Validation/Evaluation

- 1) Baseline evaluation of Forecasts
- skill (anomaly correlation, rmse, BSS, etc)
- potential predictability
- forecast spread and consistency with obs (e.g., rmse, rank histograms)
- realism of variance, autocorrelation, extremes
- climate drift

2) Forecasts of opportunity (assess conditional skill)

- ENSO/other teleconnections (e.g., soil moisture teleconnections?)
- In-depth probabilistic evaluation of forecasts
 - (reliability, resolution, ROC, etc)

3) Ability of model to simulate key modes of climate variability

- ENSO and other SST modes
- MJO, PNA,NAO, etc (important for subseasonal as signal, important for seasonal and longer as noise)

Results from Becker et al. with a focus on the GEOS-5 S2S-1.0 Model Performance

Becker, E., H. van den Dool, Q. Zhang, 2014: Predictability and Forecast Skill in NMME. J. Climate, 27, 5891-5906.

Various Measures

(see Appendix of Becker et al)

- Skill
 - anomaly correlation
 - rmse
- Potential predictability
 - homogeneous (heterogeneous) anomaly correlation
 - homogeneous (heterogeneous) rms
- Forecast Spread
 - rms about ensemble mean
 - compare to rmse to assess adequacy of dispersion
- Variance of single ensemble member
 - (singmember_SD)² = (EM_SD)²+Spread²
 - can compare to observed variance

NINO3.4 region SST: Lead 1 month, aggregated over all seasons (3month): 1982-2010; 1982-2009 for cfsv1

Estimates of Estimates of skill potential predictability A. SST Nino3.4 Region Season 1 <u>B</u> <u>A</u> cfsv1 cfsv2 cmc1 cmc2 gfdl nasa ncar obs (EM AC) EM RMSE (C) EM SD 1.3 0.65 cfsv1 EM 94 85 78 87 82 82 1.12 74 86 0.610.75 1.3 93 cfsv2 EM 69 78 76 80 86 80 82 0.58 1.03 0.41 1.0 cmc1 EM 84 76 96 93 82 90 83 87 0.430.83 1.3 0.45 83 cmc2 EM 73 92 97 80 88 82 85 0.57 1.131.3 0.46 gfdl EM 76 78 83 87 80 1.1382 94 80 0.65 0.62 1.4 87 85 83 88 84 98 1.17 nasa EM 84 88 0.56 79 82 82 79 95 ncar EM 78 85 0.67 1.14 0.22 1.3 80 NMME=89 76 76 75 78 singmem AC 84 83 85 singmem RMSE 0.74 0.70 0.460.63 0.73 0.62 0.70 singmem & obs SD 1.191.12 0.85 1.171.22 1.15 0.891.16

FIG. 9. As in Fig. 1, but for sea surface temperature in (a) the Niño-3.4 region (aggregated results for all grid points in the area 5°S-5°N, 170°-120°W) and (b) the Niño-3.4 index (area-averaged SST).

- A. dispersion/rmse
- B. Single member SD/obs SD

All models are under dispersive

Most models have too much variability

NH SST: Lead 1 month, aggregated over all seasons (3month): 1982-2010 (NMME models), 1982-2009 for cfsv1



- A. dispersion/rmse
- B. Single member SD/obs SD

All models are under dispersive

Most models have realistic variability

T2m NH land: Lead 1 month, aggregated over all seasons (3month): 1982-2010; 1982-2009 for cfsv1

Estimates of potential predictability	: Y								Estin skill	nates d	of	
TMP2m Northern Hemisp							nere Season 1					_
∠ ∠	cfsv1	cfsv2	cmc1	cmc2	gfdl	nasa	ncar	obs (EM AC)	M RMSE (C)	EM SD	<u>A</u>	<u>B</u>
cfsv1 EM	29	12	12	11	10	16	8	12	1.41	0.59	0.91	1.0
cfsv2 EM	14	38	16	24	26	28	1	29	1.32	0.62	0.91	1.0
cmc1 EM	11	14	30	19	18	21	5	17	1.38	0.60	0.80	.91
cmc2 EM	10	23	21	38	27	27	3	27	1.36	0.72	0.81	1.0
gfdl EM	9	23	16	25	36	26	0	25	1.38	0.77	0.96	1.1
nasa EM	12	25	19	24	25	39	2	23	1.37	0.69	0.84	1.0
ncar EM	6	1	4	6	3	3	19	0	1.62	0.84	0 74	1.1
singmem AC	3	13	7	16	13	11	0	NMME=29			.,	
singmem RMSE	1.91	1.79	1.77	1.76	1.90	1.79	1.99				↑	
singmem & obs SD	1.41	1.35	1.26	1.38	1.54	1.34	1.46	1.38				T

FIG. 1. Area-aggregated results for 2-m temperature, land-only Northern Hemisphere, 23°-75°N, averaged over the 12 lead-1 seasons. The seven rows and columns in black are results for predictability. The orange highlighted

- A. dispersion/rmse
- B. Single member SD/obs SD

All models are somewhat under dispersive

Variance is about right for most

Precip NH land: Lead 1 month, aggregated over all seasons (3month): 1982-2010; 1982-2009 for cfsv1

Estimates of potential predictability										Estimates of skill			
		PRATE Northern Hemisphere Season 1											
<u> </u>	cfsv1	cfsv2	cmc1	cmc2	gfdl	nasa	ncar	obs (EM AC)	EM RMSE	EM SD	<u>A</u>	<u>B</u>	
cfsv1 EM	24	11	8	12	11	10	4	10	0.41	0.21	1.54	1.3	
cfsv2 EM	13	20	7	11	9	10	4	12	0.38	0.15	1.35	1.3	
cmc1 EM	6	6	16	14	9	7	6	9	0.40	0.16	1.14	1.0	
cmc2 EM	10	6	13	25	13	10	5	11	0.40	0.18	1.16	1.0	
gfdl EM	10	8	9	14	22	11	5	12	0.40	0.20	1.3	1.1	
nasa EM	9	8	6	10	10	18	4	9	0.40	0.19	1.3	1.1	
ncar EM	4	3	5	6	5	3	12	4	0.45	0.25	1.46	1.1	
singmem AC	3	4	3	6	5	4	2	NMME=16			±.10		
singmem RMSE	0.67	0.61	0.56	0.56	0.60	0.61	0.65				↑		
singmem & obs SD	0.57	0.50	0.42	0.43	0.48	0.48	0.54	0.37				T T	

FIG. 4. As in Fig. 1, but for precipitation rate over Northern Hemisphere land. RMSE is in millimeters per day.

- A. dispersion/rmse
- B. Single member SD/obs SD

All models are over dispersive

Variance is about right for most



FIG. 10. As in Fig. 3, but for sea surface temperature: (top left) Pacific Ocean (23°-75°N); (top right) Atlantic Ocean (45°-75°N); and (bottom) Niño-3.4 region.

Blue – NDJ,DJF,JFM GREEN - FMA,MAM,AMJ RED - MJJ,JJA,JAS

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FIG. 3. The 2-m temperature homogeneous predictability AC (x axis) vs forecast skill AC (y axis). Colors indicate season: boreal winter seasons (November–January, DJF, and January–March) are blue, spring seasons (FMA, MAM, and April–June) are green, summer seasons [May–July (MJJ), JJA, and July–September (JAS)] are red, and autumn seasons (August–October, SON, and October–December) are orange. The shades of each color vary from lighter (first 3-month period; e.g., MJJ) to darker (third 3-month period; e.g., JAS). Each individual season has seven points, one for each model. The linear fit is depicted by the red line. (left) Southeastern Asia (5°–50°N, 70°–145°E) and (right) extratropical North America (north of 23°N; Greenland not included).

Lead 1 forecasts: results for each model and by season

Blue – NDJ,DJF,JFM GREEN - FMA,MAM,AMJ RED - MJJ,JJA,JAS ORANGE – ASO,SON,OND

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FIG. 6. As in Fig. 3, but for precipitation rate.

Lead 1 forecasts: results for each model and by season

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FIG. 2. Forecast skill measured by the anomaly correlation for NMME 7-model EM prediction of 2-m temperature. Four seasons are shown: lead-1 DJF, MAM, JJA, and SON. ACs are multiplied by 100. Note that the domain is near global: the NMME score of AC = 0.29 quoted in Fig. 1 corresponds to the extratropical Northern Hemisphere 11 portion of the above maps.

Prate NMME ensemble AC



JJA

SON





FIG. 5. As in Fig. 2, but for precipitation rate. The NMME skill of AC = 0.16 quoted in Fig. 4 refers to the extratropical NH portion of the above maps.

SST NMME ensemble AC









FIG. 8. As in Fig. 2, but for sea surface temperature. The NMME skill of AC = 0.50 quoted in Fig. 7 refers to the extratropical NH portion of the above maps.

Other issues (from NMME website)

Season 1 tmpsfc forecast DJF



Season 1 tmpsfc forecast JJA



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Comments

- GEOS-5 model is in the mix of the other NMME models in terms of various basic skill measures
 - It is one of the best for NINO3.4 (potential predictability suggests some room for improvement though not much)
 - model is under-dispersive for NINO3.4 SST (like all the models) despite having too much variance impacts reliability and consistency of forecasts (how do we increase the spread?)
 - other ocean basin SST show considerably less skill but also much more room for improvement based on potential predictability
 - The warm pool and eastern Indian Ocean SST show almost no skill for all models an area where the atmosphere is known to be sensitive to SST
 - skill of T2m over North America for NMME is surprisingly poor in winter and summer likely linked to errors in ENSO teleconnections in winter, and poor or no land initialization for some of the NMME models in summer
 - Skill of precipitation forecasts is generally low in NMME models (JJA over North America skill is especially low and surprising – again suggests need for improved land initialization
- The generally low skill at seasonal time scales over North America and elsewhere (T2m and Precipitation) suggests the need to focus on forecasts of opportunity- conditional skill can be much higher
 - Requires a focus on assessing impacts and predictability of various teleconnections including ENSO
 - Requires an in-depth analysis of the probabilistic quality of the forecasts (reliability, resolution, spread, BSS, etc)
- Suggest developing a comprehensive web-based baseline validation/evaluation capability (with flexibility to easily assess different regions, lags, seasons, quantities, models)