

Quantifying sources of subseasonal prediction skill in CESM2

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To <u>quantify</u> how much subseasonal predictability comes from the <u>initial state</u> of atmosphere, land, and ocean/sea-ice.



Recreated figure by Paul Dirmeyer: representative of predictability of mid-latitude surface temperature over land



Additional Sources?

- Diagram assumes ATM, LND, OCN are independent, but they are not
- Land and atmosphere are tightly coupled near the surface: Land-atmosphere interact with each other very quickly
- Atmosphere & ocean interact as well (slower timescale)





Methods

- Calculate skill for 2m Temperature and Precipitation: Weeks 1- 2, Weeks 3-4, Weeks 5 -6
 - Anomaly Correlation Coefficient (ACC)
 - "Observations" come from ERA5 (2m Temperature) and GPCP (Precipitation) although CPC is comparable
- **Standard reforecast set** (realistic ATM, LND, OCN initialization)
 - 1999 2020; weekly initializations; 11 member ensemble
- Seven additional reforecast sets with various initial states set to climatology
 - climoATM
 - climoLND
 - climoOCN

- climoOCNclimoLND
- climoOCNclimoATM
- climoATMclimoLND
- climoALL (all components climo)



Annual Mean 2m Temperature ACC

- 2m T skill over land decreases very quickly with time
- Most skill: S. America, parts of Africa
- ACC over tropical oceans ~ 1



 No skill in climoALL runs through week 4



Annual Mean 2m Temperature ACC: Role of Atmosphere



NCAR UCAR

Weeks 5-6:

Mostly not

forecast

from standard

statistically different

Annual Mean 2m Temperature ACC: Role of Land







Soil Moisture Forecast



Soil Moisture forecast as expected:

Better forecast when initialized realistically (Not translating to better 2m T forecast - why?)



By week 2 (beyond), over S. America and central Africa, better soil moisture skill with climoLND initialization



Land-atmosphere Coupling

- High-soil moisture -> increased evaporative cooling -> Negative T anomalies
- Soil drought > positive T anomalies
- **Highest interaction:** both strong soil moisture-evaporation coupling and long soil moisture memory (Guo et al 2011)
- High soil moisture predictability in N. America throughout Spring and Summer, however Low atmospheric predictability due to low atmos-land coupling strength (Guo et al 2011)





Land-atmosphere coupling





CESM2: DJF

CESM2: Strong and negative coupling in summer hemisphere

SH: similar to CESM1 (and to OBS)

NH: weaker in CESM2 than in CESM1

Weaker impact of soil moisture on surface fluxes in NH winter in CESM2

Fig from Meg Fowler



Annual Mean 2m Temperature ACC: Role of Ocean





Role of ENSO

- Increased T skill in ENSO years;
- Benefit in weeks 3-4 and week 5-6
- Mostly in the Tropics: S. America and Africa
- Also in North-West N. America

Role of ocean beginning to grow in subseasonal window







Annual Mean 2m Temperature ACC



Response is pretty linear



Annual Mean Precipitation ACC



Response is pretty linear



Mid-Latitude (30N-60N) Annual Mean 2m Temperature ACC



Predictability sources very different from hypothesis in CESM



Discussion

- Results suggest that **atmospheric initial state is the dominant source of 2m air temperature predictability through weeks 3-4** for the majority of land areas, especially in the extratropics
- Land IC plays a small role in the CESM2(CAM6) subseasonal system and higher subseasonal skill for surface temperature can be obtained with climatological land initialization
 - Land-coupling not strong enough in CESM2
- Predictability from the ocean initial state comparable to that in the atmosphere in the tropics only

 slightly increased skill during active ENSO
- Atmospheric initial state is the main driver of subseasonal **precipitation** skill over extratropical land
 - except for South America and SE Asia/Australia
- Prediction skill seems to be fairly linear

Work published on March 4 in NPJ Climate & Atmospheric Science: https://www.nature.com/articles/s41612-024-00595-4



Data & Next Steps

• Raina Roy (Monash U) looking at MJO



- Available online: <u>https://www.earthsystemgrid.org/dataset/u</u> <u>car.cgd.cesm2.s2s_hindcasts.cesm2.climo.</u> <u>html</u>
- NCAR casper: /glade/campaign/cesm/development/crosswg/S2S/CESM2/
- DOI: https://doi.org/10.5065/0s63-m767

• Yanan Duan & Sanjiv Kumar (Auburn U) looking at soil moisture in detail



Sources of Predictability





(1) standard =
$$\operatorname{Clim}_{ALL} + \mathbf{V}_{A} + \mathbf{V}_{L} + \mathbf{V}_{O} + C_{AL} + C_{AO} + C_{LO}$$

(2) <u>climoATM</u> = Clim_{AII} + V_I + V_O + C_{AL} + C_{AO} + C_{LO} (3) <u>climoLND</u> = Clim_{ALL} + V_A + V_O + C_{AL} + C_{AO} + C_{LO} (4) <u>climoOCN</u> = Clim_{A11} + V_A + V₁ + C_{A1} + C_{A0} + C₁₀

(5) <u>climoOCNclimoLND</u> = $\operatorname{Clim}_{ALL}$ + \bigvee_{A} + C_{AL} (6) <u>climoOCNclimoATM</u> = $\operatorname{Clim}_{ALL}$ + \bigvee_{L} + C_{AL} (7) <u>climoATMclimoLND</u> = $\operatorname{Clim}_{ALL}$ + \bigvee_{C} + C_{AO}

(8) <u>climoALL</u> = Clim

(9) sum = Clim_{ALL} + V_A + V_L + V_O + C_{AL} + C_{AO} (10) sum ≈ standard

When climatological initial conditions are used for a single component, we can remove that component's variability term. We assume that the average coupling between the components do not change much between the reforecast sets.

When climatological initial conditions are used for two components, we assume their two variability terms are negligible, along with their shared coupling term

Assuming that the land-ocean coupling C₁₀ is nearly zero over land, we can then use the earlier variability results (V₁ and V₀) to solve for C_{A1} and C_{A0} . If the linearity assumption holds, we should be able to retrieve the standard ACC by adding the individual components.

