Design and Framework of Seasonal Prediction Systems

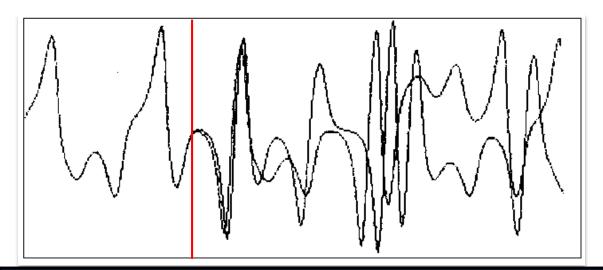
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Outline

- What is seasonal prediction and what makes seasonal prediction possible?
- Methods for making seasonal prediction
- An example of seasonal prediction system: NCEP
 Climate Forecast System version 2 (CFSv2)
- Summary

There is always a spread (uncertainty) in forecasts!

- Non-linear dynamical systems sensitivity to specification of initial conditions
- Deterministic chaos
- Uncertainty could be better quantified, but can never be removed



•
$$dx/dt = \sigma (y - x)$$

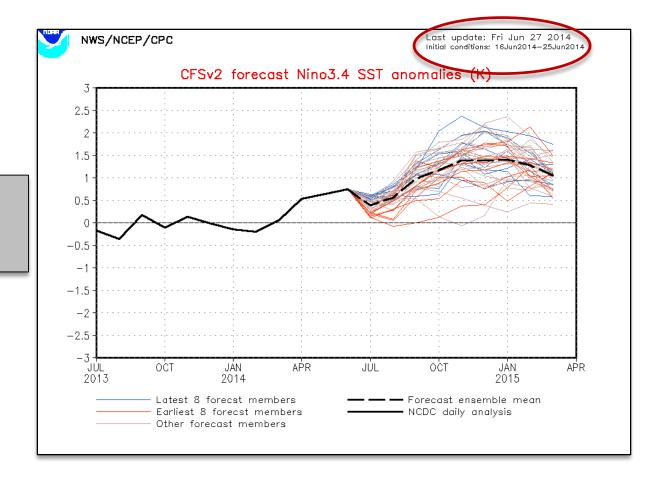
•
$$dy/dt = x(\rho - z) - y$$

•
$$dz/dt = xy - \beta z$$

- There is always a spread (uncertainty) in forecasts...
- This forecast uncertainty is quantified using ensemble prediction approach where a collection of forecasts is initiated from small perturbations in the initial conditions
- Evolution of individual forecasts in the ensemble results in a collection of future outcomes which can be quantified using a probability density function (PDF)

Example of forecast spread: ENSO Prediction

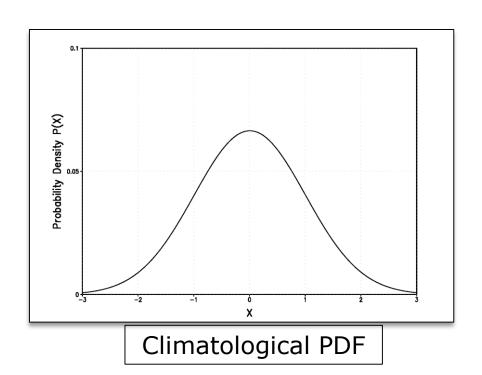
NCEP/CFS Nino 3.4 SST Prediction

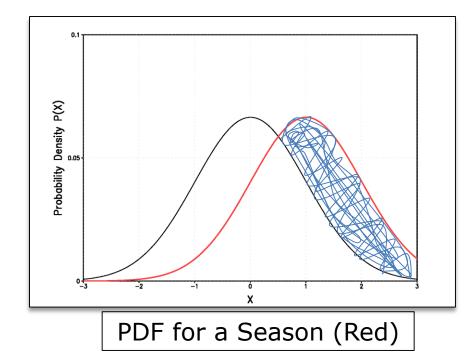


Characterizing seasonal prediction

- There is a PDF of possible outcomes of mean states for a specific season
- There is a climatological PDF based on aggregation of seasonal mean outcomes for all seasons.
- These PDF depends on
 - Season
 - Variable
 - Location
- Seasonal prediction depends our ability to differentiate PDF of forecast outcomes (for an individual season) from the climatological PDF

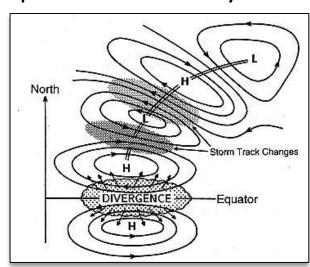
Characterizing seasonal prediction



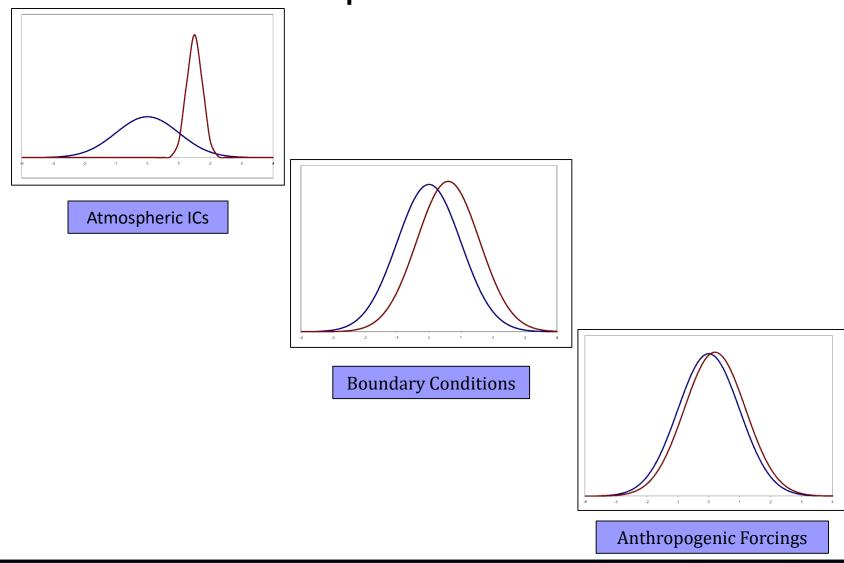


What lends predictability in long-range predictions?

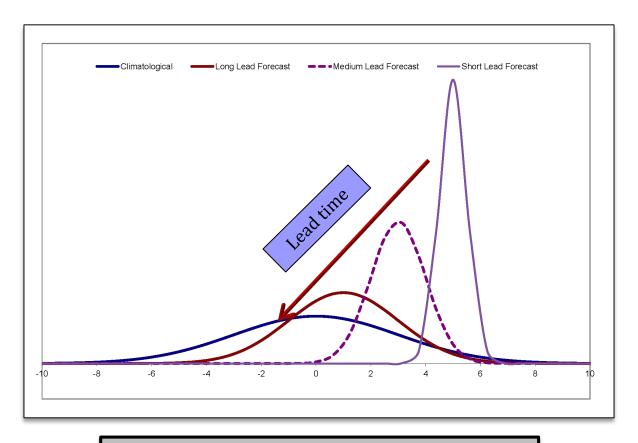
- Initial conditions
 - Weather prediction
 - ENSO prediction
- Influence of boundary conditions
 - Anomalous SSTs → Influence on atmospheric variability
- Influence of external forcings
 - Changes in CO₂



Sources of predictability and initial value predictions

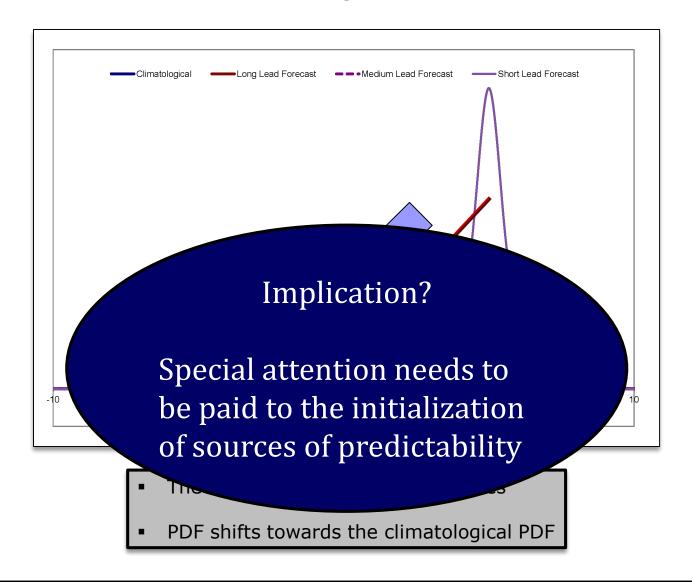


Initial value predictions with different lead time



- The spread with lead time increases
- PDF shifts towards the climatological PDF

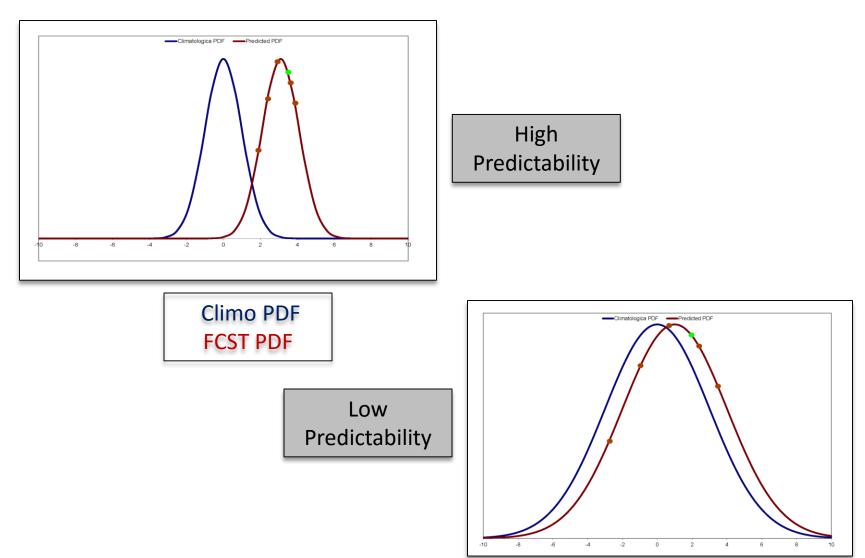
Initialized predictions



What provides <u>skill</u> in seasonal predictions

- It is our ability to distinguish PDF of outcomes for the event to be predicted from the corresponding climatological PDF
- Differences in the PDF can come from differences in various moments of the PDF
 - Mean
 - Spread
 - Skewness

Examples of high/low prediction skill



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Seasonal Prediction Methods

- Empirical prediction tools
 - Advantages
 - Trained based on historical observations
 - Unbiased
 - Simple and computationally efficient
 - Disadvantages
 - Limited by observational data
 - Mostly depend on linear relationships
 - Non-stationarity in climate is hard to include
 - Cannot handle unprecedented situations

Seasonal Prediction Methods

- Dynamical Prediction Tools
 - Advantages
 - Linearity and non-stationarity is not an issue
 - Easier to construct PDF of seasonal mean state
 - Easier to handle unprecedented situations
 - Disadvantages
 - Computationally expensive and require a large infrastructure
 - Forecast systems have biases that requires special attention
- Properties of empirical and dynamical prediction tools are complementary in nature, and in general, and generally both are used in the development of final forecast
- This is the current practice used by several operational centers, e.g.,
 prediction of monsoon rainfall by the IMD

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Components of a Seasonal Forecast System

- Forecast system components
 - Initialization
 - Hindcasts
 - Real-time forecasts
 - Skill assessment
 - Bias correction and calibration
 - Forecast dissemination

Initialization

- Various components of the forecast system need to be initialized from their observed state
 - Atmosphere (temperature; humidity; winds)
 - Ocean (temperature; salinity; ocean currents)
 - Land (soil moisture; snow)
 - Sea ice (extent; thickness)
- Initialization is done from the Climate Forecast System Reanalysis (CFSR) that provides a consistent 3-dimensional analysis of various components of the Earth System
- After initialization, forecast system is run to nine months into the future

Hindcasts

- Hindcasts Run the real-time forecast system over historical cases
- Run the forecast system over last thirty years (1981-2010)
- Four nine months forecast every 5th day of the calendar month
- 72 forecasts every year

Hindcasts

- What is the purpose of hindcasts?
 - Provides an assessment of the skill of the seasonal forecast system
 - Because of model biases
 - Real-time forecasts have to be bias corrected
 - Hindcasts provide the data set for bias correction
 - Hindcasts are used to develop initial month, and lead-time dependent model climatology
 - Calibration of real-time forecasts

Real-time forecasts: CFSv2

- Four nine month forecasts every day
- 120 seasonal forecasts in a month
- Real-time forecasts are constructed based on forecasts from latest 10 days of initial conditions, i.e., an ensemble of 40 forecasts is used for developing real-time seasonal predictions
- Lagged ensemble provides an estimate of PDF of seasonal mean states

Real-time forecasts

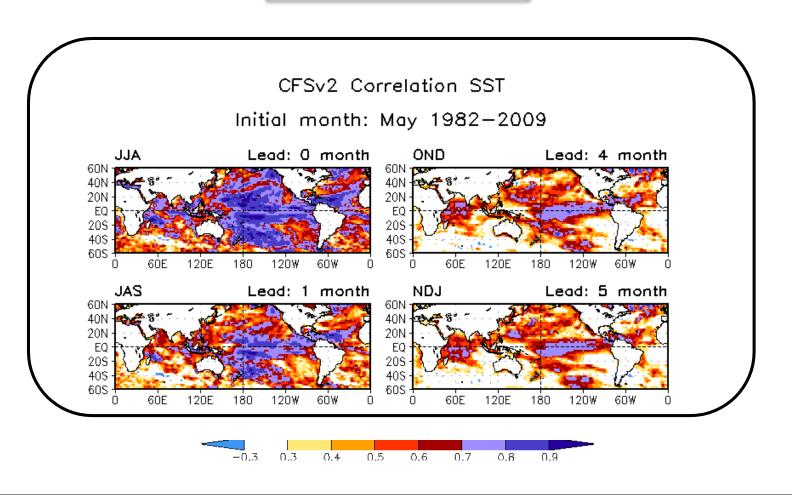
- Configuration of real-time forecasts generally differs from their hindcast counterpart
 - More frequent
 - Larger ensembles
- Consistency in the analysis of initial conditions, particularly for slowly varying components of the Earth System (SST, soil moisture) is crucial!

Skill Assessments

- Based on 30-year hindcast, skill of the CFSv2 can be assessed for
 - Predicting sea surface temperature anomalies
 - Predicting various SST indices that are important for seasonal predictions, e.g., Nino 3.4 SST index
 - Surface quantities over land, e.g., precipitation and surface temperatures
 - Other variables
 - Soil moisture
 - Sea ice

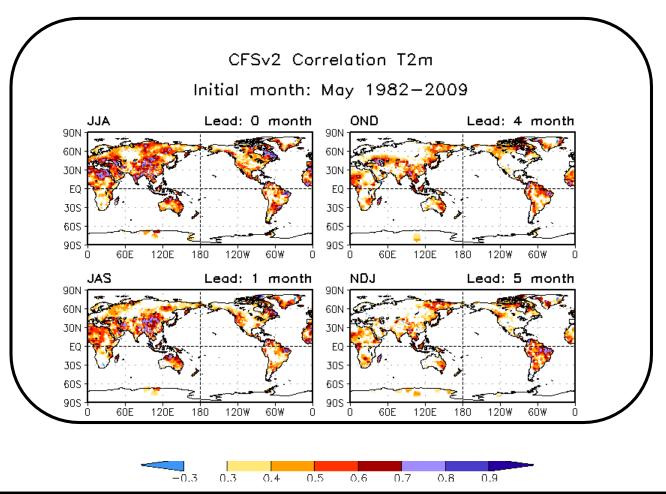
Skill Assessment: SST

Anomaly Correlation



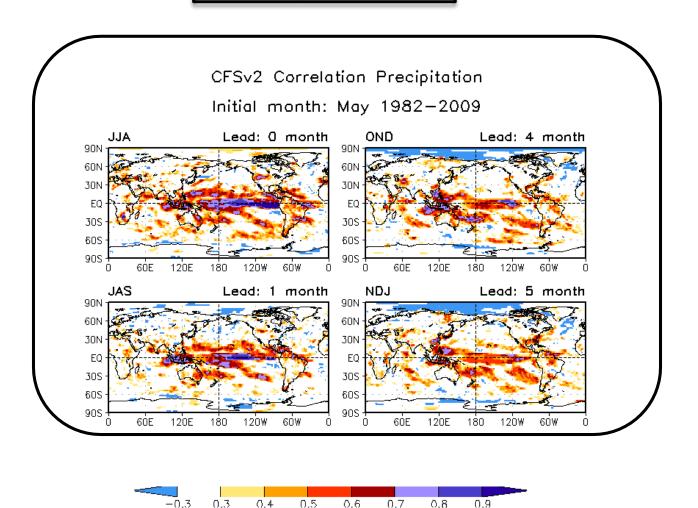
Skill Assessment: Surface Temperature

Anomaly Correlation



Skill Assessment: Precipitation

Anomaly Correlation



Bias Correction and Calibration

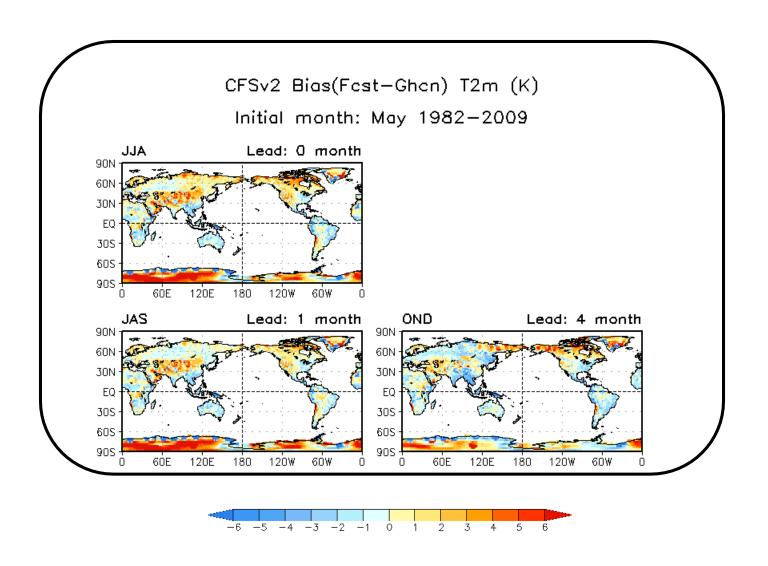
Bias correction

- Correct for differences in observed and predicted mean state
- Adjust if variability between observations and predictions differs

Calibration

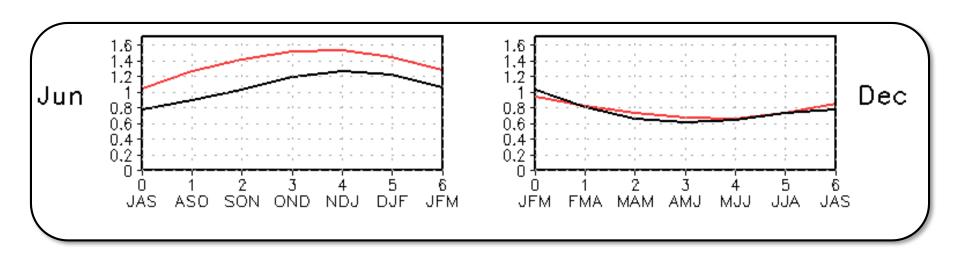
- Adjust predicted anomaly based on assessment of past skill (e.g., from hindcast data set)
- If past skill is close to zero, make the forecast PDF same as the climatological PDF

Model bias



Differences in Variability

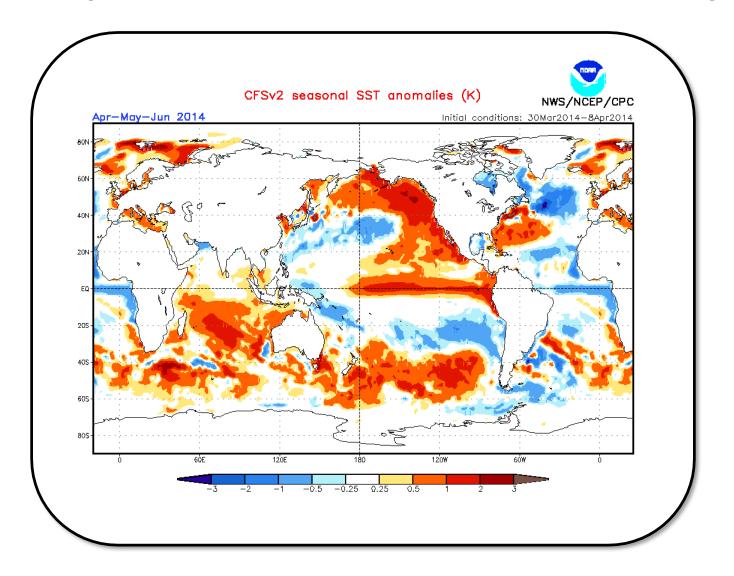
Standard Deviation Nino34 SST(K)



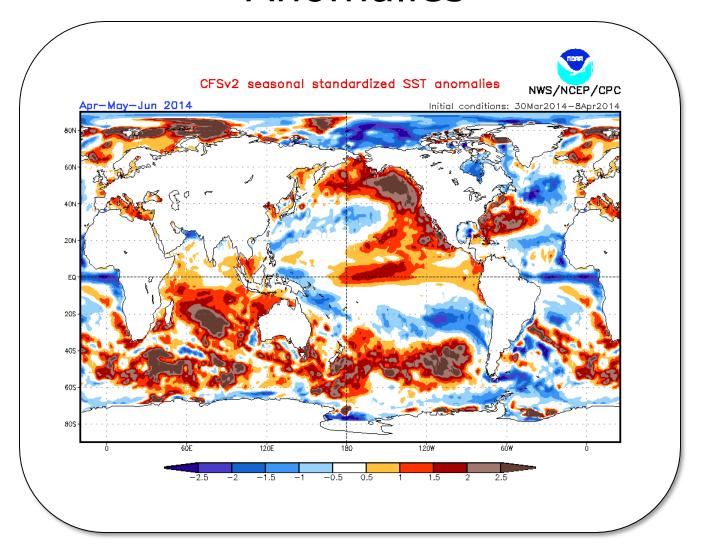
Forecast Dissemination

- Graphical products
 - Bias corrected seasonal mean anomalies
 - Normalized anomalies
 - Bias corrected anomalies with skill mask
- Forecast and hindcast gridded data
 - Real-time forecasts
 - Hindcast data available via several channels
 - Procedures could be developed for statistical downscaling

Graphical Products: SST Anomaly



Graphical Products: Standardized SST Anomalies



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- Seasonal prediction system are fairly mature
- Skill of prediction is limited, but it is better than a random guess
- Hindcast and real-time forecast data is a huge data base that can be used for various research and analyses purposes
 - Analysis and predictability of extremes
 - Influence of various climatic factors on extremes (attribution and prediction)