

Attribution of Seasonal Climate Anomalies

May-June-July 2023

(<https://www.cpc.ncep.noaa.gov/products/people/mchen/AttributionAnalysis/>)

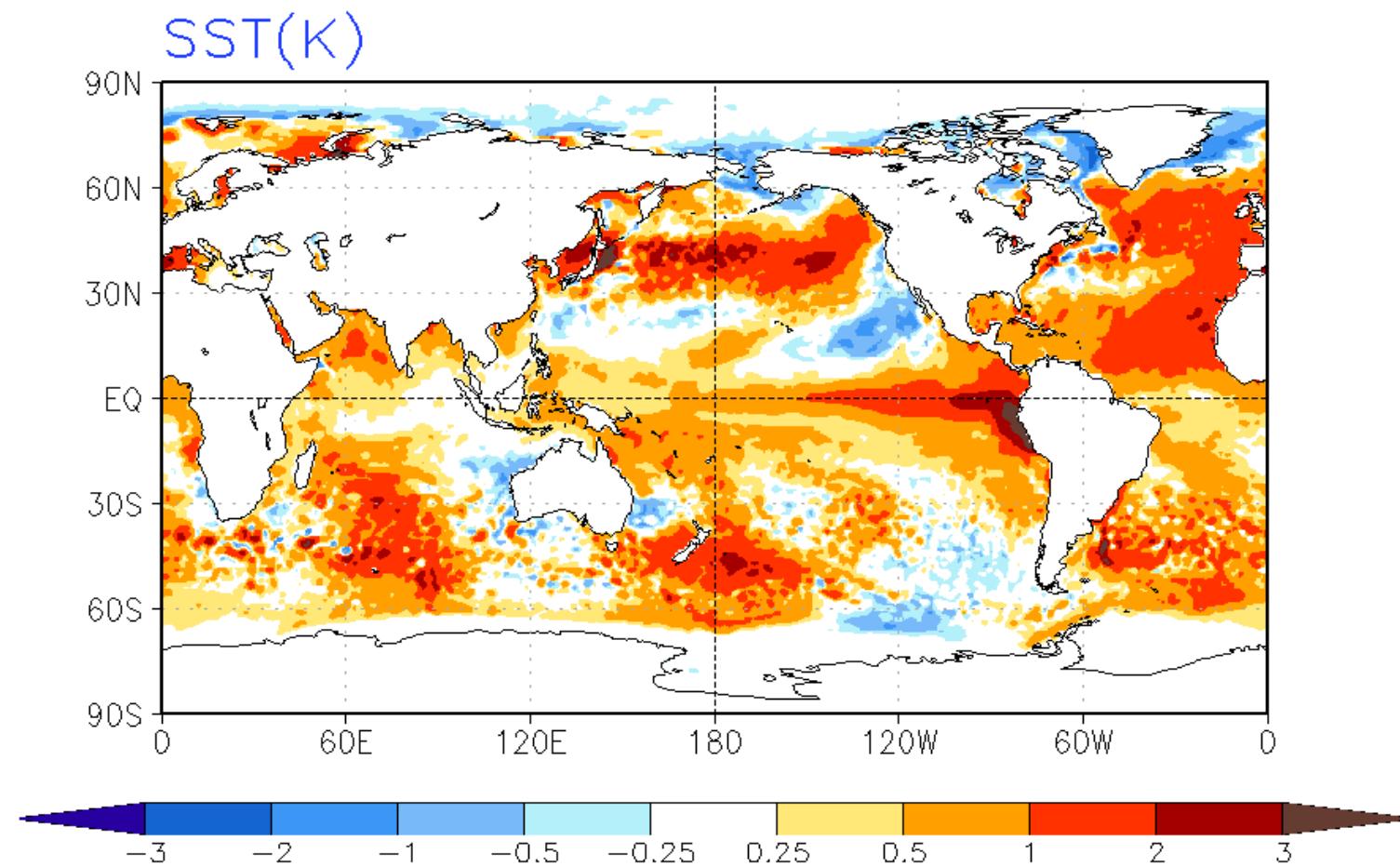
Summary of Observed Conditions and Outlooks

- The MJJ2023 SST anomalies indicated El Niño conditions over the equatorial eastern Pacific with very warm SST anomalies along the coastal regions of South America. Over the most of other ocean basins, specifically, over the North Pacific, central southern Pacific, Atlantic, northern and southern Indian Ocean, the SST warm anomalies were persisted as well (slide 4). The CFSv2 predicted well the large scale variations of observed SST anomalies (slide 10).
- The AMIP simulation and the initialized forecasts, and other MME forecasts all captured the large-scale distribution of observed precipitation anomalies in tropical latitudes – below (above) normal anomalies in the equatorial eastern Indian Ocean (Maritime Continent and equatorial western Pacific) and wetter conditions stretching in along a narrow equatorial band across the entire Pacific basin (slides 11, 37-39).
- The initialized CFSv2 predicted most of the observed z200 anomalies over tropics and extra-tropics, while missing the below-normal anomalies over the Arctic (slide 12). A good prediction of z200 anomalies over the PNA region resulted in good prediction skill of North America temperature anomalies (slide 15, 16).
- The initialized CFSv2 predicted most of the observed North America precipitation anomalies as well (slide 14).
- July 2023 monthly mean forecasts from the shortest leads predicted the z200 below-normal anomalies over the central NA region, leading to a good prediction skill of NA temperature (slides 33, 34).

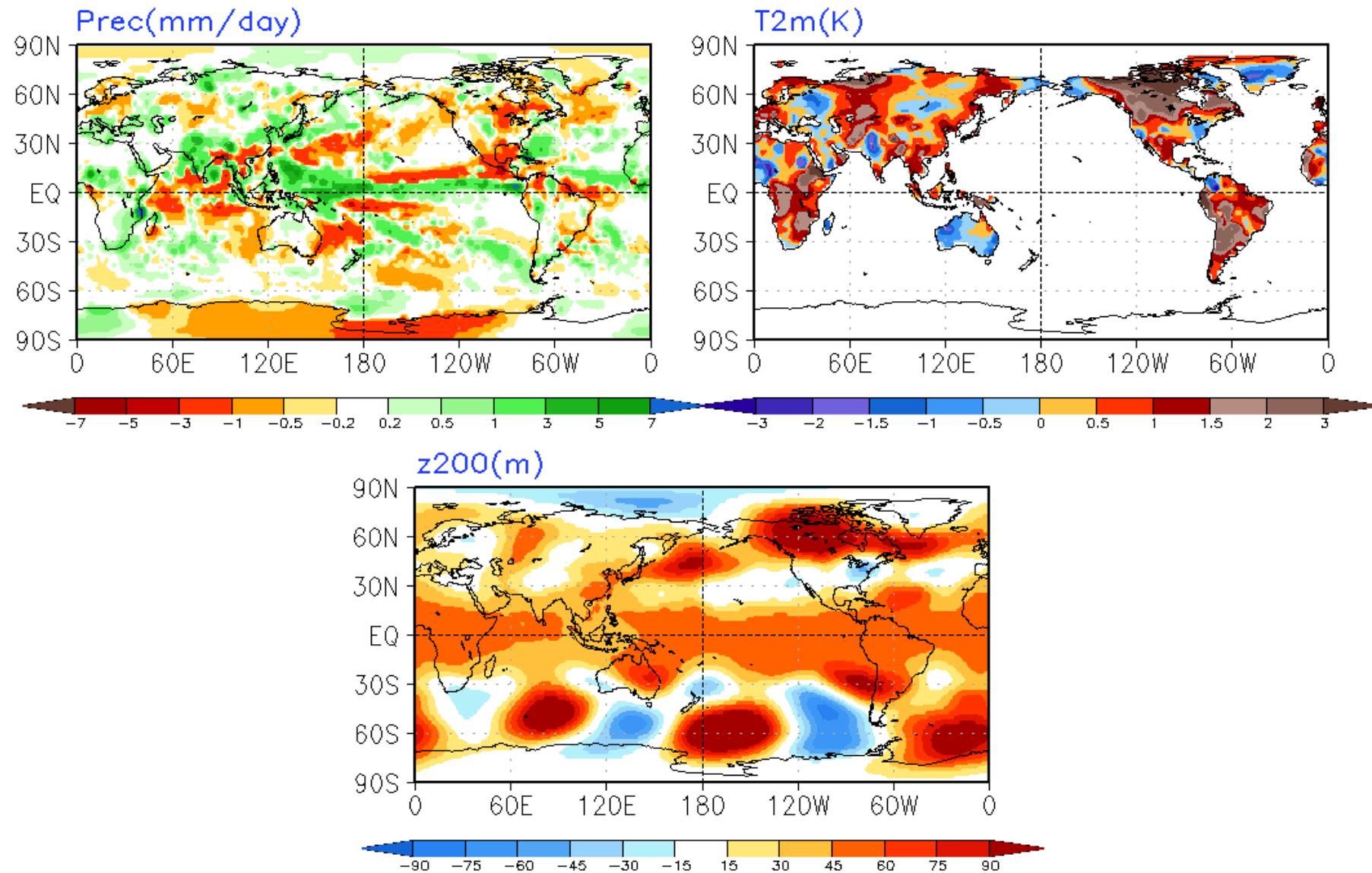
Observed Seasonal Anomalies

Global and North America

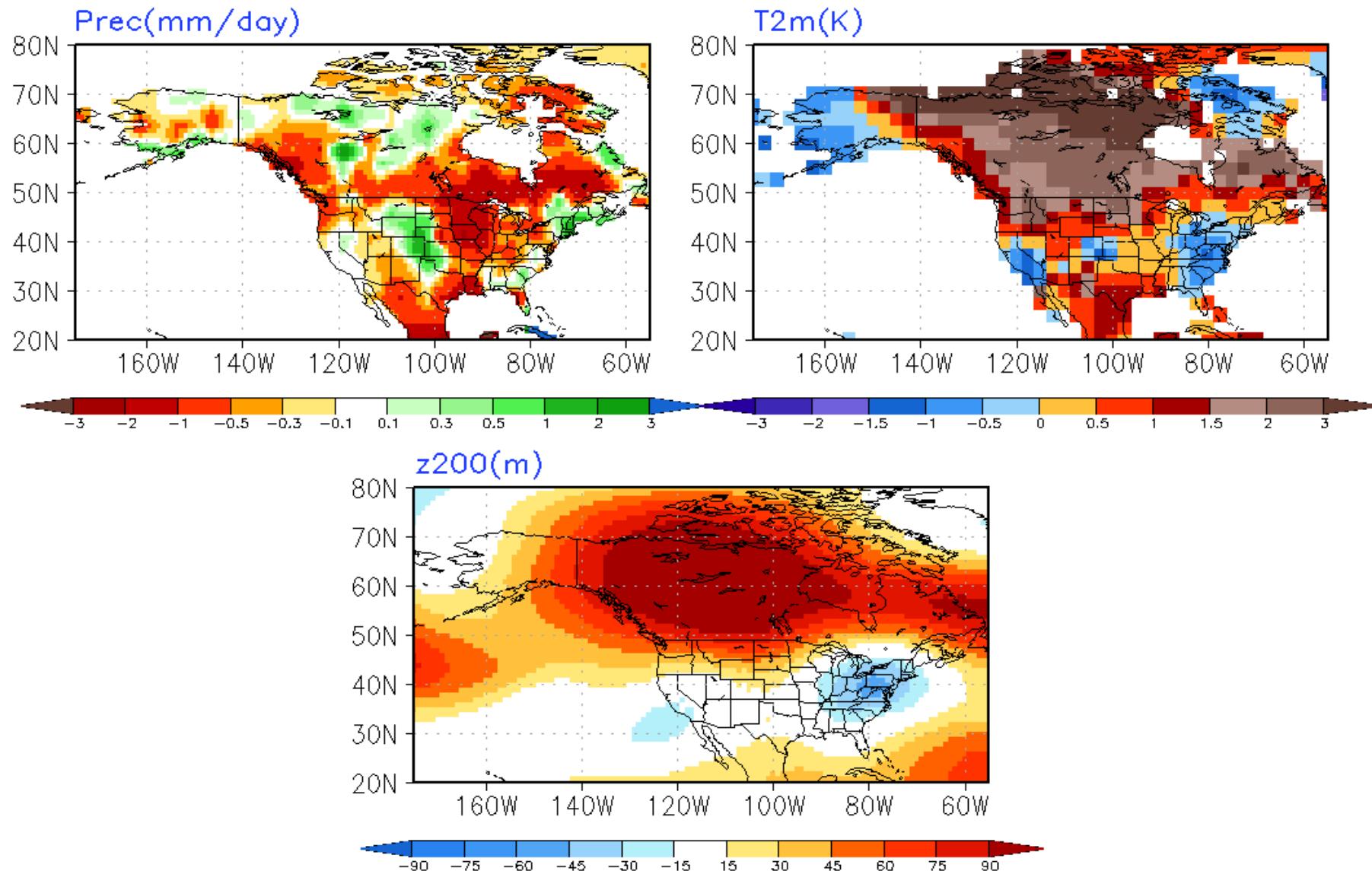
Observed Anomaly MJJ2023



Observed Anomaly MJJ2023

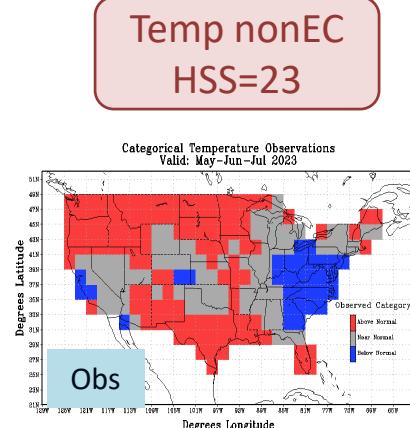


Observed Anomaly MJJ2023



CPC Seasonal Outlooks and NMME Forecasts

CPC



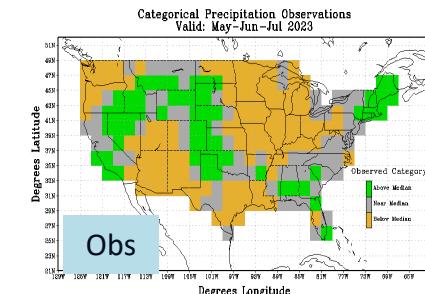
Seasonal Temperature Outlook

Valid: May-Jun-Jul 2023
Issued: April 20, 2023

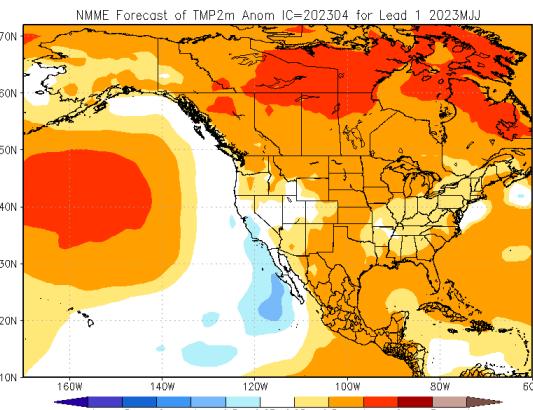
Seasonal Precipitation Outlook

Valid: May-Jun-Jul 2023
Issued: April 20, 2023

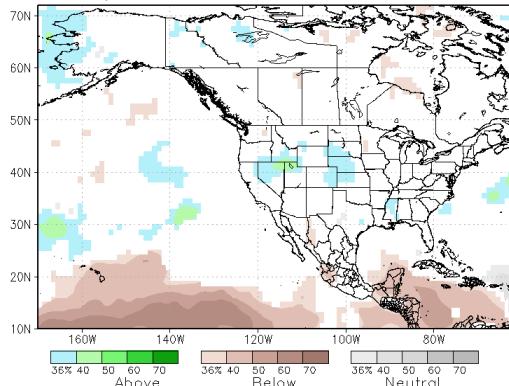
**Prec nonEC
HSS=1**



NMME



NMME prob fcst Prate IC=202304 for lead 1 2023 MJJ



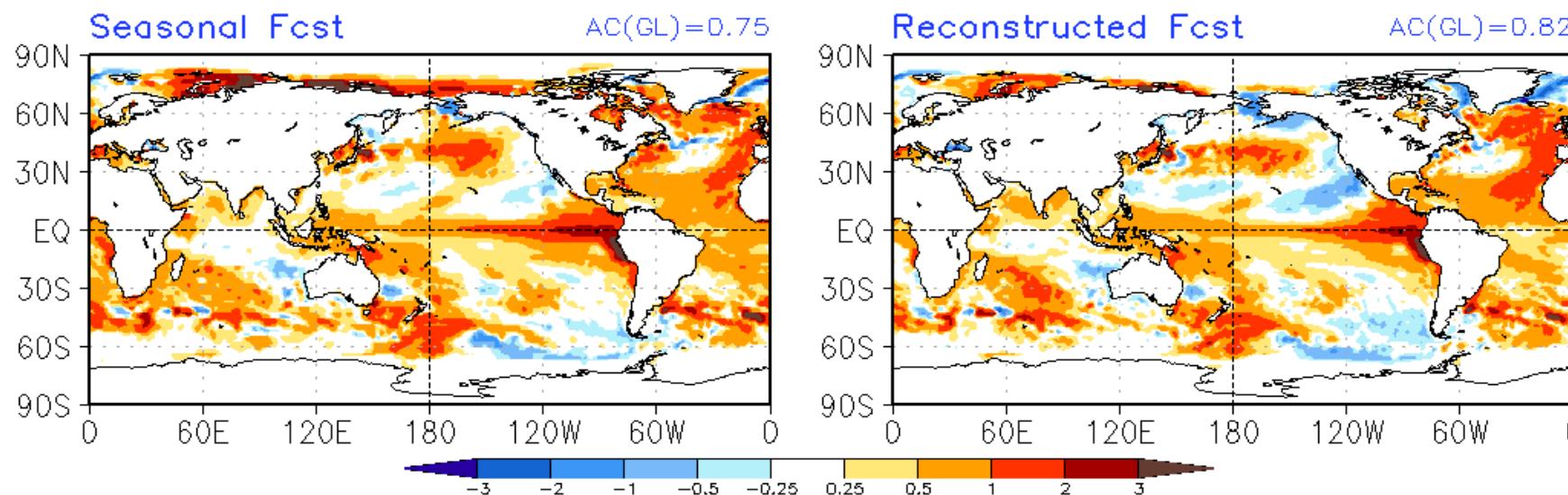
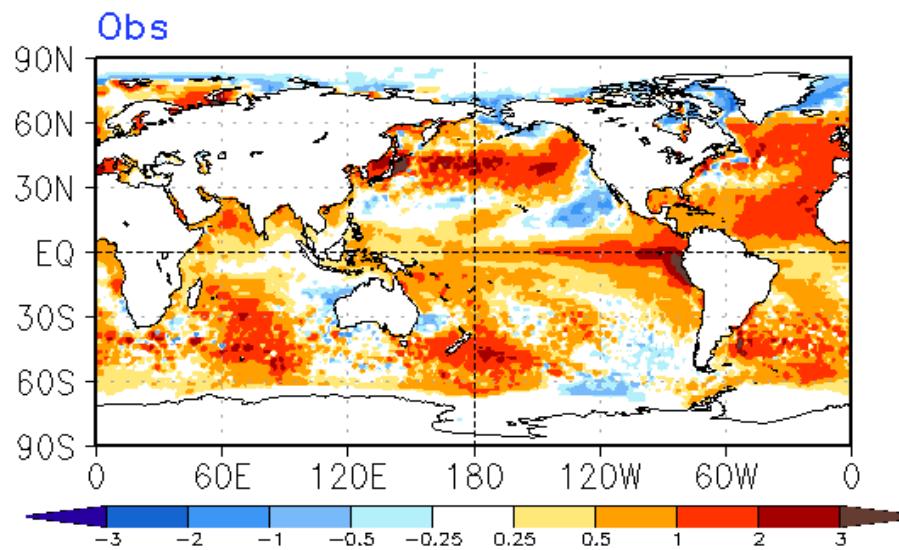
For the rationale behind CPC outlooks see: https://www.cpc.ncep.noaa.gov/products/archives/long_lead/PMD/2023/202304_PMD90D

Model Simulated/Forecast Ensemble Mean Anomalies

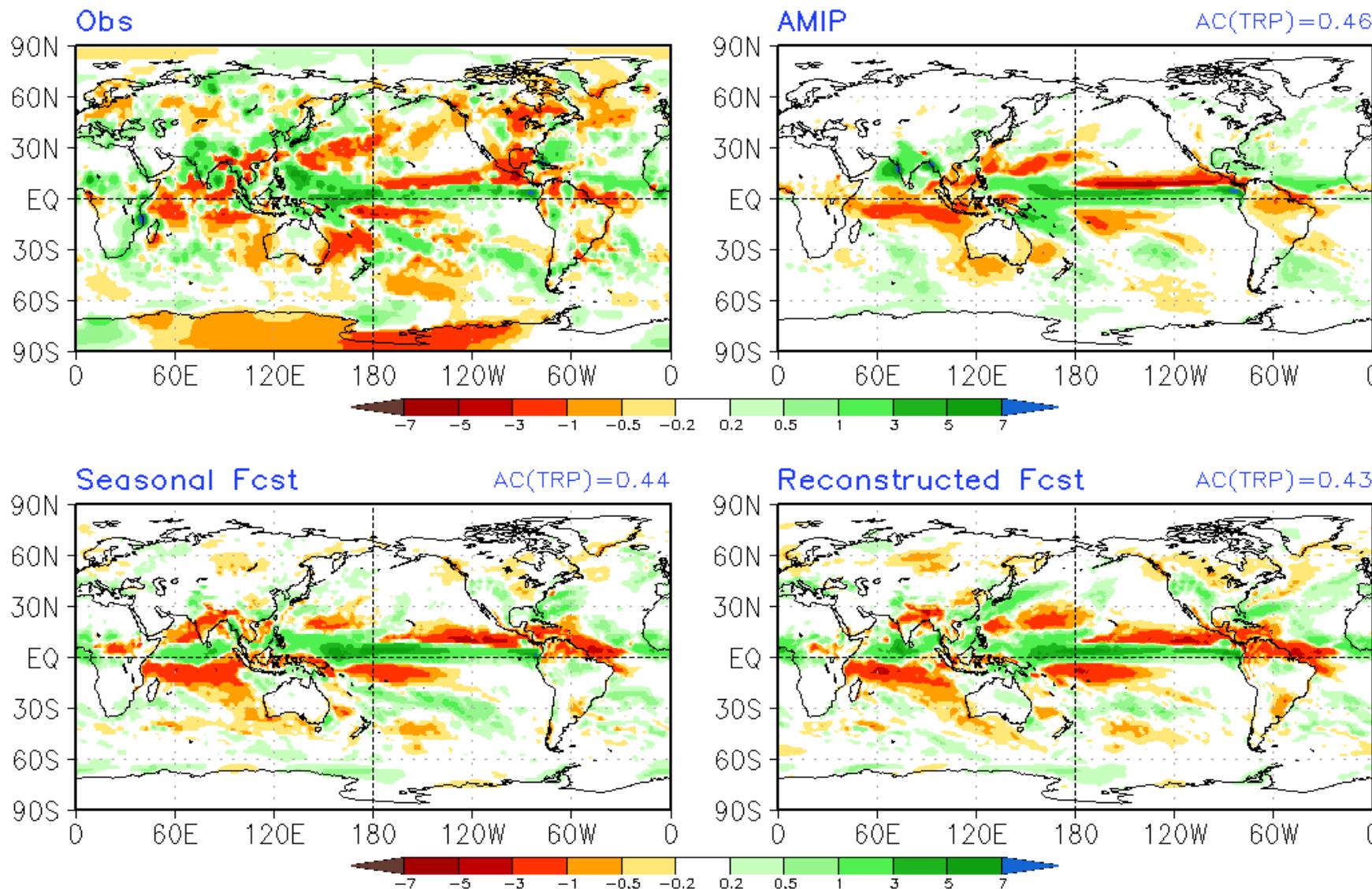
Model Simulated/Forecast Ensemble Average Anomalies

- AMIP simulations forced with observed sea surface temperatures (100 members ensemble)
- CFSv2 real time operational forecasts
 - Seasonal forecast: the seasonal mean forecasts based on 40 members from the latest 10 days before the target season (0-month-lead). For example, 2016AMJ seasonal mean forecasts are 40 members from 22-31 March2016 initial conditions.
 - Reconstructed forecast: the seasonal mean forecasts constructed from 3 individual monthly forecasts with the latest 10 days initial conditions for each individual monthly forecasts. This approach for constructing seasonal mean anomalies has more influence from the initial conditions (Kumar et al. 2013). For example, the constructed 2016AMJ seasonal mean forecasts are the average of April2016 forecasts from 22-31 March2016 initial conditions, May2016 forecasts from 21-30 April2016 initial conditions, and June2016 forecasts from 22-31 May2016 initial conditions.
- Numbers at the panels indicate the spatial anomaly correlation (AC).

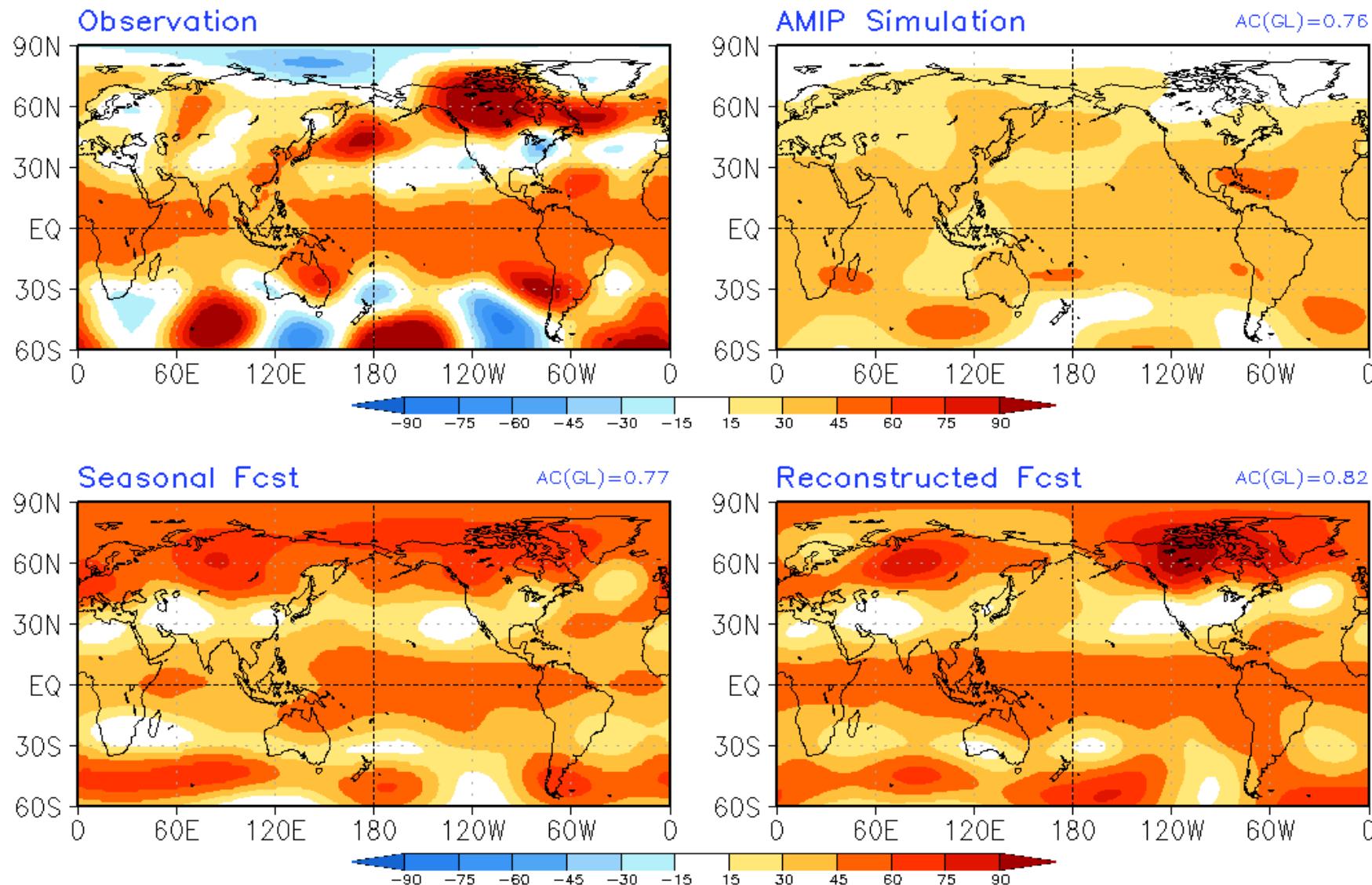
MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies SST(K)



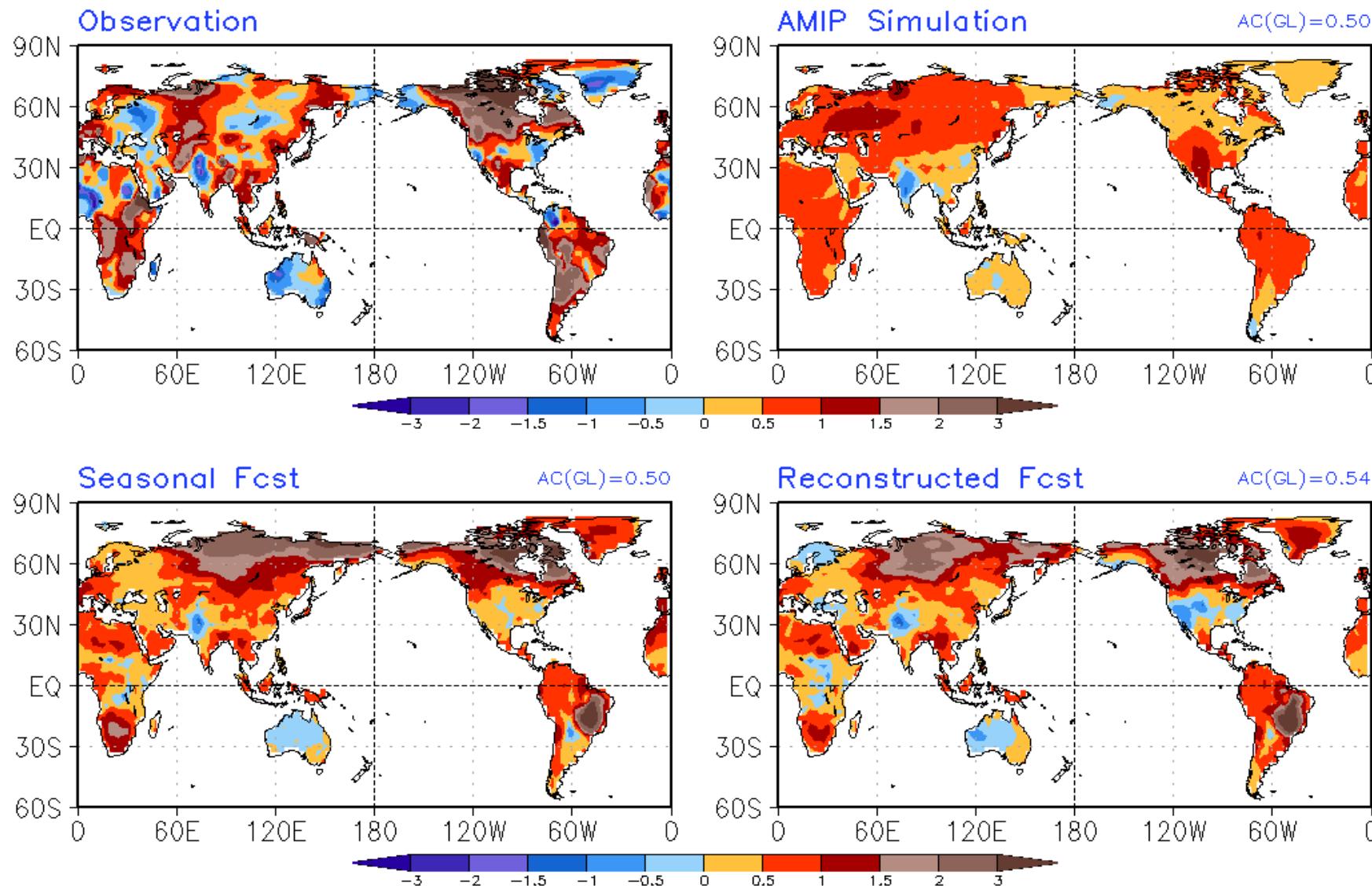
MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



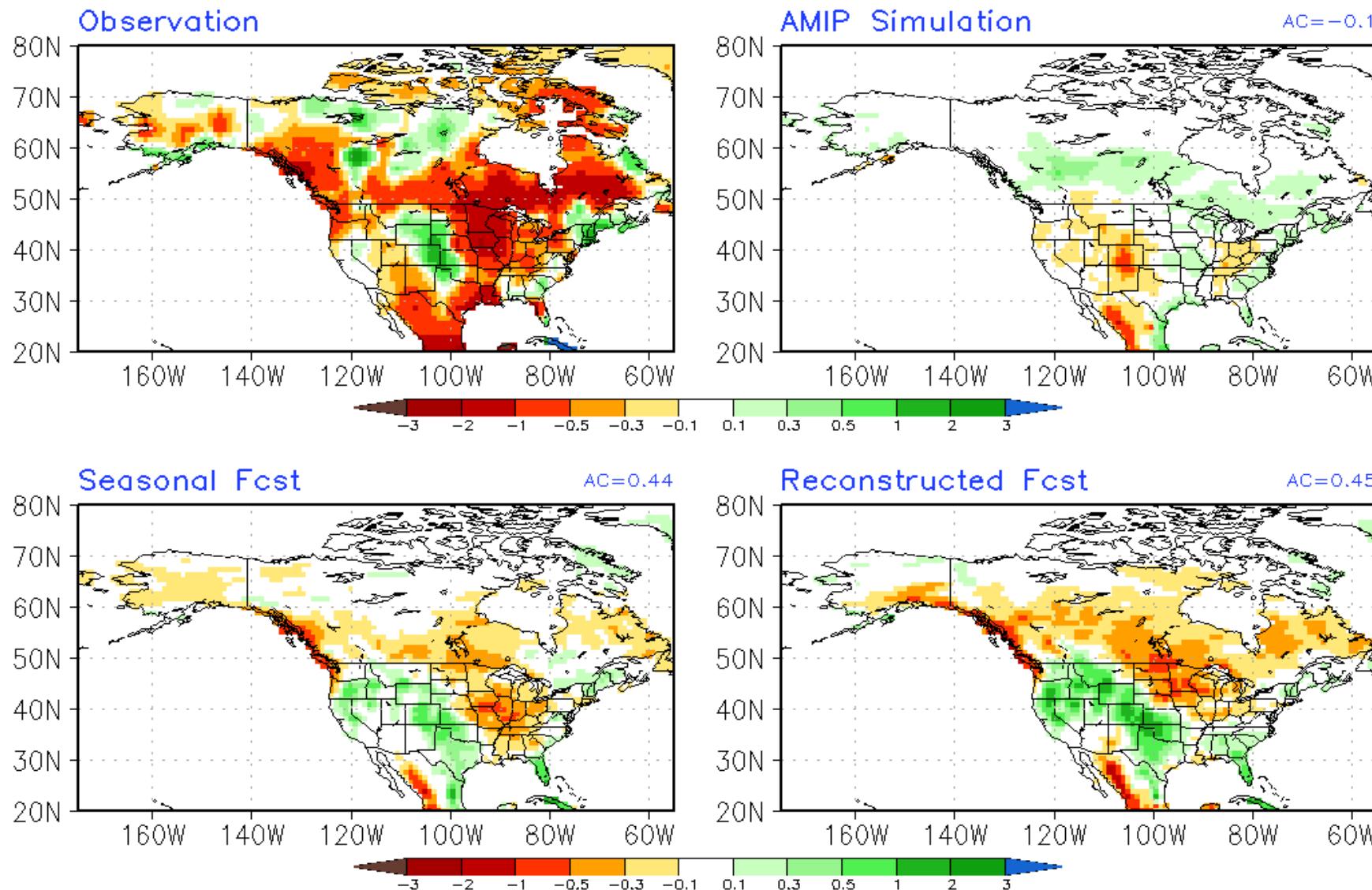
MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



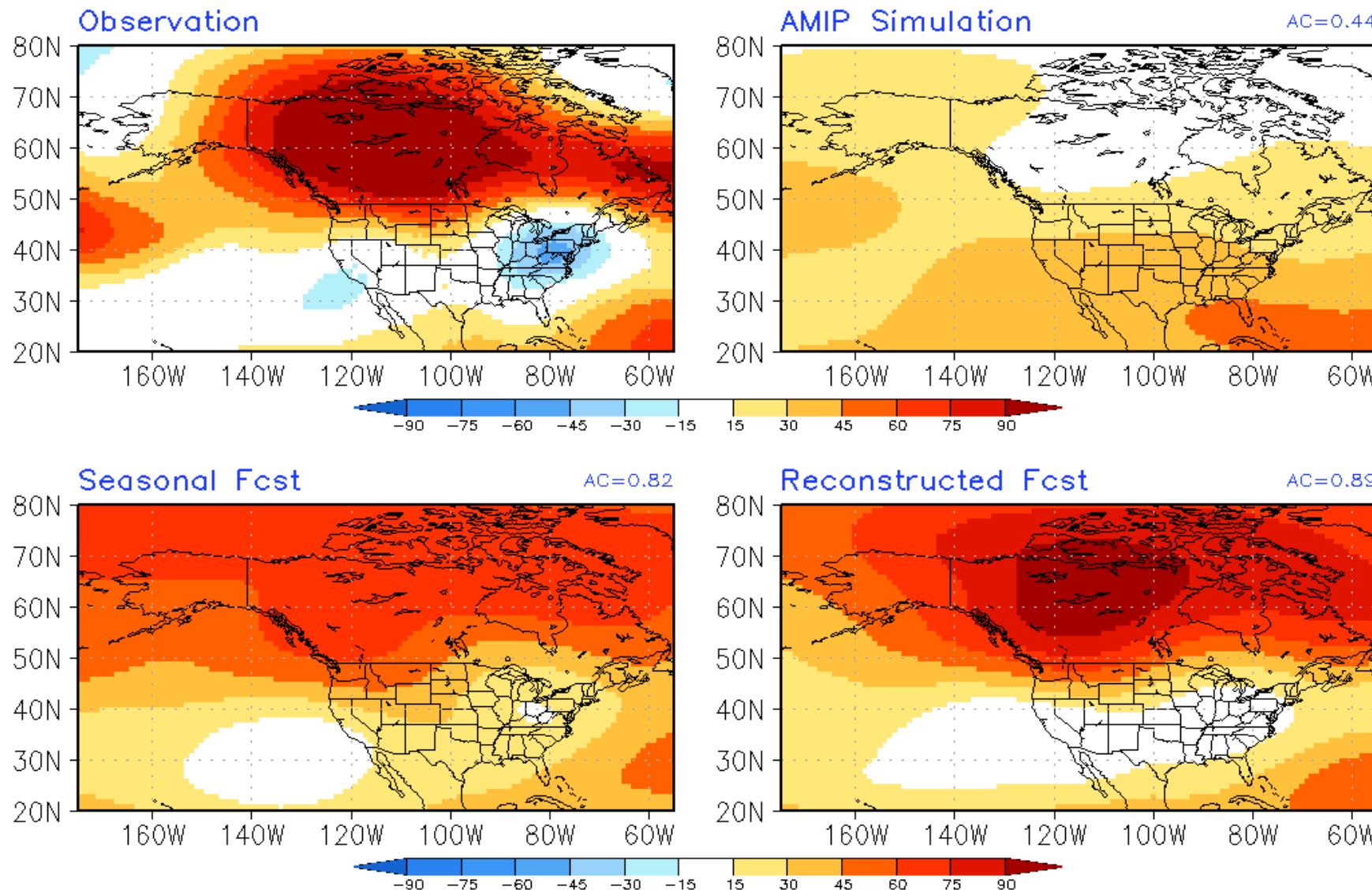
MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



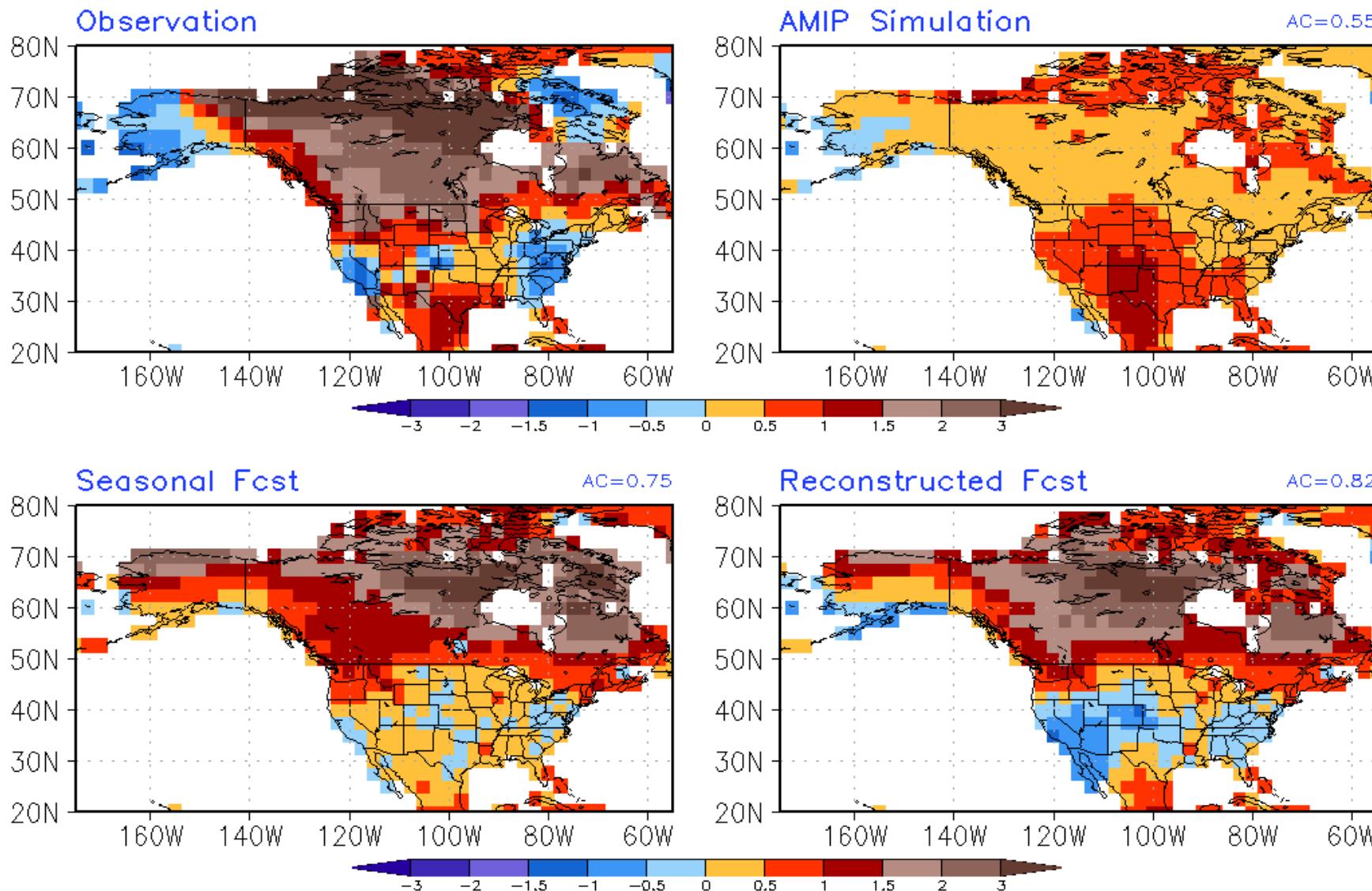
MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



MJJ2023 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)

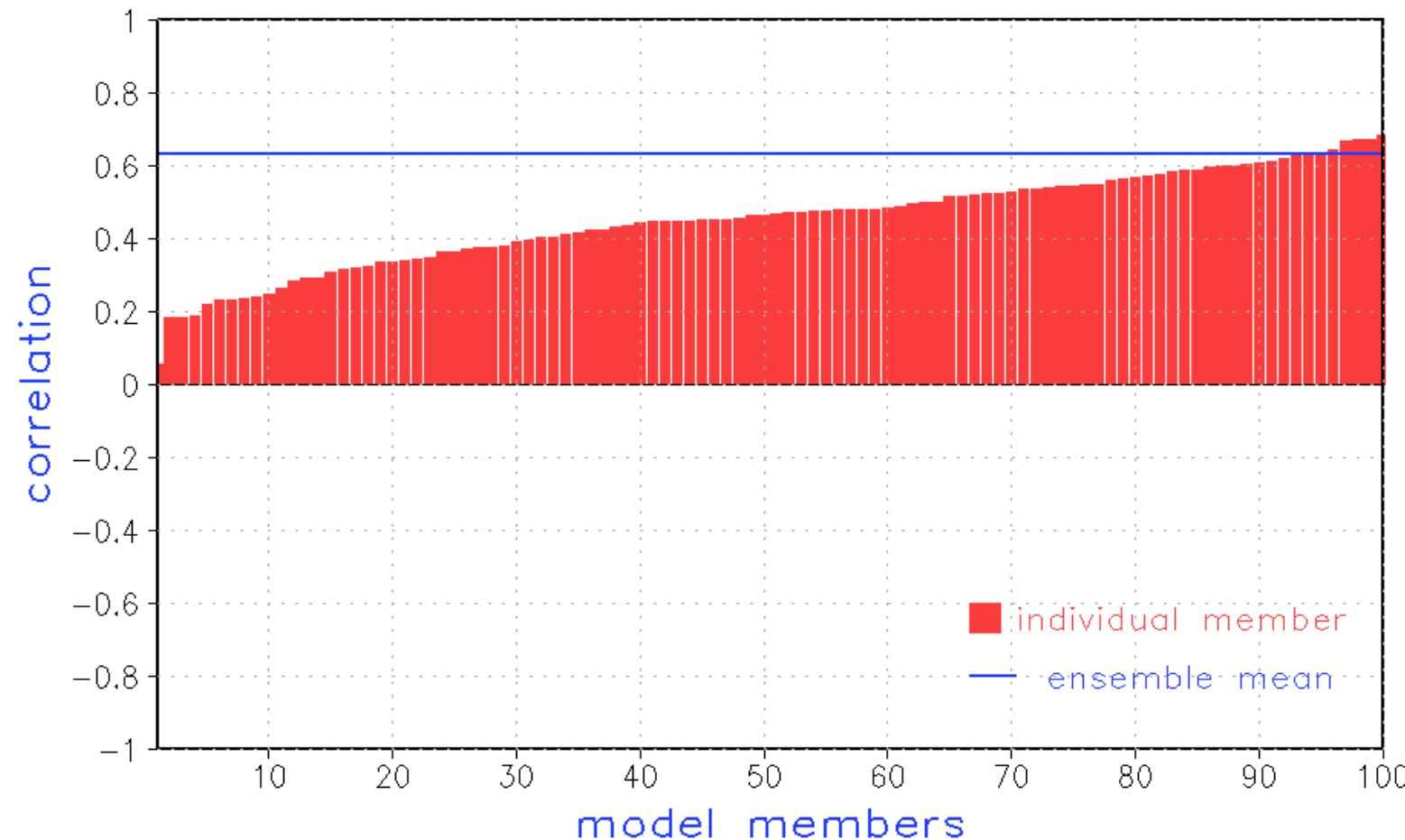


Model Simulated/Forecast Anomalies: Individual Runs

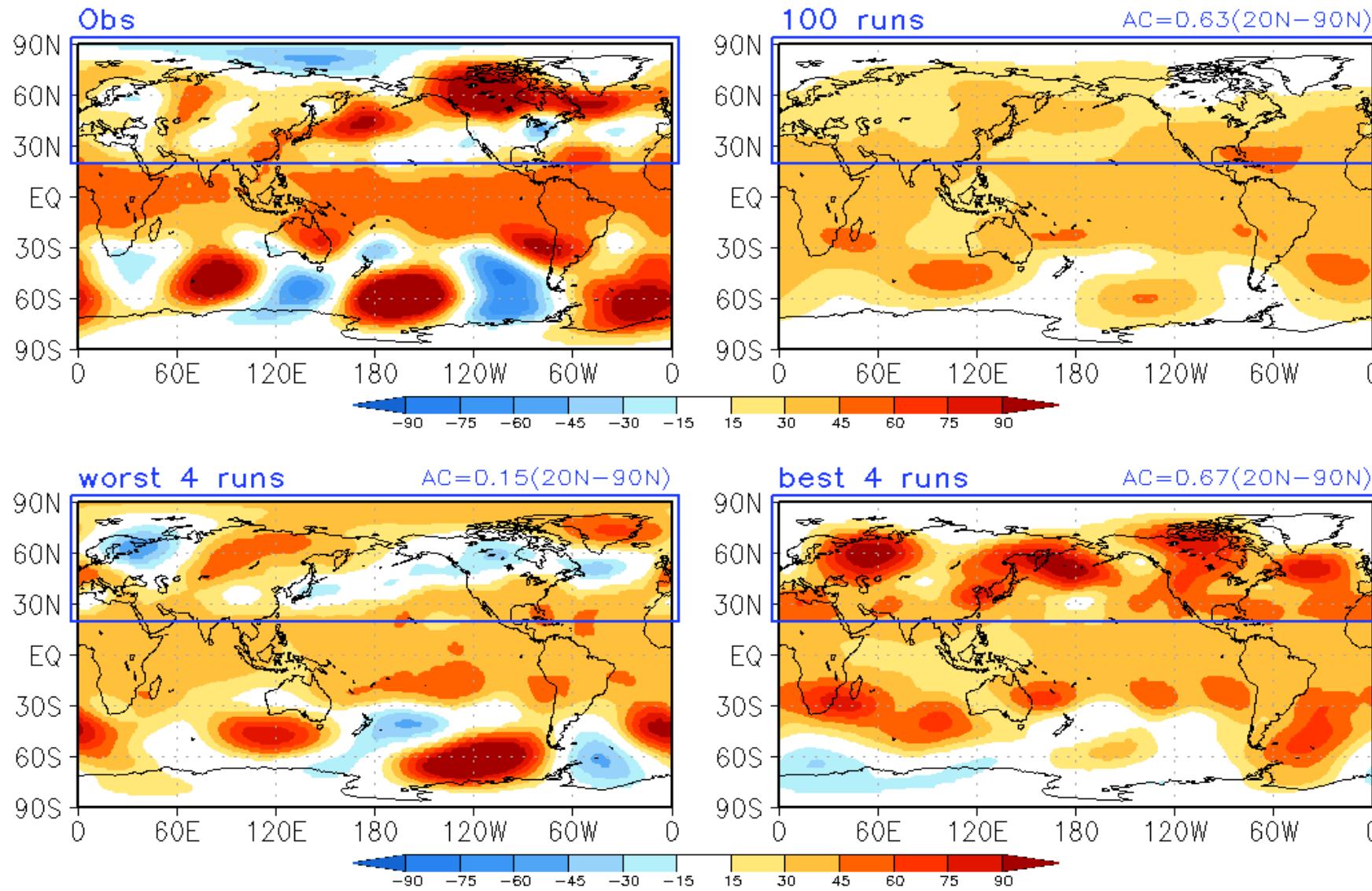
Model Simulated/Forecast Anomalies: Individual Runs

- In this analysis, anomalies from individual model runs are compared against the observed seasonal mean anomalies. The spatial resemblance between them is quantified based on anomaly correlation (AC).
- The distribution of AC across all model simulations is indicative of probability of observed anomalies to have a predictable (or attributable) component.
- One can also look at best and worst match between model simulated/forecast anomalies to assess the range of possible seasonal mean outcomes.
- For further details see: Kumar, A., M. Chen, M. Hoerling, and J. Eischeid (2013), Do extreme climate events require extreme forcings? *Geophys. Res. Lett.*, 40, 3440-3445. [doi:10.1002/grl.50657](https://doi.org/10.1002/grl.50657).

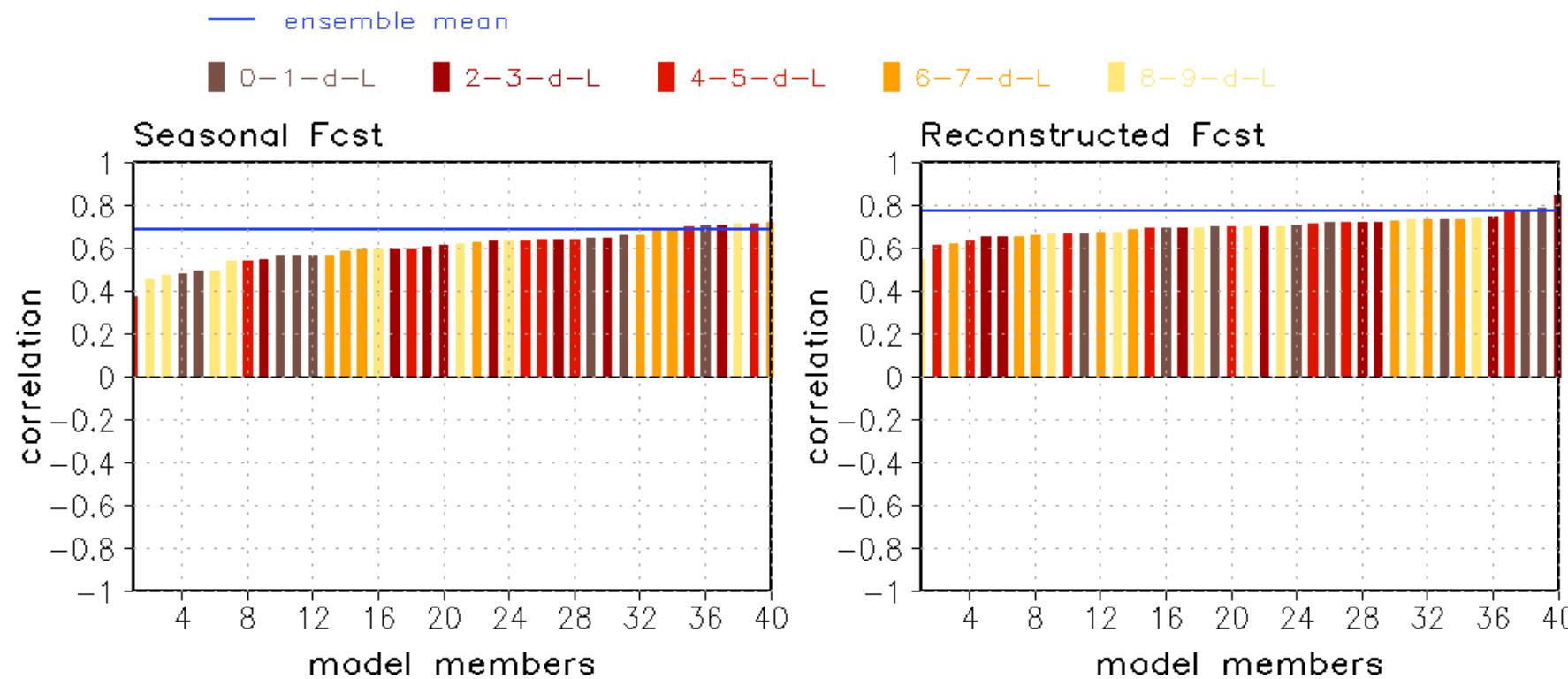
MJJ2023 Anomaly Correlation for Individual AMIP Simulation
with Observation -- z200(20N–90N)



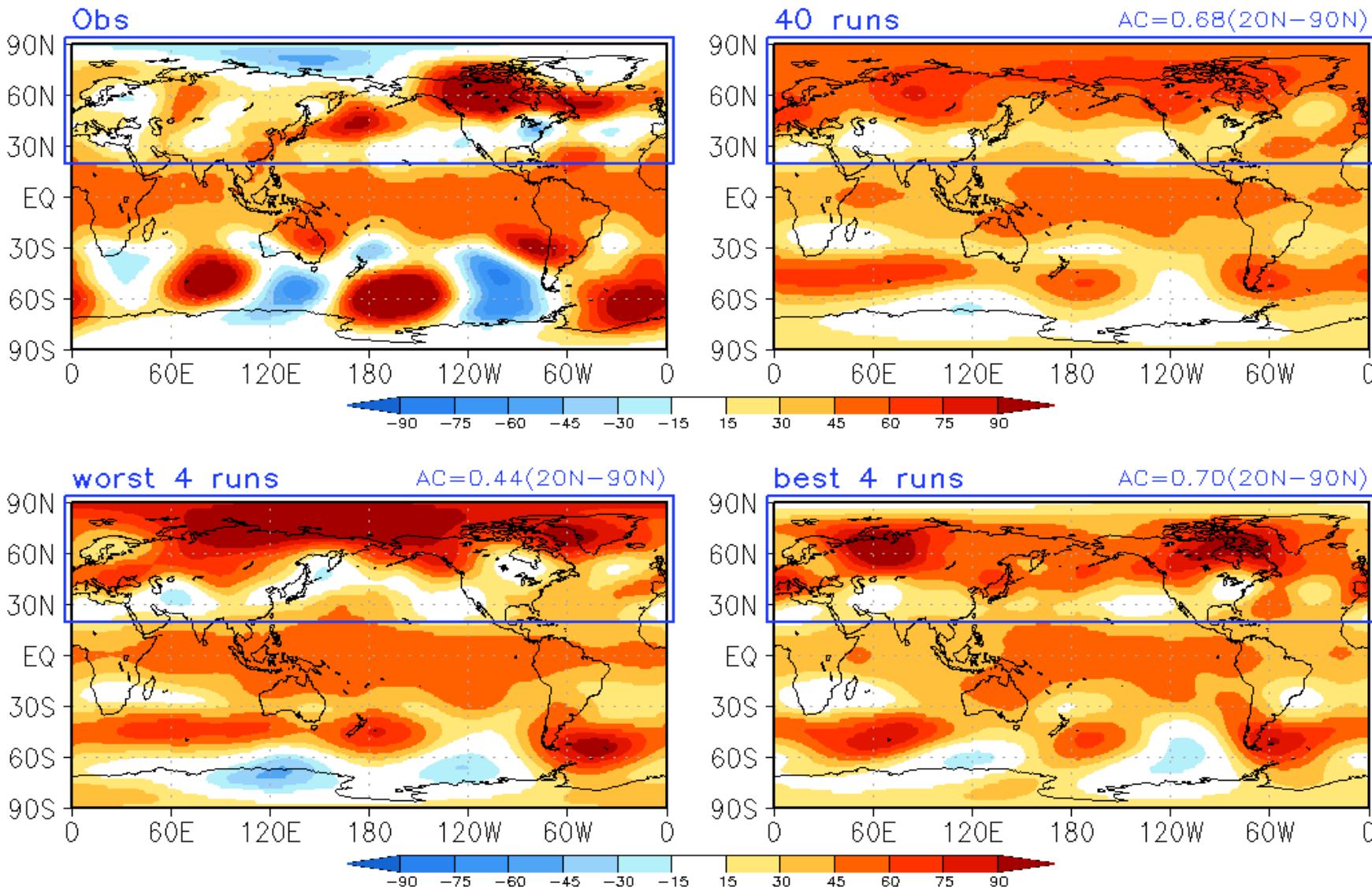
Observed & AMIP Ensemble Mean Anomalies
MJJ2023 z200(m) 100 runs/worst 4 runs/best 4 runs



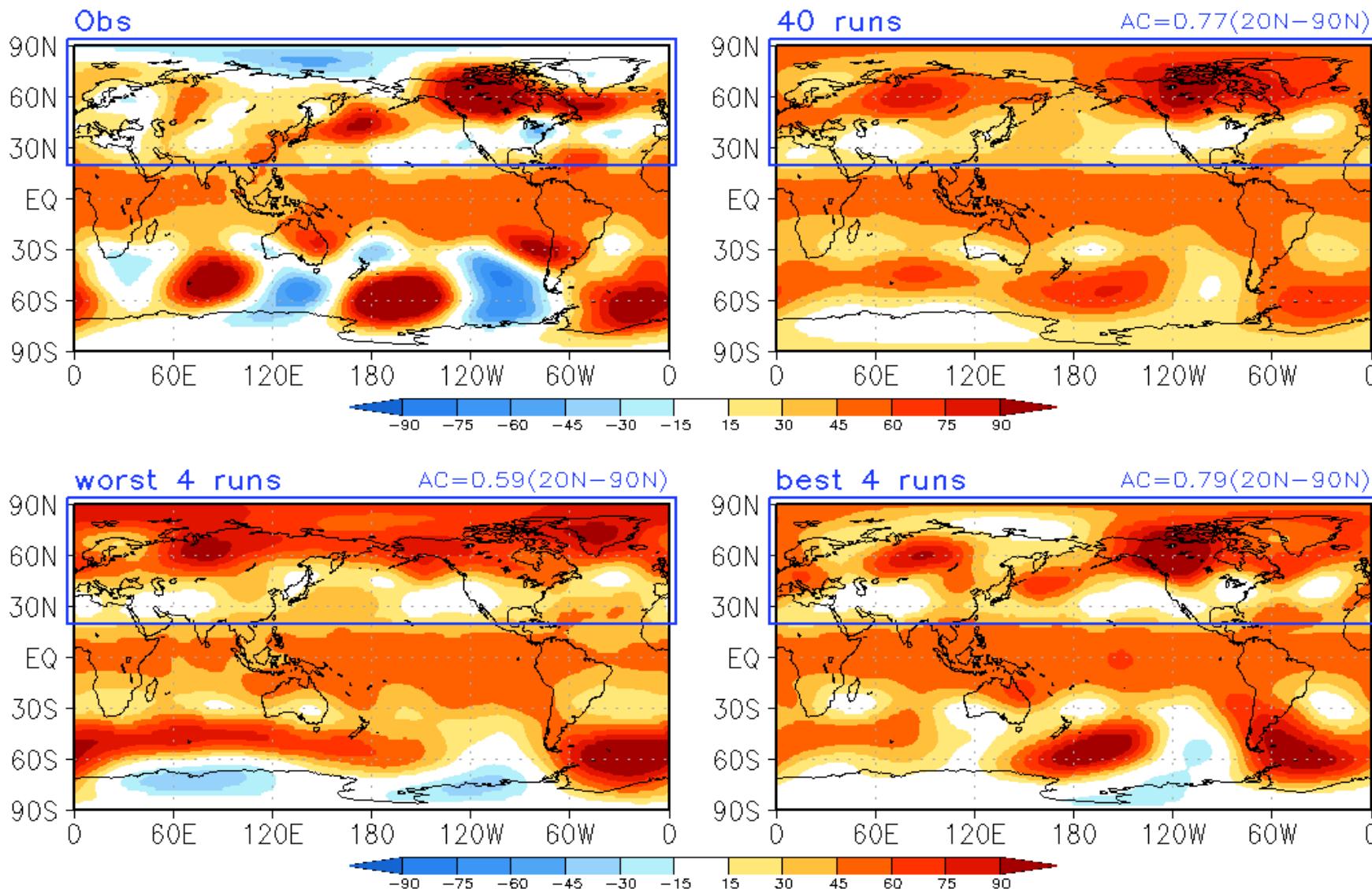
MJJ2023 Anomaly Correlation for Individual CFSv2 Forecast
with Observation -- z200 (20N–90N)



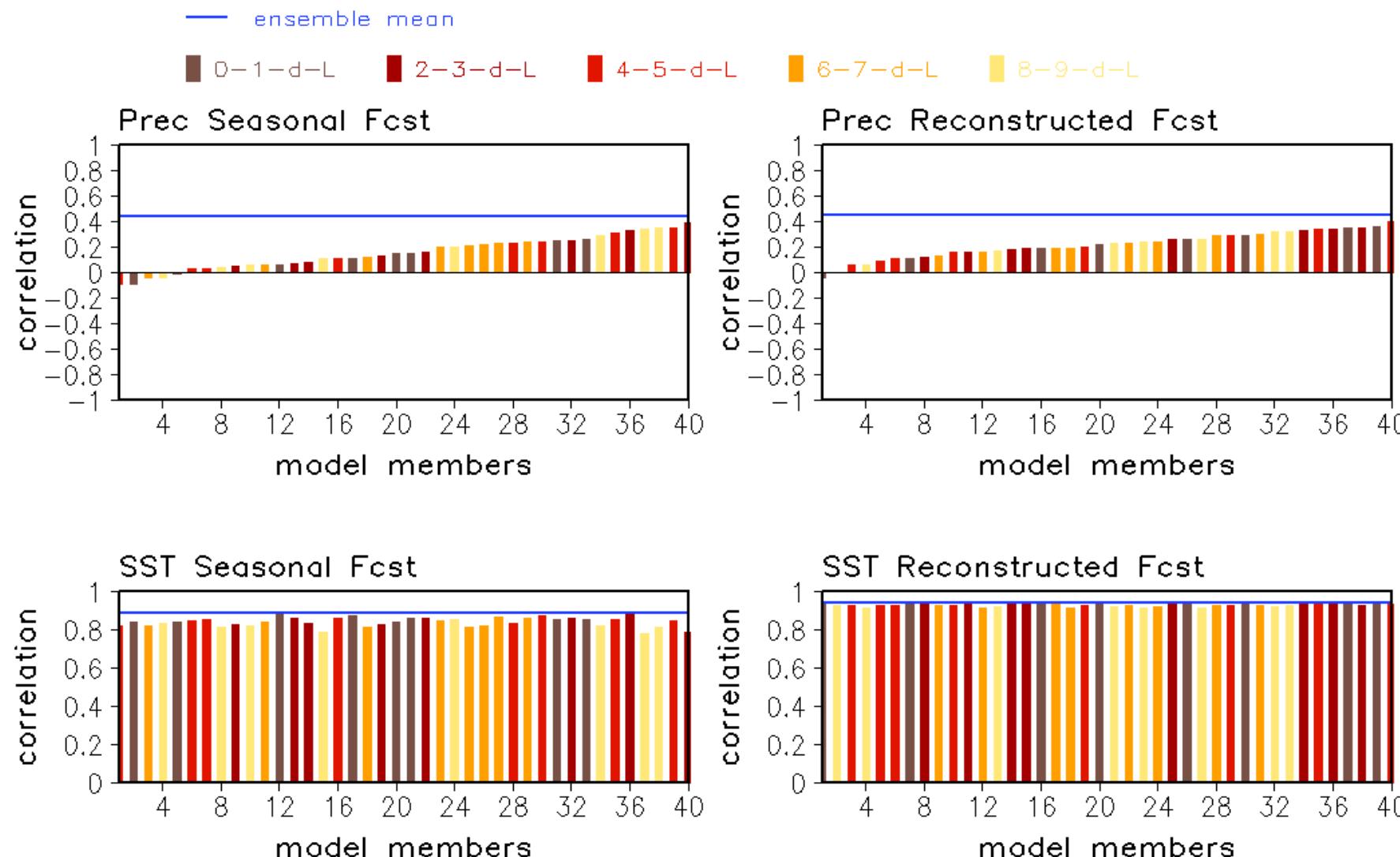
Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 z200(m) 40 runs/worst 4 runs/best 4 runs
Seasonal Forecast



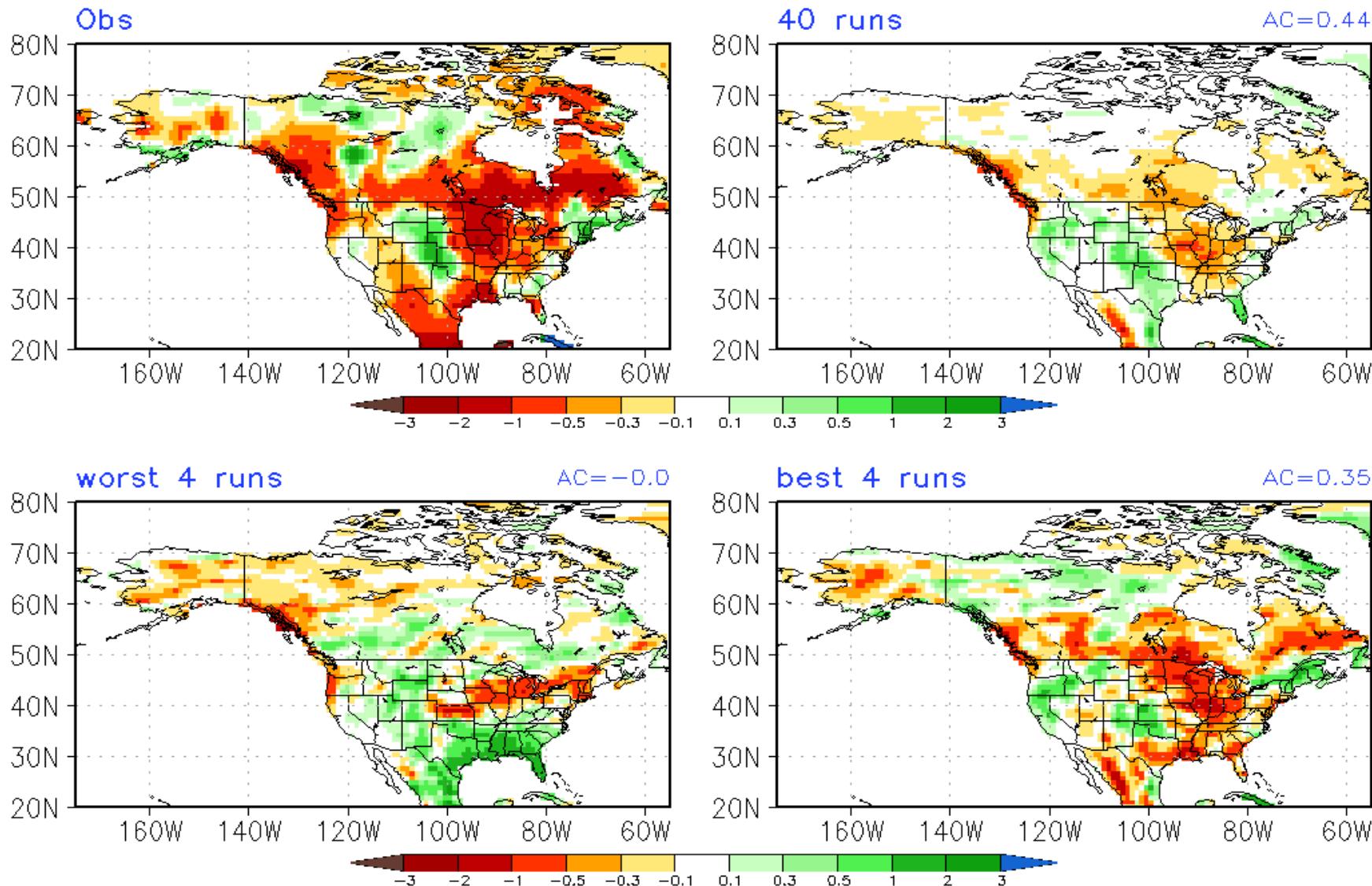
Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 z200(m) 40 runs/worst 4 runs/best 4 runs
Reconstructed Forecast



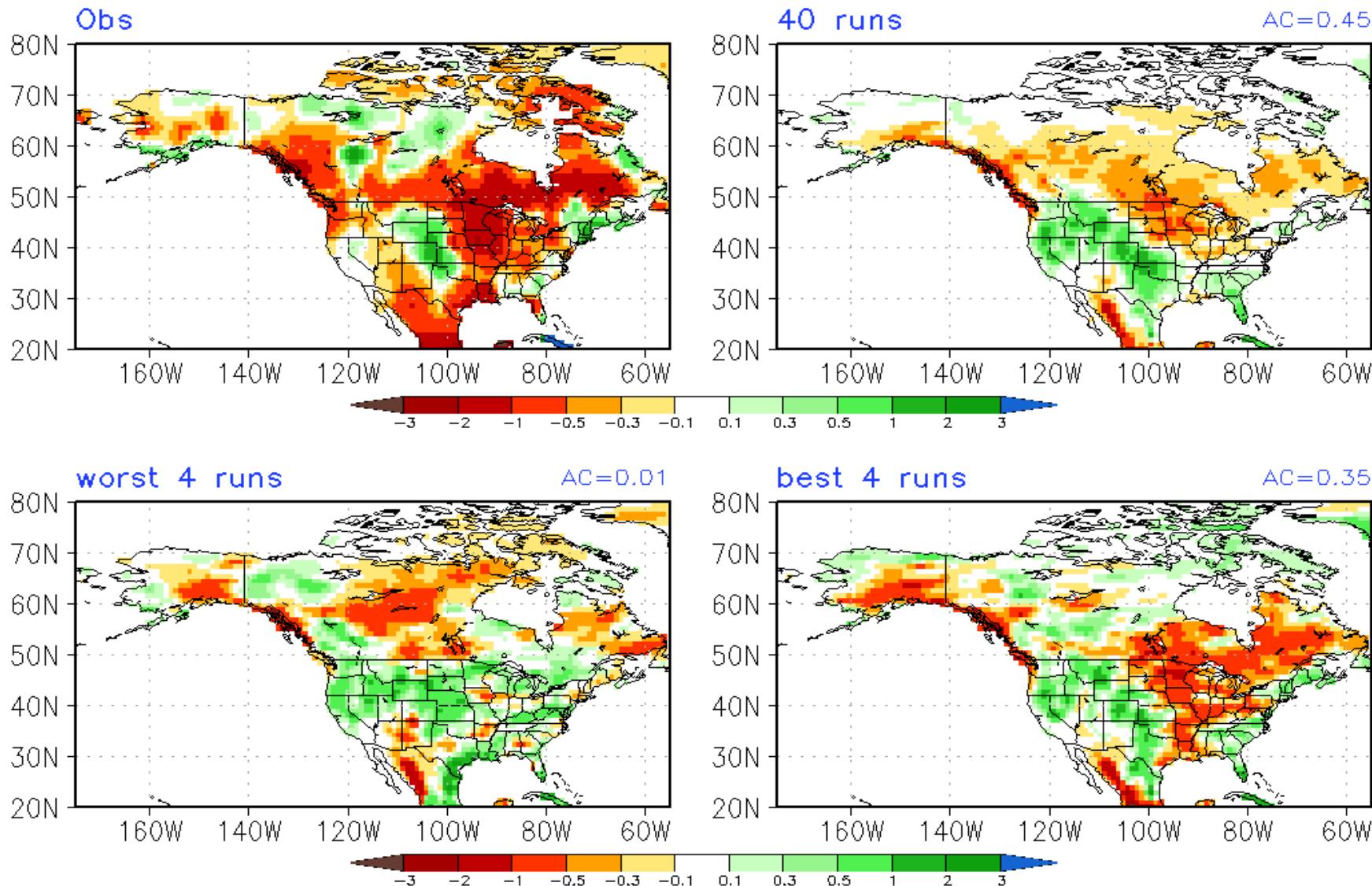
MJJ2023 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- Prec(NA)/SST(30S–30N)



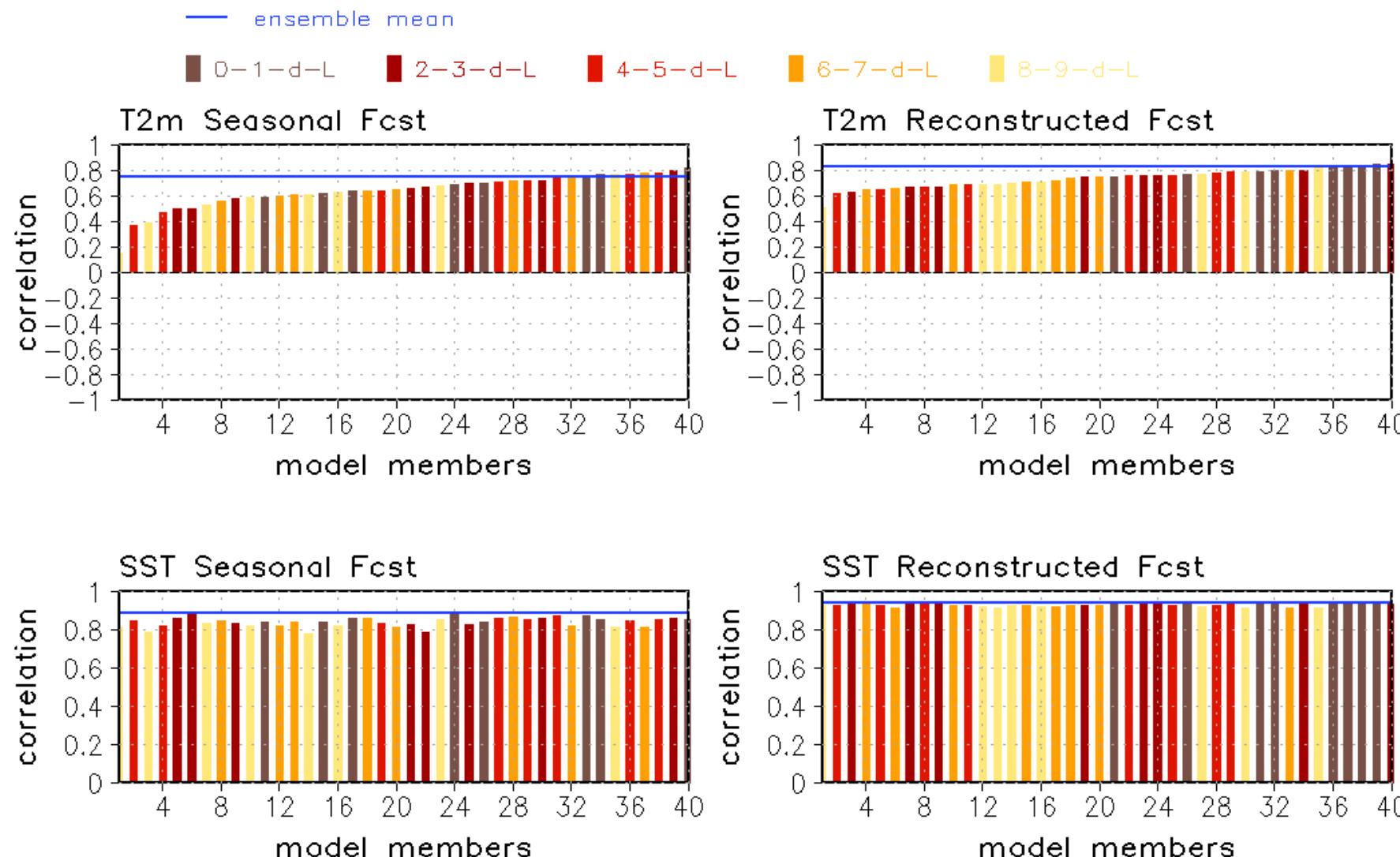
Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs
Seasonal Forecast



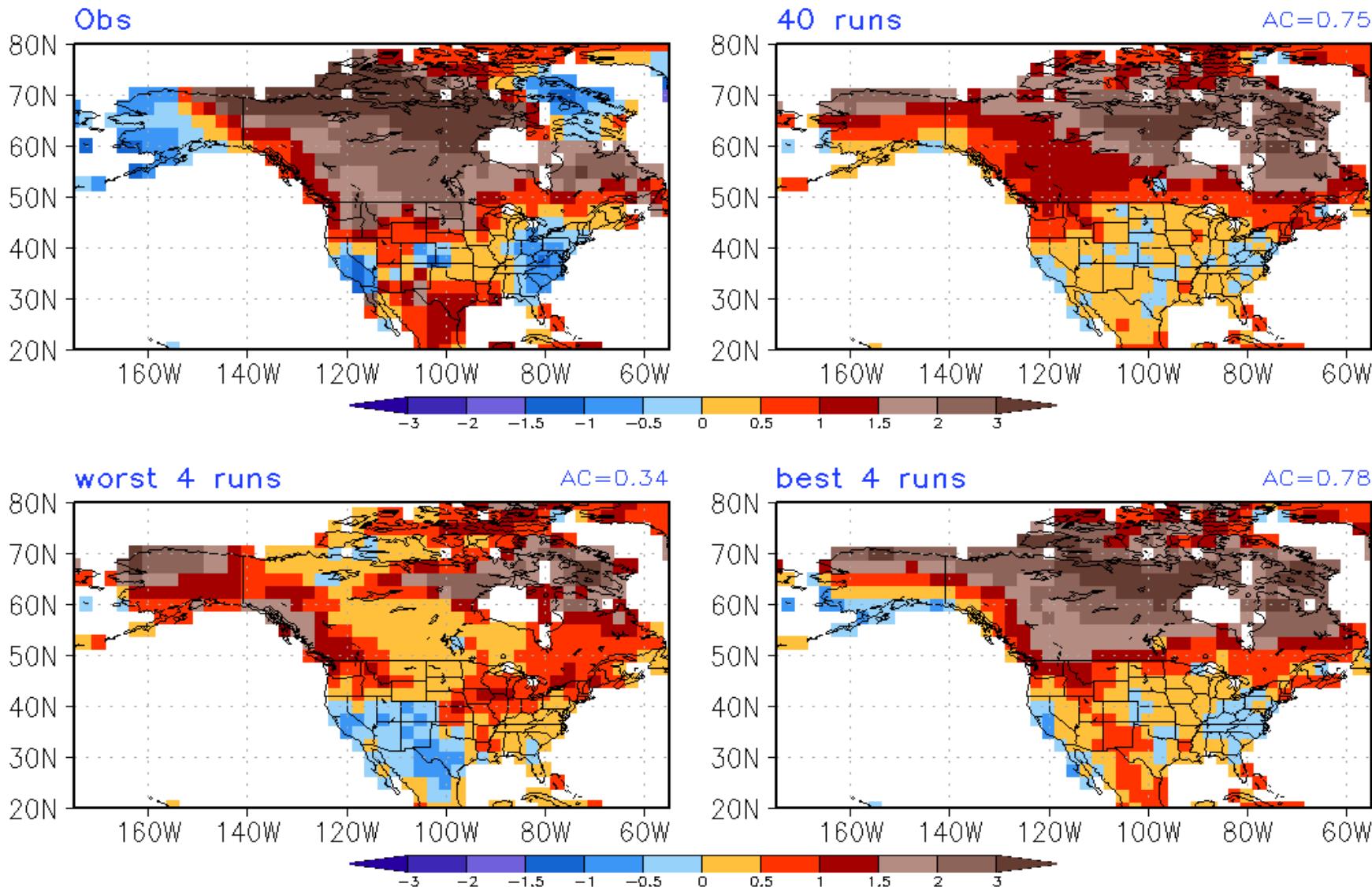
Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs
Reconstructed Forecast



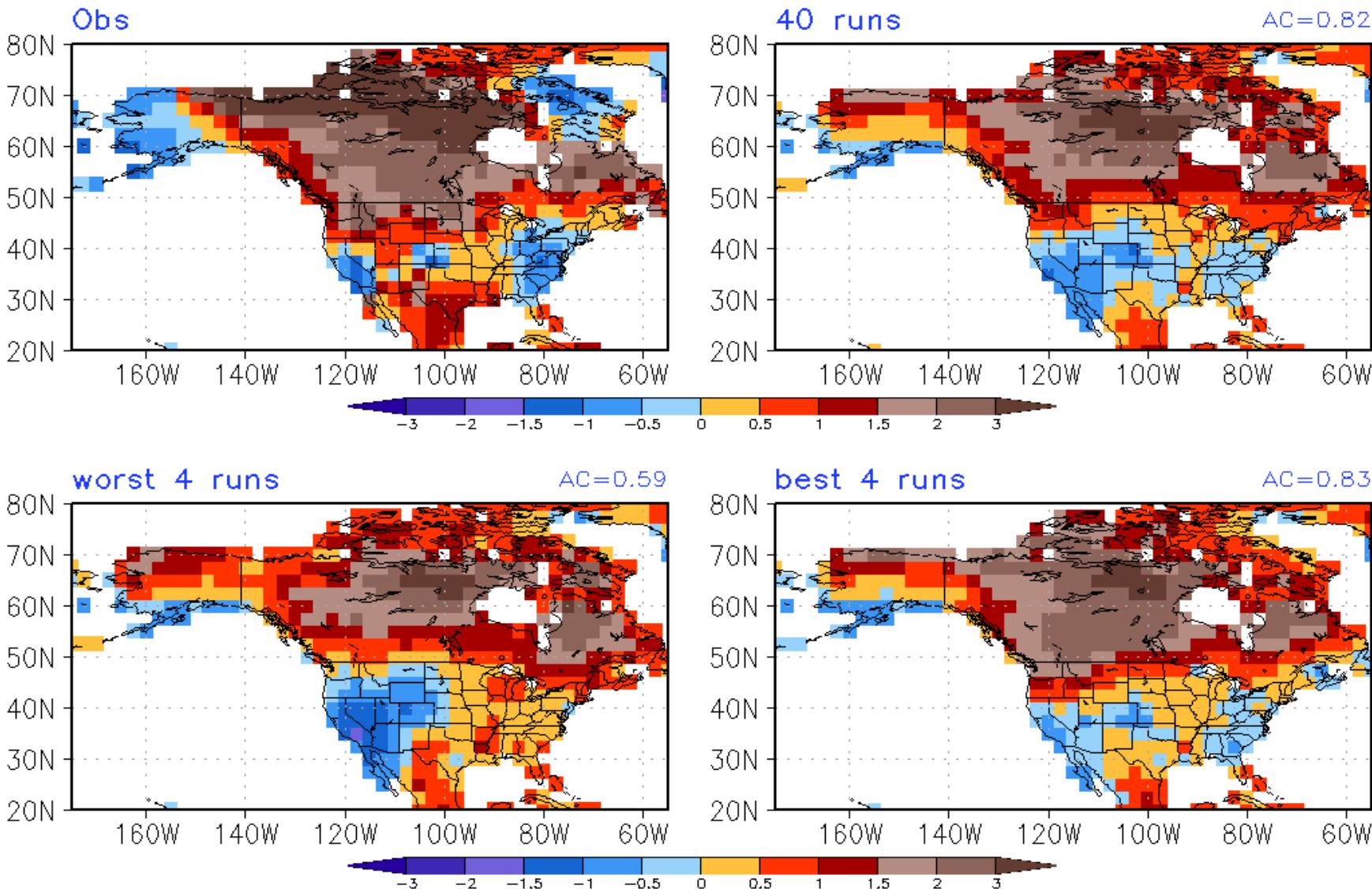
MJJ2023 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- T2m(NA)/SST(30S–30N)



Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 T2m(K) 40 runs/worst 4 runs/best 4 runs
Seasonal Forecast

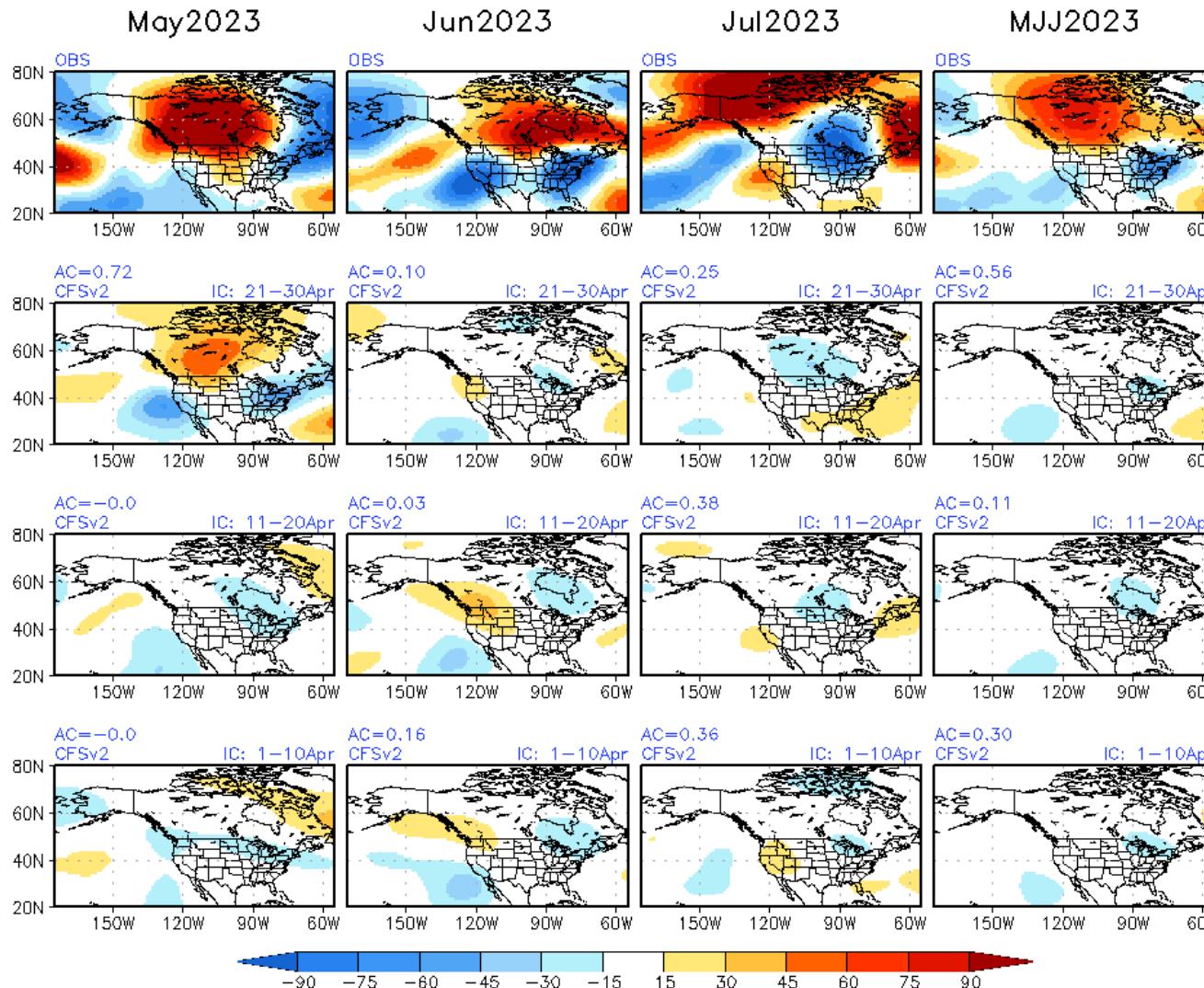


Observed & CFSv2 Forecast Ensemble Average Anomalies
MJJ2023 T2m(K) 40 runs/worst 4 runs/best 4 runs
Reconstructed Forecast



$z200(m)$ Monthly Means from Seasonal Forecast

Monthly Means from Seasonal Fcst (40ensm) MJJ2023 $z200(m)$ eddy & Obs



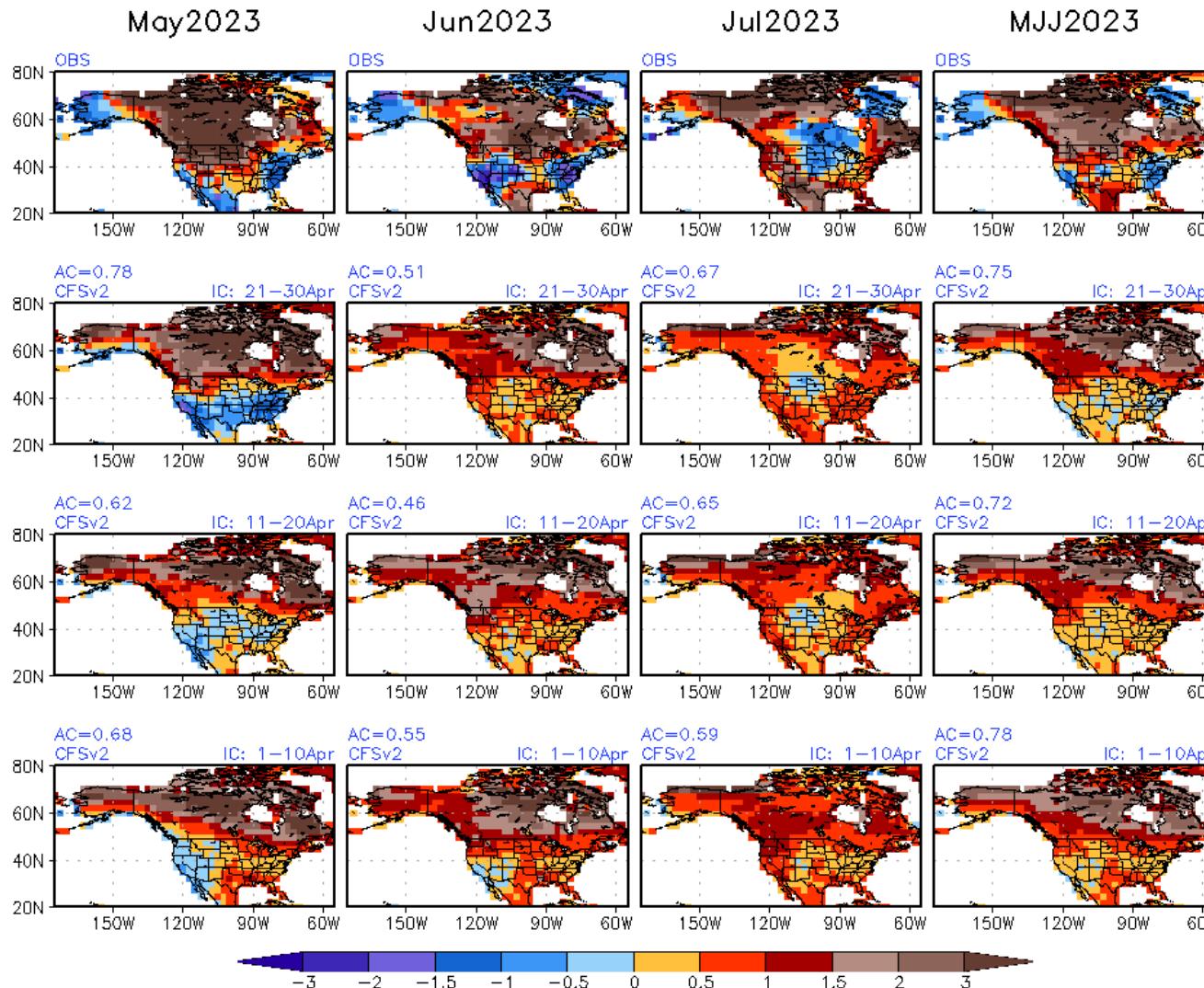
Top row: Observed anomaly.

CFSv2 seasonal forecasts from different initial conditions in the month prior to the target season:

- 2nd row: last 10 days of the prior month.
- 3rd row: 11th - 20th of the prior month.
- 4th row: 1st - 10th of the prior month.

T2m(k) Monthly Means from Seasonal Forecast

Monthly Means from Seasonal Fcst (40ensm) MJJ2023 T2m(K) & Obs



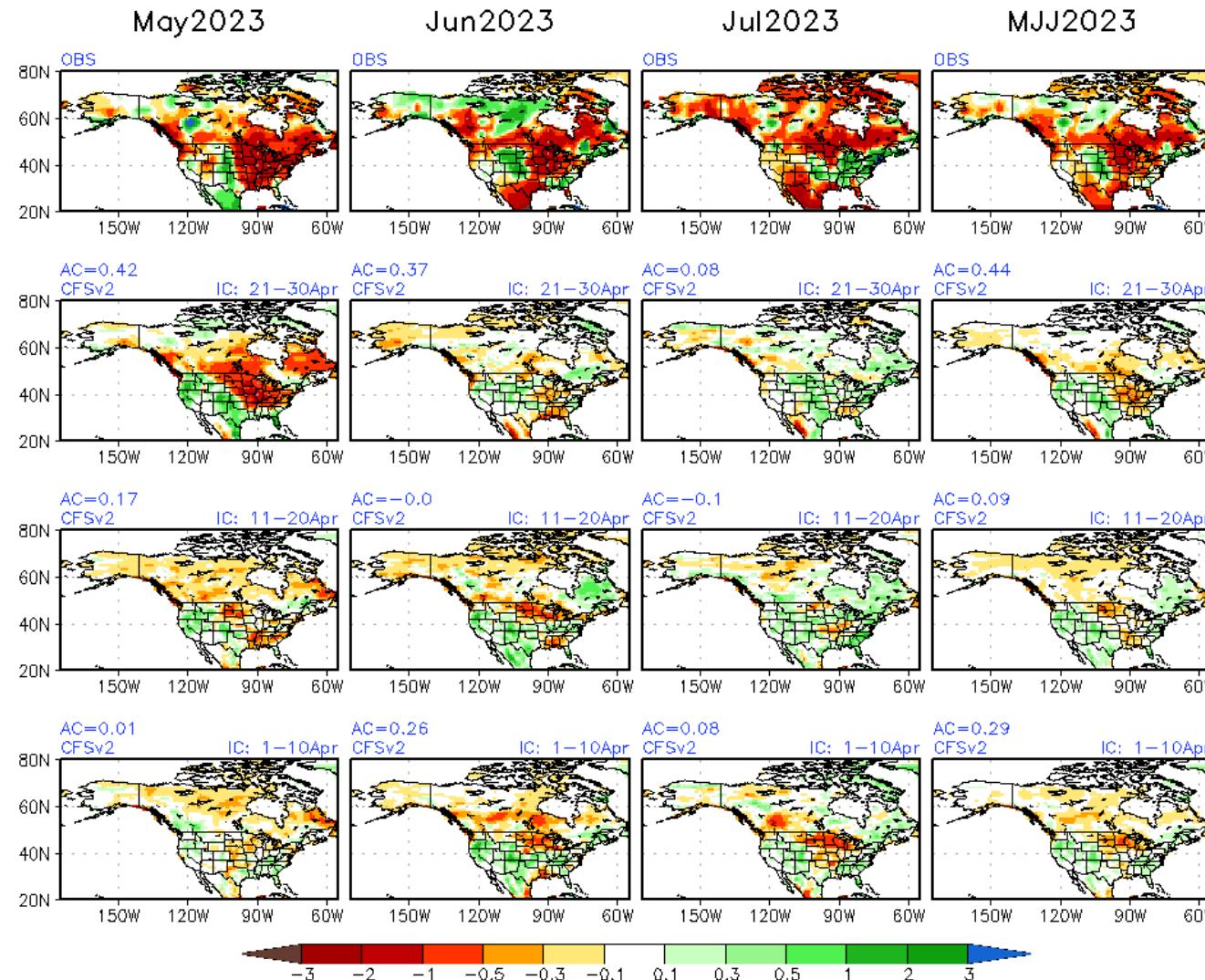
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Prec(mm/day) Monthly Means from Seasonal Forecast

Monthly Means from Seasonal Fcst (40ensm) MJJ2023 Prec(mm/day) & Obs

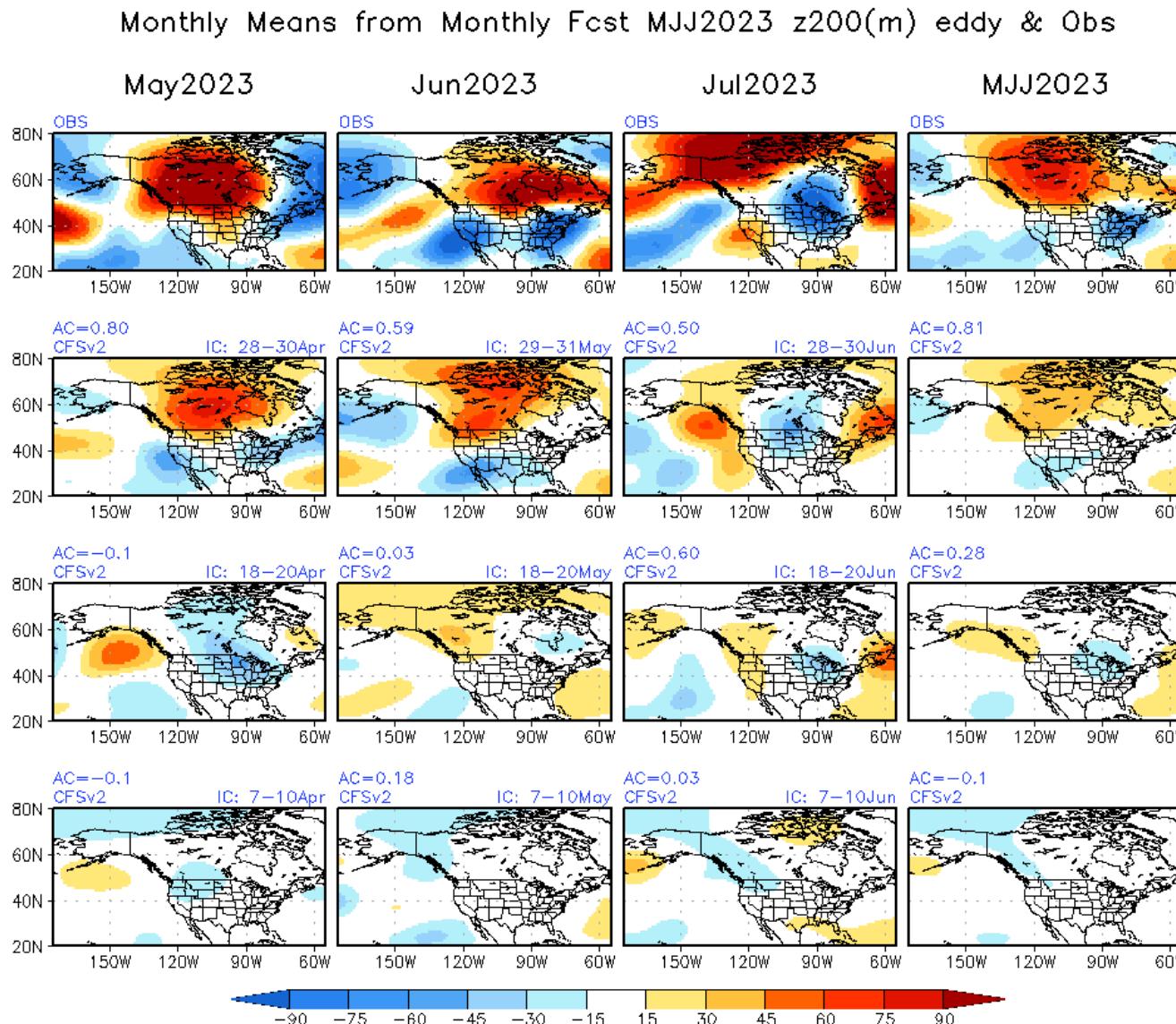


Top row: Observed anomaly.

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$z200(m)$ Monthly Means from Monthly Forecast



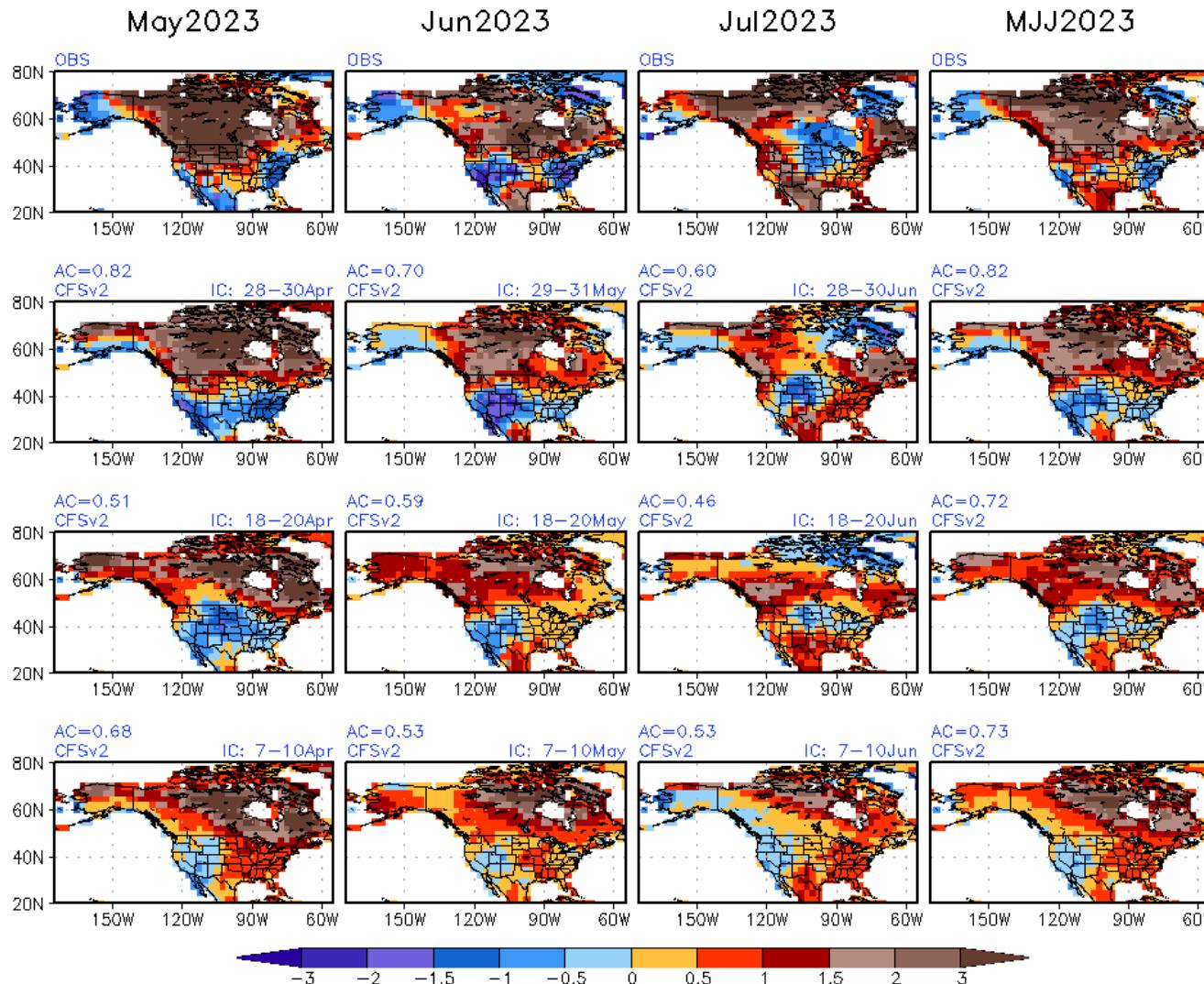
Top row: Observed anomaly.

CFSv2 monthly forecasts from different initial conditions in the month prior to the target month:

- 2nd row: last 3 days of the prior month.
- 3rd row: 18th – 20th of the prior month.
- 4th row: 7th – 10th of the prior month.

T2m(k) Monthly Means from Monthly Forecast

Monthly Means from Monthly Fcst MJJ2023 T2m(K) & Obs

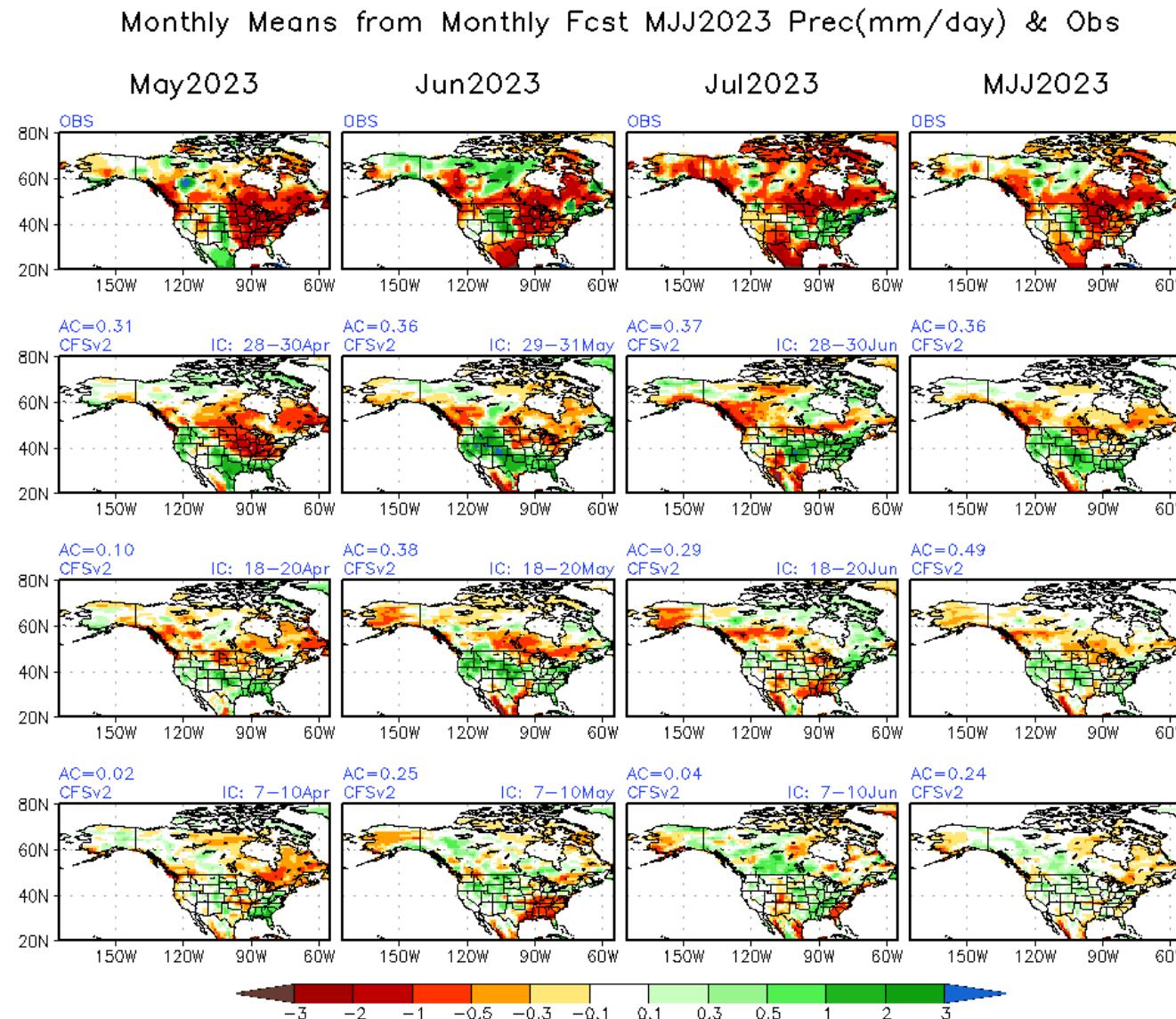


Top row: Observed anomaly.

CFSv2 monthly forecasts from different initial conditions in the month prior to the target month:

- 2nd row: last 3 days of the prior month.
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- 4th row: 7th – 10th of the prior month.

Prec(/mm/day) Monthly Means from Monthly Forecast



Top row: Observed anomaly.

CFSv2 monthly forecasts from different initial conditions in the month prior to the target month:

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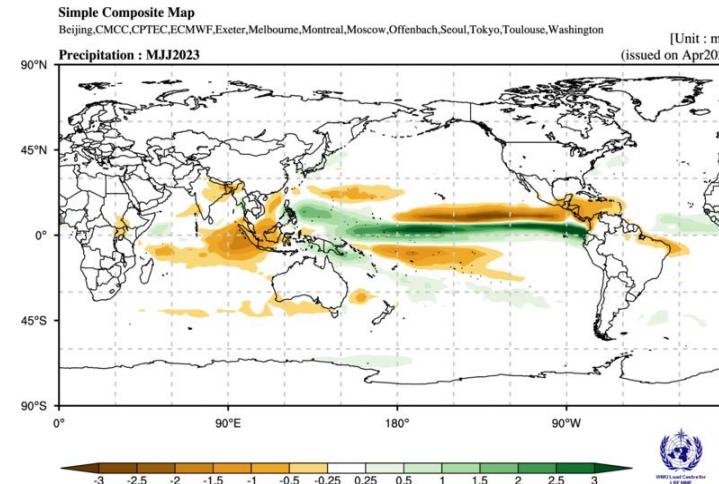
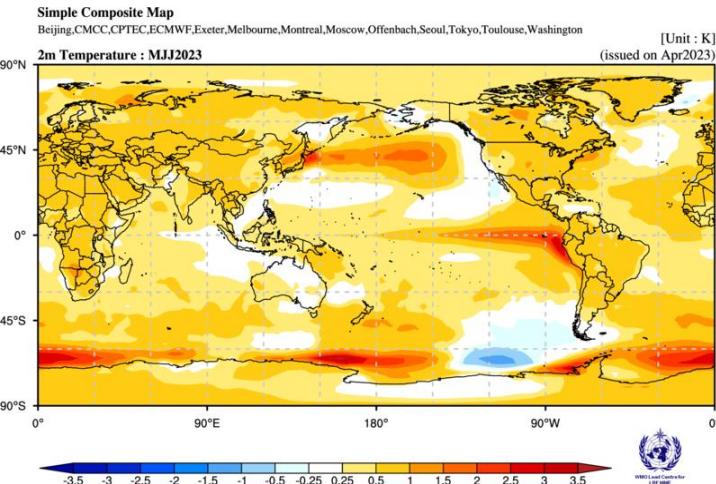
Seasonal Forecasts from Multi-Model Ensemble Systems

- WMO Lead Center for Long-Range Forecast Multi-Model Ensemble (LC-LRFMME).
<https://www.wmclc.org/>
- Copernicus Climate Change Service (C3S) Multi-model seasonal forecasts.
https://climate.copernicus.eu/charts/c3s_seasonal/
- North American Multi-Model Ensemble (NMME) seasonal forecasts.
<https://www.cpc.ncep.noaa.gov/products/NMME/>

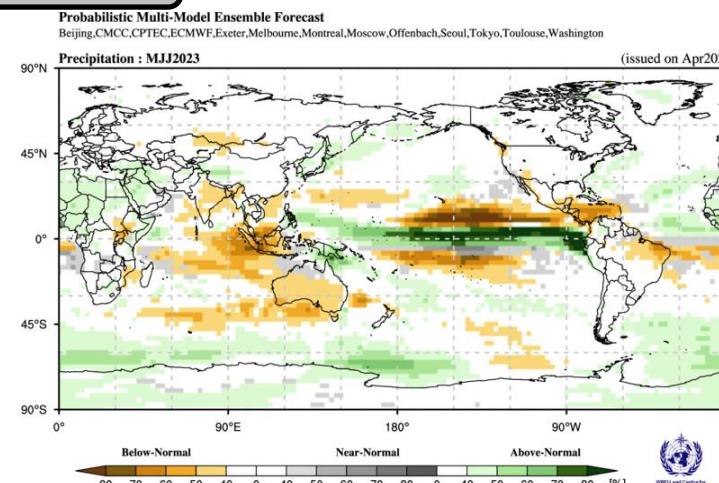
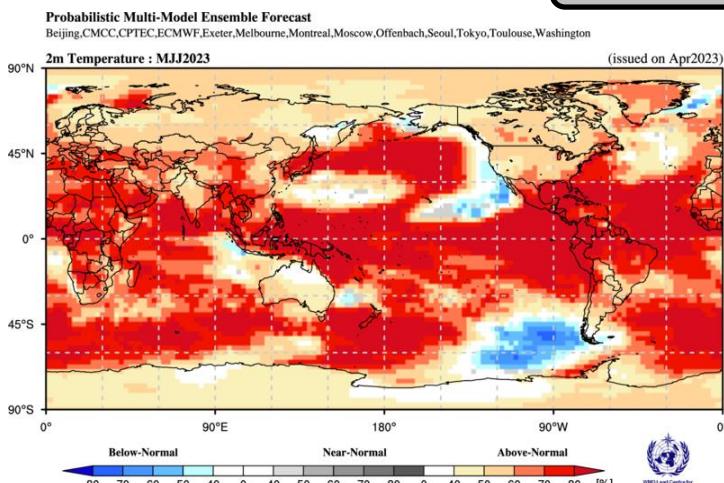
LC-LRFMM Seasonal Forecasts

(<https://www.wmoc.org/>)

Ensemble means

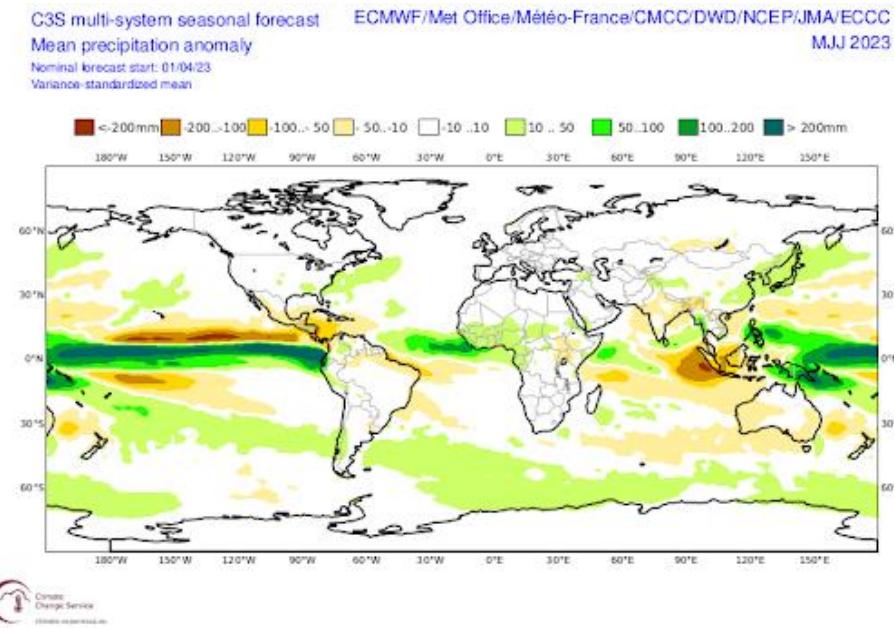
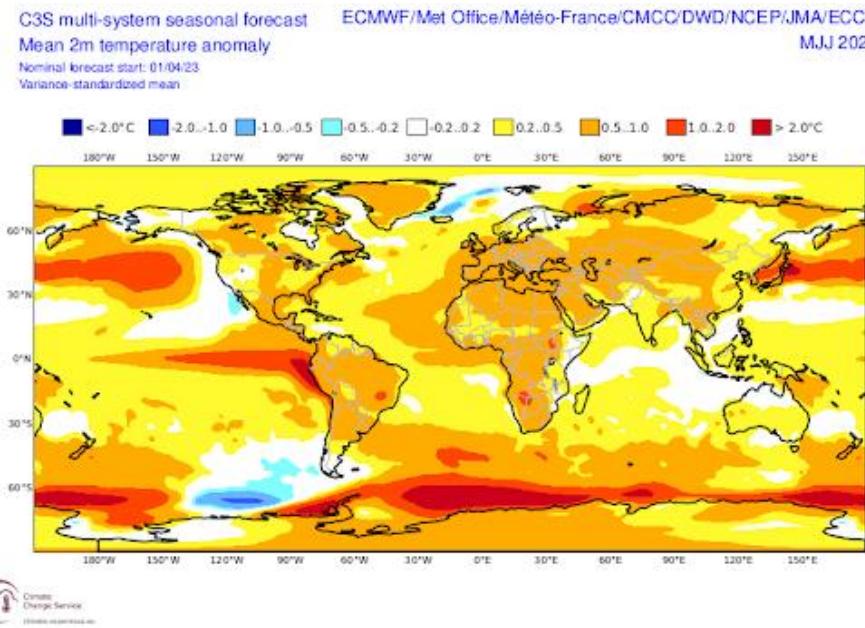


Probabilities



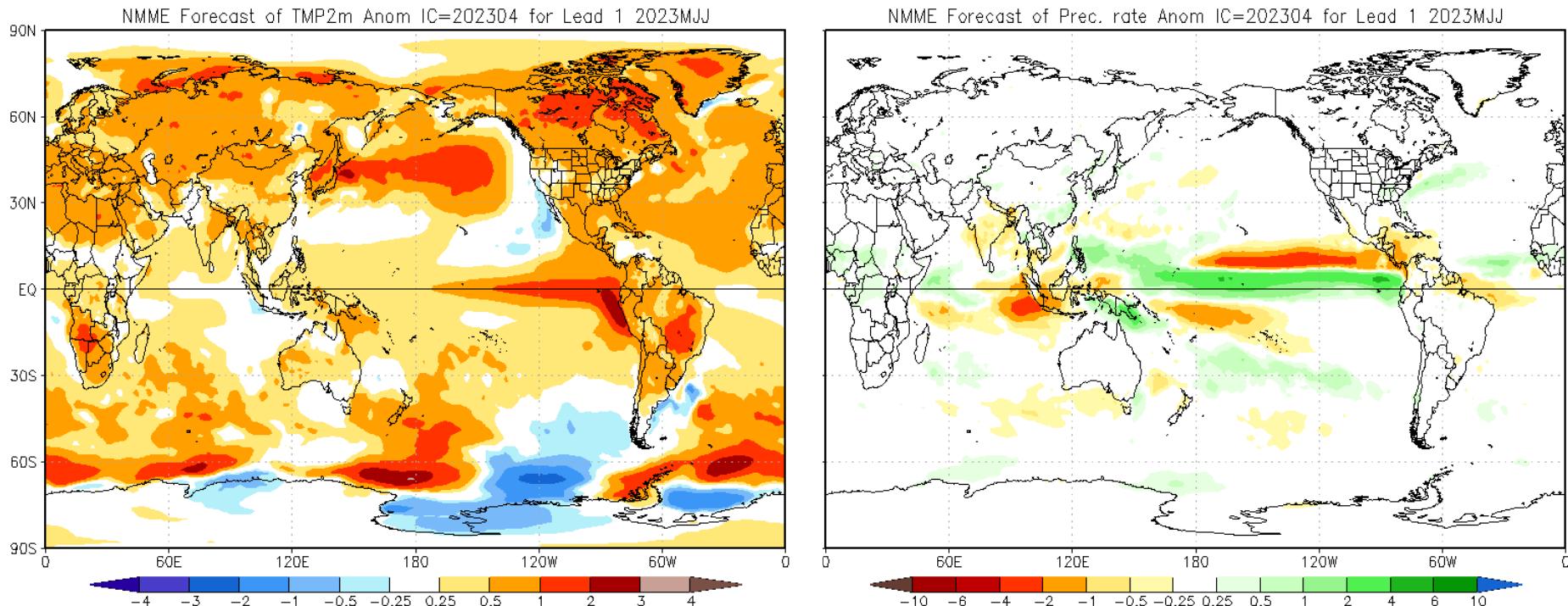
C3S Seasonal Forecast

(https://climate.copernicus.eu/charts/c3s_seasonal/)



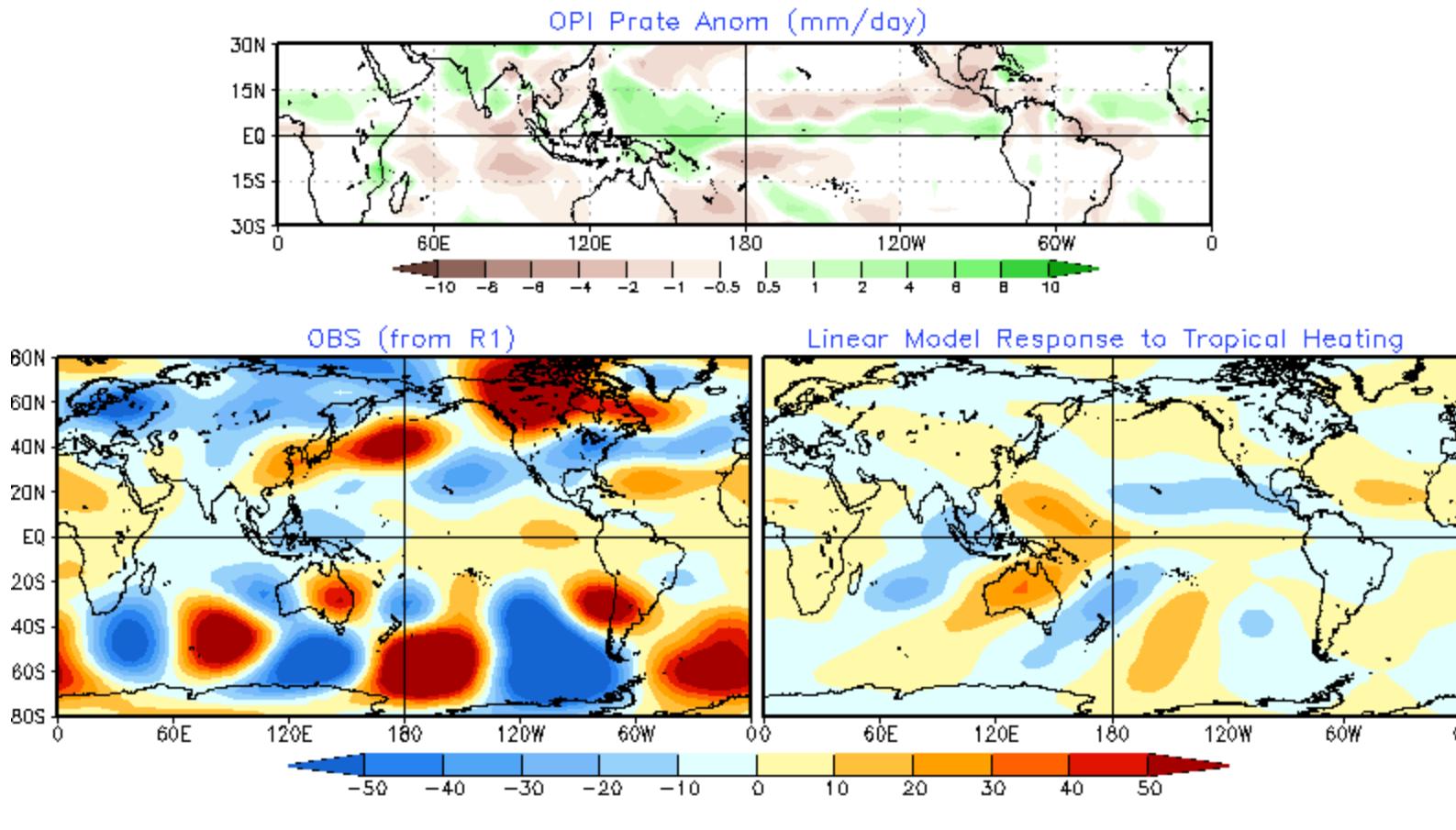
North American Multi-Model Ensemble Seasonal Forecast

(<https://www.cpc.ncep.noaa.gov/products/NMME/>)

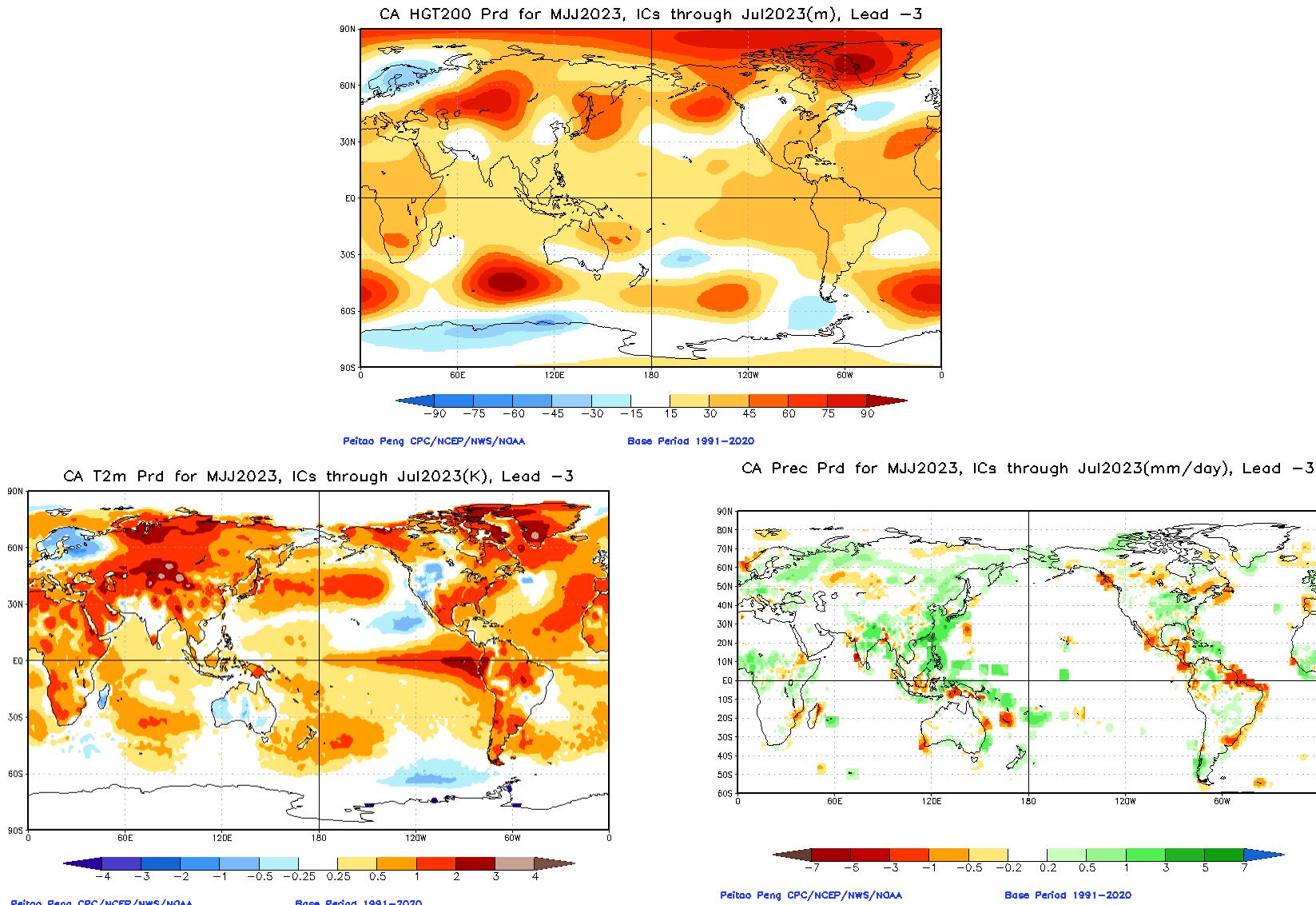


200mb Height from Linear Model

MJJ2023 200mb Eddy HGT(m)
OBS vs. Linear Model Response to Tropical Heating
Heating is converted from Prate in 15S–15N



Seasonal Forecasts from the Constructed Analog Model



Background & Methodology

Attribution of Seasonal Climate Anomalies

- Goal
 - In the context of prediction of seasonal climate variability, utilize seasonal climate forecasts and atmospheric general circulation model (AGCM) simulations to attribute possible causes for the observed seasonal climate anomalies.
 - The analysis can also be considered as an analysis of predictability of the observed seasonal climate anomalies.

Methodology - 1

- Compare observed seasonal mean anomalies with those from model simulations and forecasts.
- Ensemble averaged model simulated/predicted seasonal mean anomalies are an indication of the predictable (or attributable) component of the corresponding observed anomalies.
- For seasonal mean atmospheric anomalies, predictability could be due to
 - Anomalous boundary forcings [e.g., sea surface temperature (SSTs); soil moisture etc.];
 - Atmospheric initial conditions.
- The influence of anomalous boundary forcings (particularly due to SSTs, can be inferred from the ensemble mean of AGCM simulations forced by observed SSTs, the so called AMIP simulations). This component of predictability (or attributability) is more relevant for longer lead seasonal forecasts.

Methodology - 2

- The influence of the atmospheric initial state can be inferred from initialized predictions. This component is more relevant for short lead seasonal forecasts.
- The influence of unpredictable component in the atmospheric variability can be assessed from the analysis of individual model simulations, and the extent anomalies in individual runs deviate from the ensemble mean anomalies.
- The relative amplitude of ensemble averaged seasonal mean anomalies to the deviations of seasonal mean anomalies in the individual model runs from the ensemble average is a measure of seasonal predictability (or the extent observed anomalies are attributable).
- Observed anomalies are equivalent to a realization of a single model run, and therefore, analysis of individual model runs also gives an appreciation of how much observed anomalies can deviate from the component that is attributable (Kumar et al. 2013).

Data

- Observations
 - SST: OI version 2 analysis (Reynolds et al., 2007)
 - Prec: CMAP monthly analysis (Xie and Arkin, 1997)
 - T2m: GHCN-CAMS land surface temperature monthly analysis (Fan and van den Dool, 2008)
 - 200mb height (z200): CFSR (Saha et al., 2010)
- 0-month-lead seasonal mean forecasts from CFSv2 (Saha et al. 2014)
 - Seasonal forecast: the seasonal mean forecasts based on 40 members from the latest 10 days before the target season (0-month-lead);
 - Reconstructed forecast: the seasonal mean forecasts constructed from 3 individual monthly forecasts with the latest 10 days initial conditions for each individual monthly forecasts. This approach for constructing seasonal mean anomalies has more influence from the initial conditions (Kumar et al. 2013);
- Seasonal mean AMIP simulation based on GFS_FV3 (provided by Dr. Tao Zhang/CPC)
 - 100 members
- All above seasonal mean anomalies are based on 1991-2020 climatology.
- z200 responses to tropical heating in linear model (provided by Dr. Peitao Peng/CPC)
- Seasonal mean anomalies of z200, T2m, and Prec forecasted from the Constructed Analog Model (provided by Dr. Peitao Peng/CPC)