

# Attribution of Seasonal Climate Anomalies February-March-April 2020

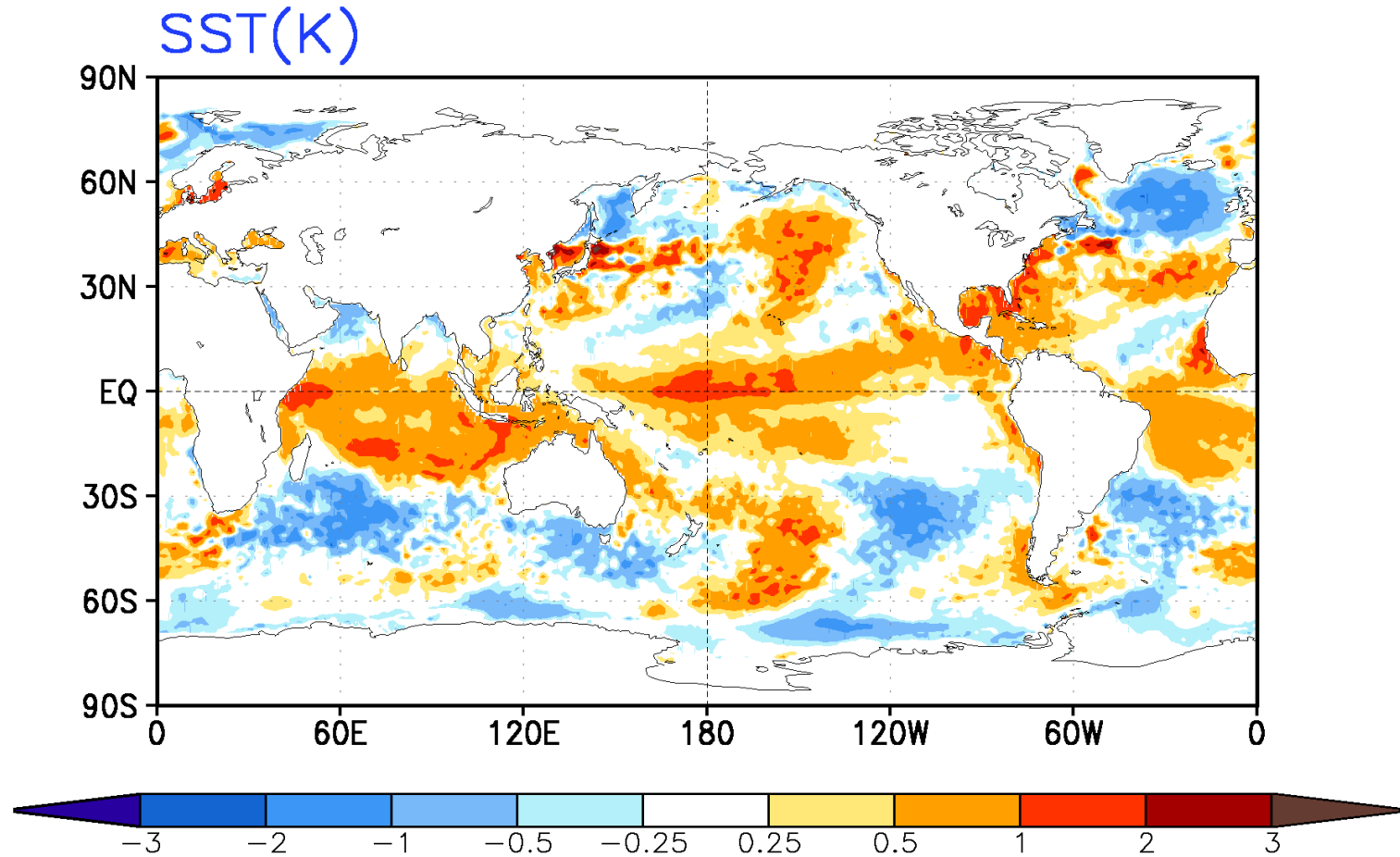
# Summary of January-February-March 2020 Observed Conditions and Outlooks

- The sea surface temperature (SST) anomalies in the equatorial central Pacific Ocean continued to be on weak warm side; the long lasting strong positive phase of Indian Ocean Dipole (IOD) SST anomaly pattern (since summer 2019) changed to positive SST anomalies over the entire Indian Ocean in FMA2020; (Slide 4).
- Except over northern regions of North America, land temperature anomalies were generally above normal.
- The large-scale pattern of the precipitation anomalies in tropics were generally well predicted in CFSv2 (slide 11) and the multi-model ensemble forecasts from the WMO Lead Center (slides 38 &40).
- The large-scale structures of the observed positive height anomalies over the tropical and N. sub-tropical region were well captured in the CFSv2 forecasts, while the observed negative height anomalies in northern hemisphere high latitudes and polar region were missed leading to missed cold surface temperature anomalies over extreme northern parts of NA including Alaska.
- The initialized seasonal forecasts well captured negative (positive ) precipitation anomalies over the US west coast extending from northern California to Pacific northwest (SE US region including Florida) .
- NA z200, precipitation, and T2m JFM2020 initialized forecasts had noticeable improvements over the AMIP simulations with further improvements in the reconstructed seasonal forecasts (slides 14-16). The monthly forecasts from the shortest 3-day-leads initial conditions captured major features of NA precip and T2m anomalies (slides 30 &31).

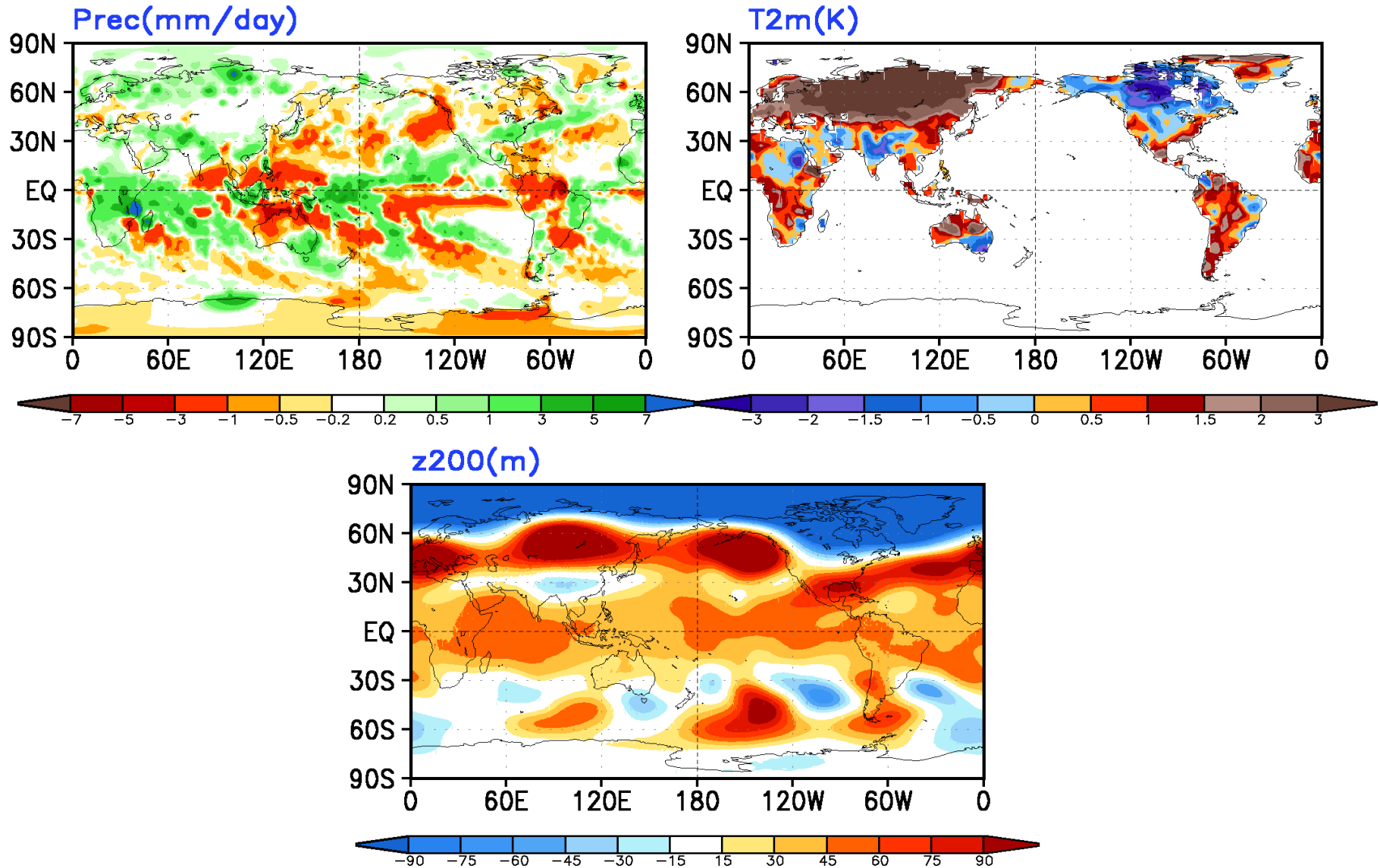
# Observed Seasonal Anomalies

## Global and North America

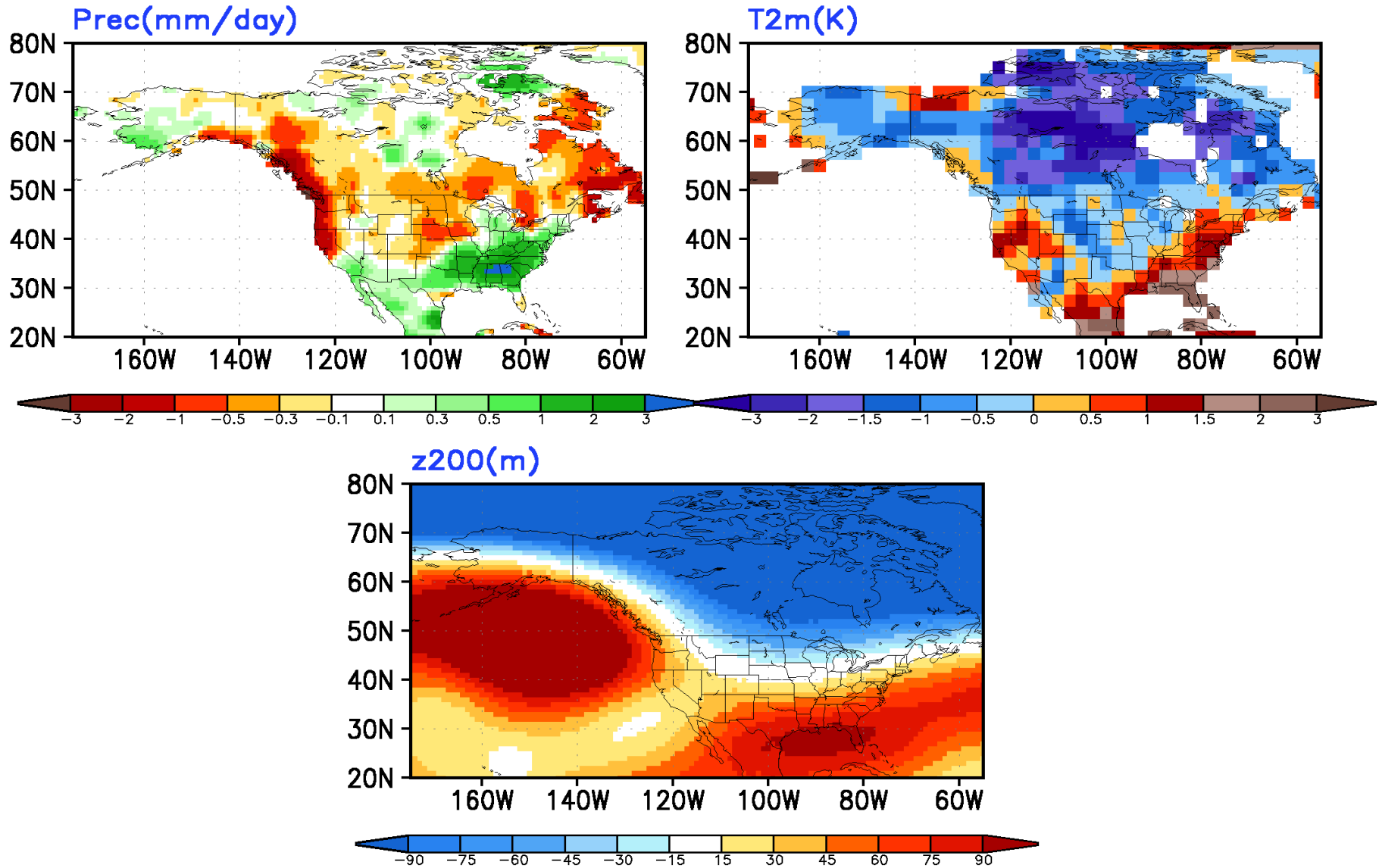
# Observed Anomaly FMA2020



# Observed Anomaly FMA2020



# Observed Anomaly FMA2020



# FMA2020 CPC Seasonal Outlooks and NMME Forecasts

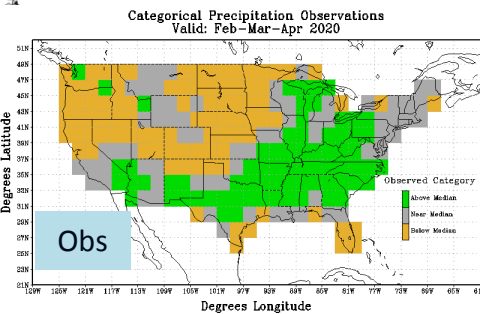
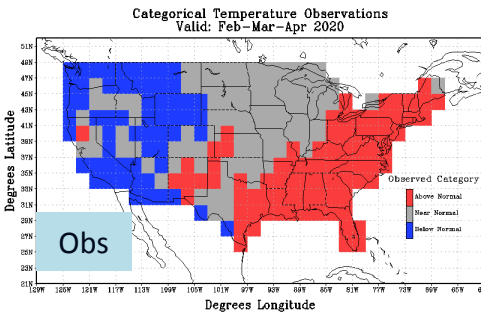
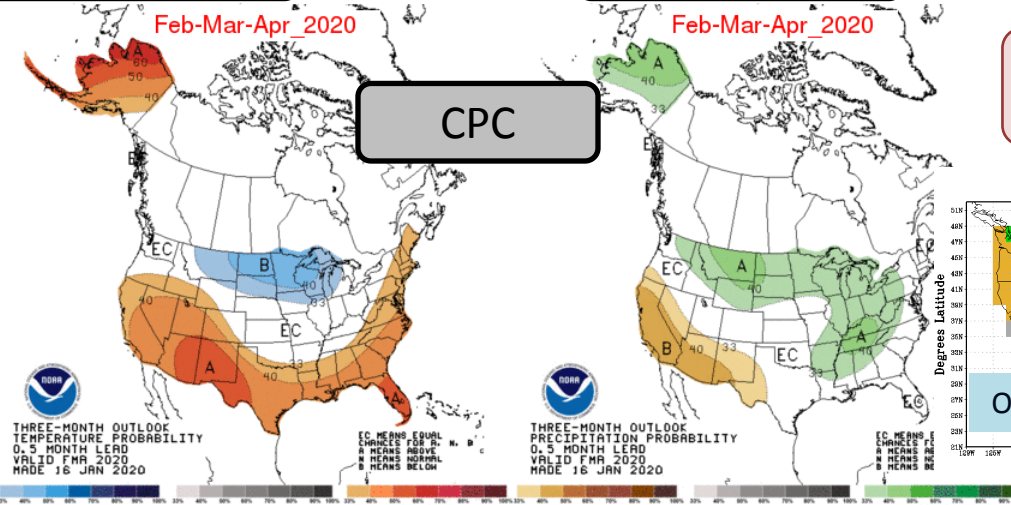
Temperature

Precipitation

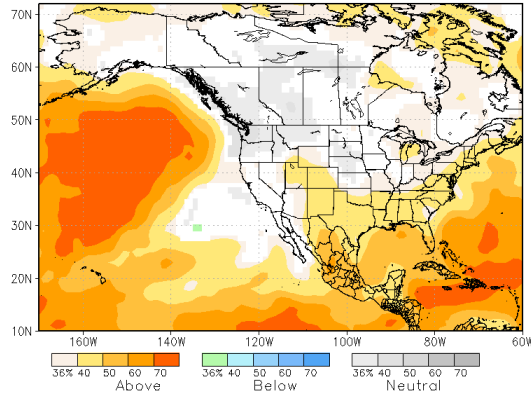
Temp nonEC  
HSS=13

Prec nonEC  
HSS=4

CPC

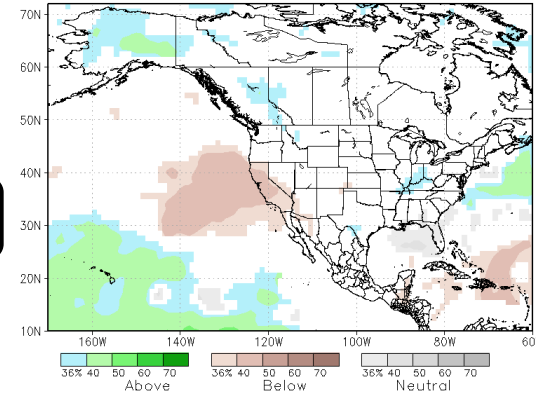


NMME prob fcst TMP2m IC=202001 for lead 1 2020 FMA



NMME

NMME prob fcst Prate IC=202001 for lead 1 2020 FMA



For the rationale behind CPC outlooks see [https://www.cpc.ncep.noaa.gov/products/archives/long\\_lead/PMD/2020/202001\\_PMD90D](https://www.cpc.ncep.noaa.gov/products/archives/long_lead/PMD/2020/202001_PMD90D)

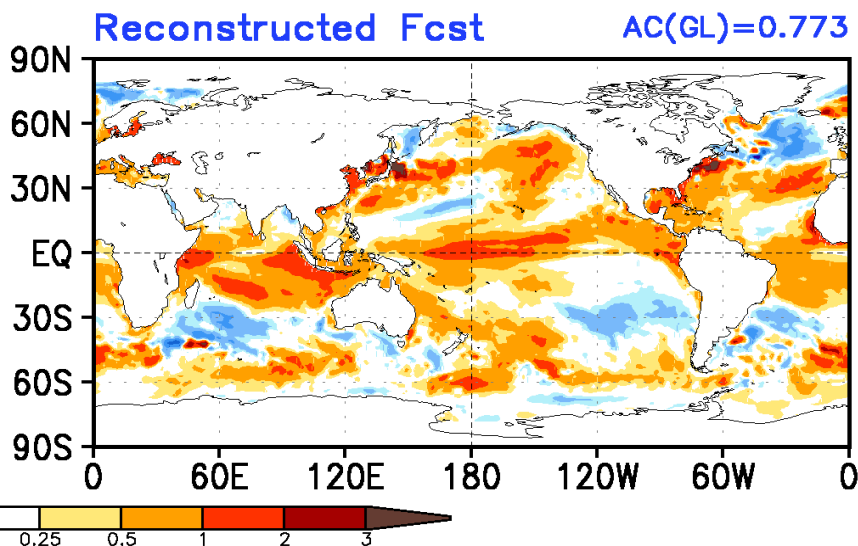
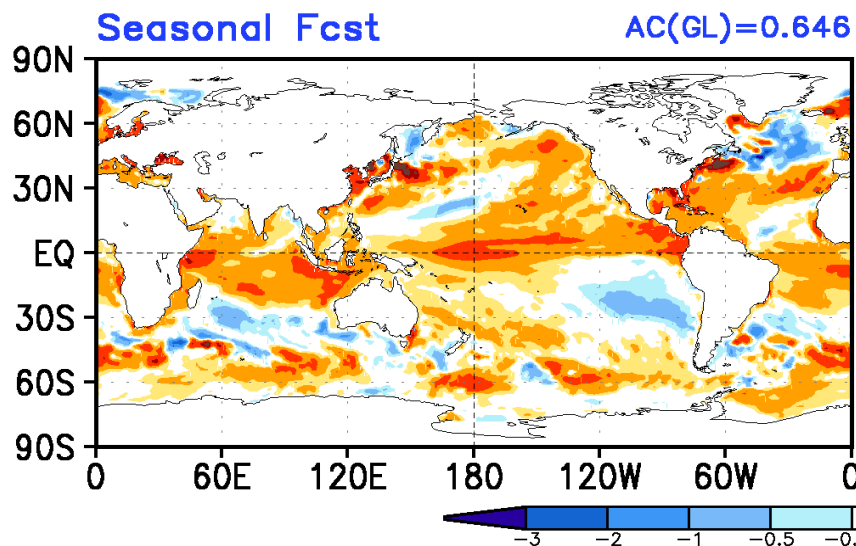
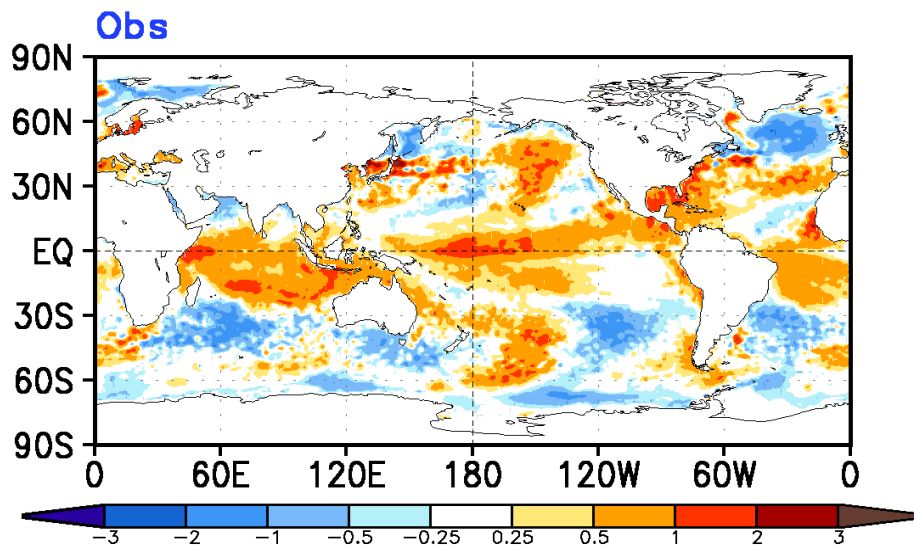
# Model Simulated/Forecast Ensemble Mean Anomalies



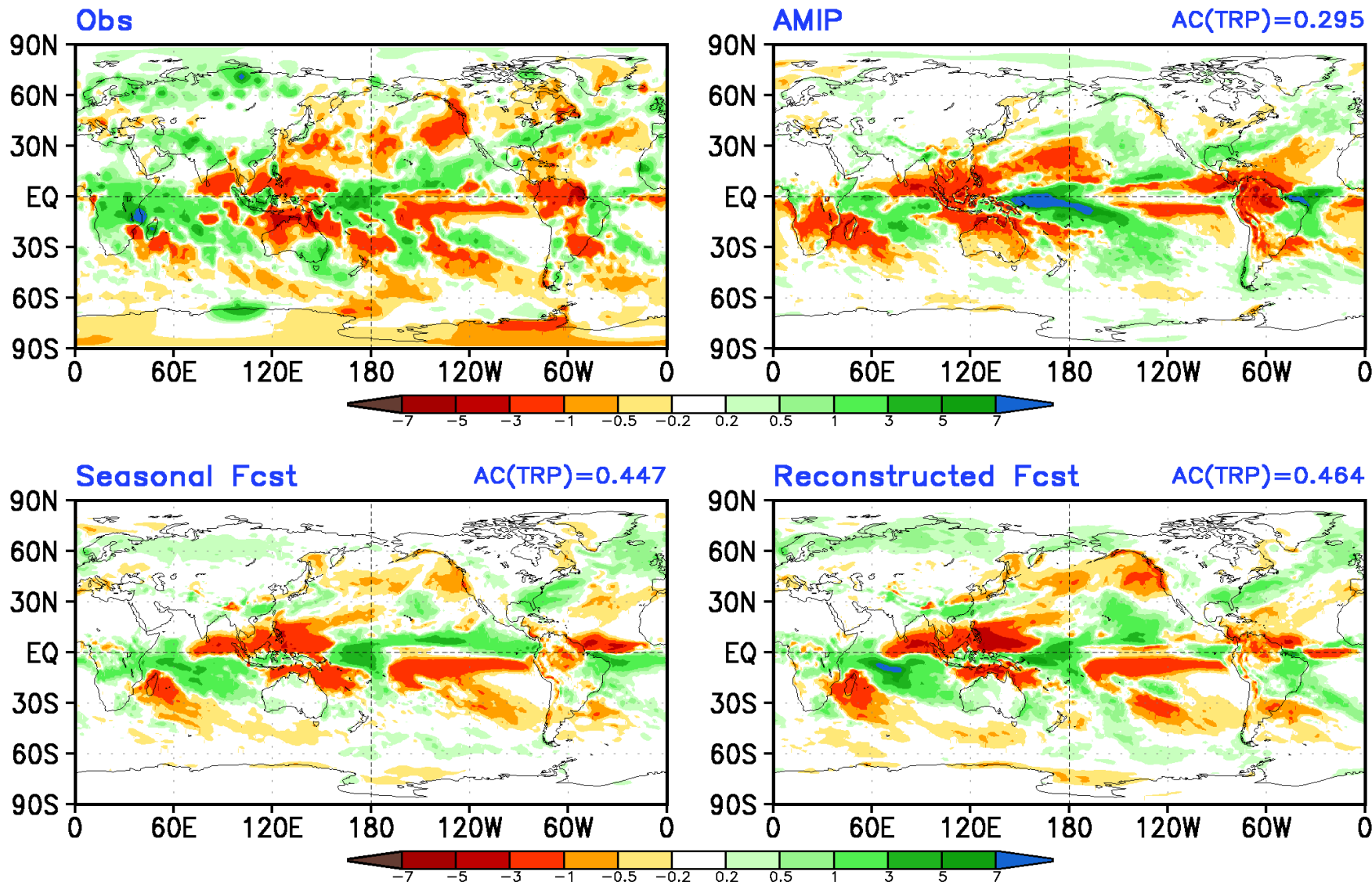
# Model Simulated/Forecast Ensemble Average Anomalies

- CFS **AMIP simulations** forced with observed sea surface temperatures (18 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).

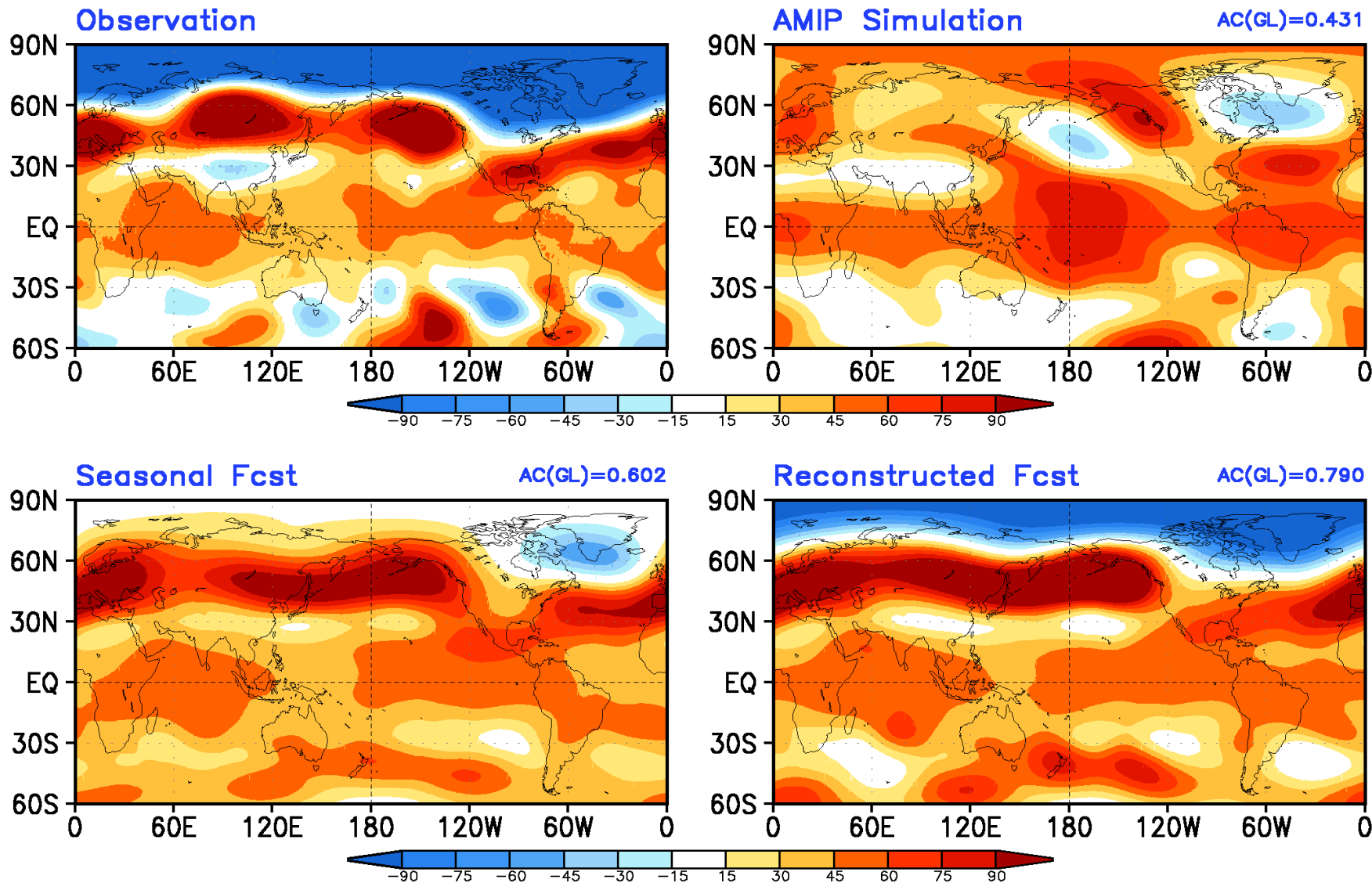
# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies SST(K)



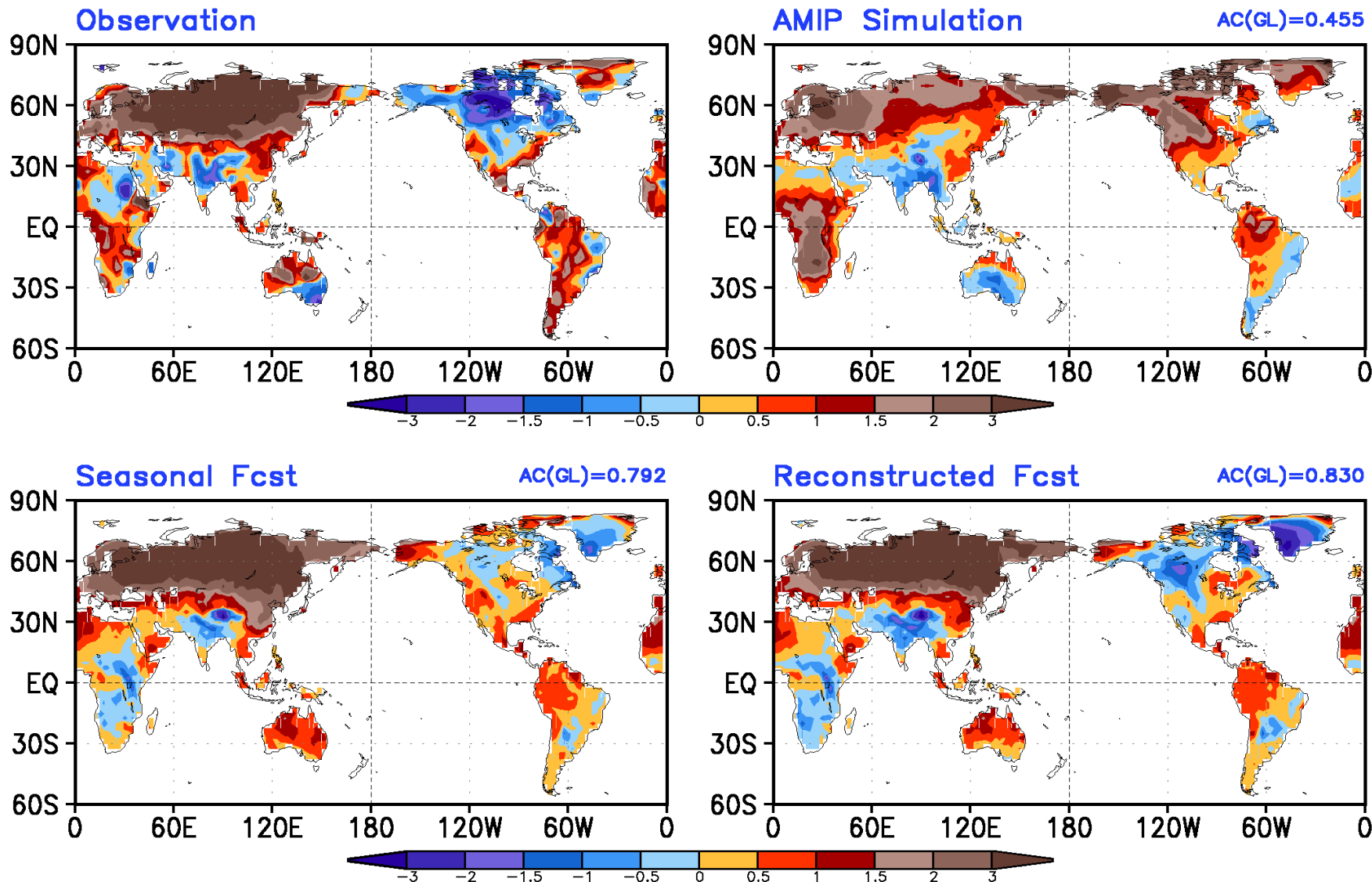
# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



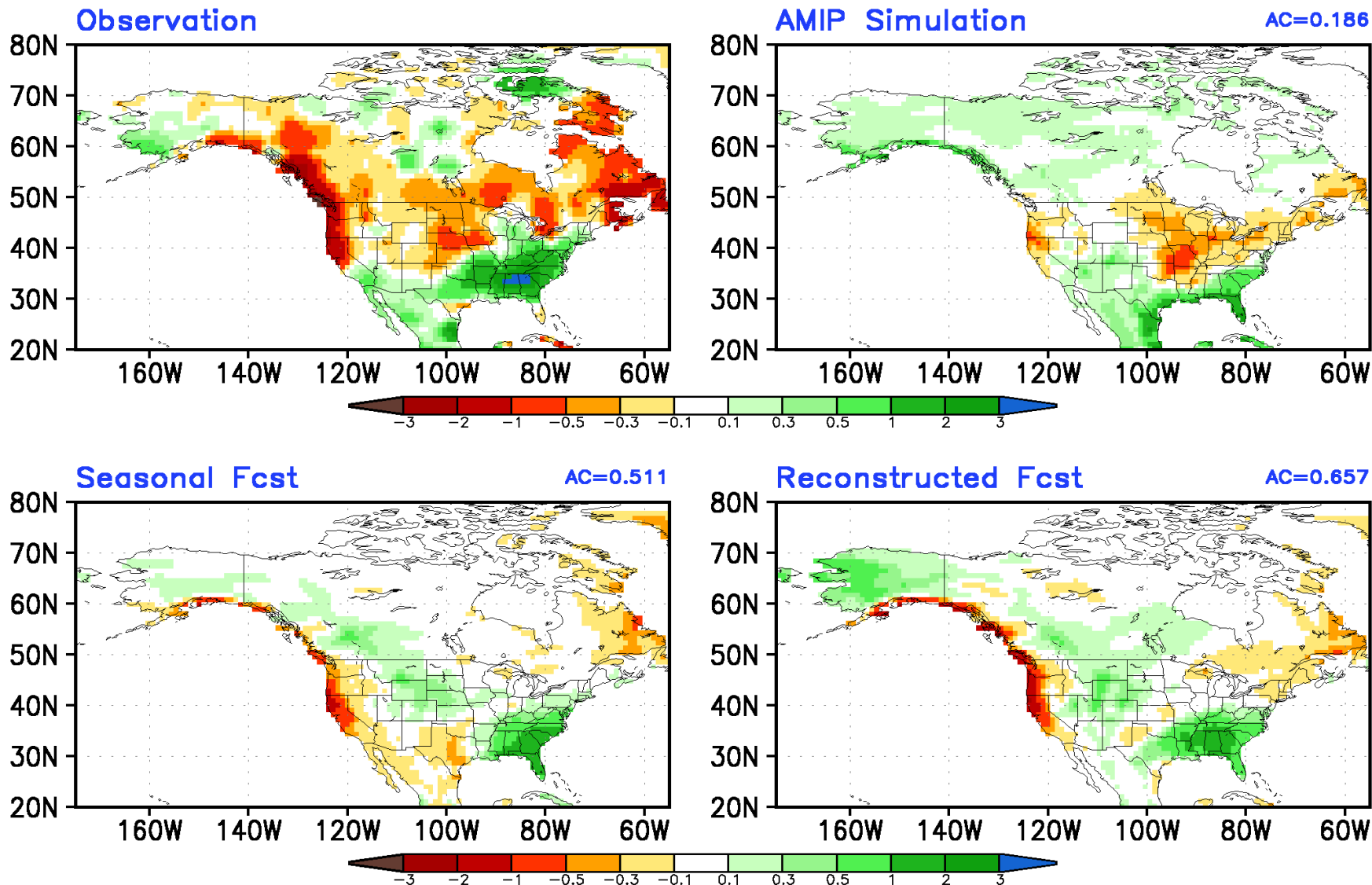
# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



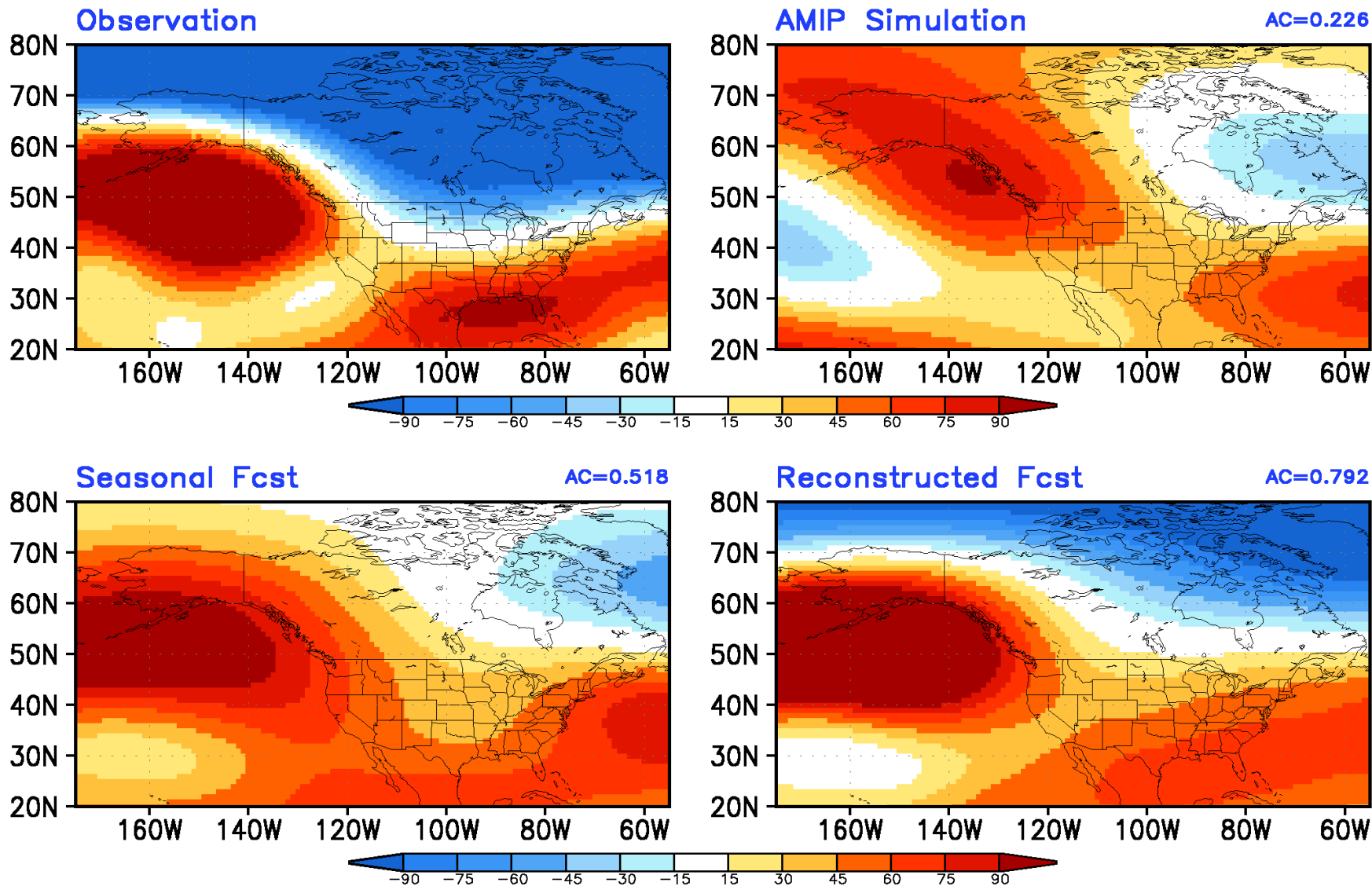
# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



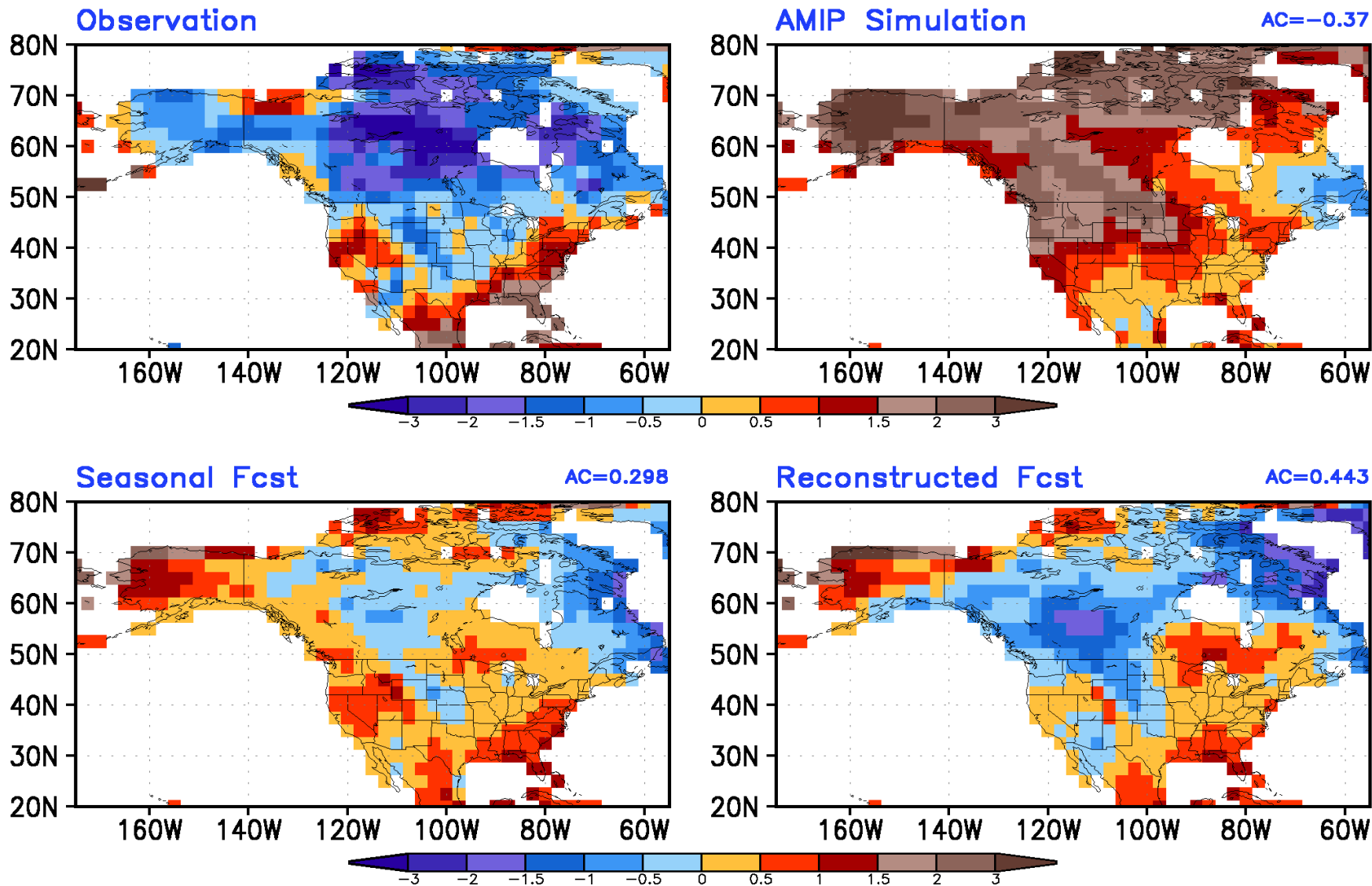
# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



# FMA2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



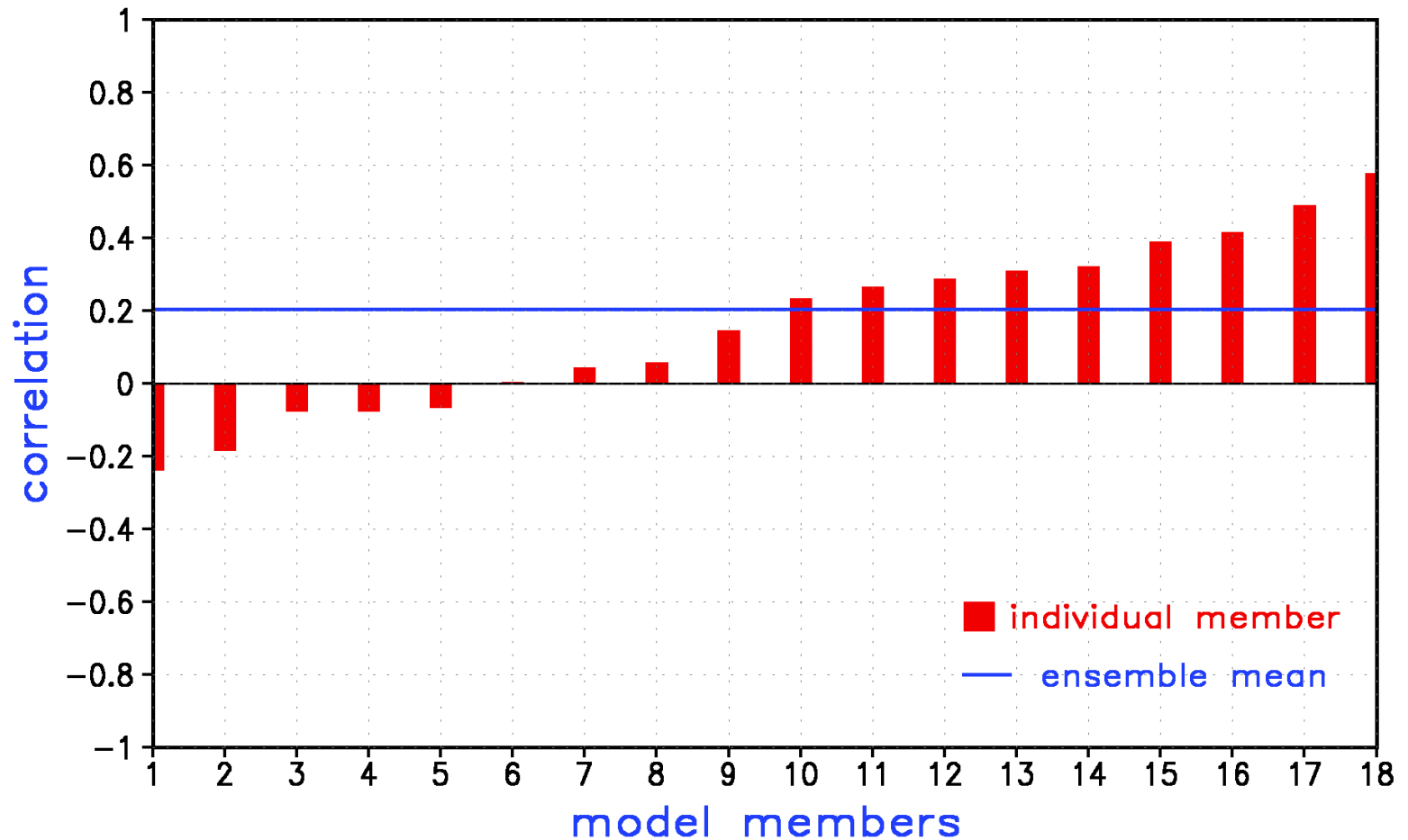


# Model Simulated/Forecast Anomalies: Individual Runs

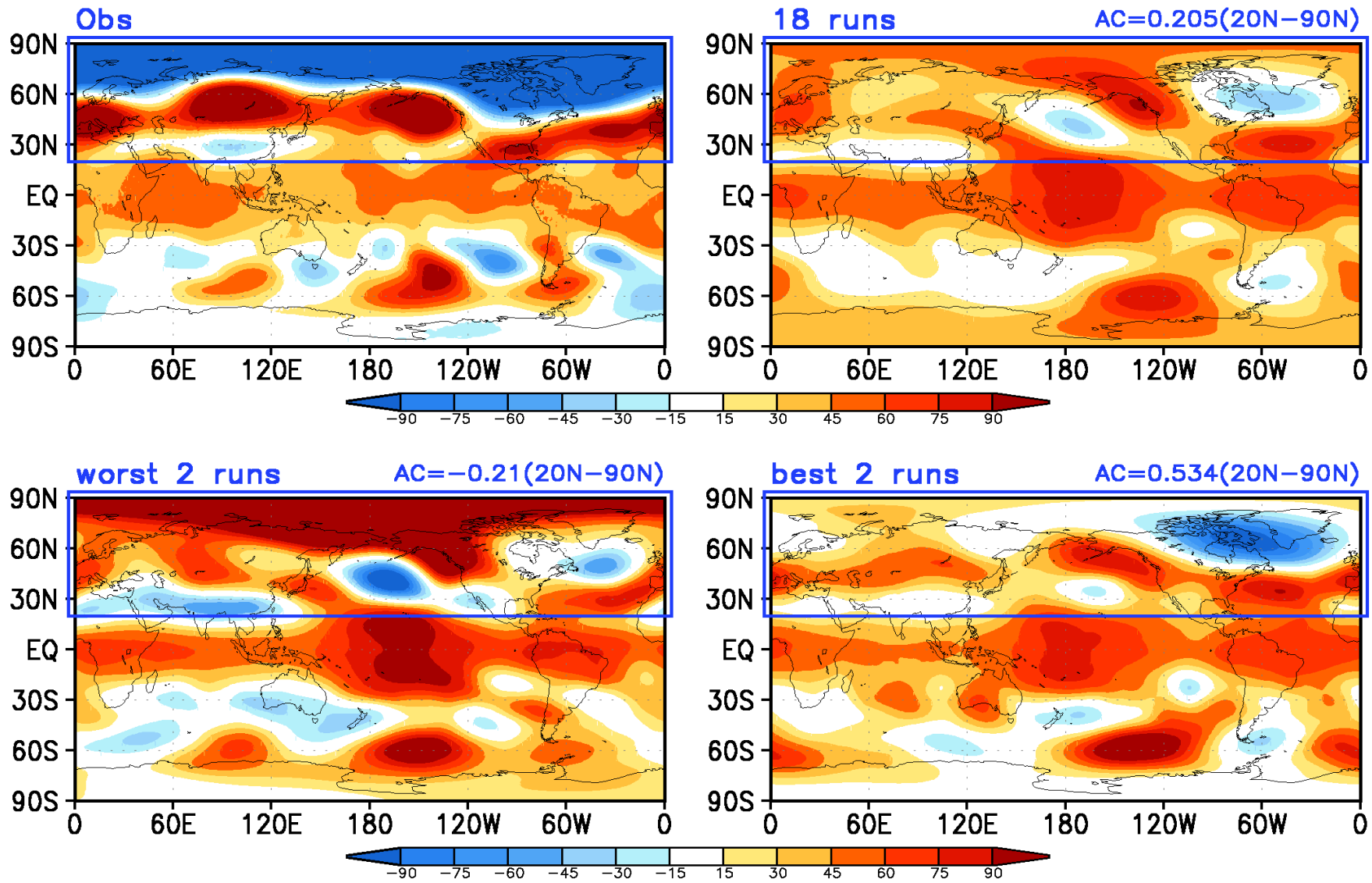
# Model Simulated/Forecast Anomalies: Individual Runs

- 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.

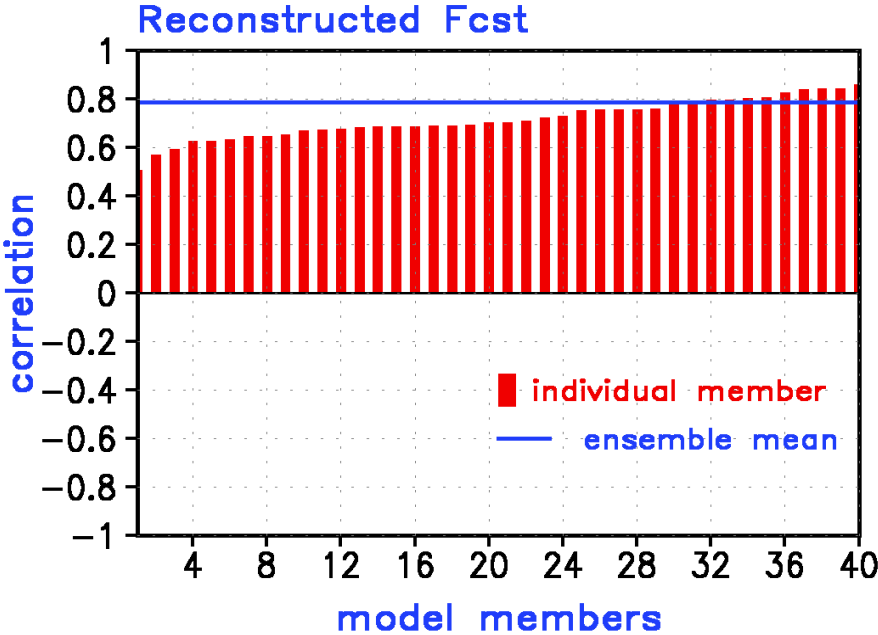
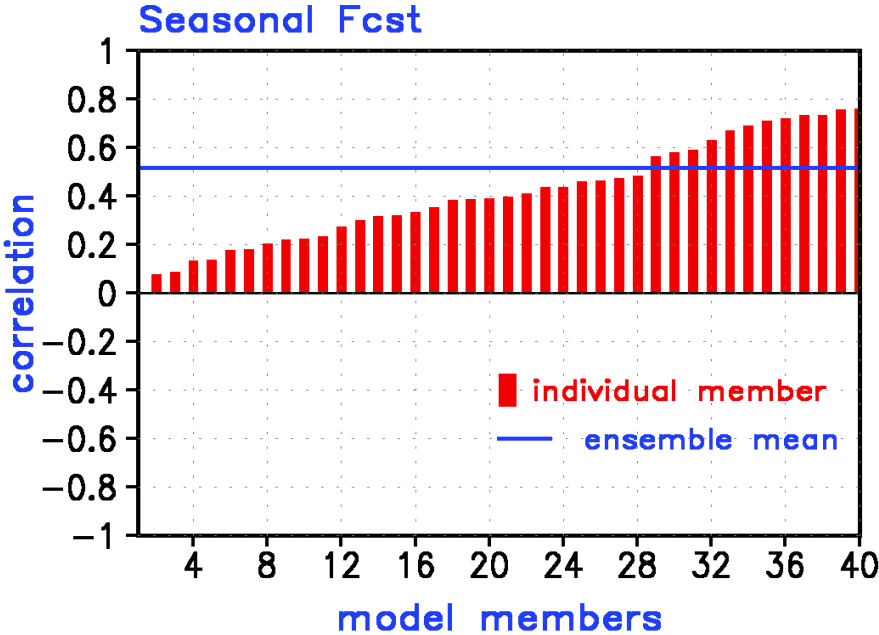
# FMA2020 Anomaly Correlation for Individual AMIP Simulation with Observation — z200(20N–90N)



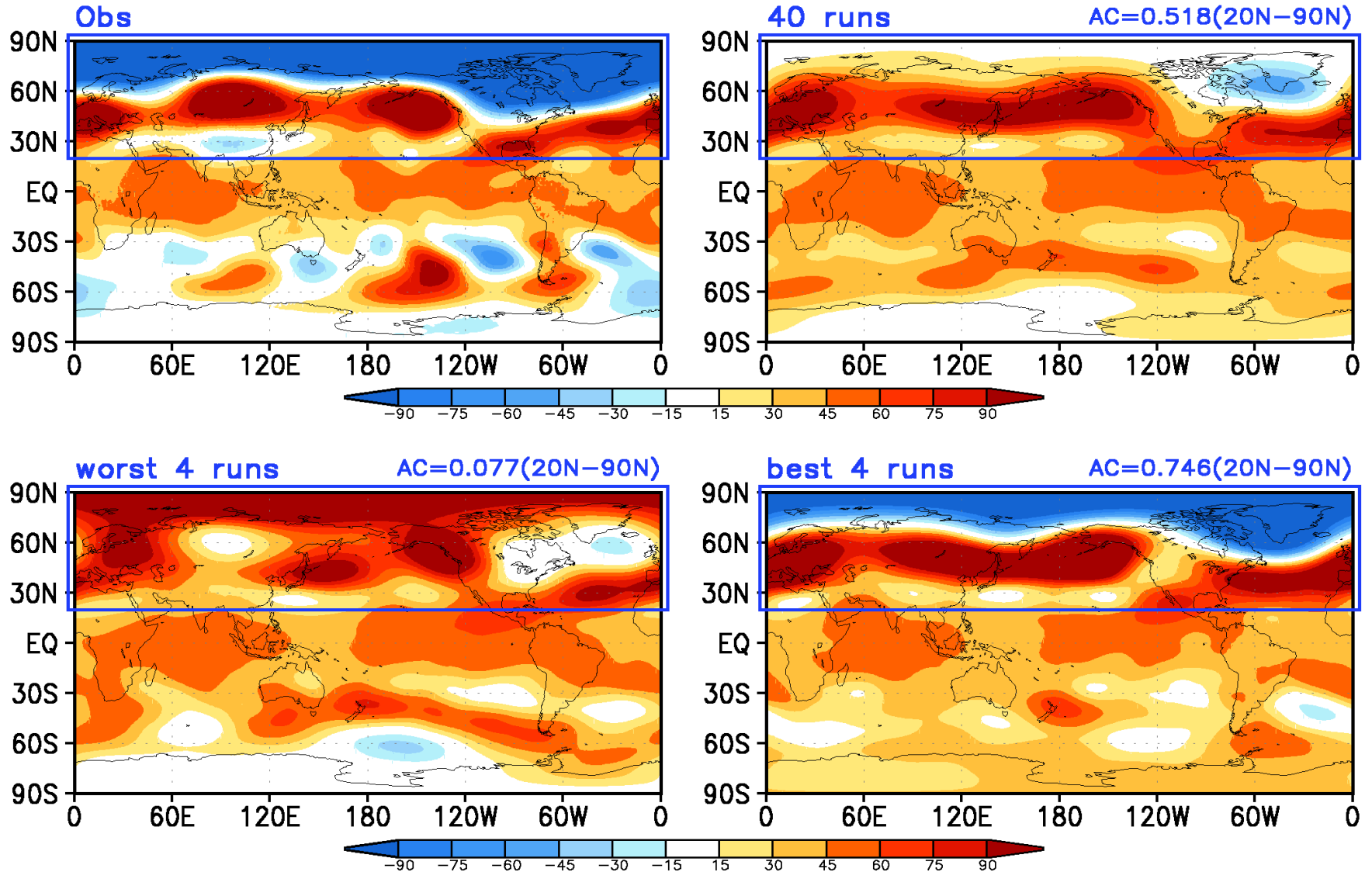
# Observed & AMIP Ensemble Average Anomalies FMA2020 z200(m) 18 runs/worst 2 runs/best 2 runs



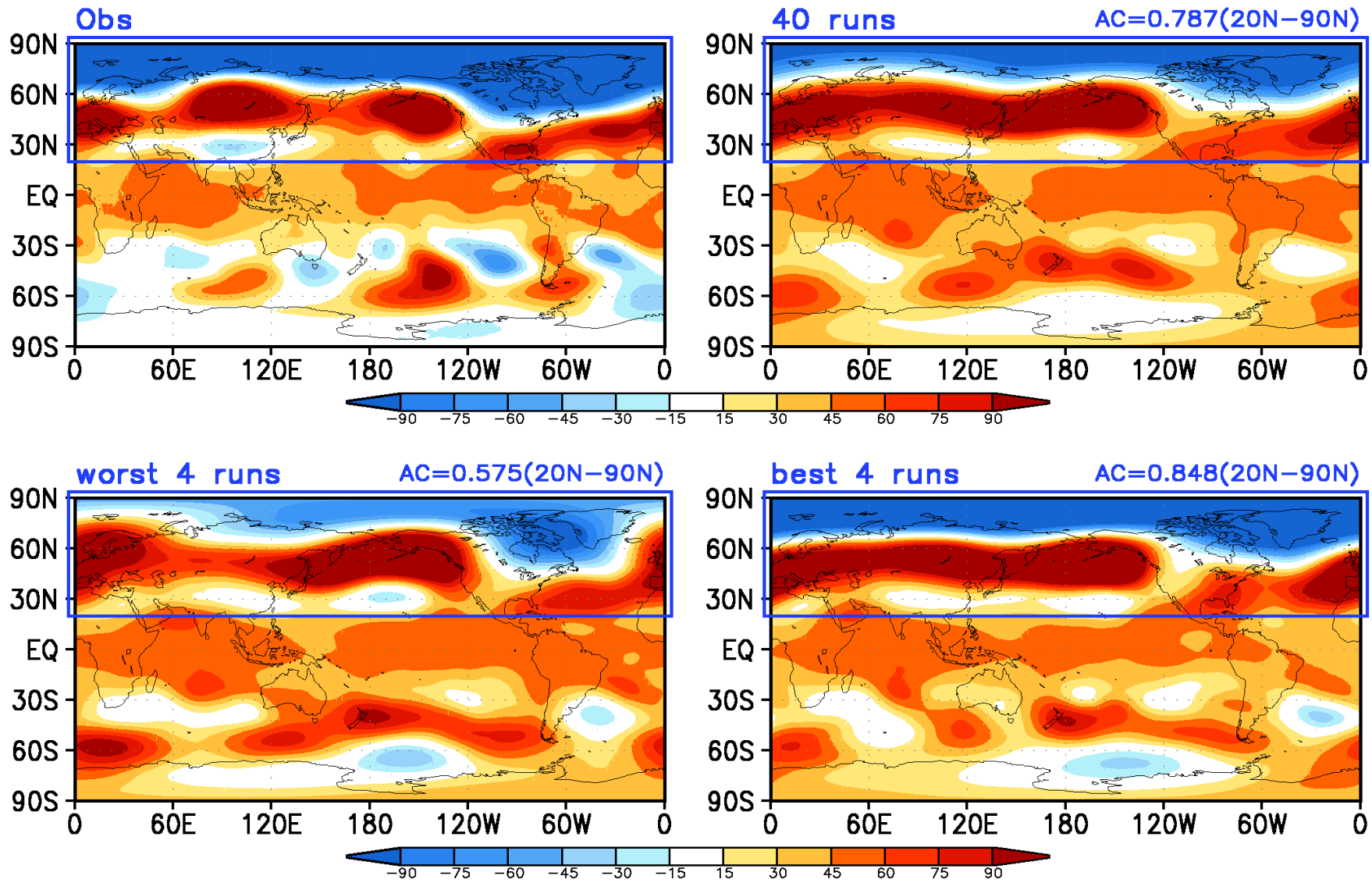
# FMA2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation — z200 (20N–90N)



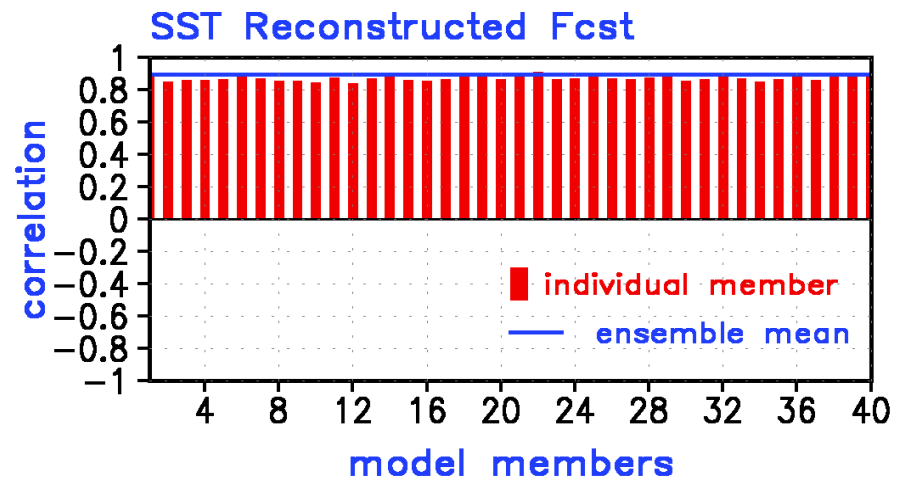
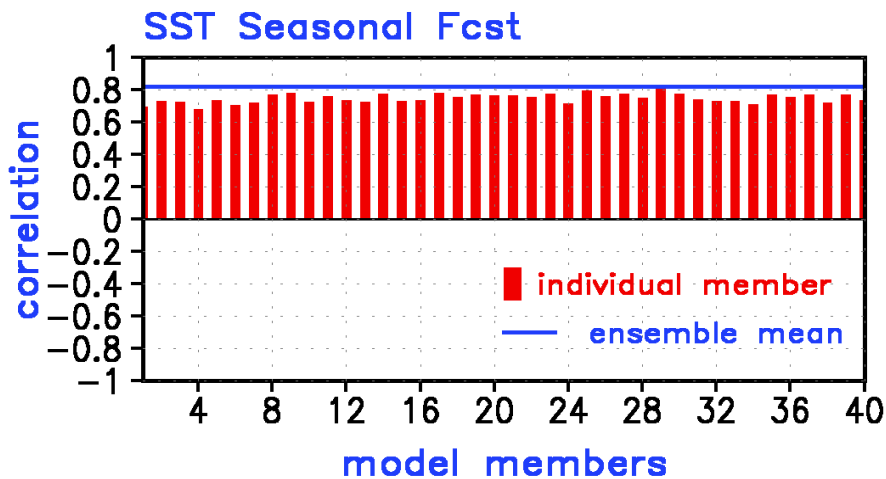
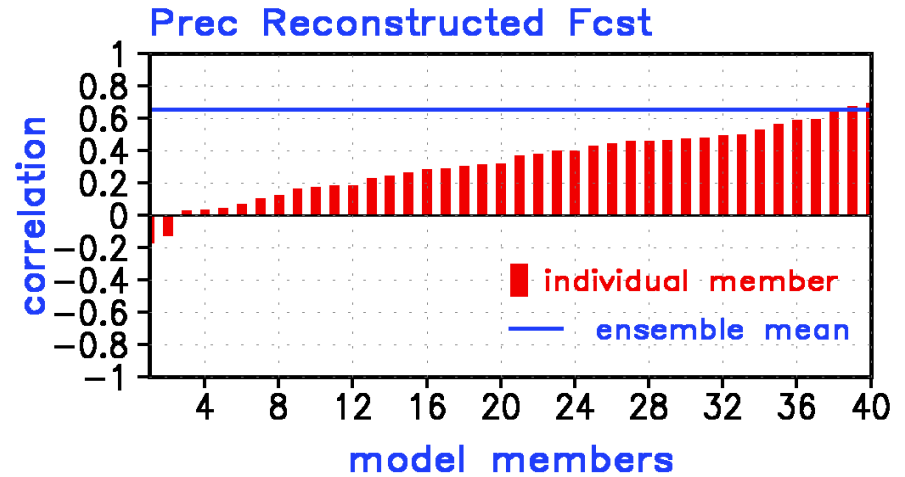
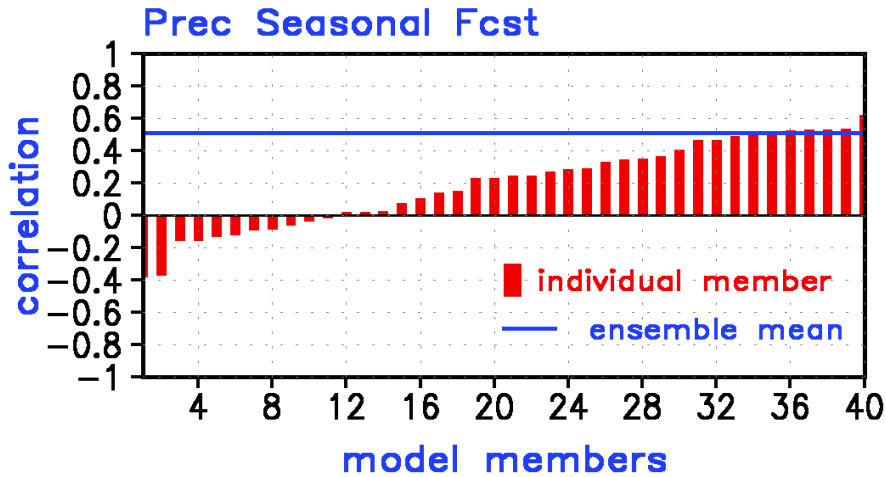
# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 z200(m) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 z200(m) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast

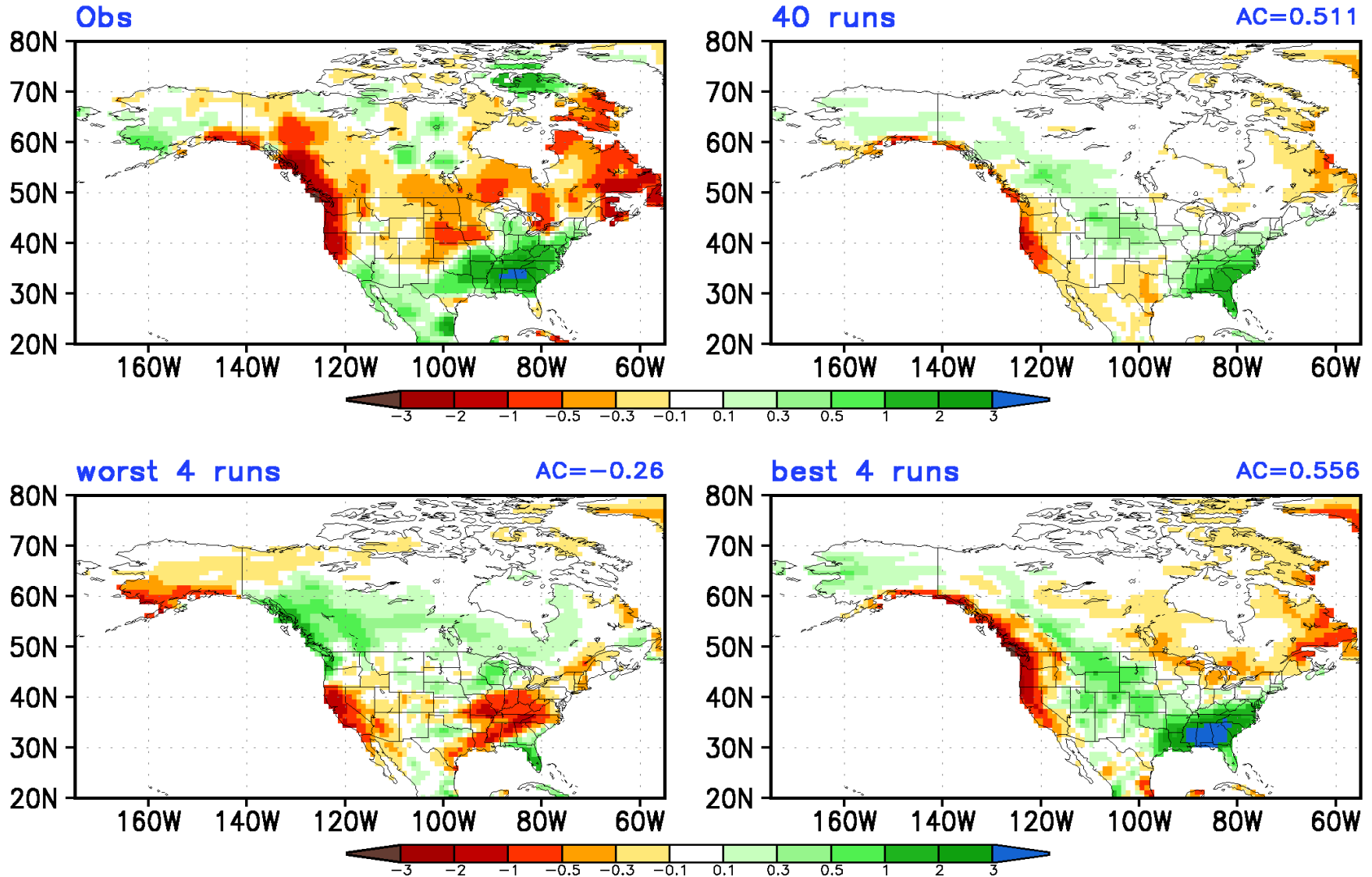


# FMA2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- Prec(NA)/SST(30S-30N)

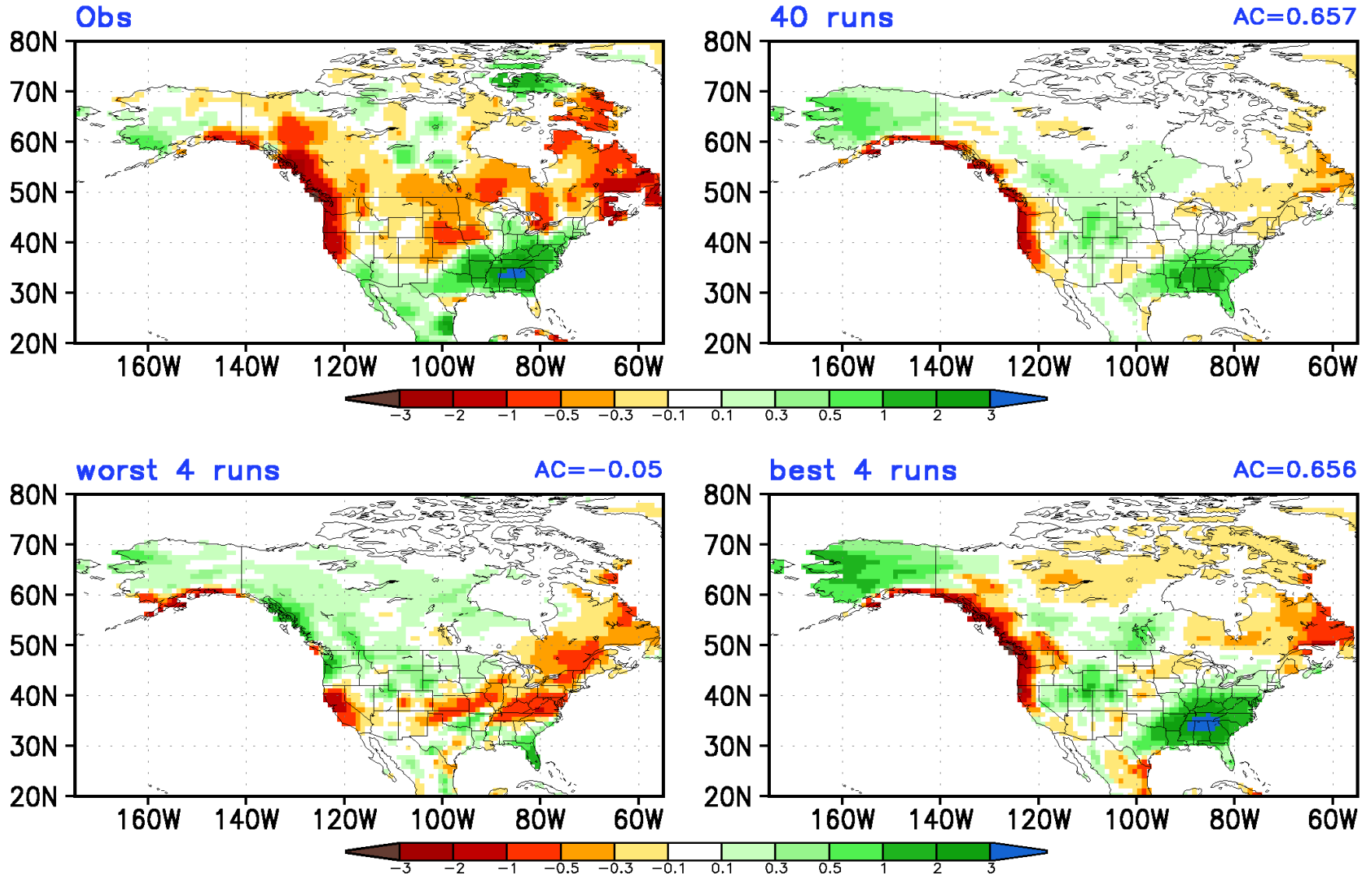




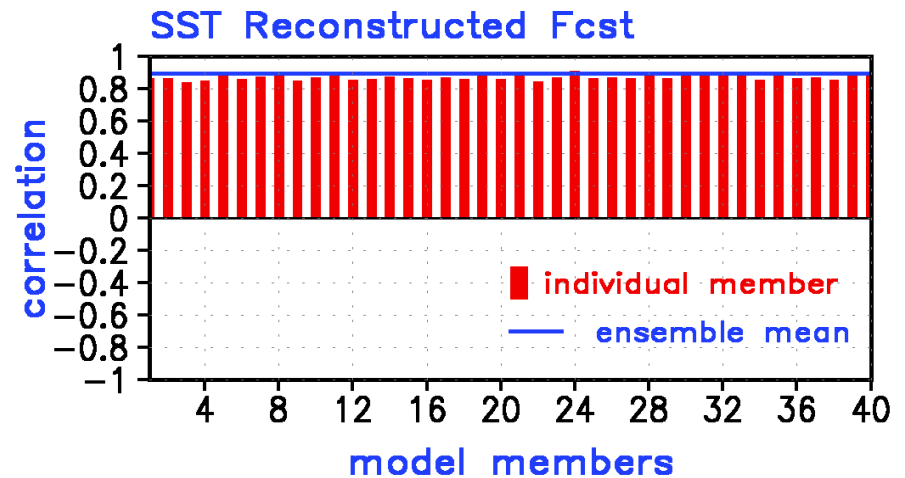
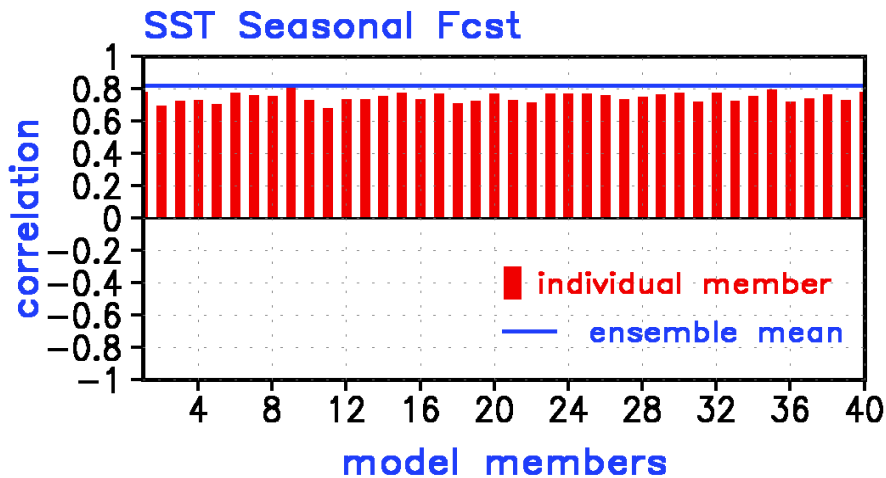
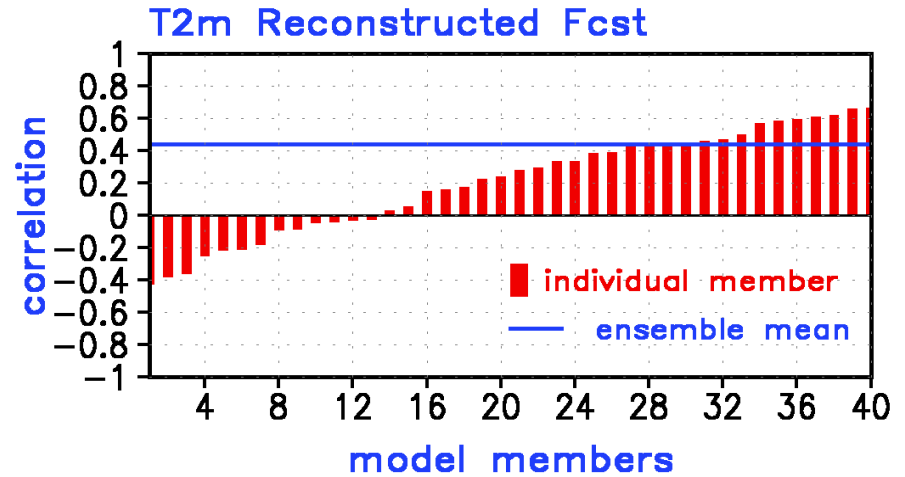
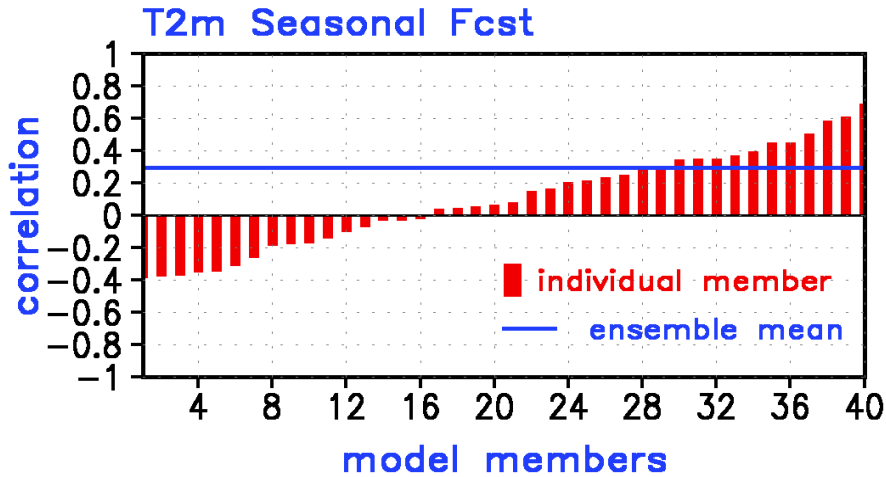
# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



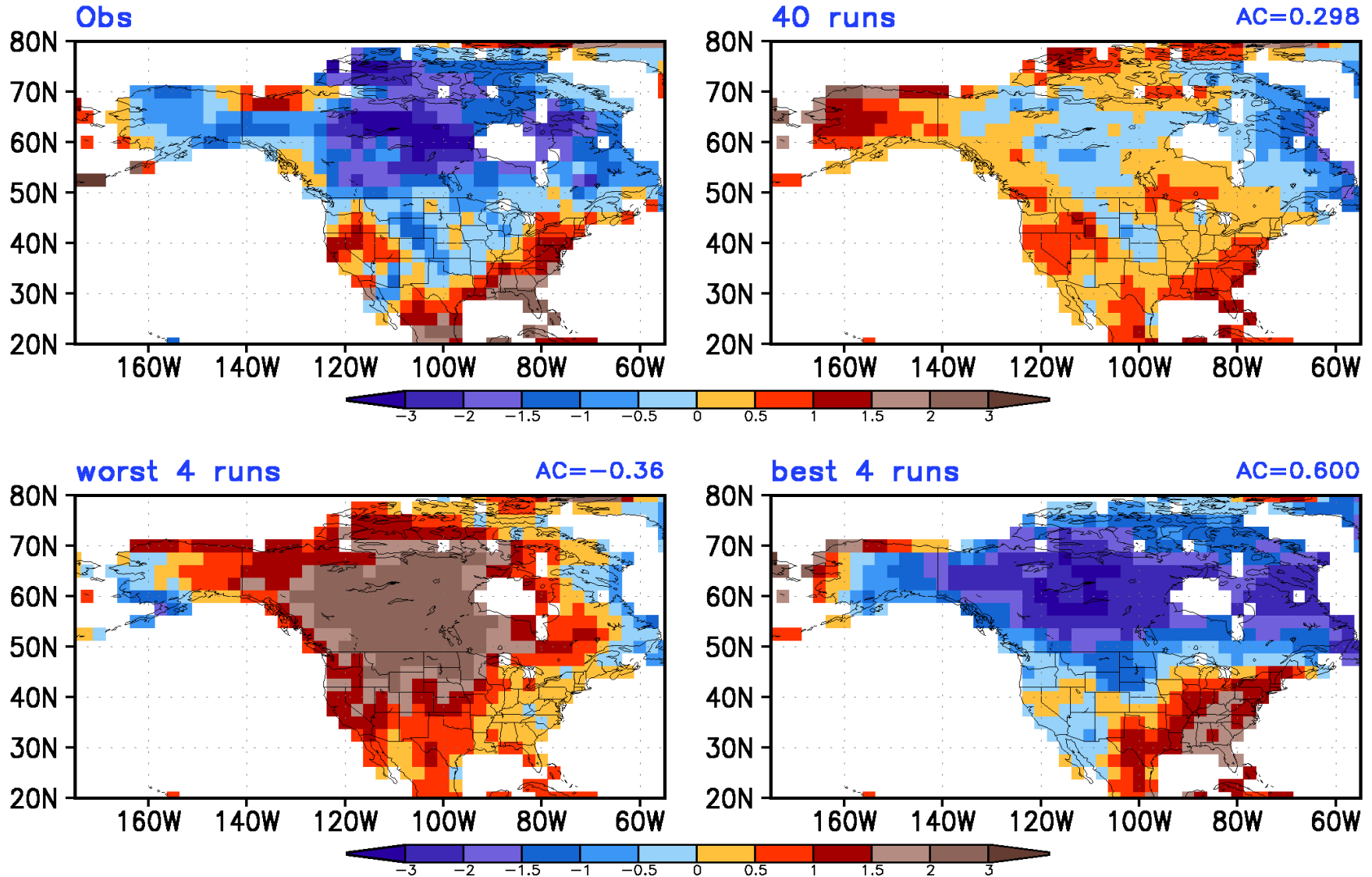
# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



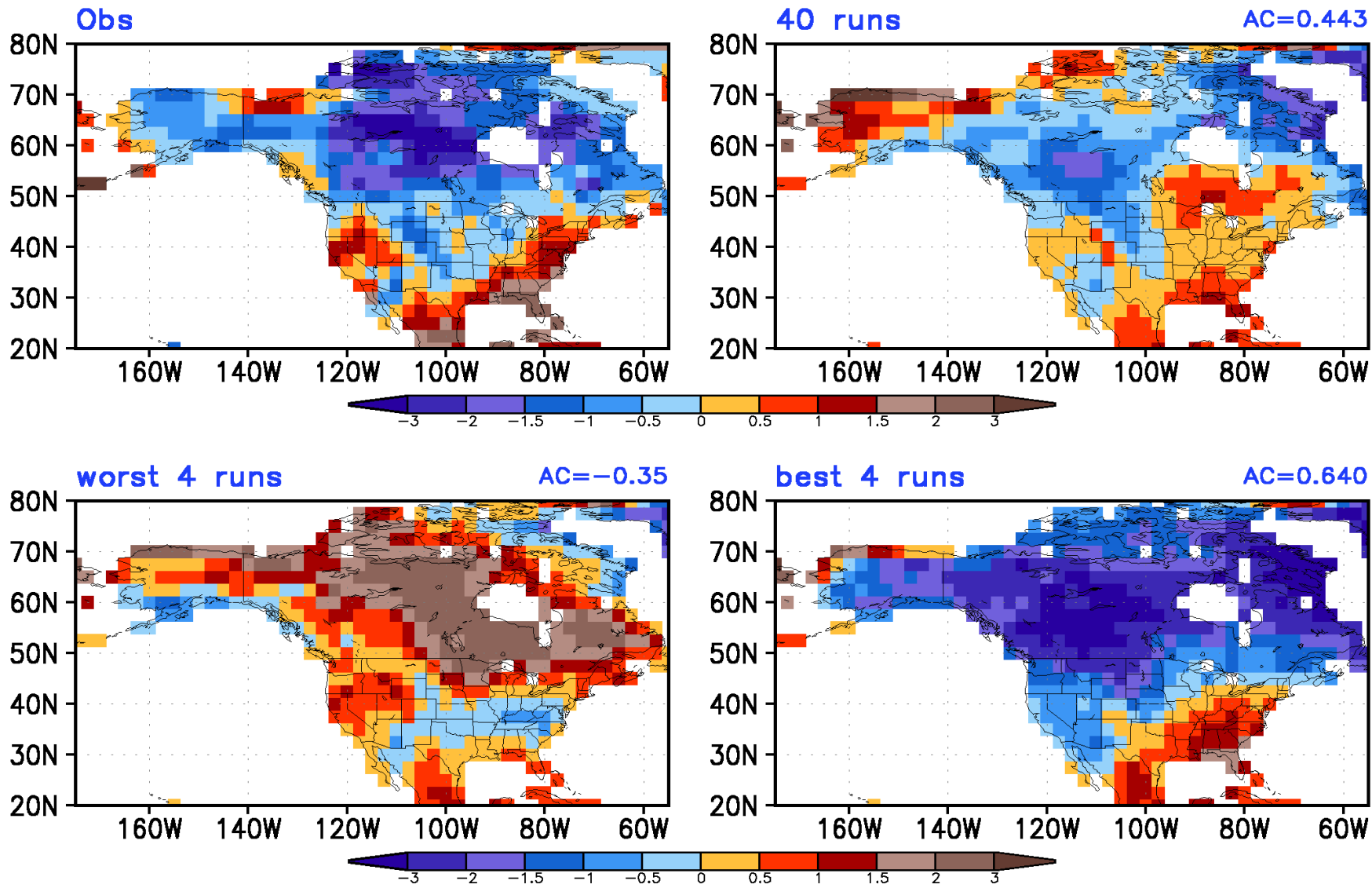
# FMA2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- T2m(NA)/SST(30S-30N)



# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 T2m(K) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast

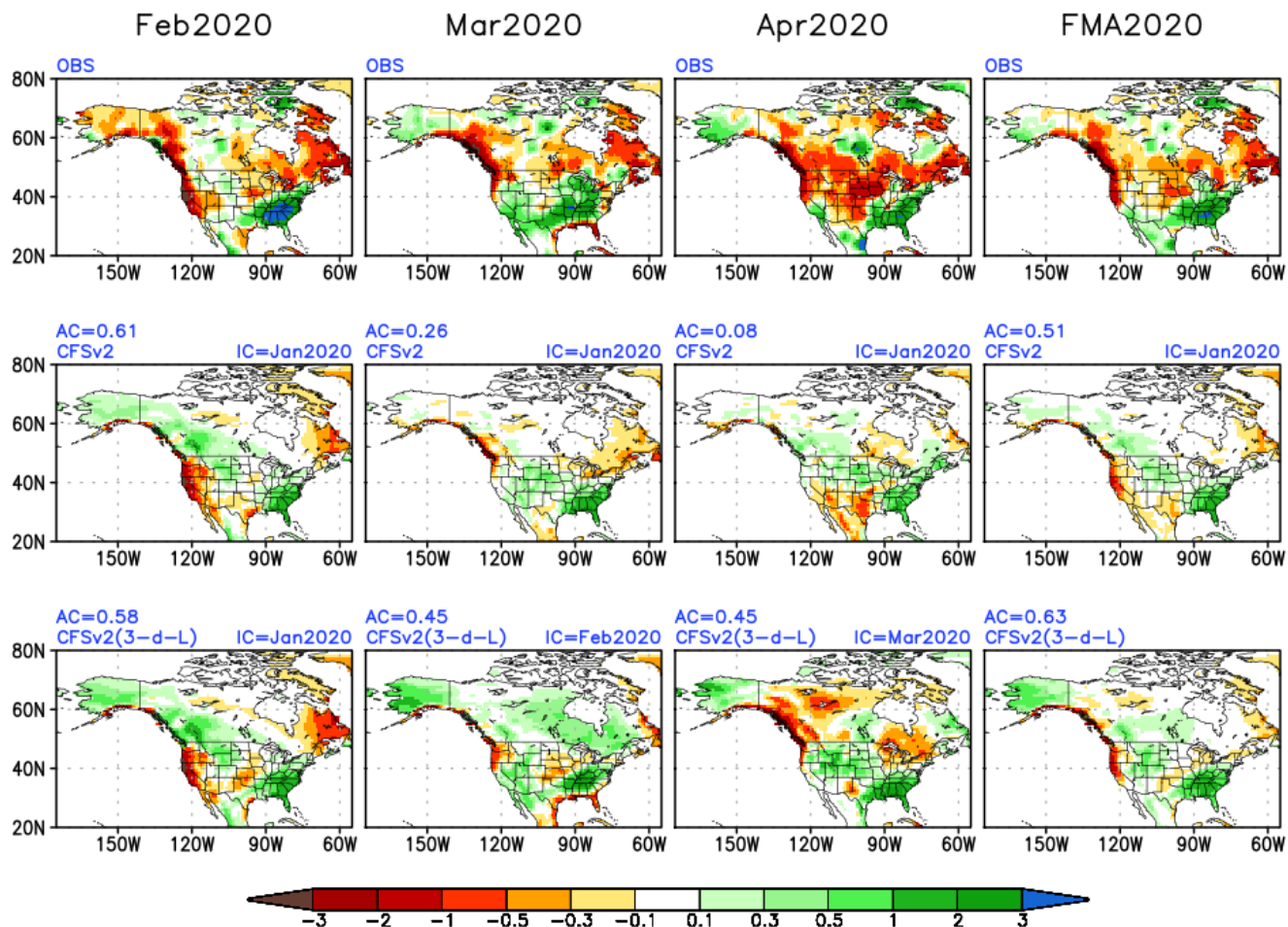


# Observed & CFSv2 Forecast Ensemble Average Anomalies FMA2020 T2m(K) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



# Monthly Means Prec(mm/day) Observed & Forecasts

## Monthly FMA2020 Prec(mm/day) Observed & Forecasts

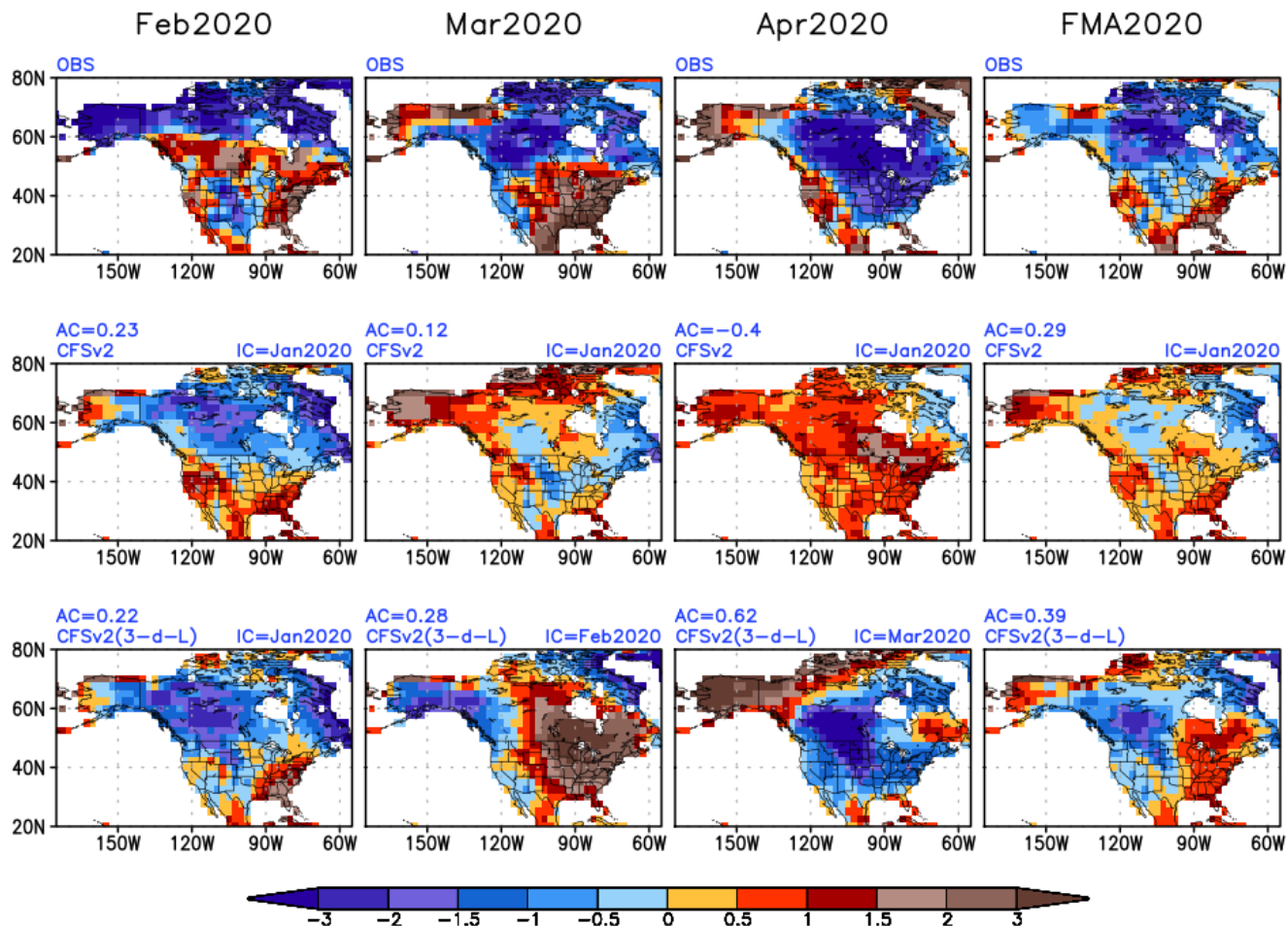


Middle row: CFSv2 seasonal forecasts from the initial conditions from the month prior to the target season.

Bottom row: CFSv2 monthly forecasts from the last three days of the month prior to the target month.

# Monthly Means T2m(K) Observed & Forecasts

## Monthly FMA2020 T2m(K) Observed & Forecasts



Middle row: CFSv2 seasonal forecasts from the initial conditions from the month prior to the target season.

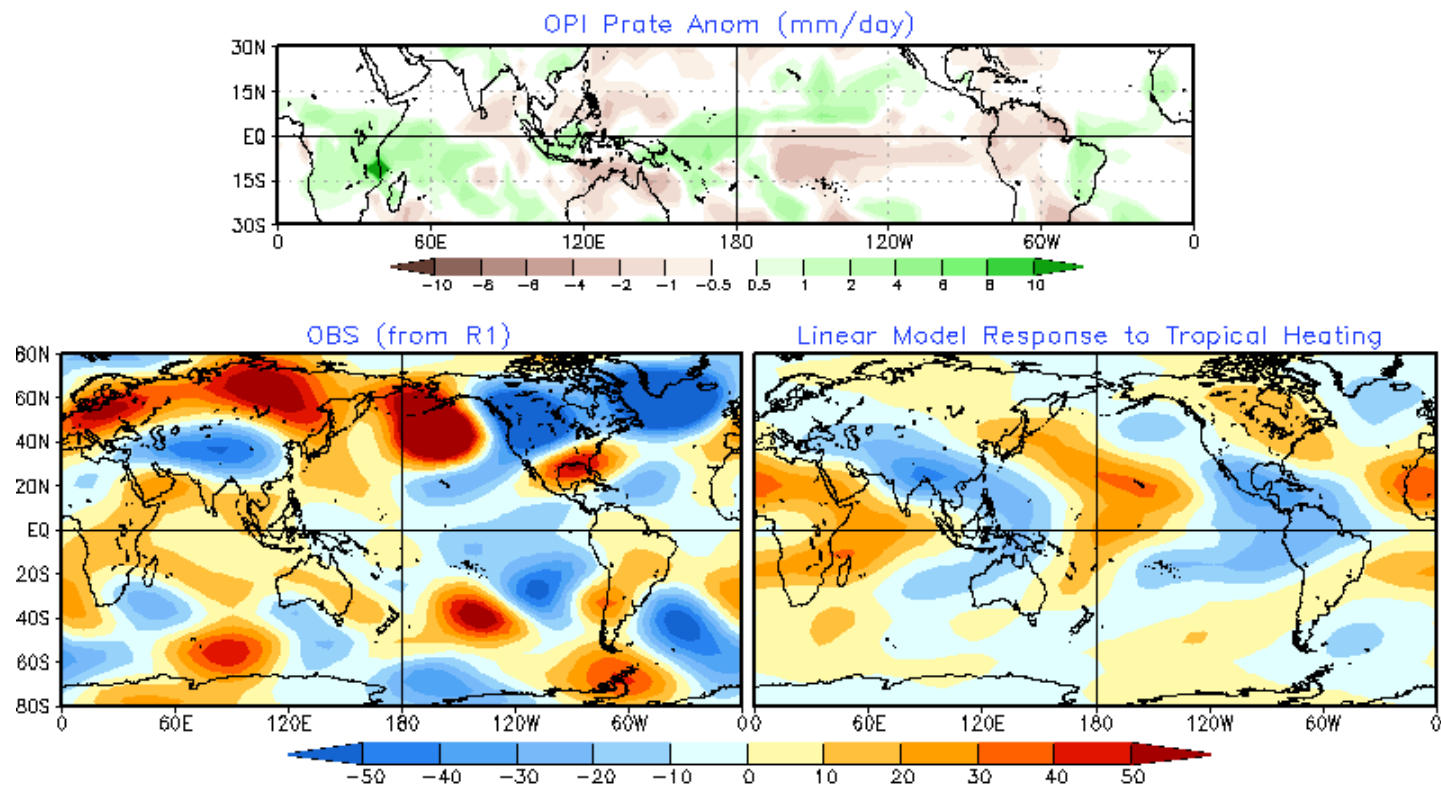
Bottom row: CFSv2 monthly forecasts from the last three days of the month prior to the target month.

# 200mb Height from Linear Model



# 200mb Height from Linear Model

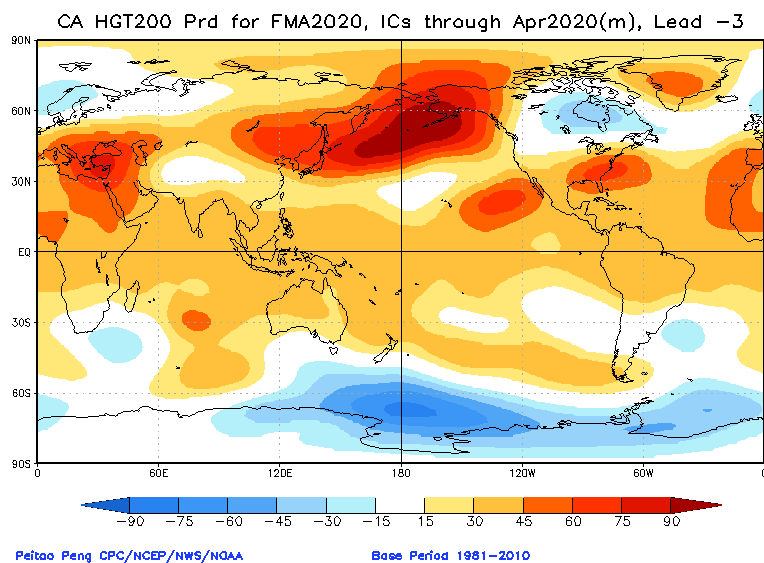
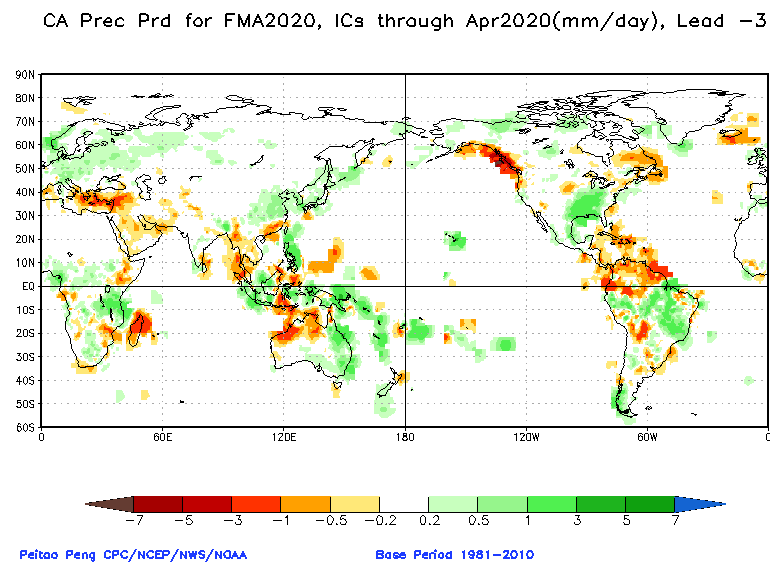
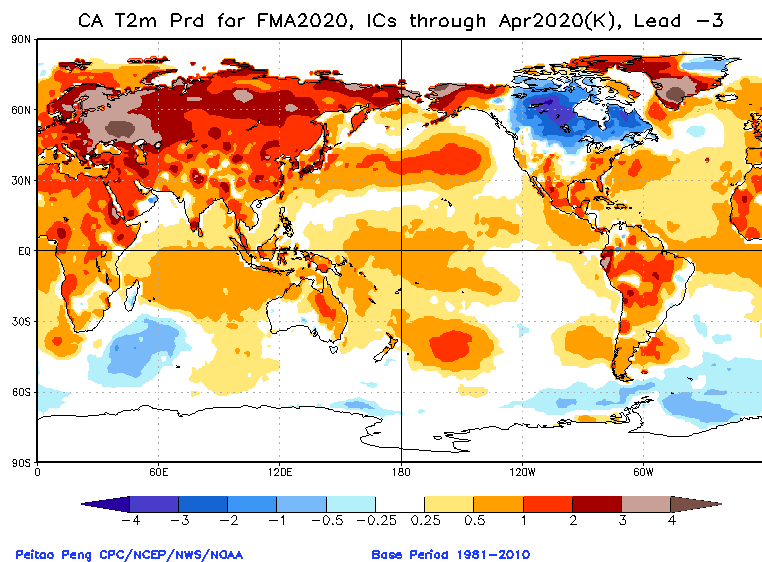
FMA2020 200mb Eddy HGT(m)  
OBS vs. Linear Model Response to Tropical Heating  
Heating is converted from Prate in 15S-15N



Pattern COR: global=0.10, tropics(30S-30N)=0.16

# Seasonal Forecasts from the Constructed Analog Model

# Seasonal Forecasts from the Constructed Analog Model



# Seasonal Forecasts from WMO Lead Center for Long-Range Forecast Multi-Model Ensemble (LC-LRFMME)

<https://www.wmolc.org/>

- LC-LRFMME seasonal forecast are based on forecasts provided by WMO recognized Global Producing Centers (GPCs) for Long-Range Forecasts to the LC-LRFMME. Contribution of all GPCs is acknowledged.
- Seasonal forecasts from GPCs are merged into a multi-model ensemble forecast.
- LC-LRFMME forecasts are based on GPC seasonal forecast systems run during the first week of the month for the next season. For example, forecasts runs in first week of January for the seasonal mean of February-March-April.
- Forecasts in slides 42-45 are from the Lead Center.
- For latest seasonal outlook guidance see <http://www.wmo.int/pages/prog/wcp/wcasp/LC-LRFMME/index.php>
- *For more information see visit Lead Center website; also see Graham, R., and Co-authors, 2011: New perspectives for GPCs, their role in the GFCS and a proposed contribution to a 'World Climate Watch'. Climate Research, 47, 47-55.*

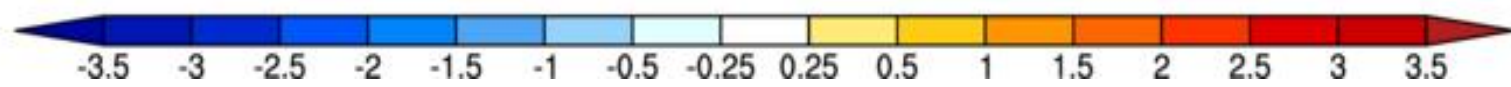
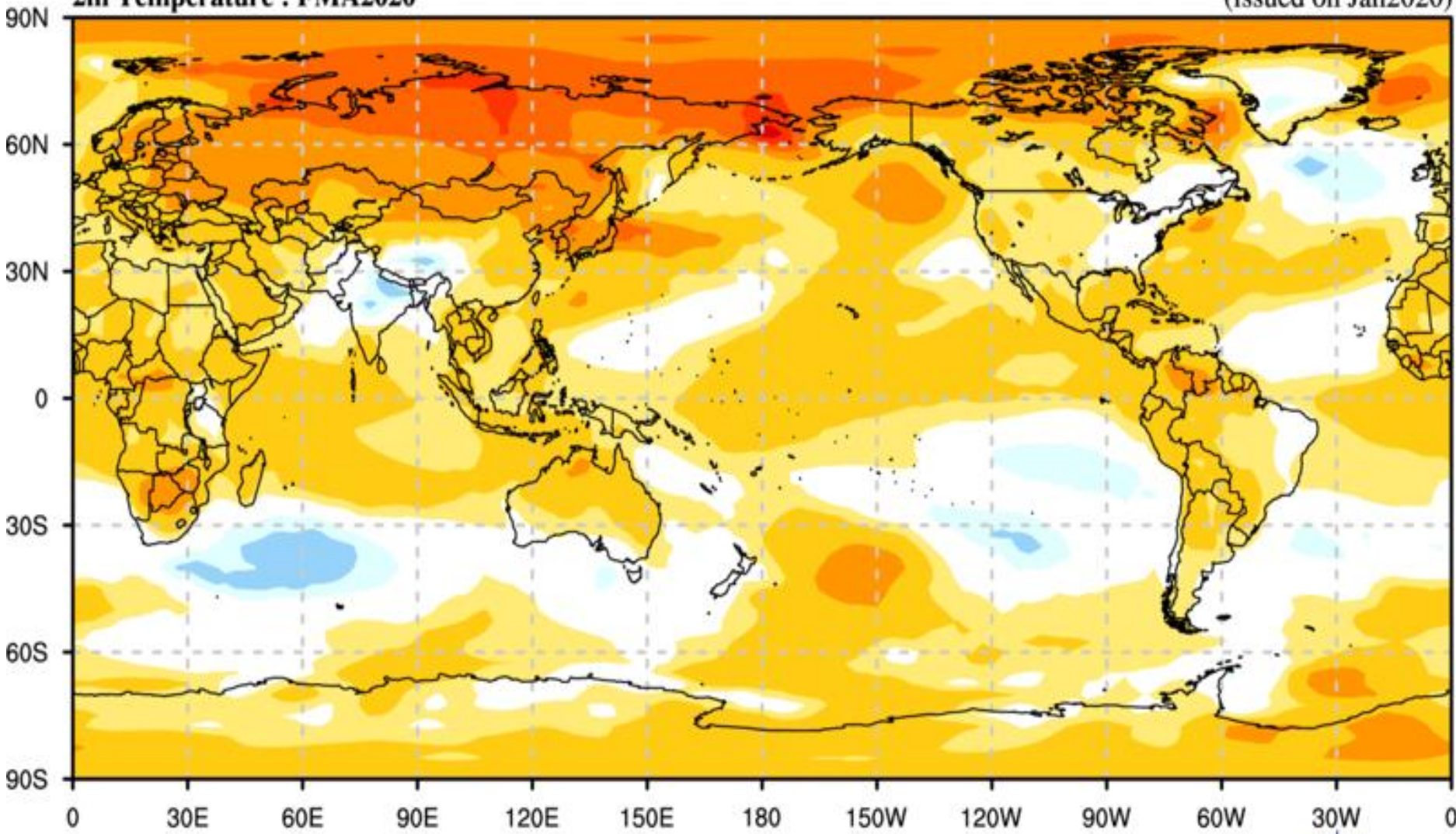
# Simple Composite Map

CPTEC,ECMWF,Melbourne,Montreal,Moscow,Offenbach,Pretoria,Seoul,Tokyo,Toulouse,Washington

[Unit : K]

## 2m Temperature : FMA2020

(issued on Jan2020)



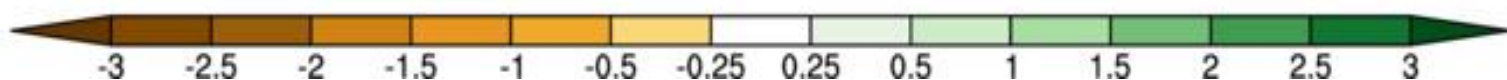
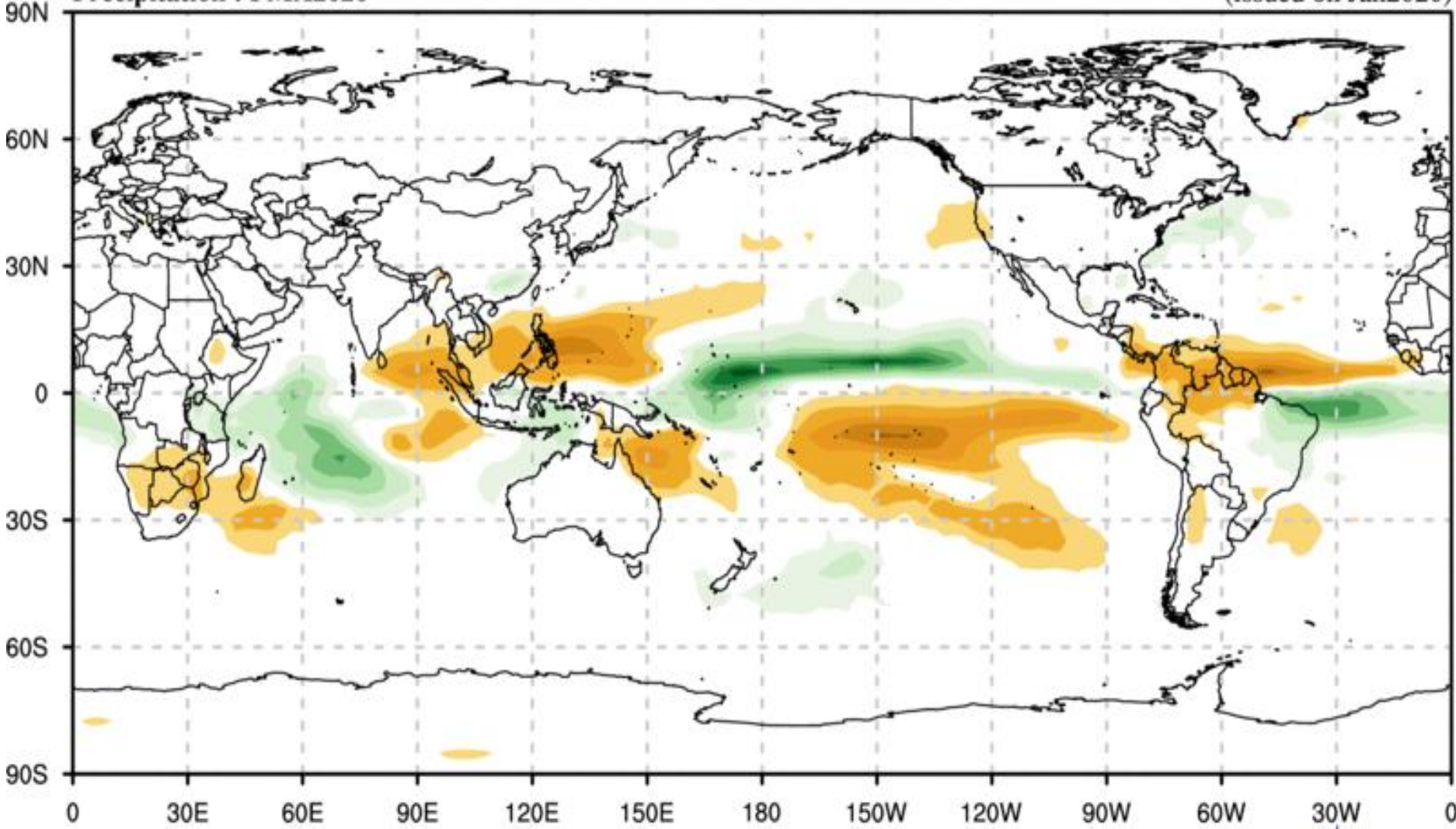
# Simple Composite Map

CPTEC,ECMWF,Melbourne,Montreal,Moscow,Offenbach,Pretoria,Seoul,Tokyo,Toulouse,Washington

[Unit : mm]

(issued on Jan2020)

## Precipitation : FMA2020

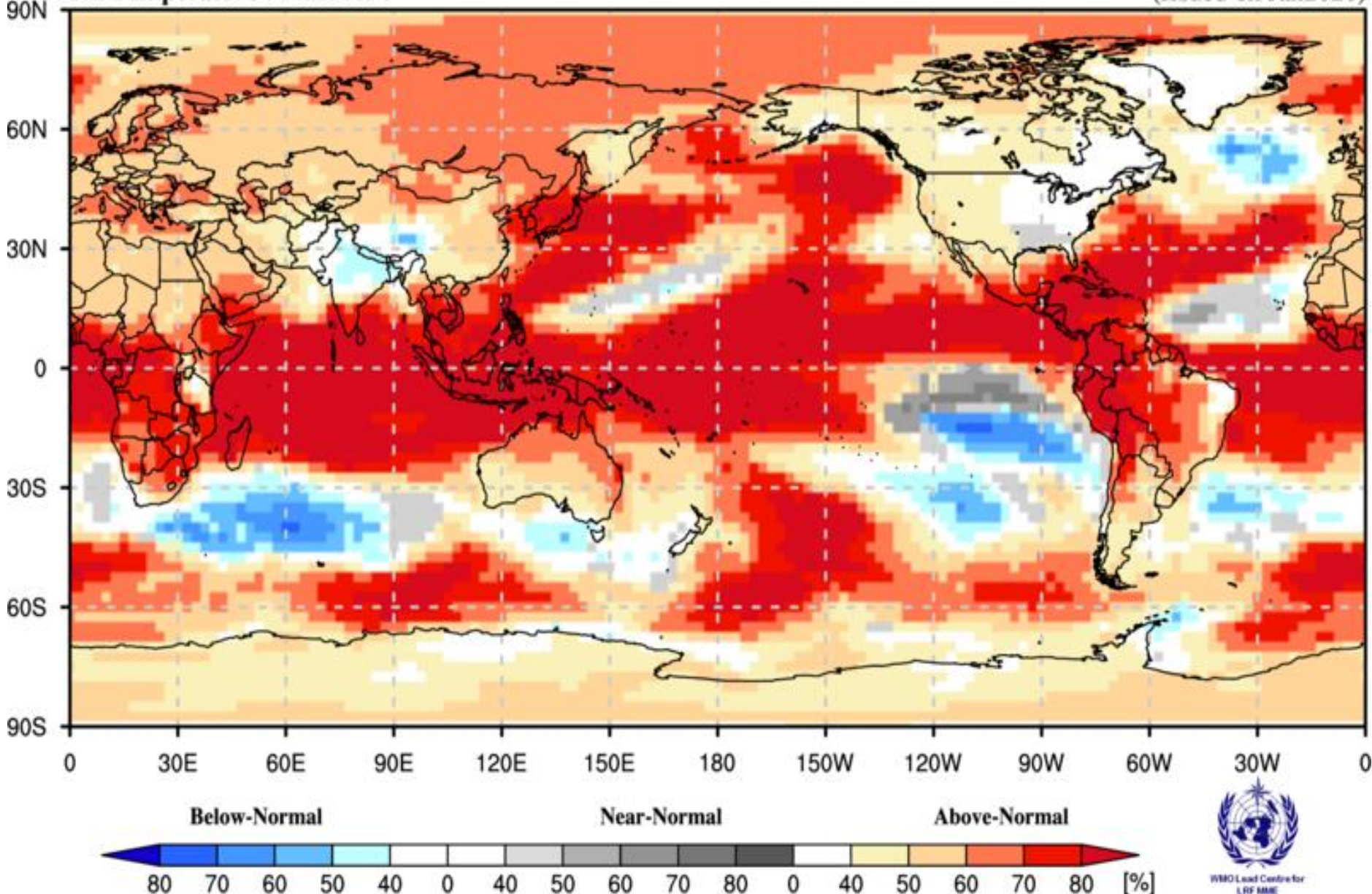


# Probabilistic Multi-Model Ensemble Forecast

CPTEC,ECMWF,Exeter,Melbourne,Montreal,Moscow,Offenbach,Pretoria,Seoul,Tokyo,Toulouse,Washington

## 2m Temperature : FMA2020

(issued on Jan2020)

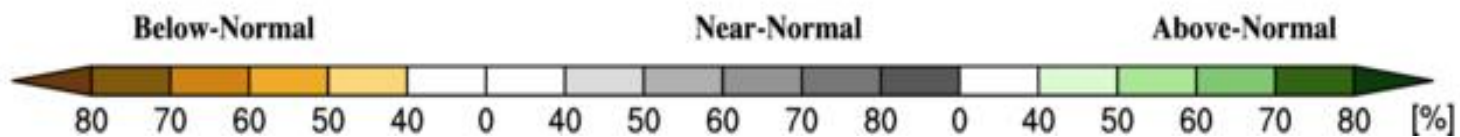
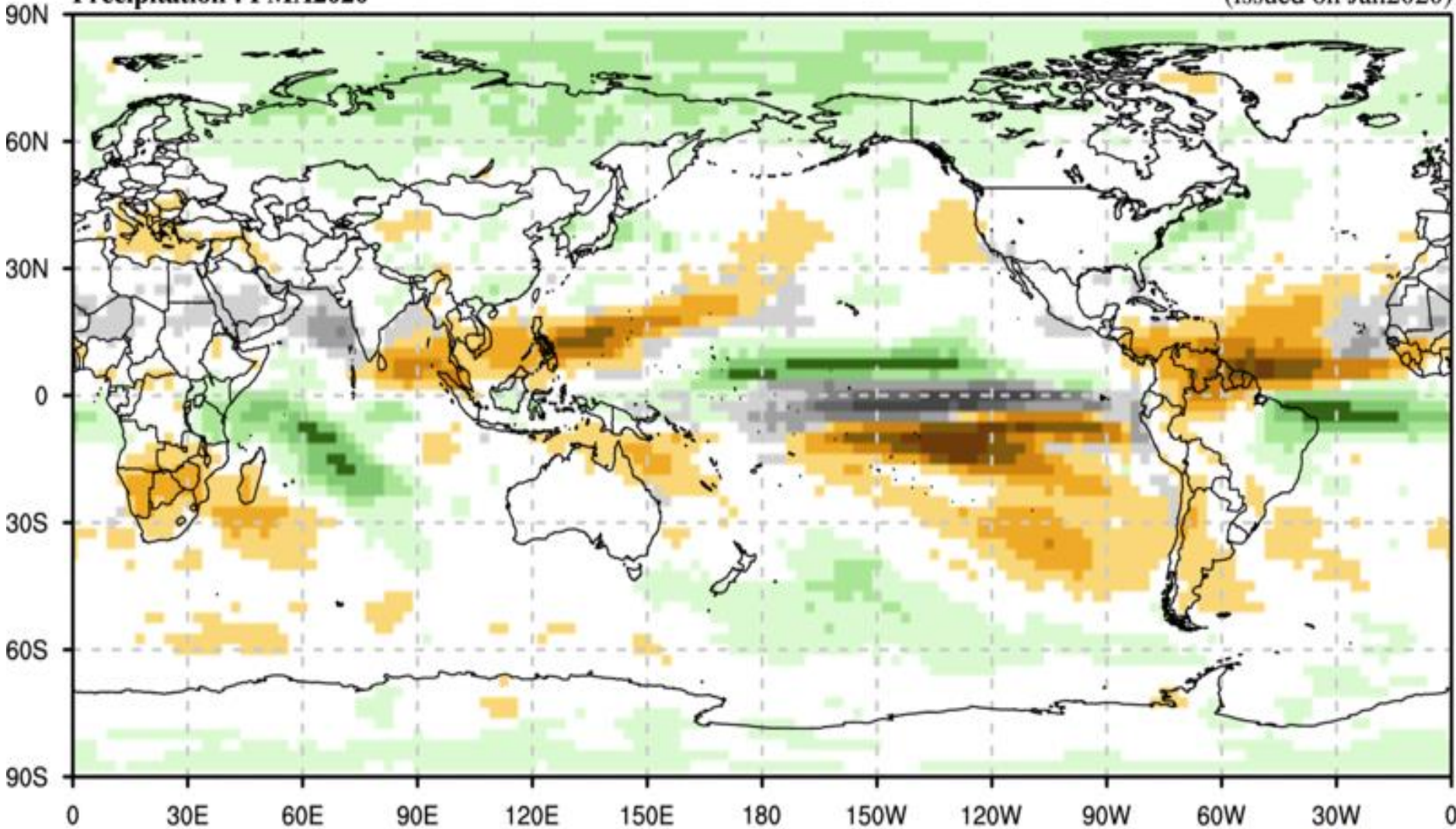


# Probabilistic Multi-Model Ensemble Forecast

CPTEC,ECMWF,Exeter,Melbourne,Montreal,Moscow,Offenbach,Pretoria,Seoul,Tokyo,Toulouse,Washington

## Precipitation : FMA2020

(issued on Jan2020)





# References

- Fan, Y., and Dool H. van den Dool (2008), A global monthly land surface air temperature analysis for 1948-present. *J. Geophys. Res.*, 113, D01103. [doi:10.1029/2007JD008470](https://doi.org/10.1029/2007JD008470).
- 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).
- Reynolds, R. W. et al (2007), Daily high resolution-blended analyses for sea surface temperature. *J. Clim.*, 20, 5473-5496. [doi:10.1175/2007JCLI1824.1](https://doi.org/10.1175/2007JCLI1824.1).
- Saha, S. et al (2010), The NCEP climate forecast system reanalysis. *Bull. Amer. Meteor. Soc.*, 91, 1015-1057. [doi:10.1175/2010BAMS3001.1](https://doi.org/10.1175/2010BAMS3001.1).
- Saha, S. et al (2014), The NCEP climate forecast system version 2. *J. Clim.*, 27, 2185-2208. [doi:10.1175/JCLI-D-12-00823.1](https://doi.org/10.1175/JCLI-D-12-00823.1).
- Xie, P, and P. A. Arkin (1997), Global precipitation: A 17-year monthly analysis based on gauge observations, satellite estimates, and numerical model outputs. *Bull. Amer. Meteor. Soc.*, 78, 2539-2558. doi: [http://dx.doi.org/10.1175/1520-0477\(1997\)078%3C2539:GPAYMA%3E2.0.CO;2](http://dx.doi.org/10.1175/1520-0477(1997)078%3C2539:GPAYMA%3E2.0.CO;2)

# 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: NCDC daily OI 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 from CFSv2 (provided by Dr. Bhaskar Jha/CPC)
  - 18 members
- All above seasonal mean anomalies are based on 1999-2010 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)