# Advancing intraseasonal climate forecasts through use of reforecasts

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Thanks to David Unger and Michelle L'Heureux of CPC and Tom Hamill of ESRL

# <u>Outline</u>

- 1. Current state of intraseasonal prediction at CPC
- 2. Motivation & New model datasets retrospective forecasts
- 3. Methods & Theory
- 4. Preliminary analysis
- 5. Skill improvement
- 6. Summary & Conclusions
- 7. Next Steps

# CPC intraseasonal forecasts and models

Extended Range Forecasts:

- Subjectively combine dynamical model output from NCEP, ECMWF and CMC
- Legacy tools that specify surface variables from upper air circulation
- With improvement in model skill, increasingly depend on model derived surface variables and statistical correction of those variables to observations or reanalysis
- 6-10 day and week-2 probabilistic precipitation and temperature forecasts
- Examining forecast skill at weeks 2, 3 and 4 in NCEP Climate Forecast System version 2

# North American Ensemble Forecast System (NAEFS)

NAEFS Models:

- NCEP Global Ensemble Forecast System (GEFS)
  - 20 members X 4 cycles per day
- Canadian Modeling Center ensemble prediction system (EPS)
  - 20 members X 2 cycles per day
- NAEFS Forecast System:
- Univariate bias correction of variables to model analysis
- Relatively short training data for adaptive system
- Bias adjustment can be spun up for model upgrades in relatively short time
- Bias =  $\alpha$  (F<sub>-1</sub> O<sub>-1</sub>) + (1  $\alpha$ )\* Bias<sub>-1</sub>

#### Winter (DJFM 2010-11) 8-14 day Temperature Forecast



8-14 Day Temperature Ranked Probability Skill Scores for all categories from 12/01/2010 - 03/31/2011



#### Summer (MJJA 2011) 8-14 day Temperature Forecast



#### Summer (MJJA 2011) 6-10 day **Temperature Forecast**



#### Week 2 Temperature using 2 Models or 1 model Probability of Above, Near (gray) and Below Normal





#### 00Z NCEP + Environment Canada

06Z NCEP Only (Note greater probabilities)

Motivation for using new retrospective forecasts

- First generation CDC/ESRL forecast tool still in use at CPC for extended range, running model more than 10-years old
- Shorter training data unable to fully capture model statistics and calibrate forecasts
  - Forecast reliability remains over-confident with under-dispersive ensembles and residual bias
- Bias estimate is always lagged from real-time forecasts, which can lead to seasonal cycle errors
- Limited independent cases for forecastobservation comparison of intraseasonal timescale variability

# Potential use of new retrospective forecasts

Next generation NOAA GEFS reforecasts created by ESRL:

- Current operational NCEP Global Ensemble Forecast System (GEFS) as of 2012
- T254L42 (about ½ degree grid spacing) in week 1 and T190L42 (about ¾-degree) in week 2
- 10 perturbation ensemble members + control
- 1985-2010 (2011 to be added)
- Using Climate Forecast System Reanalysis initial conditions (now GDAS)
- Currently nine variables are archived at CPC
- Complete data available by ftp from ESRL Comparison to CFSv2:
- T126L64
- 4 members x 4 cycles per day to 45-days lead in real time
- 4 members per day reforecasts
- 1999-2011

# Methods

- 1. Ensemble Regression
- Currently used in CPC seasonal forecasts
- Regression statistics calculated from the ensemble mean apply to individual members after accounting for the ensemble spread
- Calculate statistics for 1985-2004 reforecasts: Y<sub>mean</sub>, Y<sub>variance</sub>, E<sub>spread</sub>, R<sub>XY</sub>, X<sub>mean</sub>, X<sub>variance</sub>
- $\sigma^2_{\text{Error variance}} = \sigma^2_{\text{Observed}} (1 R^2)$
- R best member =  $R^2_{mean} / R_{individual members}$
- Initial model fields are used as observation
  - Reduces issues in matching climatologies of forecasts and observations

# Methods

- 1. Regression of ensemble mean
- 2. PDF correction using reforecast climatological mean and variance
- 3. Count-based ensemble derived PDF with kernel smoothing based on sample size

4. Climatology

# Example: Probability density functions based on linear regression



- Ensemble member forecasts
- Ensemble regression PDF
- Regression of ensemble mean

### Mean bias

#### Winter







### **Mean Correlation**

### Best Member Correlation



### Ratio best member kernel width to mean error





-2 -1 -0.5 -0.2 0.2 0.5 1 2

-2 -1 -0.5 -0.2 0.2 0.5 1



2 23 2

-2 -3 -2 -2 -015 -1 -0.2 -0.5 -0.1 -0.1 0.1 0.1 0.2 0.5 015 1 2 23 2

 $\mathsf{CRPS}_{\mathsf{ens}\,\mathsf{reg}}-\mathsf{CRPS}_{\mathsf{count}}$ 

-2 -3 -2 -2 -015 - 1 - 0.2 - 0.5 - 0.1 - 0.1 0.1 0.2 0.5 015



# Reliability of percentiles of distribution



- Ensemble regression
- Regression of ensemble mean
- PDF correction of mean and variance
- Count based PDF with kernel sampling correction

### CRPS time series for DJF 2005



- Climatological PDF forecast
- Ensemble regression
- Regression of ensemble mean
- PDF correction of mean and variance
- Count based PDF with kernel sampling correction

# Conclusions

- Correction of week 2 temperature using the NOAA GEFS reforecasts shows large gains in skill over uncorrected model ensemble PDF
- Analysis of full PDF CRPS and reliability does not indicate any advantage of using individual members in ensemble regression over regression of the ensemble mean
  - Perhaps extremes and other variables may be different
- Regression has some advantage over a simple correction of model mean and variance biases

### Next Steps

- Examine robustness of results to changes in the reforecast system: Fewer years or ensemble members
- Examine other variables and extremes
- Extensive data set will allow analysis of skill in relationship to climate modes, such as MJO