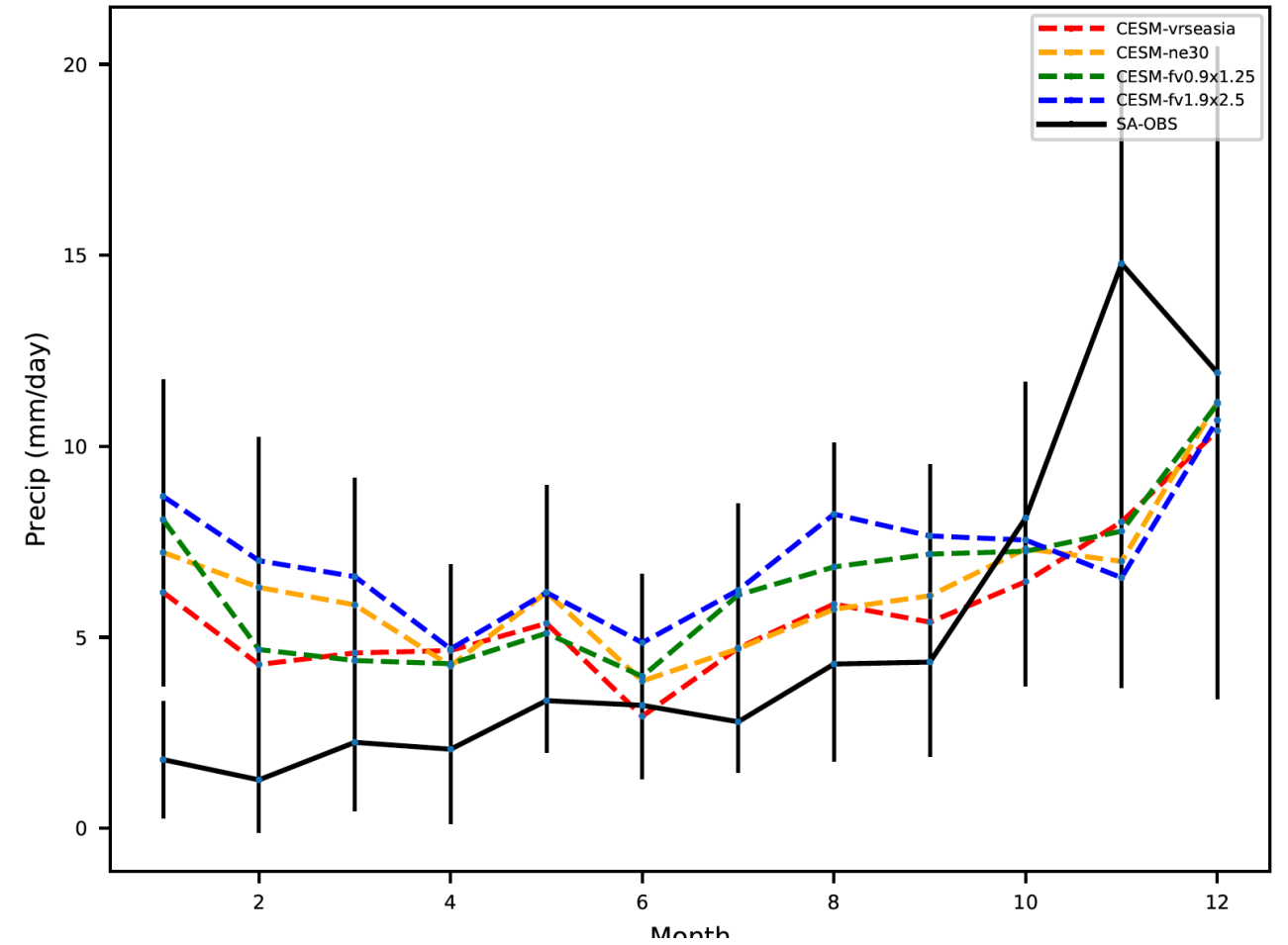
- There are significant differences in the precipitation distribution across the stations.

# Heterogeneity in the Station Data: Seasonal Cycle

Station 7: NAKHON PHANOM AGROMET in Thailand



Station 19: NAKHON SI THAMMARAT in Thailand

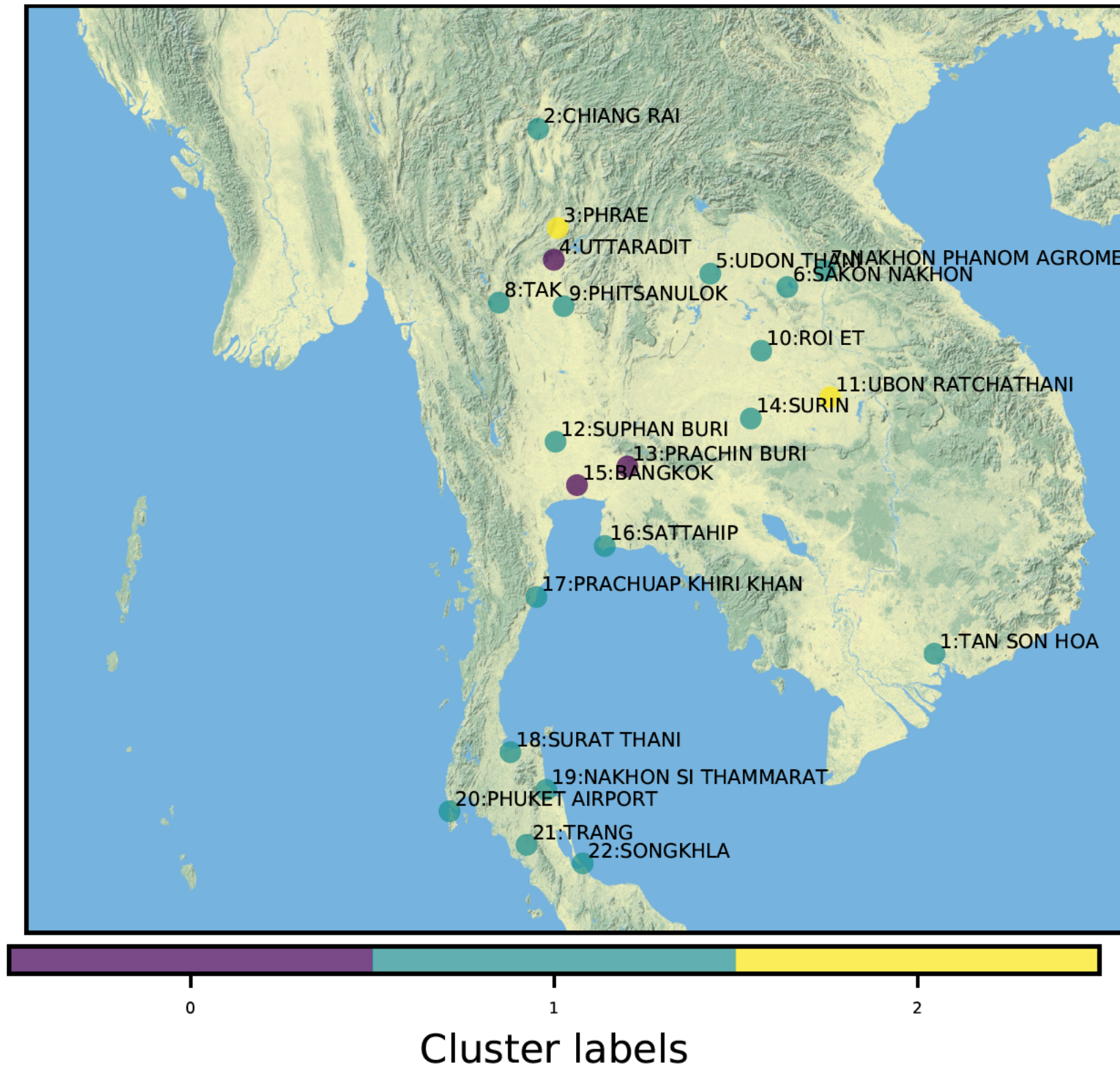


- The seasonal cycles can be also very different across the stations!



# Clustering the Station Data

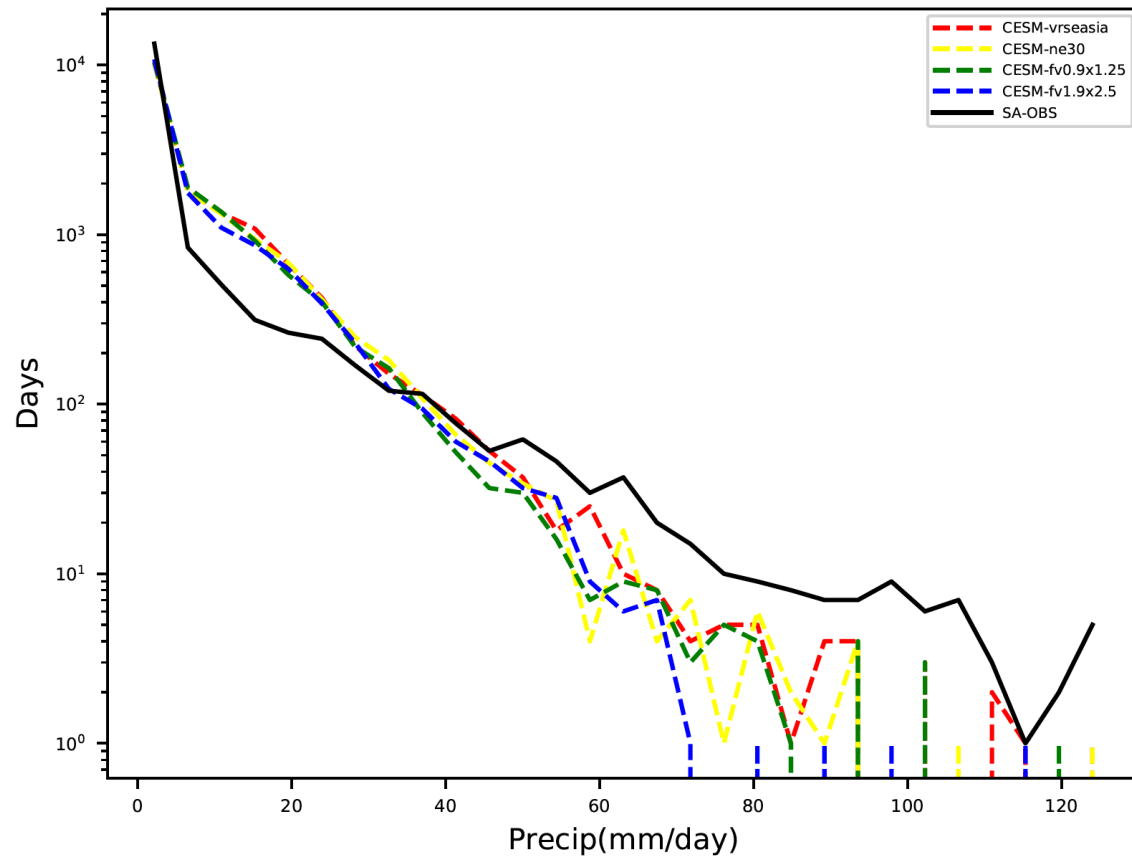
Clustering results



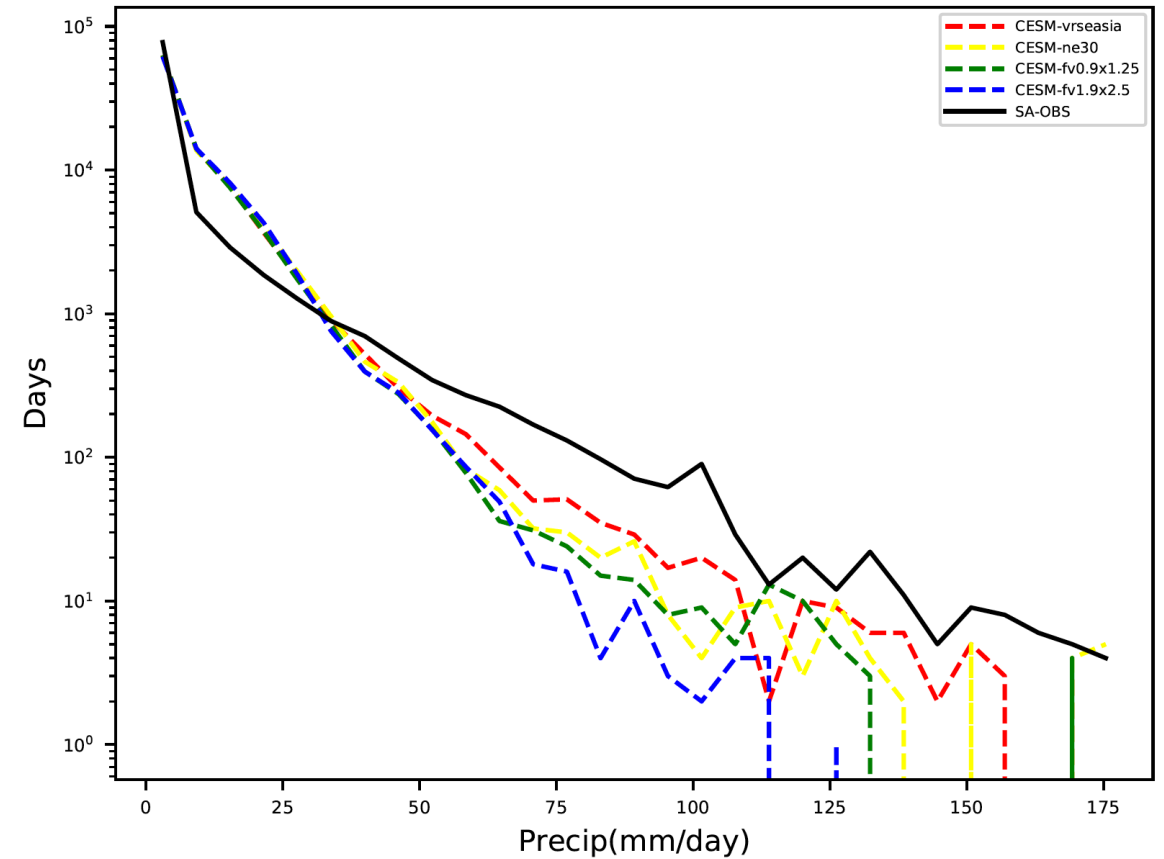
- All 22 stations are divided into 3 groups based on the differences in daily precipitation time series using K-Means clustering.
- The members in each cluster can be geographically distant.

# Histogram of Total Precipitation in the 3 Groups

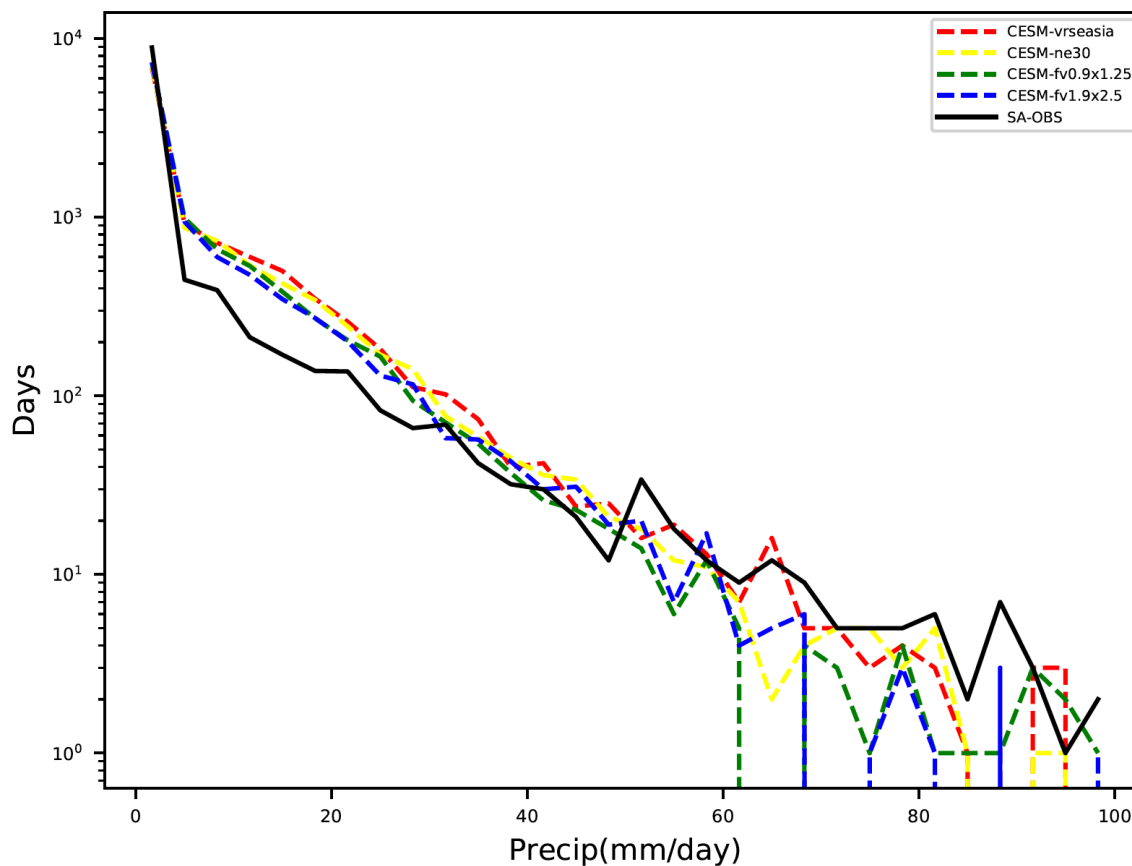
Group: 0



Group: 1



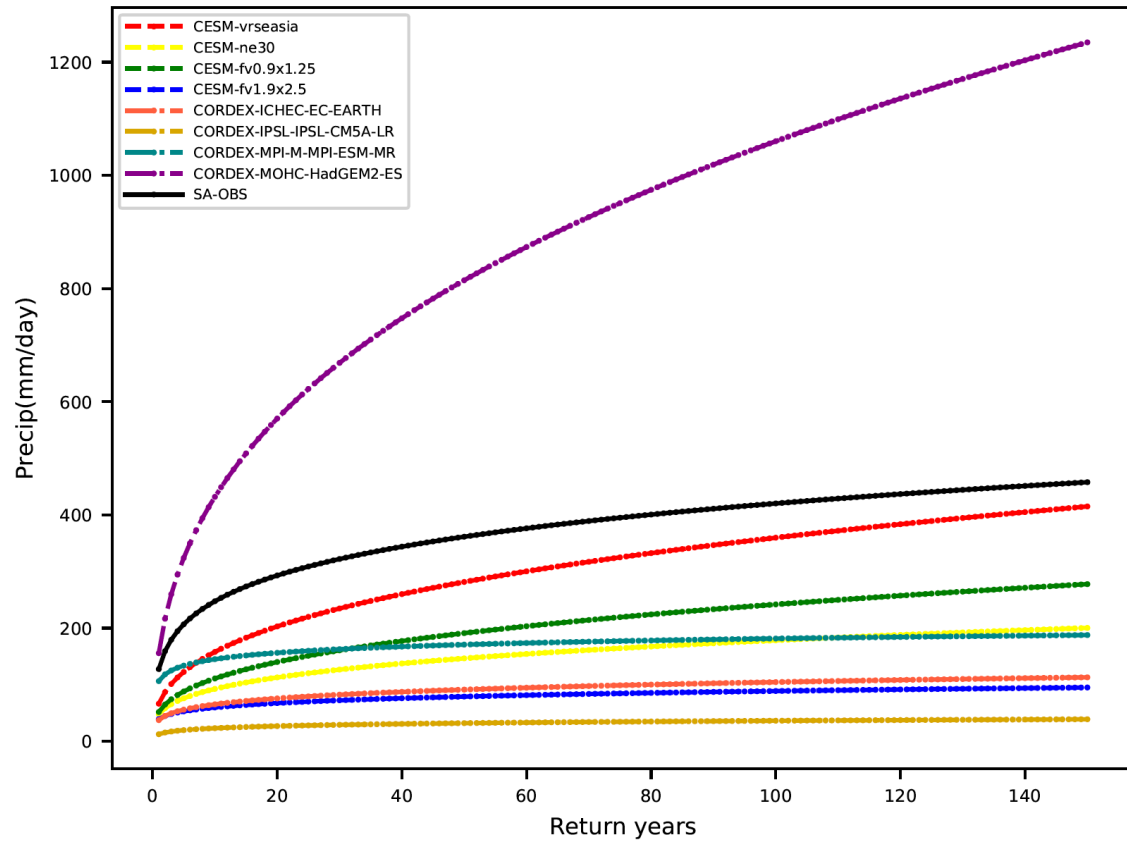
Group: 2



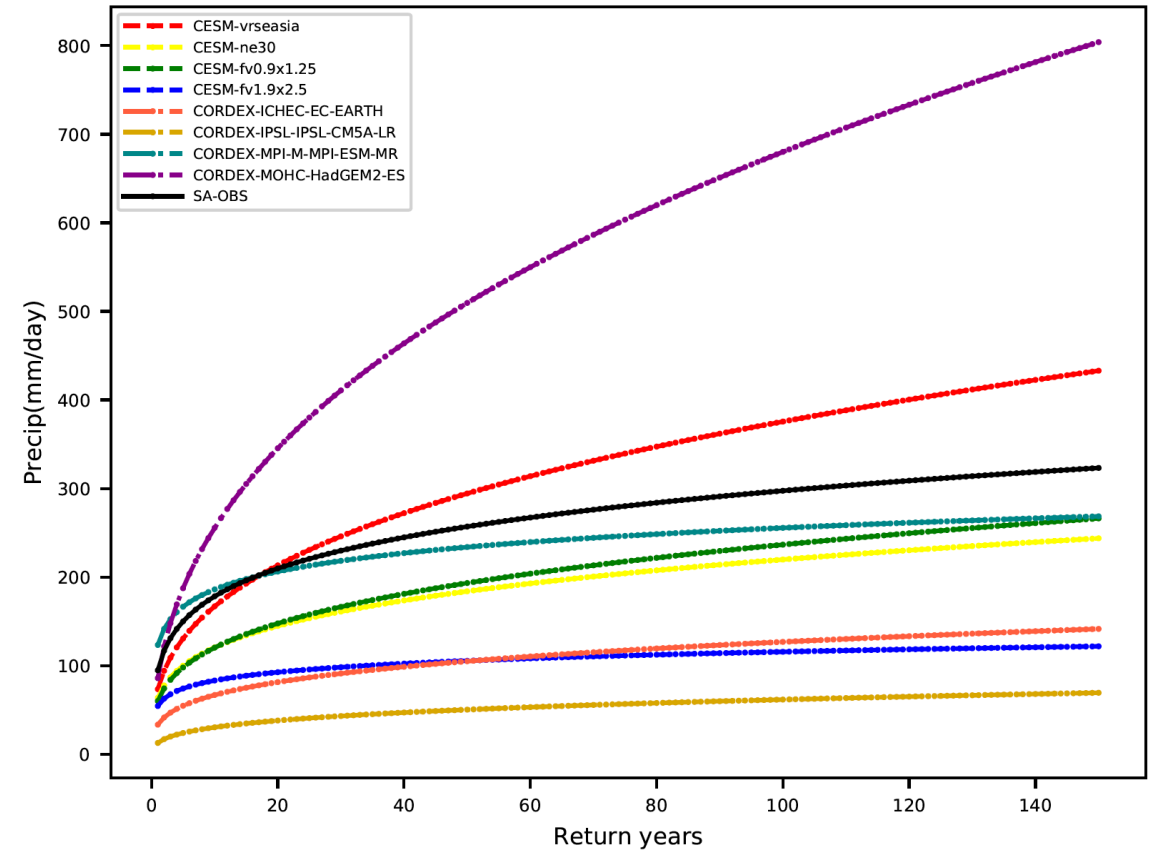
- In Group 1, which contains the most stations, VR-CESM matches the observations better.
- In Group 2 (containing Station 3 and 11), all CESM simulations fit the observations well except for the 2° runs.

# 97th Precipitation Return levels in the 3 Groups

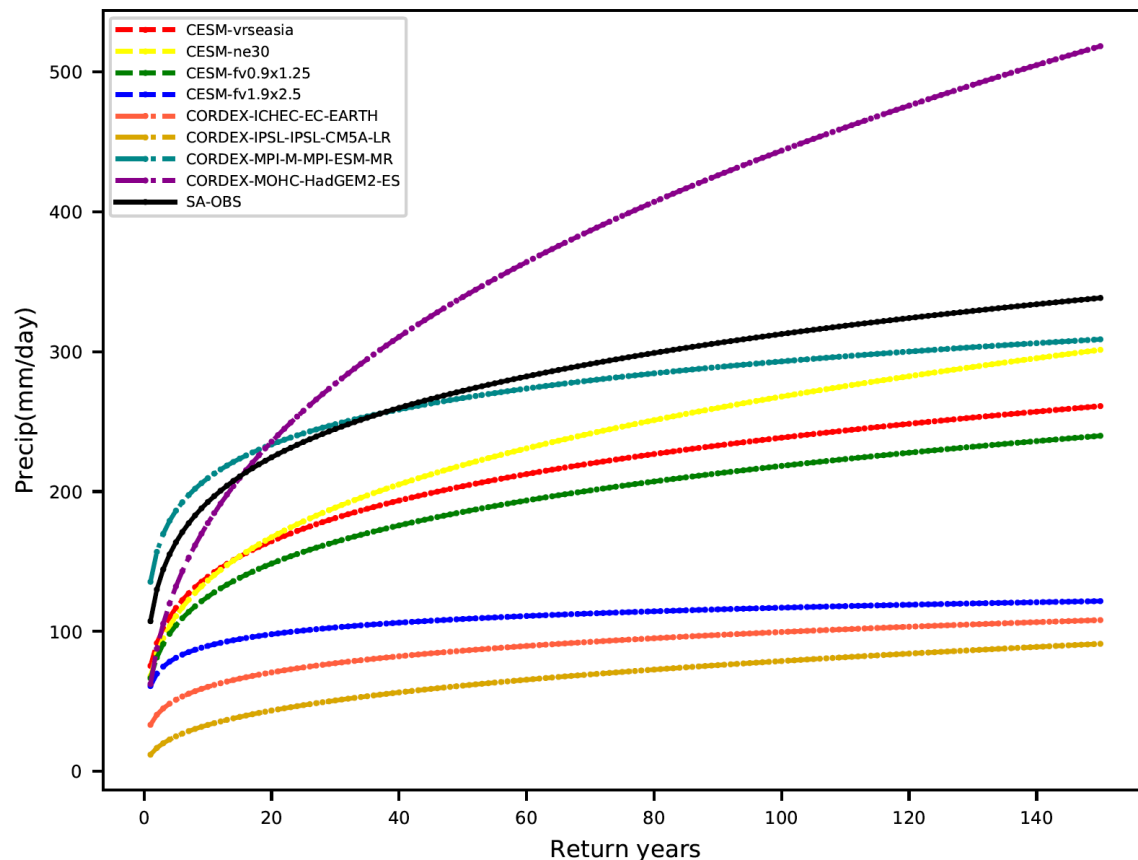
Group 0: N-year returns for 97th precipitation



Group 1: N-year returns for 97th precipitation



Group 2: N-year returns for 97th precipitation



- VR-CESM generally performs better than the other CESM simulations in simulating very high precipitation extremes.
- The MPI with RegCM4.3 CORDEX model also agrees with the extremes in the observations, but it has a significant high biases in mean precipitation.



# Summary

- The CESM simulations generally perform better than the CORDEX simulations in capturing precipitation over mainland Southeast Asia.
- VR-CESM best simulates the seasonal cycle in precipitation over mainland Southeast Asia, as seen in the CRU and GPCC data.
- VR-CESM improves the simulation of the precipitation extremes in the region (measured by the station data), though the year-to-year variability is not necessarily improved.

# Acknowledgement

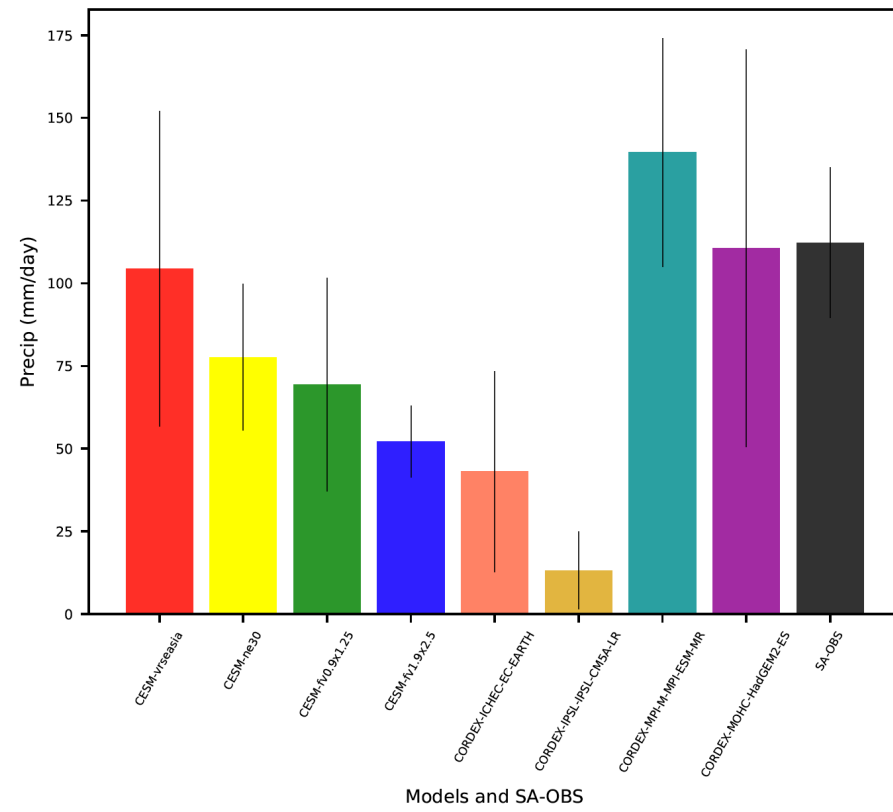
- This work was funded by the Natural Sciences and Engineering Research Council (NSERC) of Canada.
- We thank the SEACLID/CORDEX Southeast Asia Project team for making available the regional climate model data.

## References

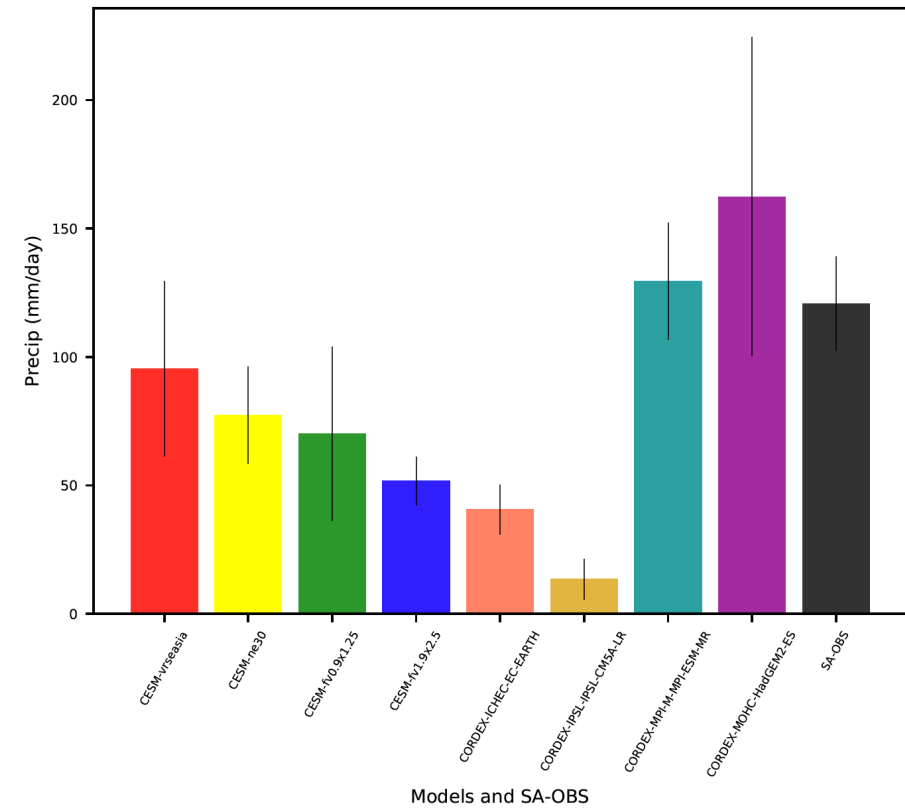
- Gent, Peter R., et al. "The community climate system model version 4." *Journal of Climate* 24.19 (2011): 4973-4991.
- Haraguchi, M, and U. Lall (2015), Flood risks and impacts: A case study of Thailand's floods in 2011 and research questions for supply chain decision making, *International Journal of Disaster Risk Reduction*, 14, 256-272.
- Harris, I. P. D. J., et al. "Updated high-resolution grids of monthly climatic observations—the CRU TS3. 10 Dataset." *International journal of climatology* 34.3 (2014): 623-642.
- Becker, Andreas, et al. "GPCC full data reanalysis Version 6.0 at 1.0: monthly land-surface precipitation from rain-gauges built on GTS-based and historic Data." *Global Precipitation Climatology Centre (GPCC): Berlin, Germany* (2011).
- Yatagai, Akiyo, et al. "APHRODITE: Constructing a long-term daily gridded precipitation dataset for Asia based on a dense network of rain gauges." *Bulletin of the American Meteorological Society* 93.9 (2012): 1401-1415.
- Van den Besselaar, Else JM, et al. "SA-OBS: a daily gridded surface temperature and precipitation dataset for Southeast Asia." *Journal of Climate* 30.14 (2017): 5151-5165.

# Supplementary slides

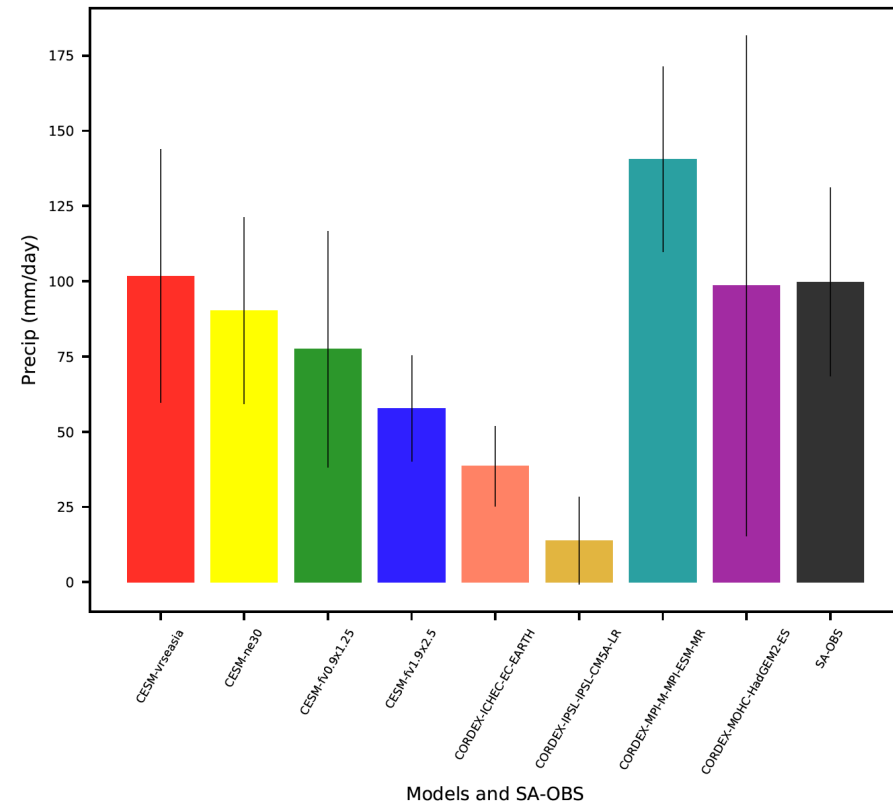
Group 0: Annual maximum precipitation



Group 1: Annual maximum precipitation



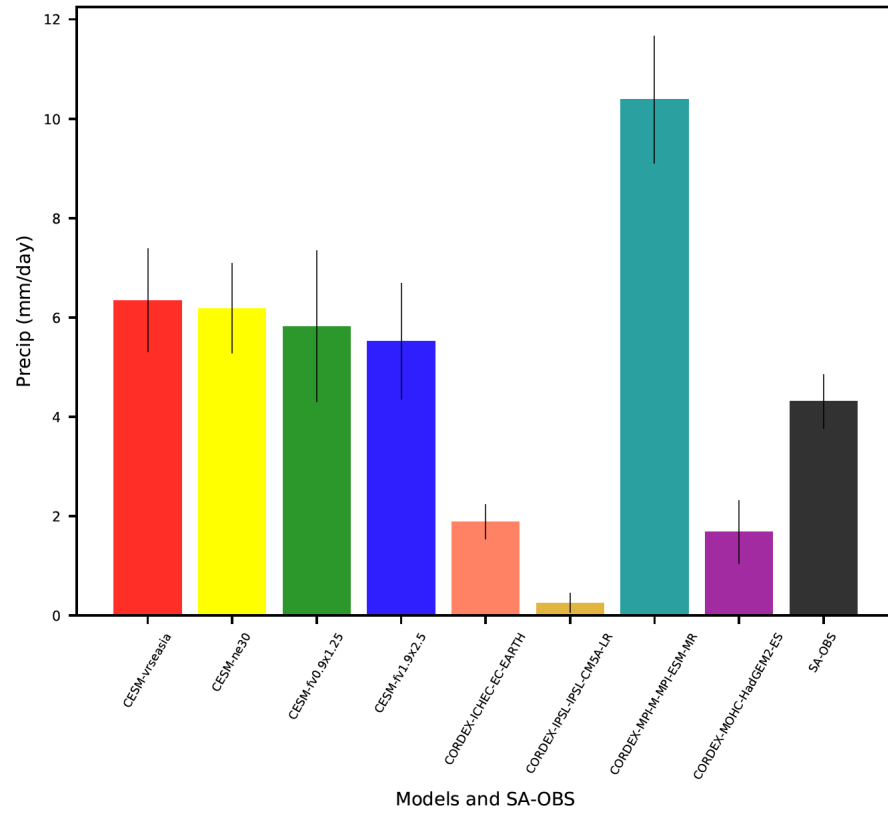
Group 2: Annual maximum precipitation



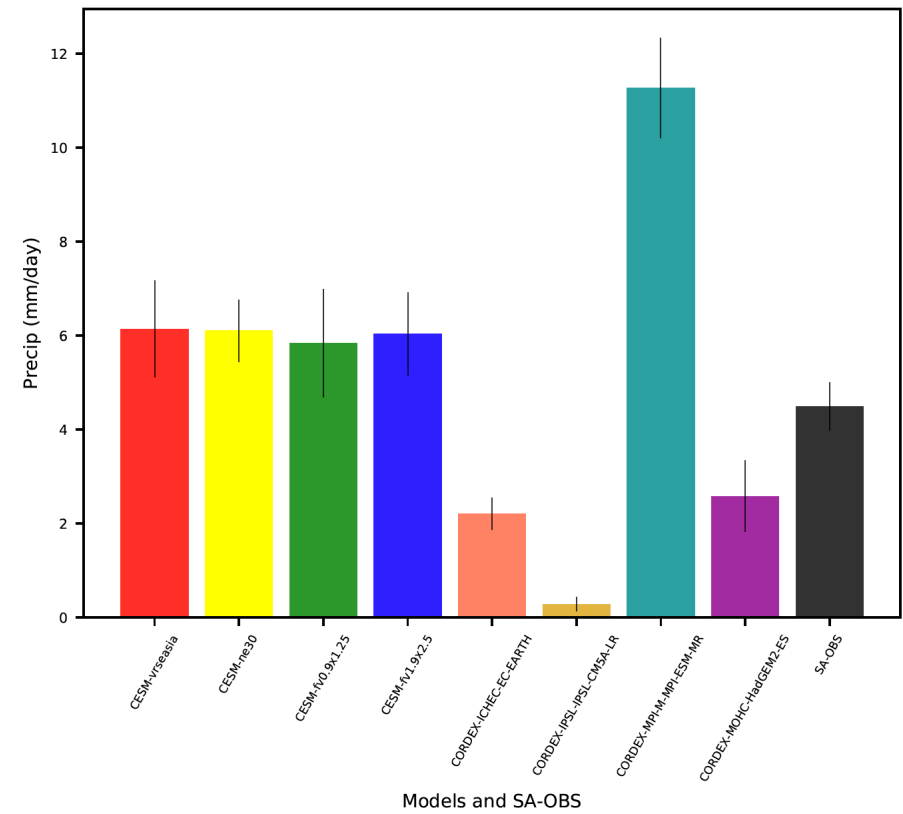


# Supplementary slides

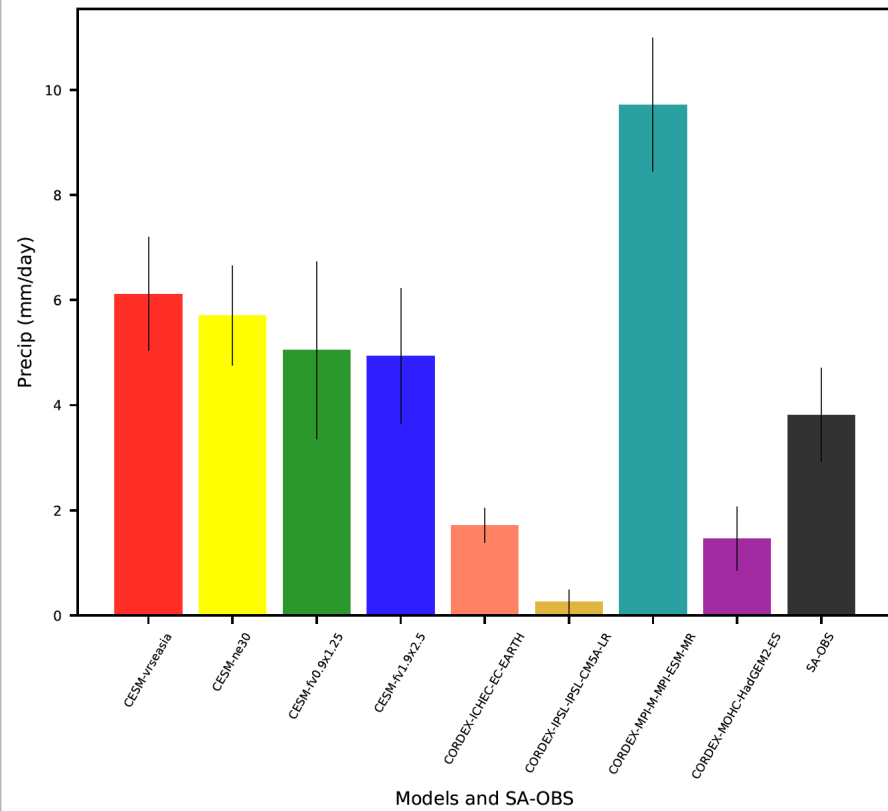
Group 0: Annual mean precipitation



Group 1: Annual mean precipitation

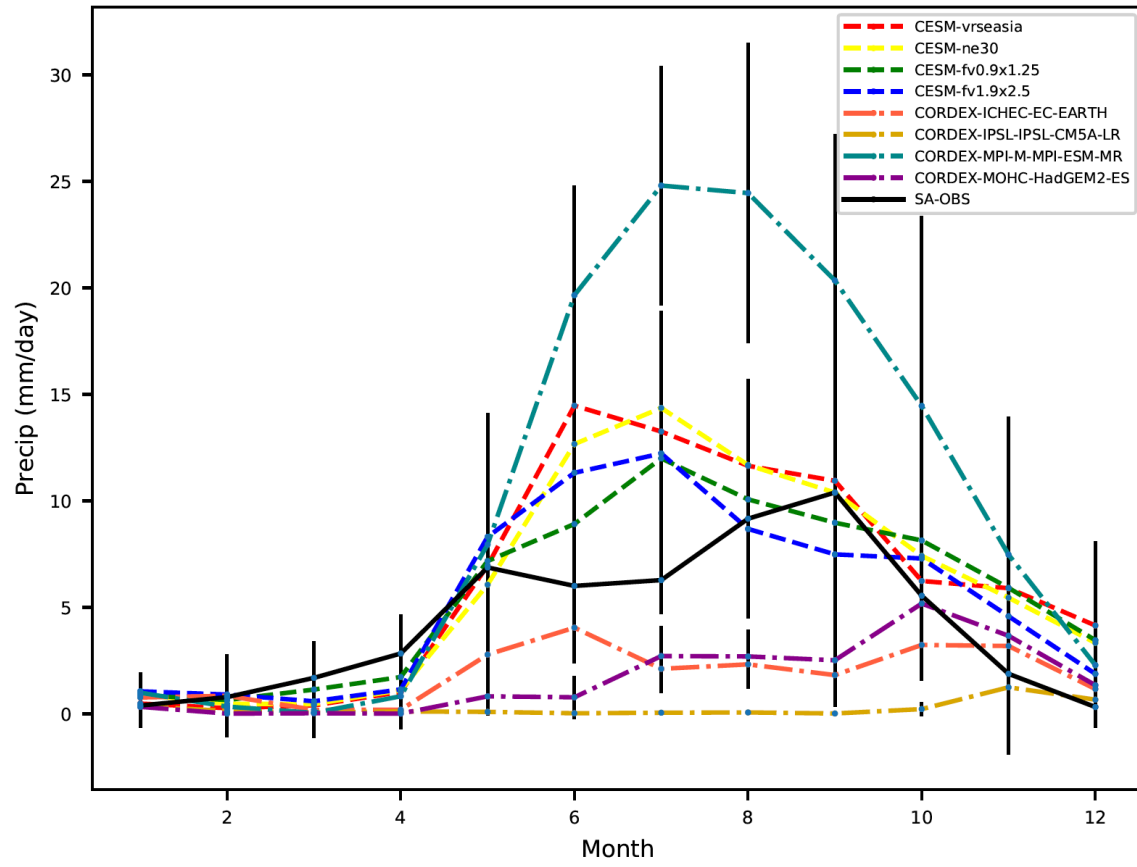


Group 2: Annual mean precipitation

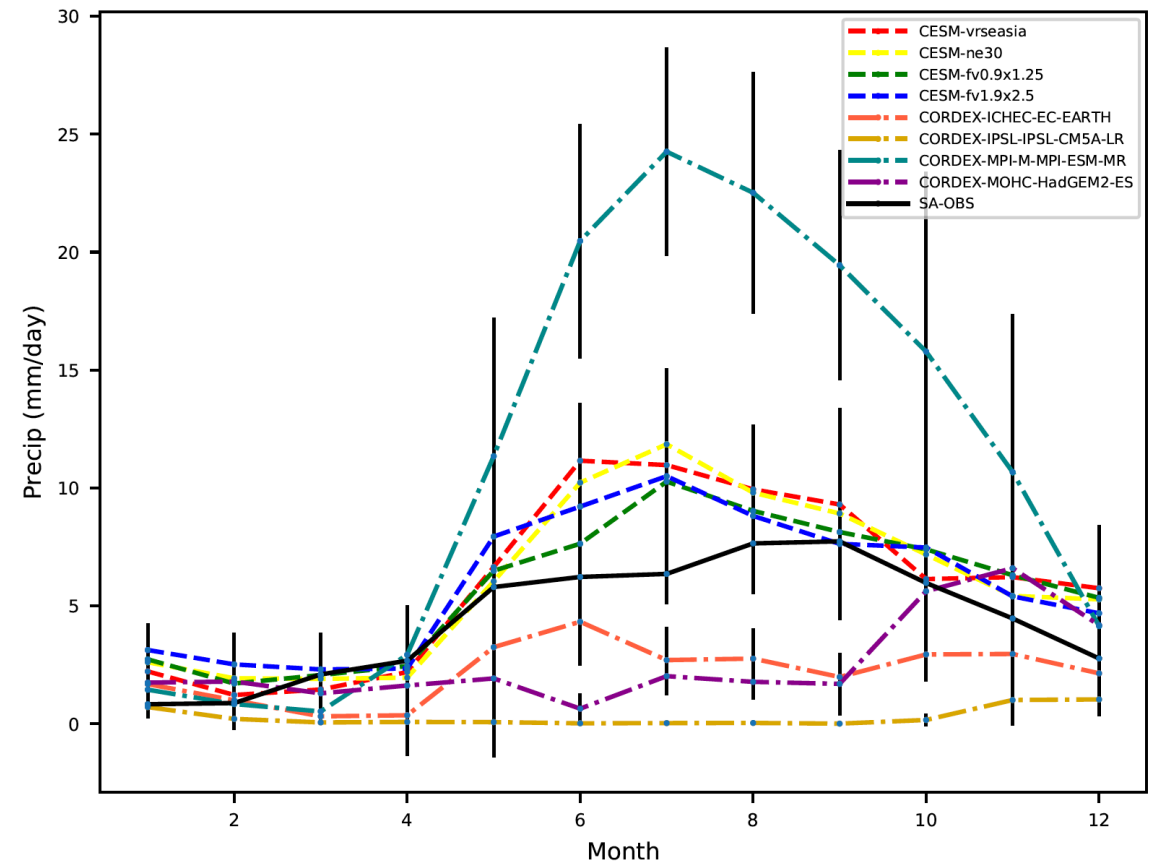


# Supplementary slides

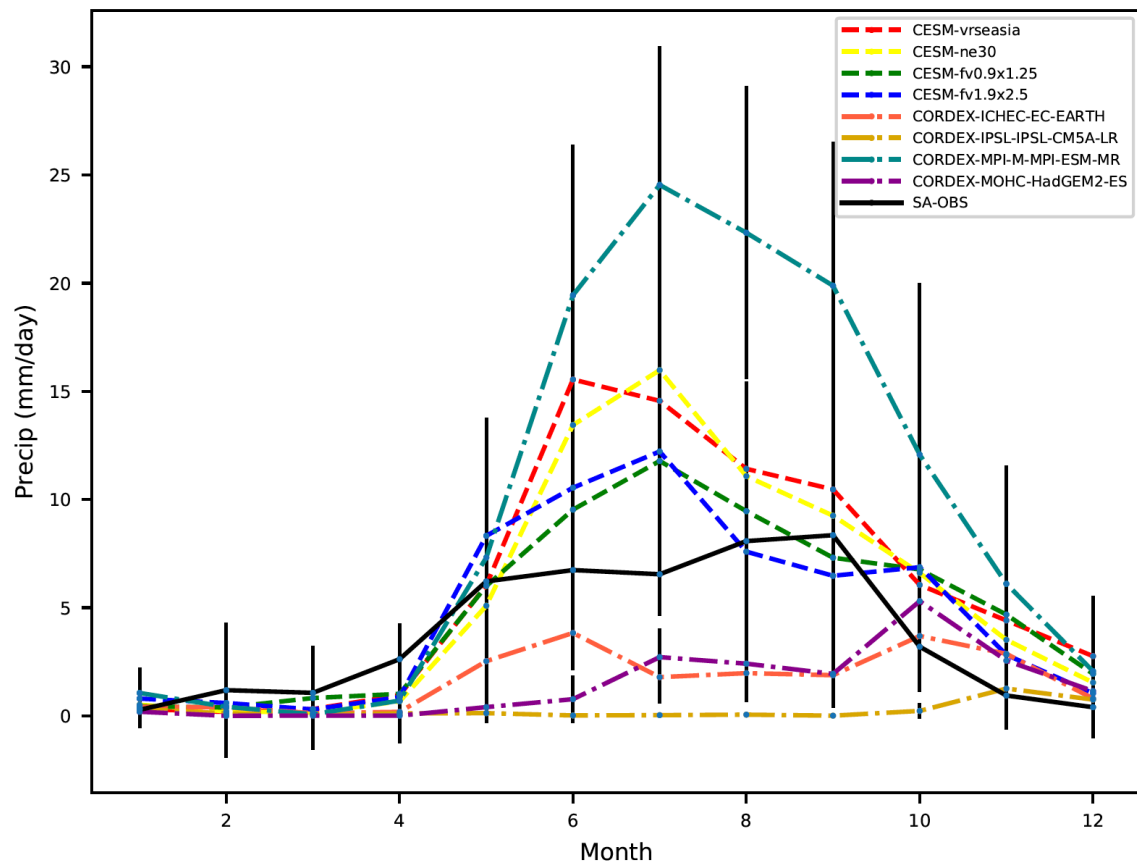
## Group 0: Seasonal cycle



## Group 1: Seasonal cycle

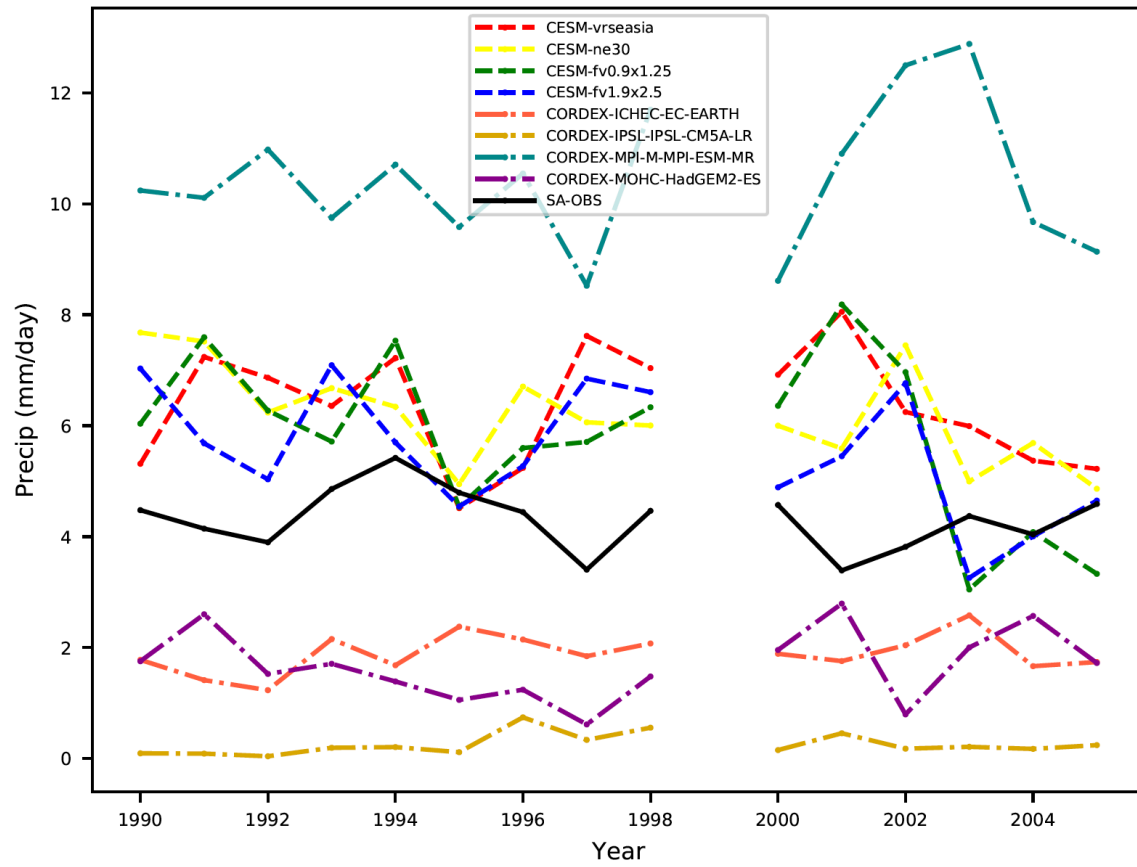


## Group 2: Seasonal cycle

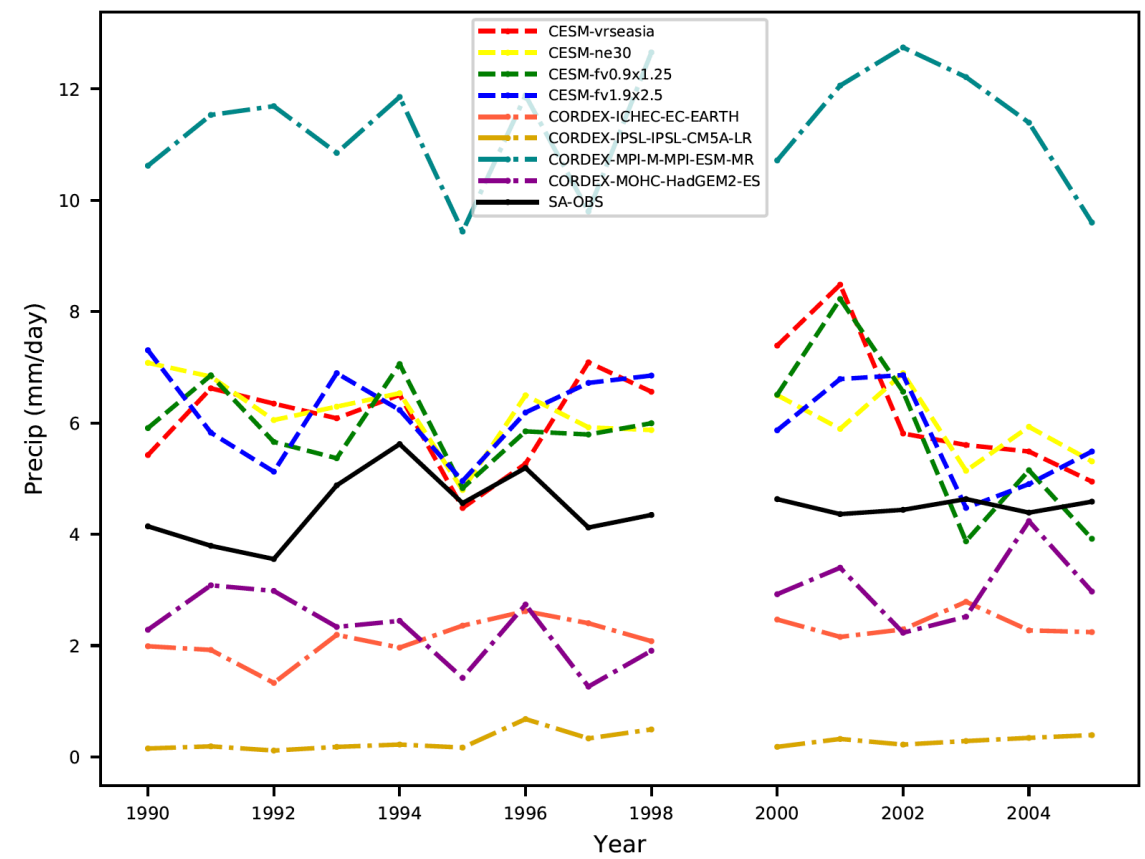


# Supplementary slides

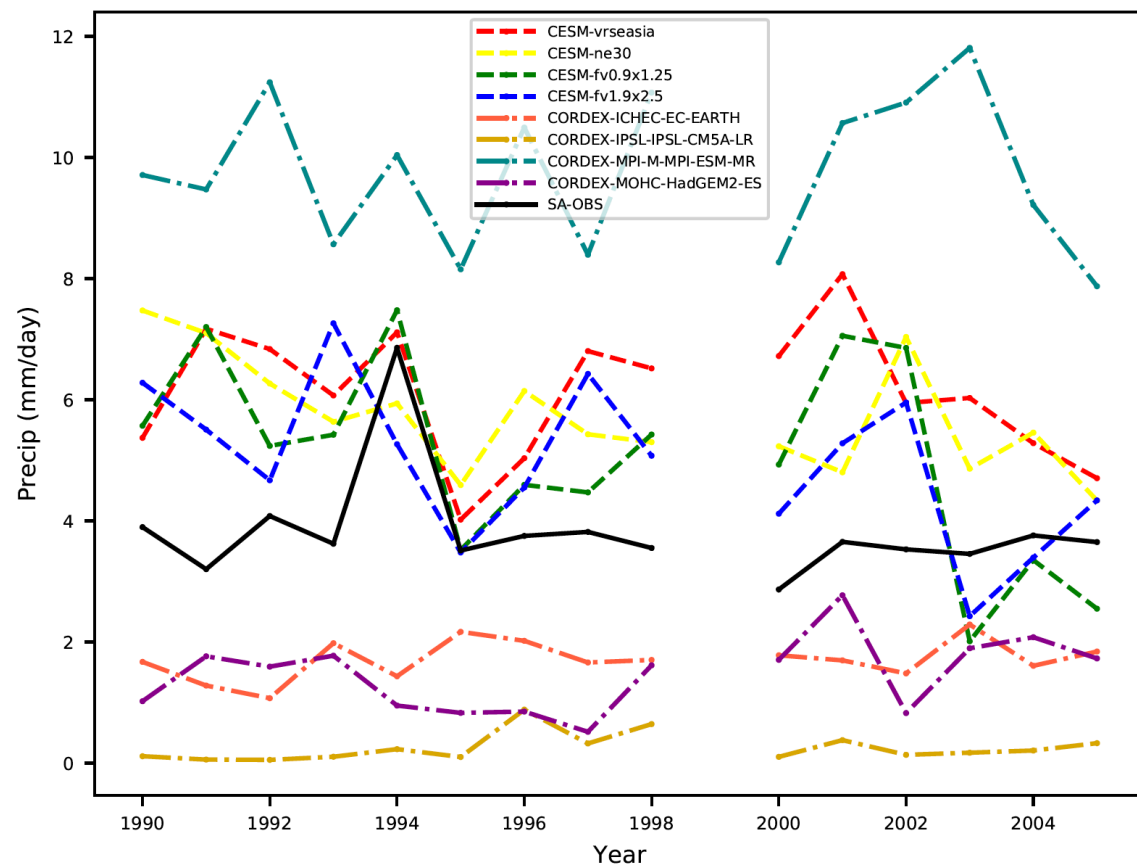
## Group 0: Annual mean precipitation



## Group 1: Annual mean precipitation

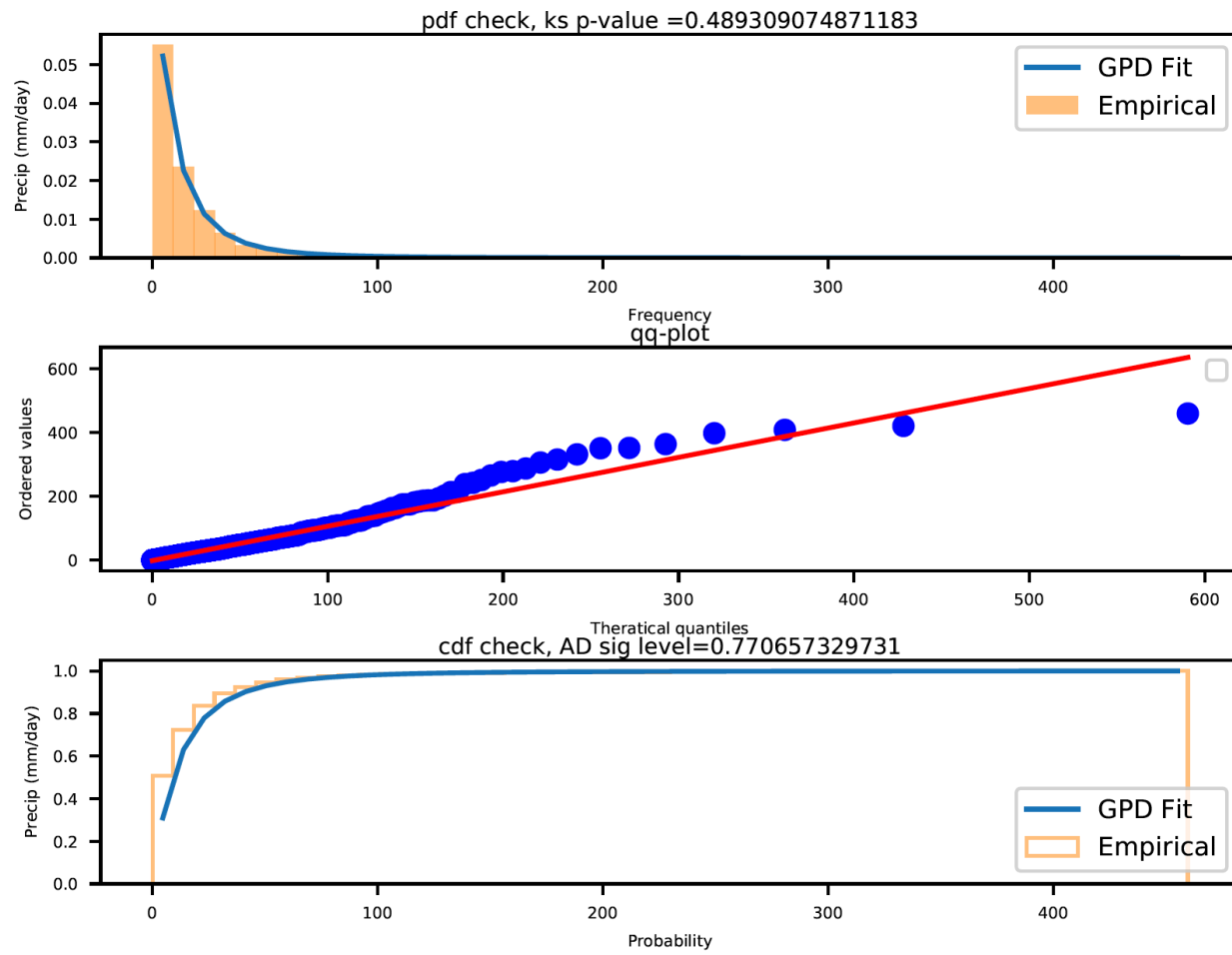


## Group 2: Annual mean precipitation

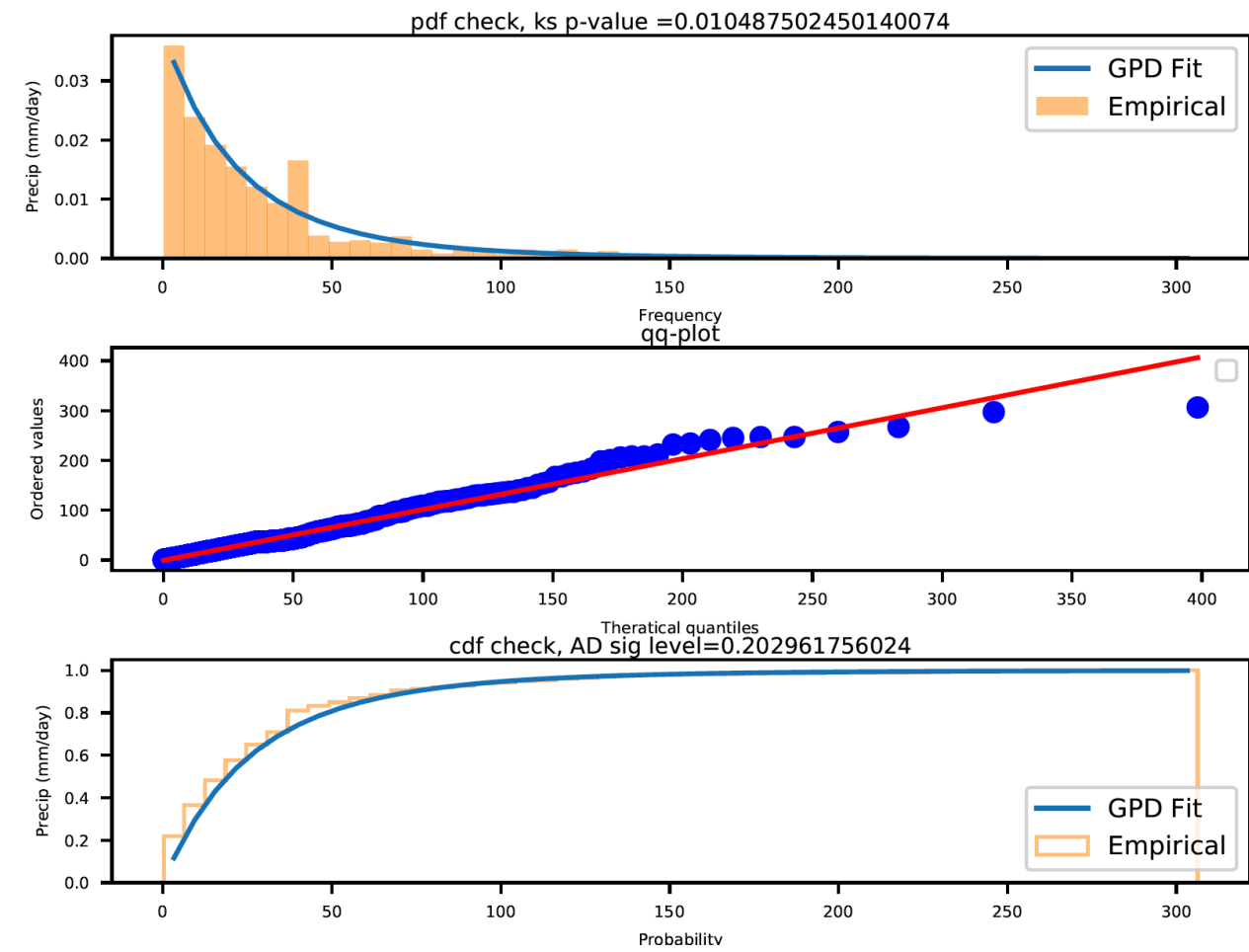


# Supplementary slides

## Goodness of GPD fitting: 97th precipitation



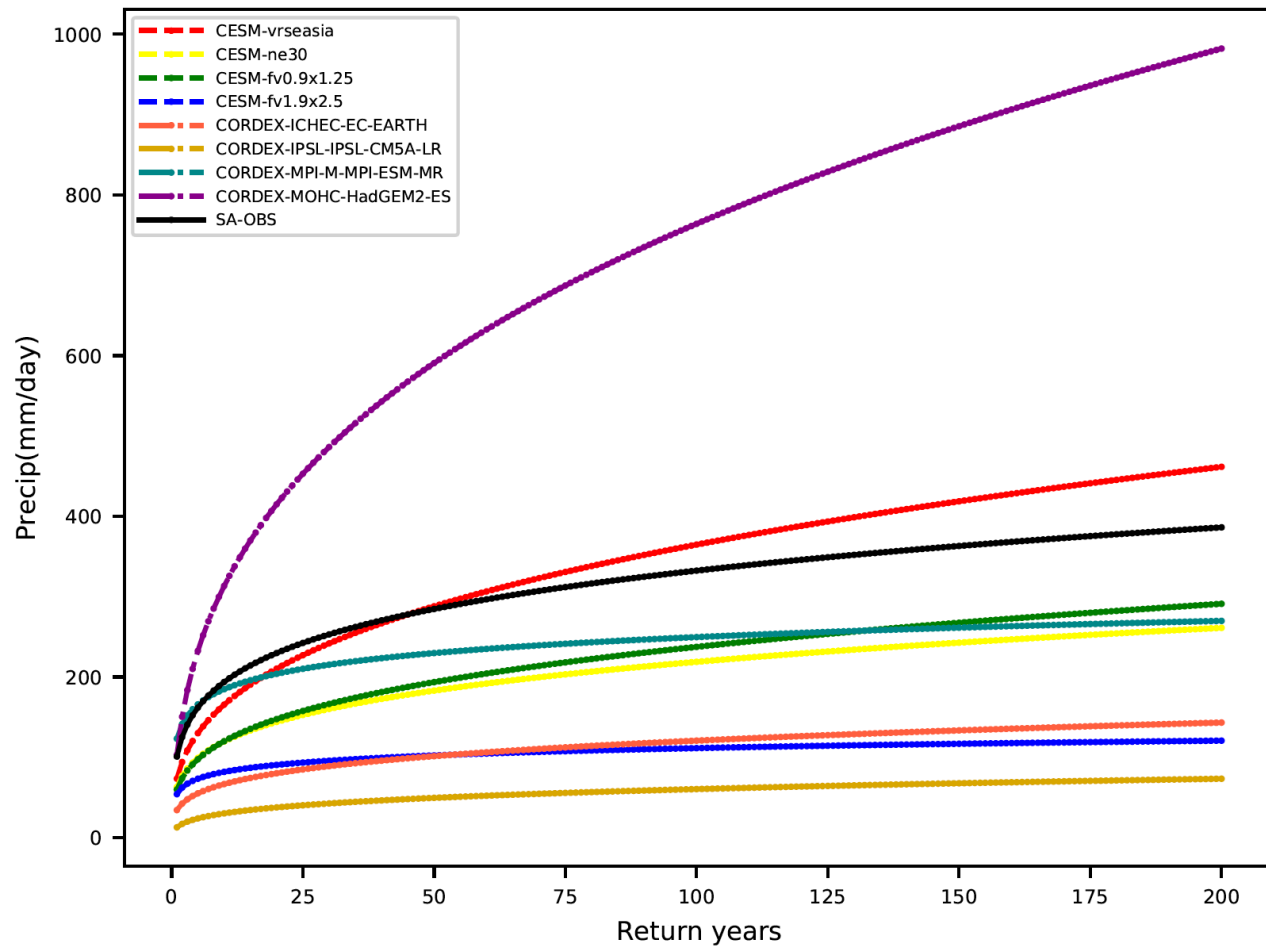
## Goodness of GPD fitting: 99th precipitation





# Supplementary slides

## N-year returns for 99th precipitation



## return levels for 99th precipitation

