



Multi-model Prediction on Subseasonal Time-scales at the US NOAA Climate Prediction Center:

Approaches to Calibration and the Identification of Forecasts of Opportunity

Dan C Collins, Sarah Strazzo, Emerson LaJoie, Emily Becker, and Jon
Gottschalck

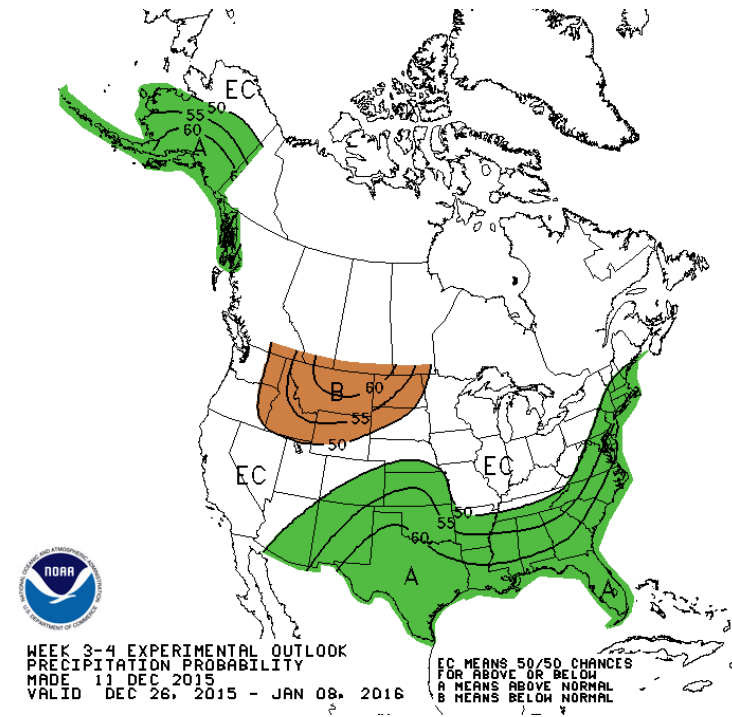
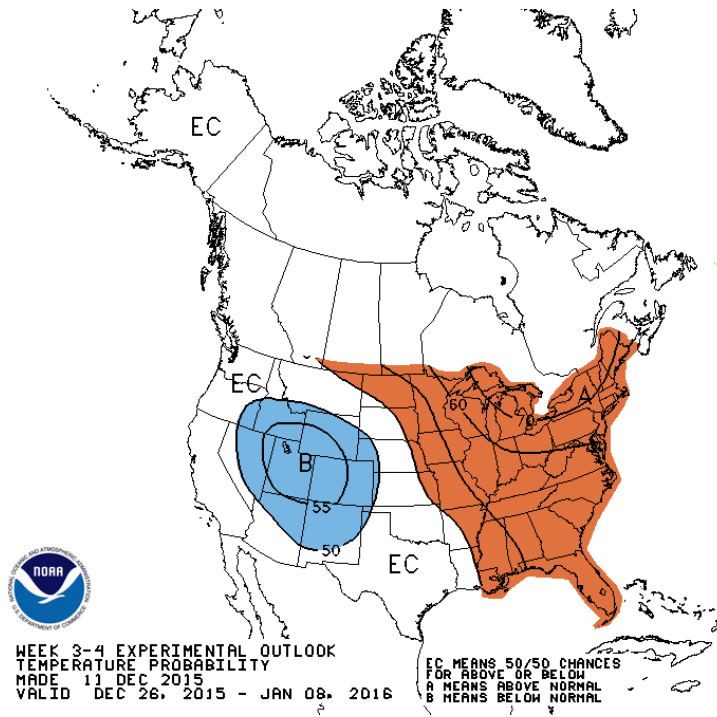
NOAA Climate Prediction Center



MME forecasts in support of week 3-4 US NOAA operational outlooks:

Above and below normal

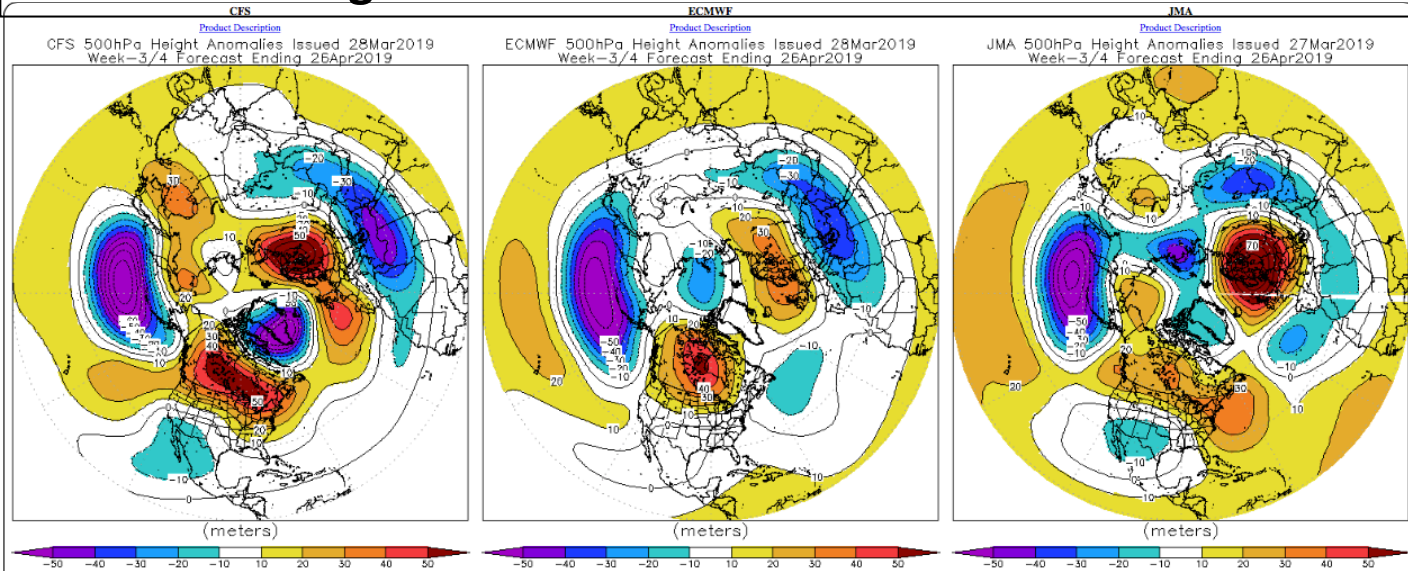
Examining extremes and hazards into week 3&4



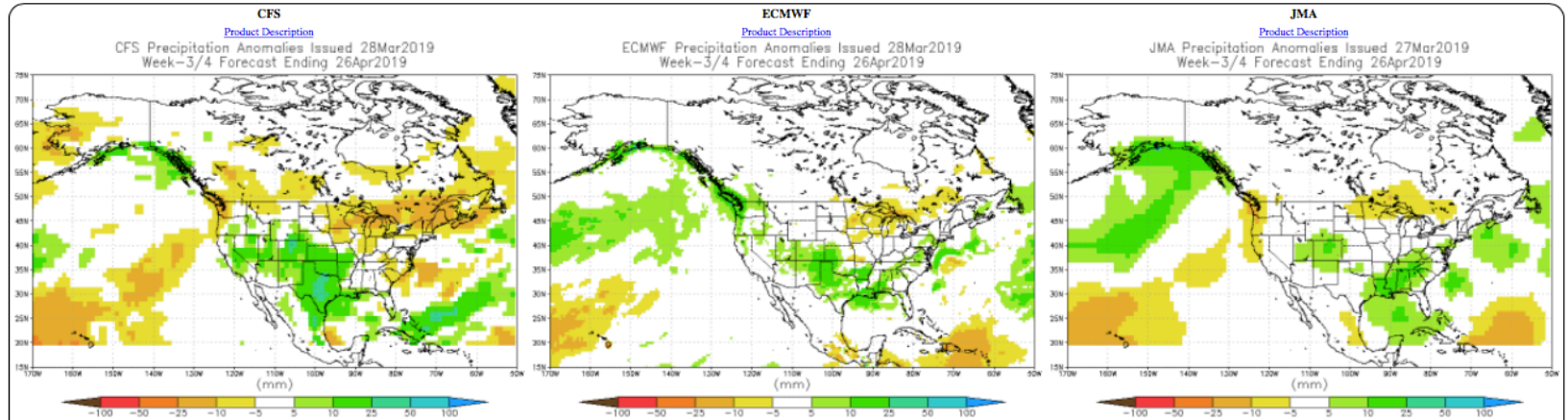
- Probability of above and below normal temperature and precipitation
- Use a combination of dynamical and statistical model forecasts
- MME guidance plays a primary role in the subseasonal forecast

Operational model guidance: NCEP CFS, ECMWF & JMA

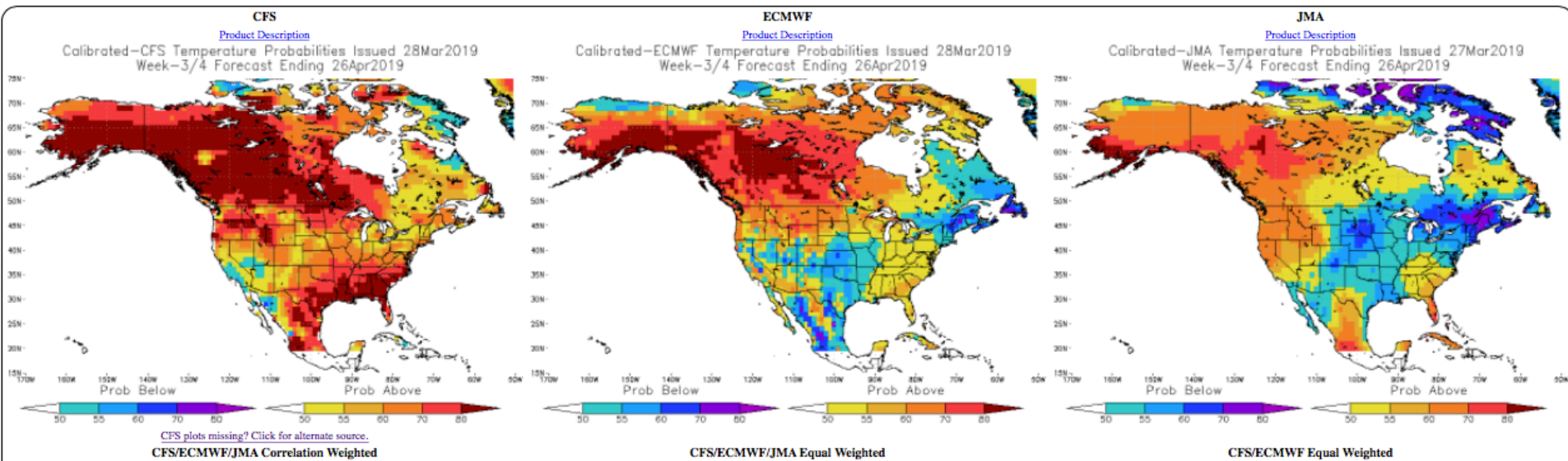
- 500 hPa height anomalies



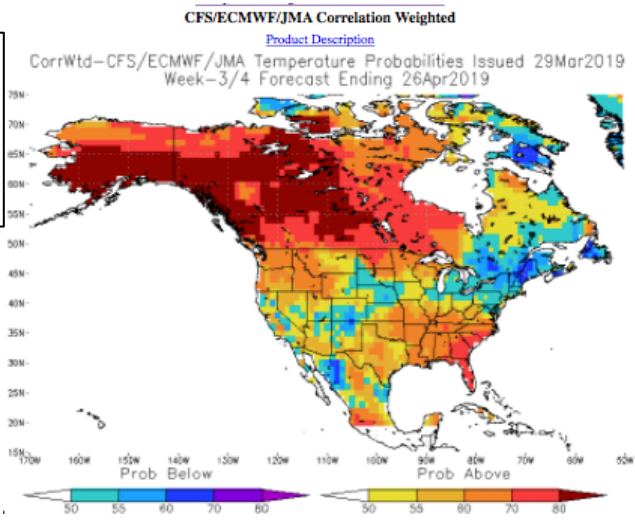
- Precipitation anomalies



- Each dynamical model is bias corrected using model hindcast
- Calibrated PDFs made using hindcast skill* (heterogeneous hindcast samples)



Correlation weighted MME:
Above and below median
2-m temperature

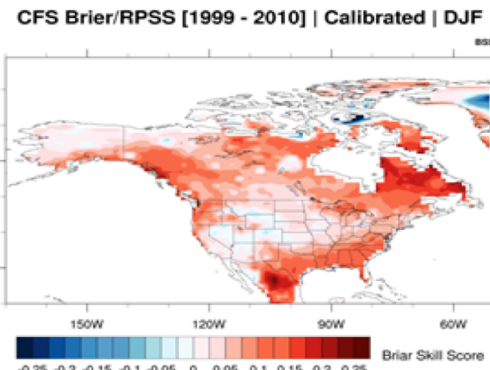
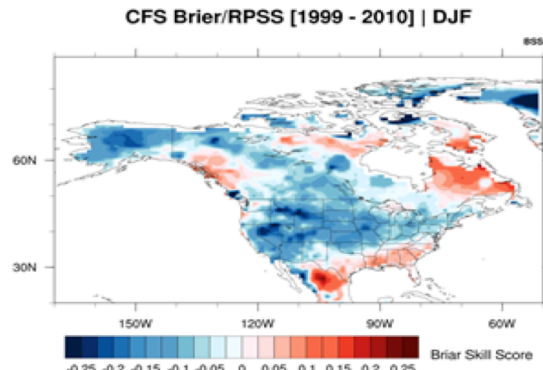


**Ensemble Regression,
Unger et al., 2009*

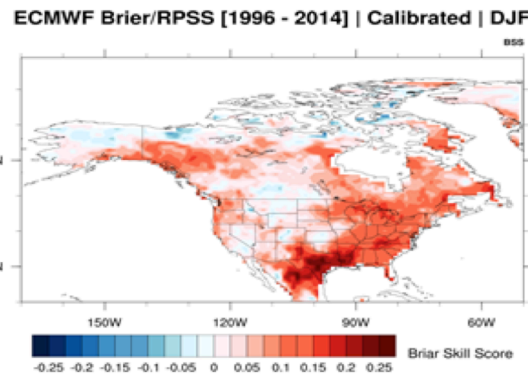
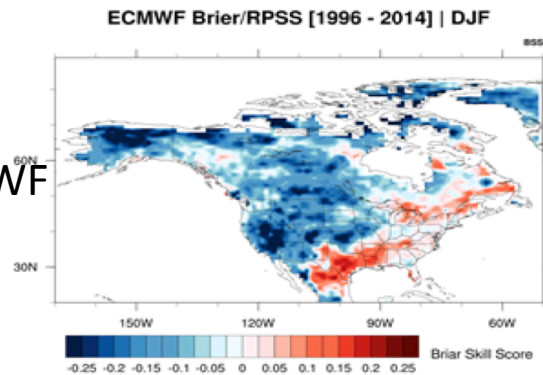


Brier Skill Scores of DJF temperature forecasts :

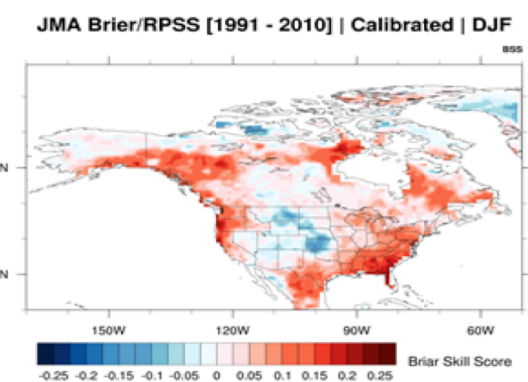
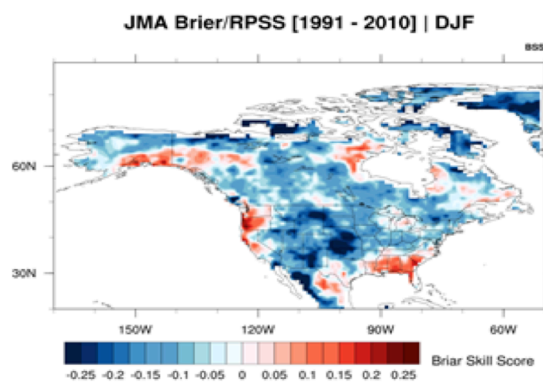
CFS



ECMWF

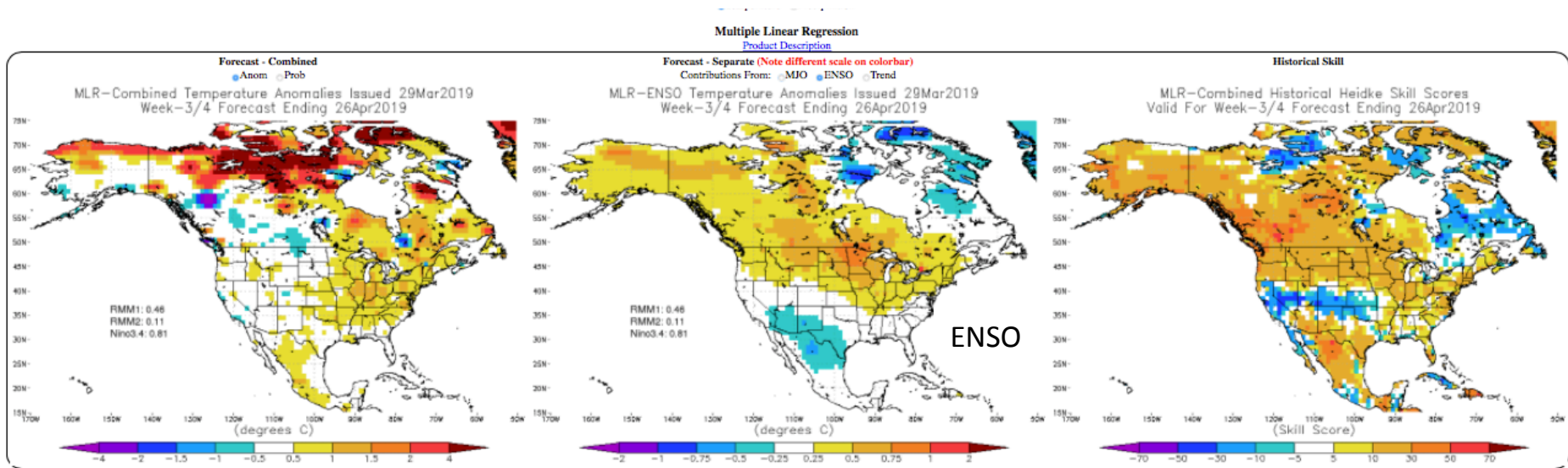


JMA



No calibration (left)
and calibrated (right)

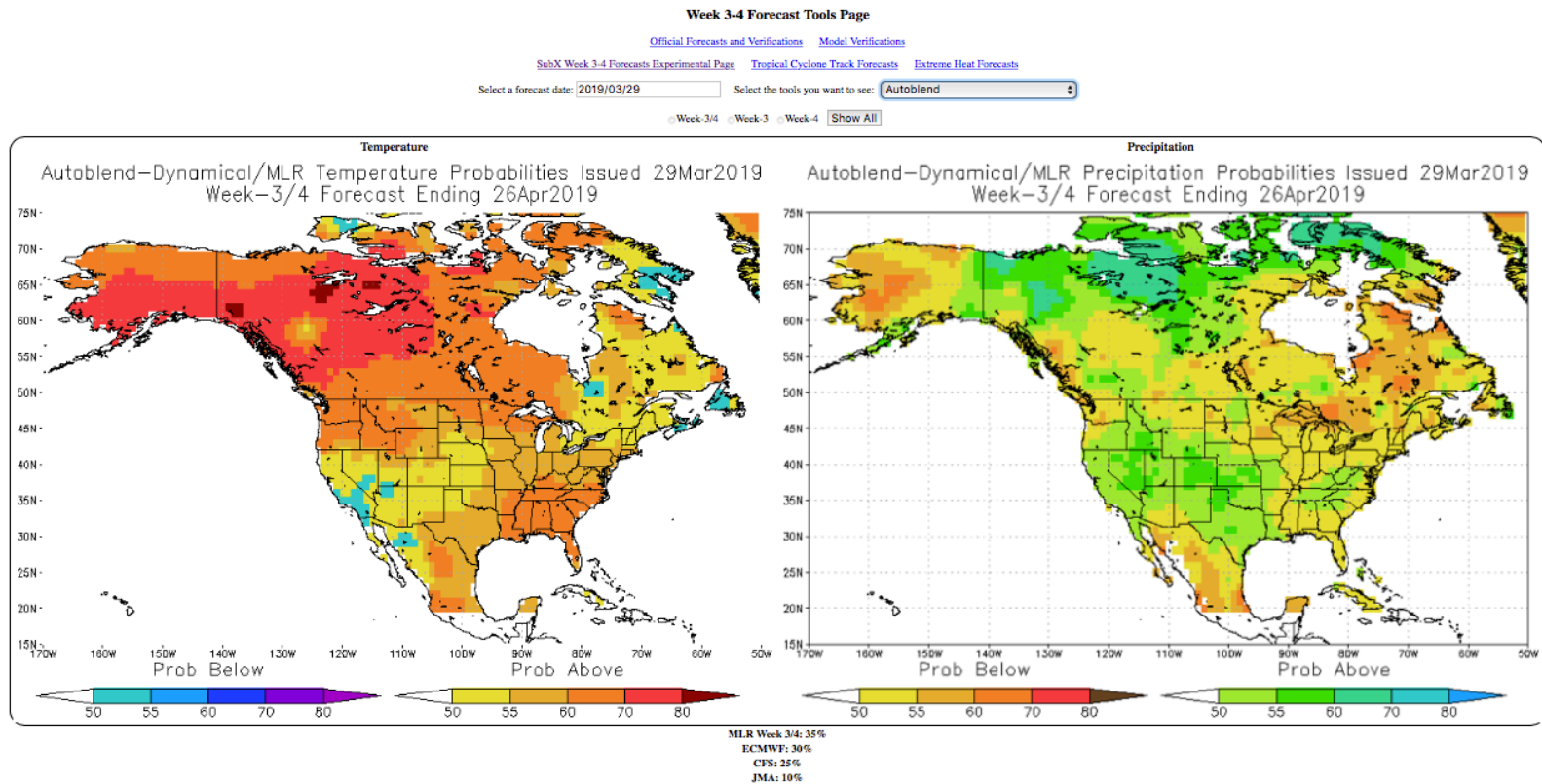
Statistical model guidance: Multiple Linear Regression



- Based on initial state of MJO and ENSO: RMM and Nino 3.4 index predictors
- Regression of climate indices to local temperature and precipitation anomalies
- Comparable skill to dynamical models

Automated Blend: Combining statistical and calibrated dynamical models

$$w_1 * \text{MLR} + w_2 * \text{ECMWF} + w_3 * \text{NCEP_CFS} + w_4 * \text{JMA}$$



- Experimenting with methods for determining weights, and...
- Combining calibrated forecasts to make a calibrated blend

SubX vs. S2S archives

- SubX protocol for hindcast years & frequency (at least weekly)
 - S2S has heterogeneous hindcasts
- SubX combines operational (NCEP, ECCC) and experimental models
 - S2S models are operational
- SubX forecast data is available in near real time to any user, through the IRI Data Library:
<http://iridl.ldeo.columbia.edu/SOURCES/.Models/.SubX/>
 - S2S forecasts available on a 3-week delay

SubX : The Subseasonal Prediction Experiment

- Providing a protocol, database and test bed for hindcast and real-time subseasonal forecasts

- Hindcasts (1999-2015)
 - More than **18 months** of **weekly** real-time forecasts
 - **7 operational or experimental** ensemble models

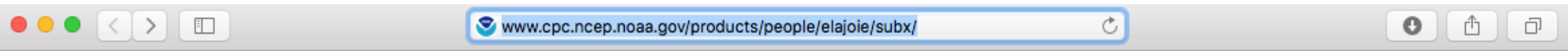
- Facilitating MME and examining the value to subseasonal
 - Model calibration and multi-model ensemble combination
 - Addition of new experimental systems (e.g. NCAR CESM2)
 - Assessing the added value to operational models

- Supporting NOAA National Weather Service /Climate Prediction Center, **Week 3-4 Outlooks**

<u>1 week of SubX</u>	Jan 2	Jan 3	Jan 4	Jan 5	Jan 6	Jan 7	Jan 8	Jan 9 Forecast Day	Week 3-4 Outlook: Jan 24 – Feb 06
Day of the week & Days to Target Dates	Fri 22:35	Sat 21:34	Sun 20:33	Mon 19:32	Tues 18:31	Wed 17:30	Thurs 16:29	Fri 15:28	2 weeks: Sat + 13 days (Fri) → WK34
Center-Model ----- Forecast Initialization Period -----									
ECCC-GEM 4 members 32 days								Forecast Day	Hindcast & Real-time forecast
EMC-GEFS 11 members 35 days								Forecast Day	Variable hindcast days
ESRL-FIMv2 4 members 32 days								Forecast Day	Variable real-time & hindcast days
NASA-GEOS 4 members 45 days								Forecast Day	
NCEP-CFSv2 4 members 44 days								Forecast Day	
NRL-NESM 4 lagged members 45 days								Forecast Day	
RSMAS-CCSM4 3 members 45 days								Forecast Day	
<i>Coming in next year:</i> CESM-46LCAM5 10 members 45 days								Forecast Day	<i>MME for week 3-4 collected for each week in the hindcast</i>
CESM-30LCAM5 10 members 45 days								Forecast Day	



CPC SubX guidance



SubX Week 3/4 forecasts

[SubX : The Subseasonal Prediction Experiment Project](#)

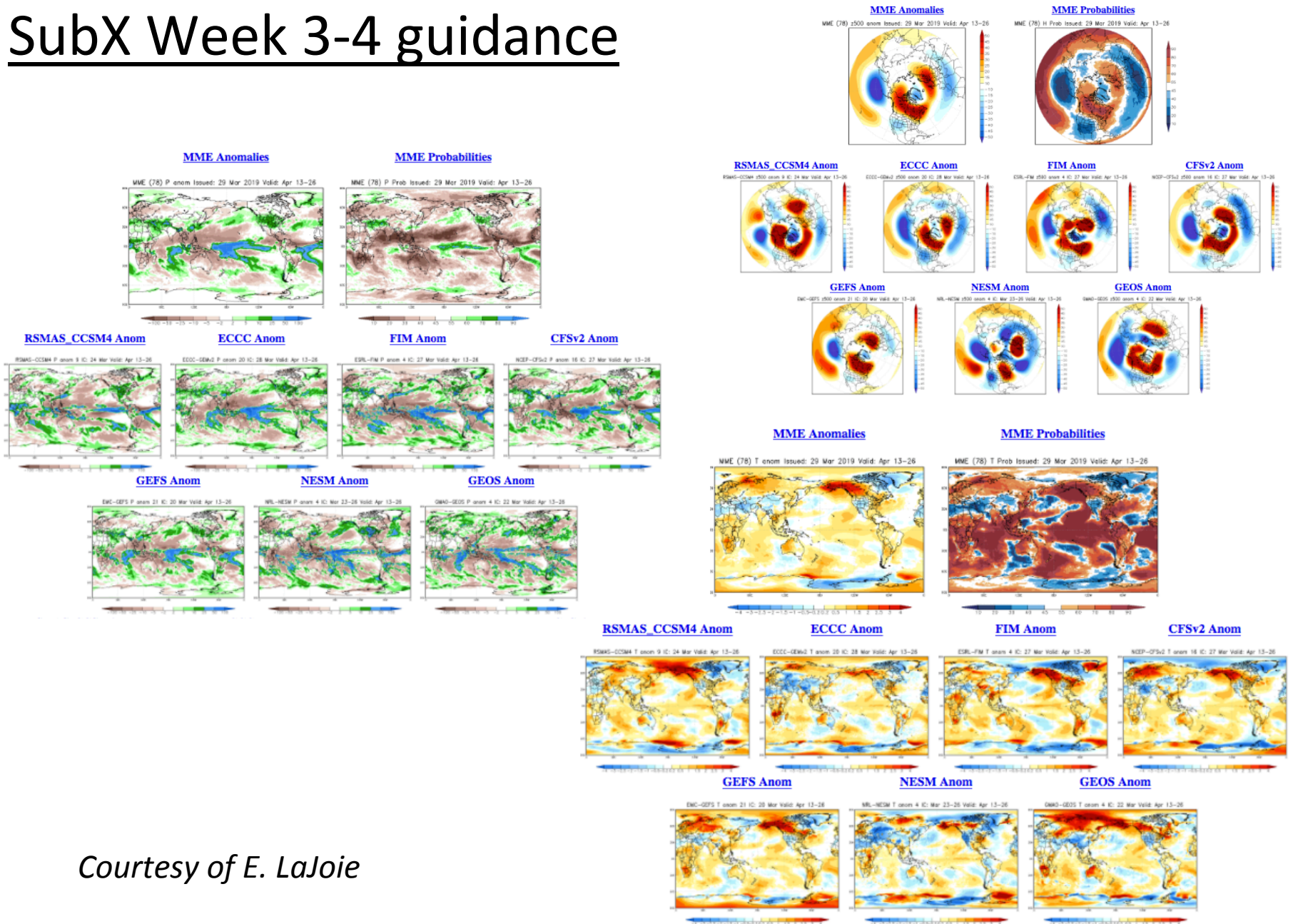
[IRI Data Library](#)

[Week 3/4 Operational Model Forecasts](#)

North America	Global
500-hPa height	500-hPa height
2-m Temperature	2-m Temperature
Precipitation	Precipitation

<http://www.cpc.ncep.noaa.gov/products/people/elajoie/subx/>

SubX Week 3-4 guidance

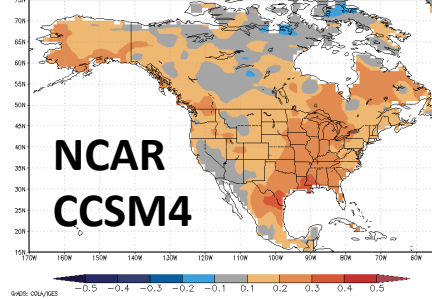


Courtesy of E. LaJoie

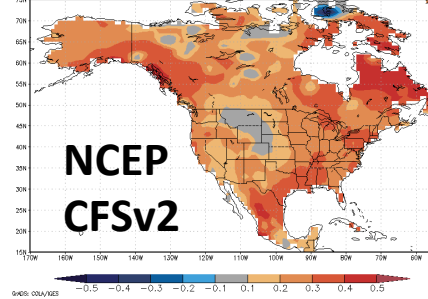
Anomaly Correlation by model & MME (DJF)

➤ MME outperforms any individual model

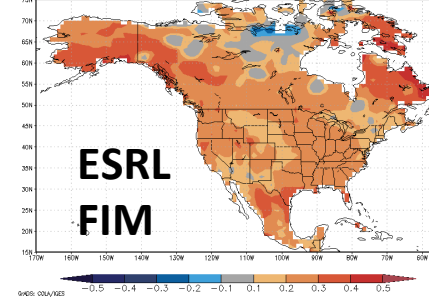
Weighted ACC-DJF TAS RSMAS-CCSM4: Area-avg Score for NA: 0.151



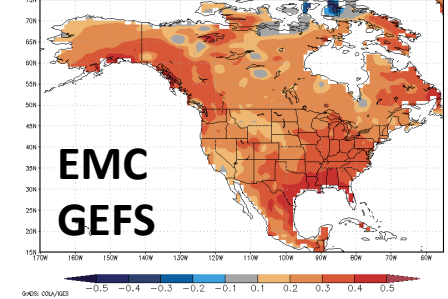
Weighted ACC-DJF TAS NCEP-CFSv2: Area-avg Score for NA: 0.275



Weighted ACC-DJF TAS ESRL-FIMv2: Area-avg Score for NA: 0.2546

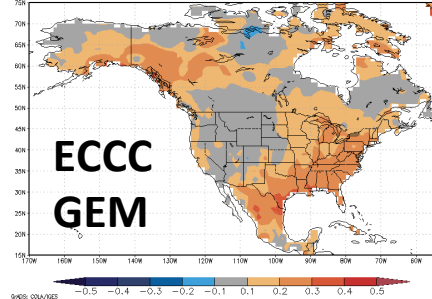


Weighted ACC-DJF TAS EMC-GEFS: Area-avg Score for NA: 0.2753

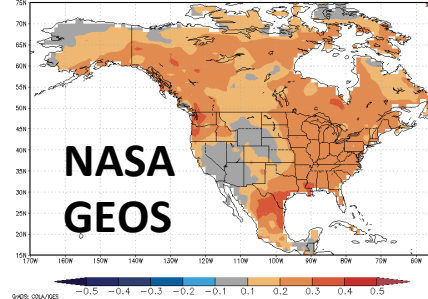


Individual Model and MME Anomaly Correlations

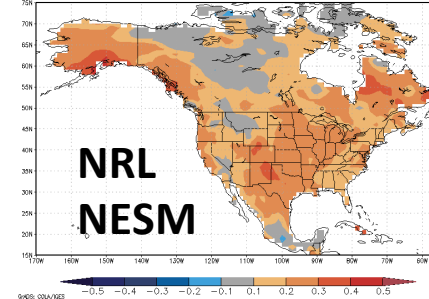
Weighted ACC-DJF TAS ECCC-GEM: Area-avg Score for NA: 0.1380



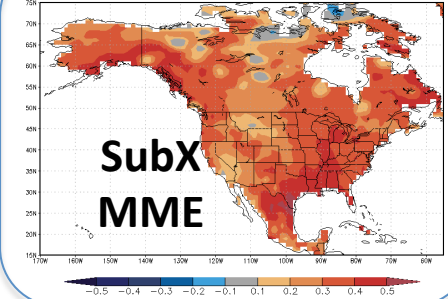
Weighted ACC-DJF TAS NASA-GEOS: Area-avg Score for NA: 0.1916



Weighted ACC-DJF TAS NRL-NESM: Area-avg Score for NA: 0.1931



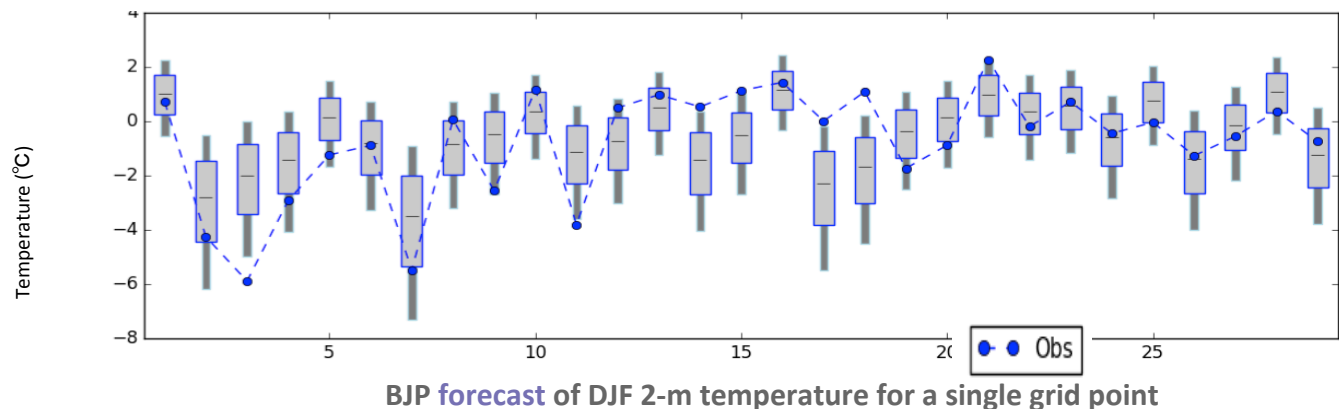
Weighted ACC-DJF TAS 7-MME: Area-avg Score for NA: 0.3252



Calibration of ensembles to obtain reliable probabilities **... while retaining skill**

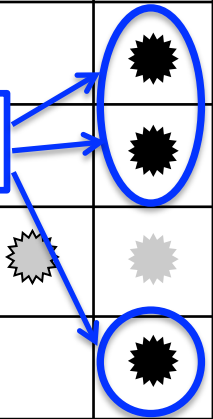
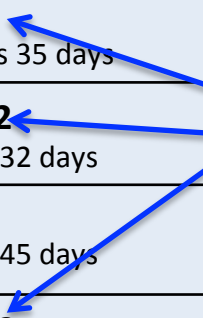
- Do forecast probabilities from an MME represent frequency of occurrence?
- Can an individual model be calibrated and produce the same skill and reliability as an MME?

- Calibration uses **Bayesian Joint Probability (BJP)** modeling (Wang et al. 2009).
 - Predictor (e.g., CFSv2 2-m T) and predictand (e.g., observed 2-m T) modeled using a bivariate normal distribution, where the distribution parameters are not fixed.
 - Individual **calibration** BJP models are developed for each SubX model ensemble mean, grid point, lead, and season.
 - Mini-MME is simple average of 3 ensemble model probabilities.
 - **Ensemble Regression (EReg) baseline** used at CPC (Unger et al. 2009).
- BJP generates a statistical ensemble by sampling from the posterior distribution of the bivariate normal parameters ($n = 1000$).



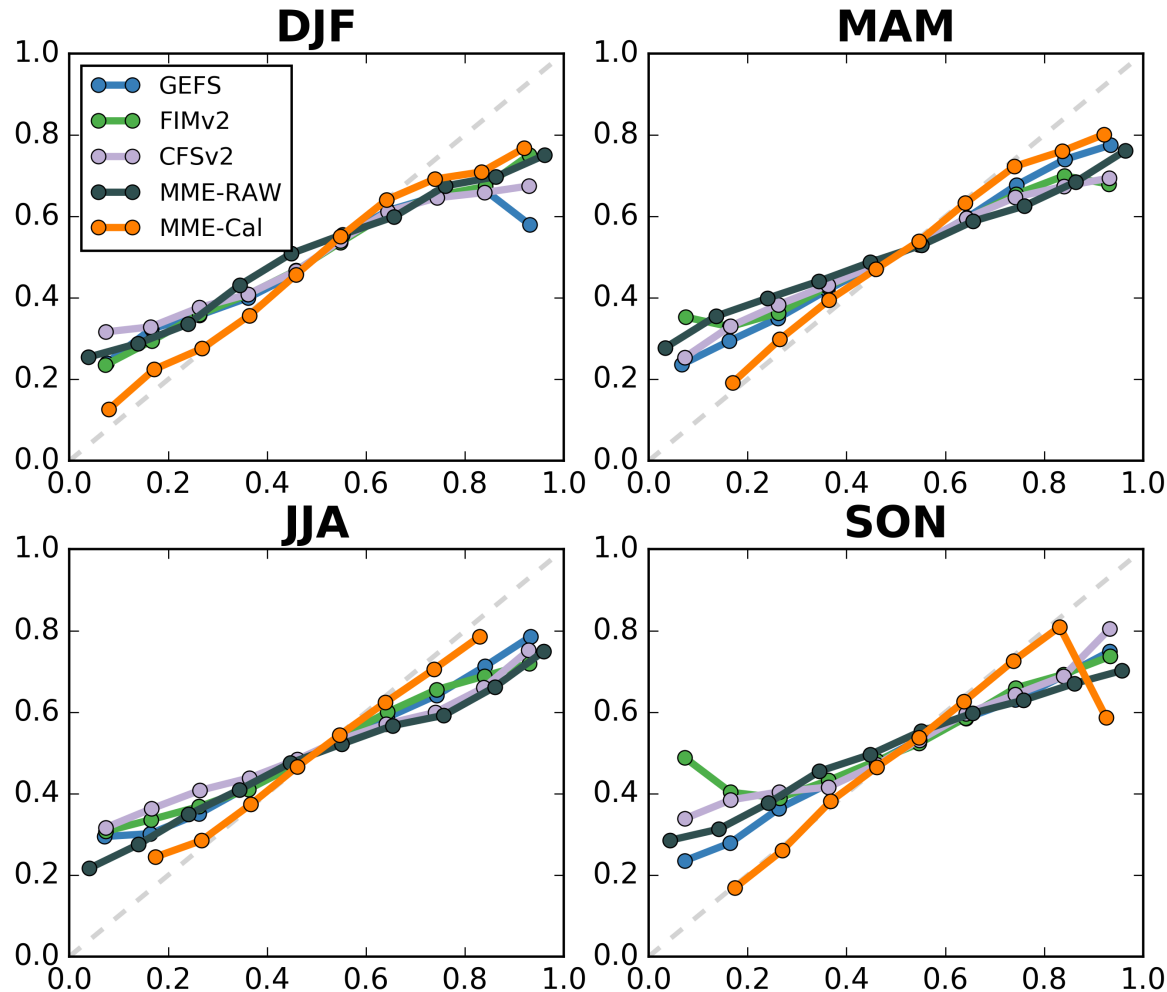
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NASA-GEOS 4 members 45 days	☀	☀	☀	☀	☀	☀		Forecast Day	
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3-Model Mini-MME



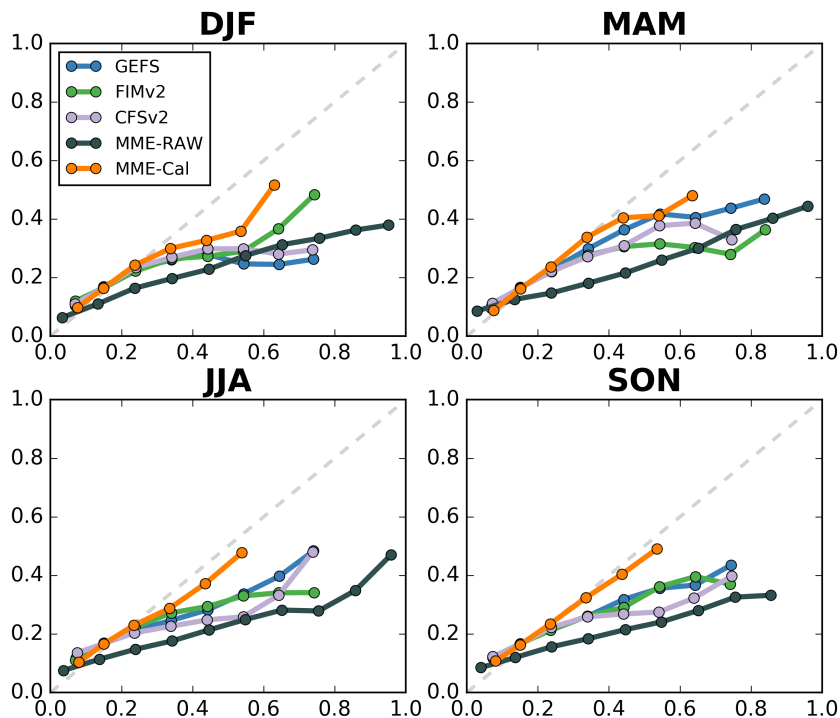
Above / below normal temperature reliability

- **Calibrated MME** more reliable than calibrated **GEFS**, **FIMv2** or **CFSv2**, (small ensemble size), or **MME member count (raw)** probability in all seasons

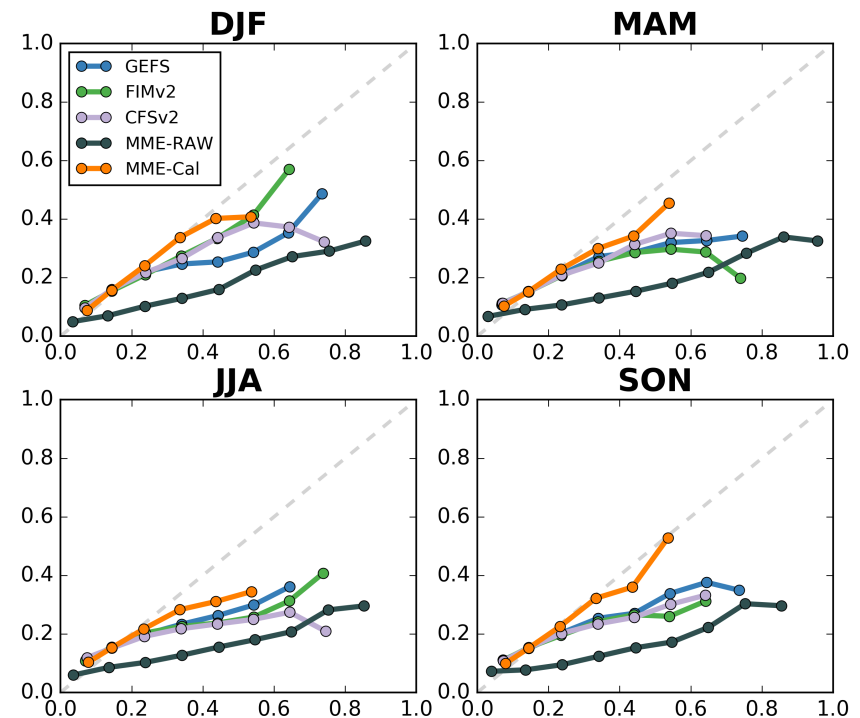


Extreme above/below normal reliability (high and low 15th percentile)

- **Calibrated MME** essential to reliability of probabilities of extremes
- **Raw MME** has much less reliable probabilities
- Individual calibrated **SubXGEFS**, **FIMv2** or **CFSv2** are less reliable than **MME**



Extreme **below** normal



Extreme **above** normal

Are Calibrated Probabilistic MME forecasts more skillful than individual models and un-calibrated ?

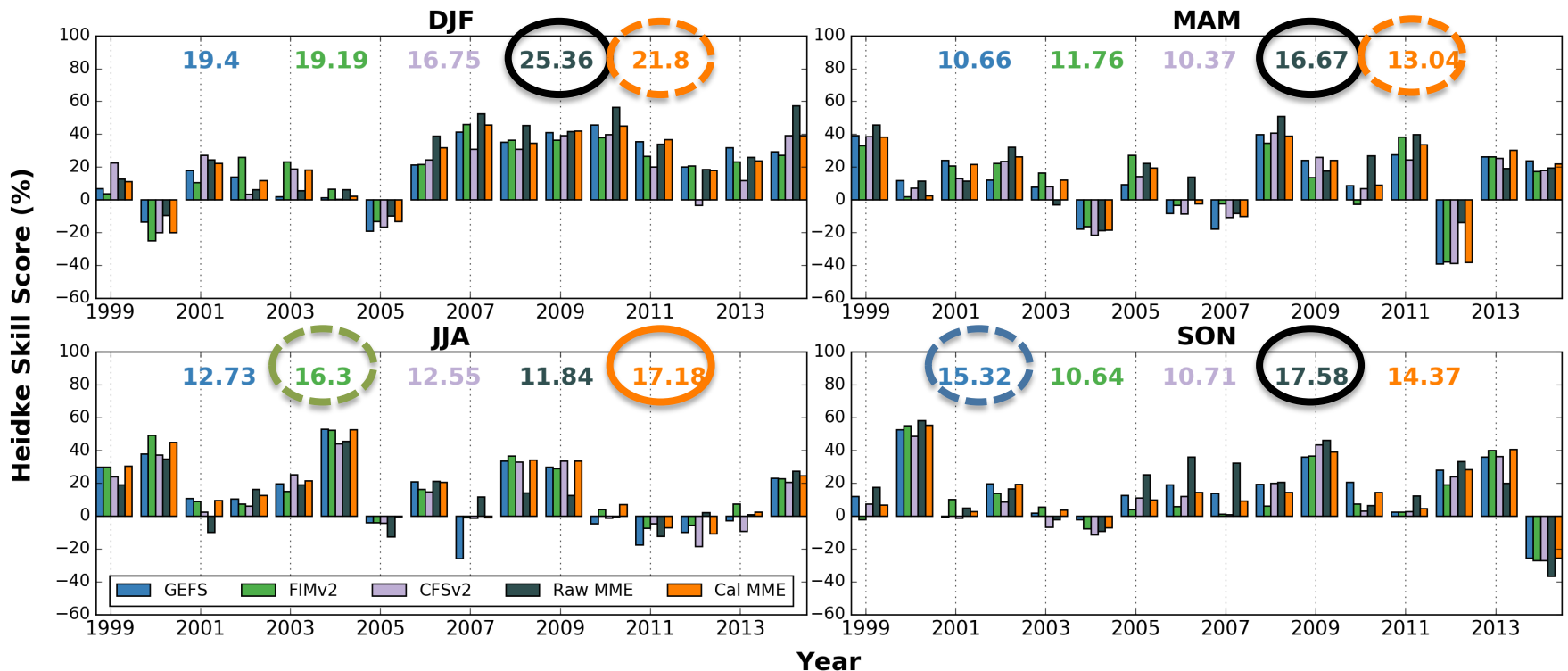
- Heidke Skill Score : Improvement of hit rate relative to random forecast of 15% frequency of occurrence

Above/below median Heidke Skill Score

(% Hit rate improvement)

1st  & 2nd  ranked models

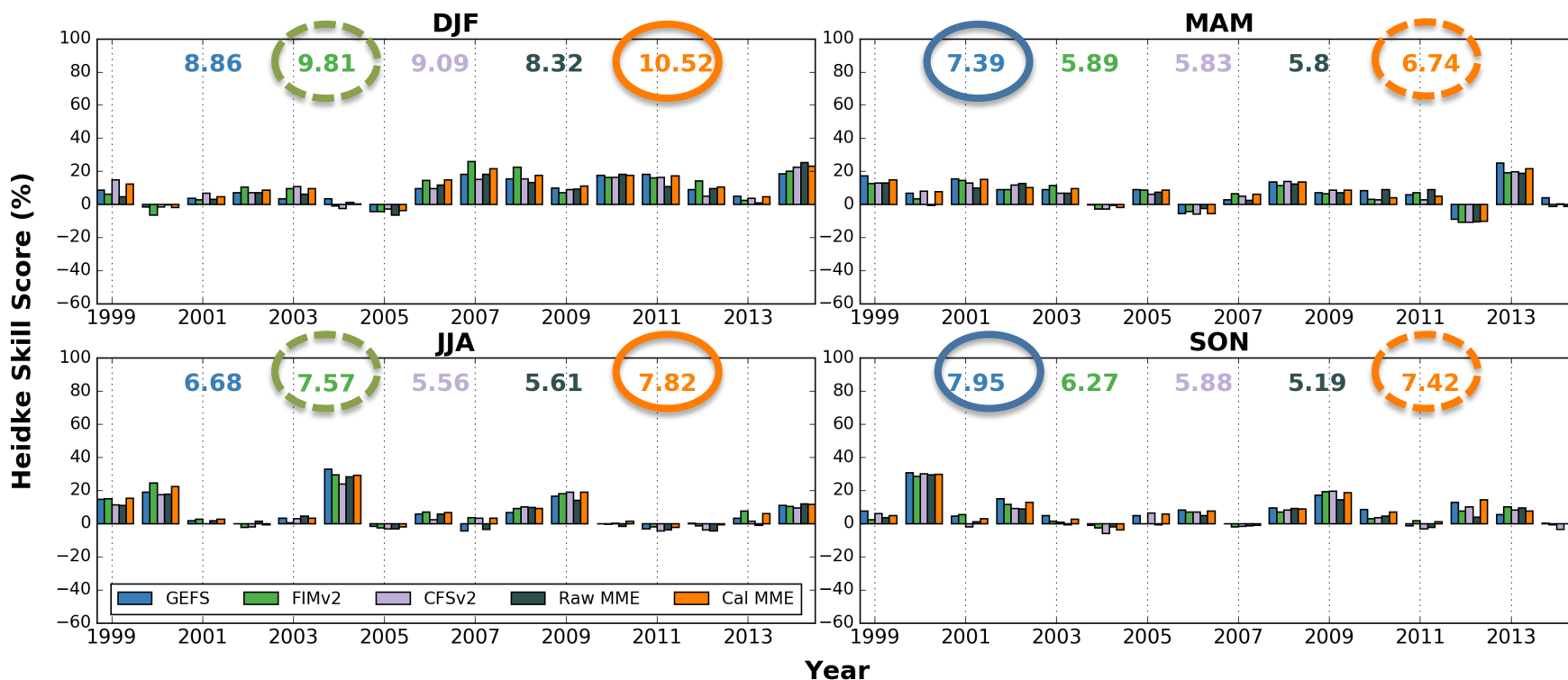
- Raw mini-MME has less reliable probabilities but occasionally better hit rate
- MME more skillful in most months & years than **GEFS**, **FIMv2** or **CFSv2**



Extreme below normal Heidke Skill Score

1st  & 2nd  ranked models

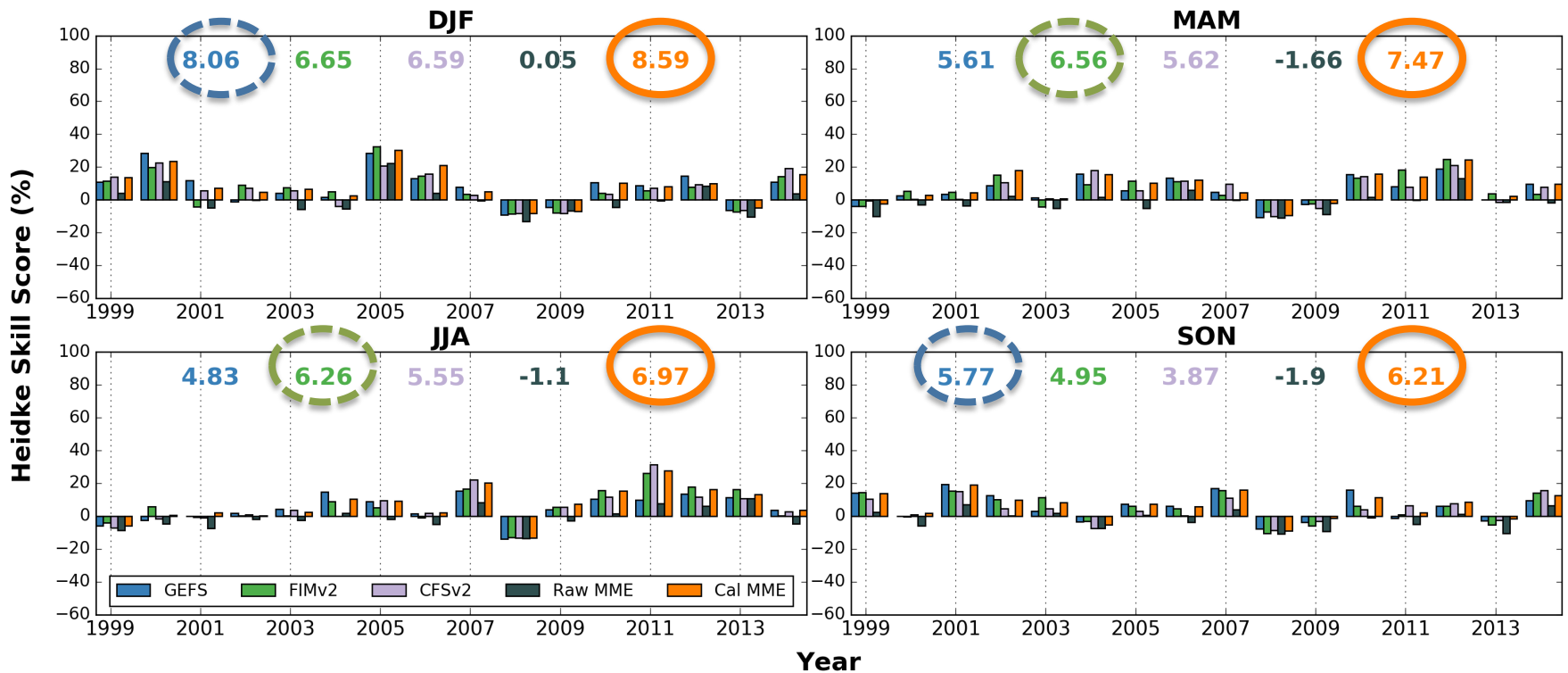
- **Calibration** of raw mini-MME probabilities improves overall Heidke Skill Score
- **Raw mini-MME** has less reliable probabilities AND lower hit rate
- MME more skillful in most months & years than **GEFS**, **FIMv2** or **CFSv2**



Extreme above normal Heidke Skill Score

1st  & 2nd  ranked models

- **Calibration** of raw mini-MME probabilities *improves overall Heidke Skill Score*
- **Raw mini-MME** has less reliable probabilities AND lower hit rate
- MME more skillful in most months / years than **GEFS**, **FIMv2** or **CFSv2**

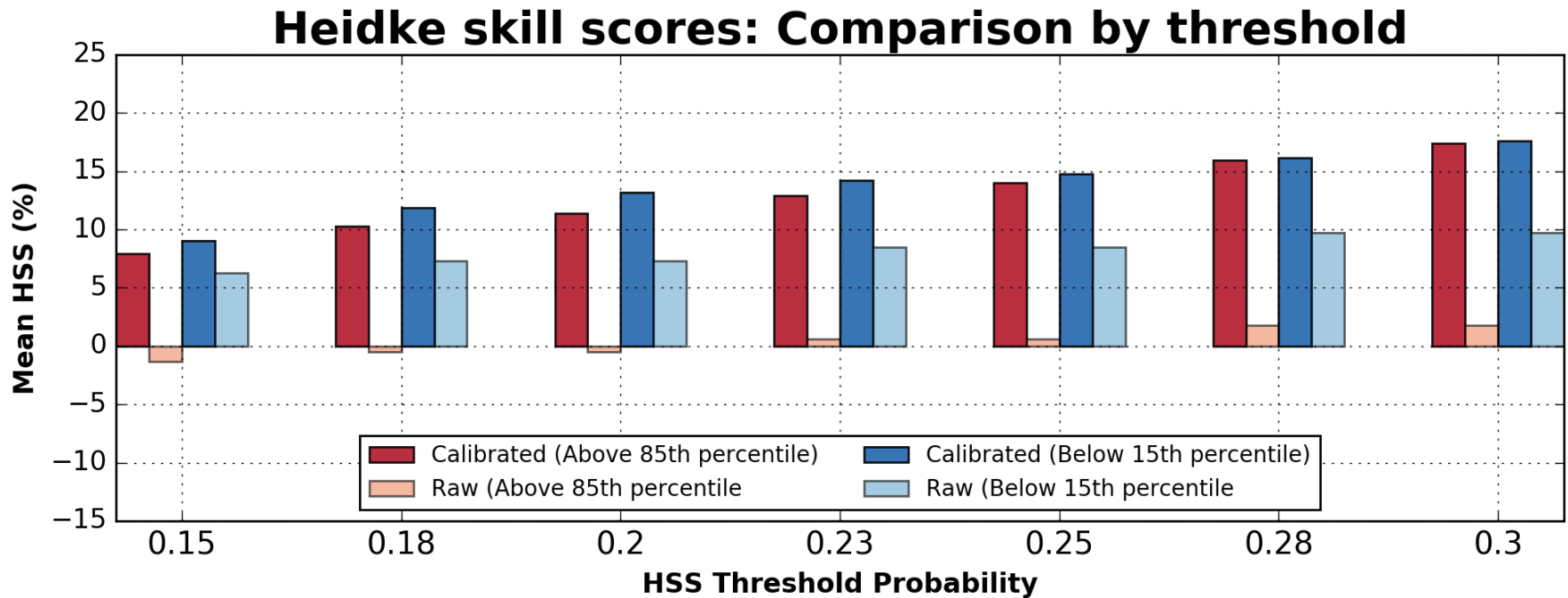


“Forecasts of Opportunity”

Can intermittent skill of forecasts be identified prior to forecast?

Extreme **above** / **below** normal temperature

- **Calibration** improves Heidke Skill Score of **raw** extremes forecast at all probability levels
- **Greater probability implies greater skill**



Summary

- MME of calibrated models produced reliable probabilities
- Calibration improves probabilistic skill (Brier and RPSS)
 - MME improves skill over individual models
- Higher probabilities represent periods of greater skill for extremes, or forecasts of opportunity
- Future work:
 - Optimize MME combination weighting
 - Identify conditional skill and forecasts of opportunity using possibly weather regimes or climate modes of variability