



Environment and  
Climate Change Canada

Environnement et  
Changement climatique Canada

Canada

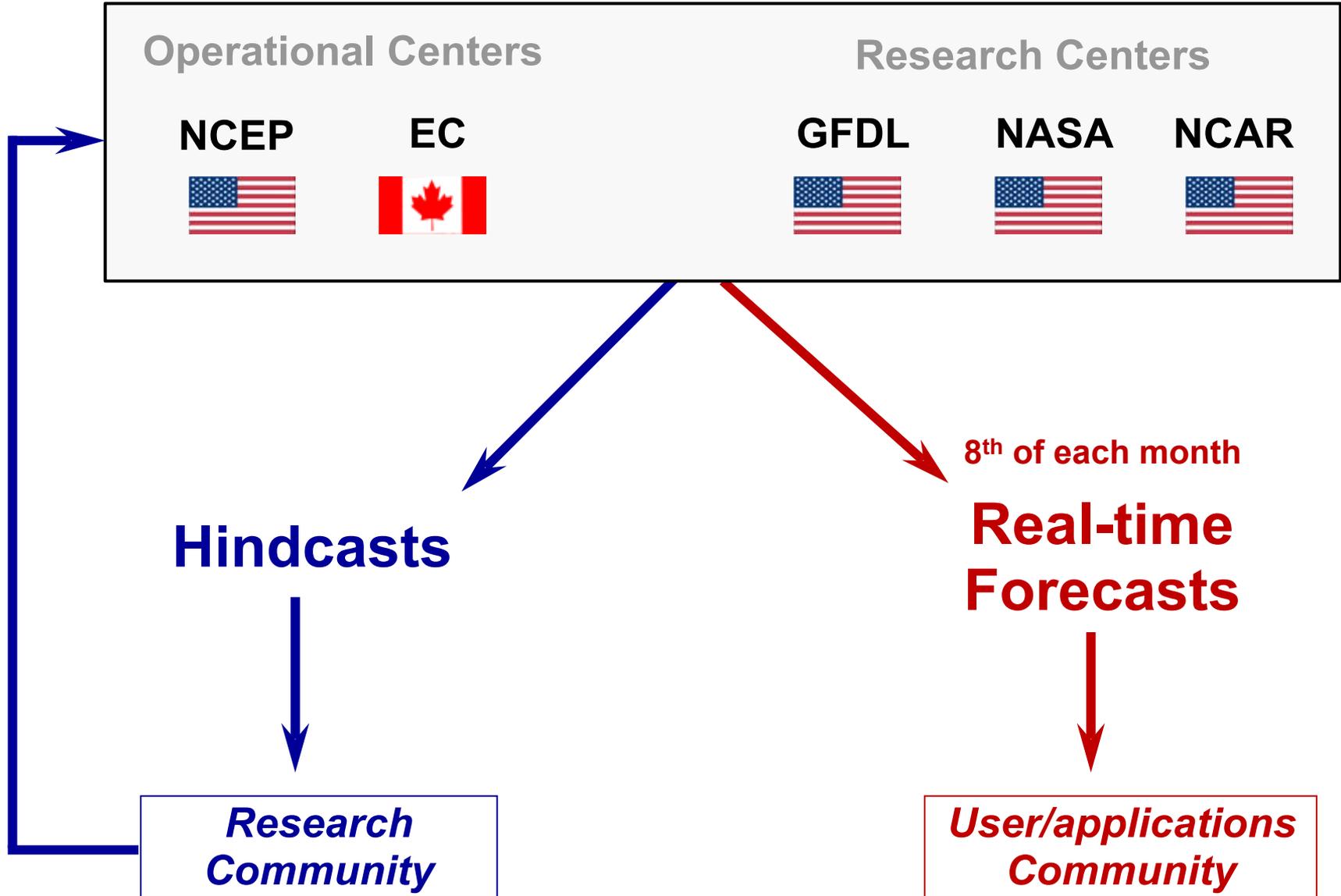


# North American Multi-Model Ensemble (NMME) update

**Bill Merryfield**

**Canadian Centre for Climate Modelling and Analysis (CCCma)**

# NMME



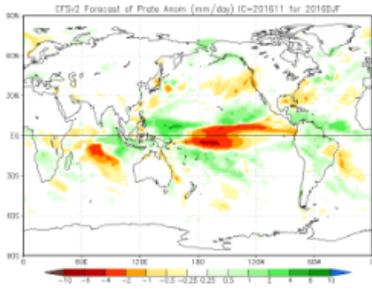
# Currently contributing models

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
<b>NEW</b> CESM1	NCAR	10
Total ensemble size		109 (113)

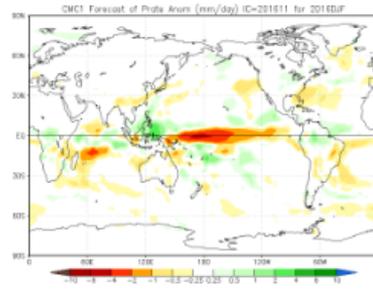
**Individual model forecasts**

2016 DJF  
Lead 1 mon

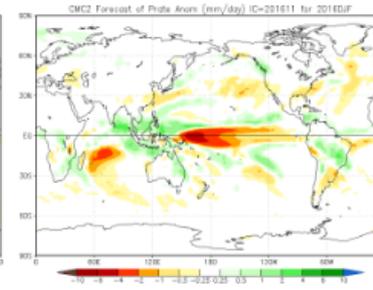
NCEP\_CFSv2



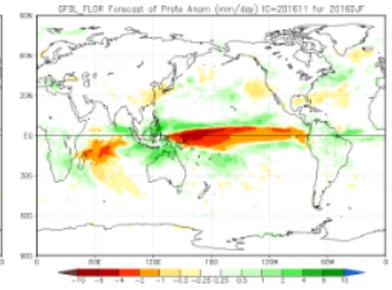
CMC1\_CanCM3



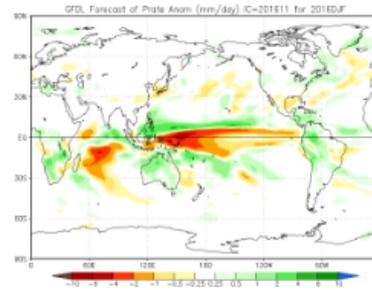
CMC2\_CanCM4



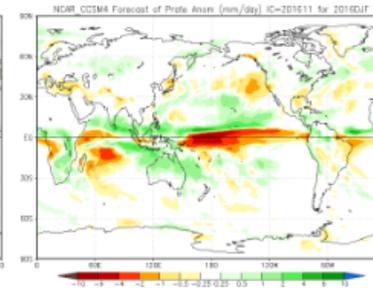
GFDL\_FLOR



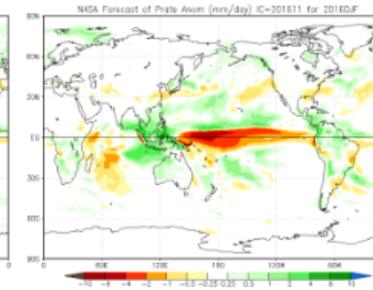
GFDL\_CM2.1



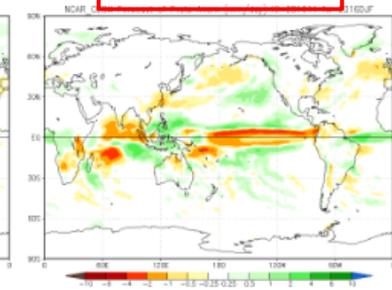
NCAR\_CCSM4



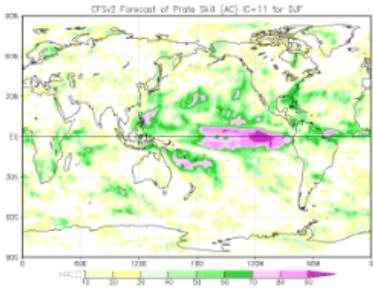
NASA\_GEOS5



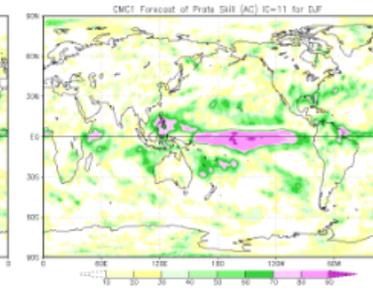
**NCAR\_CESM**



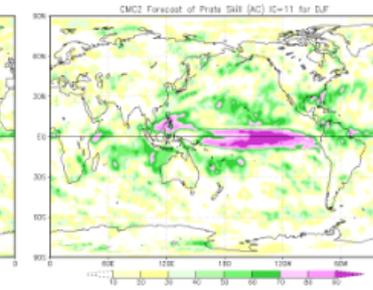
NCEP\_CFSv2



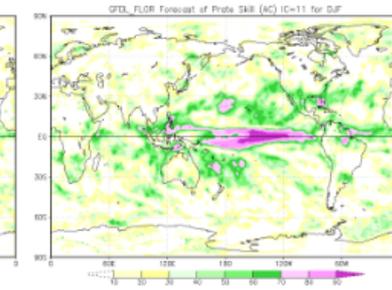
CMC1\_CanCM3



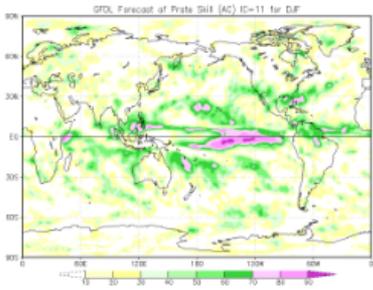
CMC2\_CanCM4



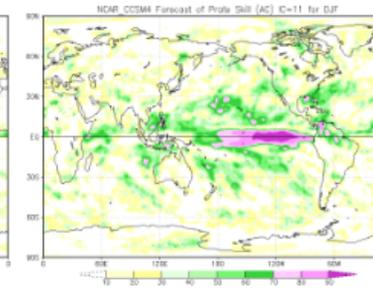
GFDL\_FLOR



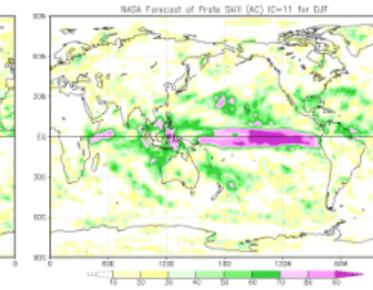
GFDL\_CM2.1



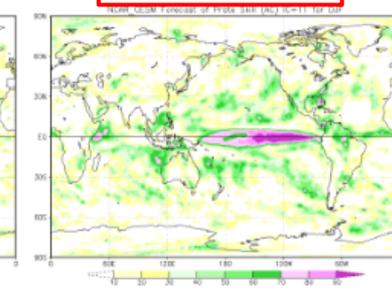
NCAR\_CCSM4



NASA\_GEOS5



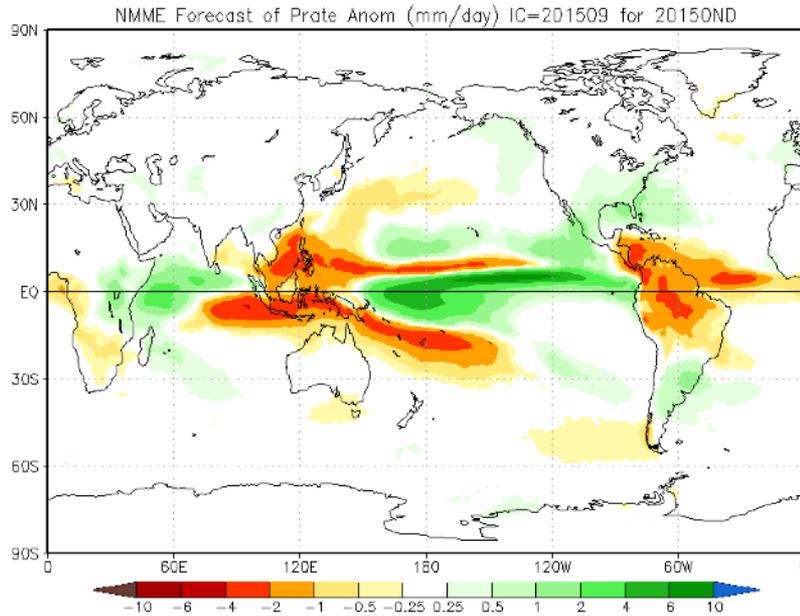
**NCAR\_CESM**



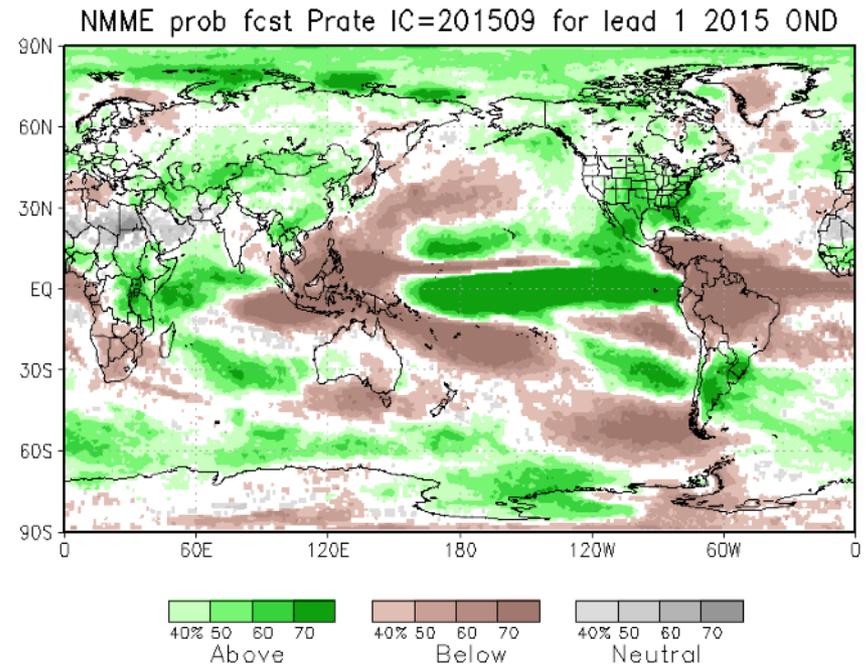
**Individual model skills**

# Deterministic and probabilistic forecasts

## Prate 2015 OND from 201509



Deterministic  
*Models weighted equally*

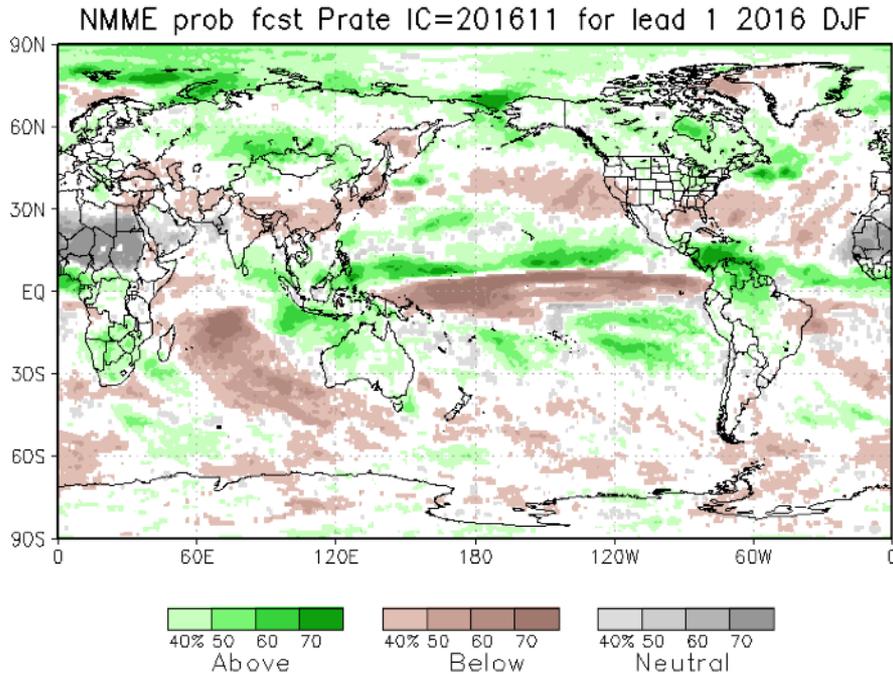


Probabilistic  
*Ensemble members weighted equally\**

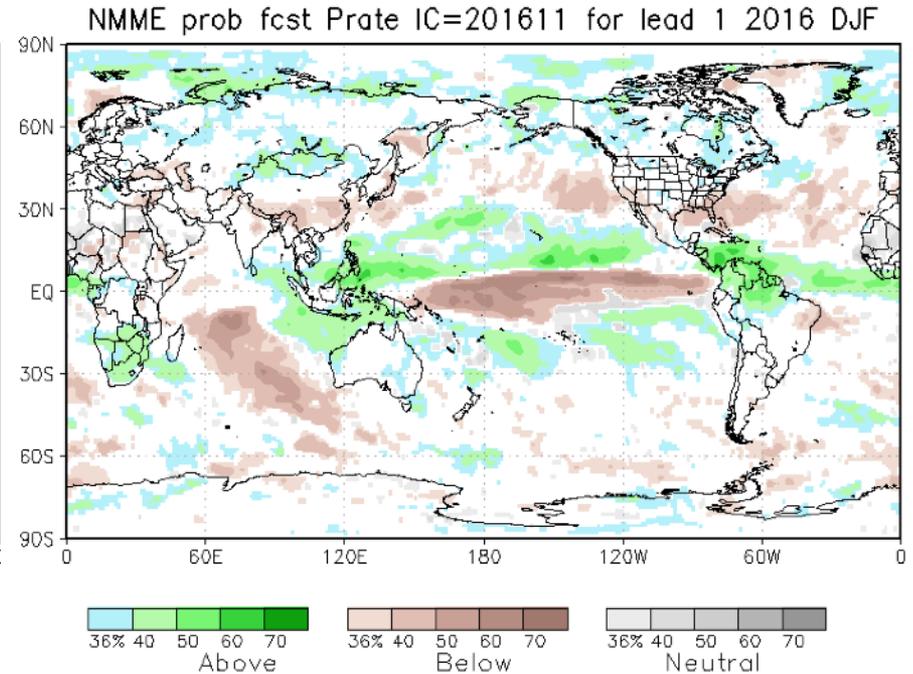
\*Anomalies and tercile boundaries computed separately for each model

# Raw and calibrated probabilistic forecasts

Prate 2016 DJF from 201511



Raw probabilistic  
(overconfident)



Calibrated probabilistic  
(more reliable)

# NMME Subseasonal Experiment

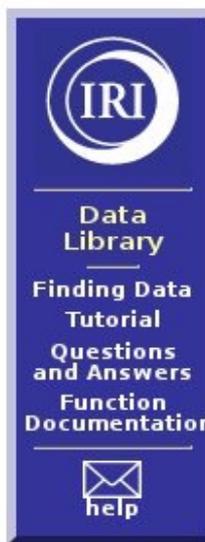
- Weekly initialization
- Forecast length  $\geq 32$  days (45 days encouraged)
- Hindcast period 1999-2015 (additional years encouraged)
- $\geq 4$  ensemble members (more encouraged)
- Hindcasts and real-time forecasts (product based, like seasonal NMME)
- Models can differ from seasonal NMME, e.g. CMC GEM model contributing to S2S will replace CanCM3/4

# NMME Phase 1 Data at IRI

## *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



Models NMME options [Help](#) [Expert Mode](#)

[SOURCES](#) [Models](#) [NMME](#)

## Models NMME

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

### Documents

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

### Semantic Documents

[auxinfo.owl](#)

### Datasets and variables

<a href="#">CMC1-CanCM3</a>	Models NMME CMC1-CanCM3[ <b>FORECAST HINDCAST</b> ]
<a href="#">CMC2-CanCM4</a>	Models NMME CMC2-CanCM4[ <b>FORECAST HINDCAST</b> ]
<a href="#">COLA-RSMAS-CCSM3</a>	Models NMME COLA-RSMAS-CCSM3[ <b>MONTHLY</b> ]
<a href="#">CPC-CMAP</a>	Models NMME CPC-CMAP[ <b>prate</b> ]
<a href="#">CPC-PRECIP</a>	Models NMME CPC-PRECIP[ <b>prate</b> ]
<a href="#">GFDL-CM2p1</a>	Models NMME GFDL-CM2p1 [ <b>MONTHLY</b> ]
<a href="#">GFDL-CM2p1-aer04</a>	Models NMME GFDL-CM2p1-aer04[ <b>MONTHLY</b> ]
<a href="#">GFDL-CM2p5-FLOR-A06</a>	Models NMME GFDL-CM2p5-FLOR-A06[ <b>MONTHLY</b> ]
<a href="#">GFDL-CM2p5-FLOR-B01</a>	Models NMME GFDL-CM2p5-FLOR-B01 [ <b>MONTHLY</b> ]
<a href="#">GHCN_CAMS</a>	Models NMME GHCN_CAMS[ <b>temp</b> ]
<a href="#">IRI-ECHAM4p5-AnomalyCoupled</a>	Models NMME IRI-ECHAM4p5-AnomalyCoupled[ <b>MONTHLY</b> ]
<a href="#">IRI-ECHAM4p5-DirectCoupled</a>	Models NMME IRI-ECHAM4p5-DirectCoupled[ <b>MONTHLY</b> ]
<a href="#">LSMASK</a>	Models NMME LSMASK[ <b>land</b> ]
<a href="#">NASA-GMAO</a>	Models NMME NASA-GMAO[ <b>MONTHLY</b> ]
<a href="#">NASA-GMAO-062012</a>	Models NMME NASA-GMAO-062012[ <b>MONTHLY</b> ]
<a href="#">NCDC-OISST</a>	Models NMME NCDC-OISST[ <b>sst</b> ]
<a href="#">NCEP-CFSv1</a>	Models NMME NCEP-CFSv1 [ <b>MONTHLY</b> ]
<a href="#">NCEP-CFSv2</a>	Models NMME NCEP-CFSv2[ <b>MONTHLY</b> ]

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



## Search Results

# NMME Phase 2 Data on the ESG

<https://www.earthsystemgrid.org/search.html?Project=NMME>

### Project

NMME [remove](#)

[Search Help](#)

1 - 20 of 8474 results

Show: 20 [50](#) [100](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=day, modeling\\_realm=atmos](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=day, modeling\\_realm=land](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=mon, modeling\\_realm=atmos](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=mon, modeling\\_realm=land](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=mon, modeling\\_realm=landIce](#)

[project=NMME, model=CanCM3, experiment=19810101, time\\_frequency=mon, modeling\\_realm=ocean](#)

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[project=NMME, model=CanCM3, experiment=19810201, time\\_frequency=day, modeling\\_realm=atmos](#)

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### Institute

[CCMA](#) (5331)

[NASA-GMAO](#) (365)

[NCEP](#) (3)

[NOAA-GFDL](#) (1660)

[UM-RSMAS](#) (1115)

### Model

[CCSM4](#) (1115)

[CFSV2-2011](#) (3)

[CanCM3](#) (2660)

[CanCM4](#) (2671)

[FLORB-01](#) (1660)

[GEOS-5](#) (365)

[...Show Fewer](#)

### Experiment

[19800101](#) (4)

[19800201](#) (4)

[19800301](#) (4)

[19800401](#) (4)

[19800501](#) (4)

[...Show More](#)

### Frequency

[3-Hourly](#) (339)

[6-Hourly](#) (3)

[Daily](#) (2679)

[Monthly](#) (5453)

NMME (North-American Multi-Model Ensemble) is to improve intra-seasonal to interannual (ISI) operational predictions based on the leading US and Canada climate models.

# NMME

NORTH AMERICAN MULTI-MODEL ENSEMBLE

## Publications

2015

### Data Analyses and Applications

Antonietta Capotondi, Andrew T. Wittenberg, Matthew Newman, Emanuele Di Lorenzo, Jin-Yi Yu, Pascale Braconnot, Julia Cole, Boris Dewitte, Eric Guilyardi, Fei-Fei Jin, Christopher Karamuskas, Benjamin Kirtman, Tong Lee, Niklas Schneider, Yan Xue, and Sang-Wook Yeh, 2015: Understanding ENSO diversity. *Bull. Amer. Meteor. Soc.*, 96, 921–938. doi: <http://dx.doi.org/10.1175/BAMS-D-13-00117.1>

Yoo-Geun Ham and Jong-Seong Kug, 2015: Improvement of ENSO simulation based on intermodel diversity. *J. Climate*, 28, 998–1015. doi: <http://dx.doi.org/10.1175/JCLI-D-14-00376.1>

Johnna M. Infanti, Ben P. Kirtman, 2015: North American rainfall and temperature prediction response to the diversity of ENSO. *Clim. Dyn.*, e-View. doi: <http://dx.doi.org/10.1007/s00382-015-2748-0>

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Ben Kirtman, 2015: Current status of ENSO prediction and predictability. *US CLIVAR VARIATIONS*, 13, 10–15. [http://www.usclivar.org/sites/default/files/documents/2015/Annations2015/Winter\\_D.pdf#page=10](http://www.usclivar.org/sites/default/files/documents/2015/Annations2015/Winter_D.pdf#page=10)

Sarah M. Larson and Ben P. Kirtman, 2015: Revisiting ENSO coupled instability theory and SST error growth in a fully coupled model. *J. Climate*, 28, 4724–4742. doi: <http://dx.doi.org/10.1175/JCLI-D-14-00731.1>

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Kingtse C. Mo and Bradford Lyon, 2015: Global meteorological drought prediction using the North American Multi-Model Ensemble. *J. Hydrometeorol.*, 16, 1409–1424. doi: <http://dx.doi.org/10.1175/JHM-D-14-0102.1>

Andrew W. Robertson, Arun Kumar, Malaquias Peña, and Frederic Utari, 2015: Improving and promoting subseasonal to seasonal prediction. *Bull. Amer. Meteor. Soc.*, 96, ES49–ES53. doi: <http://dx.doi.org/10.1175/BAMS-D-14-00139.1>

Joshua K. Roundy, Xing Yuan, John Schaake, and Eric F. Wood, 2015: A framework for diagnosing seasonal prediction through canonical event analysis. *Mon. Wea. Rev.*, 143, 2404–2418. doi: <http://dx.doi.org/10.1175/MWR-D-14-00190.1>

Richard Seager, Martin Hoerling, Siegfried Schubert, Hailan Wang, Bradford Lyon, Arun Kumar, Jennifer Nakamura, and Naomi Henderson, 2015: Causes of the 2011–14 California drought. *J. Climate*, 28, 6997–7024. doi: <http://dx.doi.org/10.1175/JCLI-D-14-00880.1>

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Stephan Thober, Rohini Kumar, Justin Sheffield, Juliane Mai, David Schäfer, Luis Samaniego, 2015: Seasonal soil moisture drought prediction over Europe using the North American Multi-Model Ensemble (NMME). *J. Hydrometeorol.*, e-View. doi: <http://dx.doi.org/10.1175/JHM-D-15-0053.1>

Eric F. Wood, Siegfried D. Schubert, Andrew W. Wood, Christa D. Peters-Lidard, Kingtse C. Mo, Annarita Mariotti, and Roger S. Pulwarty, 2015: Prospects for advancing drought understanding, monitoring, and prediction. *J. Hydrometeorol.*, 16, 1636–1657. doi: <http://dx.doi.org/10.1175/JHM-D-14-0104.1>

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### Model and System Improvements

Anthony G. Barnston, Michael K. Tippett, Huug M. van den Dool, and David A. Unger, 2015: Toward an improved multimodel ENSO prediction. *J. Appl. Meteorol. Climatol.*, 54, 1579–1595. doi: <http://dx.doi.org/10.1175/JAMC-D-14-01088.1>

Timothy DelSole, Michael K. Tippett, 2015: Forecast comparison based on random walks. *Mon. Wea. Rev.*, e-View. doi: <http://dx.doi.org/10.1175/MWR-D-15-0218.1>

Liwai Jia, Xiaosong Yang, Gabriel A. Vecchi, Richard G. Gudgel, Thomas L. Delworth, Anthony Rosati, William F. Stern, Andrew T. Wittenberg, Lakshmi Krishnamurthy, Shaogang Zhang, Rym Msadek, Sarah Kapnick, Seth Underwood, Wenh G. Anderson, Venkatesan Balaji, and Keith Dixon, 2015: Improved seasonal prediction of temperature and precipitation over land in a high-resolution GFDL climate model. *J. Climate*, 28, 2044–2062. doi: <http://dx.doi.org/10.1175/JCLI-D-14-00112.1>

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Rajesh R. Shrestha, Markus A. Schnorbus, and Alex J. Cannon, 2015: A dynamical climate model–driven hydrologic prediction system for the Fraser River, Canada. *J. Hydrometeorol.*, 16, 1273–1292. doi: <http://dx.doi.org/10.1175/JHM-D-14-0107.1>

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## Publications

2014

### NMME Overview Paper

Ben P. Kirtman, Dughong Min, Johnna M. Infanti, James L. Kinter III, Daniel A. Paolino, Qin Zhang, Huug van den Dool, Suranjana Saha, Malaquias Pena Mendez, Emily Becker, Peitao Peng, Patrick Tripp, Jin Huang, David G. DeWitt, Michael K. Tippett, Anthony G. Barnston, Shuhua Li, Anthony Rosati, Siegfried D. Schubert, Michele Rienecker, Max Suarez, Zhao E. Li, Jelena Marshak, Young-Kwon Lim, Joseph Tribbia, Kathleen Pegion, William J. Merryfield, Bertrand Denis, and Eric F. Wood, 2014: The North American multimodel ensemble: phase-1 seasonal-to-interannual prediction; phase-2 toward developing intraseasonal prediction. *Bull. Amer. Meteor. Soc.*, 95, 585–601. doi: <http://dx.doi.org/10.1175/BAMS-D-12-00050.1>

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Johnna M. Infanti and Ben P. Kirtman, 2014: Southeastern U.S. rainfall prediction in the North American Multi-Model Ensemble. *J. Hydrometeorol.*, 15, 529–550. doi: <http://dx.doi.org/10.1175/JHM-D-13-072.1>

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Sarah M. Larson and Ben P. Kirtman, 2014: The Pacific meridional mode as an ENSO precursor and predictor in the North American multimodel ensemble. *J. Climate*, 27, 7018–7032. doi: <http://dx.doi.org/10.1175/JCLI-D-14-00065.1>

Haiqin Li, Vasubandhu Misra, 2014: Global seasonal climate predictability in a two tiered forecast system. part II: boreal winter and spring seasons. *Climate Dyn.*, 42, 1449–1469. doi: <http://dx.doi.org/10.1007/s00382-013-1813-y>

Hosmay Lopez and Ben P. Kirtman, 2014: WWBS: ENSO predictability, the spring barrier and extreme events. *J. Geophys. Res.*, doi: <http://dx.doi.org/10.1002/2014JD022109>

Vasubandhu Misra, and H. Li, 2014: The seasonal climate predictability of the Atlantic Warm Pool and its teleconnections. *Geophys. Res. Lett.*, 41, 691–699. doi: <http://dx.doi.org/10.1002/2013GL058740>

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Kingtse C. Mo and Dennis P. Lettenmaier, 2014: Hydrologic prediction over the Conterminous United States using the National Multi-Model Ensemble. *J. Hydrometeorol.*, 15, 1467–1472. doi: <http://dx.doi.org/10.1175/JHM-D-13-0107.1>

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# NMME Phase 2 Data at NCAR

- **Common 1° grid**
- **NetCDF4**
- **Available 2014-15**

## Daily atmospheric and land surface fields (22)

Variable	Var. Name
Surface temperature (SST+land)	Ts
2m T daily max	Tasmax
2m T daily min	Tasmin
Mean sea level pressure	Ps1
Snow water equivalent	swe
Total soil moisture	Mrsov
Total precipitation*	prlr
Downward surface solar	Rsds
Downward surface longwave	Rlds
Net surface solar	Rss
Net surface longwave	Rls
Top net solar	Rst
Top net longwave	Rlt
Surface latent flux	Hflsd
Surface sensible flux	Hfssd
Surface stress (x)	Tauu
Surface stress (y)	Tauv
2m temperature	Tas
Total cloud cover	Clt
10m wind (u)	Uas
10m wind (v)	Vas
Surface specific humidity	huss

## Daily atmospheric pressure level fields (5)

Provided at 850, 500, 200, 100, 50 hPa

Variable	Var. Name
Geopotential	G
Temperature	Ta
Zonal velocity	ua
Meridional velocity	va
Specific humidity	hus

## Monthly sea ice fields (2)

Variable	Var. Name
Sea ice concentration	sic
Sea ice thickness	sit

## Monthly ocean fields (7)

3D ocean fields thetao/so/uo/vo/wo are provided at 125.0, 150.0, 200.0, 250.0, 300.0, and 400.0 m

Variable	Var. Name
Potential temperature	thetao
Salinity	so
Zonal velocity	uo
Meridional velocity	vo
Vertical velocity	wo
Sea level	zoh
Mixed layer depth	zmlo