Attribution of Seasonal Climate Anomalies September-October-November 2020

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

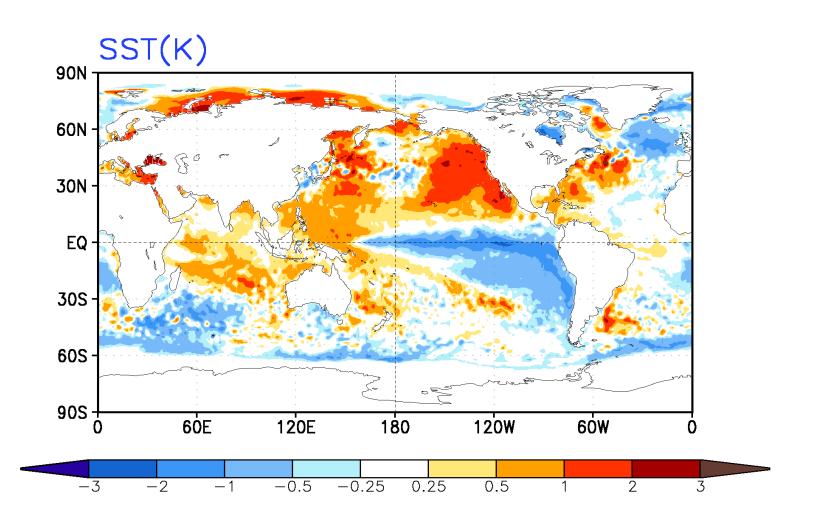
Summary of Observed Conditions and Outlooks

- <u>La Niña condition</u> continued in the central/eastern Pacific; tropical Indian Ocean and western Pacific Ocean continued to be on the warm side (slide 4). SSTs in extratropical northeast Pacific were above normal and maybe indicative of an ongoing <u>Marine Heat Wave (MHW)</u>.
- The observed positive land temperature anomalies over northern Asia and Europe, far north Canada, Australia and S. America were well predicted in CFSv2 and other multi-model ensemble MME forecasts (slides 5, 13, 33, 34, & 38).
- In general, the large-scale distribution of negative and positive precipitation anomalies in the western Indian Ocean and equatorial Pacific (a reflection of La Niña conditions) was well predicted in CFSv2 and other MME forecast systems (slides 11, 33, 34, & 39).
- The initialized CFSv2 captured the large-scale structure of the observed positive height anomalies over tropics, while missed the large negative height anomalies over the region of central-east Canada that led terrors in forecasting cold temperature anomalies over the central NA (slides 12, 13, 15, & 16). This observed large cold temperature anomalies were also not predicted in various MME systems (Slides 33, 34 & 39).
- The CPC outlook anticipated the observed precipitation anomalies over most of US, while the NMME only indicated negative anomalies over the central part of US; CFSv2 missed most of observed anomalies (slides 7 & 14).
- The Nov 2020 monthly forecasts from the shortest 3-day-leads initial conditions predicted the observed negative
 precipitation anomalies over the southern US and west coast and cold anomalies over the region from Canada to the US
 northern boundary (slide 30 &31).

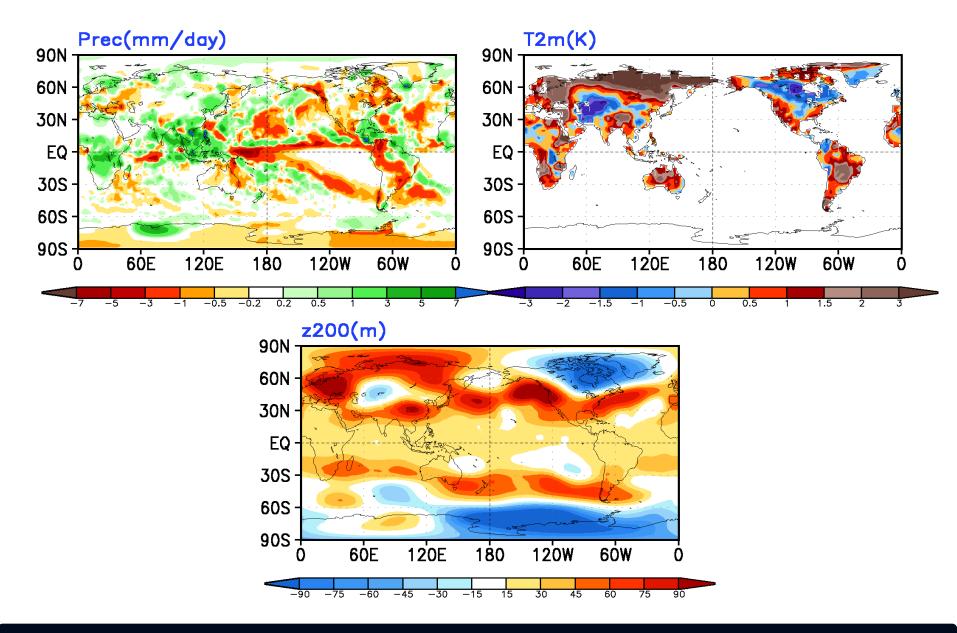
Observed Seasonal Anomalies

Global and North America

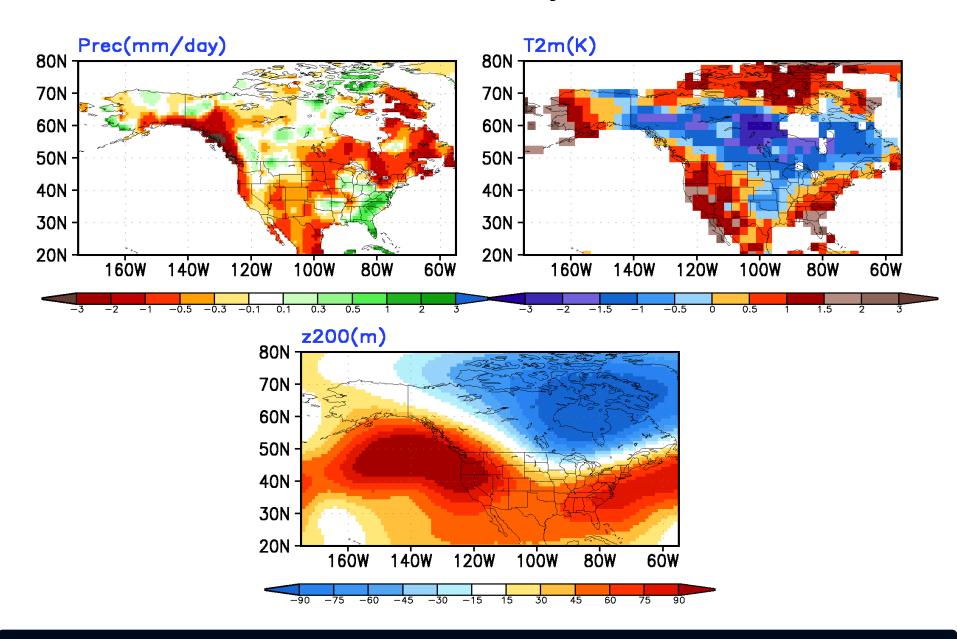
Observed Anomaly SON2020



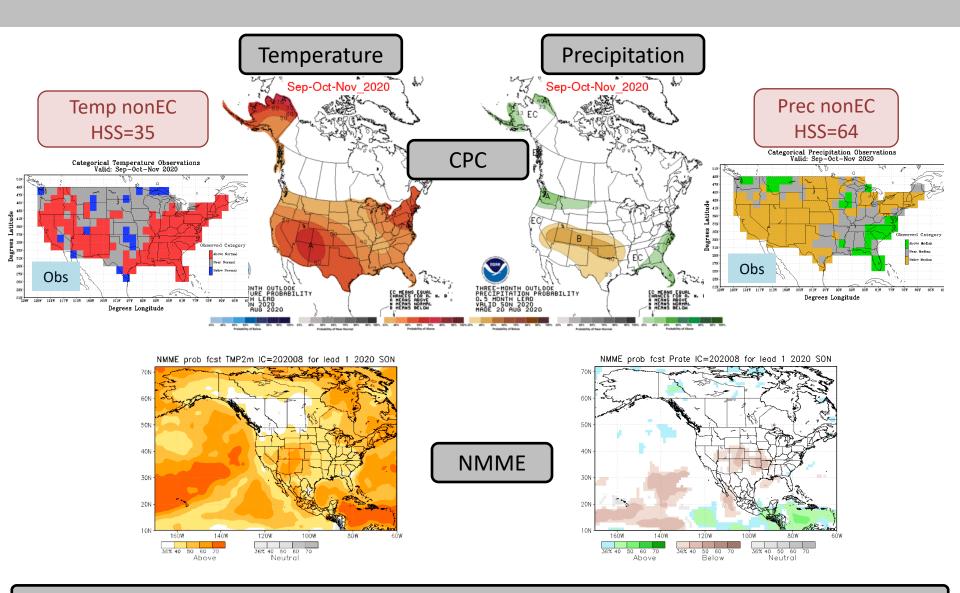
Observed Anomaly SON2020



Observed Anomaly SON2020



SON2020 CPC Seasonal Outlooks and NMME Forecasts



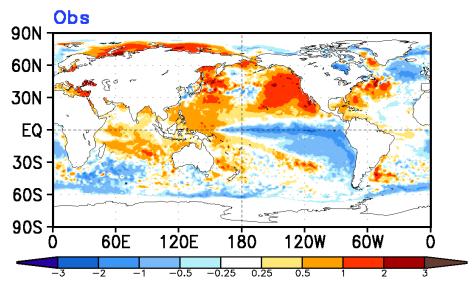
For the rationale behind CPC outlooks see https://www.cpc.ncep.noaa.gov/products/archives/long lead/PMD/2020/202008 PMD90D

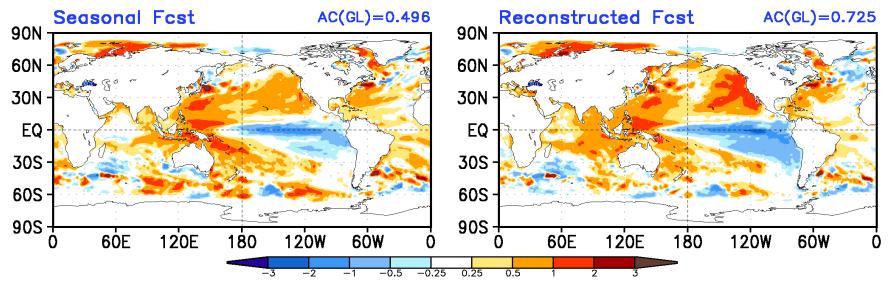
Iodel Simulated/Forecast Ensemble Mean Anomalie	2 S

Model Simulated/Forecast Ensemble Average Anomalies

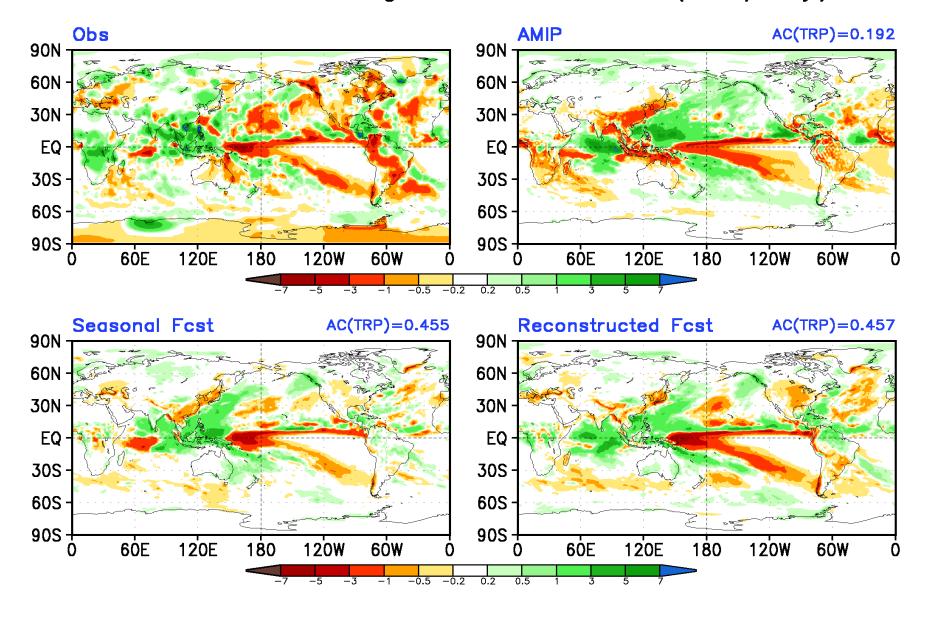
- CFS AMIP simulations forced with observed sea surface temperatures (18 members ensemble)
- CFSv2 real time operational forecasts
 - Seasonal forecast: the seasonal mean forecasts based on 40 members from the latest
 10 days before the target season (0-month-lead). For example, 2016AMJ seasonal
 mean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

SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies SST(K)

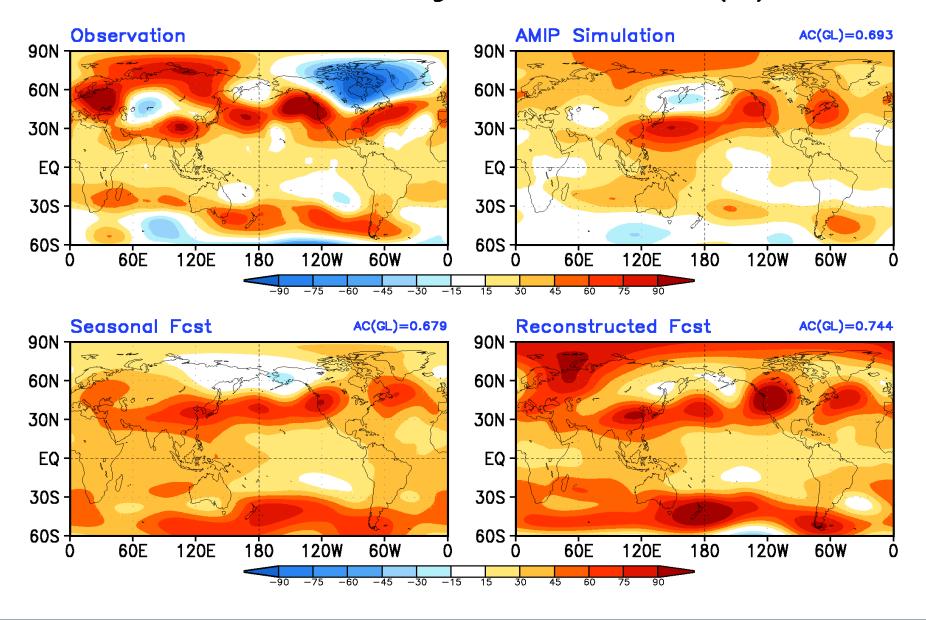




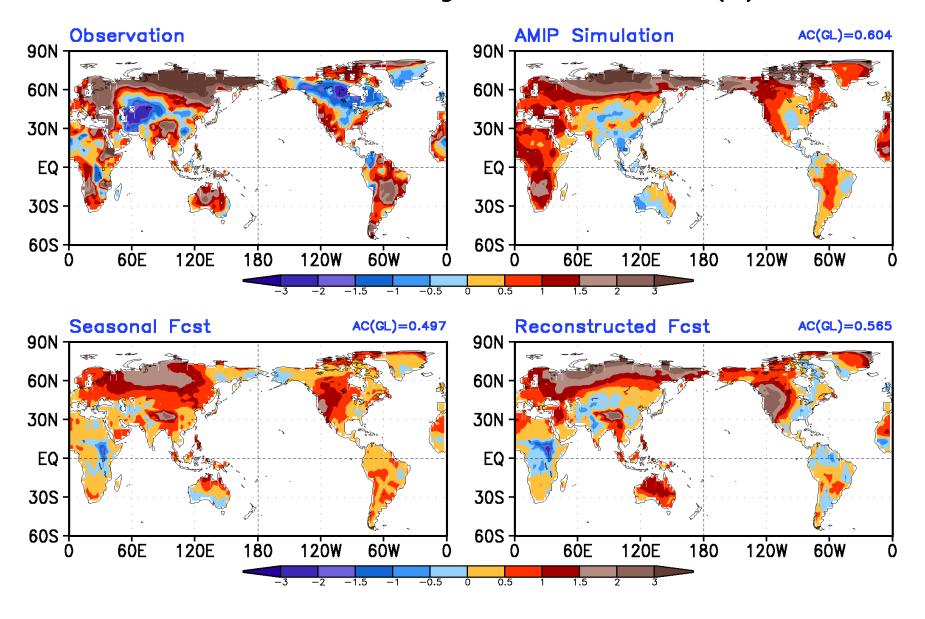
SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)



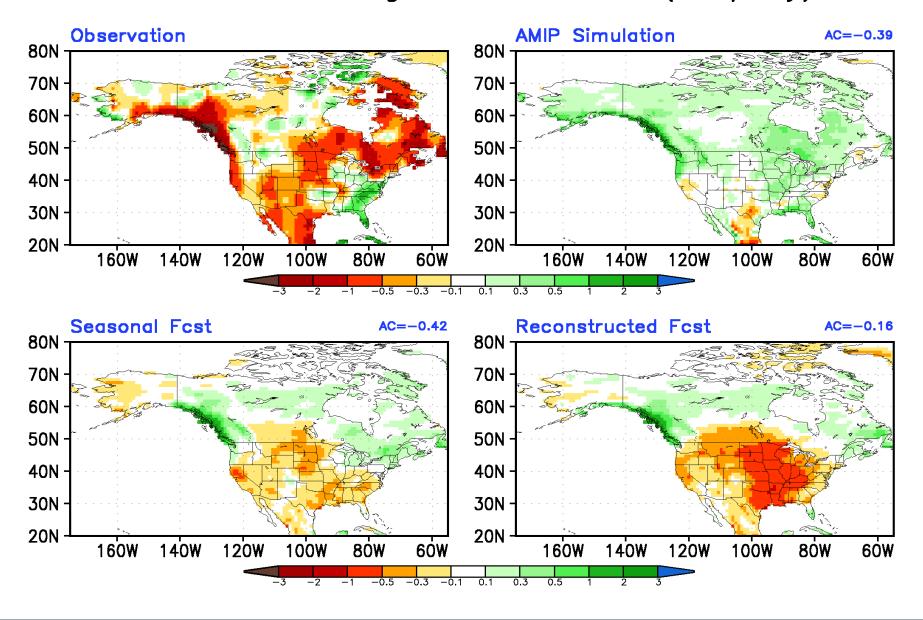
SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



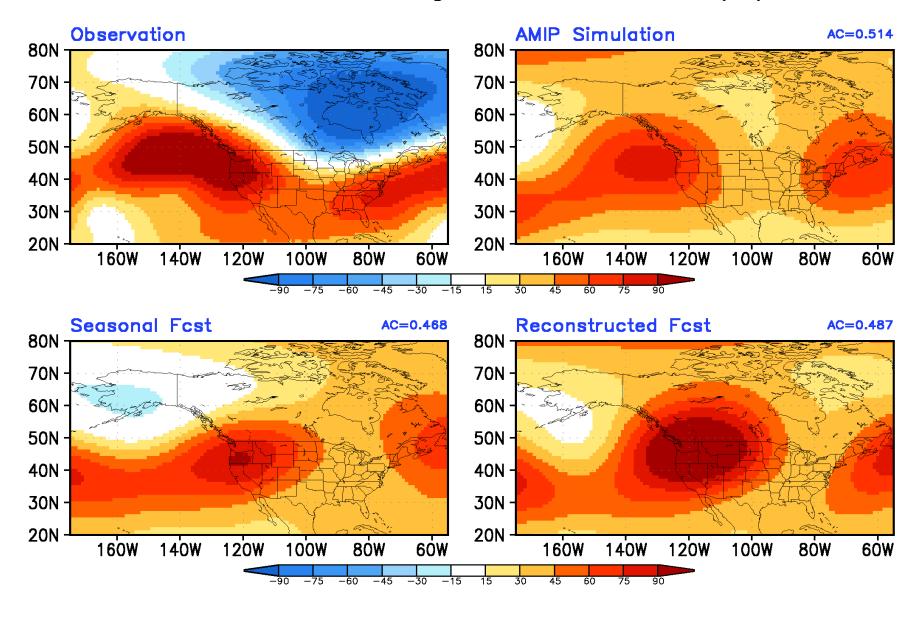
SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)



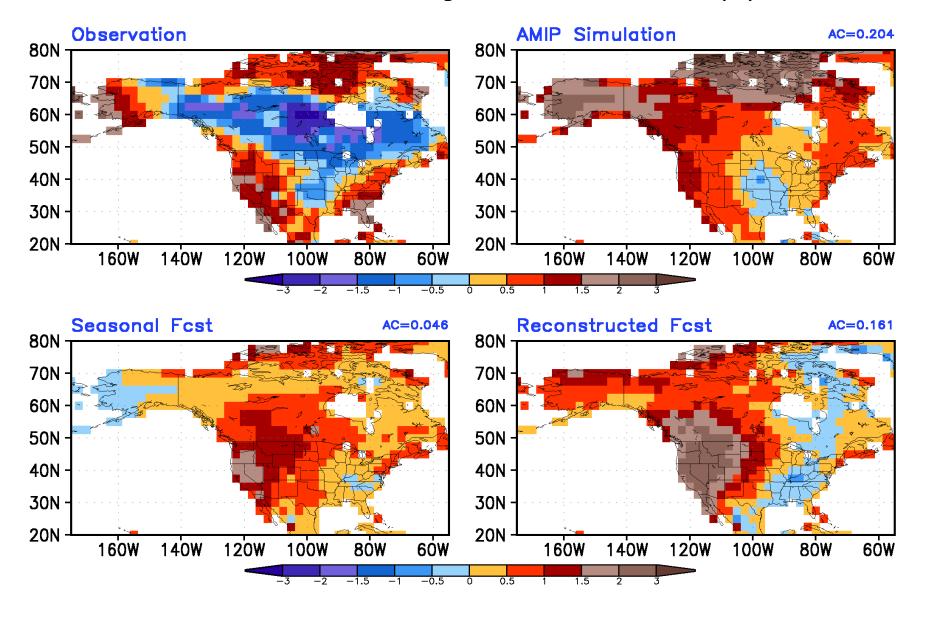
SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies Prec(mm/day)

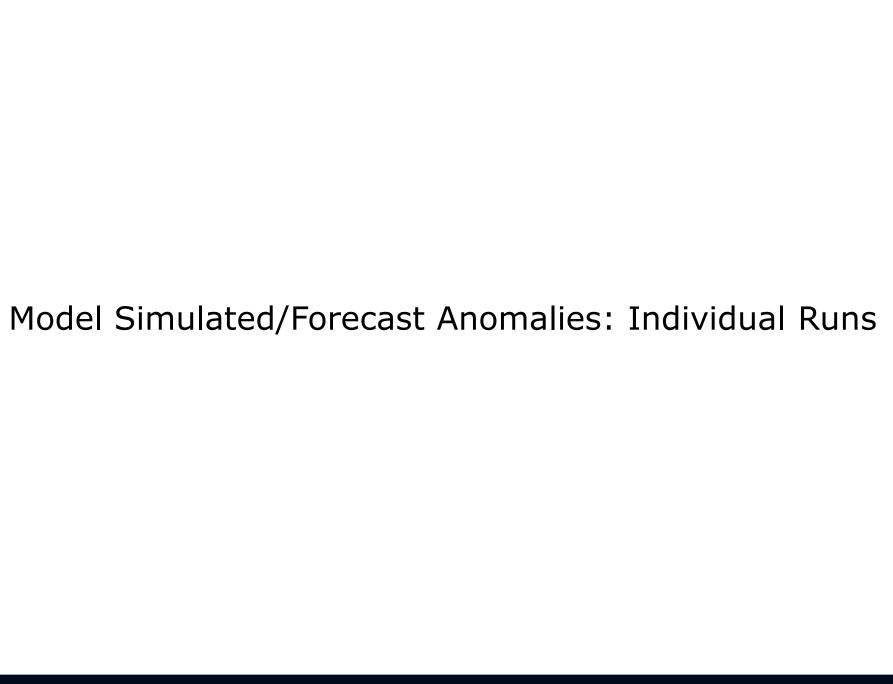


SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies z200(m)



SON2020 Observed & Model Simulated/Forecast Ensemble Average Anomalies T2m(K)

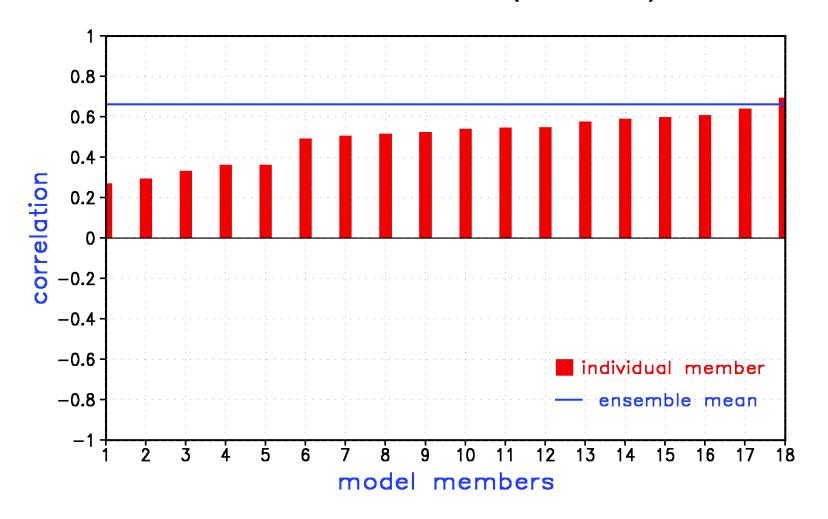




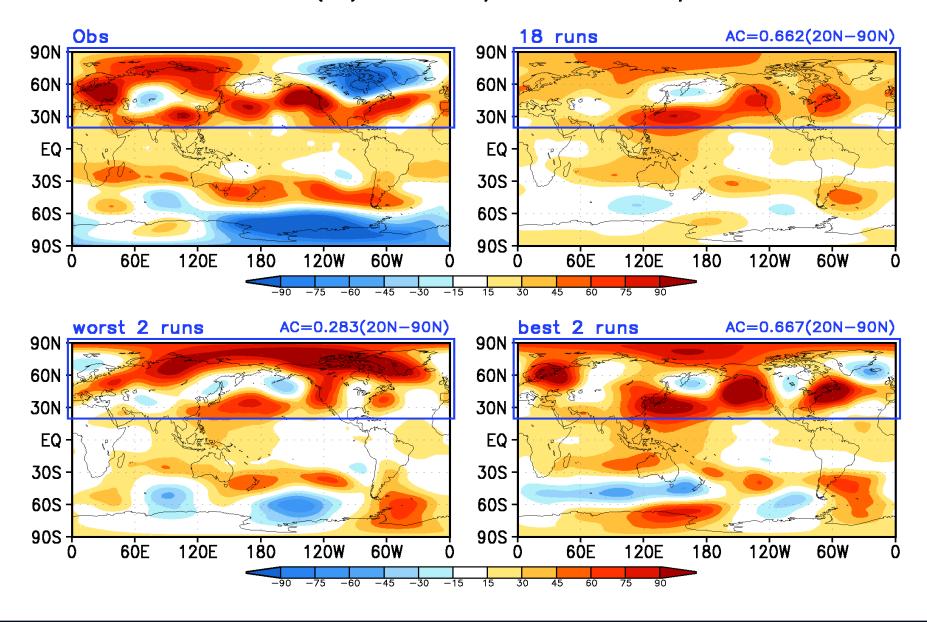
Model Simulated/Forecast Anomalies: Individual Runs

- 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.

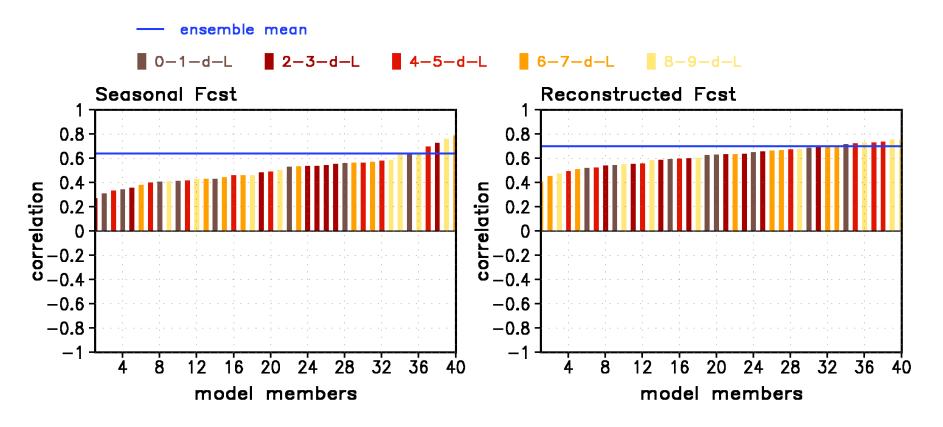
SON2020 Anomaly Correlation for Individual AMIP Simulation with Observation —— z200(20N—90N)



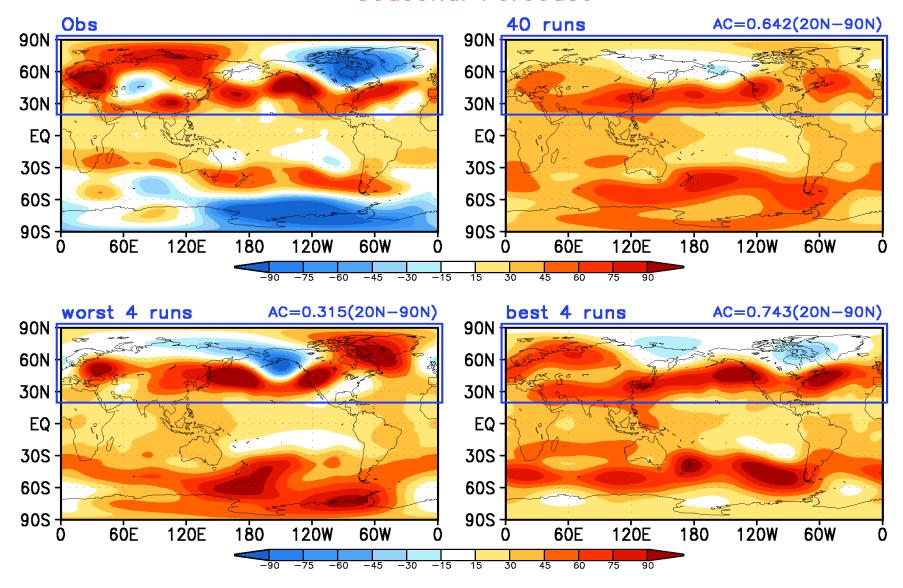
Observed & AMIP Ensemble Average Anomalies SON2020 z200(m) 18 runs/worst 2 runs/best 2 runs



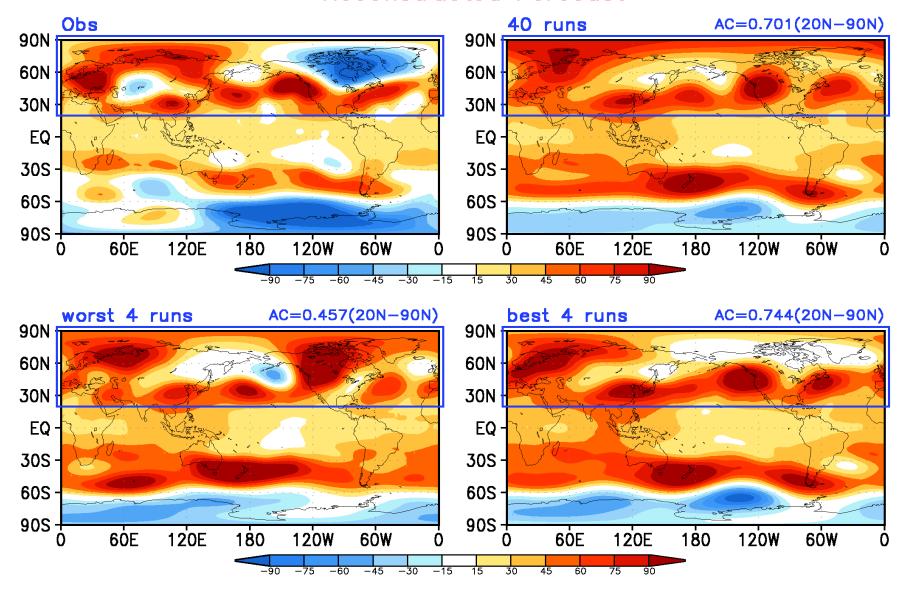
SON2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation -- z200 (20N-90N)



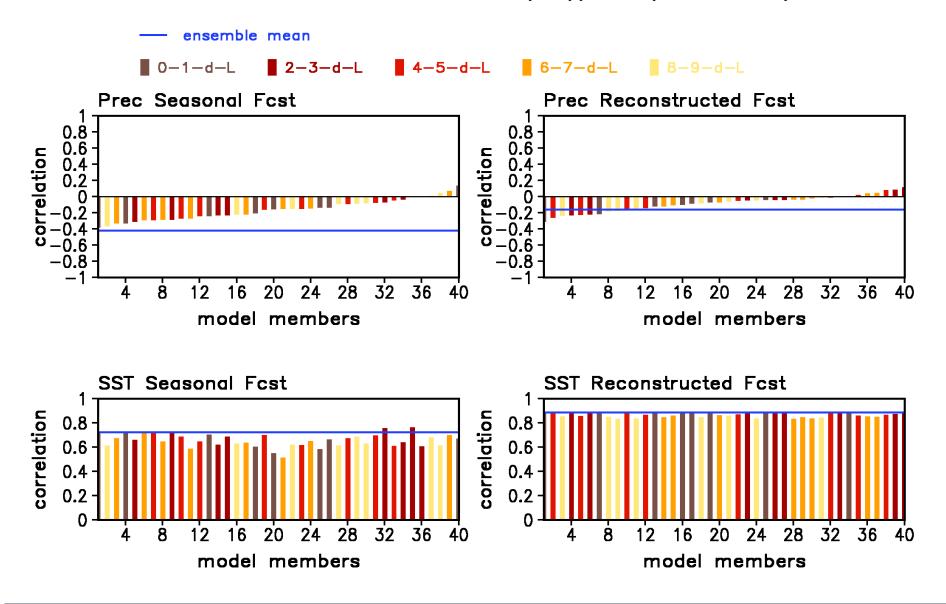
Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 z200(m) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



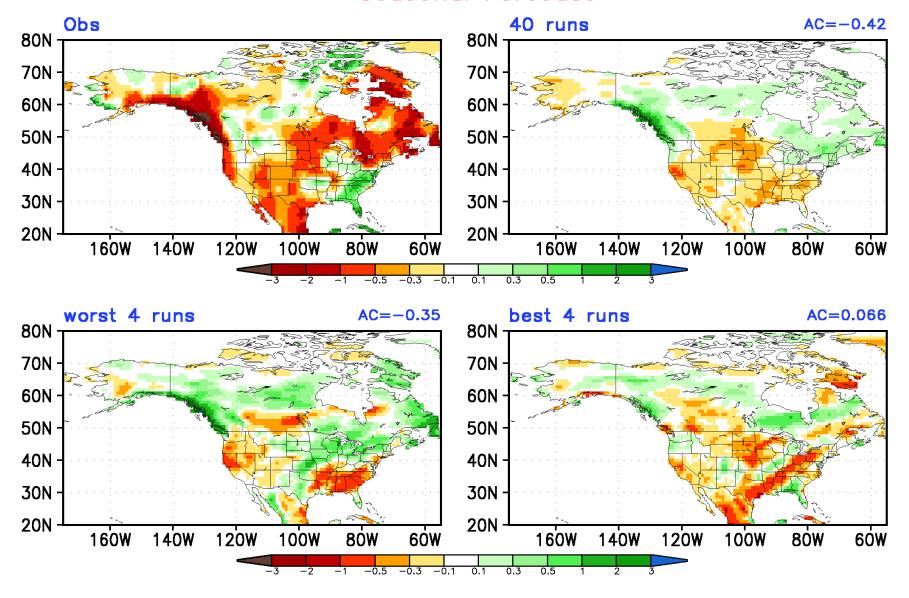
Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 z200(m) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



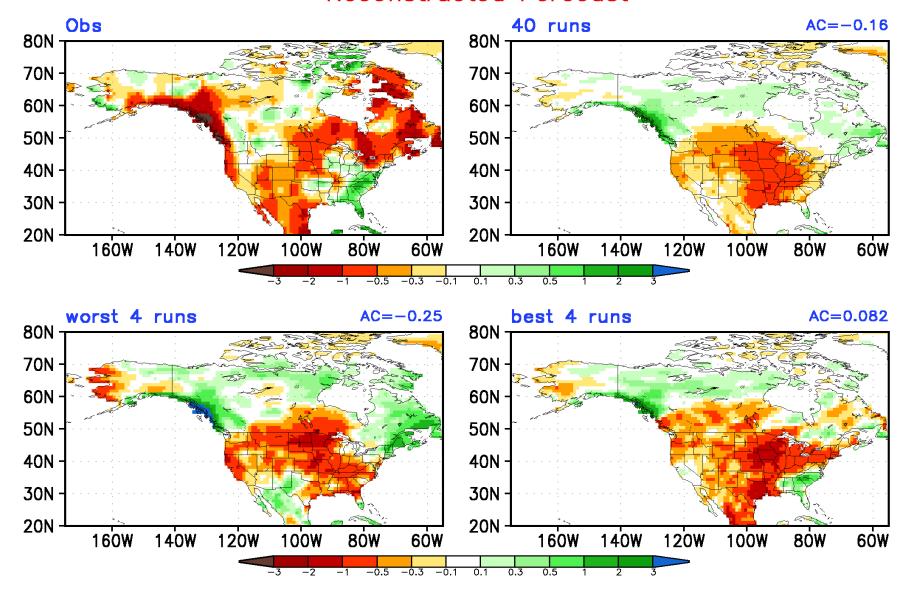
SON2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation —— Prec(NA)/SST(30S—30N)



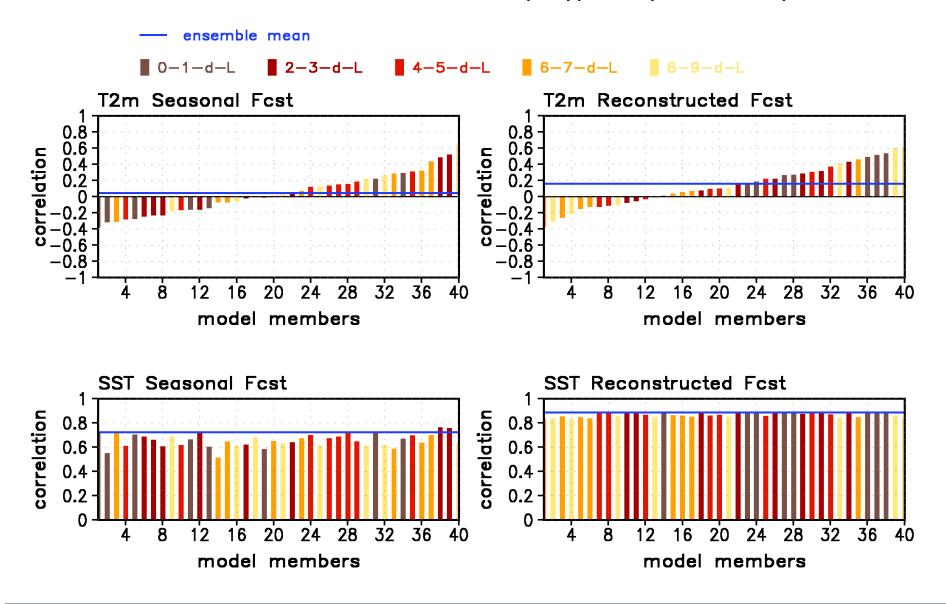
Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast



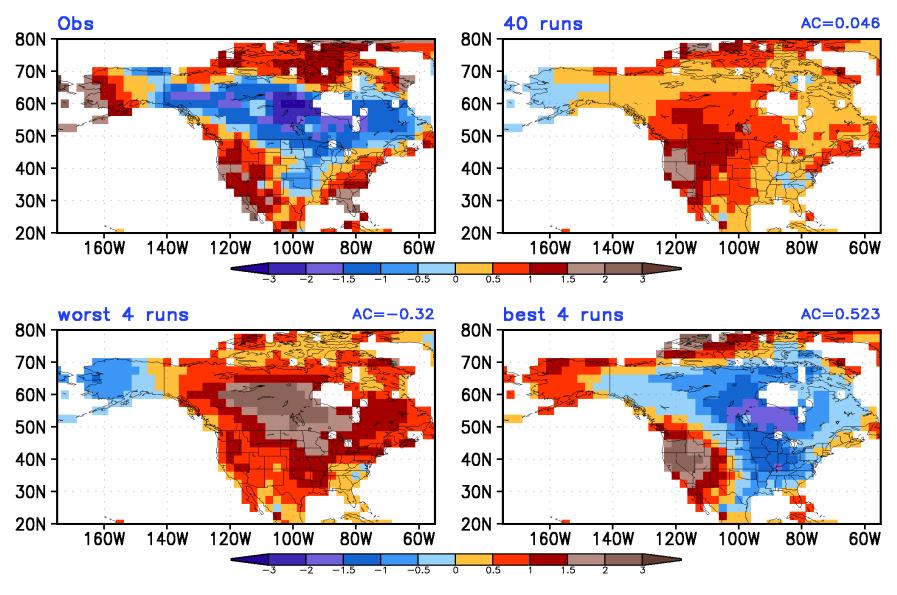
Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 Prec(mm/day) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



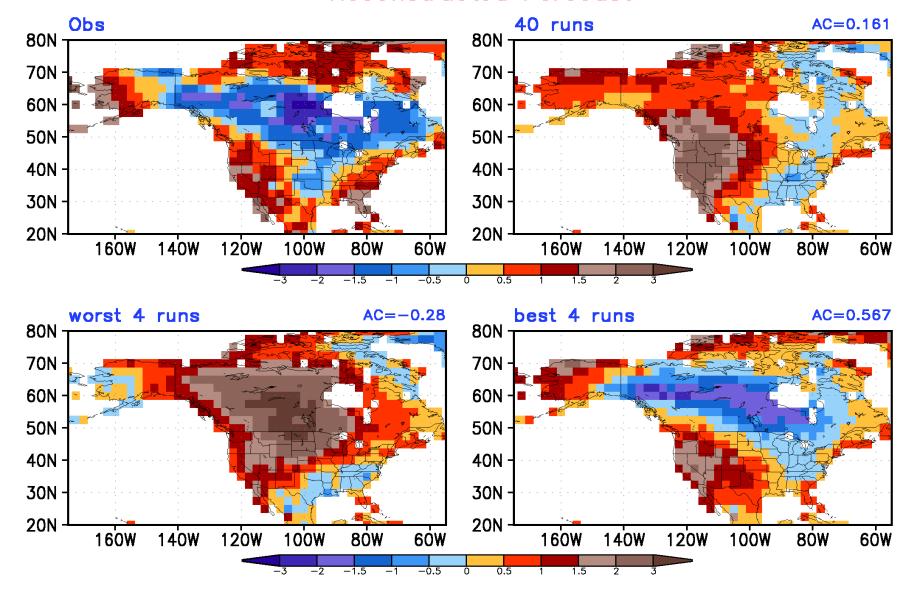
SON2020 Anomaly Correlation for Individual CFSv2 Forecast with Observation — T2m(NA)/SST(30S-30N)



Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 T2m(K) 40 runs/worst 4 runs/best 4 runs Seasonal Forecast

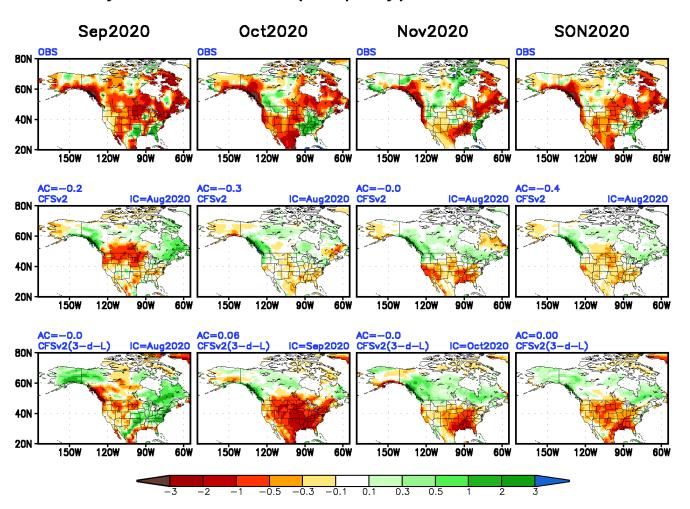


Observed & CFSv2 Forecast Ensemble Average Anomalies SON2020 T2m(K) 40 runs/worst 4 runs/best 4 runs Reconstructed Forecast



Monthly Means Prec(mm/day) Observed & Forecasts

Monthly SON2020 Prec(mm/day) Observed & Forecasts



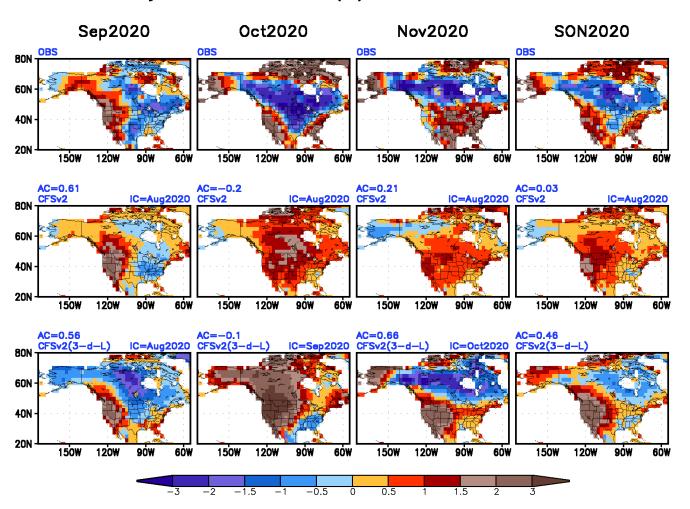
Top row: Observed anomaly

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

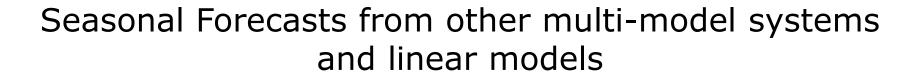
Monthly SON2020 T2m(K) Observed & Forecasts



Top row: Observed anomaly

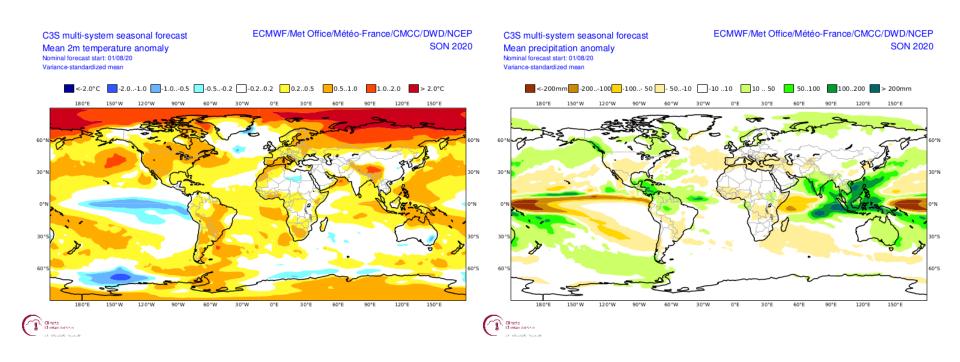
Middle row: CFSv2 seasonal forecasts from the initial conditions from the month prior to the target season.

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



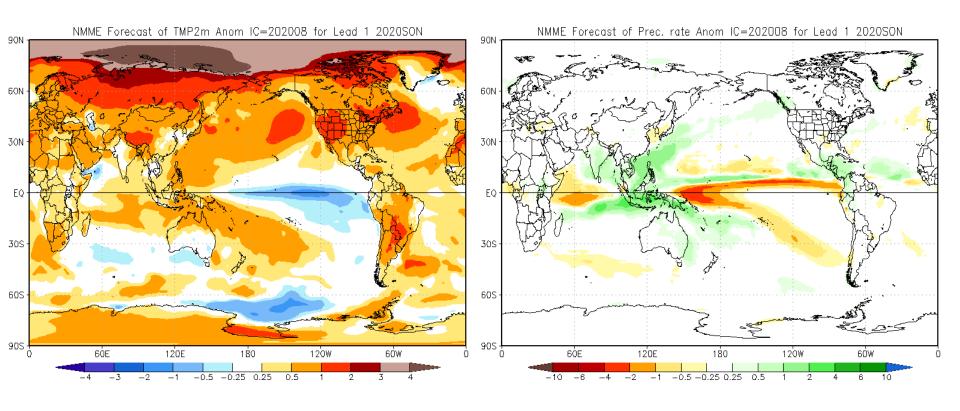
C3S Seasonal Forecast

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



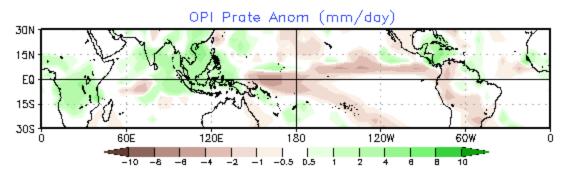
North American Multi-Model Ensemble Seasonal Forecast

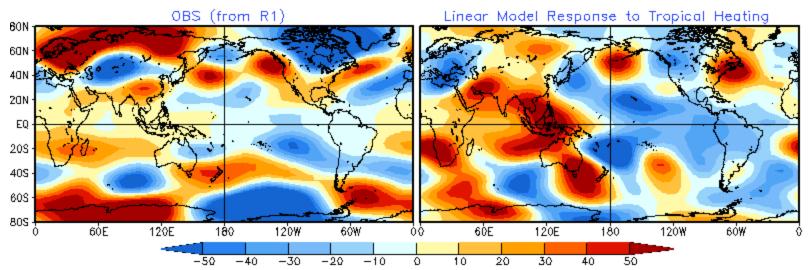
(https://www.cpc.ncep.noaa.gov/products/NMME/)



200mb Height from Linear Model

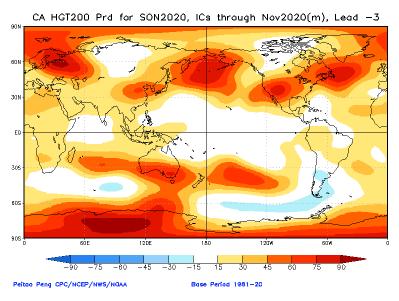
SON2020 200mb Eddy HGT(m)
OBS vs. Linear Model Response to Tropical Heating
Heating is converted from Prate in 15S—15N

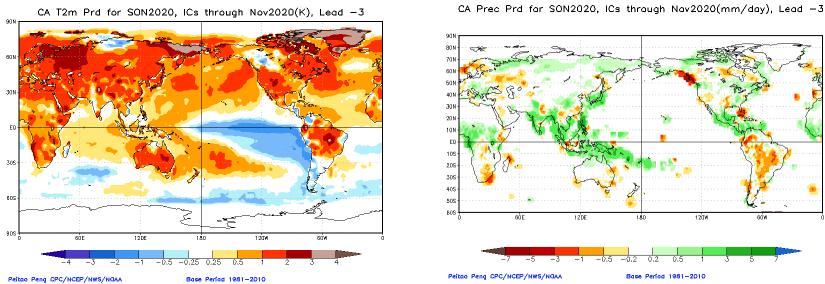




Pattern COR: global=0.20, tropics(30S-30N)=0.52

Seasonal Forecasts from the Constructed Analog Model



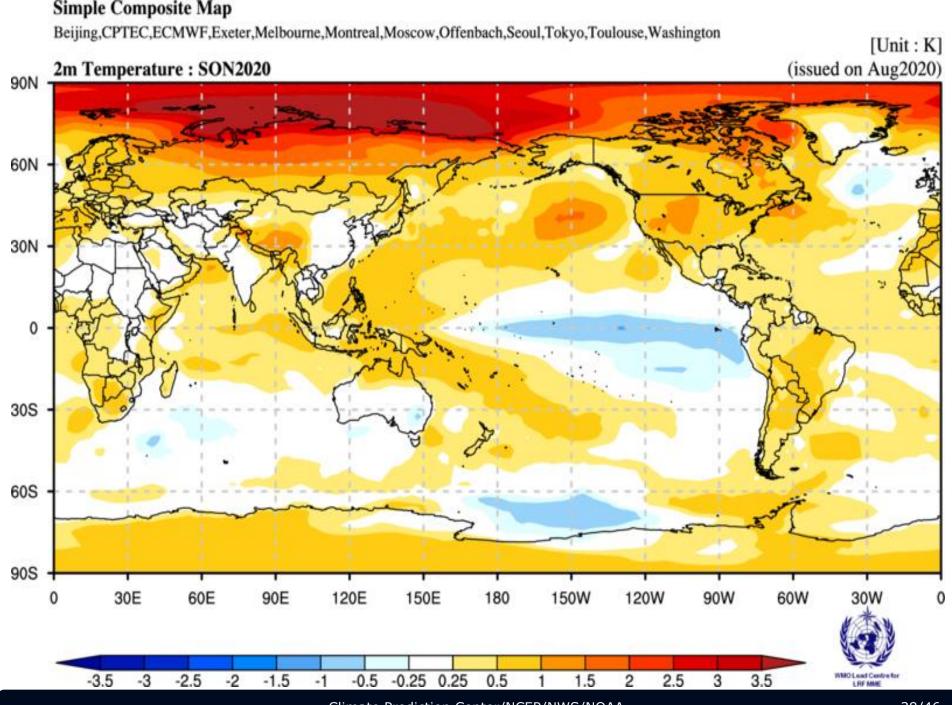


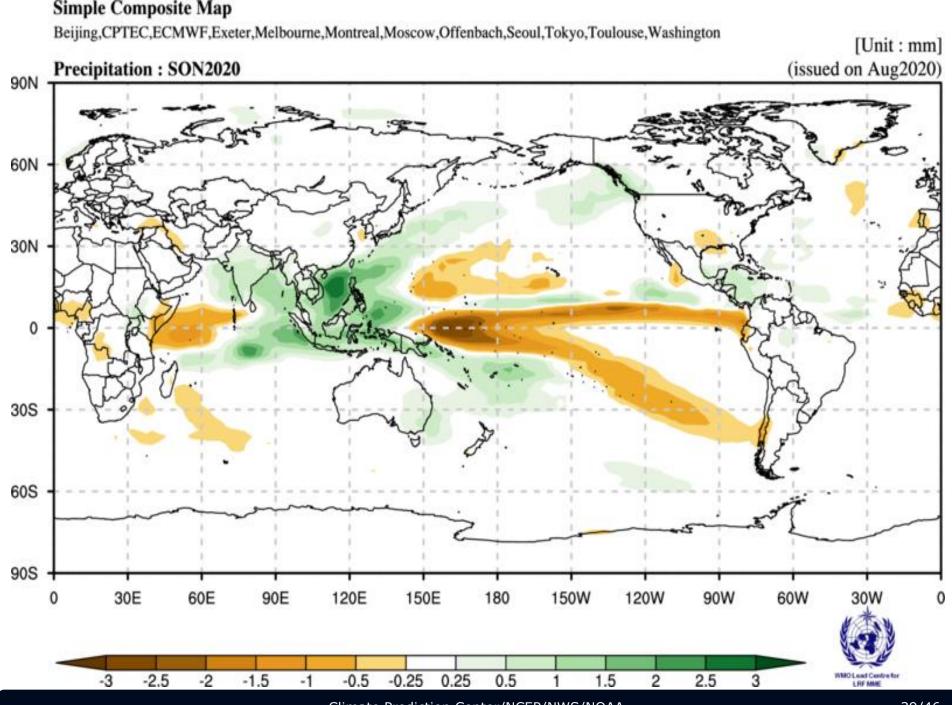
Seasonal Forecasts from WMO Lead Center for Long-Range Forecast Multi-Model Ensemble (LC-LRFMME)

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

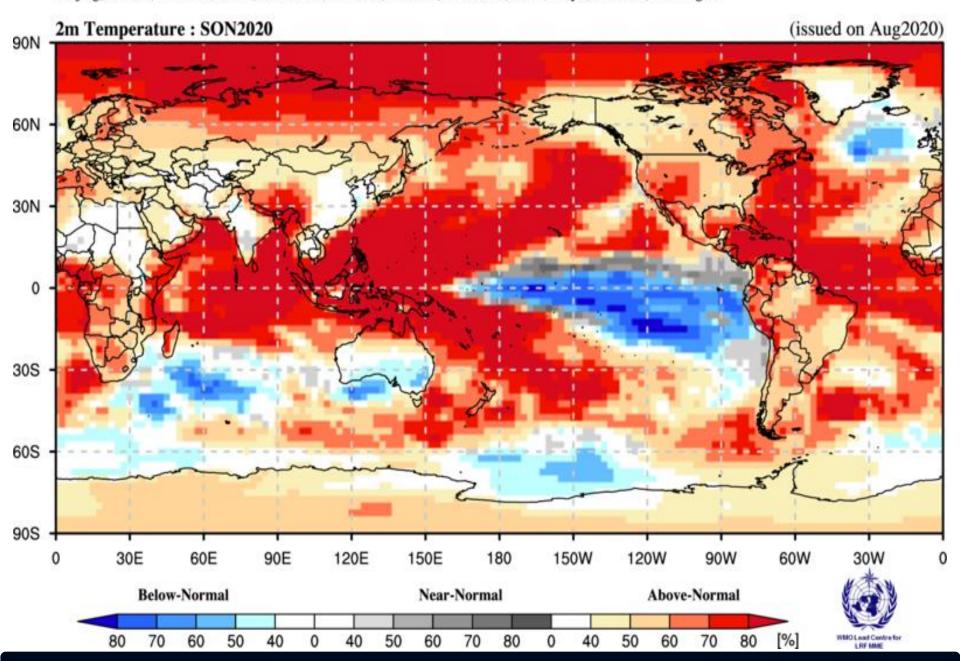
[•] For more information see visit Lead Center website; also see Graham, R., and Co-authors, 2011: New perspectives for GPCs, their role in the GFCS and a proposed contribution to a 'World Climate Watch'. Climate Research, 47, 47-55.





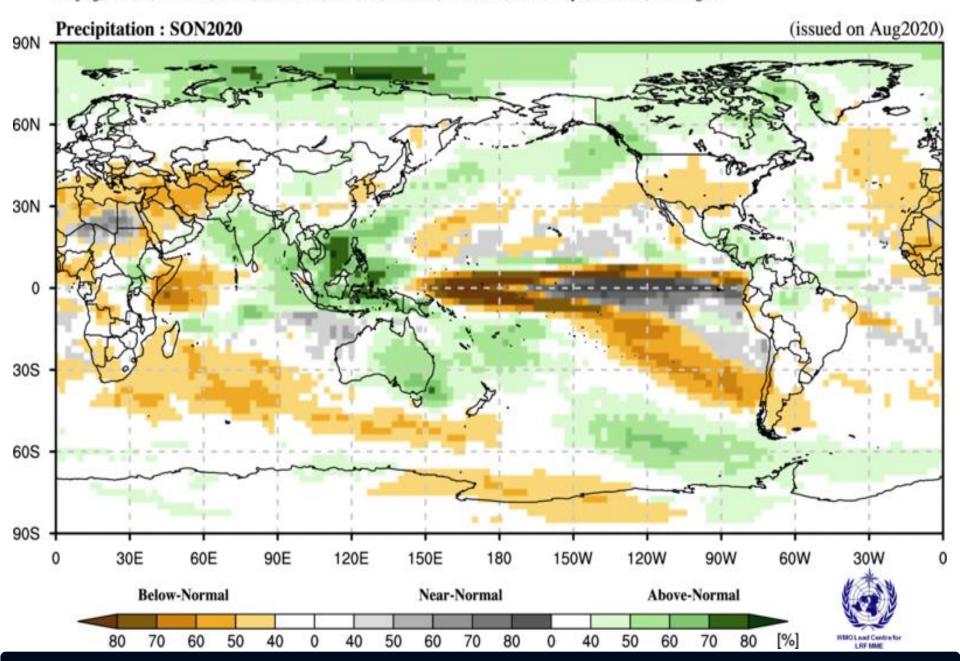
Probabilistic Multi-Model Ensemble Forecast

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)