

Attribution of Seasonal Climate Anomalies April-May-June 2022

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

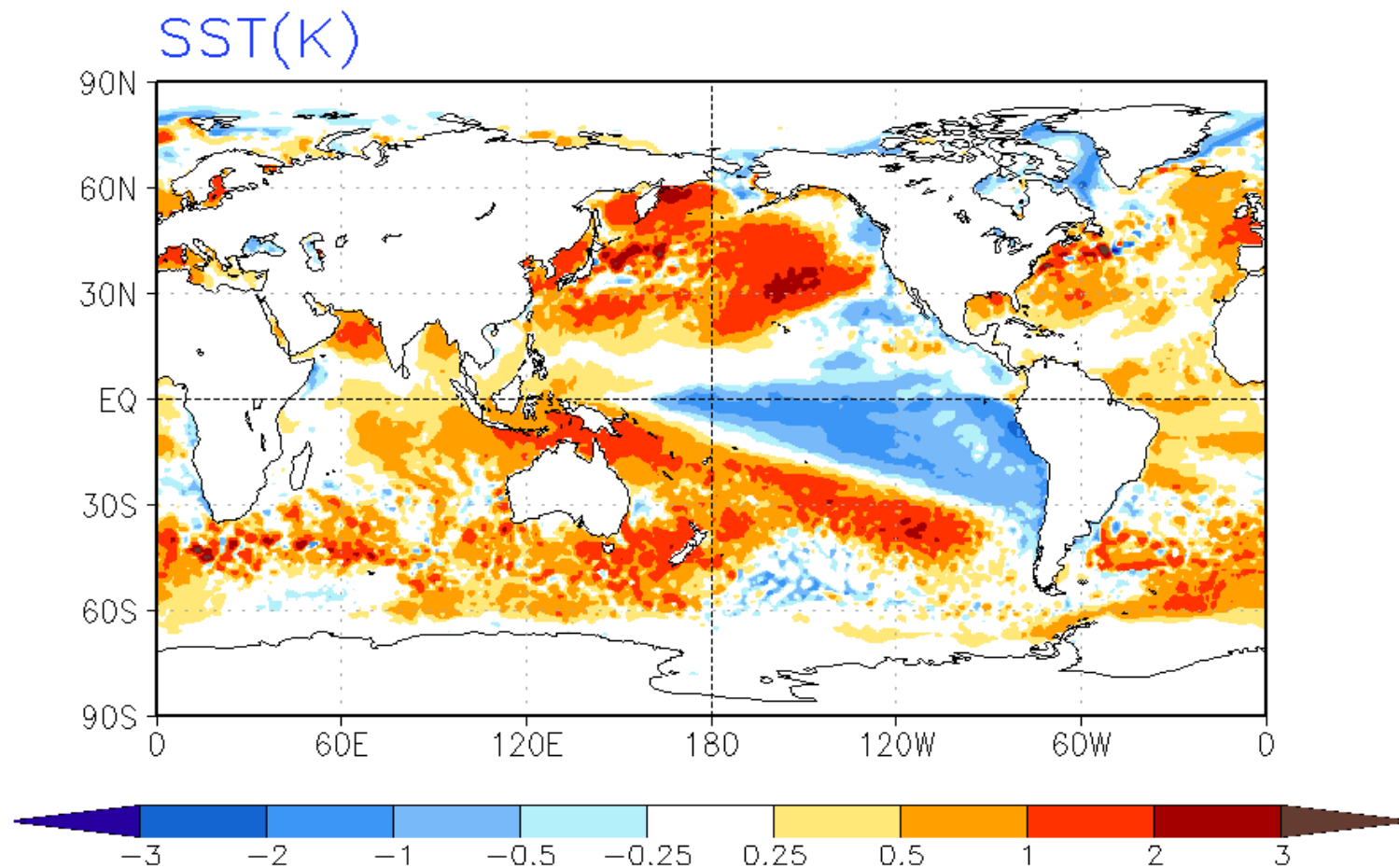
Summary of Observed Conditions and Outlooks

- Tropical SST anomalies continued in La Nina conditions; the equatorial Atlantic and North Pacific SST anomalies remained on the warm side (slide 4). In general, the large-scale distribution of SST anomalies was predicted well (slide 10);
- Large-scale distribution of above normal precipitation anomalies in the equatorial eastern Indian Ocean, Maritime Continent and dry conditions in the equatorial western, central Pacific Ocean ([a reflection of La Niña conditions](#)) were predicted well in the initialized CFSv2 and other MME models (slides 37-39).
- Initialized CFSv2 forecasts predicted the large-scale distribution of observed 200mb height anomalies over tropical and sub-tropical areas (slide 12), while they misplaced the location and overestimated the amplitude of the observed height positive anomalies over the north-eastern Canada, leading to erroneous prediction of much above normal warm temperature anomalies (slide 12, 13, 15, 16).
- CFSv2 and MME predictions of AMJ2022 North American precipitation and temperature anomalies were consistent with the AMIP simulations (response to SSTs) (slides 7, 14, 16).
- For increasing lead-time, the NA monthly anomalies from seasonal forecasts rapidly converged to the pattern that is consistent with longer lead seasonal mean forecasts (Slide 30-32).
- For June 2022, monthly mean forecasts from the shortest leads didn't improve the prediction skill for the NA precipitation and temperature (Slide 34,35).

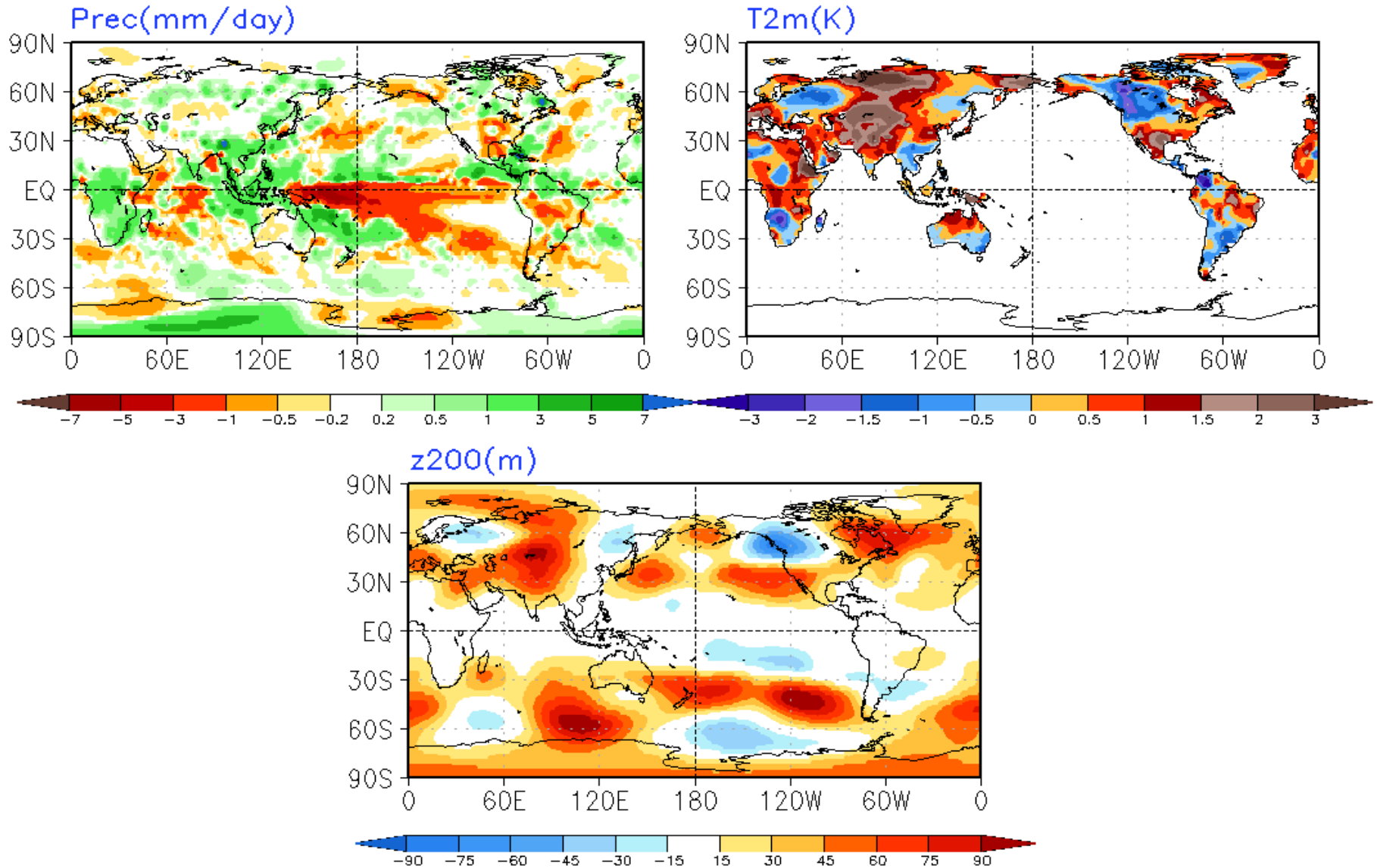
Observed Seasonal Anomalies

Global and North America

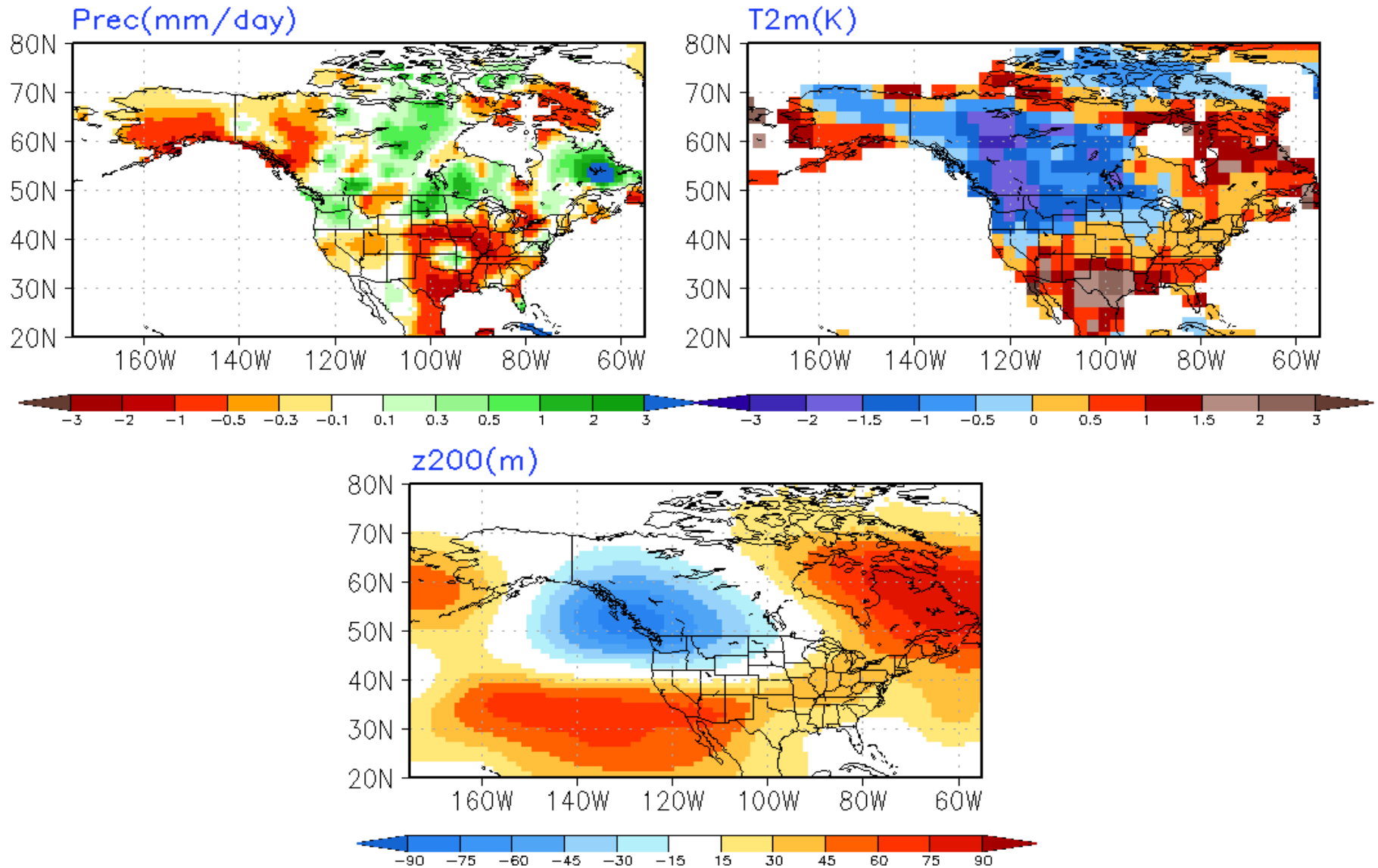
Observed Anomaly AMJ2022



Observed Anomaly AMJ2022



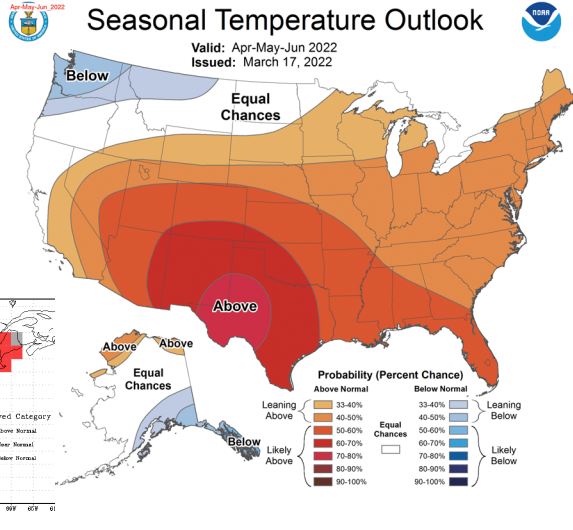
Observed Anomaly AMJ2022



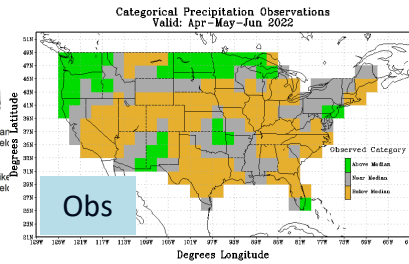
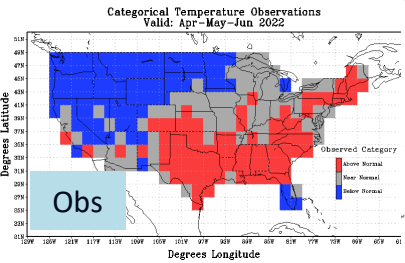
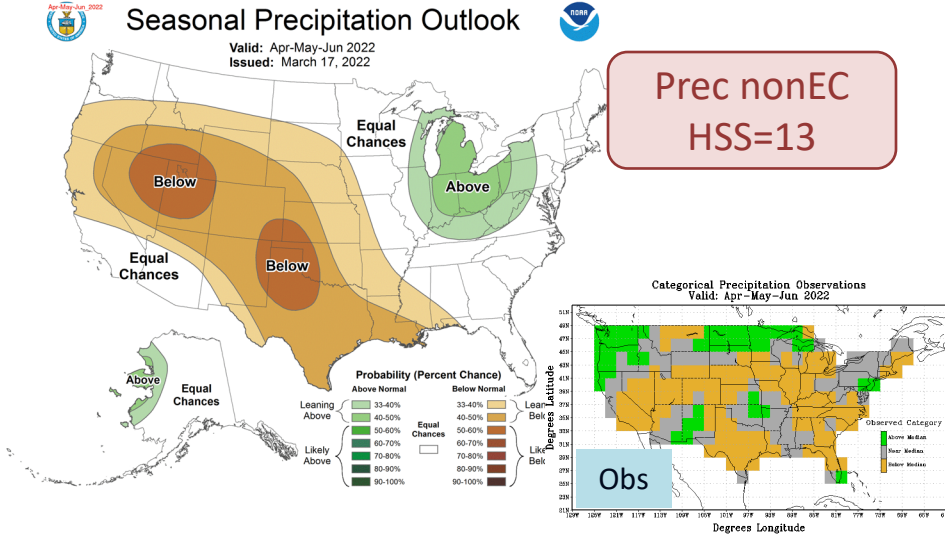
CPC Seasonal Outlooks and NMME Forecasts

CPC

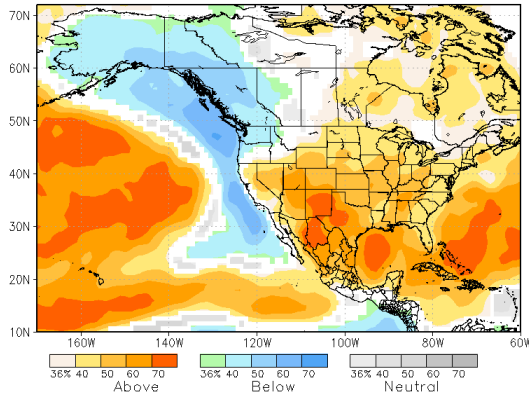
Temp nonEC
HSS=17



Prec nonEC
HSS=13

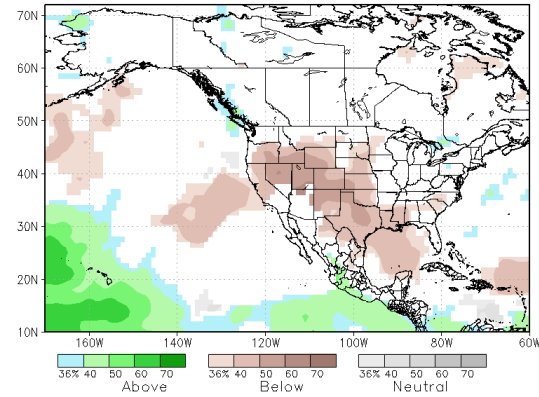


NMME prob fcst TMP2m IC=202203 for lead 1 2022 AMJ



NMME

NMME prob fcst Prate IC=202203 for lead 1 2022 AMJ



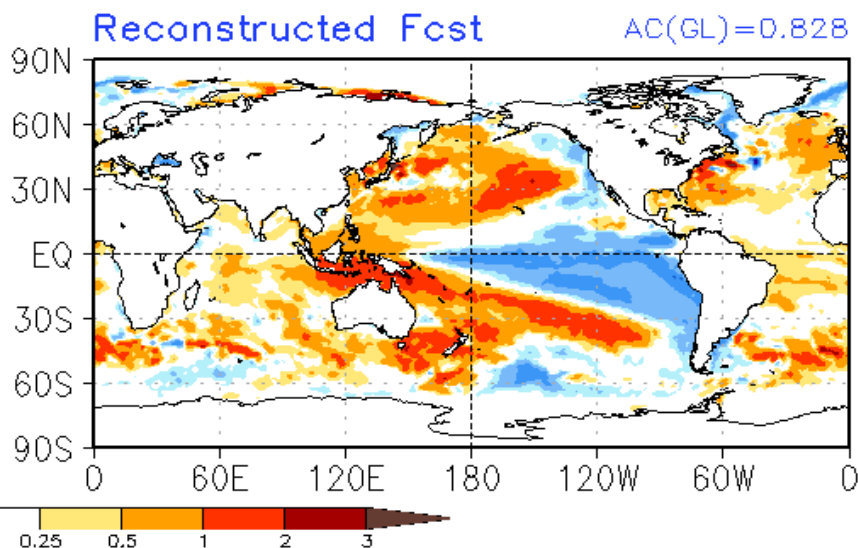
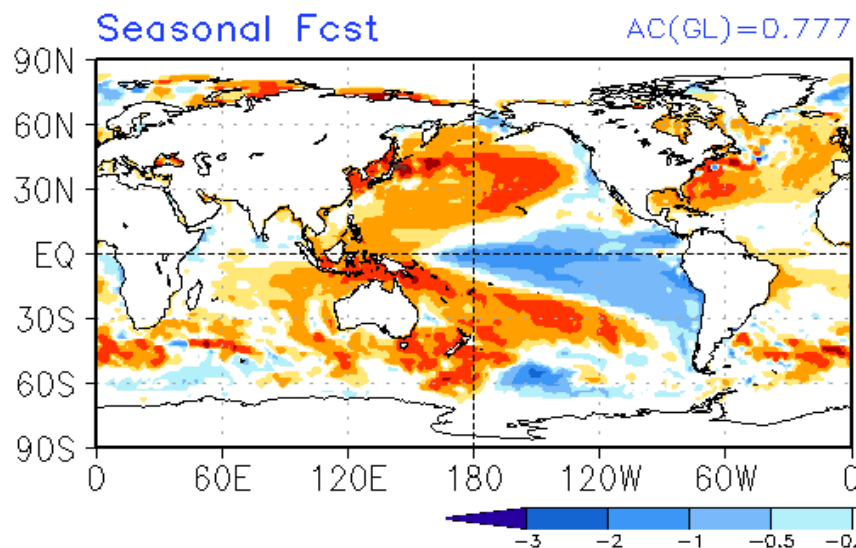
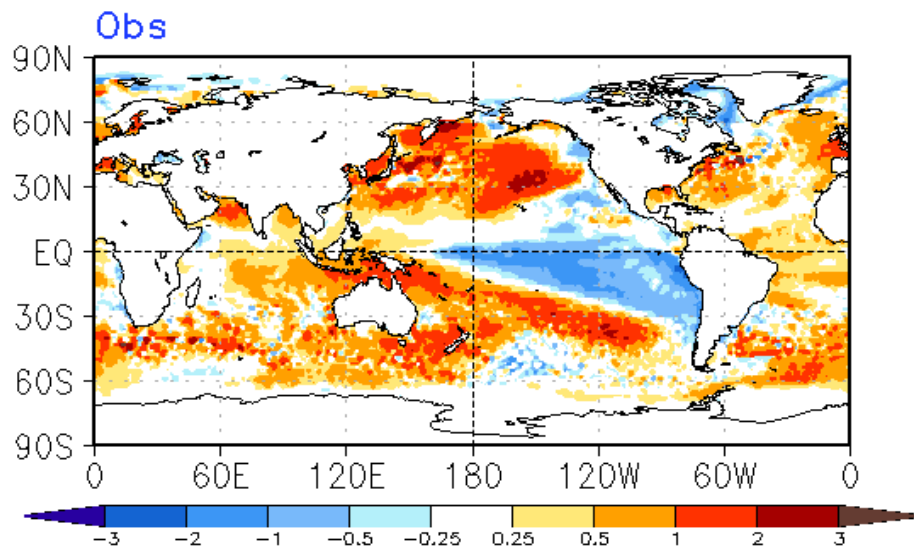
For the rationale behind CPC outlooks see: https://www.cpc.ncep.noaa.gov/products/archives/long_lead/PMD/2022/202203_PMD90D

Model Simulated/Forecast Ensemble Mean Anomalies

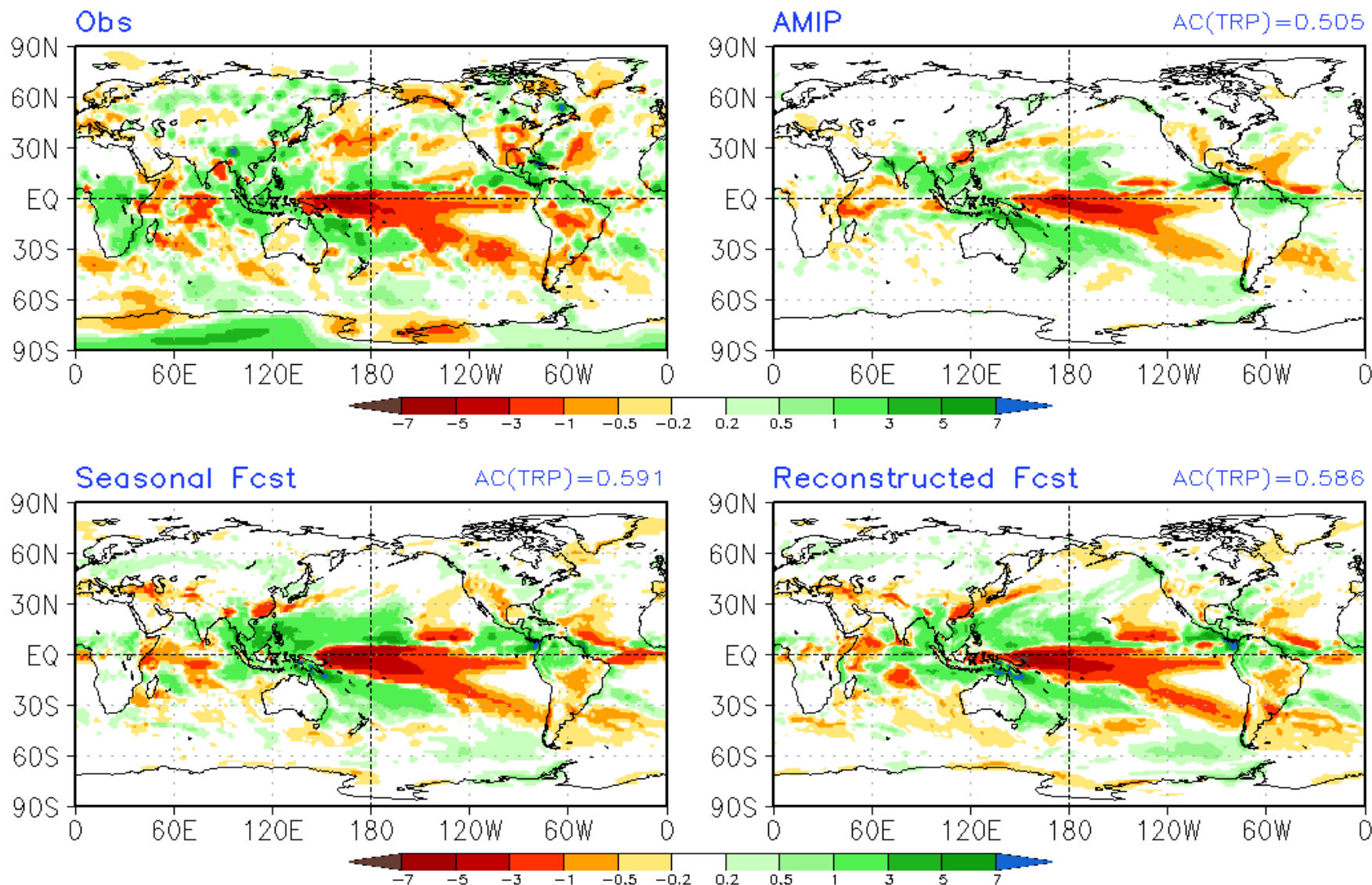
Model Simulated/Forecast Ensemble Average Anomalies

- **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).

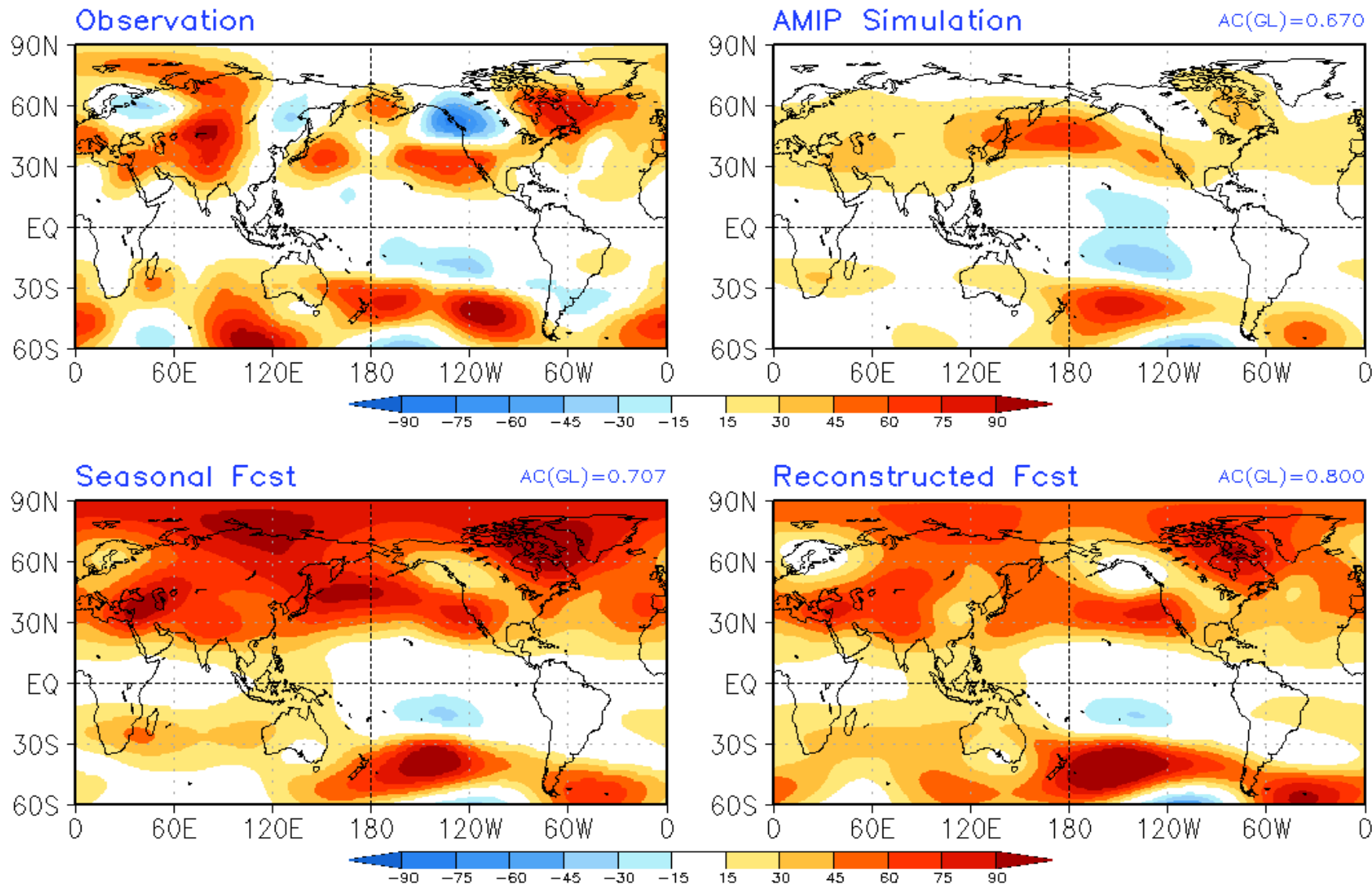
AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies SST(K)



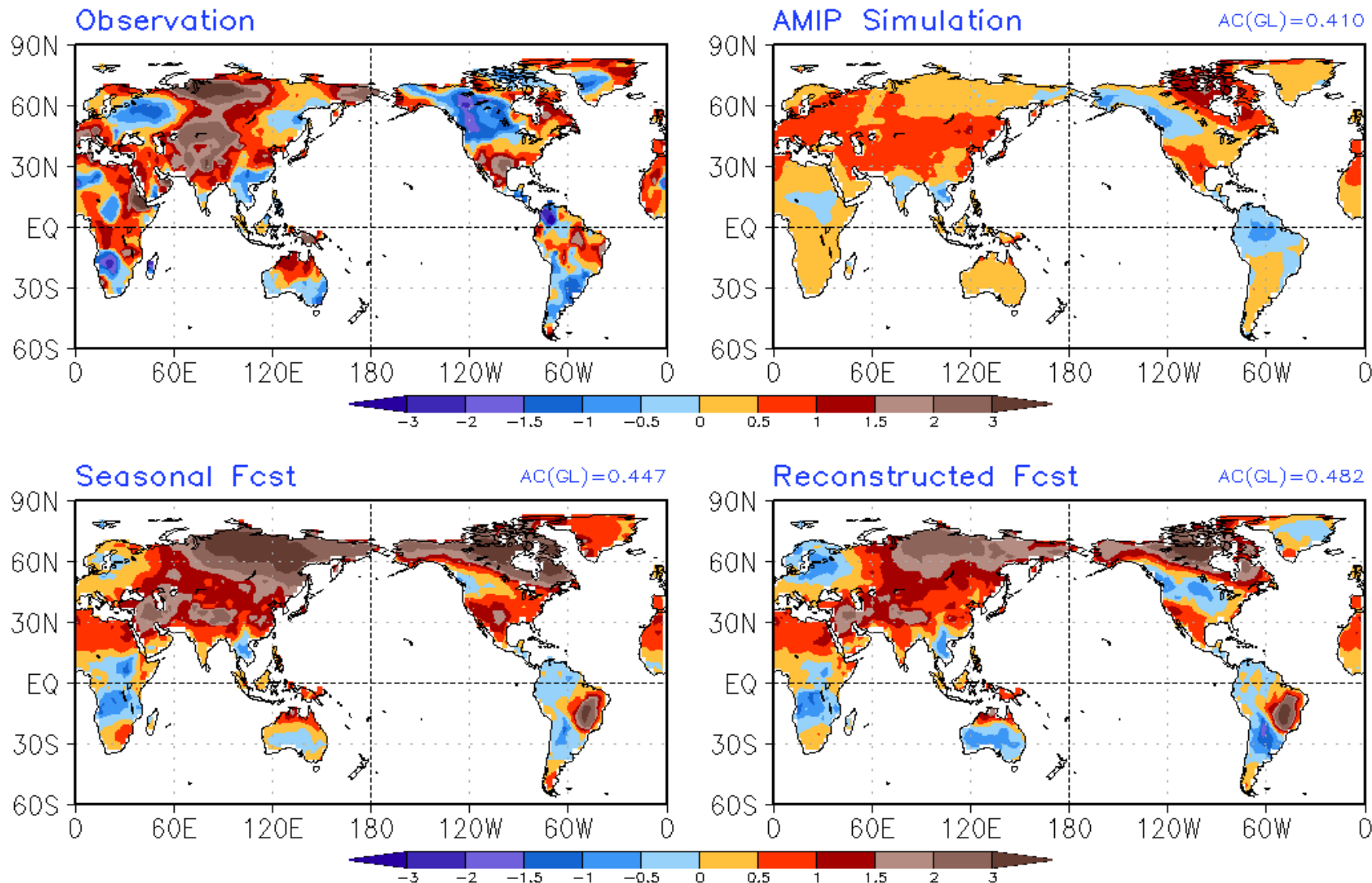
AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



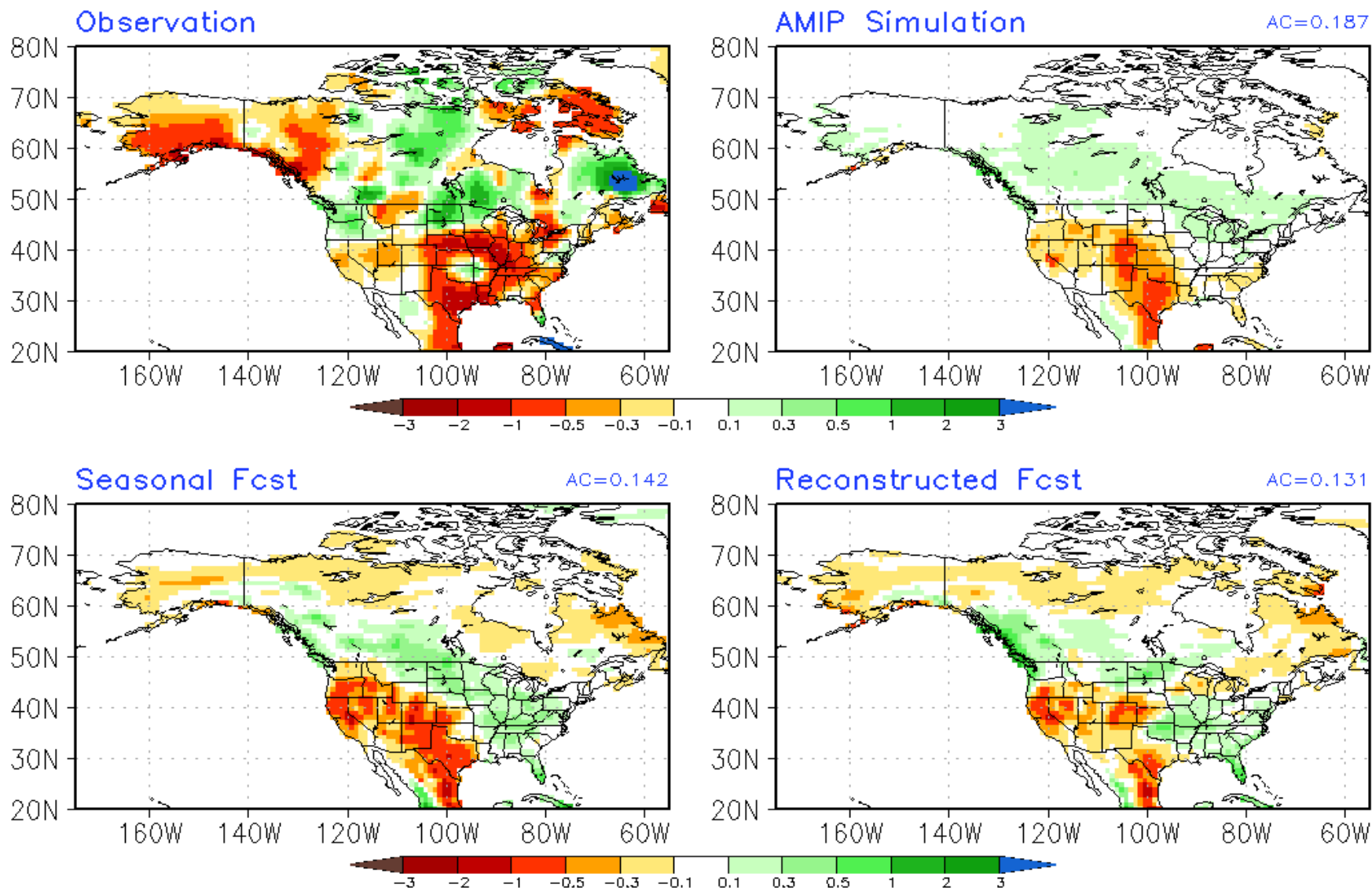
AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



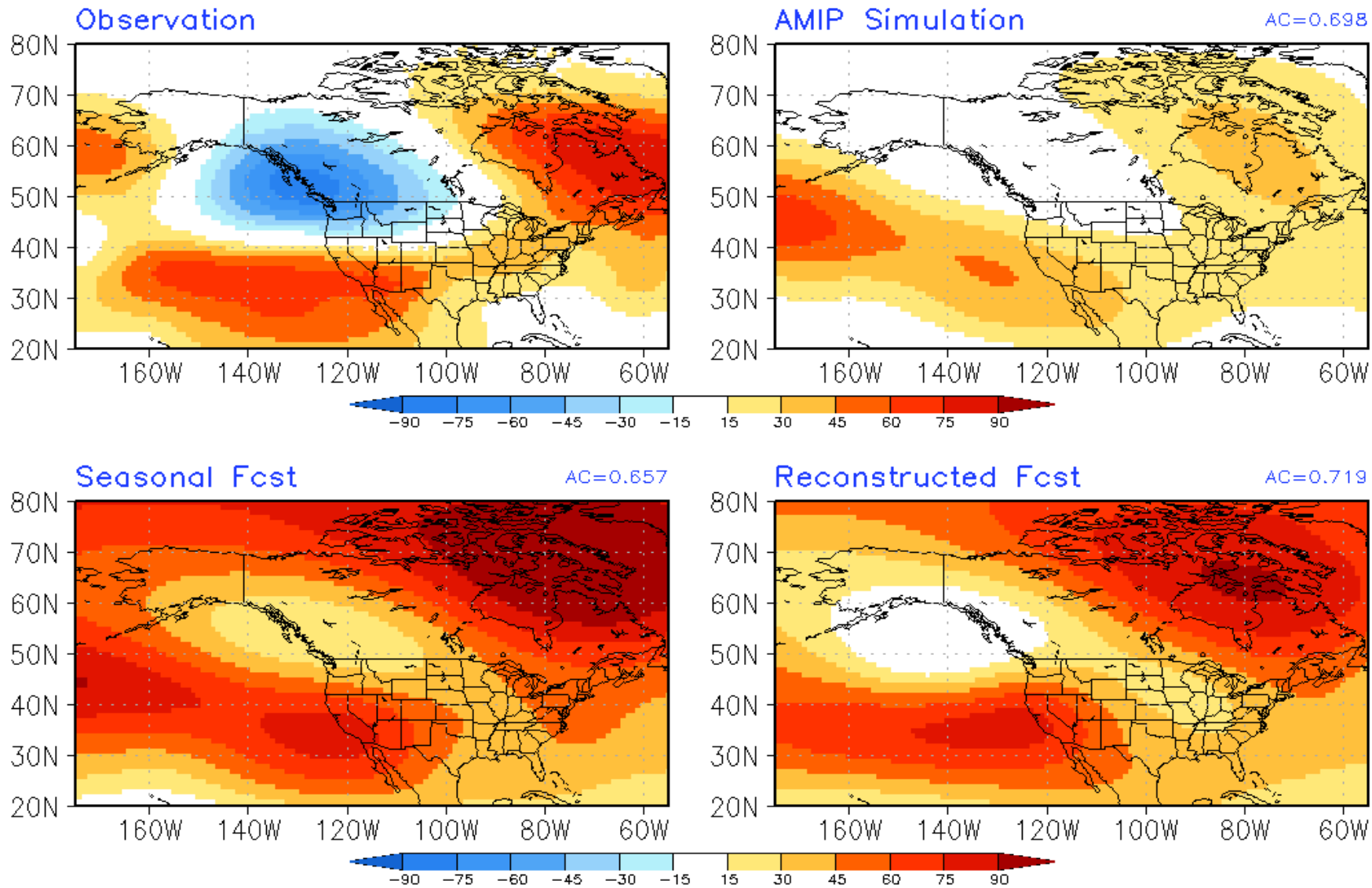
AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



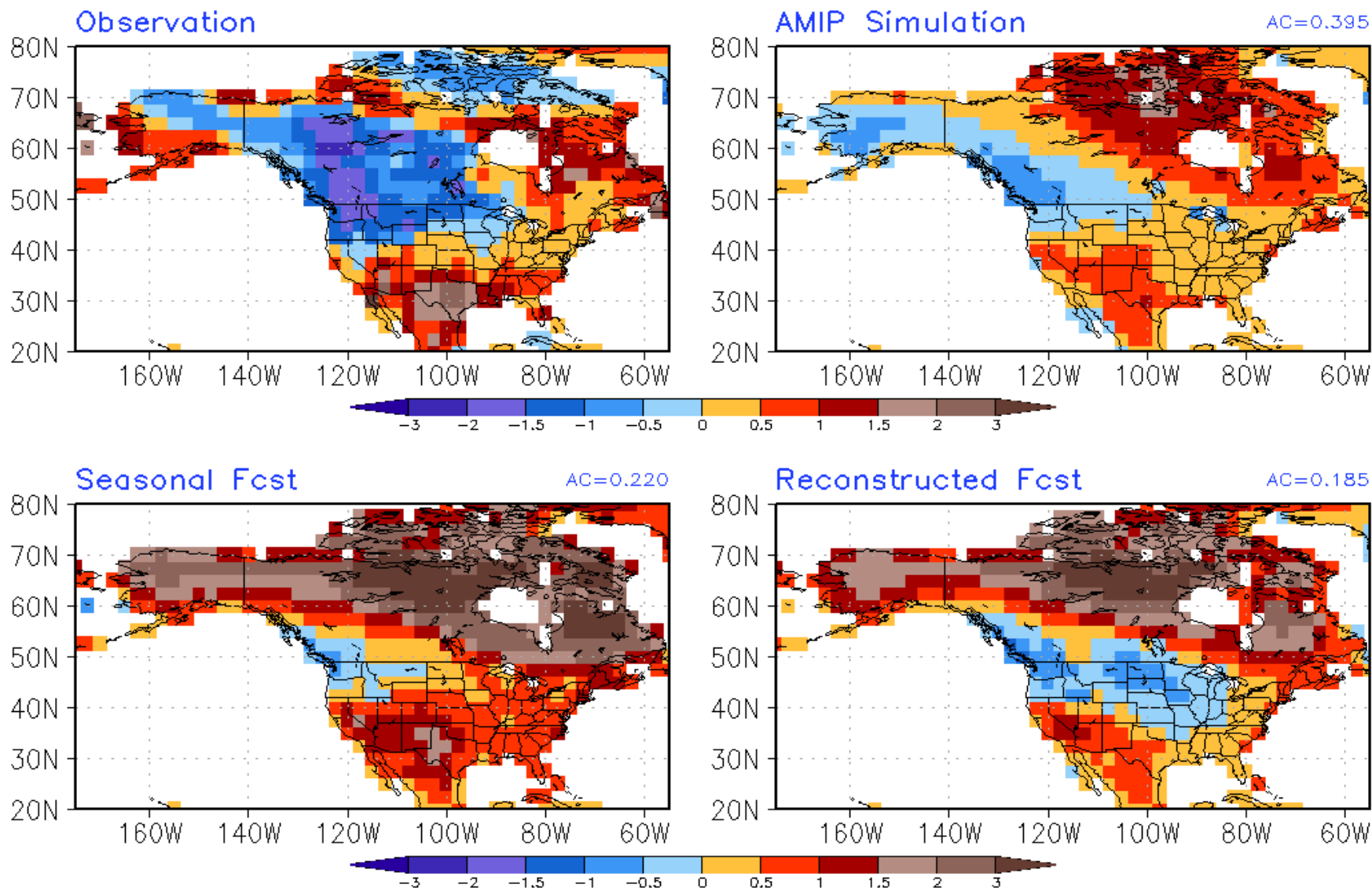
AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



AMJ2022 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)

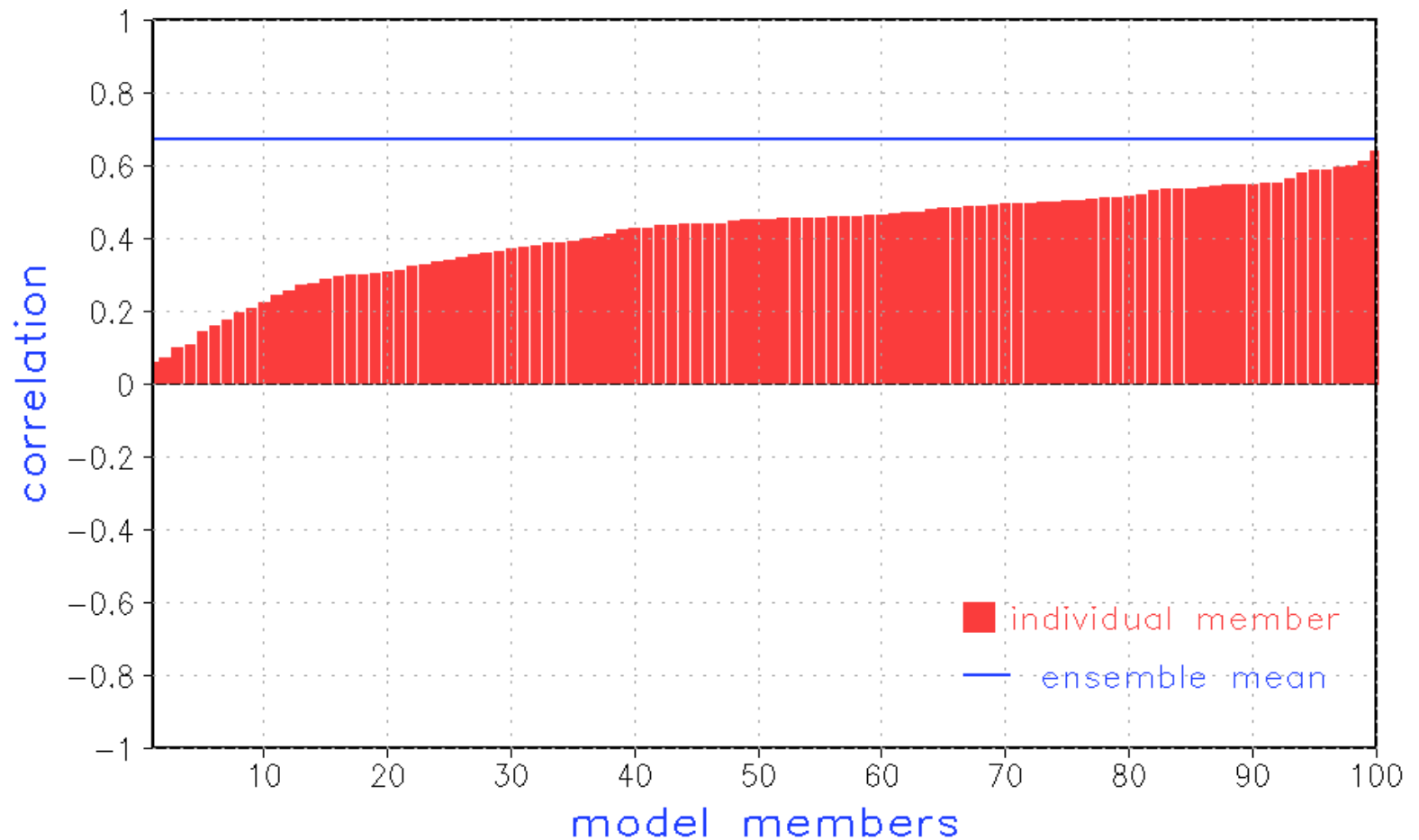


Model Simulated/Forecast Anomalies: Individual Runs

Model Simulated/Forecast Anomalies: Individual Runs

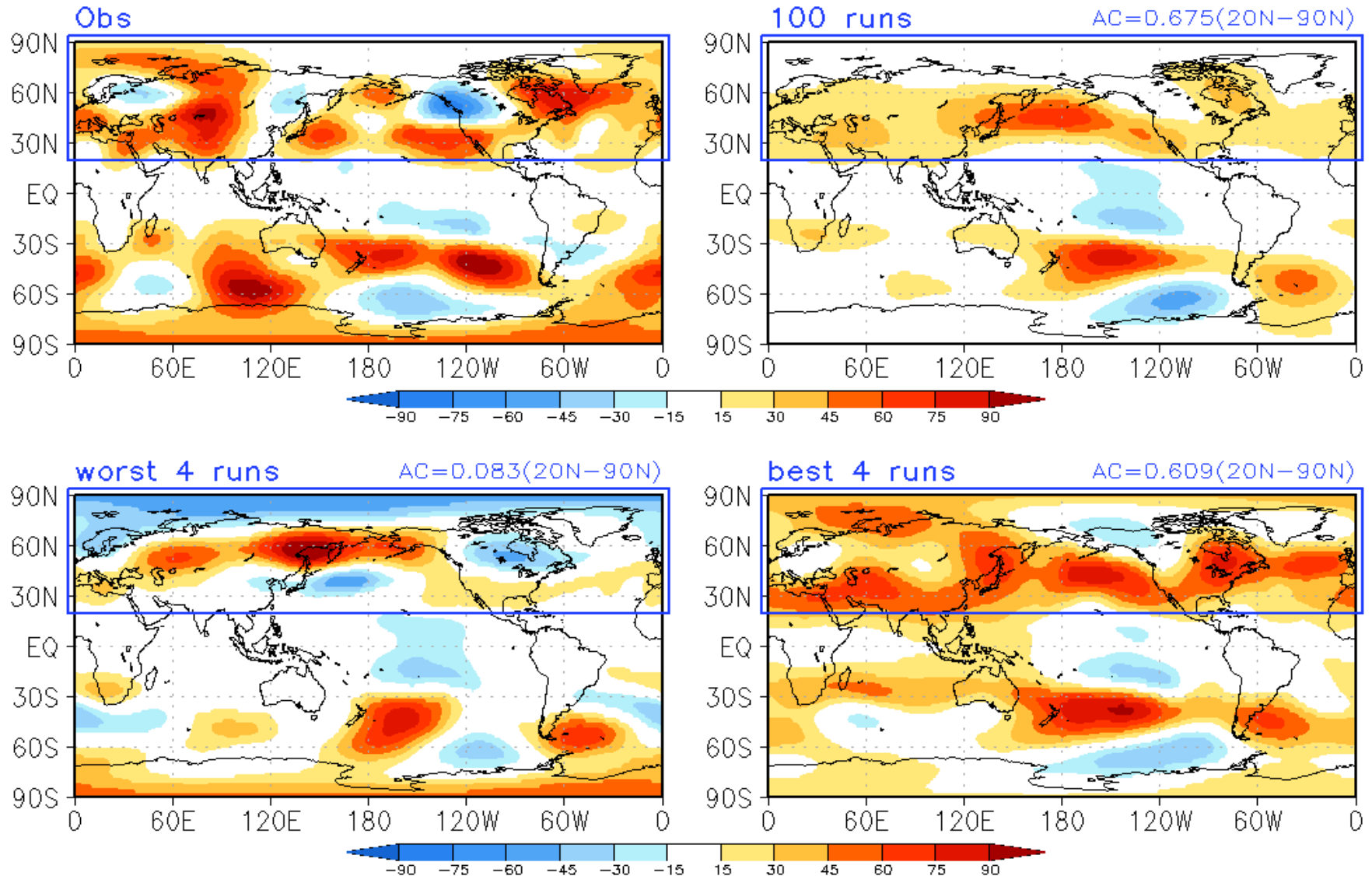
- In this analysis, anomalies from individual model runs are compared against the observed seasonal mean anomalies. The spatial resemblance between them is quantified based on anomaly correlation (AC).
- The distribution of AC across all model simulations is indicative of probability of observed anomalies 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).

AMJ2022 Anomaly Correlation for Individual AMIP Simulation with Observation -- z200(20N-90N)

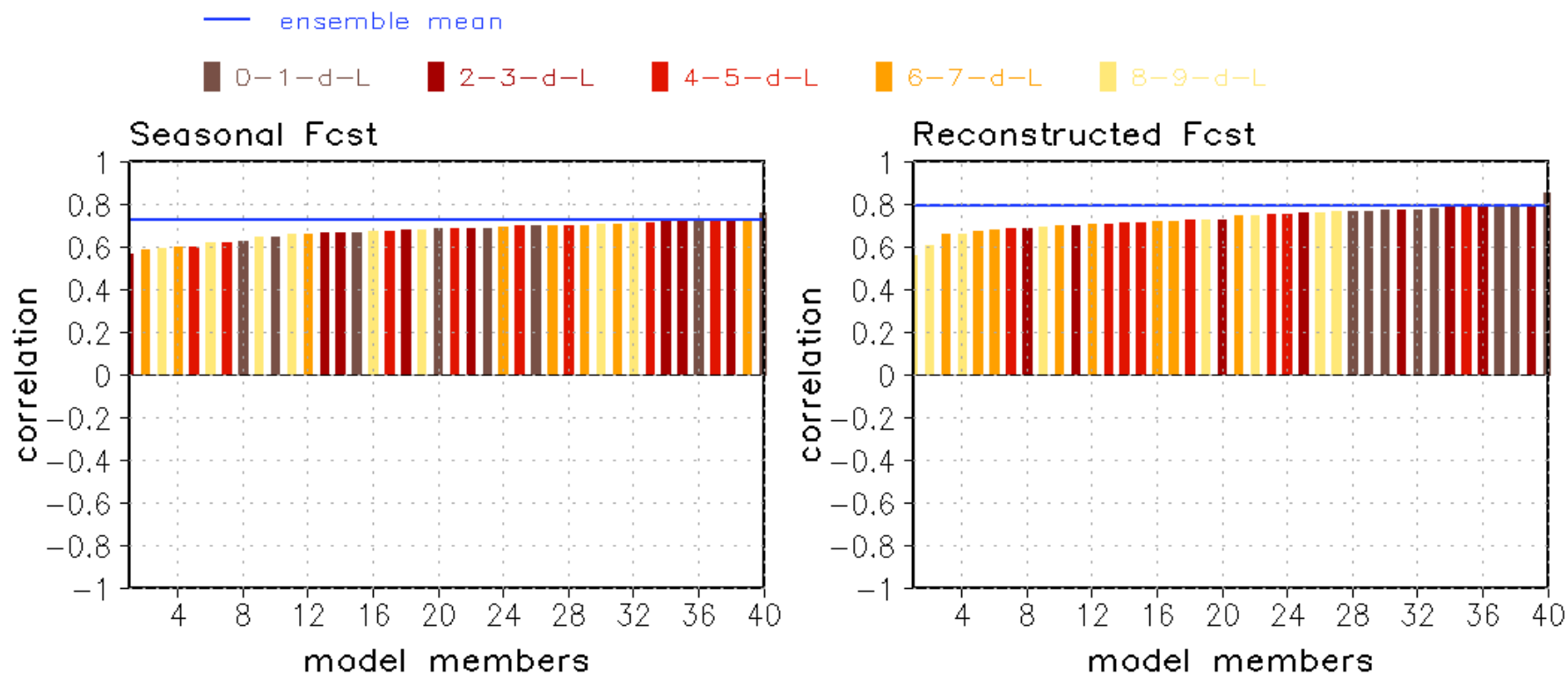


Observed & AMIP Ensemble Mean Anomalies

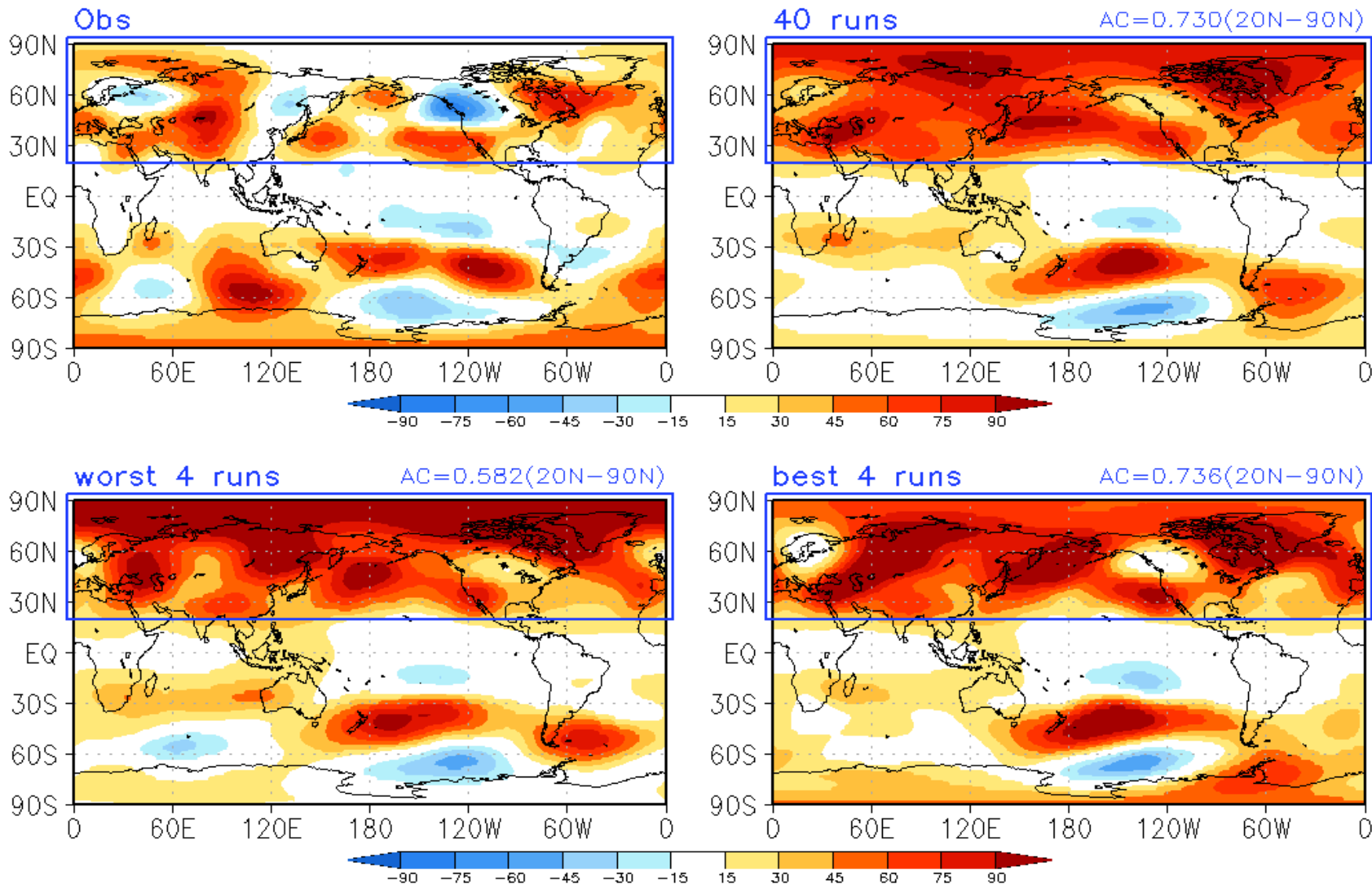
AMJ2022 z200(m) 100 runs/worst 4 runs/best 4 runs



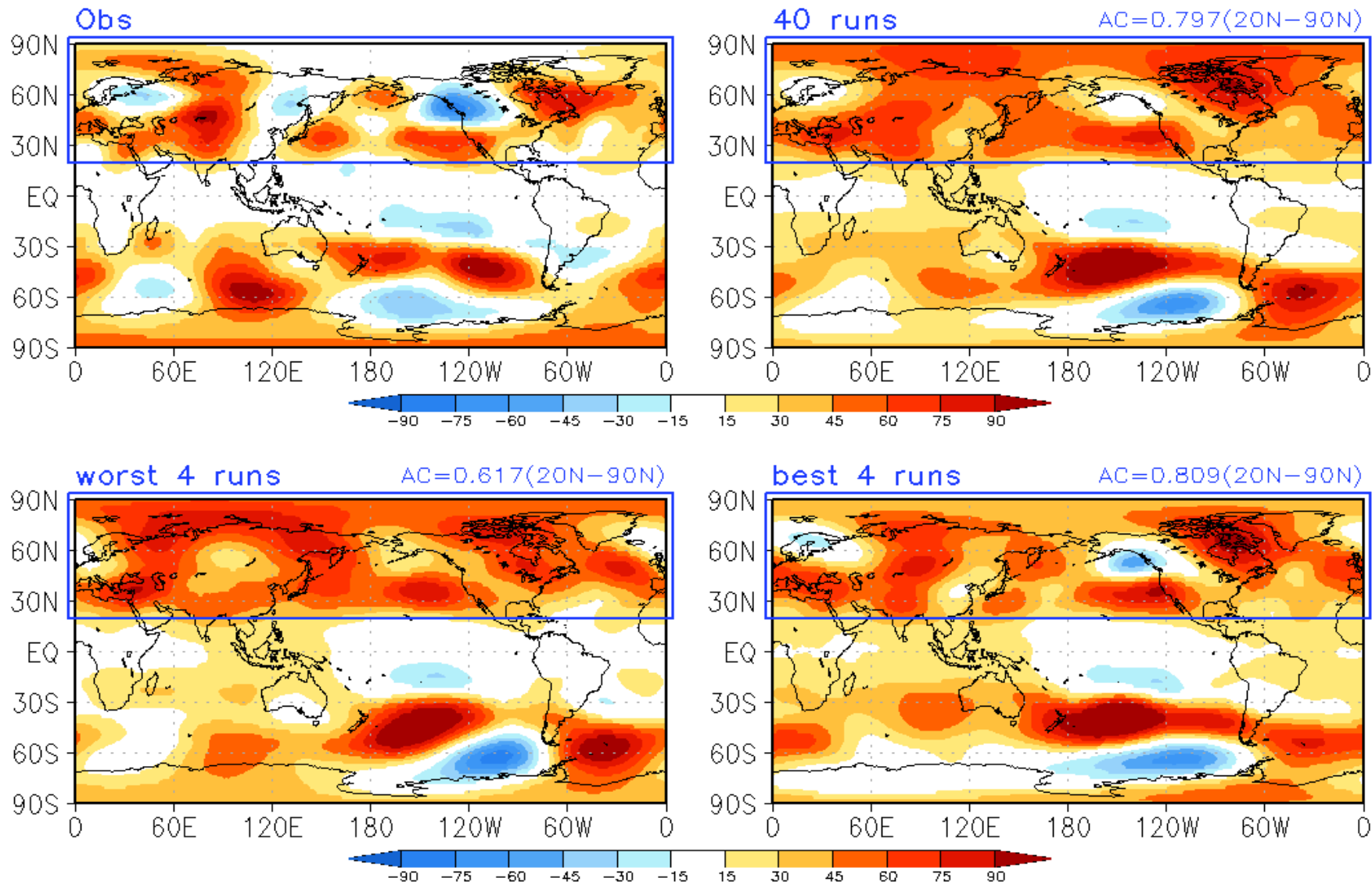
AMJ2022 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- z200 (20N-90N)



Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 z200(m) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



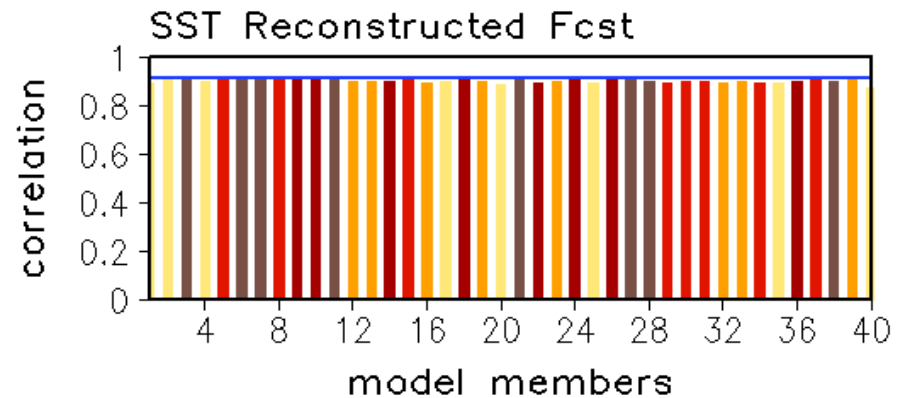
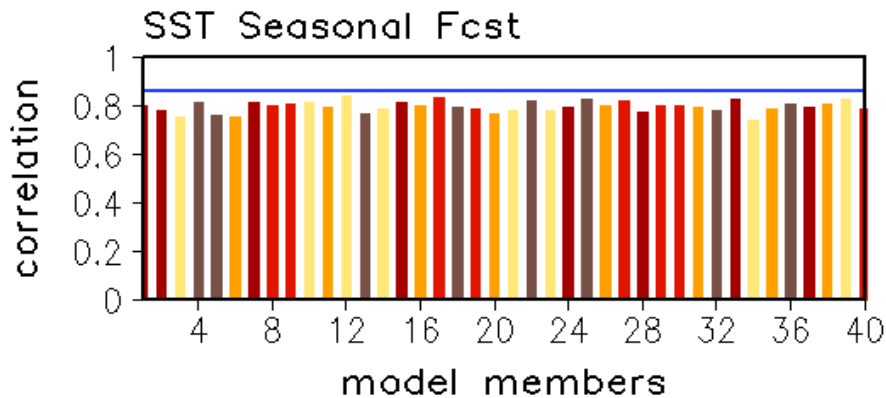
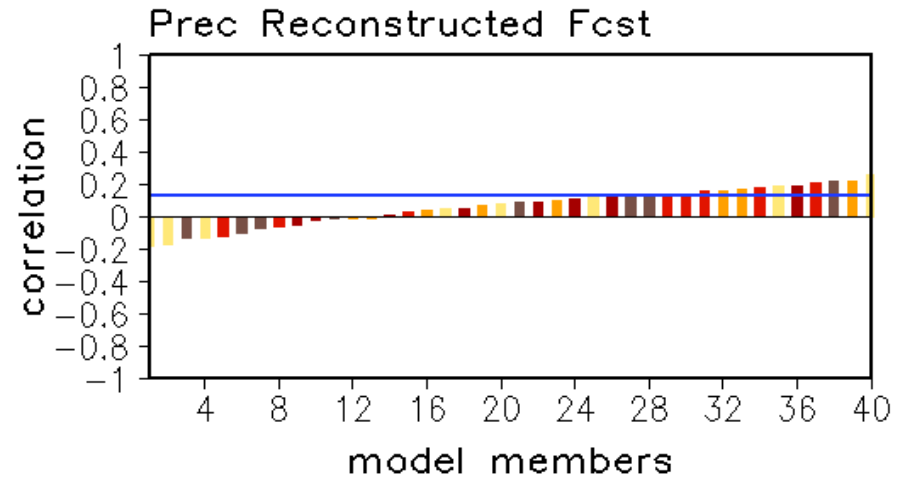
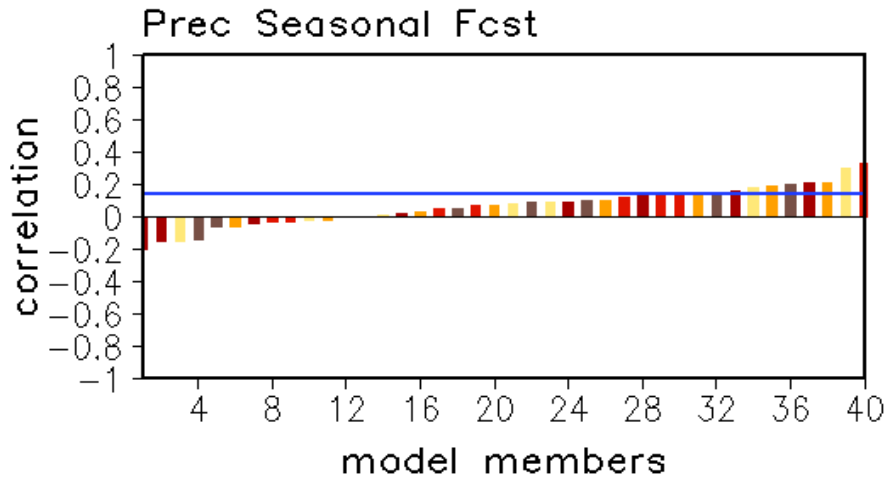
Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 z200(m) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



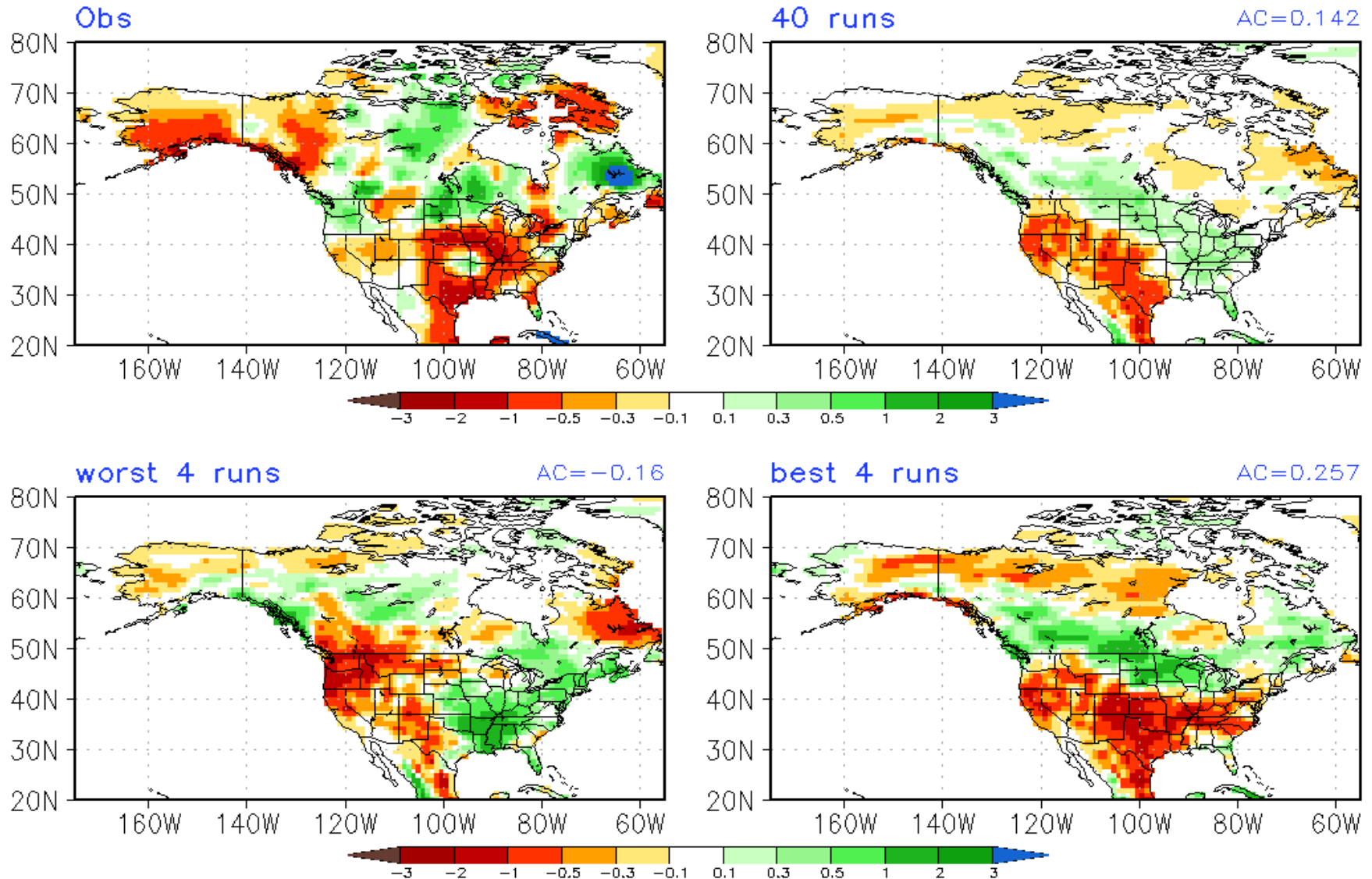
AMJ2022 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- Prec(NA)/SST(30S-30N)

— ensemble mean

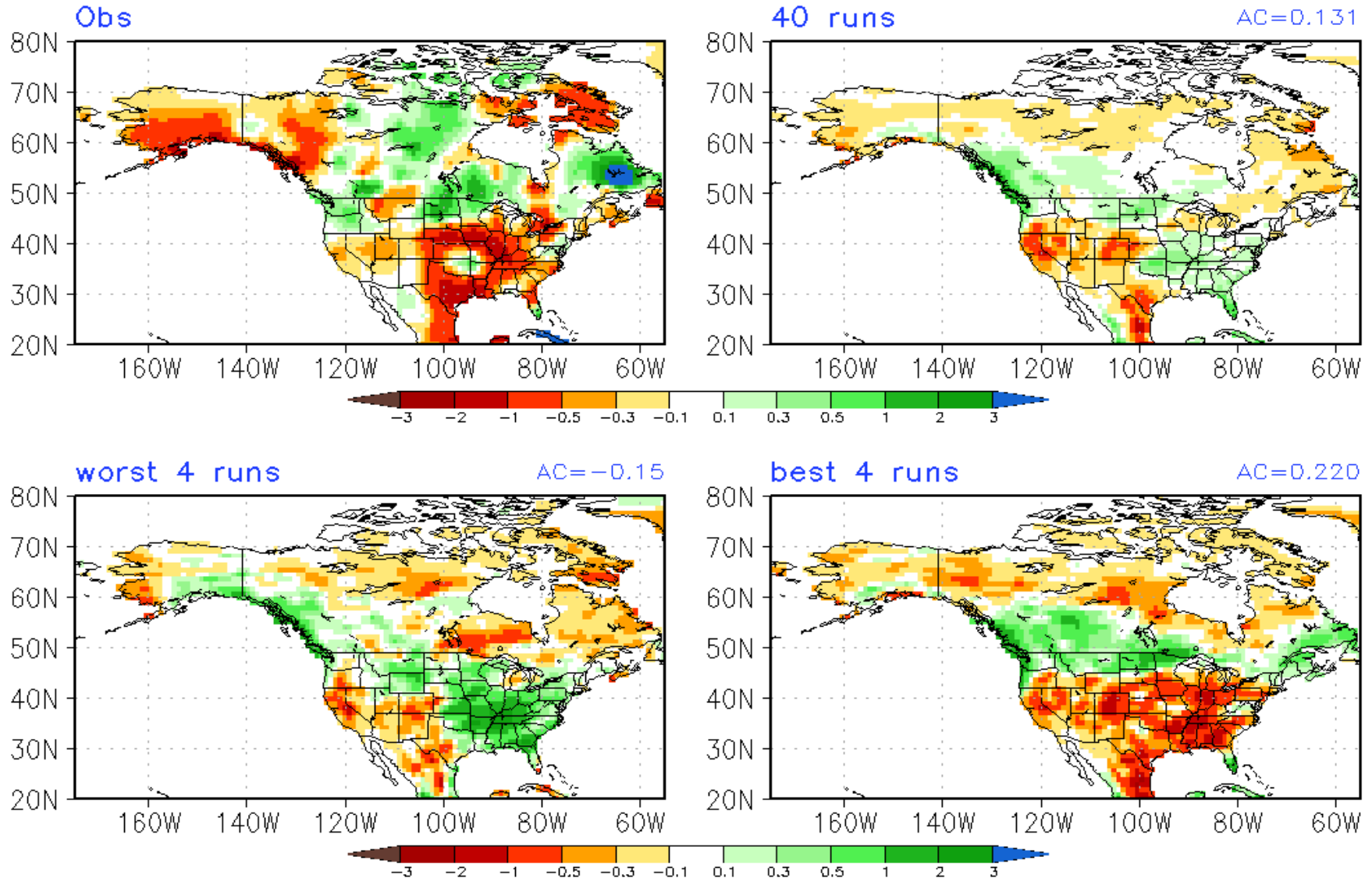
■ 0-1-d-L ■ 2-3-d-L ■ 4-5-d-L ■ 6-7-d-L ■ 8-9-d-L



Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



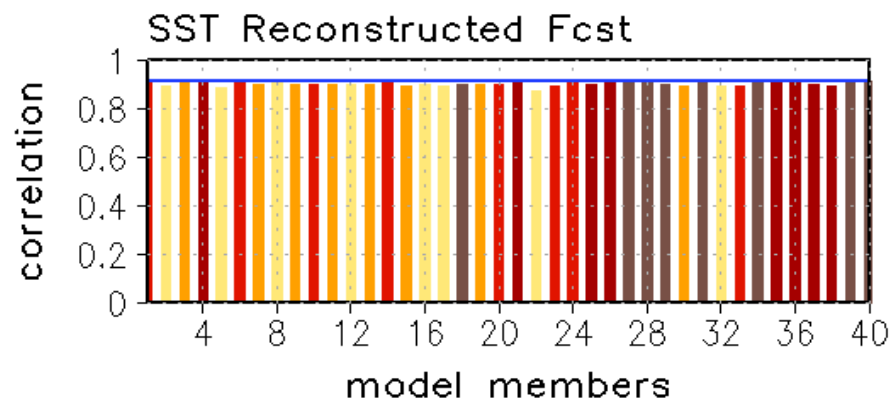
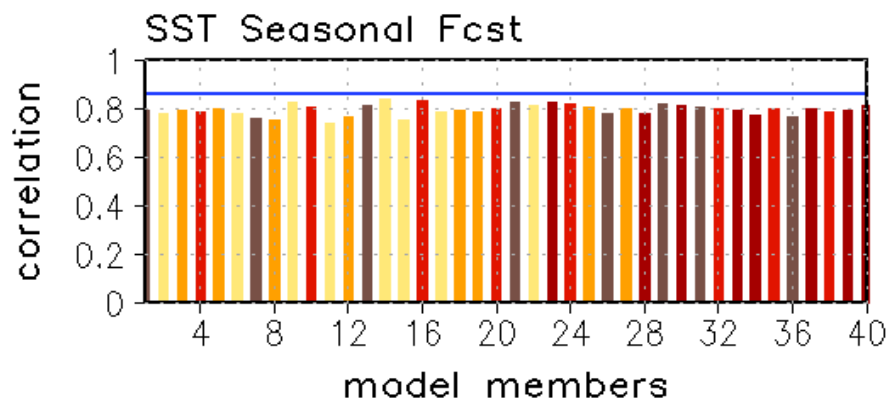
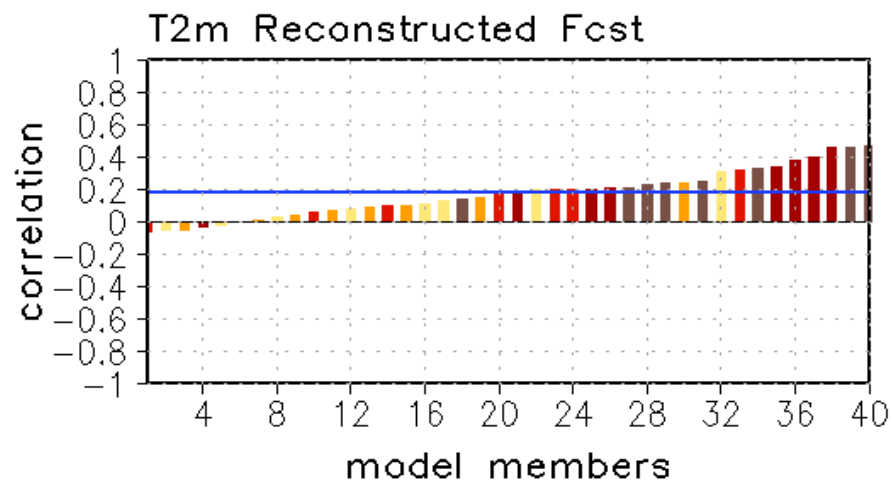
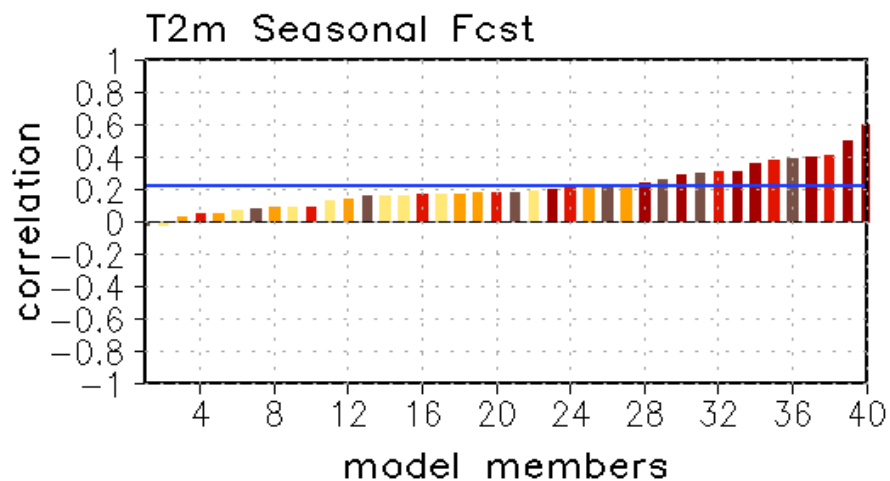
Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



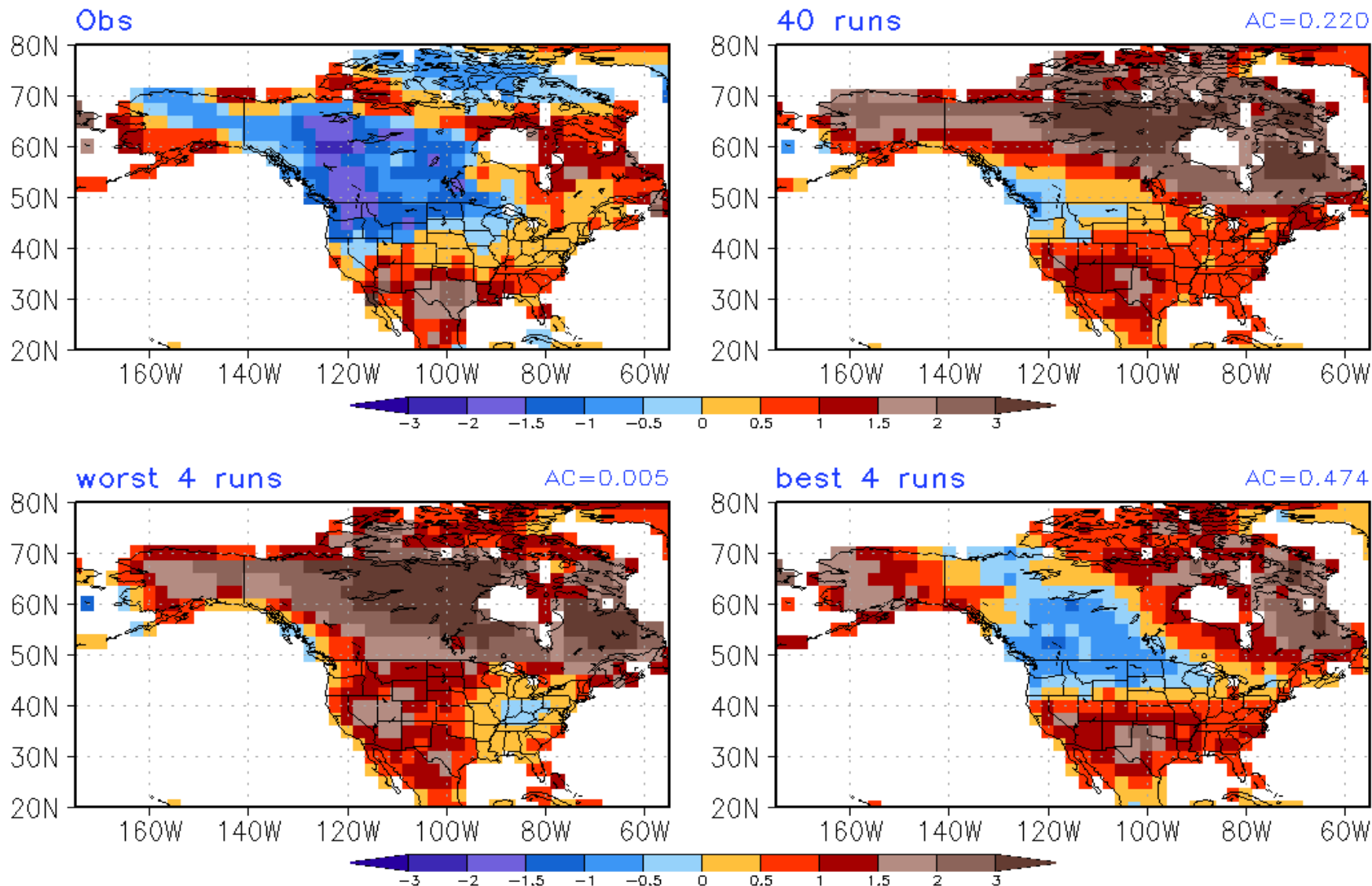
AMJ2022 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- T2m(NA)/SST(30S-30N)

— ensemble mean

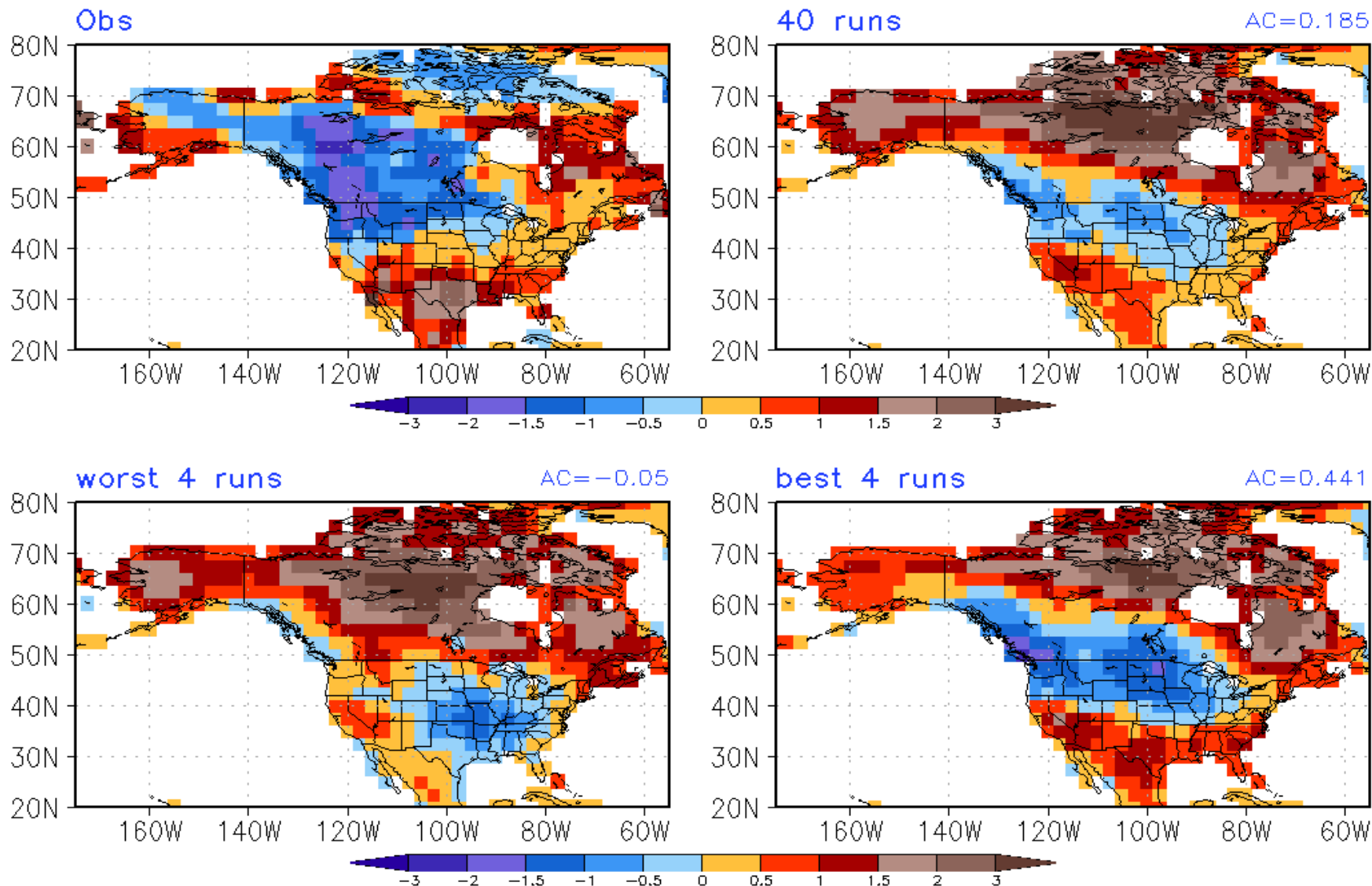
0-1-d-L 2-3-d-L 4-5-d-L 6-7-d-L 8-9-d-L



Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 T2m(K) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast

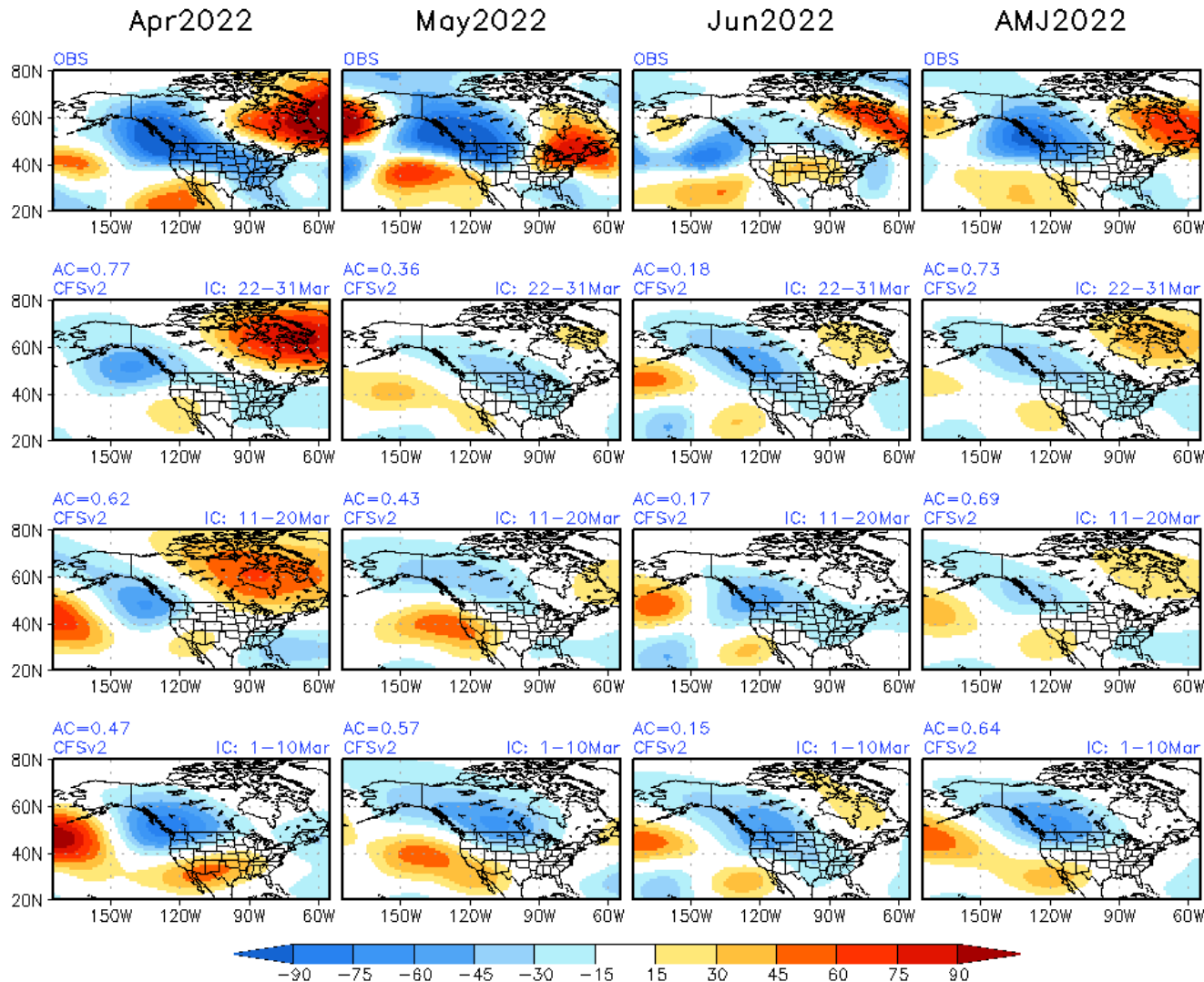


Observed & CFSv2 Forecast Ensemble Average Anomalies AMJ2022 T2m(K) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



z200(m) Monthly Means from Seasonal Forecast

Monthly Means from Seasonal Fcst (40ensm) AMJ2022 z200(m) eddy & Obs



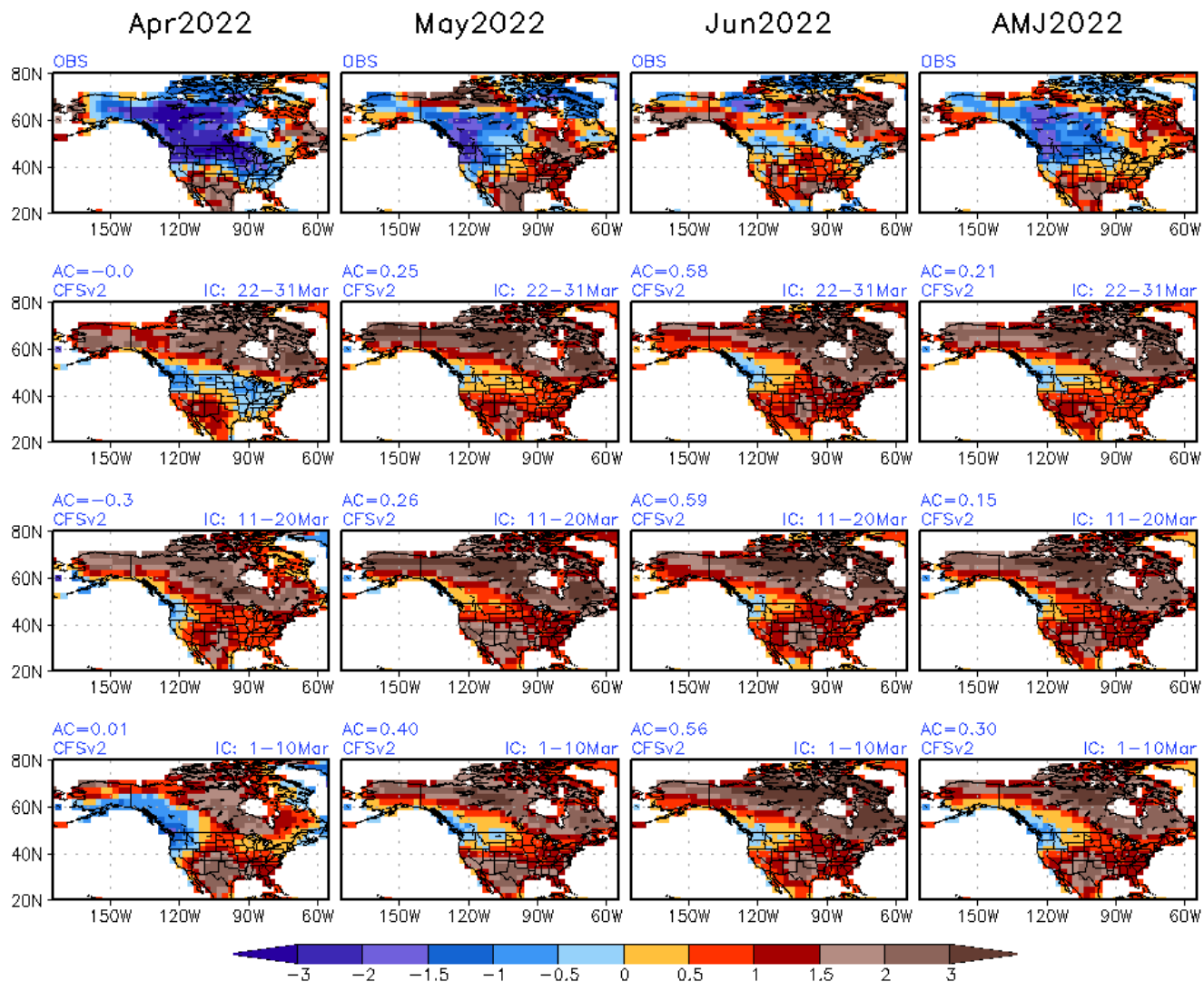
Top row: Observed anomaly.

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) AMJ2022 T2m(K) & Obs



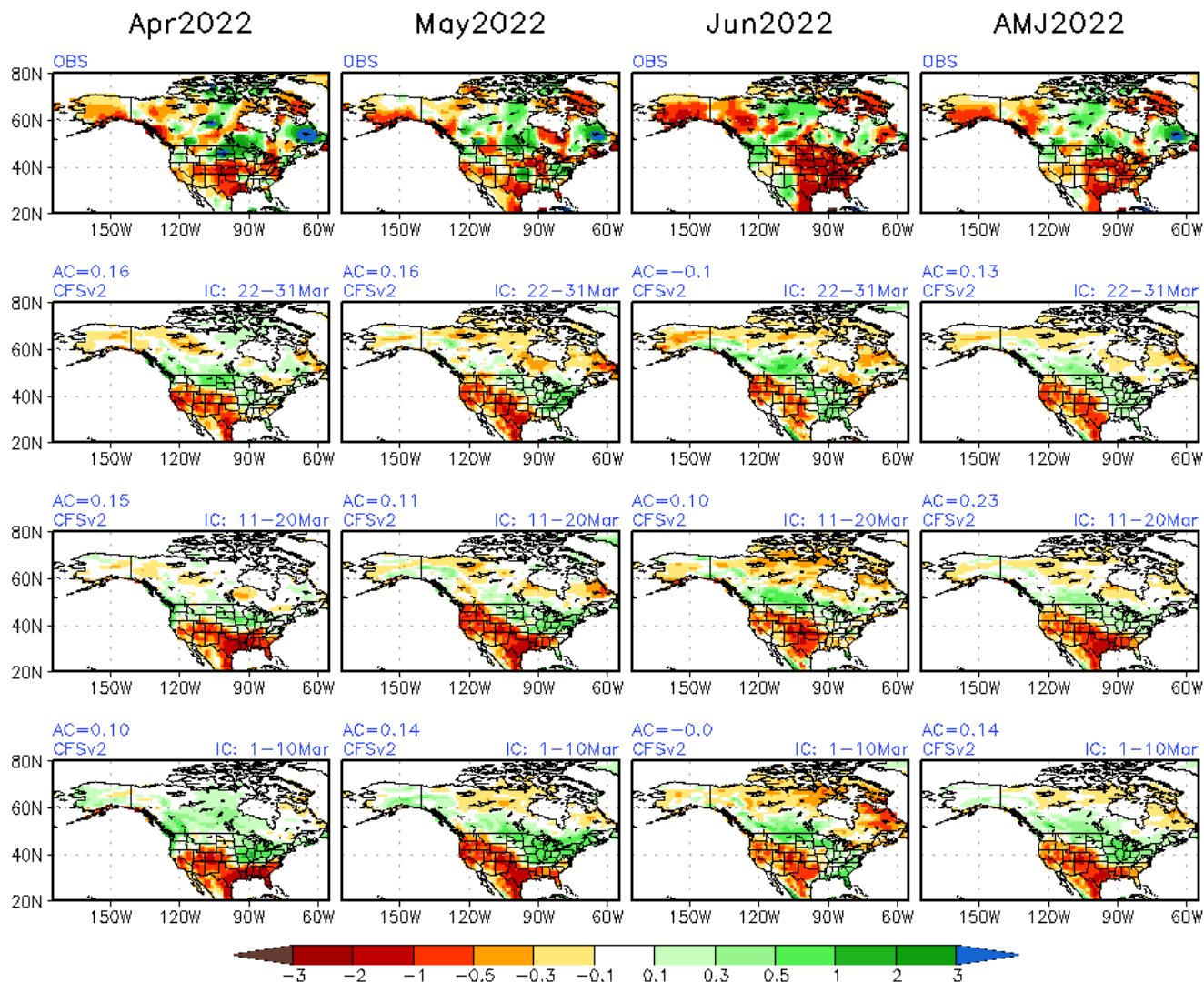
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Prec(mm/day) Monthly Means from Seasonal Forecast

Monthly Means from Seasonal Fcst (40ensm) AMJ2022 Prec(mm/day) & Obs



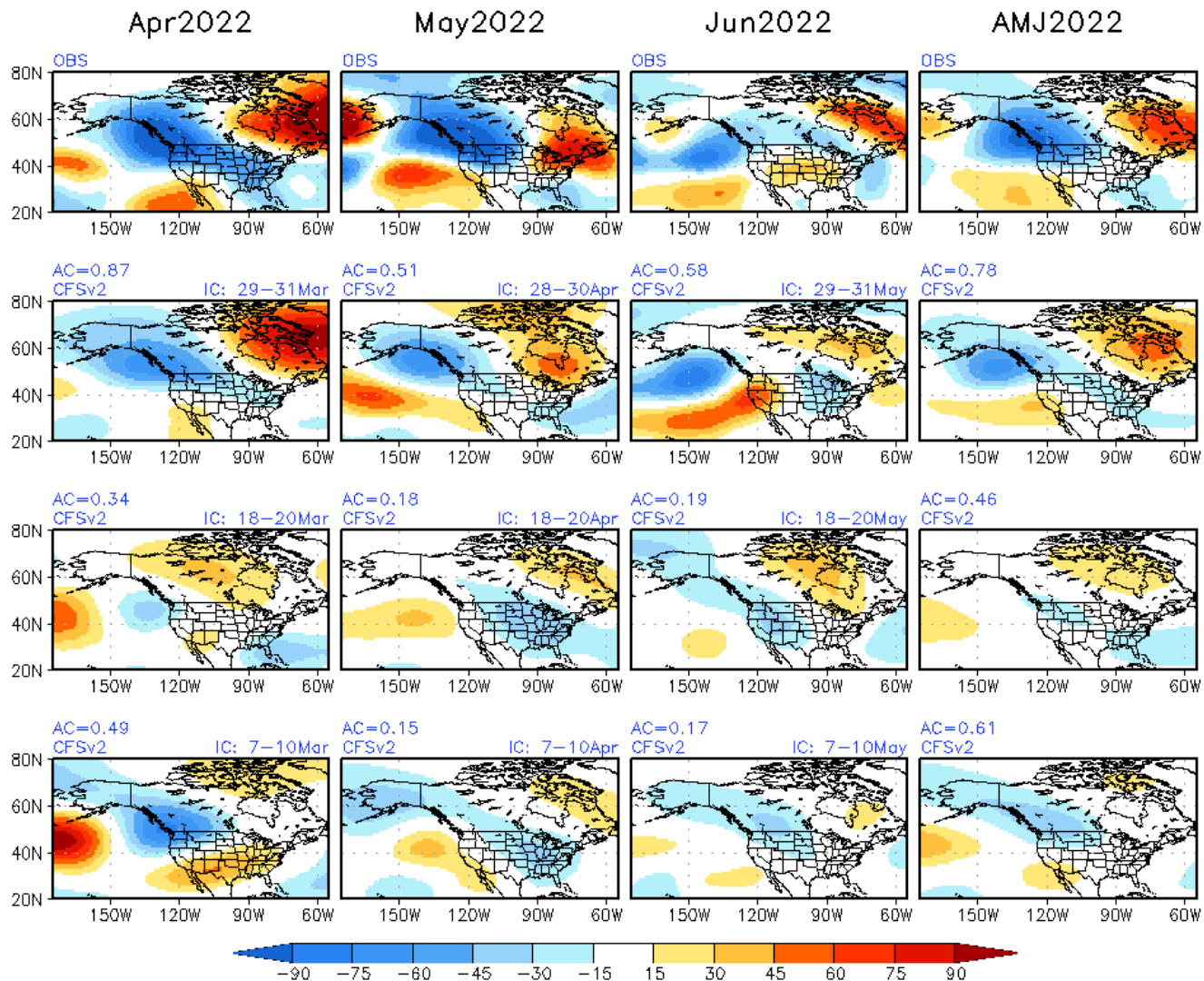
Top row: Observed anomaly.

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z200(m) Monthly Means from Monthly Forecast

Monthly Means from Monthly Fcst AMJ2022 z200(m) eddy & Obs



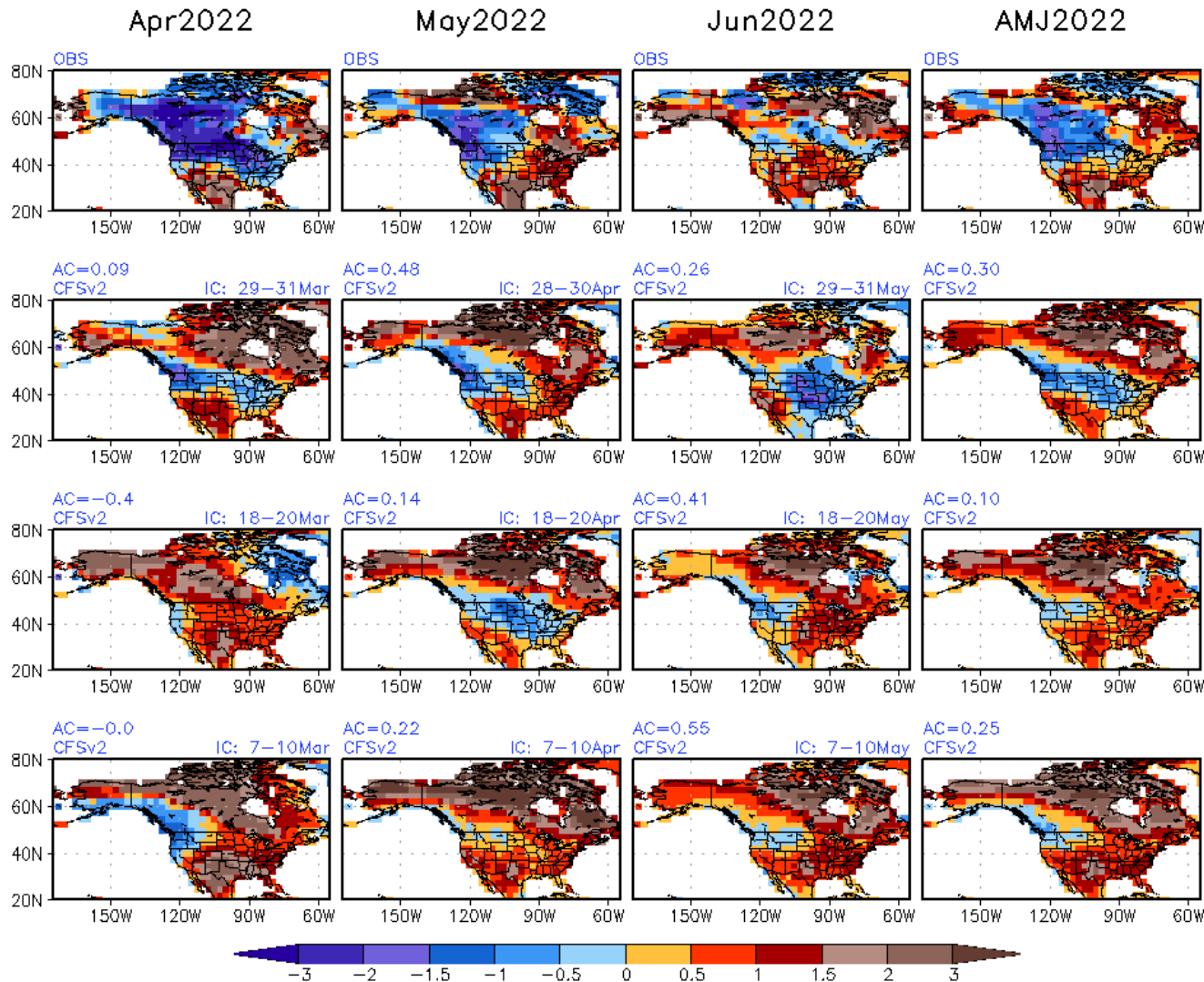
Top row: Observed anomaly.

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

- 2nd row: last 3 days of the prior month.
- 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 AMJ2022 T2m(K) & Obs



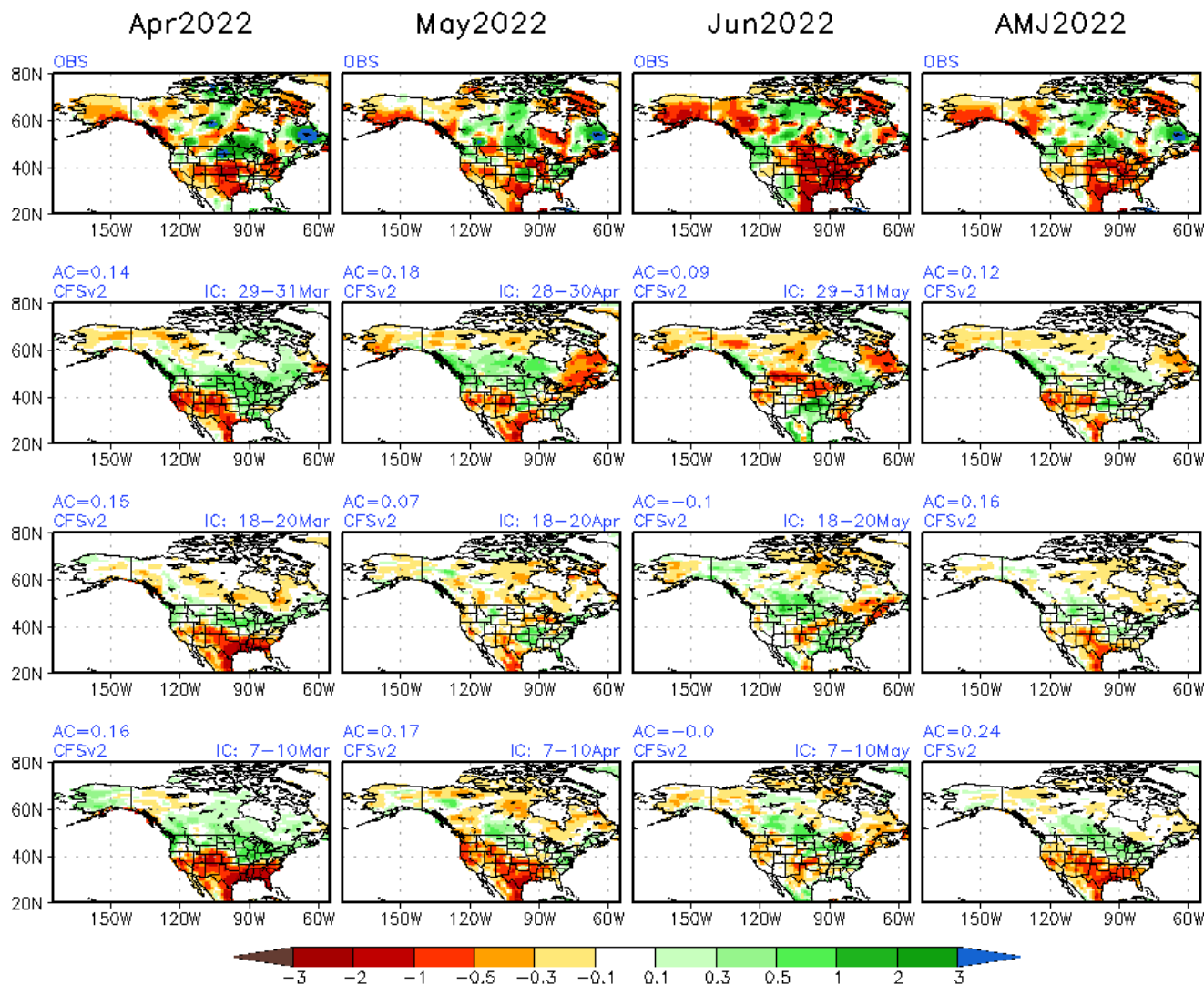
Top row: Observed anomaly.

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

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T2m(k) Monthly Means from Monthly Forecast

Monthly Means from Monthly Fcst AMJ2022 Prec(mm/day) & Obs



Top row: Observed anomaly.

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

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- 4th row: 7th – 10th of the prior month.

Seasonal Forecasts from Multi-Model Ensemble Systems

- WMO Lead Center for Long-Range Forecast Multi-Model Ensemble (LC-LRFMME). <https://www.wmolc.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.wmolc.org/>)

Ensemble means

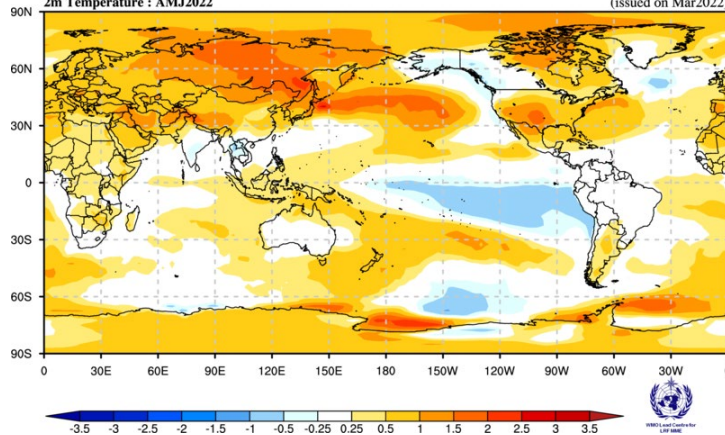
Simple Composite Map

Beijing,CMCC,CPTec,ECMWF,Exeter,Melbourne,Monreal,Moscow,Offenbach,Seoul,Tokyo,Toulouse,Washington

[Unit : K]

2m Temperature : AMJ2022

(issued on Mar2022)



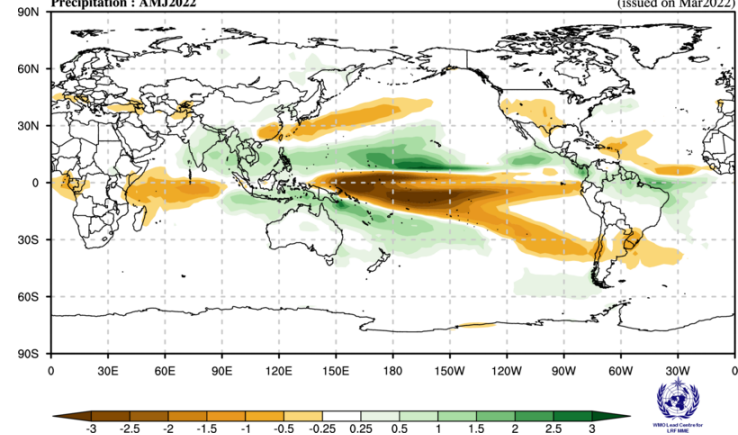
Simple Composite Map

Beijing,CMCC,CPTec,ECMWF,Exeter,Melbourne,Monreal,Moscow,Offenbach,Seoul,Tokyo,Toulouse,Washington

[Unit : mm]

Precipitation : AMJ2022

(issued on Mar2022)



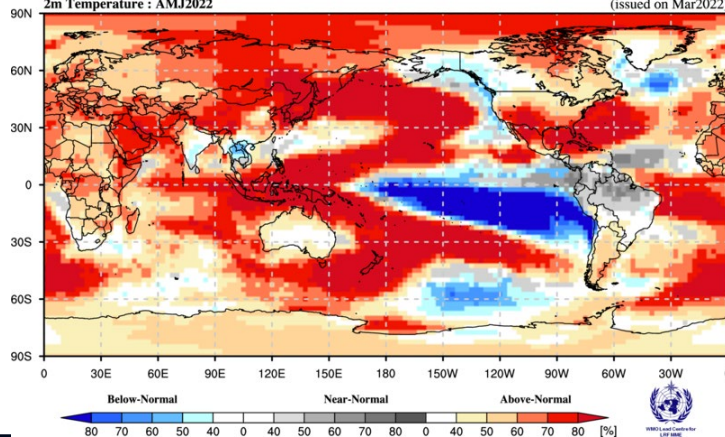
Probabilities

Probabilistic Multi-Model Ensemble Forecast

Beijing,CMCC,CPTec,ECMWF,Exeter,Melbourne,Monreal,Moscow,Offenbach,Seoul,Tokyo,Toulouse,Washington

2m Temperature : AMJ2022

(issued on Mar2022)

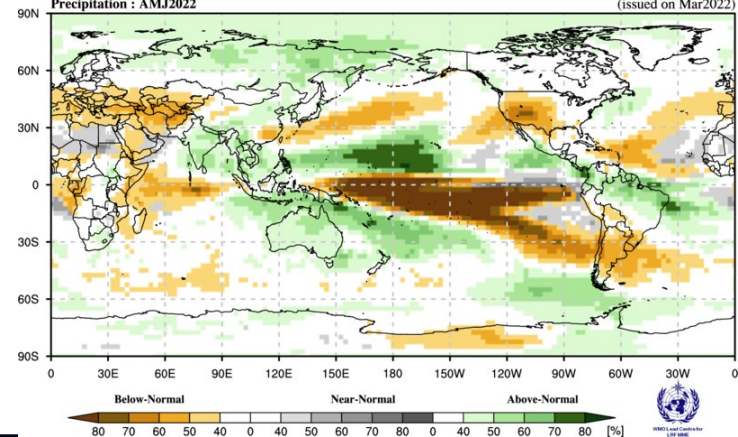


Probabilistic Multi-Model Ensemble Forecast

Beijing,CMCC,CPTec,ECMWF,Exeter,Melbourne,Monreal,Moscow,Offenbach,Seoul,Tokyo,Toulouse,Washington

Precipitation : AMJ2022

(issued on Mar2022)

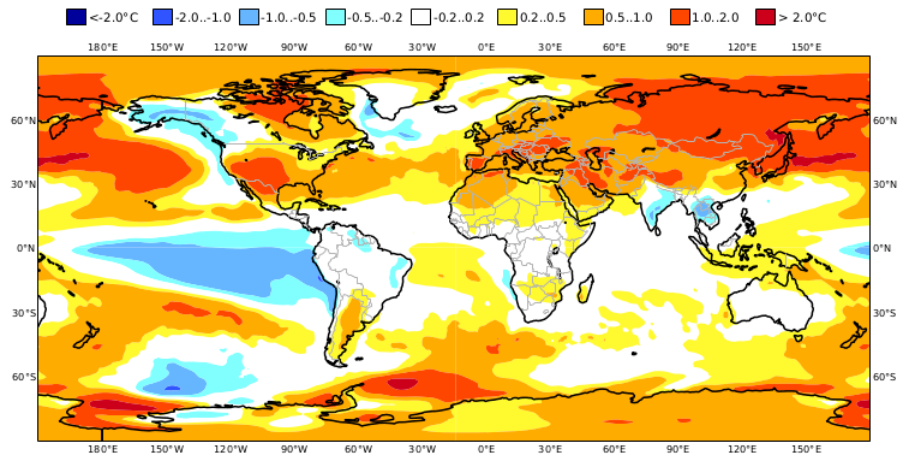


C3S Seasonal Forecast

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

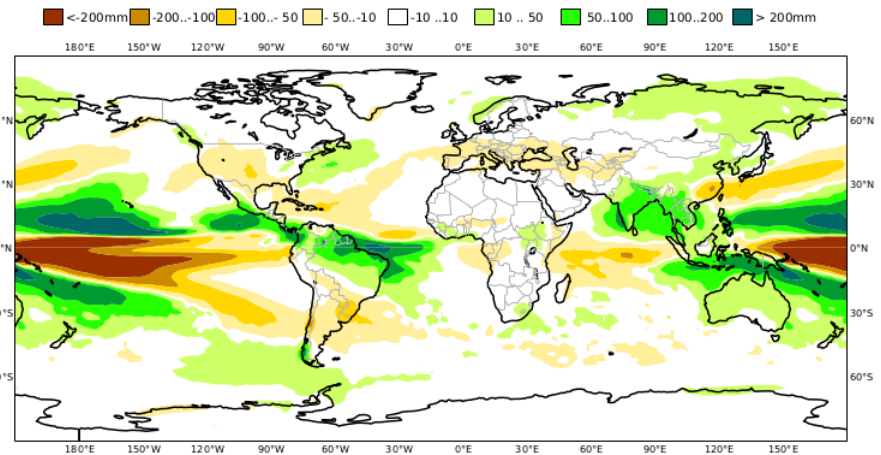
C3S multi-system seasonal forecast
Mean 2m temperature anomaly
Nominal forecast start: 01/03/22
Variance-standardized mean

ECMWF/Met Office/Météo-France/CMCC/DWD/NCEP/JMA/ECCC
AMJ 2022



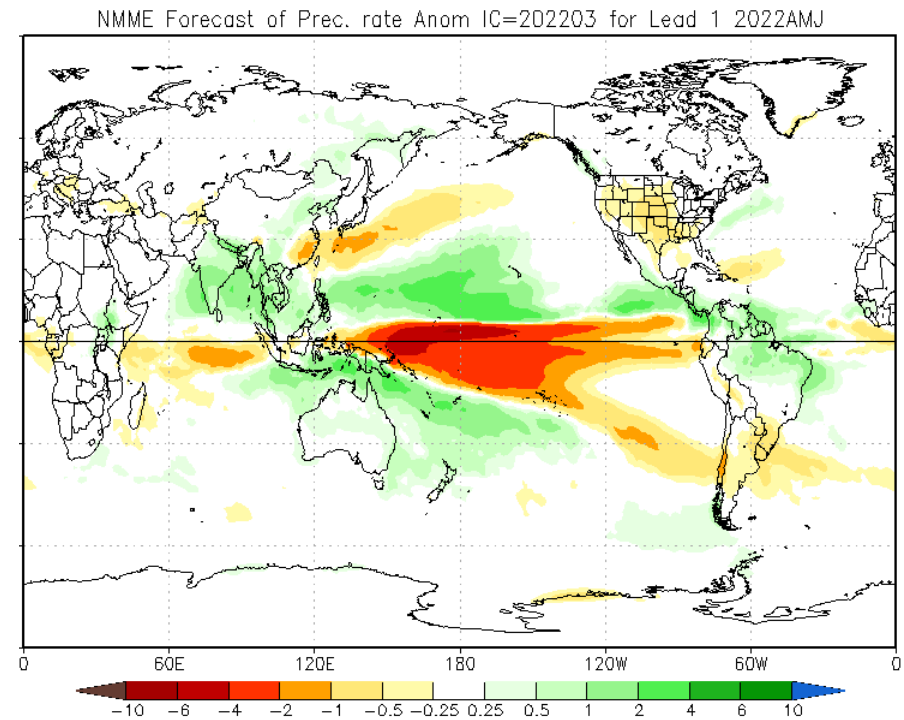
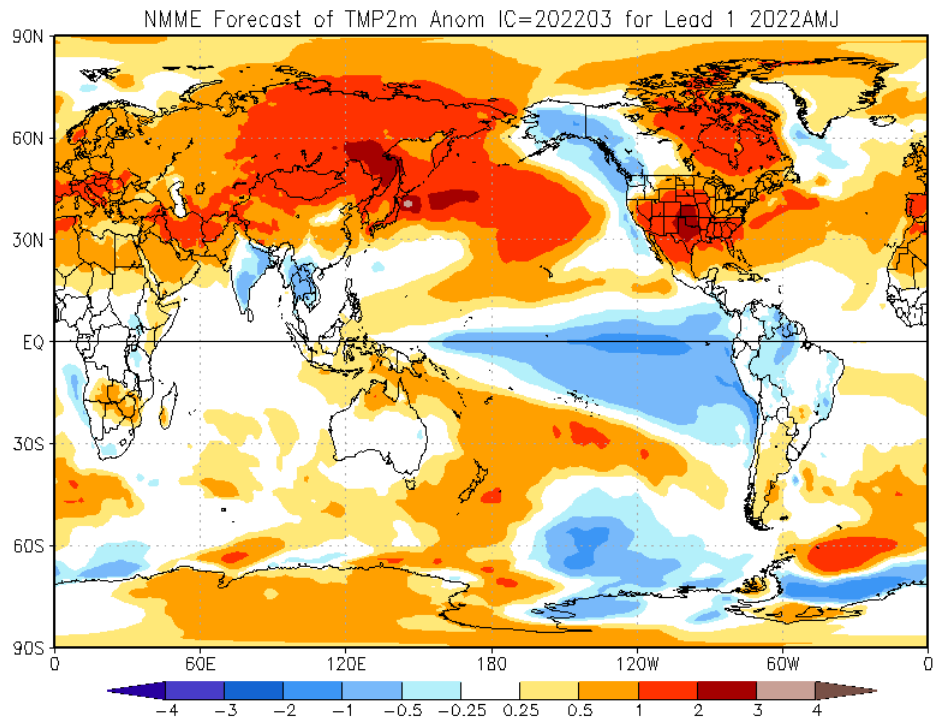
C3S multi-system seasonal forecast
Mean precipitation anomaly
Nominal forecast start: 01/03/22
Variance-standardized mean

ECMWF/Met Office/Météo-France/CMCC/DWD/NCEP/JMA/ECCC
AMJ 2022



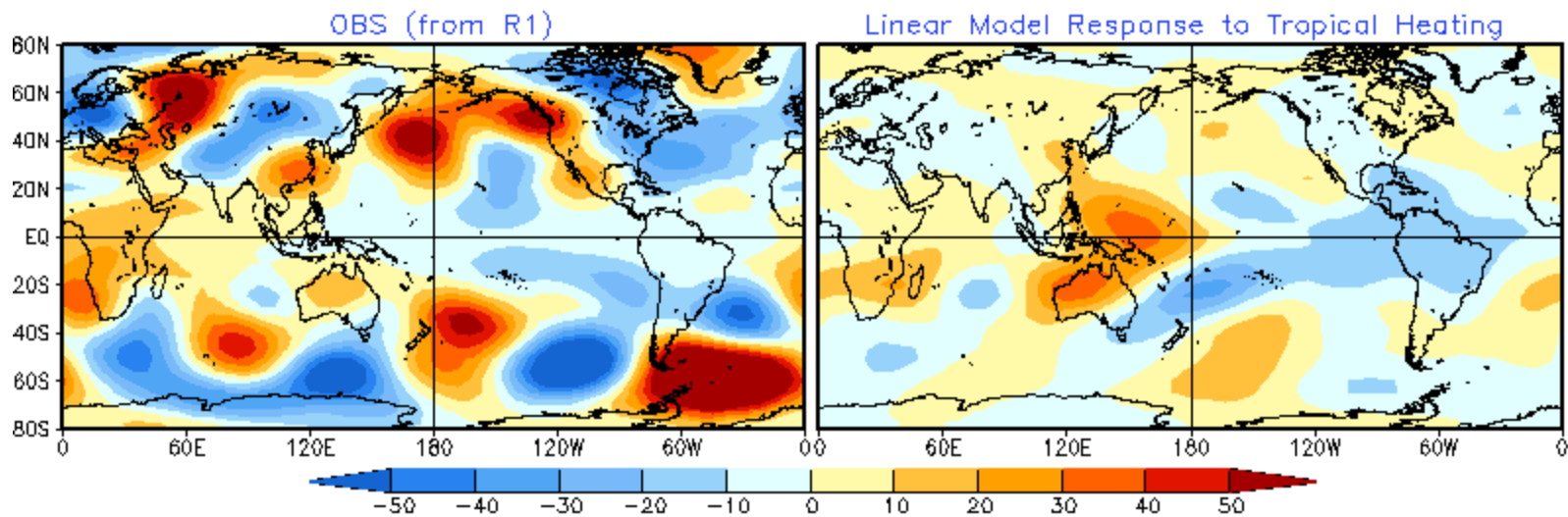
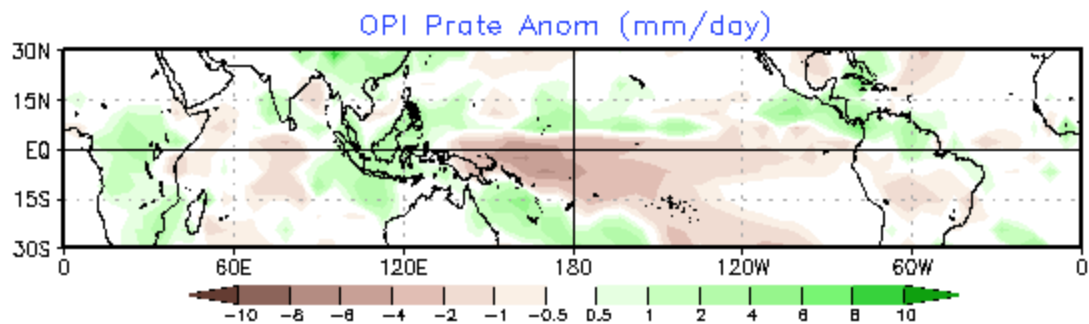
North American Multi-Model Ensemble Seasonal Forecast

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



200mb Height from Linear Model

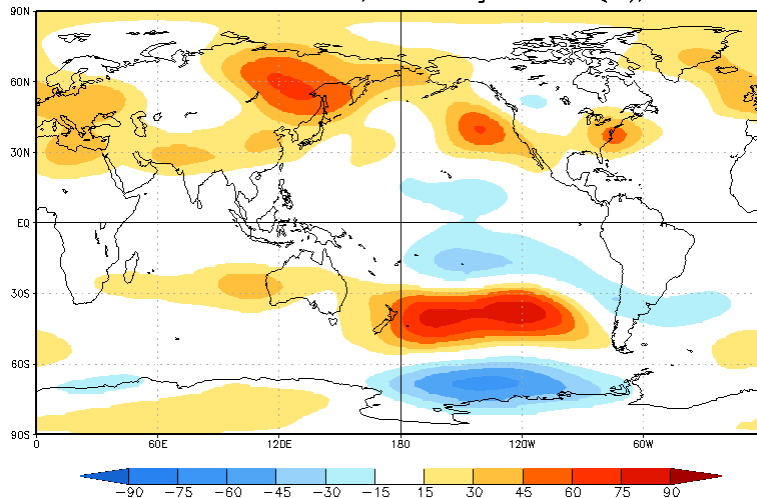
AMJ2022 200mb Eddy HGT(m)
OBS vs. Linear Model Response to Tropical Heating
Heating is converted from Prate in 15S-15N



Pattern COR: global=0.12, tropics(30S-30N)=0.25

Seasonal Forecasts from the Constructed Analog Model

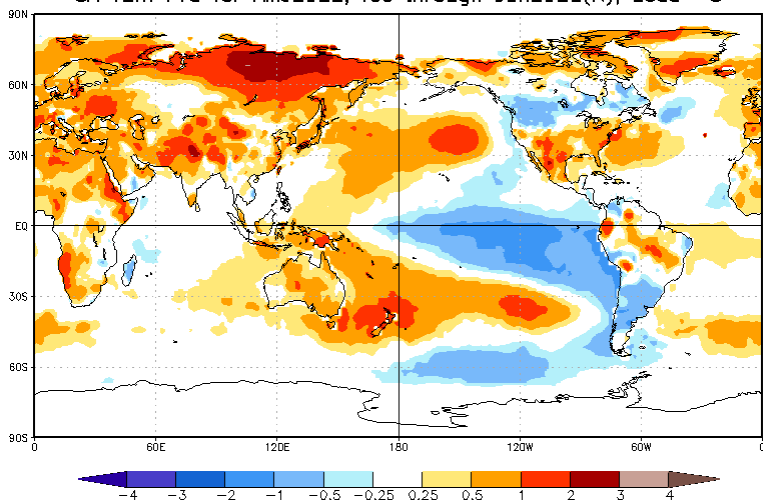
CA HGT200 Prd for AMJ2022, ICs through Jun2022(m), Lead -3



Peitao Peng CPC/NCEP/NWS/NOAA

Base Period 1991-2020

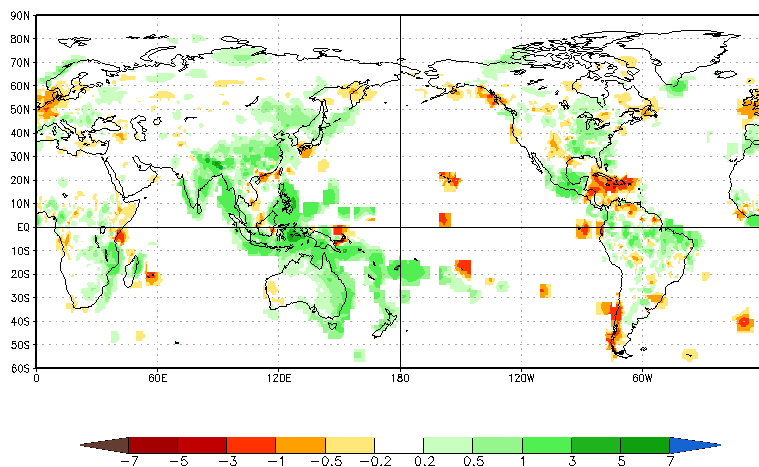
CA T2m Prd for AMJ2022, ICs through Jun2022(K), Lead -3



Peitao Peng CPC/NCEP/NWS/NOAA

Base Period 1991-2020

CA Prec Prd for AMJ2022, ICs through Jun2022(mm/day), Lead -3



Peitao Peng CPC/NCEP/NWS/NOAA

Base Period 1991-2020

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)