

Seasonal forecasting with the NMME with a focus on Africa

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Topics covered

- Fundamentals of seasonal forecasting
 - Deterministic vs probabilistic ensemble forecasts
 - Forecast skill
 - How seasonal forecasts are produced
 - El Niño impacts
 - ENSO prediction
- Multi-model ensembles (MMEs)
- North American Multi-Model Ensemble (NMME)

Fundamentals of seasonal forecasting

Necessary conditions for useful climate predictions

1) The phenomenon being forecast must be *predictable*

2) Prediction method must have ability to capitalize on natural predictability

→ If these two conditions are met then there is potential for skillful predictions

Daily weather is not very predictable after 7-10 days

Example: Daily weather predictions for Paris in February 2017, retrieved on 20 November 2016!

France Weather Paris, F		France Local Weather 🔇 7°C 🝷 Per		Perso	onalized Forecasts: Migraine 🔻	
Now	Weekend	Extended	Month R	Radar Min	uteCast®	
(!) Yellow Warning for Wind Although rather usual in this region,						
G January 2017		View:	February - 2017 -		March 2017 🔿	
SUN 1/29	MON 1/30	TUE 1/31	WED 2/1	THU 2/2	FRI 2/3	SAT 2/4
- Č	-)	- 🏹	- 🏹	-ờợ-	-`Ċ	-\\
6°/-3° Mostly sunny	4° _{/-3°} Mostly sunny	3°/-3° Mostly sunny	4° _{/-3°} Mostly sunny	5°/-3° Sunshine	5°/-1° Sunshine	7°/-2° Sunshine
Hist. Avg. 6°/1°	Hist. Avg. 6°/1°	Hist. Avg. 6°/1°	Hist. Avg. 6°/1°	Hist. Avg. 6°/1°	Hist. Avg. 6°/1°	Hist. Avg. 6°/1°
SUN 2/5	MON 2/6	TUE 2/7	WED 2/8	THU 2/9	FRI 2/10	SAT 2/11
-)	-)	- 🏹		-)	-)	
7°/-2° Sun and some clouds	5°/3° Sunshine and some clouds	8° /1° Mostly sunny	9°/-1° Periods of rain	8° /4° Partly sunny	9°/4° Partly sunny	12°/1° Mostly sunny
Hist. Avg.	Hist. Avg.	Hist. Avg.	Hist. Avg.	Hist. Avg.	Hist. Avg.	Hist. Avg.

Daily weather is not very predictable after 7-10 days

Example: Daily weather predictions for Paris in February 2017, retrieved on 20 November 2016!



However, longer term averages over a week, month or season may be predictable depending on location, lead time, etc.

Predictability and Prediction



Probabilities based on an ensemble of forecasts



- When uncertainties are large, a single deterministic forecast tells us very little → need an *ensemble* of forecasts to estimate the probabilities of different outcomes
- Ensemble average provides a deterministic forecast for the average outcome
- Better are probabilistic forecasts describing the likelihood of different outcomes

Deterministic vs probabilistic ensemble forecasts

Ensemble deterministic forecasts

Example: Seasonal mean temperature for JFM 2016

Deterministic forecast (single location)

"The average temperature in Victoria, Canada during JFM 2016 will be 0.85°C above normal relative to the average of all years in 1981-2010."

Deterministic forecast map





Uncalibrated ensemble mean anomaly forecast.

However, these products contain no indication of uncertainty

Probabilistic forecast (single location)



Here the forecast probability distribution or PDF is described in terms of probabilities that forecast seasonal mean temperature will fall into climatologically equi-probable tercile categories: **below normal near normal above normal**

Probabilistic forecast (single location)



Here the forecast probability distribution or PDF is described in terms of probabilities that forecast seasonal mean temperature will fall into climatologically equi-probable tercile categories: below normal near normal above normal

Probabilistic forecast (single location)



Here the forecast probability distribution or PDF is described in terms of probabilities that forecast seasonal mean temperature will fall into climatologically equi-probable tercile categories: below normal near normal above normal

Probabilistic forecast maps

Probabilities in each category



Reliability of probabilistic forecasts

- Consider many probabilistic forecasts from different times, locations
- Compare forecast probabilites with observed frequencies



Reliability of probabilistic forecasts

- Consider many probabilistic forecasts from different times, locations
- Compare forecast probabilites with observed frequencies



Advantages of calibrated probability forecasts

- uncalibrated probabilities:
 - high probabilities predicted far more frequently than observed
 - overconfident, especially for precipitation and nearnormal category
 - near-normal grossly overpredicted
- calibrated probabilities:
 - much more reliable (forecast probability ≈ observed frequency)
 - less overconfident
 - near-normal less overpredicted







Seasonal precipitation forecast

Growth of uncertainty with increasing lead



Lead 0 months

Lead 3 months



Lead 6 months



Lead 9 months



Growth of uncertainty with increasing lead

Lead 3 months



Lead 0 months

Flexible probabilistic forecasts from IRI

- Useful if tercile below/near/above normal probabilities are not specific enough
- Example: probability that JFM 2016 mean temperature will exceed 80th percentile relative to 1981-2010 (Options are 10, 15,...85, 90 percentiles)



http://iridl.ldeo.columbia.edu/maproom/Global/Forecasts/Flexible_Forecasts/temperature.html

Forecast skill

Some terminology

0'

Forecast lead time



0'

0'

Global anomaly correlation skills

(from Canadian Seasonal to Interannual Prediction System)

DJF (Lead 0 months)

ths) JJA (Lead 0 months) Near-surface temperature





Precipitation



General behavior

- Higher in tropics than extratropics
- Higher over oceans than land



- Higher in winter than summer
- *Much* lower for precipitation then temp

Skill dependence on lead time and averaging period

(from Canadian Seasonal to Interannual Prediction System)

Example: Anomaly correlation for near-surface temperature from Dec



Skill dependence on lead time and averaging period

(from Canadian Seasonal to Interannual Prediction System)

Example: Anomaly correlation for near-surface temperature from Dec





Guiding principles of climate (e.g. seasonal) forecasting

1) Forecasts should communicate uncertainty





Probabilities



ensemble forecasts

- 2) Forecasts should be interpreted in the context of past performance (skill)
 - need many years of hindcasts to calculate skill



Purposes of hindcasts

Hindcasts enable us to...

- Estimate lead-time dependent model biases ("drift") so that they can be corrected for – more in lab session
- Estimate historical skill
- Calibrate probabilistic forecasts

Notes:

- When estimating in-sample corrections and skill, cross validation should be applied to avoid inflated estimates of skill (won't worry about it in the lab, unless you want to)
- WMO currently recommends 1981-2010 as hindcast base period
- 30 years × 12 initialization months × 10 ensemble members = 3600 years of model integration per hindcast ! (assuming 12 mon range)

How seasonal forecasts are produced



How dynamical seasonal forecasts are made

Weather forecast

1-10 days



- Atmosphere/land models
- Observations of current global conditions used to initialize model



Climate projection 10-100 years

Global average surface temperature change



- Atmosphere/ocean/ landsea ice models
- Initial conditions not crucial

How dynamical seasonal forecasts are made



1 tier (coupled) vs 2 tier forecasts

1 tier forecast



- atmosphere interacts with land
- SSTs *specified* (no ocean model)
- For example, some systems simply persist the SST anomaly present before the forecast
- 1 tier systems cannot forecast El Niño/La Niña



2 tier forecast

- atmosphere interacts with land and ocean
- coupled climate model includes ocean component
- future SSTs are forecast by model
- 2 tier systems potentially can predict El Niño/La Niña (and often do)

Problem with 1 tier forecasts

 <u>C</u>onsider 2-tier forecast (persisted SSTA) from 1 April 2006





Oct 2006 (lead 6)

Problem with 1 tier forecasts

- <u>C</u>onsider 2-tier forecast (persisted SSTA) from 1 April 2006
- The 1 tier forecast persists the La Niña present before the start of the forecast



Oct 2006 (lead 6)

Problem with 1 tier forecasts

Consider 2-tier SST "forecast" **Observed SST anomaly** ٠ (persisted SST anomaly) forecast (persisted SSTA) from 1 April Mar 2006 2006 The 1 tier forecast ٠ Apr 2006 (lead 0)persists the La Niña present before the start of the forecast May 2006 (lead 1) But the La Niña ٠ soon disappears! Jun 2006 (lead 2) Jul 2006 (lead 3)Oct 2006

(lead 6)
Steps for producing seasonal forecasts

- Run ensemble of forecasts from slightly different initial conditions
- Subtract the *hindcast climatology* to obtain anomalies, from the climatological mean and correct for model biases and drift
- Deterministic forecast = ensemble mean anomaly
- To construct probabilistic forecast, must
 - count ensemble members in each tercile of observed climatology, or
 - fit ensemble values to a distribution (better), or
 - calibrate the forecast distribution to produce a more reliable forecast (best)

El Niño impacts

Equatorial Pacific climate



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Equatorial Pacific climate



Equatorial Pacific climate



El Niño direct impacts





El Niño teleconnections



El Niño conditions in the tropical Pacific

El Niño teleconnections



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Historical El Niño/La Niña variability

- A widely used indicator of El Niño/La Niña activity is Nino3.4 = mean SST anomaly in 5N-5S, 120W-170W
- The Oceanic Nino Index (ONI) consists of a 3-month rolling average of Nino3.4





Historical El Niño/La Niña variability



Global El Niño impacts

Global El Niño impacts

Global La Niña impacts

ENSO impacts on Africa precipitation

ENSO impacts on Africa precipitation

ENSO Prediction

Nino3.4 ensemble plumes from Nov 2016

Historical CanSIPS ENSO predictions

Seasonal mean Nino3.4 index: observed vs 0-9 month lead times

Multi-model ensembles (MMEs)

Why multi-model ensembles?

- 1) Different models have different strengths and weaknesses
 - model errors will tend to cancel each other out
 - higher skill for multi models than for single model, for a given ensemble size N
 - this example considers 4 models with 10 ensemble members each
- 2) More ensemble members available by combining models than from individual models

TAS, Land, zero-month lead, Corr

Kharin et al., Atm.-Ocn. (2009)

WMO multi-model ensemble

https://www.wmolc.org/

- 12 Global Producing Centres (GPCs) representing different meteorological services
- Forecast information provided to Regional Climate Centres (RCCs) and Climate Outlook Forums (COFs)
- Maps and data password protected

APCC multi-model ensemble

http://www.apcc21.net

Climate Outlook for November 2016 - April 2017

BUSAN, 25 October 2016 – Synthesis of the latest model forecasts for November 2016 to April 2017 (NDJFMA) at the APEC Climate Center (APCC), located in Busan, Korea, indicates the ENSO state to be neutral. The positive temperature anomalies are most likely to prevail over the globe. Highly probable above normal rainfalls over the maritime continent are predicted for NDJ.

- Models include CMCC, CanCM3, CanCM4, NASA, NCEP, PMU, POAMA
- Month 1-3 and 4-6 probabilistic & deterministic forecast maps publicly available
- Data password protected

EUROSIP multi-model ensemble

- Four models at ECMWF:
 - ECMWF System 4
 - Met Office HADGEM model, Met Office ocean analyses
 - Météo-France Météo-France model, Mercator ocean analyses
 - NCEP CFSv2
- Common operational schedule (products released at 12Z on 15th)
- Some charts (ENSO plumes, tropics) publicly available at <u>http://www.ecmwf.int/en/forecasts/charts/seasonal/</u>

North American Multi-Model Ensemble (NMME)

What is the NMME?

- Ensemble of opportunity: US, Canadian operational forecast systems + US research systems
- Hindcasts and real time forecasts
- Real time forecasting since Aug 2011
- All data openly accessible
- Requirements for inclusion
 - ensemble system, range \ge 9 months
 - must provide hindcast data for 1982-2010
 - commitment to provide real time forecasts by 8th of each month (CPC operational schedule)

NMME

Currently contributing models

Model	Center	Ensemble size
CFSv2	NCEP	24 (28)
CanCM3	EC/CMC	10
CanCM4	EC/CMC	10
FLOR	GFDL	24
CM2.1	GFDL	10
CCSM4	NCAR	10
GEOS-5	NASA	11
CESM1	NCAR	10
Total ensemble size		109 (113)

NMME home page

http://www.cpc.ncep.noaa.gov/products/NMME/

(web search "nmme")

Deterministic and probabilistic forecasts

Prate 2015 OND from 201509

Deterministic Models weighted equally

Probabilistic Ensemble members weighted equally*

*Anomalies and tercile boundaries computed separately for each model

Raw and calibrated probabilistic forecasts Prate 2016 DJF from 201511

Raw and calibrated probabilistic forecasts Prate 2016 DJF from 201511

(overconfident)

Calibrated probabilistic (more reliable)

NMME Nino3.4 plumes

All Members

Ensemble Mean NMME Forecast for Nino 3.4 IC= 201611 weak La Niña in

NMME Forecast for Nino 3.4 IC= 201611 CFSk2 CMC1 CMC2 OFDS. NASA NCAR CESM GTDL_FLOR A NCAR CESNA statel DEC 20 158 -

Ens Mean + IMME

CFSv2 CFSv2

CMC1 CanCM3

GFDL FLOR

GFDL CM2.1 GFDL Forecast for Nino 3.4 IC= 201611

101 AN 101 MA

NASA GEOS5

35

site and any rise and site

NCAR CESM

NMME for International Regions

http://www.cpc.ncep.noaa.gov/products/international/nmme/nmme.shtml

(web search "nmme international")

NMME SEASONAL FORECASTS FOR INTERNATIONAL REGIONS

SEASONAL FORECASTS		MONTHLY	FORECASTS DA	TA DOWNLOAD	S VERI	VERIFICATION	
SEA SURFACE TEMPERATURE							
Region Model	Anomalies	StdAnom	Masked StdAnom	SkillMaps	ProbAnom	3Category Prob	
Global	0	0	0	0	0	0	
Pacific	0	0	0	0	0	0	
Atlantic	0	0	0	0	0	0	
Indian Ocean	0	0	0	0	0	0	
Atlantic&Indian	0	0	0	0	0	0	
			PRECIPITATION				
Region Model	Anomalies	StdAnom	Masked StdAnom	SkillMaps	ProbAnom	3Category Prob	
Global	0	0	0	0	0	0	
Africa	0	0	0	0	0	0	
CAM&Caribbean	0	0	0	0	0	0	
Maritime-CONT	0	0	0	0	0	0	
Central Asia	0	0	0	0	0	0	
East Asia	0	0	0	0	0	0	
South Asia	0	0	0	0	0	0	
South America	0	0	0	0	0	0	
2-METER AIR TEMPERATURE							
Region Model	Anomalies	StdAnom	Masked StdAnom	SkillMaps	ProbAnom	3Category Prob	
Global	0	0	0	0	0	0	
Africa	0	0	0	0	0	0	
CAM&Caribbean	0	0	0	0	0	0	
Maritime-CONT	0	0	0	0	0	0	
Central Asia	0	0	0	0	0	0	
East Asia	0	0	0	0	0	0	
South Asia	0	•	0	0	0	0	
South America	0	0	0	0	0	0	
Precipitation forecasts from Nov 2016

AFRICA NMME SEASONAL PRECIPITATION ANOMALIES November2016 INITIAL CONDITIONS

	CFSv2	CMC1	СМС2	GFDL	GFDL-FLOR	NASA	NCAR-CCSM4	ENSEMBLE MEAN
Dec2016-Feb2017								
Jan2017-Mar2017								
Feb2017-Apr2017								
Mar2017-May2017								
Apr2017-Jun2017								

NMME International Data

http://ftp.cpc.ncep.noaa.gov/International/nmme/

Index of /International/nmme

Name	Last modified	<u>Size</u>
Parent Directory		_
binary_monthly/	09-Nov-2016 17:23	-
binary_seasonal/	09-Nov-2016 17:22	-
monthly_nmme_forecast_in_cpt_forma	<u>t/</u> 09-Nov-2016 18:28	-
monthly_nmme_hindcast_in_cpt_forma	<u>at/</u> 09-Nov-2016 18:34	-
readme	16-Oct-2015 17:29	823
seasonal_nmme_forecast_in_cpt_forma	<u>tt/</u> 09-Nov-2016 18:02	-
seasonal_nmme_hindcast_in_cpt_forma	<u>at/</u> 09-Nov-2016 18:08	-

Data is freely accessible!

NMME Data at RI

Hindcasts + real time forecasts

Real-Time Monthly fields (8)

2m T daily max

2m T daily min

2m temperature

200 mb Geopotential

Total precipitation

Total soil moisture

Surface temperature (SST-land)

Surface runoff

Data is freely accessible!

IRI



Finding Data Tutorial Questions and Answers Function Documentation



Models NMME options Help Expert Mode

SOURCES Models NMME

Models NMME

Models NMME from SOURCES: the IRI/LDEO collection of climate data.

Documents

an outline showing sub-datasets of this dataset overview CTB home Climate Test Bed NMME Home Information about the NMME project

Semantic Documents

auxinfo.owl

Datasets and variables

CMC1-CanCM3	Models	NMME	CMC1-CanCM3[FORECAST HINDCAST]
CMC2-CanCM4	Models	NMME	CMC2-CanCM4[FORECAST HINDCAST]
COLA-RSMAS-CCSM3	Models	NMME	COLA-RSMAS-CCSM3[MONTHLY]
CPC-CMAP	Models	NMME	CPC-CMAP[prate]
CPC-PRECIP	Models	NMME	CPC-PRECIP[prate]
GFDL-CM2p1	Models	NMME	GFDL-CM2p1[MONTHLY]
GFDL-CM2p1-aer04	Models	NMME	GFDL-CM2p1-aer04[MONTHLY]
GFDL-CM2p5-FLOR-A06	Models	NMME	GFDL-CM2p5-FLOR-A06[MONTHLY]
GFDL-CM2p5-FLOR-B01	Models	NMME	GFDL-CM2p5-FLOR-B01 [MONTHLY]
GHCN_CAMS	Models	NMME	GHCN_CAMS[temp]
IRI-ECHAM4p5-AnomalyCoupled	Models	NMME	IRI-ECHAM4p5-AnomalyCoupled[MONTHLY]
IRI-ECHAM4p5-DirectCoupled	Models	NMME	IRI-ECHAM4p5-DirectCoupled[MONTHLY]
LSMASK	Models	NMME	LSMASK[land]
NASA-GMAO	Models	NMME	NASA-GMA0[MONTHLY]
NASA-GMA0-062012	Models	NMME	NASA-GMA0-062012[MONTHLY]
NCDC-0ISST	Models	NMME	NCDC-0ISST[sst]
NCEP-CFSv1	Models	NMME	NCEP-CFSv1 [MONTHLY]
NCEP-CFSv2	Models	NMME	NCEP-CFSv2[MONTHLY]

http://iridl.ldeo.columbia.edu/SOURCES/.Models/.NMME

NMME PRECIPITATION SEASON1 Standard Anomalies

NMME PRECIPITATION SEASON1 Skill Maps

Initial Conditions: 01Sep2015-08Sep2015 Oct-Dec2015 Forecast

Initial Conditions: 01Sep2015-08Sep2015 Oct-Dec2015 Forecast



-1.5 -1 -0.5 -0.25 0.25 0.5



206

40X 50X



ENSMEAN



CFSv2

CMC2 CMC2 Precipitation Forecast Skill



GFDL-FLOR GFDL-FLOR Precipitation Forecast Skill



CanSIPS Explorer

- Developed and maintained at CCCma by Slava Kharin
- Displays all monthly, seasonal hind/forecasts + verifications 1979-present + skills
- Probabilistic/deterministic forecasts (maps & local PDFs) for many variables, regions (including Africa), indices



username: cccmasf password: seasforum

Daily N-day, monthly & seasonal forecasts

What is El Niño?

















cooler water

Typical buildup of a strong El Niño: the role of westerly wind bursts (WWB)



Typical buildup of a strong El Niño: the role of westerly wind bursts (WWB)



Typical buildup of a strong El Niño: the role of westerly wind bursts (WWB)



2-3 months later



2-3 months later



2-3 months later



SST anomaly - "El Niño"



SLP anomaly - "Southern Oscillation"



El Niño Southern Oscillation (ENSO)

2°S-2°N Average, 3 Pentad Running Mean



2°S-2°N Average, 3 Pentad Running Mean



2°S-2°N Average, 3 Pentad Running Mean



-2.1-1.8-1.5-1.2-0.9-0.6-0.3 0.3 0.6 0.9 1.2 1.5 1.8 2.1

2°S-2°N Average, 3 Pentad Running Mean



-2.1-1.8-1.5-1.2-0.9-0.6-0.3 0.3 0.6 0.9 1.2 1.5 1.8 2.1

NOAA/NCEP/CPC Monthly Ocean Briefing http://www.cpc.ncep.noaa.gov/products/GODAS/

2°S-2°N Average, 3 Pentad Running Mean



-2.1-1.8-1.5-1.2-0.9-0.6-0.3 0.3 0.6 0.9 1.2 1.5 1.8 2.1

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