Attribution of Seasonal Climate Anomalies January-February-March 2021

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

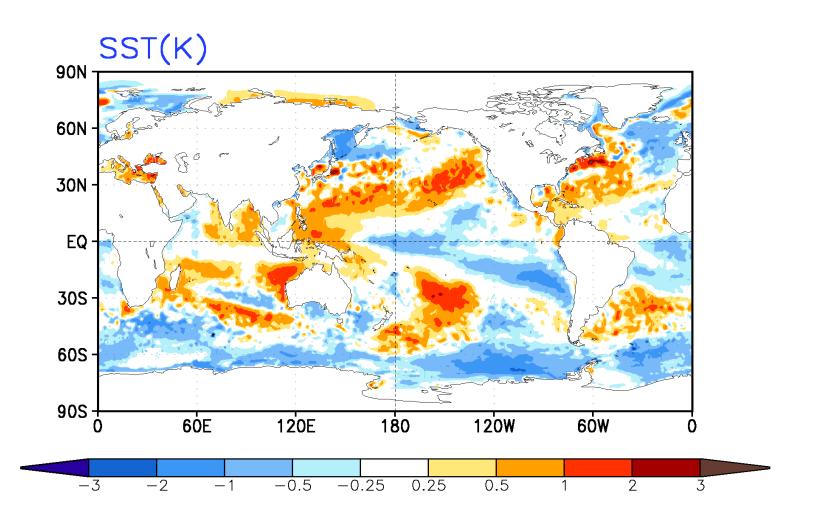
Summary of Observed Conditions and Outlooks

- <u>La Niña condition</u> continued in the central/eastern Pacific; tropical Indian Ocean and western and NE Pacific Ocean remained on the warm side (slide 4).
- In general, the large-scale distribution of negative and positive precipitation anomalies in the equatorial Pacific and over the Maritime Continent (a reflection of La Niña conditions) was well predicted in CFSv2 and other MME forecast systems (slides 11, 33, 34, & 39).
- The JFM2021 amplitude of <u>negative phase of AO</u> pattern of the height anomalies between 60°N-90°N reduced from DJF20/21 as <u>the AO index changed to positive since mid-Feb2021</u>. The initialized CFSv2 and other multi-model ensemble MME forecast didn't capture the observed height anomalies over the mid to high latitudes leading to errors in forecasting cold temperature anomalies over the NA and the northern part of Asia and Europe (slides 12, 13, 15, 16, 33, 34, & 38).
- The forecast NA temperature anomalies in the MME models were in the pattern consistent with the La Nina composite, while it was different in the CFSv2 alone (slides 7, 16, 33, 34, 38, 40).
- The forecast NA precipitation anomalies in the CFSv2 and MME models (slides 7, 14) were in the pattern consistent with the <u>La Nina composite</u>, however, were very different from the observed anomalies (slide2s, 7, 14. 39, 41)
- The Mar2021 monthly forecasts from the shortest 3-day-leads initial conditions captured the observed positive precipitation anomalies over the SE areas of US and captured the most observed temperature anomalies (slides 30, 31).

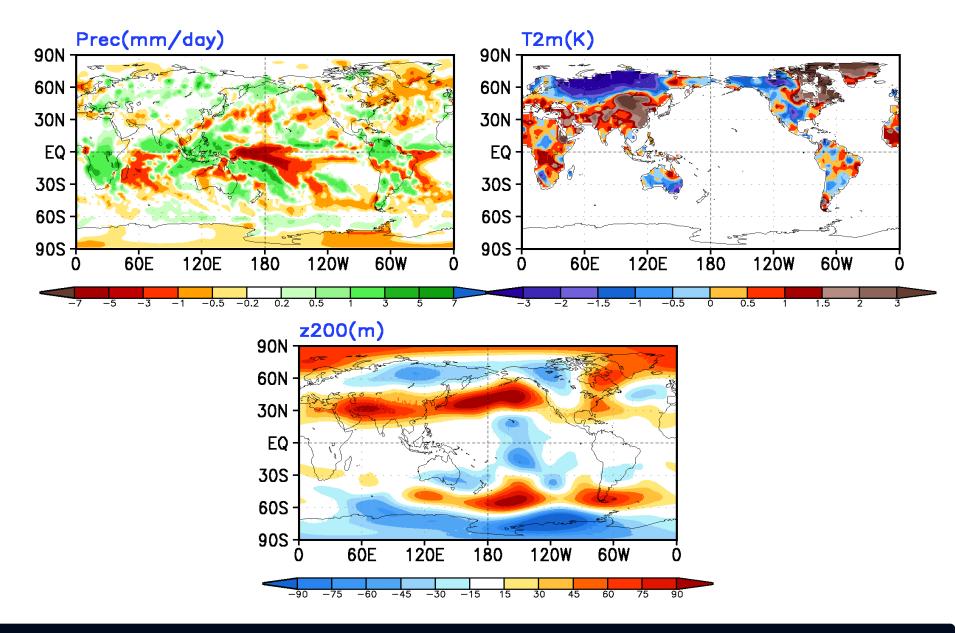
Observed Seasonal Anomalies

Global and North America

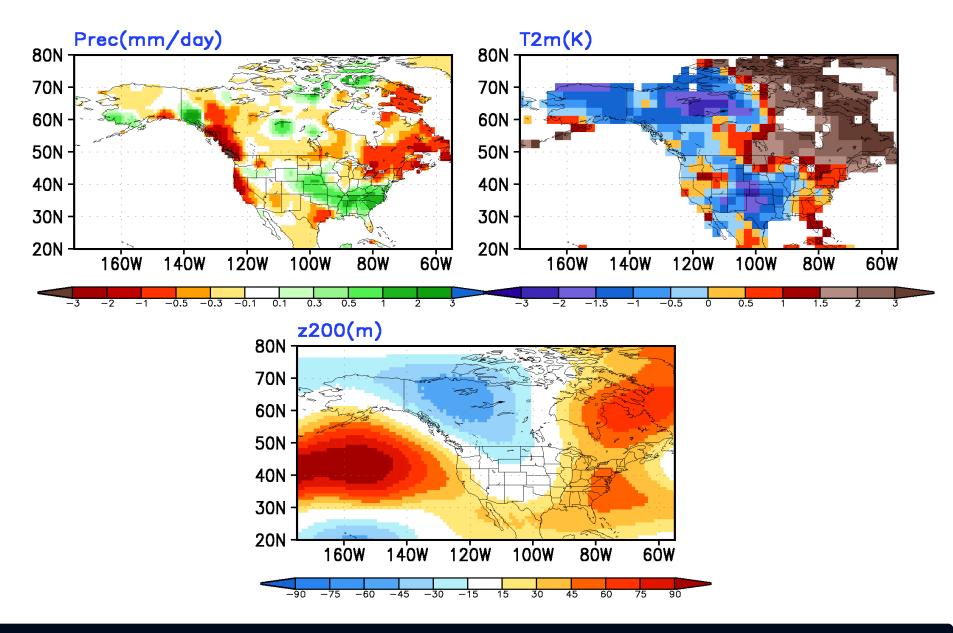
Observed Anomaly JFM2021



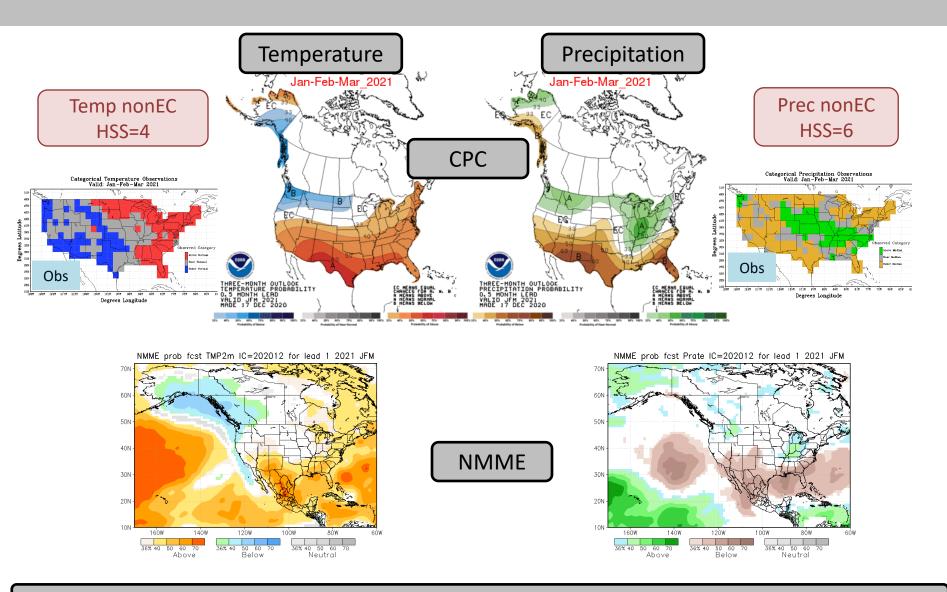
Observed Anomaly JFM2021



Observed Anomaly JFM2021



CPC Seasonal Outlooks and NMME Forecasts



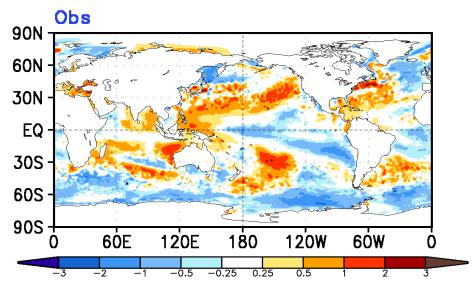
For the rationale behind CPC outlooks see: https://www.cpc.ncep.noaa.gov/products/archives/long lead/PMD/2020/202012 PMD90D

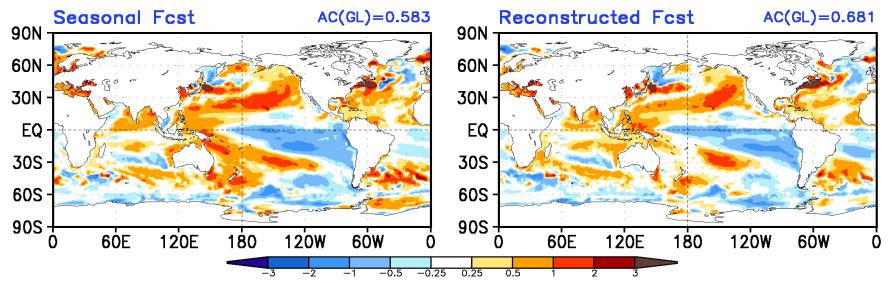
Iodel Simulated/Forecast Ensemble Mean Anomalie	2 S

Model Simulated/Forecast Ensemble Average Anomalies

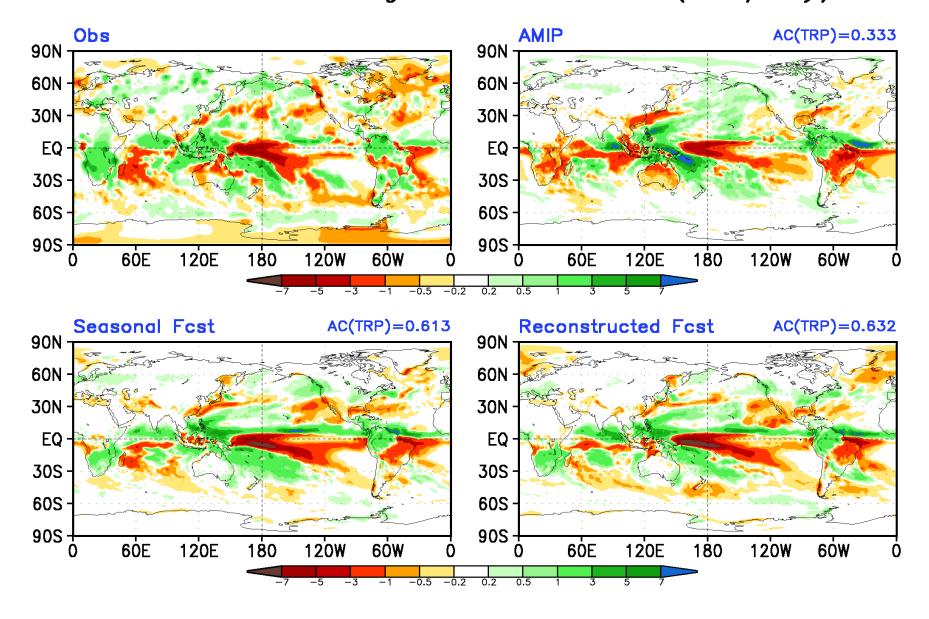
- CFS AMIP simulations forced with observed sea surface temperatures (18 members ensemble)
- CFSv2 real time operational forecasts
 - Seasonal forecast: the seasonal mean forecasts based on 40 members from the latest
 10 days before the target season (0-month-lead). For example, 2016AMJ seasonal
 mean 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 fr 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).

JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies SST(K)

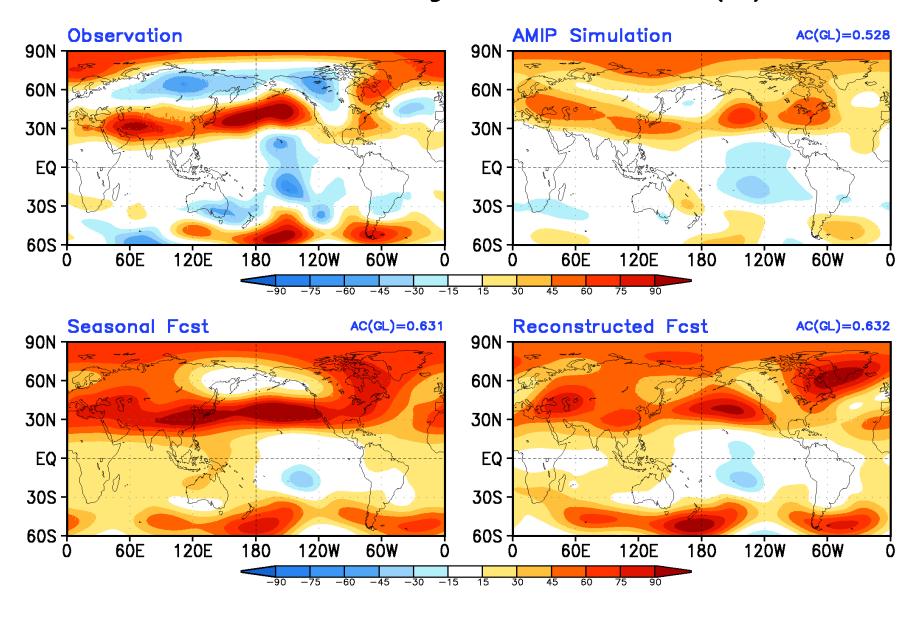




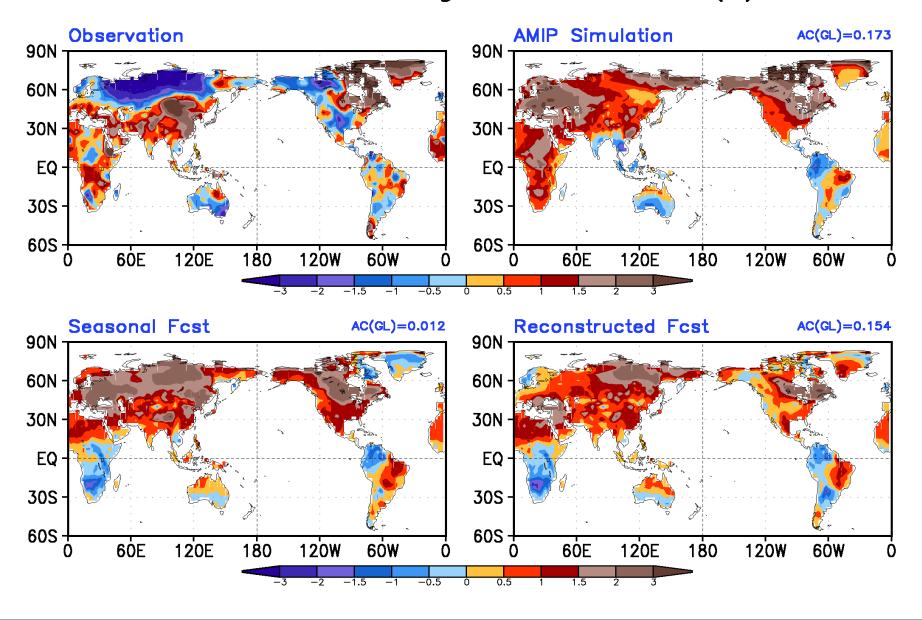
JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



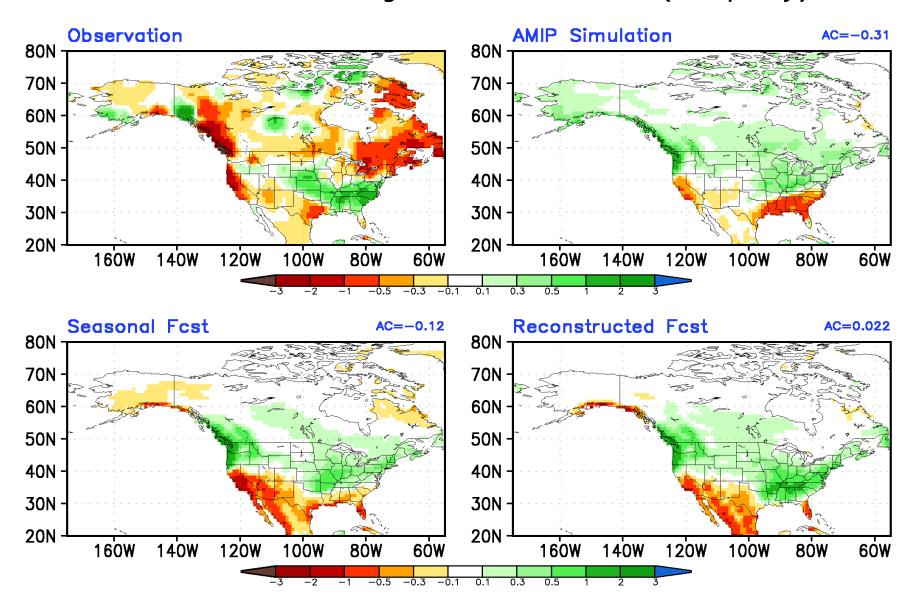
JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



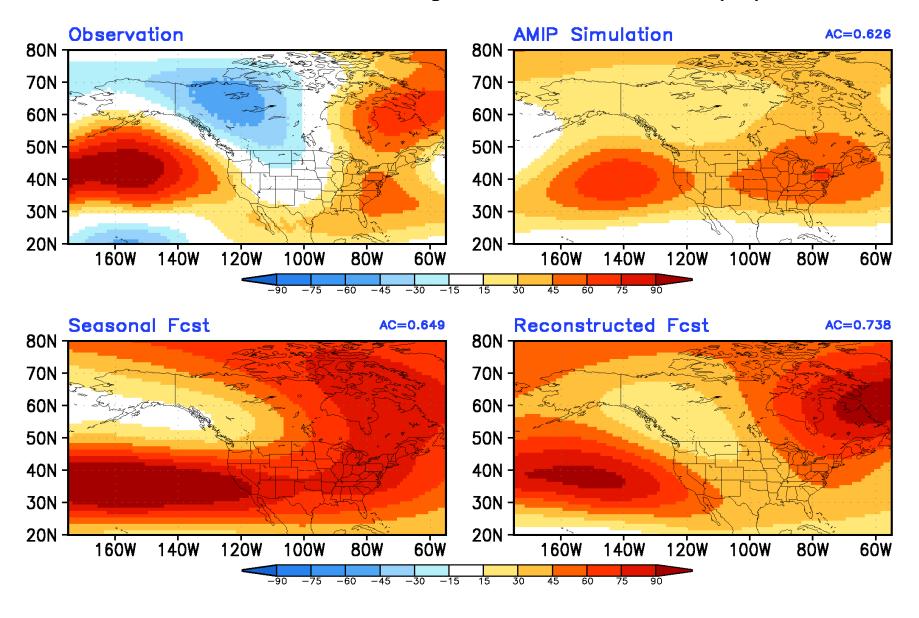
JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



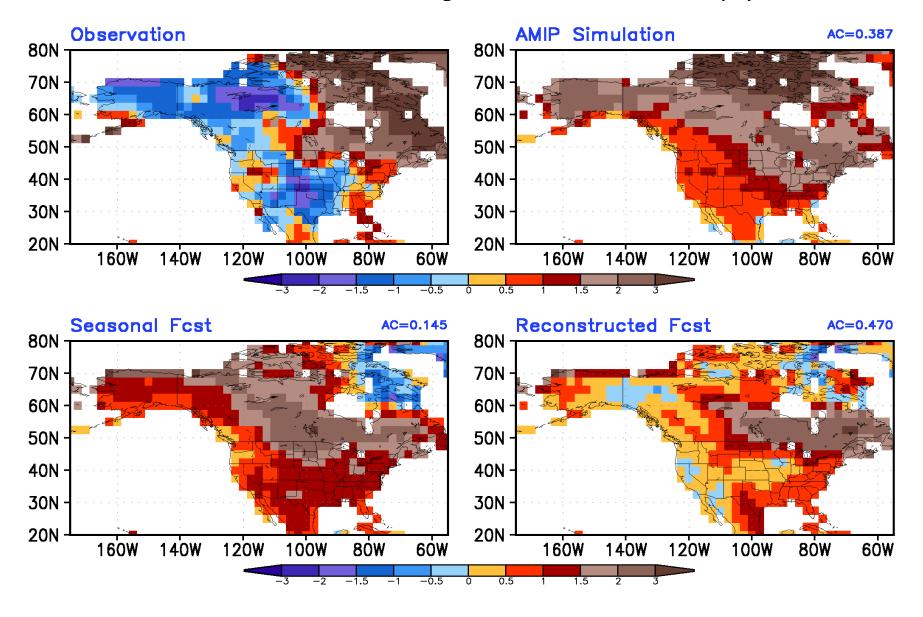
JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)

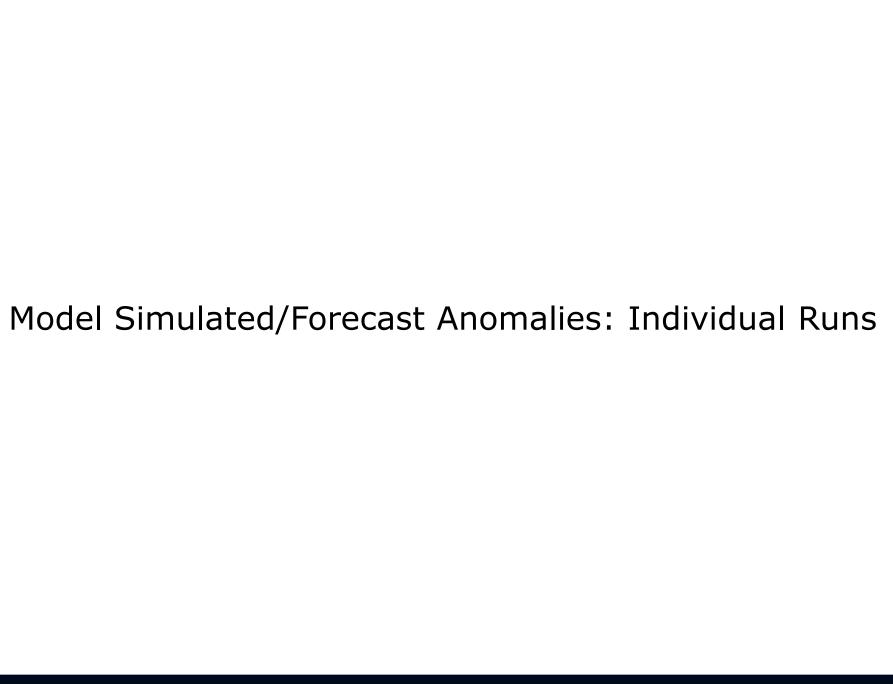


JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



JFM2021 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)

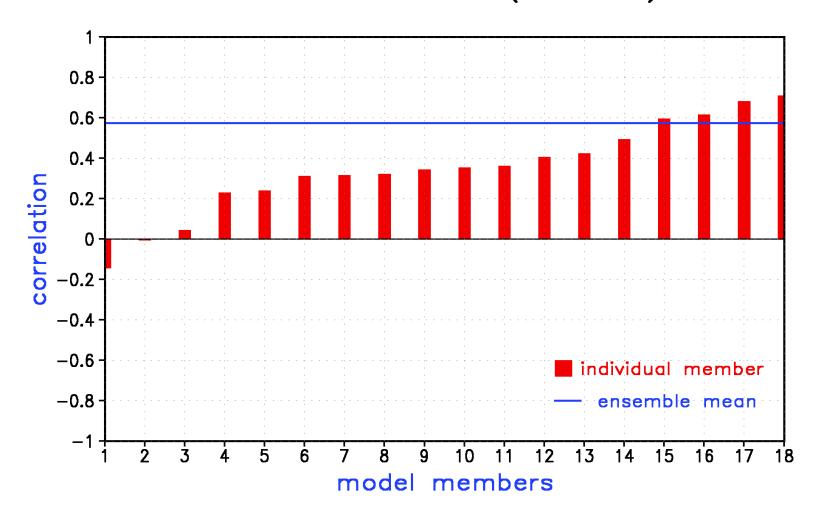




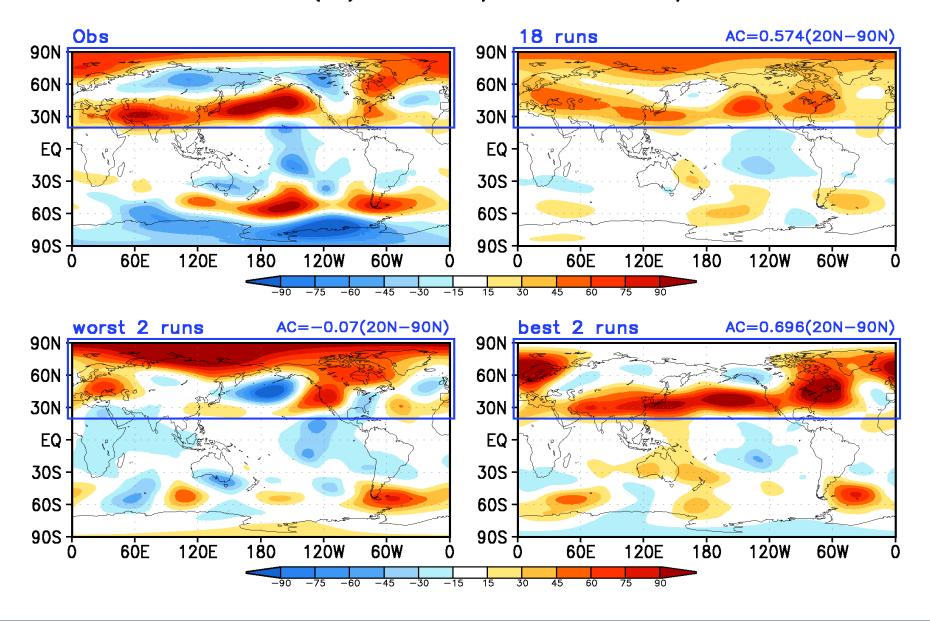
Model Simulated/Forecast Anomalies: Individual Runs

- 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.

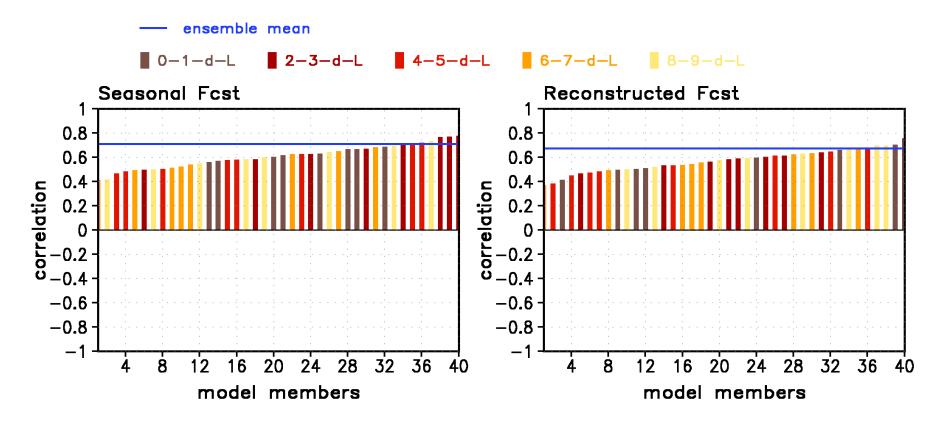
JFM2021 Anomaly Correlation for Individual AMIP Simulation with Observation —— z200(20N—90N)



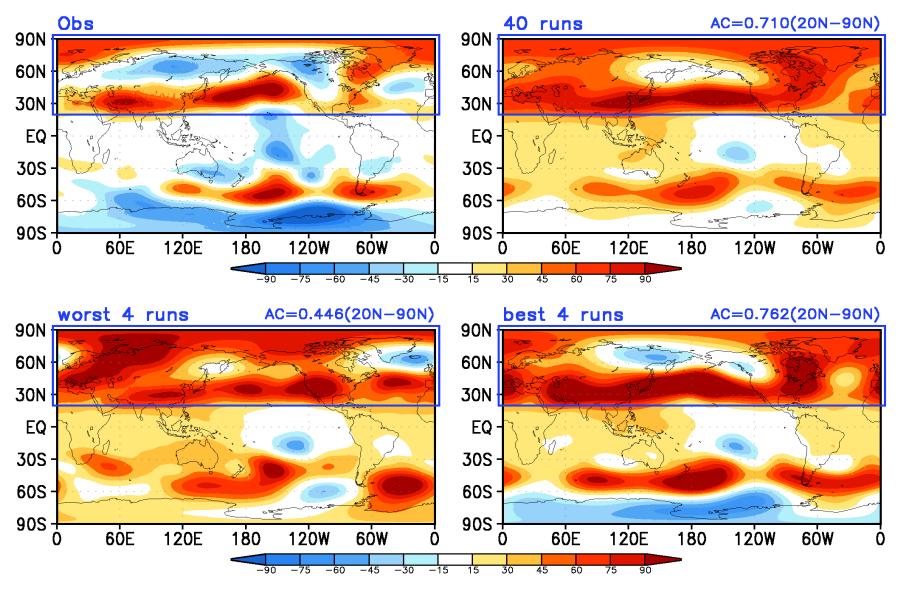
Observed & AMIP Ensemble Average Anomalies JFM2021 z200(m) 18 runs/worst 2 runs/best 2 runs



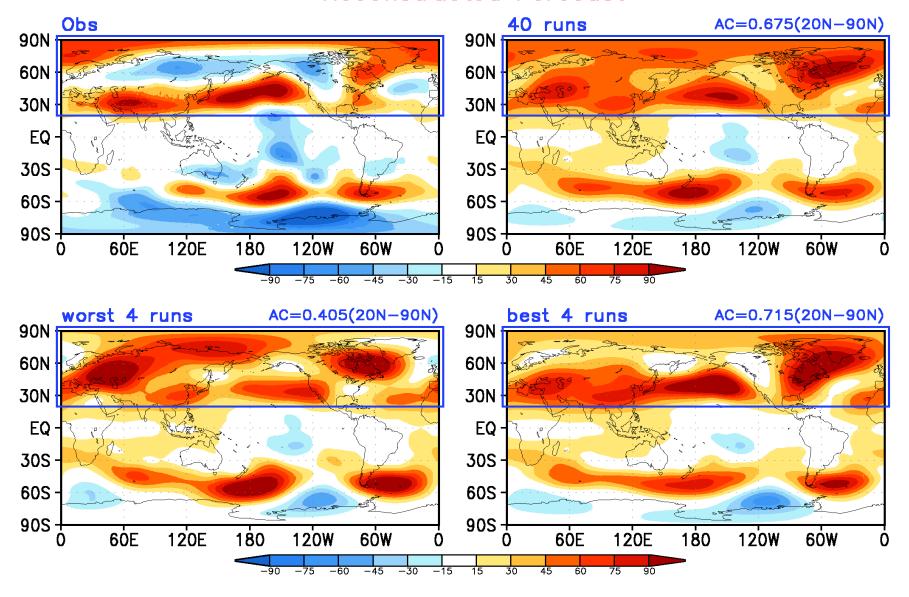
JFM2021 Anomaly Correlation for Individual CFSv2 Forecast with Observation —— z200 (20N—90N)



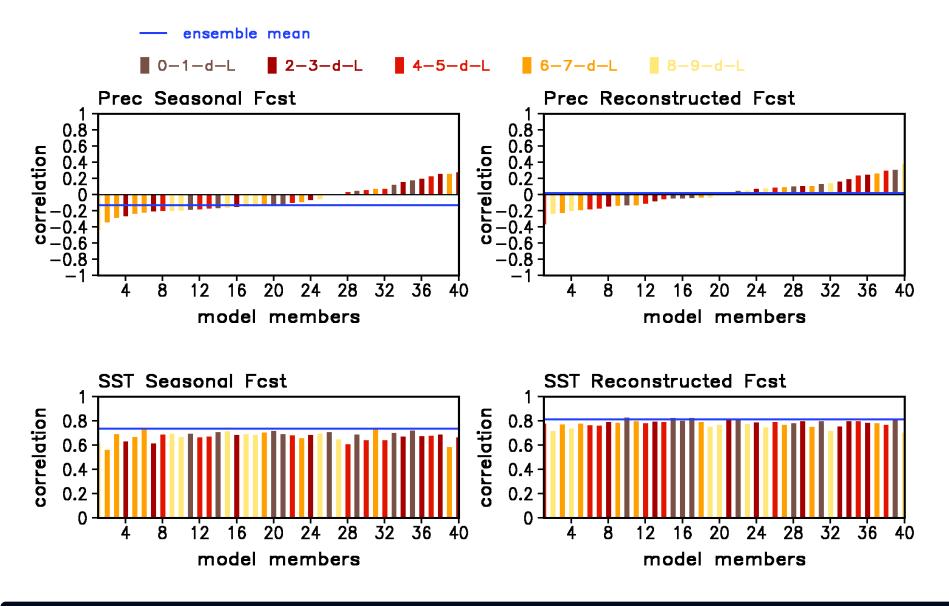
Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 z200(m) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



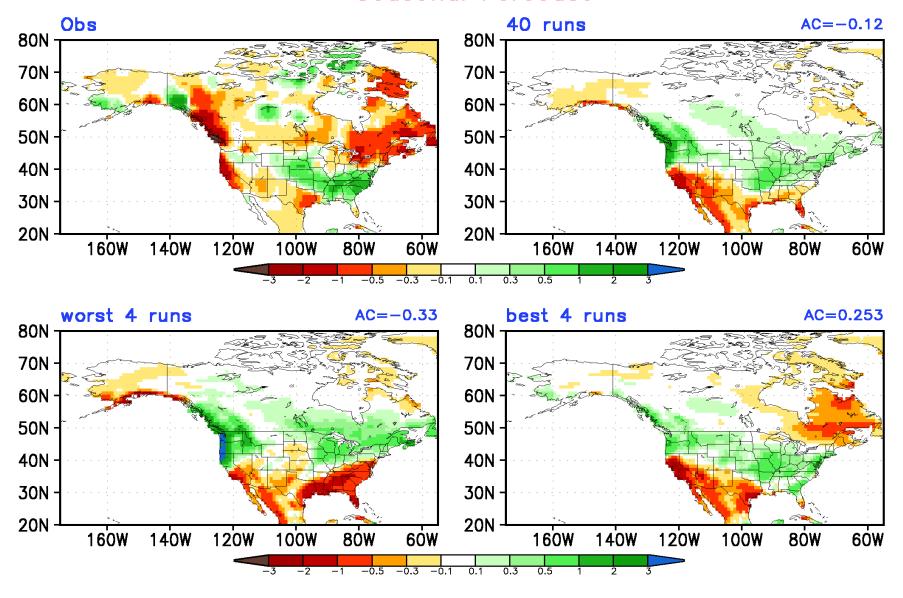
Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 z200(m) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



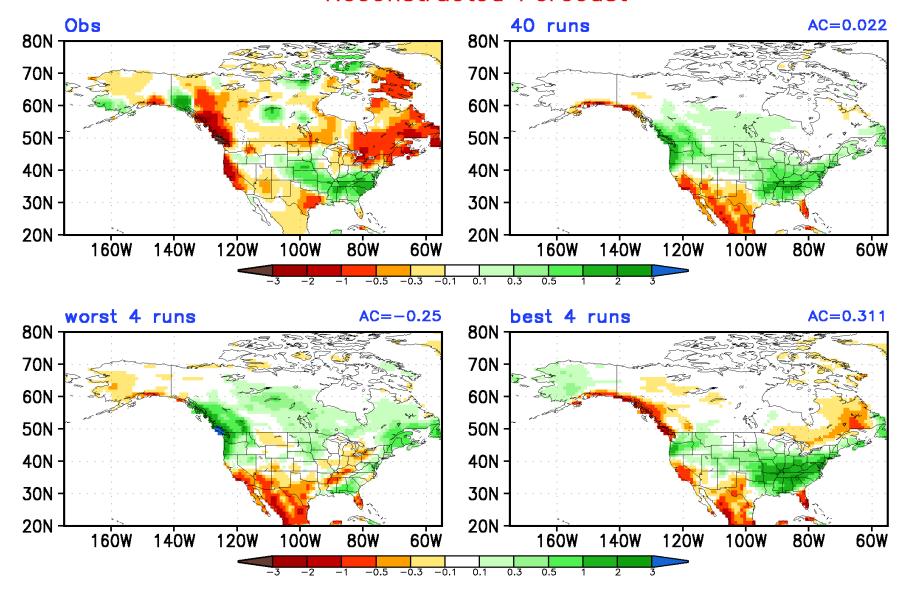
JFM2021 Anomaly Correlation for Individual CFSv2 Forecast with Observation —— Prec(NA)/SST(30S—30N)



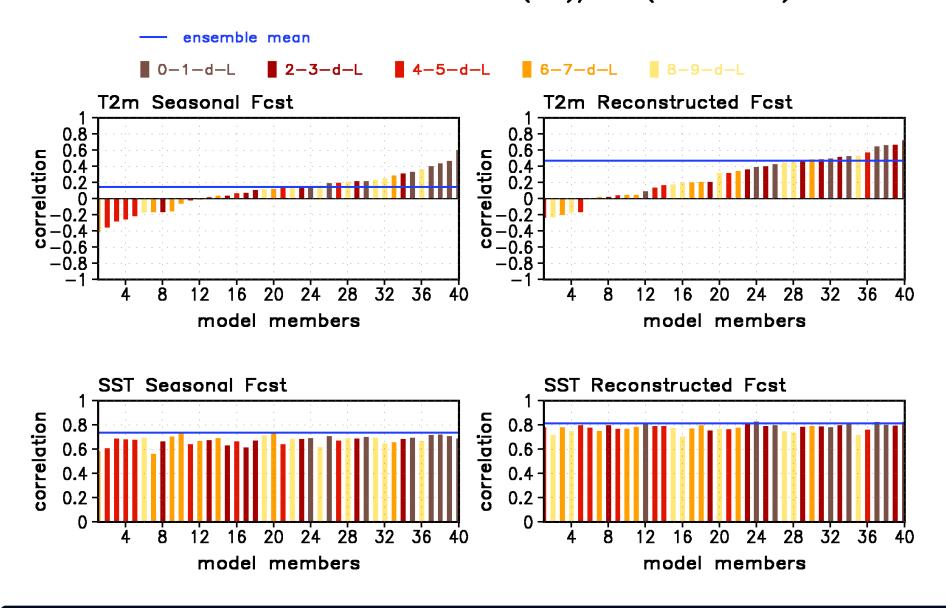
Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



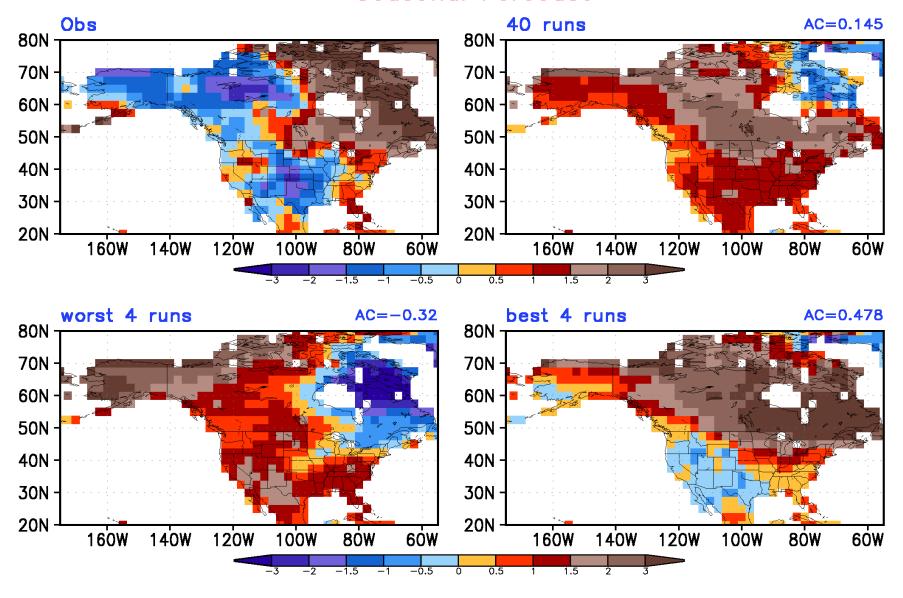
Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



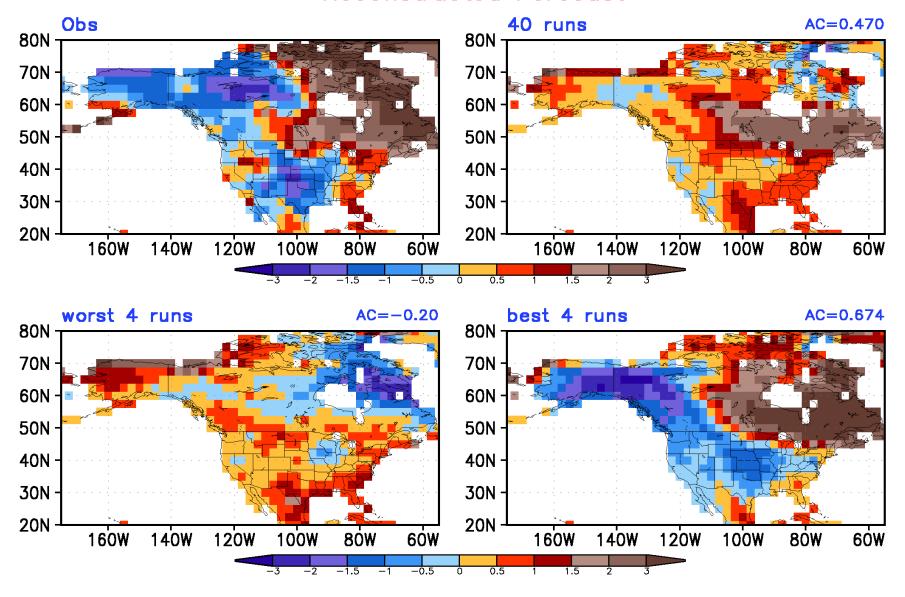
JFM2021 Anomaly Correlation for Individual CFSv2 Forecast with Observation — T2m(NA)/SST(30S-30N)



Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 T2m(K) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast

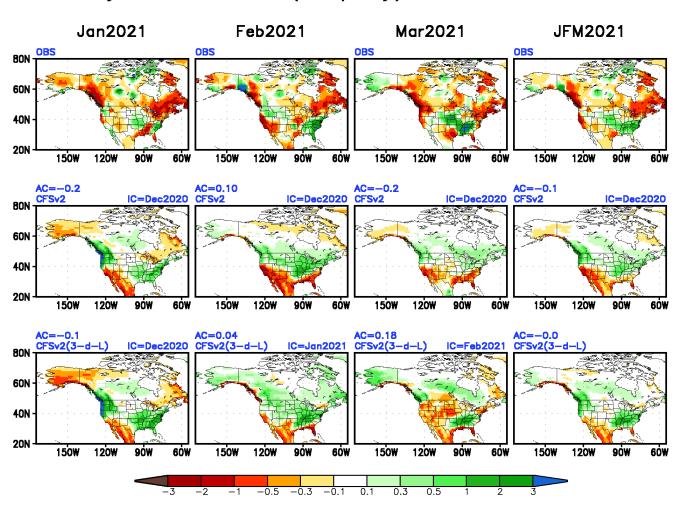


Observed & CFSv2 Forecast Ensemble Average Anomalies JFM2021 T2m(K) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



Monthly Means Prec(mm/day) Observed & Forecasts

Monthly JFM2021 Prec(mm/day) Observed & Forecasts



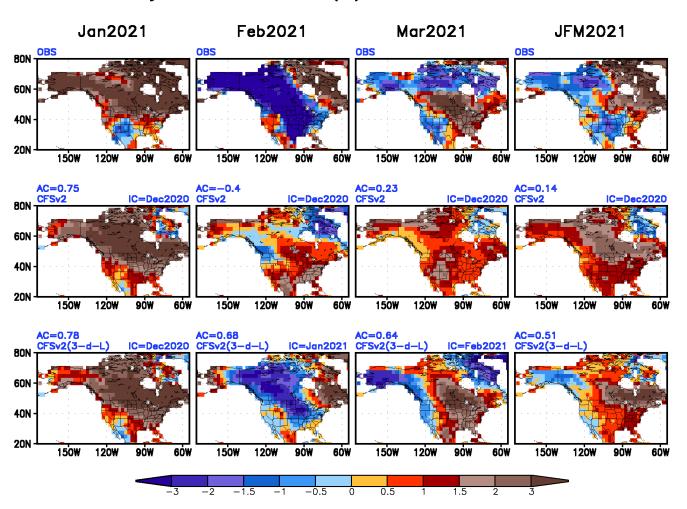
Top row: Observed anomaly

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 <u>target</u> month.

Monthly Means T2m(K) Observed & Forecasts

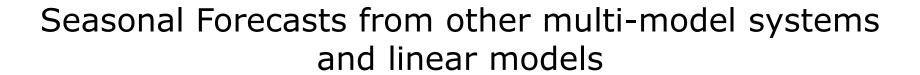
Monthly JFM2021 T2m(K) Observed & Forecasts



Top row: Observed anomaly

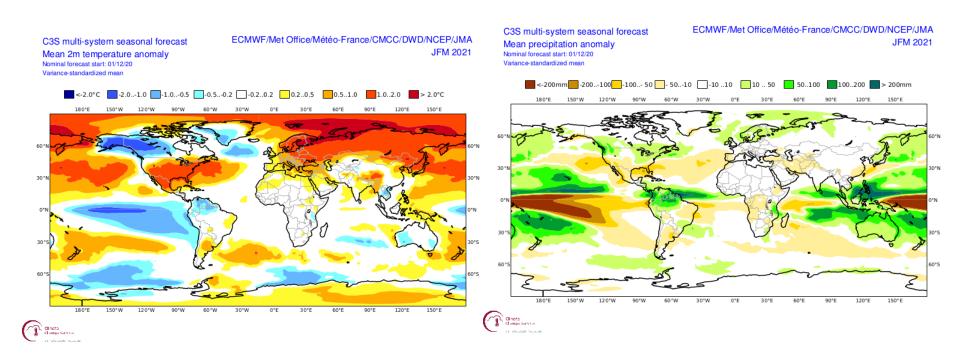
Middle row: CFSv2 seasonal forecasts from the initial conditions from the month prior to the target season.

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



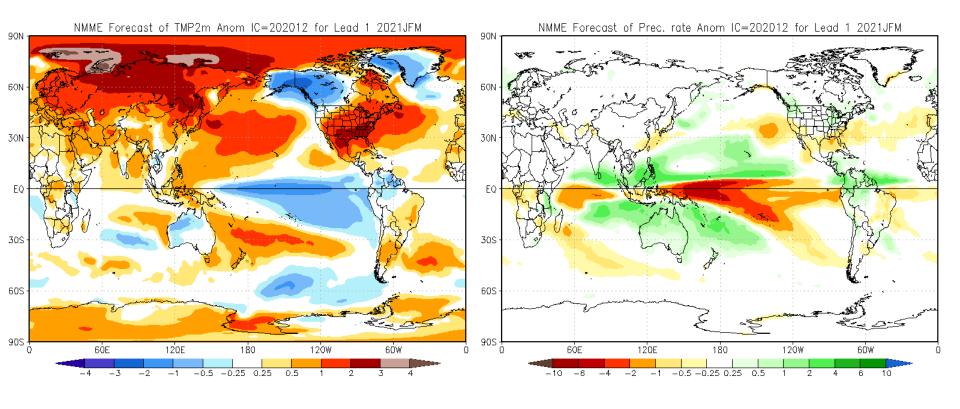
C3S Seasonal Forecast

(https://climate.copernicus.eu/charts/c3s seasonal/)



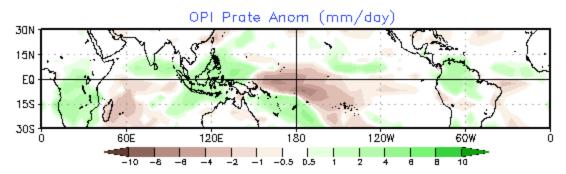
North American Multi-Model Ensemble Seasonal Forecast

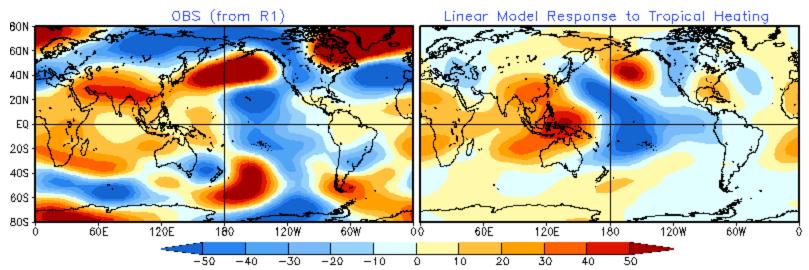
(https://www.cpc.ncep.noaa.gov/products/NMME/)



200mb Height from Linear Model

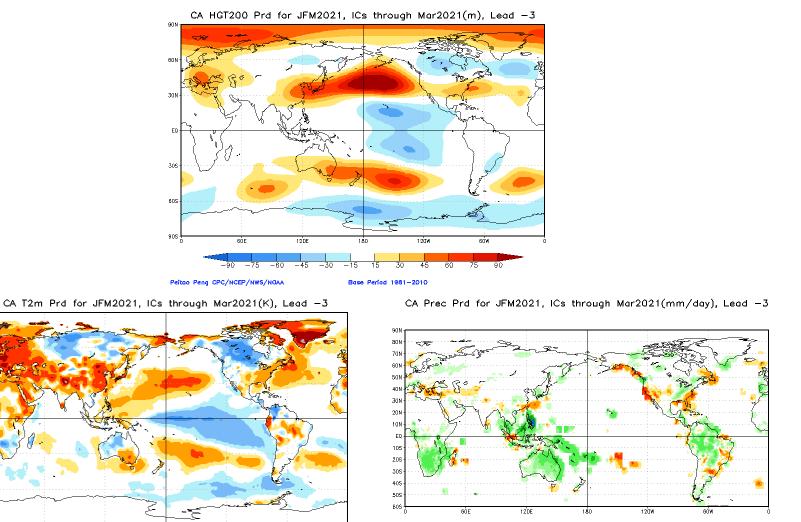
JFM2021 200mb Eddy HGT(m)
OBS vs. Linear Model Response to Tropical Heating
Heating is converted from Prate in 15S—15N





Pattern COR: global=0.36, tropics(30S-30N)=0.62

Seasonal Forecasts from the Constructed Analog Model



Peitoo Peng CPC/NCEP/NWS/NGAA

Base Period 1981-2010

Peitoo Peng CPC/NCEP/NWS/NGAA

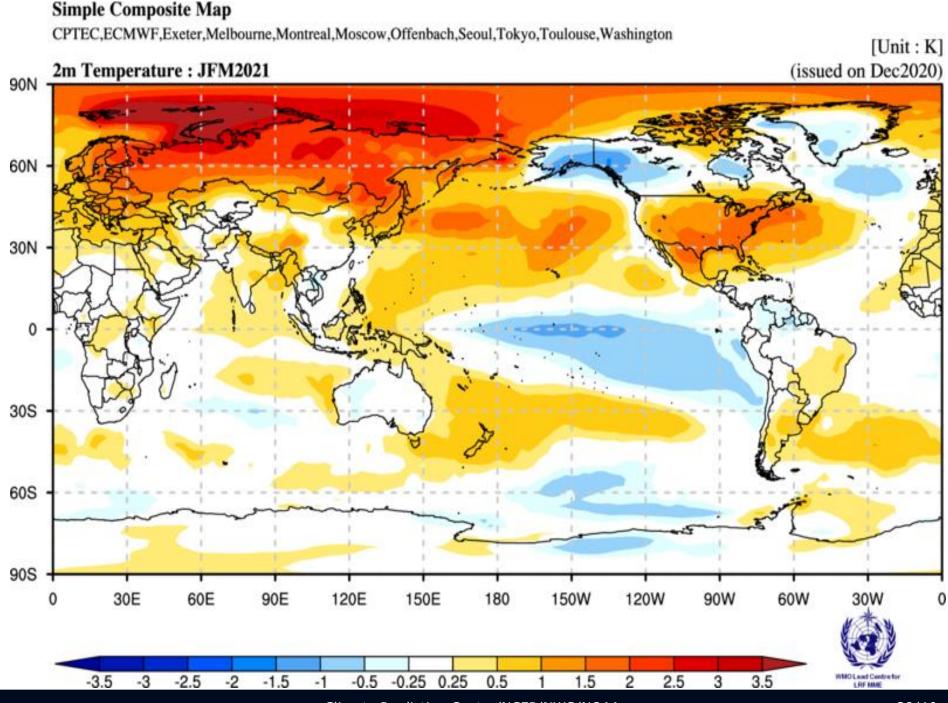
Base Period 1981-2010

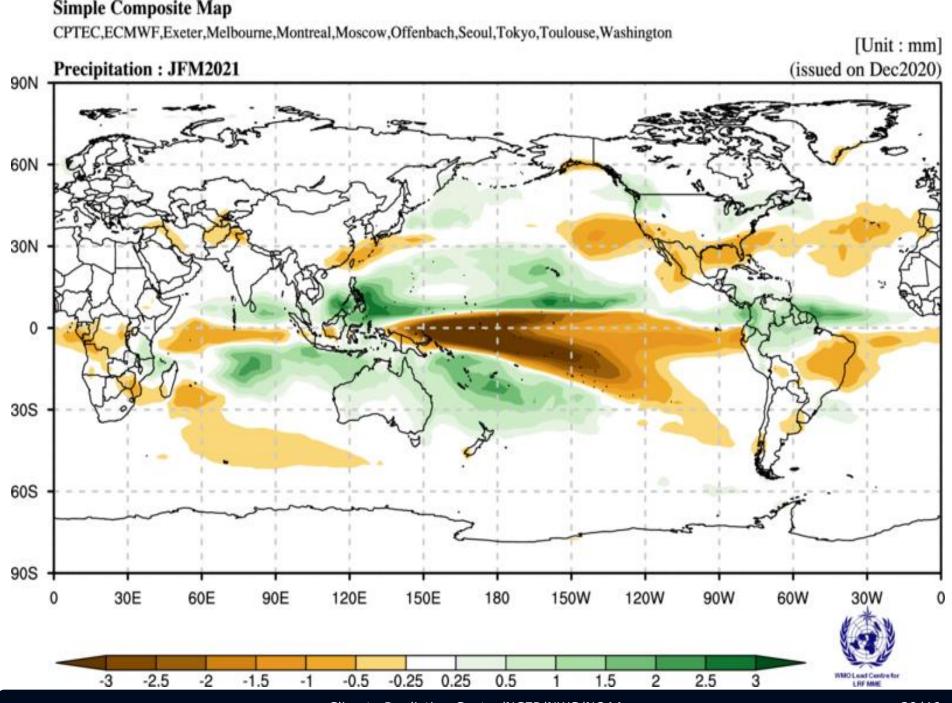
Seasonal Forecasts from WMO Lead Center for Long-Range Forecast Multi-Model Ensemble (LC-LRFMME)

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

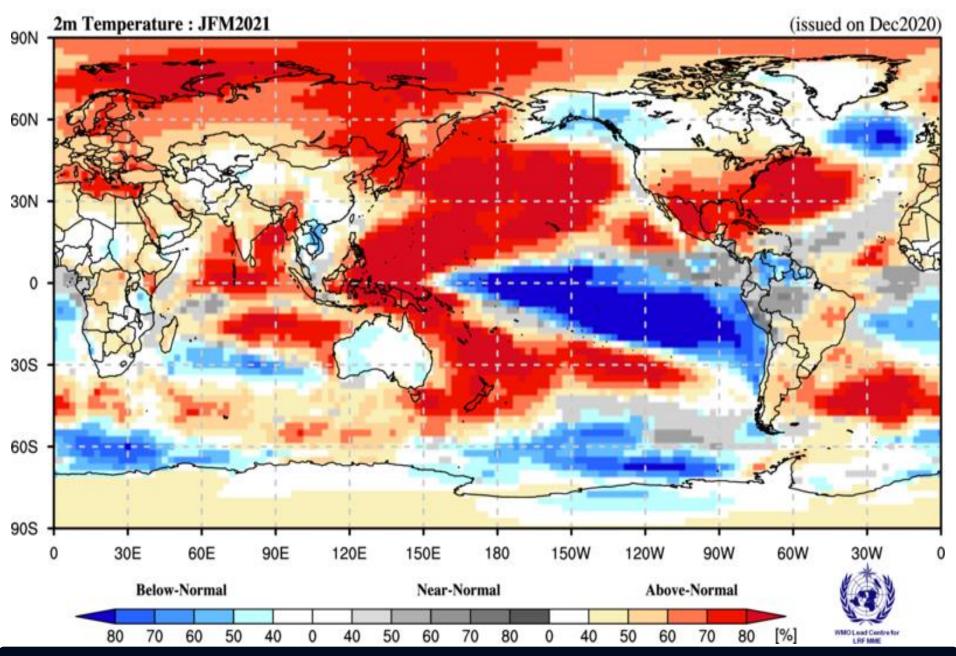
[•] For more information see visit Lead Center website; also see Graham, R., and Co-authors, 2011: New perspectives for GPCs, their role in the GFCS and a proposed contribution to a 'World Climate Watch'. Climate Research, 47, 47-55.





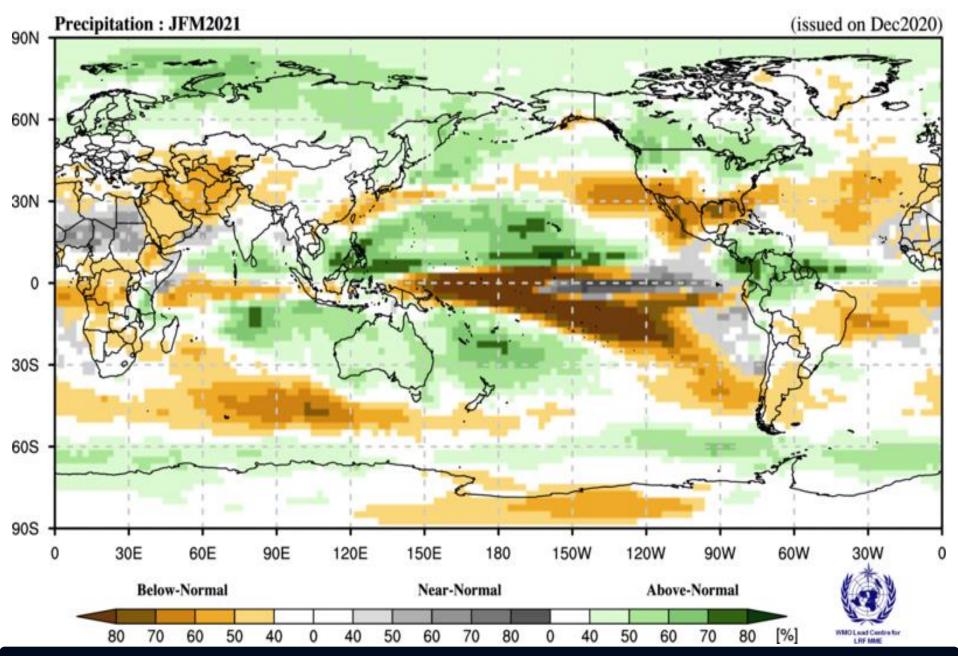
Probabilistic Multi-Model Ensemble Forecast

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



Probabilistic Multi-Model Ensemble Forecast

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





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)
 - <u>Seasonal forecast:</u> 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)