

Understanding climate information from APCC

October 15, 2018

Ji-Hyun Oh

APEC Climate Center, Climate Prediction Department

What does APEC Climate Center do?

- Established in 2005 as a climate prediction center during the 13th APEC Leaders' Meeting in Busan, South Korea
- Aim: enhance socio-economic wellbeing of APEC countries using climate information

MISSION

Use climate science and its applications to contribute to safer, more prosperous, and more resilient communities through four interrelated themes:



Climate Prediction



Climate Information Services

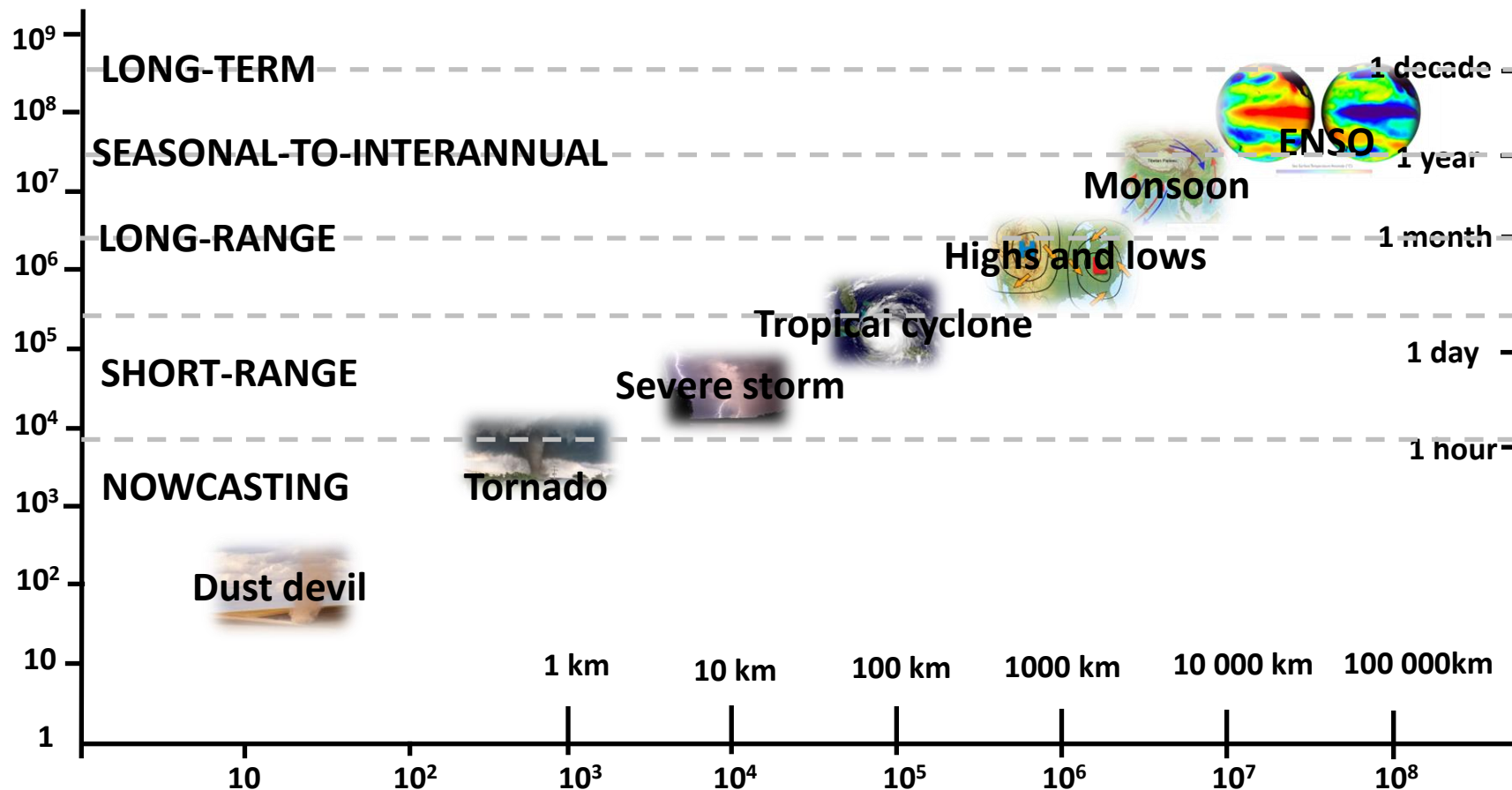


Interdisciplinary Research



International Cooperation

Spatial and temporal scale



(Adapted from: J.W., Zillman, WMO Bulletin 48 (2) April 1999)

What does APEC Climate Center do?



LOGIN CONTACT US SITE MAP 한국어 Q

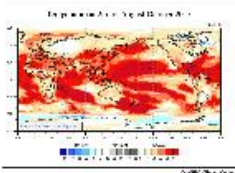
Research Climate Information Services International Cooperation Media Notices About us



CLIMATE INFORMATION SERVICES



FORECASTS



Climate Outlook for August 2017 - January 2018

BUSAN, 25 July 2017 - The synthesis of the latest model forecasts for August 2017 to January 2018 (ASONDJ) from the APEC Climate Center (APCC), located in Busan, South Korea, indicates a persistent we...

NOTICE

APCC News	Employment	APCC Seminars
Notice	APEC Climate Center's 2017 Pacific I...	07.13
Notice	Dr. Woo-Seop Lee (Team Leader, CLI...	07.13
Notice	APEC Climate Center Signed an MO...	07.04
Notice	Application Deadline for the 2017 A...	06.20
Notice	APEC Climate Center held the 2017...	06.20

Activity Schedule



P.R.



Add our website to your favorites!



<http://www.apcc21.org>

What does APEC Climate Center do?

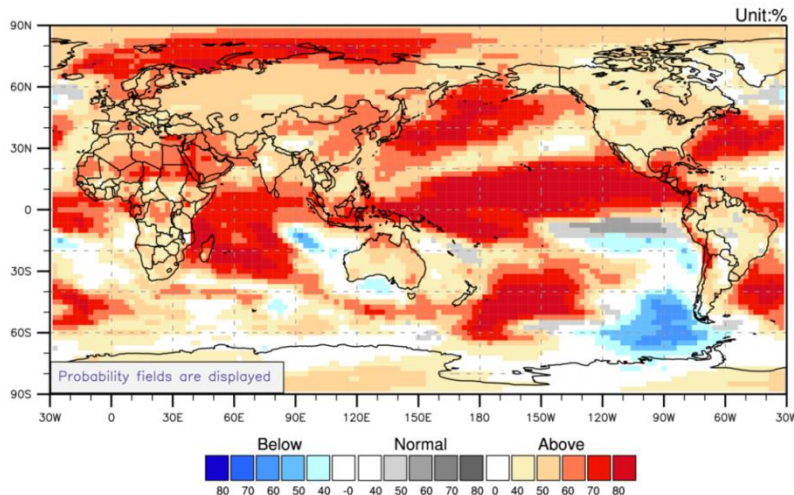
CLIMATE INFORMATION SERVICES



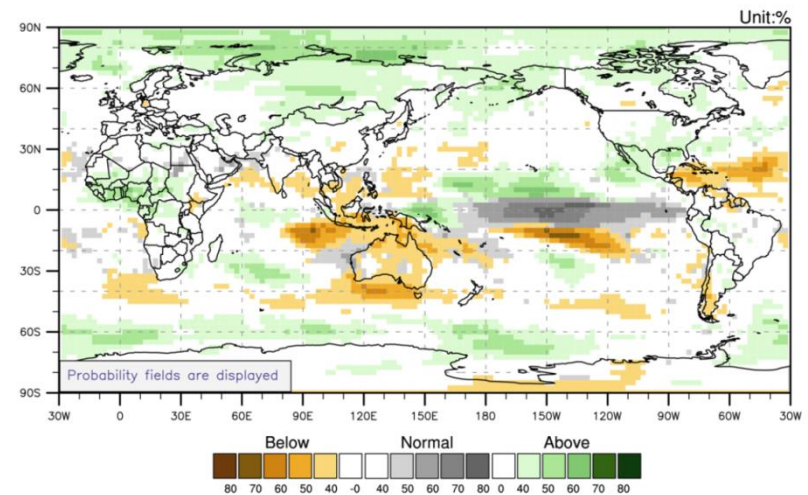
- Information about the “climate” that can be expected in the coming months

Forecast for October-December 2018

Temperature at 2m for October-December 2018



Precipitation for October-December 2018



Seasonal forecast: Temperature and Precipitation Outlook (Oct–Dec 2018)

Strongly enhanced probability for above normal temperatures is predicted for the eastern Arctic Ocean, Bering Sea, tropical North Pacific, northern North Pacific, central South Pacific, equatorial and subtropical Atlantic, and the western and central Indian Ocean. **Enhanced probability for above normal temperatures is expected for** the Arctic, Antarctic, Eurasia, Africa, eastern and northern Australia, Gulf of Mexico, Caribbean Sea, Central America (excluding northern Mexico), and northern Chile. A tendency for above normal temperatures is predicted for North and South America. **Enhanced probability for below normal temperatures is predicted for** the southeastern South Pacific and the eastern Indian Ocean. A tendency for below normal temperatures is expected for the eastern subtropical South Pacific and the Great Australian Bight. **Enhanced probability for near normal temperatures is predicted for** the eastern off-equatorial South Pacific. A tendency for near normal temperatures is expected for the northern North Atlantic and the Coral Sea. **Enhanced probability for above normal precipitation is predicted for** the eastern Arctic Ocean, central off-equatorial North Pacific, and West Africa. **A tendency for above normal precipitation is expected for** the Arctic, Antarctic Ocean, Bering Sea, eastern and central Russia, Mongolia, Alaska, northern Canada, eastern USA, Mexico, and the Gulf of Mexico. **Enhanced probability for below normal precipitation is predicted for** the central off-equatorial South Pacific, off-equatorial North Atlantic, Caribbean Sea, eastern Indian Ocean, and the Great Australian Bight. **A tendency for below normal precipitation is expected for** the Arabian Sea, Bay of Bengal, South China and Philippine Seas, Australia, the Coral Sea, and the seas surrounded by Indonesia and Australia. **Strongly enhanced probability for near normal precipitation is predicted for** the central equatorial Pacific. Enhanced probability for near normal precipitation is expected for the equatorial belt of the Pacific. **A tendency for near normal precipitation is predicted for** some parts of the eastern Indian Ocean near western Australia.

How do we predict?



Seasonal forecast: Method

	Statistical prediction	Dynamical prediction
	<ul style="list-style-type: none">▪ Use observed relationship of climate system to predict future	<ul style="list-style-type: none">▪ Based on physical laws of the climate system
Pros	<ul style="list-style-type: none">▪ Simple▪ Cheap	<ul style="list-style-type: none">▪ Nonlinearity can be considered.▪ Spatiotemporally coherent variables.
Cons	<ul style="list-style-type: none">▪ Stationarity – climate is changing.▪ Limited data	<ul style="list-style-type: none">▪ Complicated & expensive▪ Not real – biased & need correction.

Seasonal forecast: Statistical prediction

- Use past observational record and statistical methods

(0) Climatology (use 30 year average as a baseline)

$$x(t + 1) = \bar{x}$$

(1) Persistence (Assume that future will be same as it is now)

$$x(t + 1) = x(t)$$

(2) Regression (The most popular method and many variations)

$$x(t + 1) = ax(t) + b$$

Anomalies (deviation) vs Climatology (mean)

Climatology:

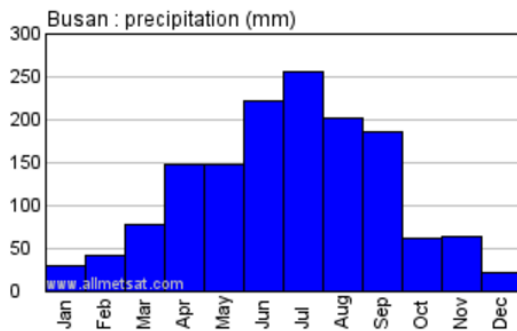
1. total mean of all data over the observing record - for instance 20 years of a particular data product.
2. a monthly climatology, which would be the monthly mean over the years in the observing record



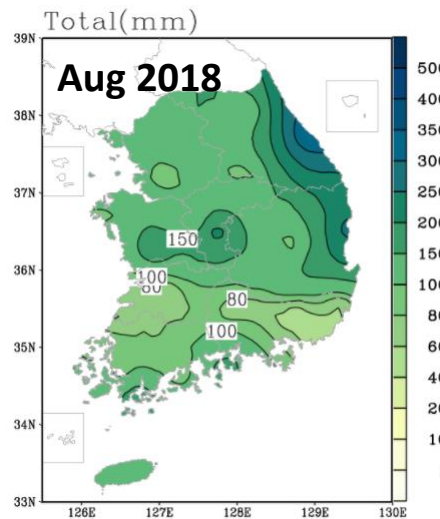
Seasonal forecast: Statistical prediction

- Predict rainfall in Busan for coming December
 - (0) Climatology (use 30 year average as a baseline)
 - (1) Persistence (Assume that future will be same as it is now)
 - (2) Regression (The most popular method and many variations)

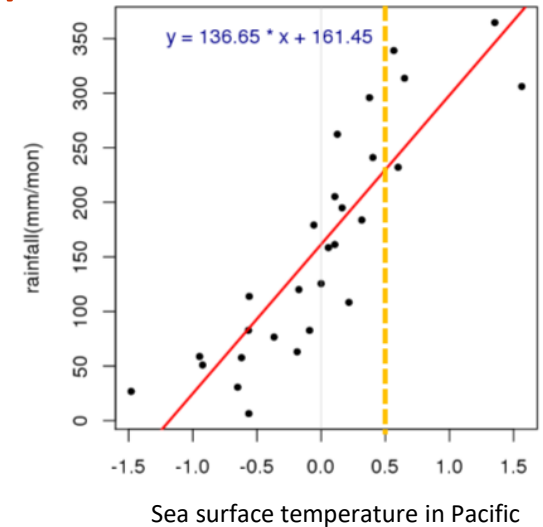
(0)



(1)

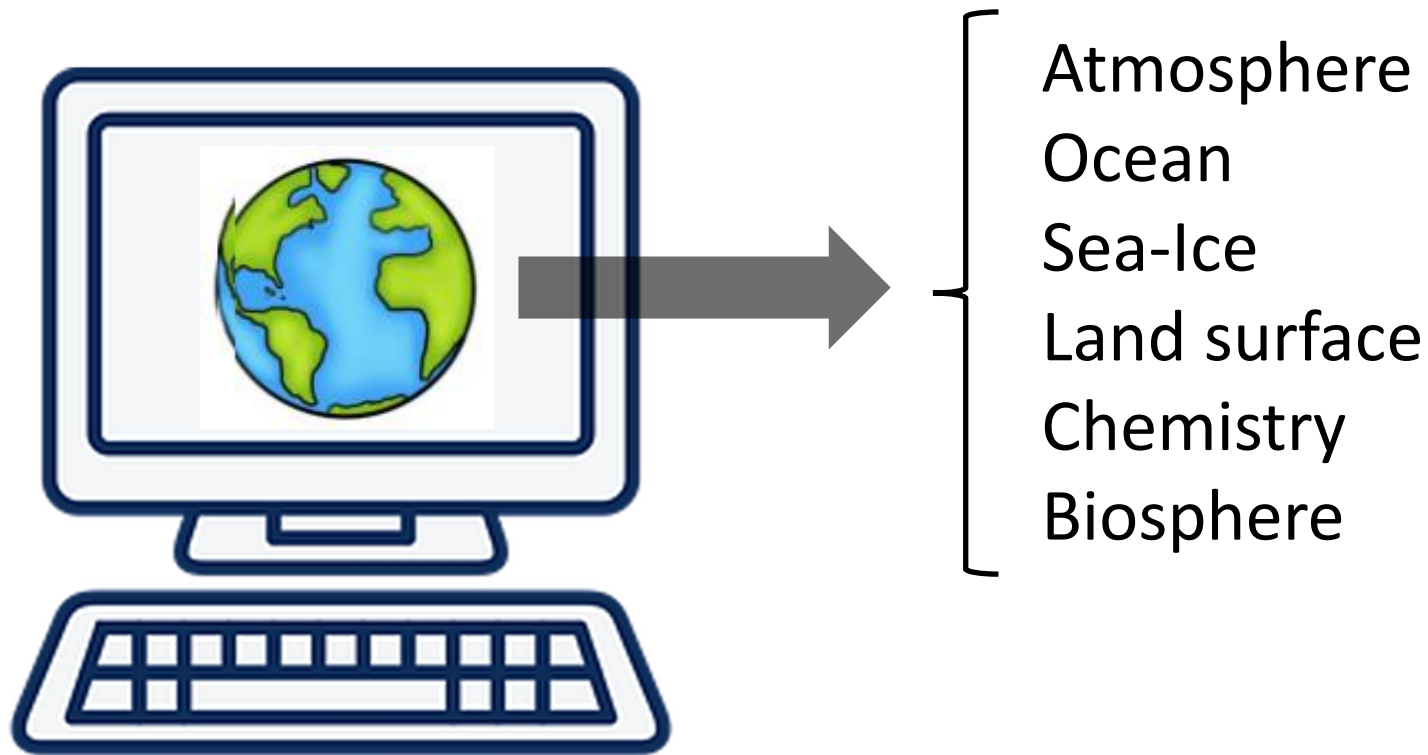


(2)



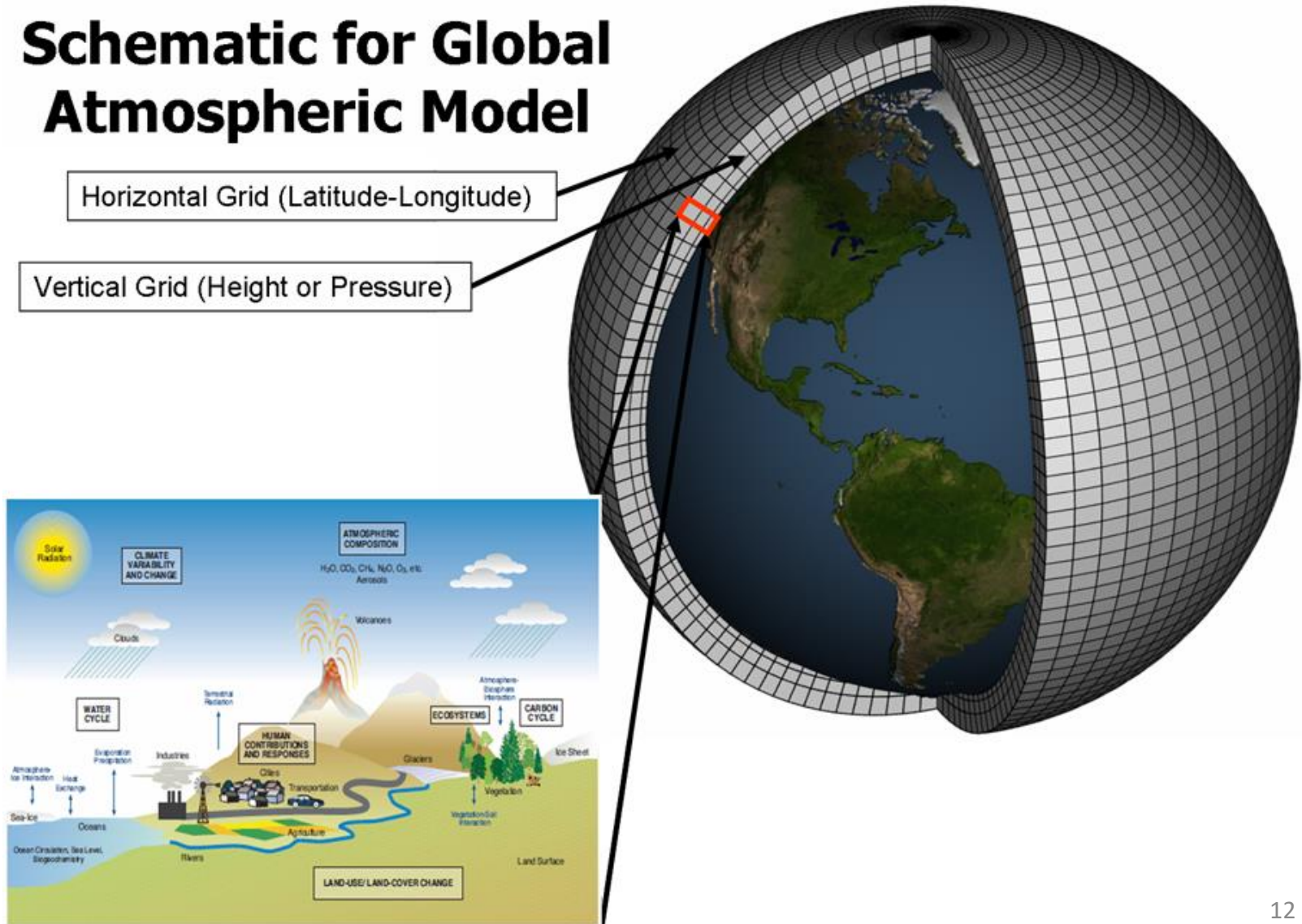
Seasonal forecast: Dynamical prediction

- Use GCM : Global Climate Model

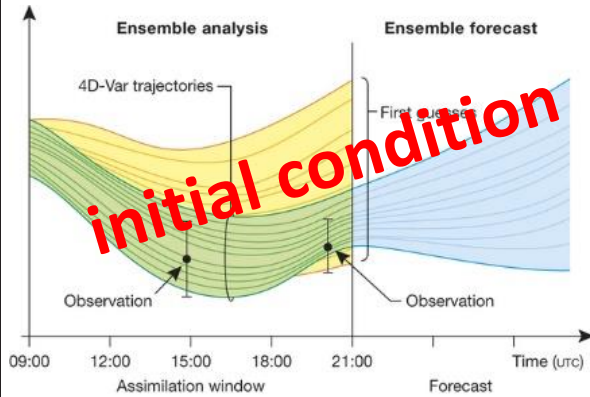


Seasonal forecast: Dynamical prediction

Schematic for Global Atmospheric Model



Seasonal forecast: Dynamical prediction



$$r : \rho \left(\frac{\partial u_r}{\partial t} + u_r \frac{\partial u_r}{\partial r} + \frac{u_\phi}{r \sin(\theta)} \frac{\partial u_r}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_r}{\partial \theta} - \frac{u_\phi^2 + u_\theta^2}{r} \right) = -\frac{\partial p}{\partial r} + \rho g_r +$$

$$\mu \left[\frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \frac{\partial u_r}{\partial r} \right) + \frac{1}{r^2 \sin(\theta)^2} \frac{\partial^2 u_r}{\partial \phi^2} + \frac{1}{r^2 \sin(\theta)} \frac{\partial}{\partial \theta} \left(\sin(\theta) \frac{\partial u_r}{\partial \theta} \right) - 2 \frac{u_r + \frac{\partial u_\theta}{\partial \theta} + u_\theta \cot(\theta)}{r^2} - \frac{2}{r^2 \sin(\theta)} \frac{\partial u_\phi}{\partial \phi} \right]$$

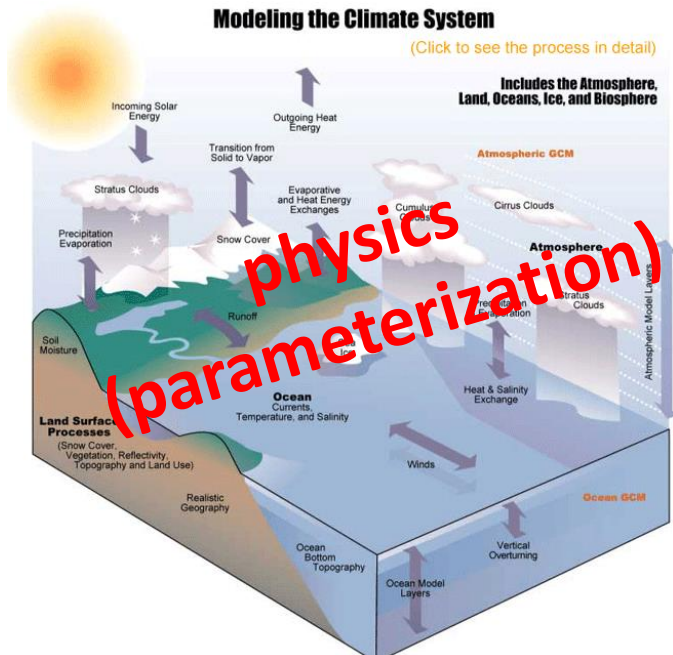
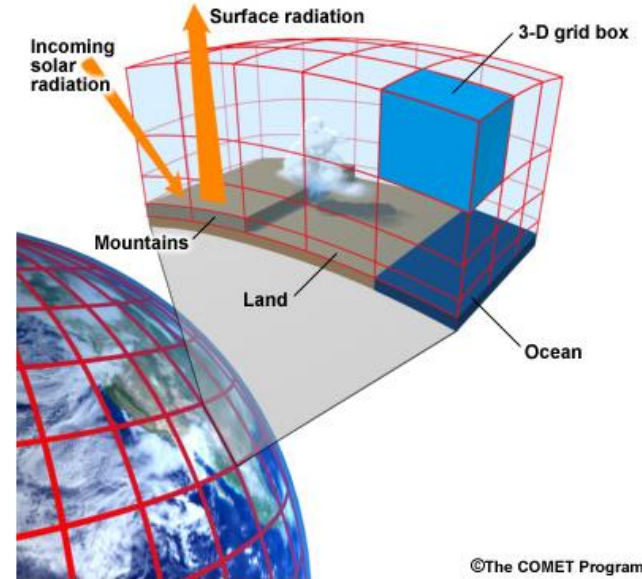
$$\phi : \rho \left(\frac{\partial u_\phi}{\partial t} + u_r \frac{\partial u_\phi}{\partial r} + \frac{u_\phi}{r \sin(\theta)} \frac{\partial u_\phi}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_\phi}{\partial \theta} + \frac{u_r u_\phi + u_\phi u_\theta \cot(\theta)}{r} - \frac{1}{r \sin(\theta)} \frac{\partial p}{\partial \phi} + \rho g_\phi +$$

$$\mu \left[\frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \frac{\partial u_\phi}{\partial r} \right) + \frac{1}{r^2 \sin(\theta)^2} \frac{\partial^2 u_\phi}{\partial \phi^2} + \frac{1}{r^2 \sin(\theta)} \frac{\partial}{\partial \theta} \left(\sin(\theta) \frac{\partial u_\phi}{\partial \theta} \right) + \frac{2 \sin(\theta) \frac{\partial u_r}{\partial \phi} + 2 \cos(\theta) \frac{\partial u_\theta}{\partial \phi} - u_\phi}{r} \right]$$

$$\theta : \rho \left(\frac{\partial u_\theta}{\partial t} + u_r \frac{\partial u_\theta}{\partial r} + \frac{u_\phi}{r \sin(\theta)} \frac{\partial u_\theta}{\partial \phi} + \frac{u_\theta}{r} \frac{\partial u_\theta}{\partial \theta} + \frac{u_r u_\theta - u_\phi^2 \cot(\theta)}{r} + \rho g_\theta +$$

$$\mu \left[\frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \frac{\partial u_\theta}{\partial r} \right) + \frac{1}{r^2 \sin(\theta)^2} \frac{\partial^2 u_\theta}{\partial \phi^2} + \frac{1}{r^2 \sin(\theta)} \frac{\partial}{\partial \theta} \left(\sin(\theta) \frac{\partial u_\theta}{\partial \theta} \right) + \frac{2 \cos(\theta) \frac{\partial u_r}{\partial \theta} - 2 u_\theta \cot(\theta) - u_\theta}{r} \right]$$

dynamics



Which approach does APCC take for Seasonal forecast? Multi-model ensemble (use many GCMs)



Seasonal forecast: Predictability

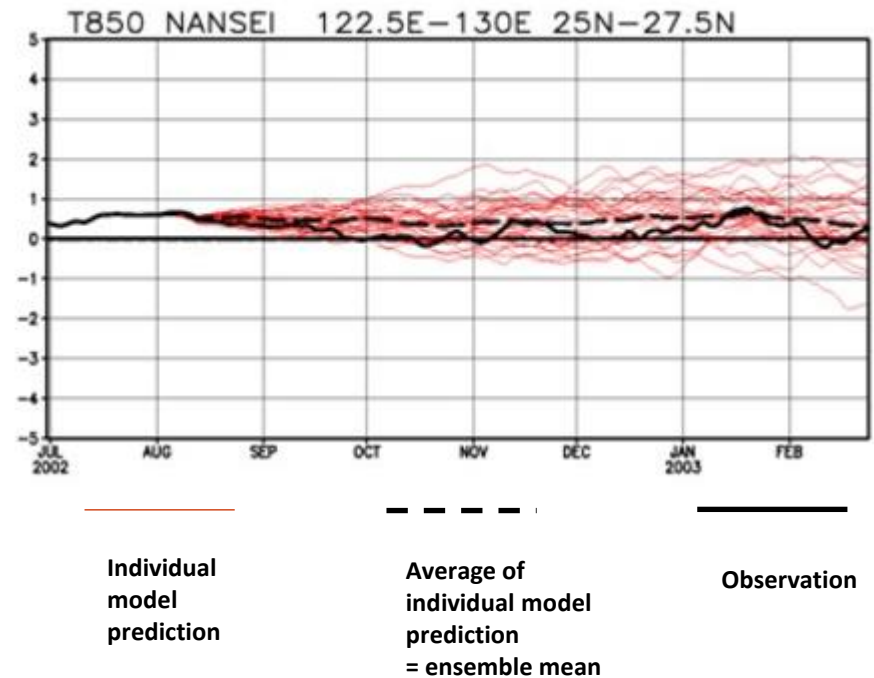
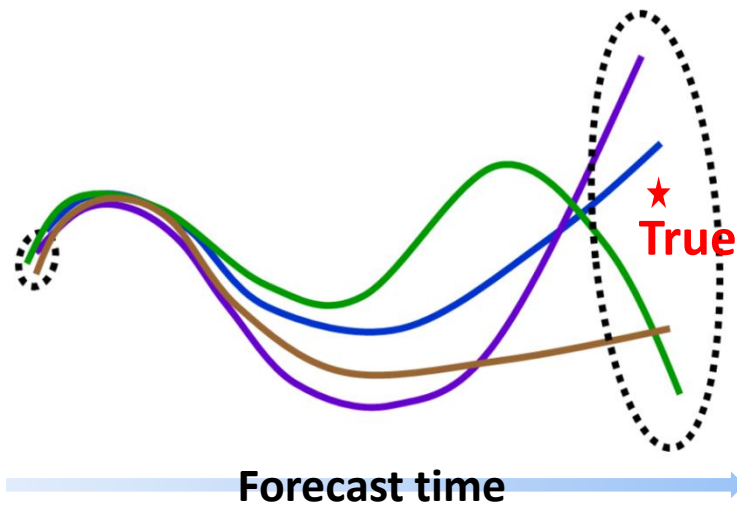
- Prediction: Action of forecast
- Prediction skill: Model (or human) capability of making forecast
- Predictability: Characteristics of Nature – How fast two states of the system would departure from each other in time

Lead time(τ)	<ol style="list-style-type: none">1. <i>Temperature of this room tomorrow</i>2. <i>Temperature of this room in 30days later</i>3. <i>Temperature of this room in 30years later</i>
Location	<ol style="list-style-type: none">1. <i>Temperature of Seoul (Korea)</i>2. <i>Temperature of Jakarta (Indonesia)</i>3. <i>Temperature of Villa Las Estrellas (Antarctica)</i>
Physical variables	<ol style="list-style-type: none">1. <i>Temperature</i>2. <i>rainfall</i>3. <i>wind speed</i>

Trying to reach potential predictability with state-of-art prediction system

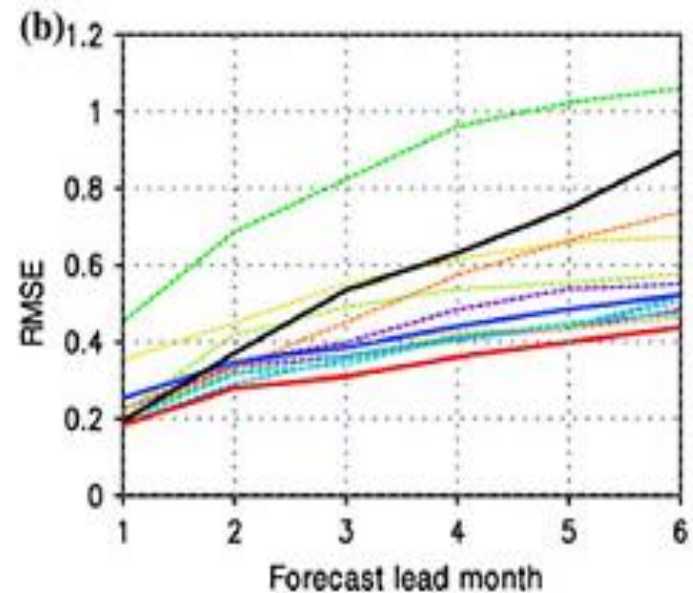
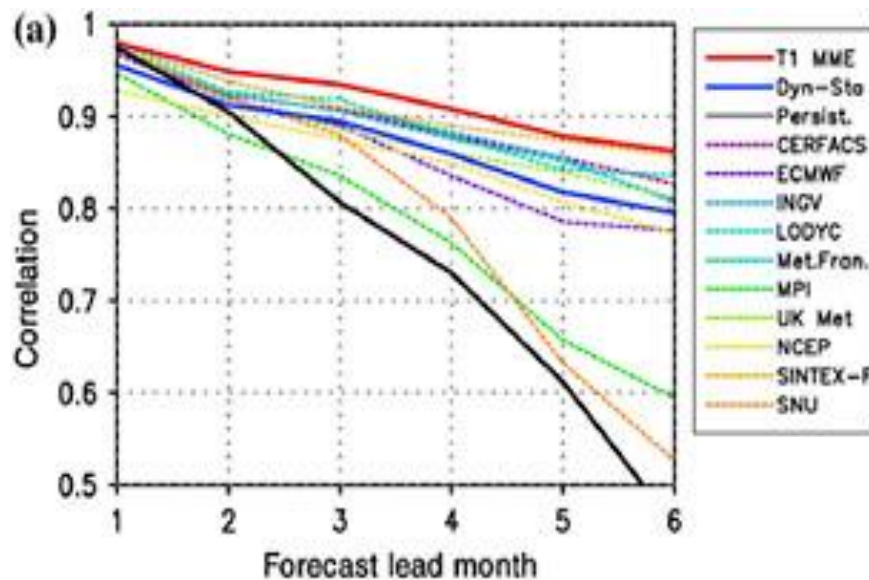
Seasonal forecast: Multi-model ensemble

- Averaging across a number of models
- To reduce noise by averaging large ensemble members
(The Earth's atmosphere is chaotic, and GCM is not perfect)



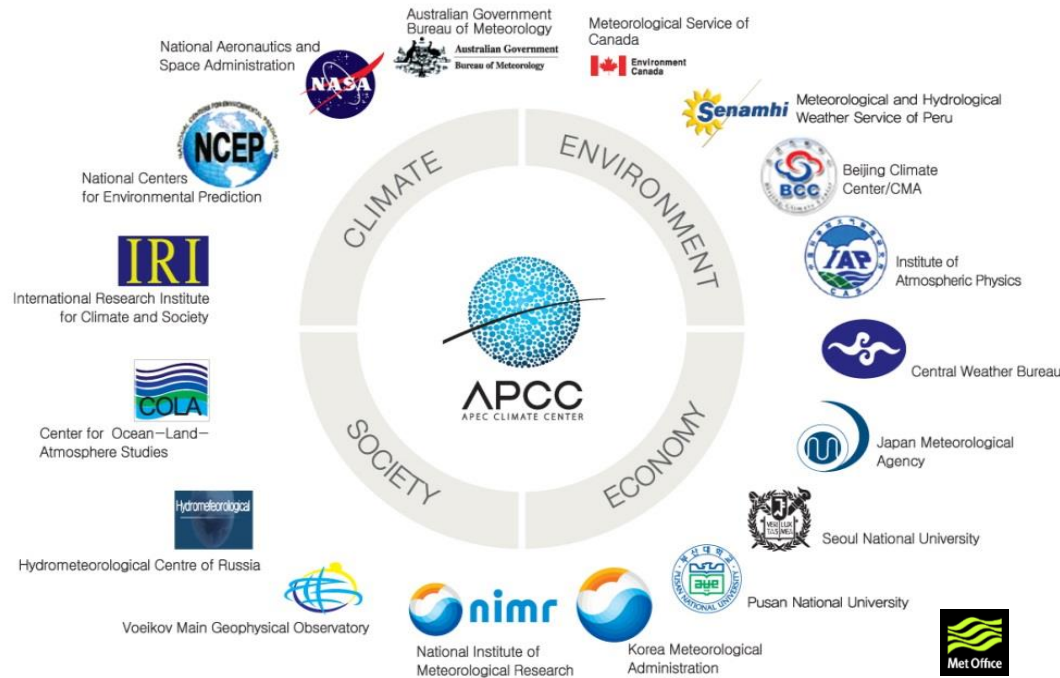
Seasonal forecast: Multi-model ensemble

An example of a MME outperforming individual models in forecasting
(Jin et al. (2008))



Seasonal forecast: Multi-model ensemble

- Collect Global climate forecast data from 17 institutes and disseminate MME forecast



Graphical products

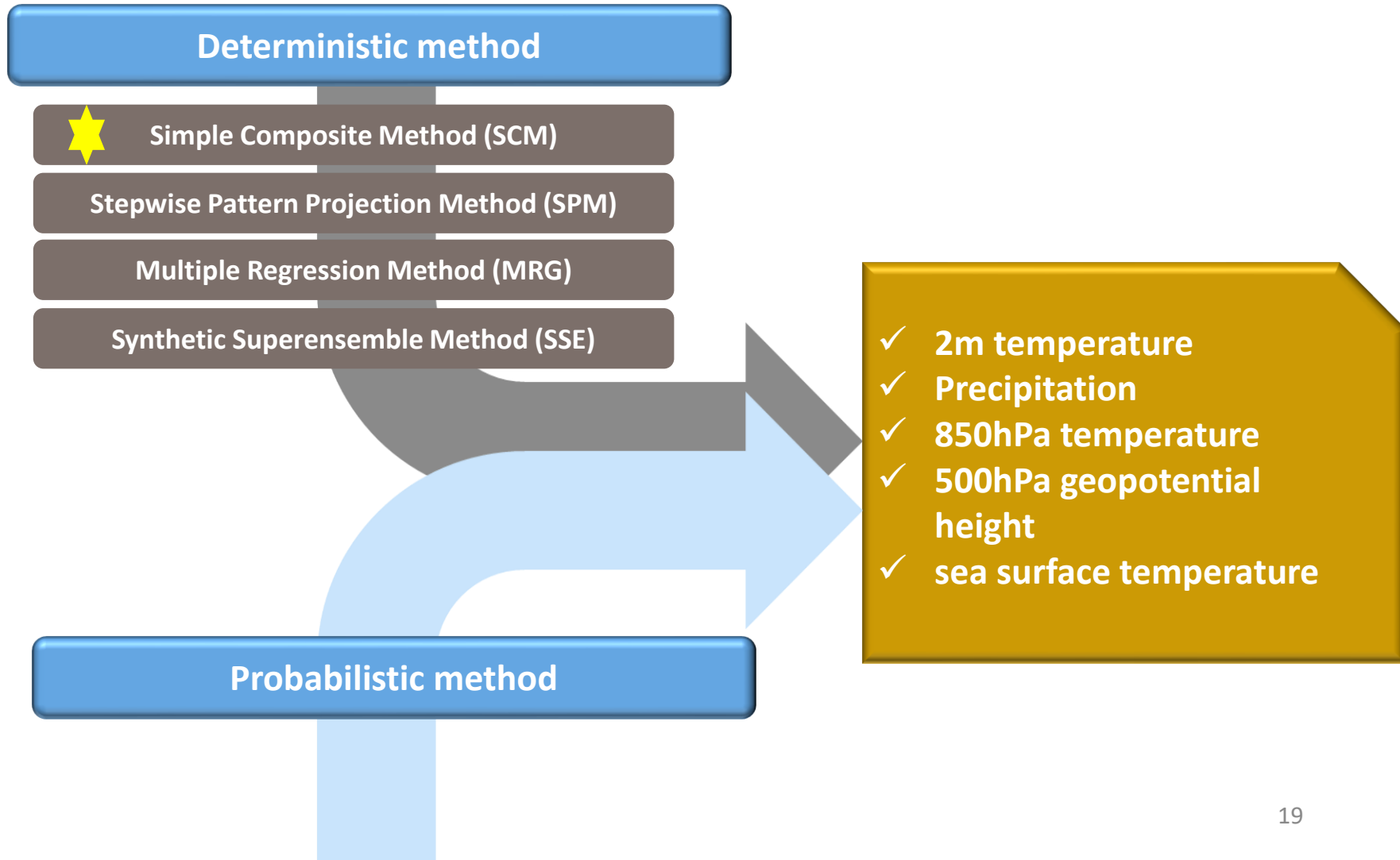
APCC website: apcc21.org

Digital products

APCC Data Service System website (ADSS): adss.apcc21.org

Seasonal forecast: Multi-model ensemble

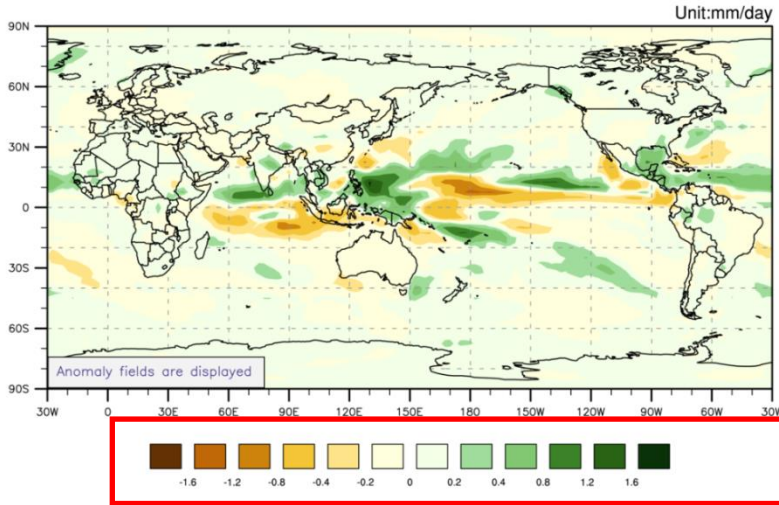
- Methodology of the APCC MME Prediction System (Min et al. 2014)



Seasonal forecast: Multi-model ensemble

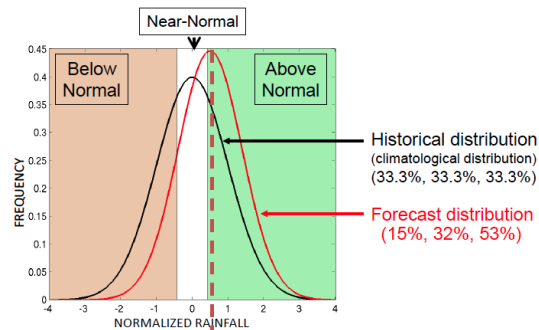
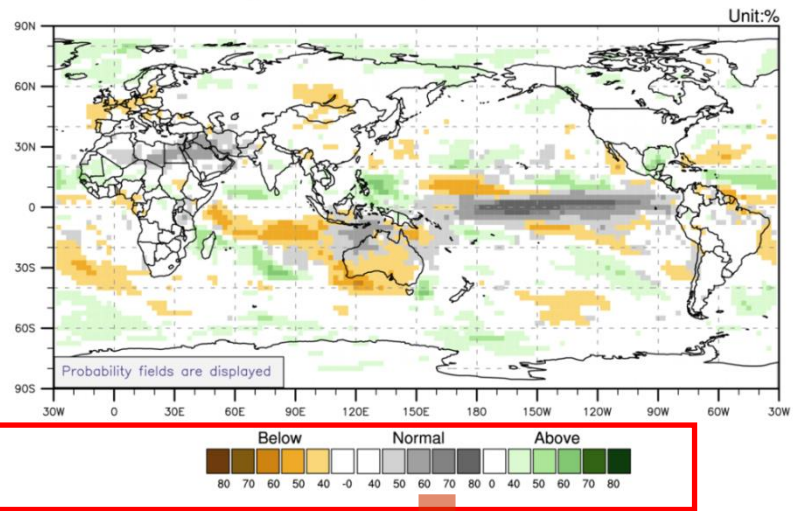
Deterministic

Precipitation for August-October 2017



Probabilistic

Precipitation for August-October 2017



Map showing probability of rainfall falling in one of three categories (with respect to climatology) : BN vs NN vs AN

How good are they?

What makes a good forecast?

1. Consistency
2. Quality
3. Value

(Murphy AH 1993; Wea. Forecasting 8, 281)



■ Verification for Deterministic MME method

Anomaly Correlation Coefficient	Root Mean Square Error
---------------------------------	------------------------

■ Verification for Probabilistic MME method

Reliability Diagram	Relative Operating Characteristics Curve	Heidke Skill Score	Ranked Probability Skill Score
---------------------	--	--------------------	--------------------------------

How good are they?

■ Verification for Deterministic MME method

Anomaly Correlation Coefficient	Root Mean Square Error
---------------------------------	------------------------

$$ACC = \frac{1}{n} \sum_{i=1}^n \frac{(x_i - \bar{x})}{\sigma_x} \frac{(y_i - \bar{y})}{\sigma_y}$$

-1 ≤ ACC ≤ 1, ACC = 1 best score

$$RMSE = \sqrt{\frac{1}{N} \sum (y_i - o_i)^2}$$

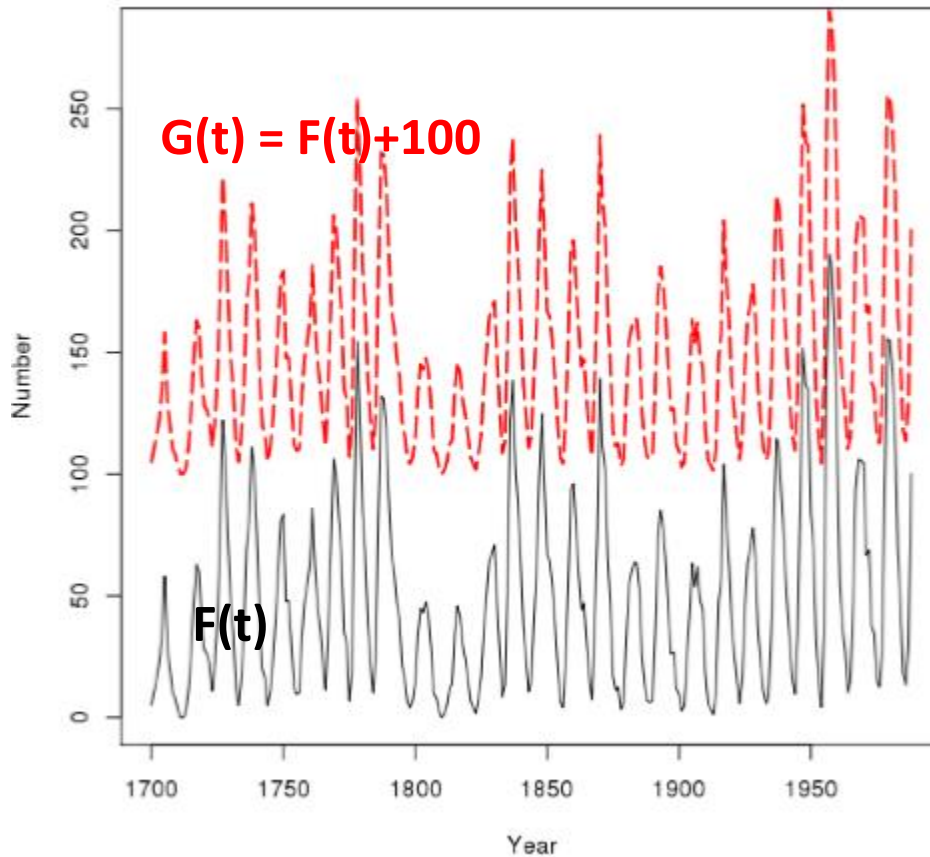
The smaller, the better

How good are they?

Verification for Deterministic MME method

Anomaly Correlation Coefficient	Root Mean Square Error
---------------------------------	------------------------

Sunspot Number



Correlation coefficient?
RMSE?

How good are they?

■ Verification for Probabilistic MME method

Reliability Diagram	Relative Operating Characteristics Curve	Heidke Skill Score	Ranked Probability Skill Score
---------------------	--	--------------------	--------------------------------

		Observed		Total
		Yes	No	
Forecast	Yes	$a = \text{hits}$	$b = \text{false alarms}$	$a + b = \text{forecast yes}$
	No	$c = \text{misses}$	$d = \text{correct negatives}$	$c + d = \text{forecast no}$
Total		$a + c = \text{observed yes}$	$b + d = \text{observed no}$	$a + b + c + d = \text{total}$

$$\text{HSS} = \frac{\text{FC} - \text{FC}_{\text{random}}}{\text{FC}_{\text{perfect}} - \text{FC}_{\text{random}}}$$

FC: Fraction correct = fraction of hits (a/N) + fraction of correct negatives (d/N)

$\text{FC}_{\text{random}}$: random forecast

$$\text{FC}_{\text{random}} = \left(\frac{a+b}{N} \right) \left(\frac{a+c}{N} \right) + \left(\frac{c+d}{N} \right) \left(\frac{b+d}{N} \right)$$

Yes_f
Yes_o
No_f
No_o

Fraction of hits **Fraction of correct negatives**

HSS = 1 perfect forecast

HSS = 0 forecast as good as random

HSS = worse than random

$\text{FC}_{\text{perfect}} = 1$

How good are they?

■ Verification for Probabilistic MME method

Reliability Diagram	Relative Operating Characteristics Curve	Heidke Skill Score	Ranked Probability Skill Score
---------------------	--	--------------------	--------------------------------

- Ranked probability score (RPS): squared probability error

$$RPS = \frac{1}{ncat - 1} \sum_{icat=1}^{ncat} (Pcumfct_{icat} - Pcumobs_{icat})^2$$

icat: category number (1.below normal, 2.near normal, 3.above normal)

ncat: number of categories

Ex) 0.20/0.35/0.45 (fcst:below/near/above) vs (0.0/0.0/1.0)

$$RPS = 1/(3-1) \times \{(0.20-0.0)^2 + (0.55-0.0)^2 + (1.00-1.00)^2\} = 0.17125$$

- Ranked probability skill score (RPSS)

$$RPSS = 1 - \frac{RPS_{fct}}{RPS_{cli}}$$

RPSS > 0 : Better than climatology

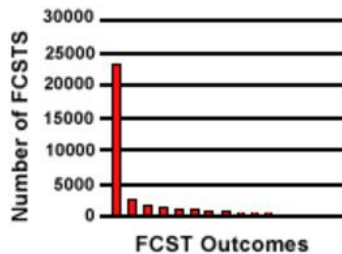
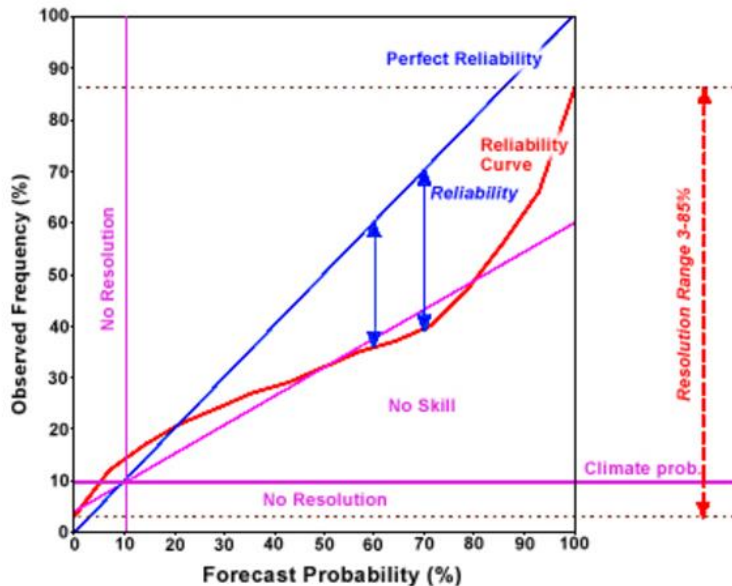
RPSS = 1 : skill equal to climatology

RPSS < 0: Worse than climatology

How good are they?

Verification for Probabilistic MME method

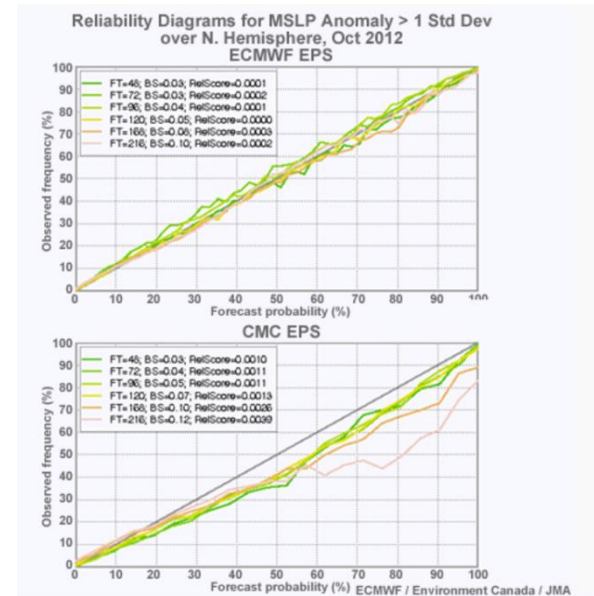
Reliability Diagram	Relative Operating Characteristics Curve	Heidke Skill Score	Ranked Probability Skill Score
---------------------	--	--------------------	--------------------------------



Reliability diagram for 3-day lead time ensembles for January 1996. Forecast probabilities are based on observed frequencies associated with the same number of ensemble members falling in a particular bin during 1-20 December 1995.

NCEP

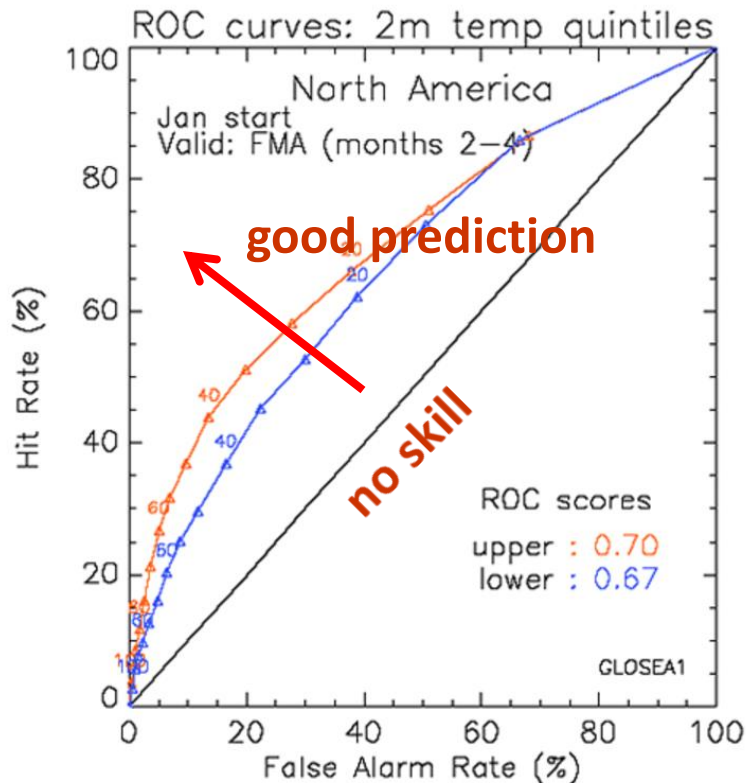
- Forecast probability versus the observed frequency
- For a perfectly reliable ensemble, the two should match, yielding a straight line from the lower-left corner to upper-right corner



How good are they?

Verification for Probabilistic MME method

Reliability Diagram	Relative Operating Characteristics Curve	Heidke Skill Score	Ranked Probability Skill Score
---------------------	--	--------------------	--------------------------------



- Information on the hit rates and false alarm rates

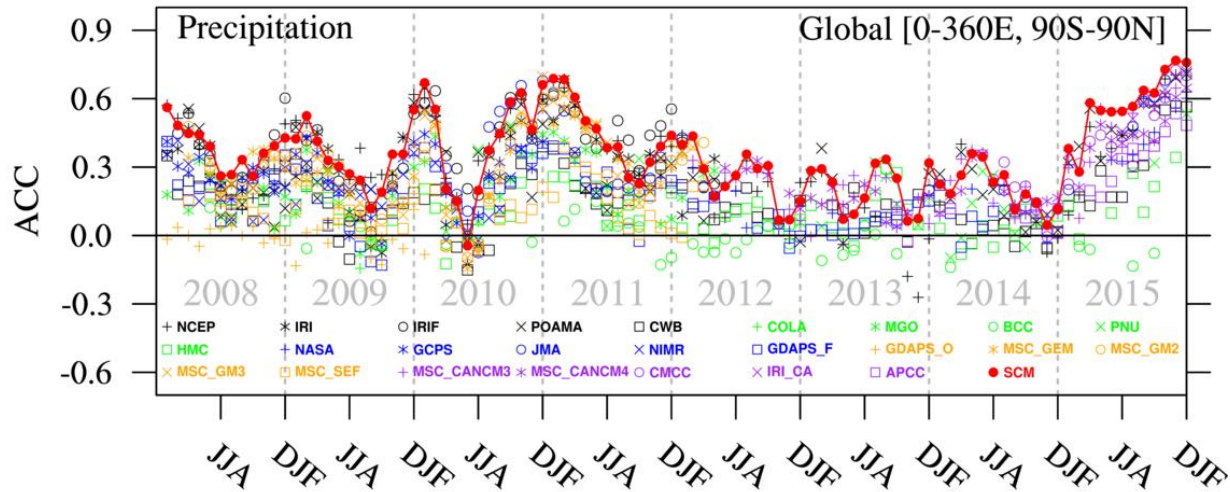
hit rate: the number of times the event was forecast (with probability above a certain threshold) and later observed to occur

false alarm rate: the number of times the event was forecast but did not occur

		Observed		Total
		Yes	No	
Forecast	Yes	$a = \text{hits}$	$b = \text{false alarms}$	$a + b = \text{forecast yes}$
	No	$c = \text{misses}$	$d = \text{correct negatives}$	$c + d = \text{forecast no}$
Total		$a + c = \text{observed yes}$	$b + d = \text{observed no}$	$a + b + c + d = \text{total}$

How good are they?

- APCC operational forecast



	APCC	WMOLC	ECMWF	NCEP	UKMO	JMA
AN	0.569457	0.541897	0.535531	0.52996	0.528975	0.531497
NN	0.520962	0.521424	0.537661	0.519823	0.524022	0.514656
BN	0.567702	0.533777	0.516511	0.535767	0.516994	0.534244


Realtime rainfall forecast for last 4 years (12-15)
 ROC score : Perfect = 1, Meaningless(no skill) =0.5,

How good are they?

Clim Dyn
DOI 10.1007/s00382-017-3576-2



Skill of real-time operational forecasts with the APCC multi-model ensemble prediction system during the period 2008–2015

Young-Mi Min¹  · Vladimir N. Kryjov¹ · Sang Myeong Oh¹ · Hyun-Ju Lee¹

- The skill of the APCC forecasts strongly depends on seasons and regions that it is higher for the tropics and boreal winter than for the extratropics and boreal summer
- forecast skill for precipitation is more seasonally and regionally dependent than that for temperature
- The skill of both temperature and precipitation forecasts strongly depends upon the ENSO strength. (the highest forecast skill noted in 2015/2016 boreal winter is associated with the strong forcing of an extreme El Nino event)

What else does APCC provide?

CLIMATE INFORMATION SERVICES



ADSS APCC Data Service System

[LOGIN](#) | [REGISTER](#) | [CONTACT US](#)



The APEC Climate Center Data Service System was developed for real-time climate monitoring and provision of digital data service to APEC member economies. This system underscores the role of APCC in playing an important role as a hub of climate data and services in the region. The main objective of the ADSS is to provide a comprehensive set of models and observational climate data to various researchers and users to establish a scientific basis for climate prediction. ADSS also aims to monitor climate information using near real-time in-situ observation and prediction data in a standardized and accessible format for various users.

What else does APCC provide?

DataSet	Areal Coverage	Grid Size	Time Step	Access	Source	Requirements
APCC-MME(6-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
APCC-MME(3-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
INDIVIDUAL-MODEL(6-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
INDIVIDUAL-MODEL(3-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
CORDEX-SEA25	Regional	25km	Daily		APCC	
CORDEX-SEA50	Regional	50km	Daily		APCC	
Clipped CMIP5	National level (22 Countries)	Depending on GCMs	Daily		ESGF	
IRI_DATA_LIBRARY	Global	2.5×2.5(degree)	Various		IRI	
NCEP	Global	2.5×2.5(degree)	Daily		NOAA	
NCEP-SFC	Global	2.5×2.5(degree)	Daily		NOAA	
NOAA-OLR	Global	2.5×2.5(degree)	Daily		NOAA	
TMI	Global	2.5×2.5(degree)	Daily		REMSS	
QUICKSCAT	Global	0.25×0.25(degree)	Daily		REMSS	
GPCP	Global	1.0×1.0(degree)	Daily		NASA	
GHCN	Global	5.0×5.0(degree)	Monthly		NOAA	
UD	Global	0.5×0.5(degree)	Monthly		University of Delaware	

Seasonal forecast data

Climate change scenario data

Observed data

What else does APCC provide?

Climate change scenario data

- **Coupled Model Intercomparison Project (CMIP) Phase 5**

20 climate modeling groups from around the world involved to

- 1) Assess the mechanisms responsible for model differences in poorly understood feedbacks associated with the carbon cycle and with clouds
- 2) Examine climate “predictability” and exploring the ability of models to predict climate on decadal time scales
- 3) Determine why similarly forced models produce a range of responses

Realistic scenarios of climate forcing for both historical, paleoclimate and future scenarios

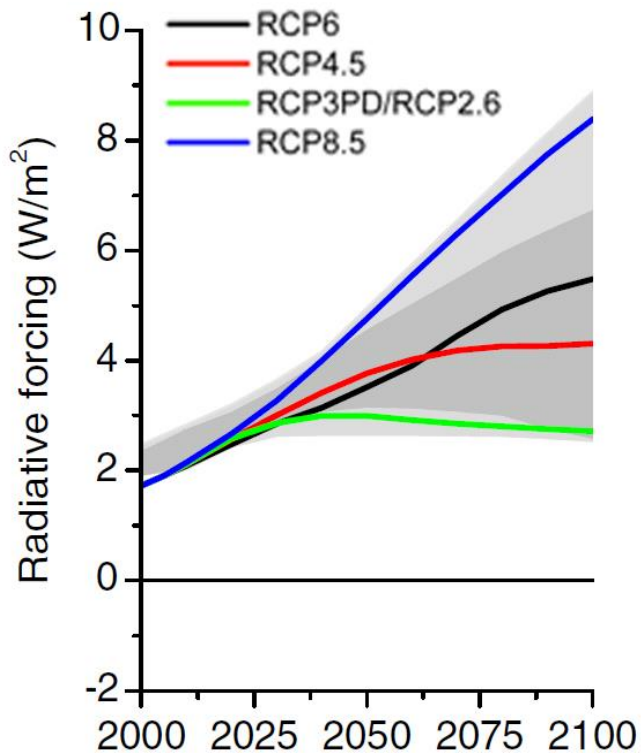
Provide simulations for assessment in the Assessment Report (AR)5 of IPCC (Intergovernmental Panel on Climate Change)

What else does APCC provide?

Climate change scenario data

Coupled Model Intercomparison Project (CMIP) Phase 5

Representative Concentration Pathways (RCPs) scenarios



Source: IPCC

RCPs	Description
RCP8.5	Rising radiative forcing pathway leading to 8.5 W/m ² in 2100.
RCP6.0	Stabilization without overshoot pathway to 6 W/m ² at stabilization after 2100
RCP4.5	Stabilization without overshoot pathway to 4.5 W/m ² at stabilization after 2100
RCP2.6	Peak in radiative forcing at ~ 3 W/m ² before 2100 and decline

Total radiative forcing

cumulative measure of human emissions of GHGs from all sources expressed in Watts per square meter

What else does APCC provide?

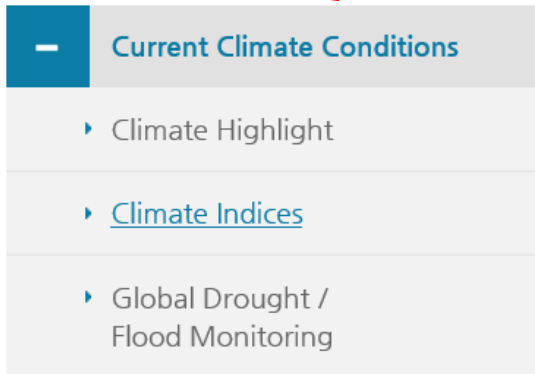
DataSet	Areal Coverage	Grid Size	Time Step	Access	Source	Requirements
APCC-MME(6-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
APCC-MME(3-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
INDIVIDUAL-MODEL(6-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
INDIVIDUAL-MODEL(3-MON)	Global	2.5×2.5(degree)	Monthly		APCC	Login
CORDEX-SEA25	Regional	25km	Daily		APCC	
CORDEX-SEA50	Regional	50km	Daily		APCC	
Clipped CMIP5	National level (22 Countries)	Depending on GCMs	Daily		ESGF	

— National level data based on clipped CMIP5 data (29 GCMs)

Scenarios	Countries	Variables
Historical	Bangladesh, Burma, Chile, Cuba, Egypt, Ethiopia, Federated States of Micronesia, India, Indonesia, Kenya, Malaysia, Marshall Islands, Mongolia, Nepal, Philippines, Pakistan, Samoa, Tanzania, Thailand, Tonga, Vietnam, Zambia	Precipitation, max/min temperature, wind speed, relative humidity, solar radiation
RCP4.5		
RCP8.5		

What else?

CLIMATE INFORMATION SERVICES



Pacific SST	Atlantic SST	Atmosphere	Monsoon	Index Forecast
-------------	--------------	------------	---------	----------------

Slowly (w.r.t. atmosphere) varying ocean, land surface, sea ice conditions may provide some predictable signals...

Assume you want to predict $X = X_{\text{signal}} + X_{\text{noise}}$

Signal \gg Noise : more predictable

Signal \ll Noise : less predictable

Monsoon

- Most of the world's population live in monsoon regions
- While the global monsoon system responds to net heating on planetary scales, the evolution of the regional monsoons depends on the distribution of land and ocean as well as SST gradients and topography.

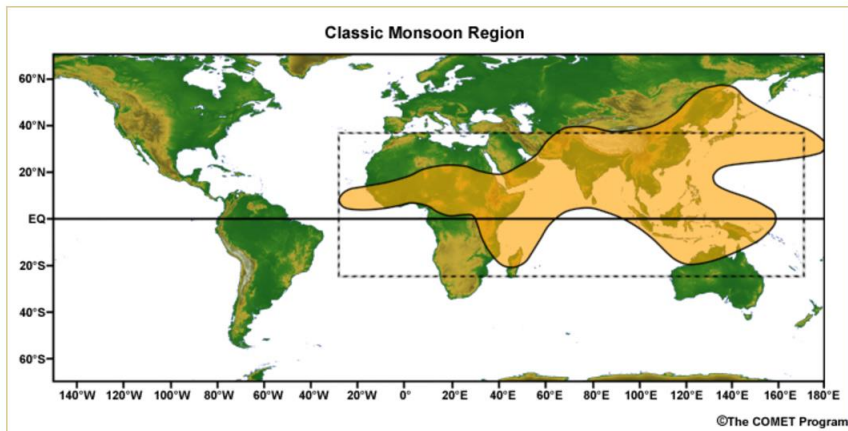
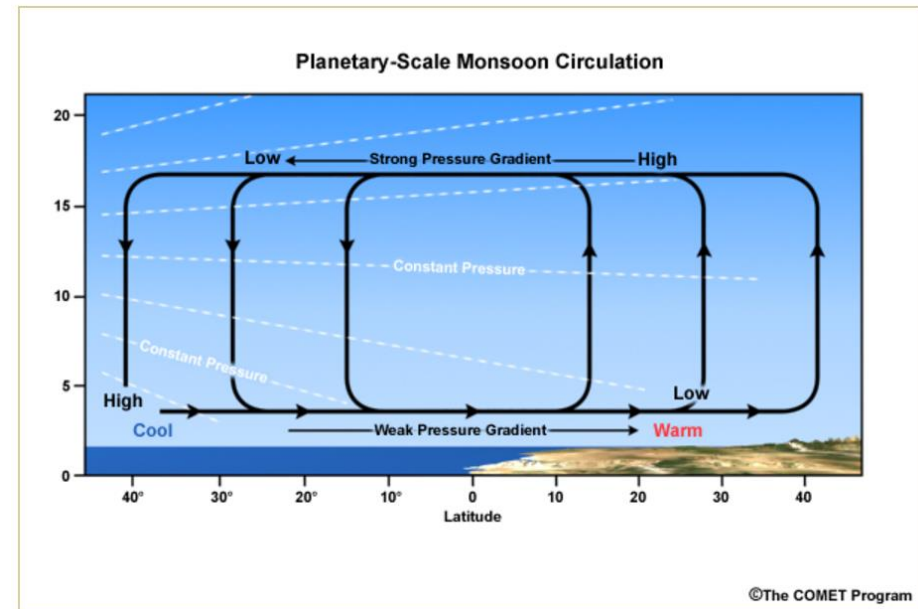
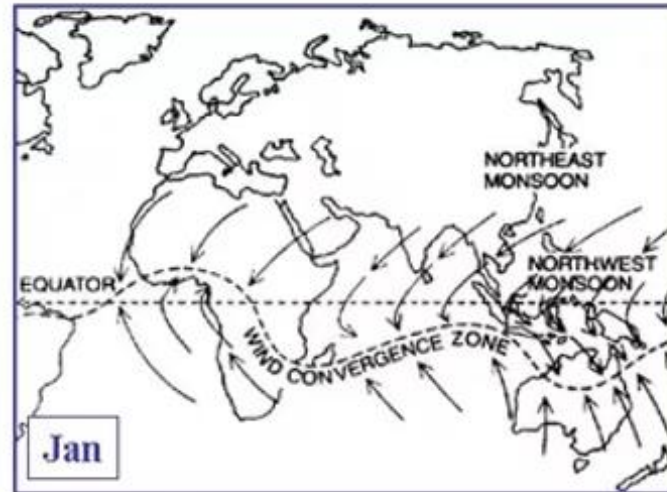


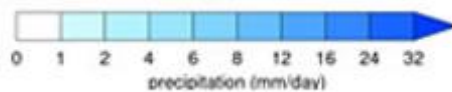
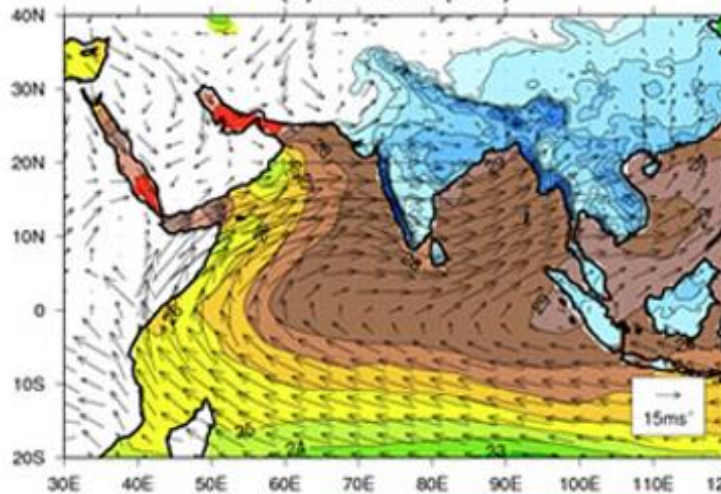
Fig. 3.29. The monsoon regions as defined by Ramage (1971).



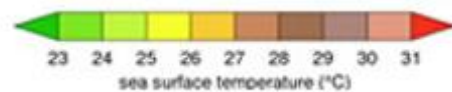
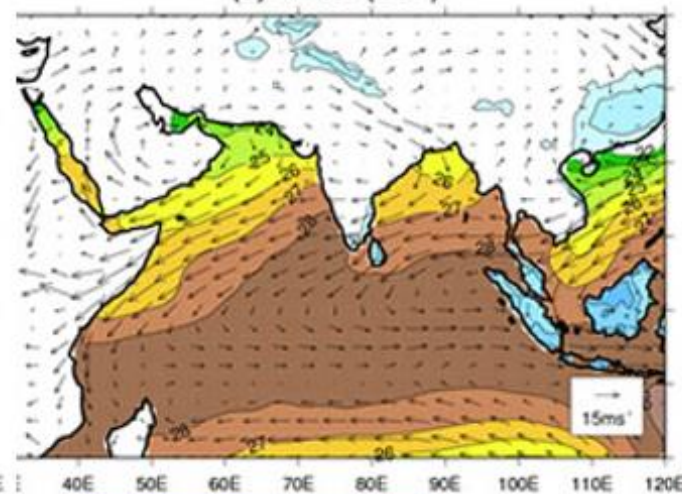
Monsoon



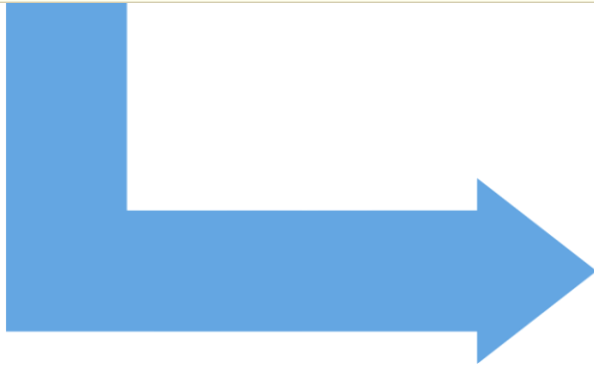
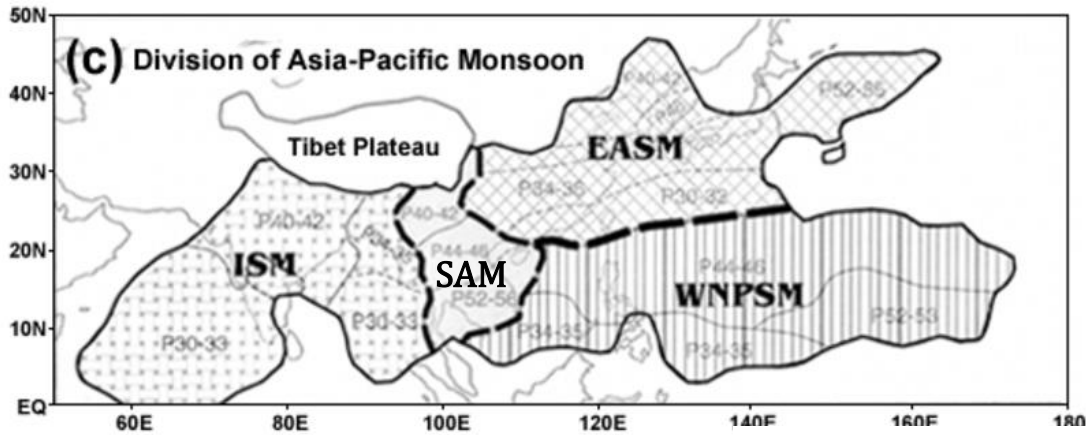
(b) summer (JJA)



(a) winter (DJF)

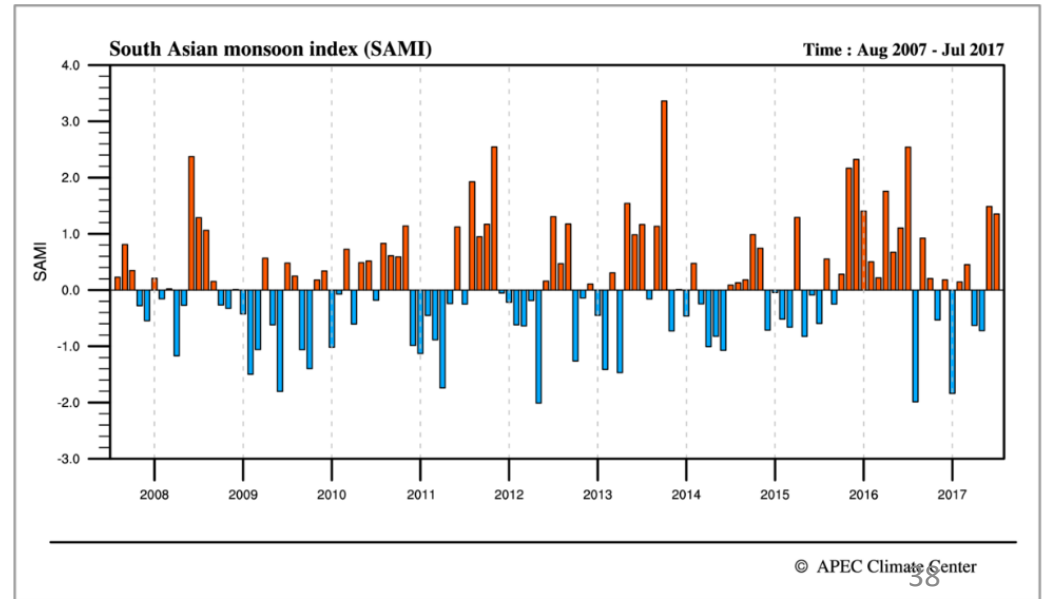


Monsoon



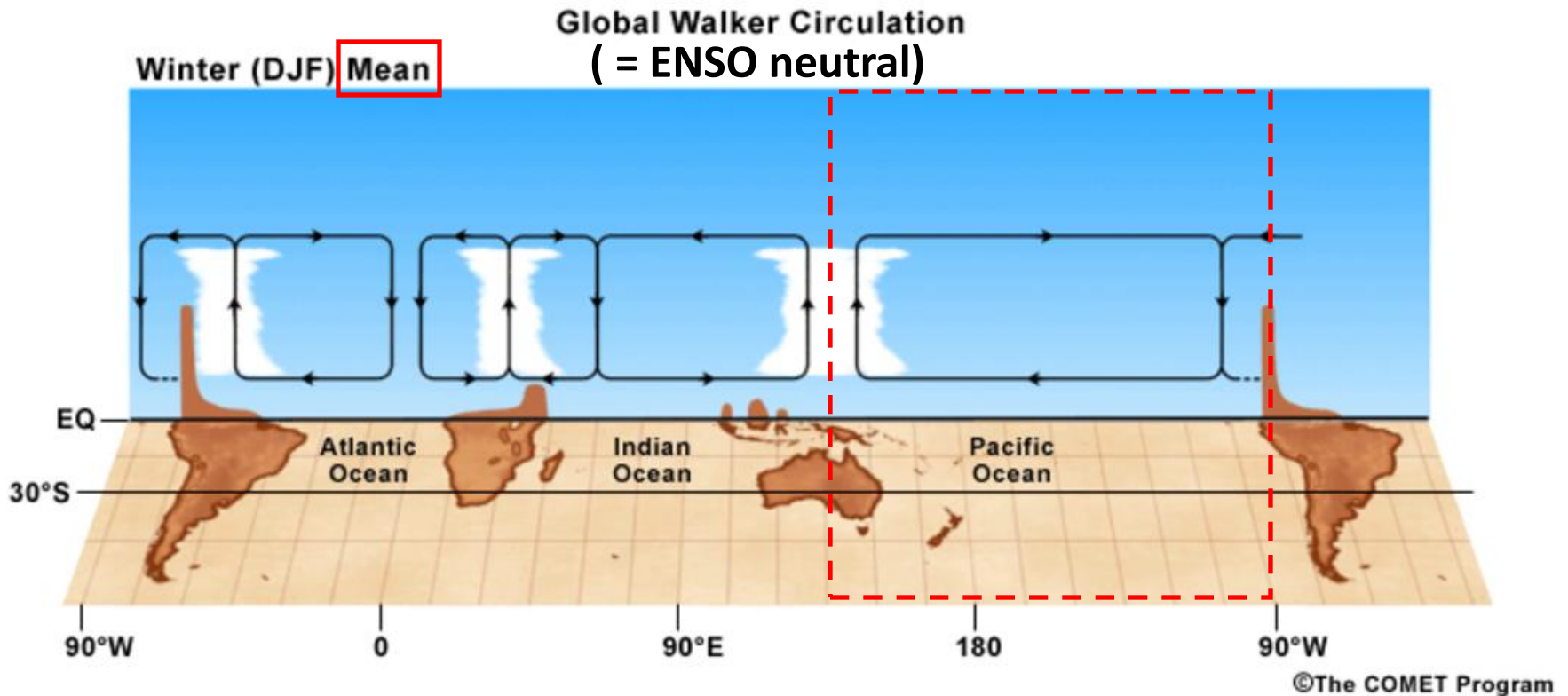
Monsoon indices Monitoring

WYI	AUSMI	SAMI	IMI	WNPMI
-----	-------	-------------	-----	-------



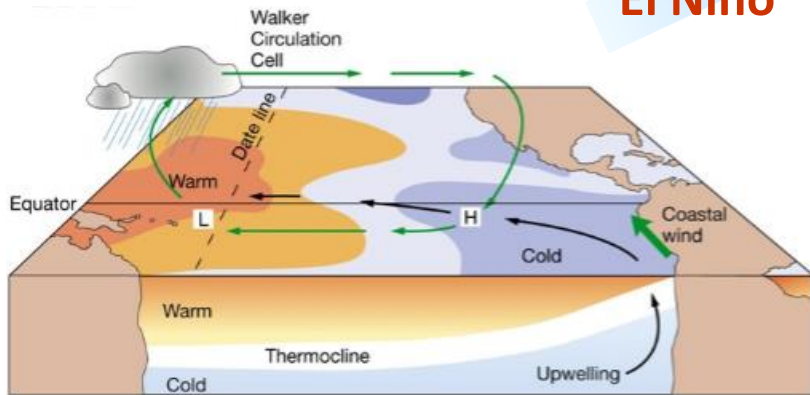
ENSO

- El Nino Southern Oscillation: the biggest signal on the globe
- Strong couplings between the ocean and atmosphere
- Interannual times scale occurs in the tropical Pacific
- Has three states: El Nino vs La Nina vs Neutral
- Perturbing Walker circulation



ENSO

El Nino

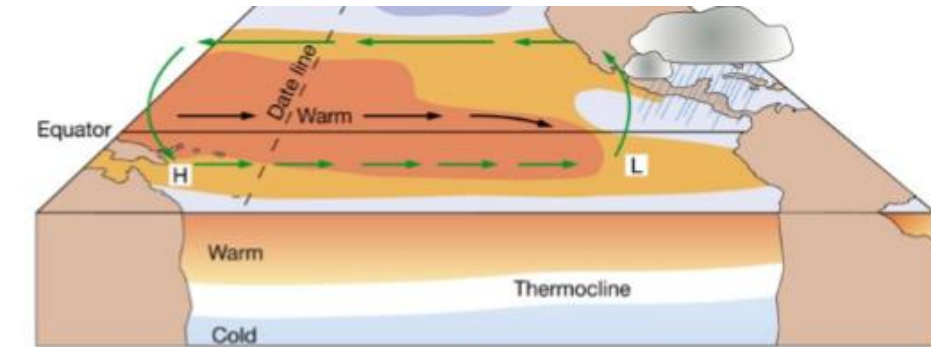


(a) Normal conditions

Copyright © 2005 Pearson Prentice Hall, Inc.

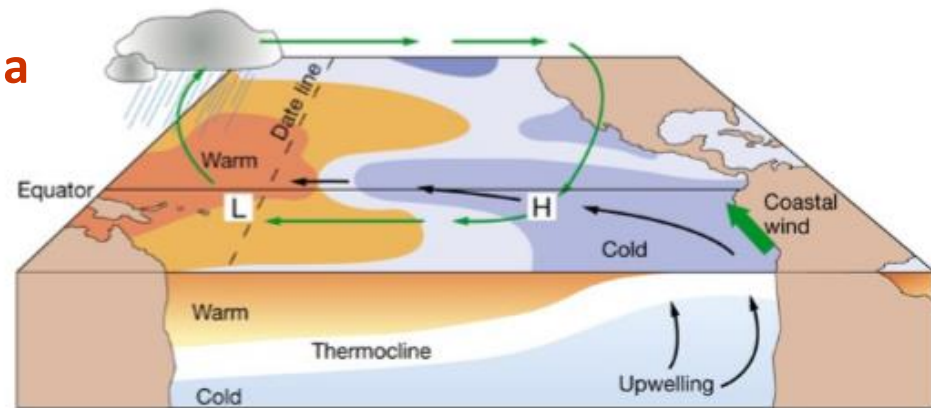
Normal

La Nina



(b) El Niño conditions

Copyright © 2005 Pearson Prentice Hall, Inc.



(c) La Niña conditions

Copyright © 2005 Pearson Prentice Hall, Inc.



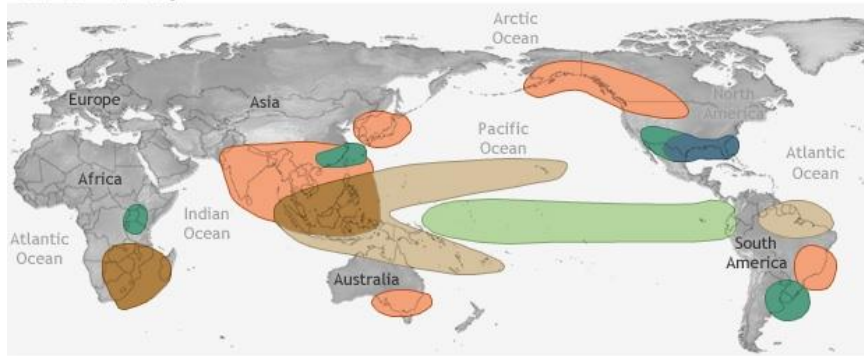
Easterlies: wind blowing from the east
 * Think about New Yorker, Parisian

ENSO

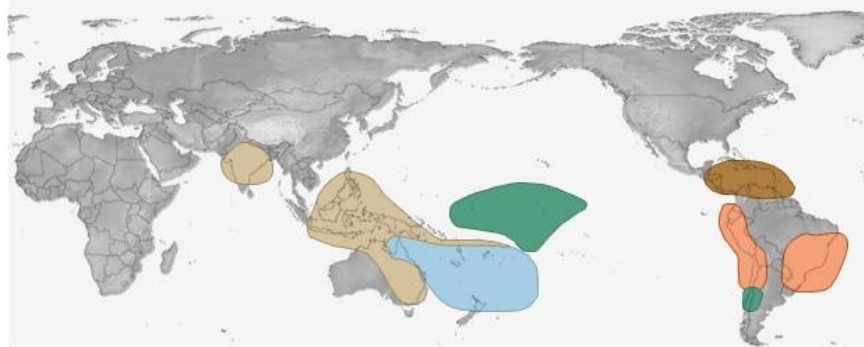
- Why do we care ENSO?

EL NIÑO CLIMATE IMPACTS

December-February



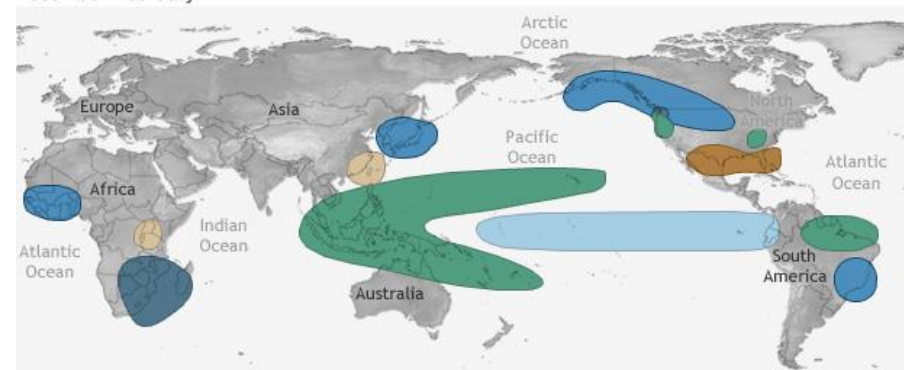
June-August



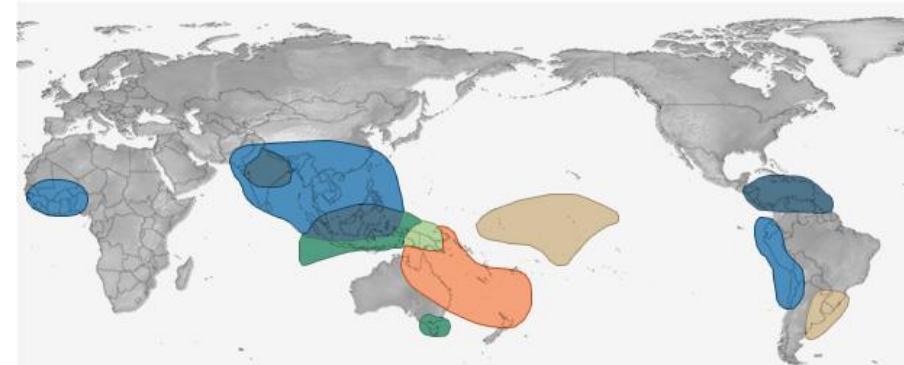
NOAA Climate.gov

LA NIÑA CLIMATE IMPACTS

December-February



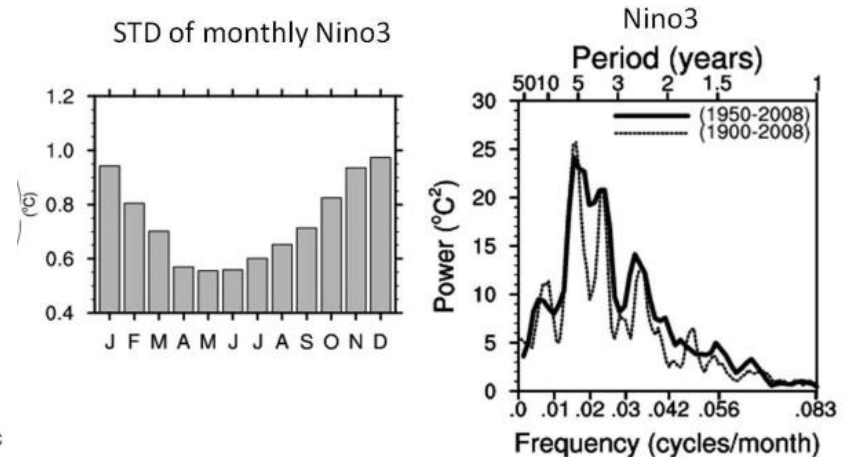
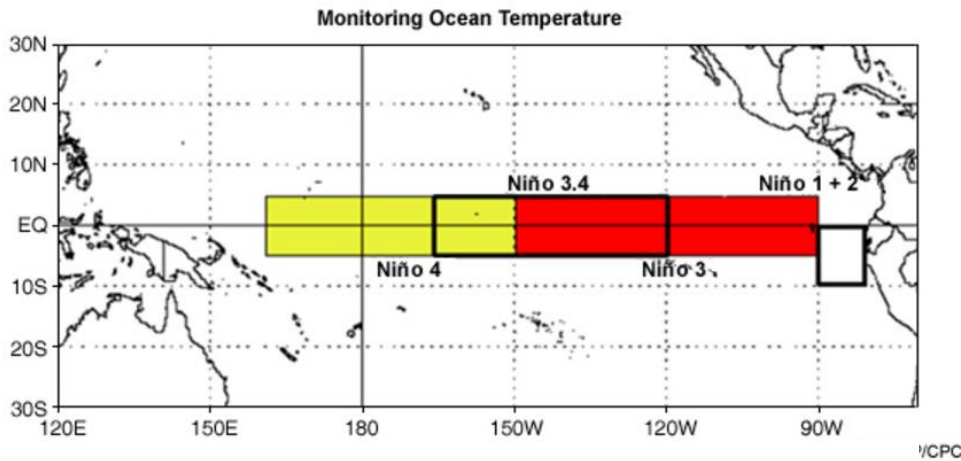
June-August



NOAA Climate.gov

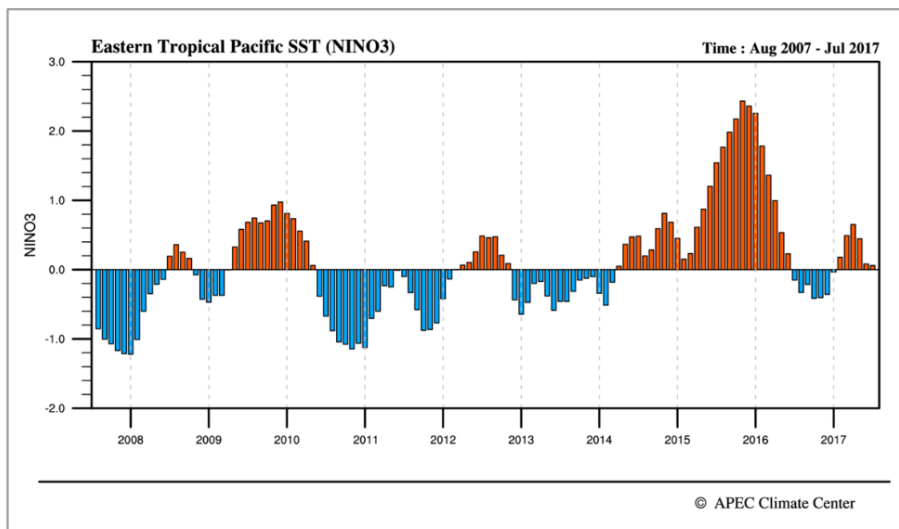
ENSO

- ENSO monitoring



Pacific SST indices Monitoring

NINO1+2	NINO3	NINO3.4	NINO4	ONI	TNI	PACWARM	EOFPAC	EMI
---------	--------------	---------	-------	-----	-----	---------	--------	-----

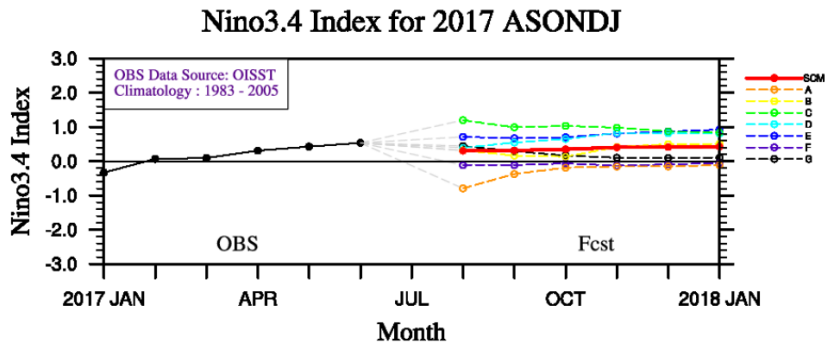


El Niño and La Niña events tend to develop during the period Apr-Jun and they

- Tend to reach their maximum strength during October - February
- Typically persist for 9-12 months, though occasionally persisting for up to 2 years
- Typically recur every 2 to 7 years

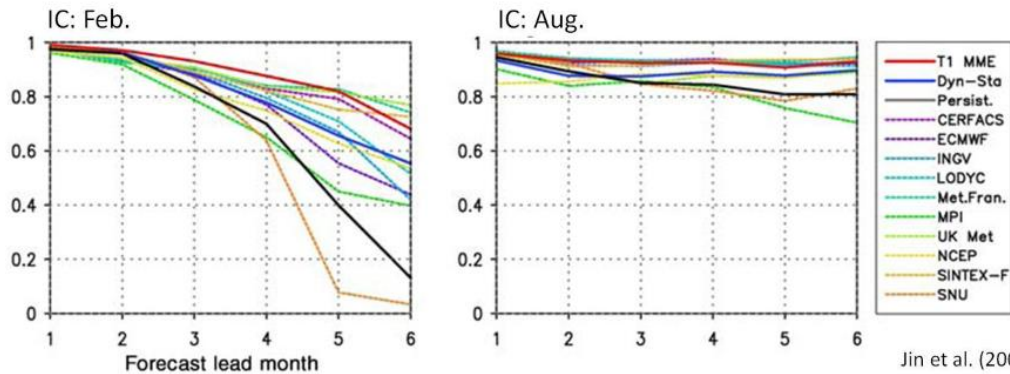
ENSO

- ENSO prediction



© APEC Climate Center

- The skill of both temperature and precipitation forecasts strongly depends upon the ENSO strength (Min et al., 2017)



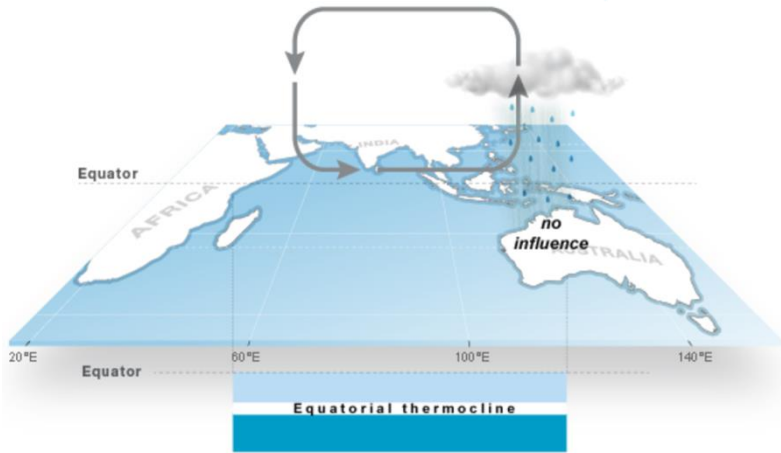
Jin et al. (2008)

- ENSO can be predicted even 6 months ahead
- ENSO forecast is difficult from Feb-Apr, called the “spring predictability barrier”

IOD

- Indian Ocean Dipole

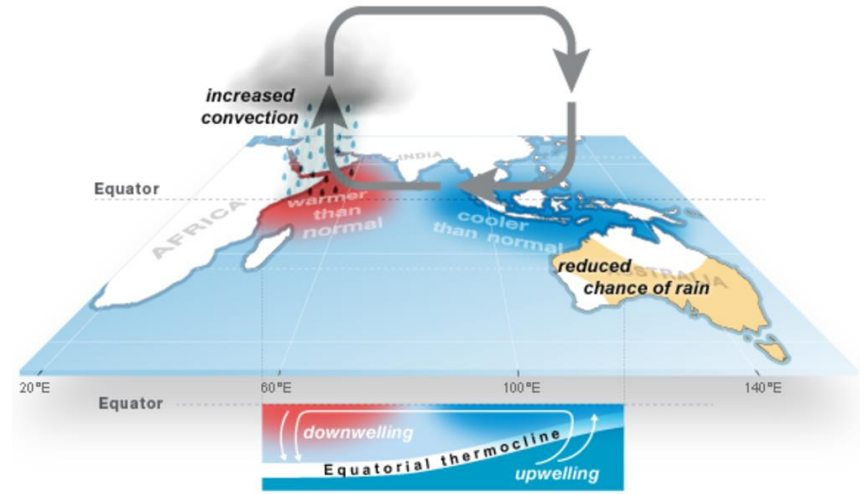
Positive



Indian Ocean Dipole (IOD): Neutral phase

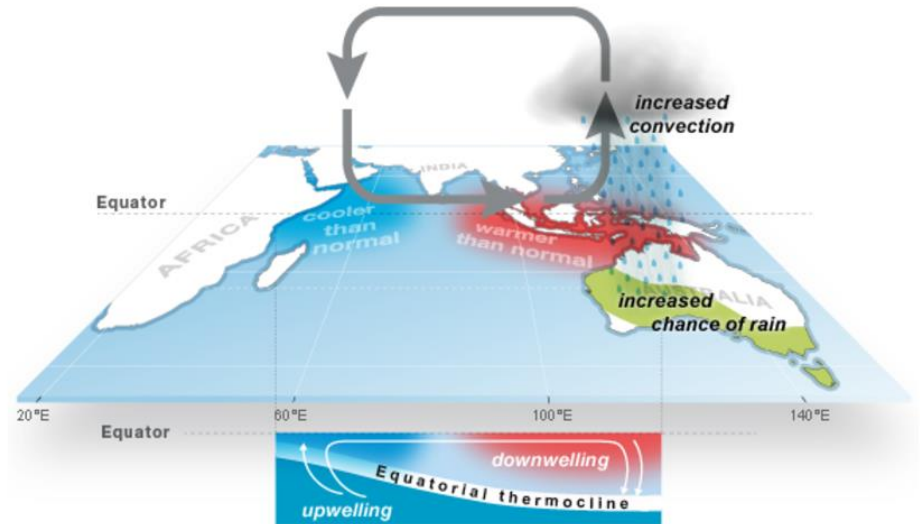
© Commonwealth of Australia 2013.

Negative



Indian Ocean Dipole (IOD): Positive phase

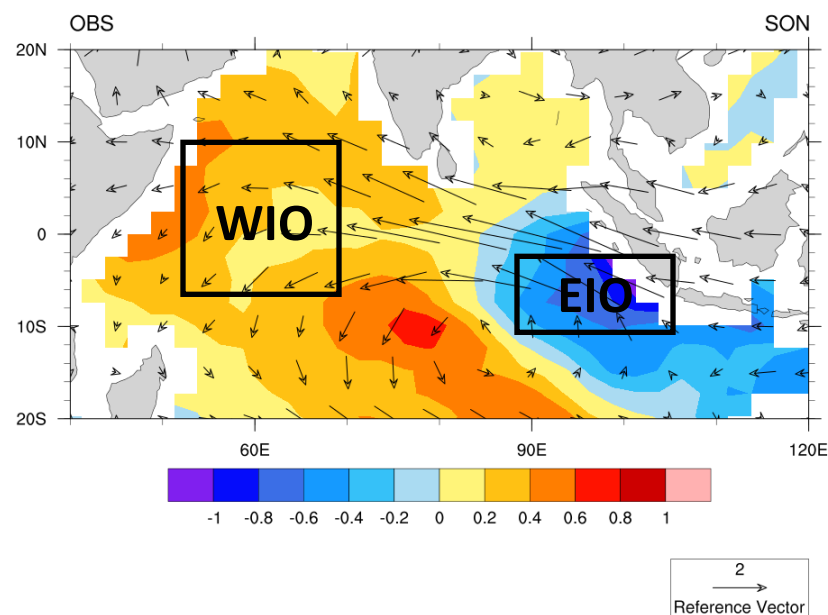
© Commonwealth of Australia 2013.



Indian Ocean Dipole (IOD): Negative phase

IOD

- IOD develops rapidly in boreal summer and reaches its mature phase in October
- IOD tends to have a biennial tendency, that is, the zonal wind and SST gradient anomalies change the sign from one year to the following year (i.e., A negative IOD usually follows a positive IOD)
- Co-evolution of El Nino (La Nina) with positive (negative) IOD
- Dipole Mode Index
= $SST_{WIO} - SST_{EIO}$



IOD

- Dipole Mode Index = SST_{WIO} - SST_{EIO}

Indian & Atlantic SST indices Monitoring

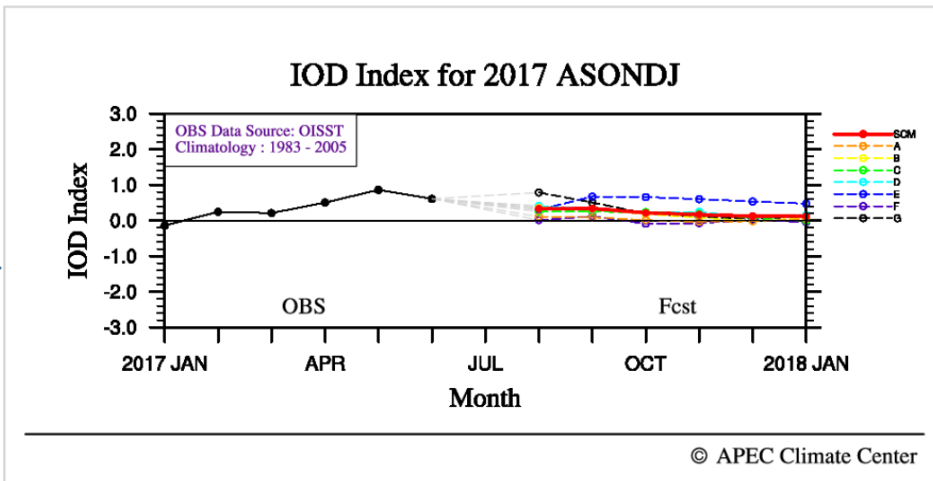
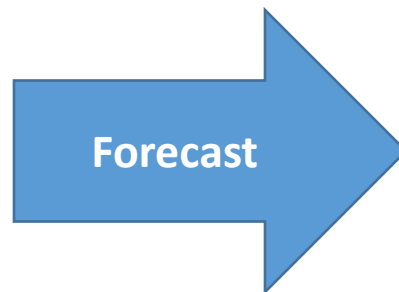
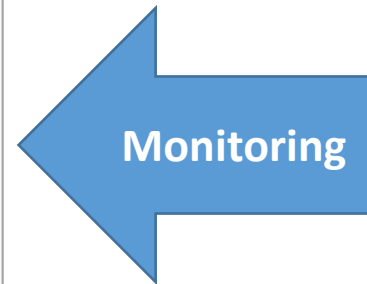
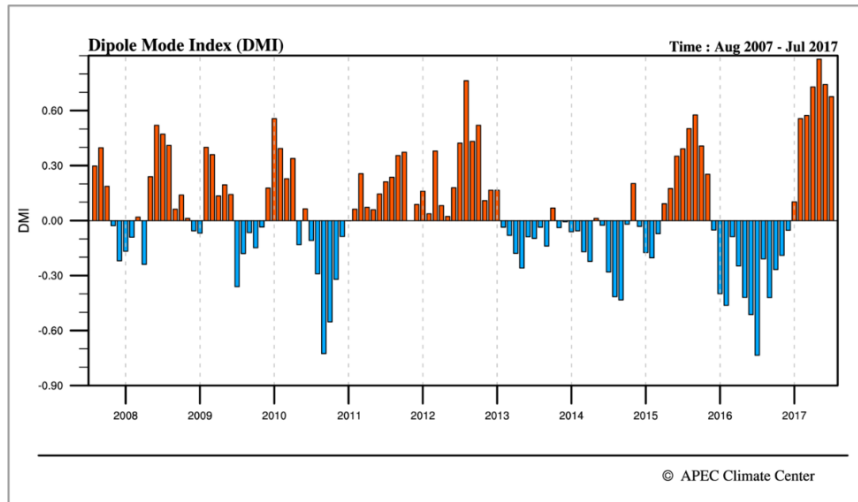


Fig. 2. Predicted Indian Ocean Dipole mode index (IODMI) from individual models (A, B, C, D, E, F and G) and the SCM.

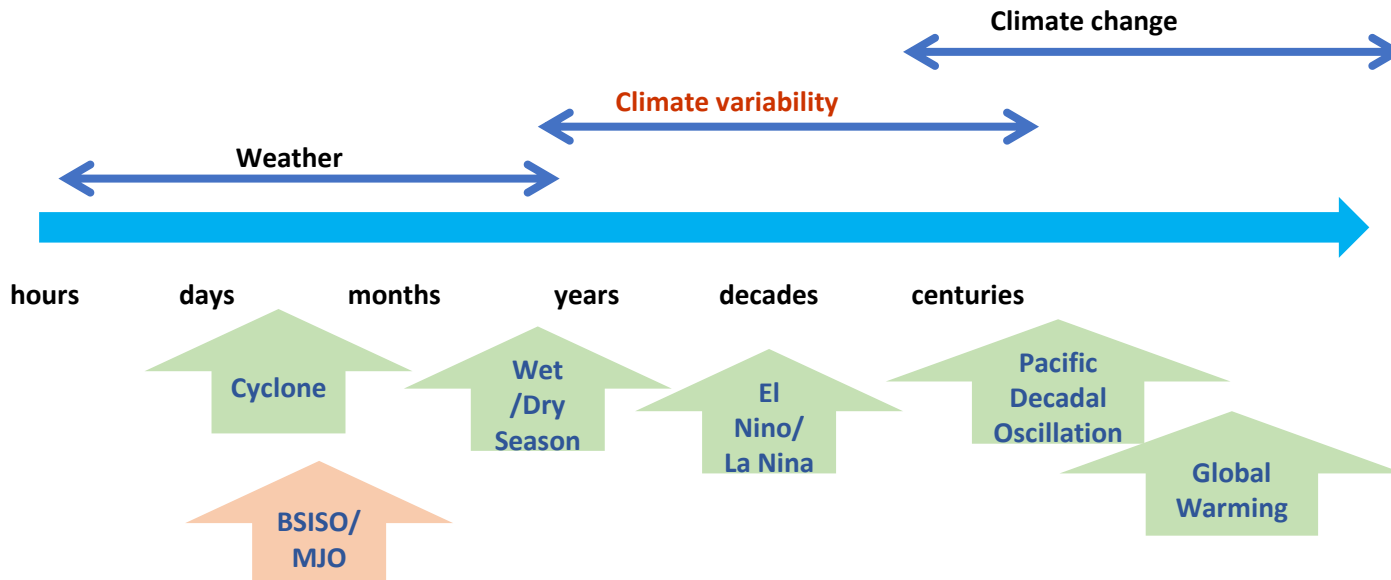
Last but not least

- Don't forget our BSISO forecast!

CLIMATE INFORMATION SERVICES

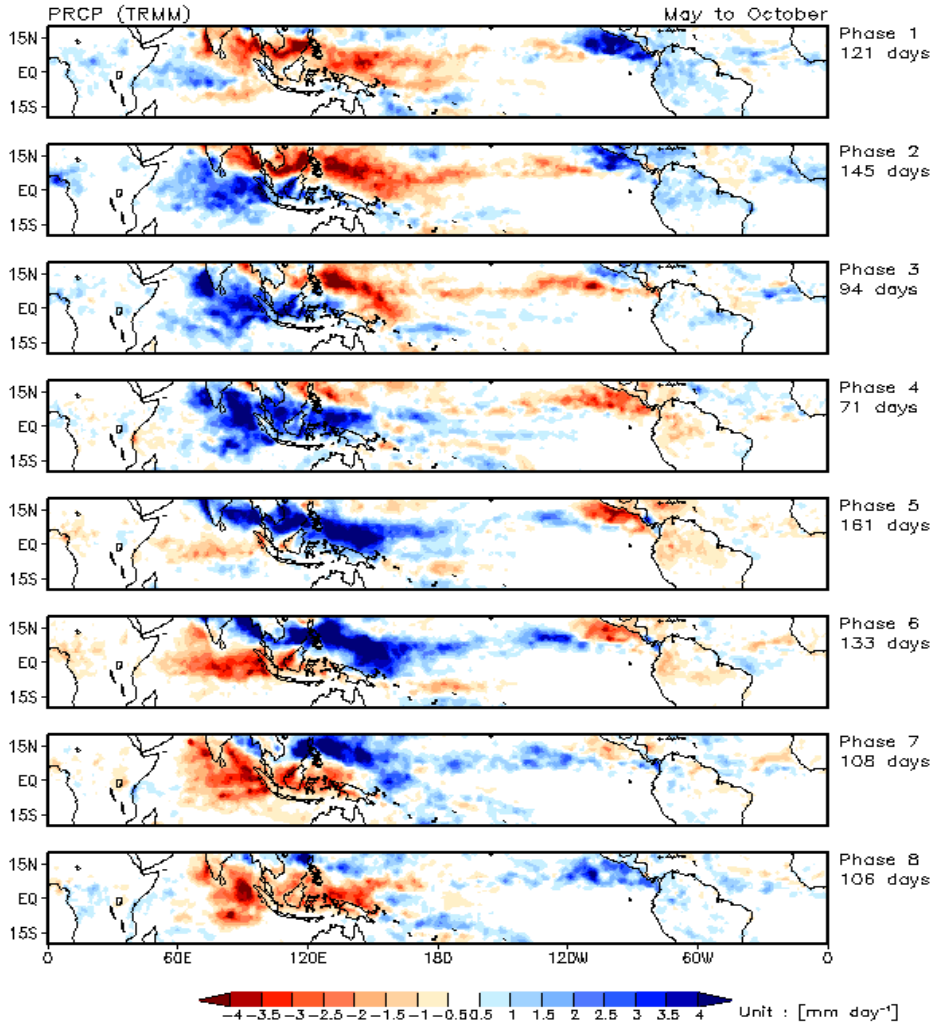


- Boreal Summer **Intraseasonal** Oscillation (2 weeks upto a season = *subseasonal*)

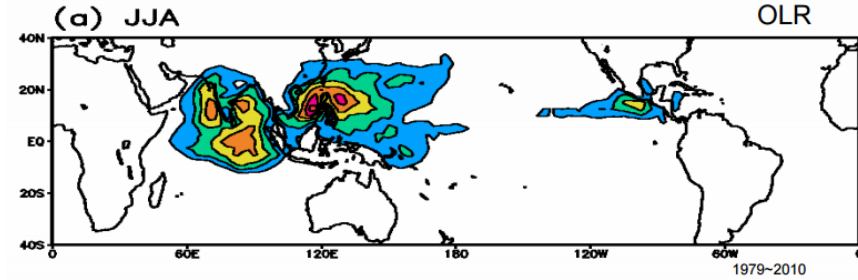


BSISO

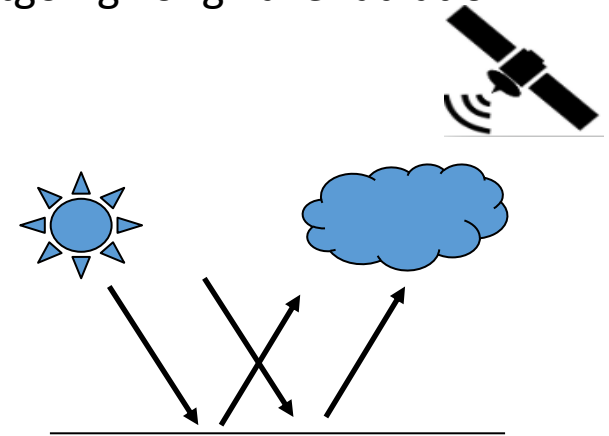
BSISO Life cycle composite



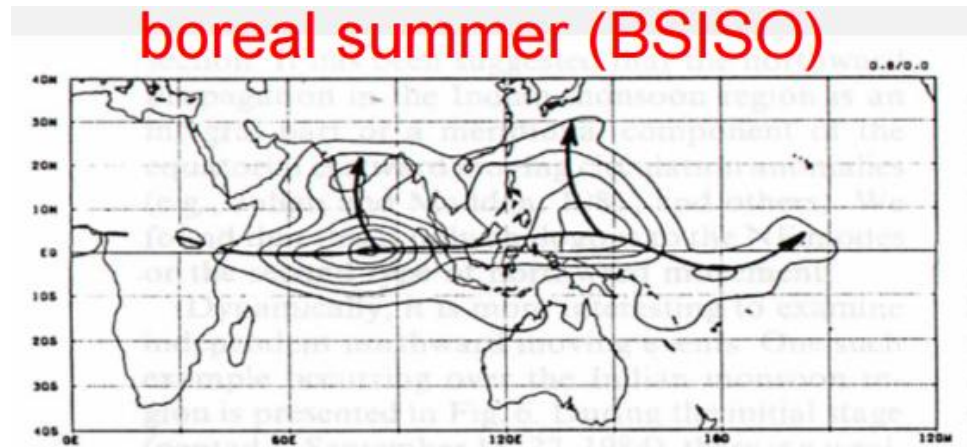
Intraseasonal Variance



- ❖ Intraseasonal: 2 weeks upto a season (20-90 days)
- ❖ OLR: Outgoing Longwave radiation



BSISO

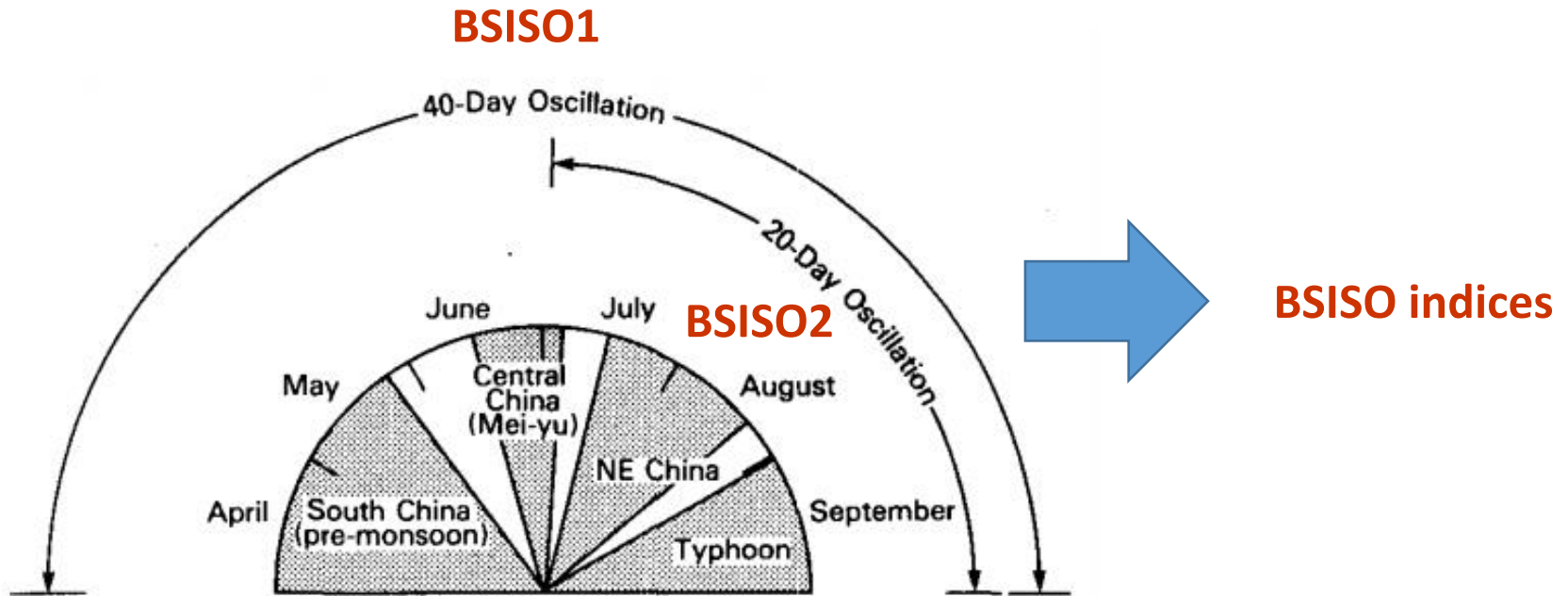


Wang and Rui 1990

- In northern summer, signal maximizes in the northern Indian Ocean and South China Sea.
- Propagation is both northward and eastward.
- **Northward propagation is related to the onset/break of Asia Monsoon**

BSISO

- Two modes of BSISO



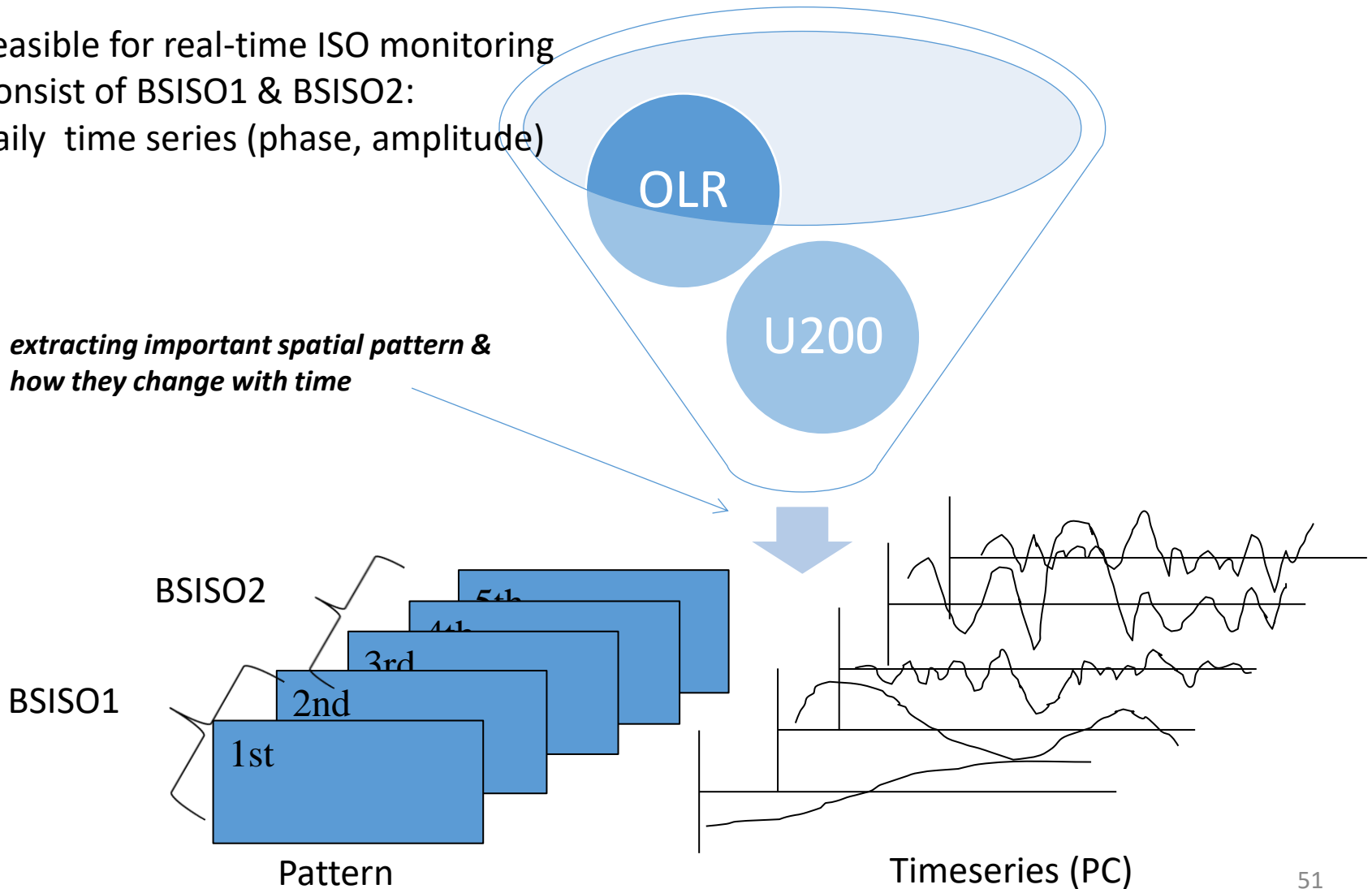
Lau et al (1988) found that the 40-day & 20-day Oscillations are related to rainfall fluctuations over East Asia.

BSISO

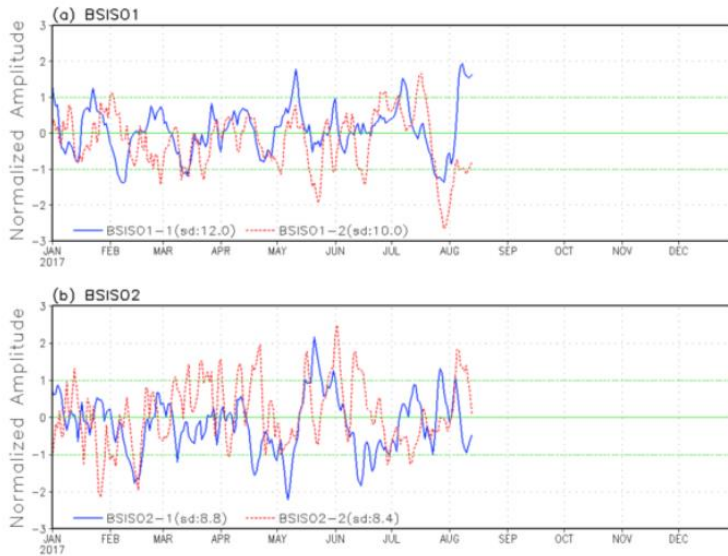
- BSISO indices: Multivariate EOF (A statistical tool)

Feasible for real-time ISO monitoring
Consist of BSISO1 & BSISO2:
daily time series (phase, amplitude)

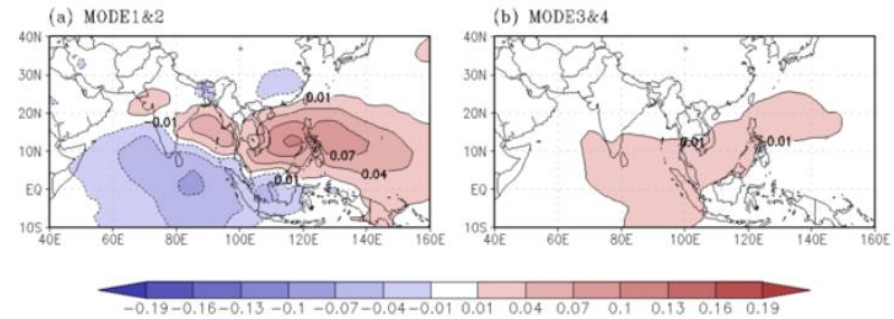
*extracting important spatial pattern &
how they change with time*



BSISO

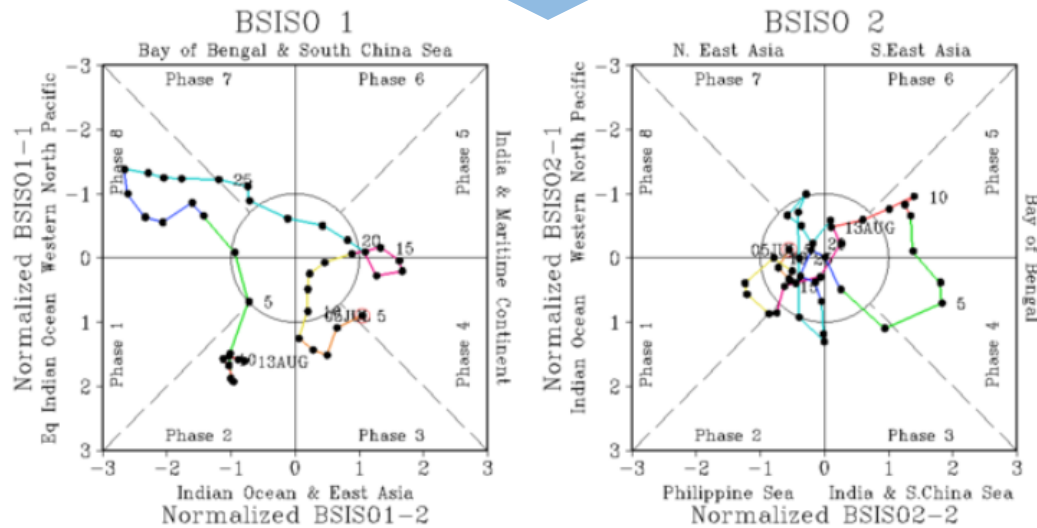


Reconstructed OLR anomaly based on the BSISO indices (13Aug2017)



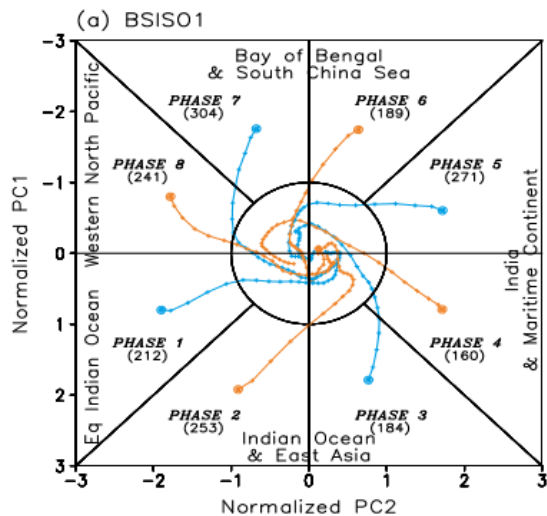
Monitoring

BSISO Monitoring for 05Jul2017~13Aug2017

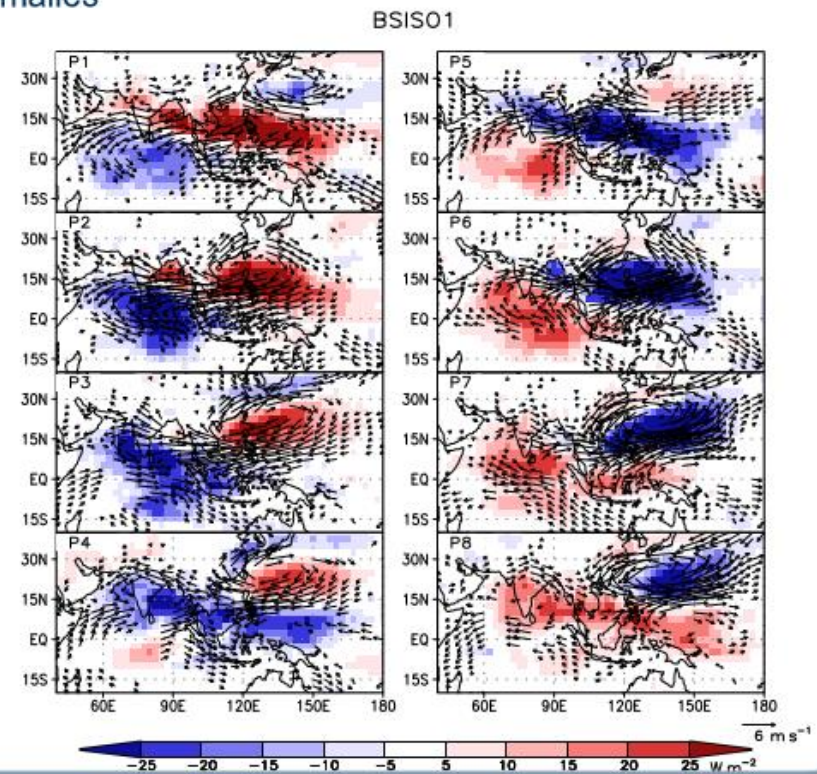


BSIS01

- Consists of EOF1 and EOF2
- Represent canonical northward and northeastward propagating ISO
- Periods of 30-60 days

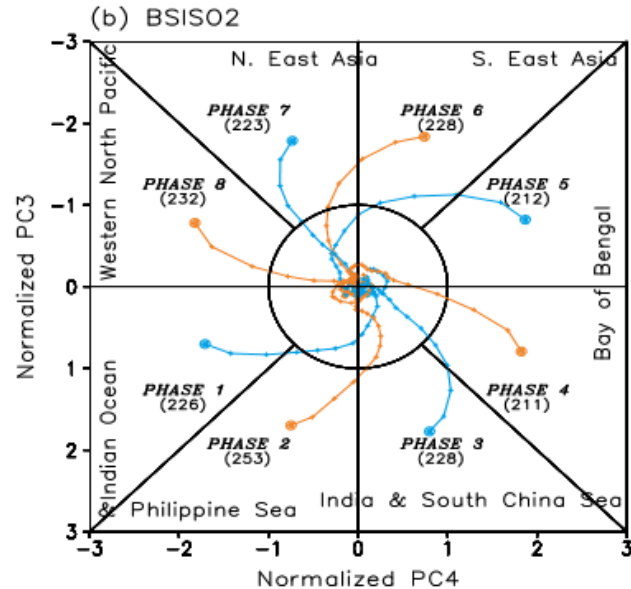


Life cycle composite of OLR (shading) and 850-hPa wind anomalies



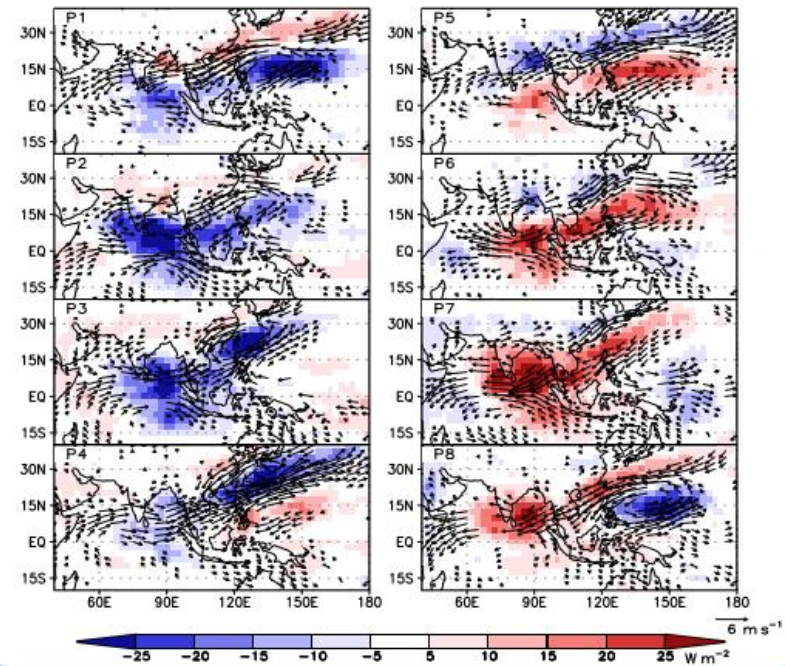
BSISO2

- Consists of EOF3 and EOF4
- Represent northward/northwestward propagating variability
- Periods of 10-30 days



Life cycle composite of OLR (shading) and 850-hPa wind anomalies

BSISO2



BSISO forecast

CLIMATE INFORMATION SERVICES



Forecast




Monitoring

Methodology

- The BSISO forecast activity has been initiated in 2013 with the goal of improving our ability to understand and forecast the BSISO based on numerical models
- The forecast is updated everyday with the latest information and is available from May to October

BSISO forecast

- Participating models

Institute	Model	Ensemble Size	Forecast Period	Update frequency	Resolution
 NCEP	Climate Forecast System	4	40 days	Once a day	T126 L64
	Global Forecast System	1	16 days	Once a day	T574, T190 L64
 Australia	POAMA 2.4 multi-week model	33	40 days	Twice per week	T47 L17
 ECMWF	ECMWF Ensemble Prediction System	51	32 days	Twice per week	T639, T319 L62
Taiwan CWB	CWB EPS T119	1	40 days	Every 5 days	T119 L30

BSISO forecast

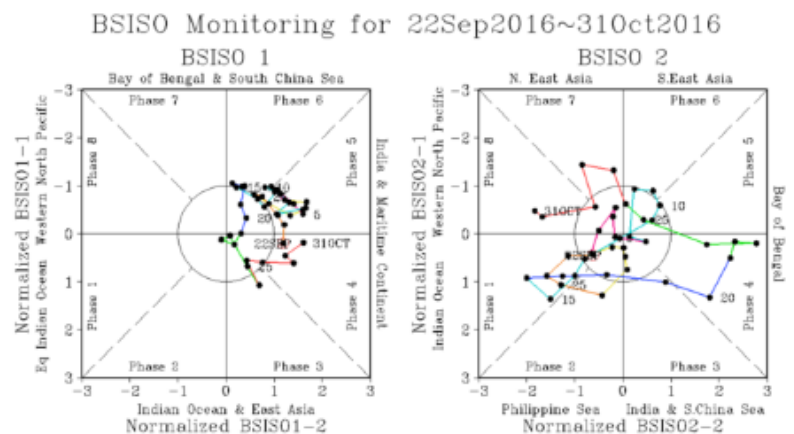
- Monitoring

Home > Climate Information Service > BSISO Forecast > **Monitoring**

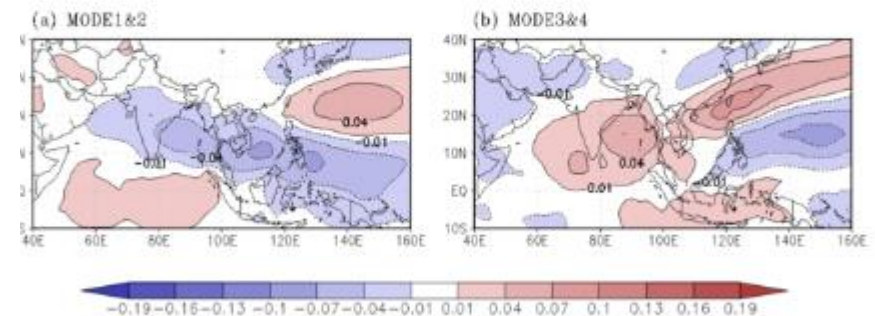
Monitoring

Welcome to the Boreal Summer Intraseasonal Oscillation (BSISO) monitoring website. The BSISO, one of the dominant phenomena over the Asian summer monsoon region, is characterized by northward/northeastward propagation over the Indian summer monsoon region and northward/northwestward propagation over the Western North Pacific-East Asian region, including equatorial eastward propagation. **This monitoring information is available from May to October.**

[Text file of Normalized Time Series for the BSISO1 and BSISO2 index](#)



Reconstructed OLR anomaly based on the BSISO indices (31Oct2016)



BSISO forecast

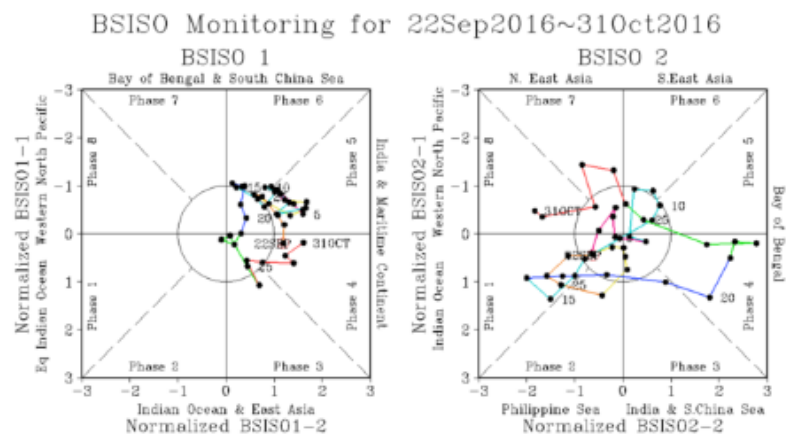
- Monitoring

Home > Climate Information Service > BSISO Forecast > Monitoring

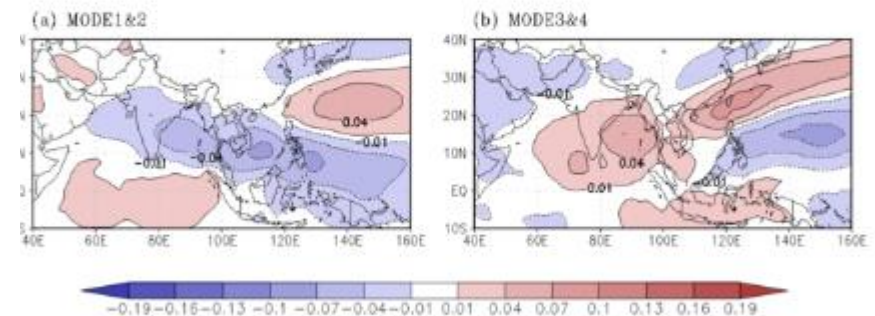
Monitoring

Welcome to the Boreal Summer Intraseasonal Oscillation (BSISO) monitoring website. The BSISO, one of the dominant phenomena over the Asian summer monsoon region, is characterized by northward/northeastward propagation over the Indian summer monsoon region and northward/northwestward propagation over the Western North Pacific-East Asian region, including equatorial eastward propagation. **This monitoring information is available from May to October.**

[Text file of Normalized Time Series for the BSISO1 and BSISO2 index](#)



Reconstructed OLR anomaly based on the BSISO indices (31Oct2016)



BSISO forecast

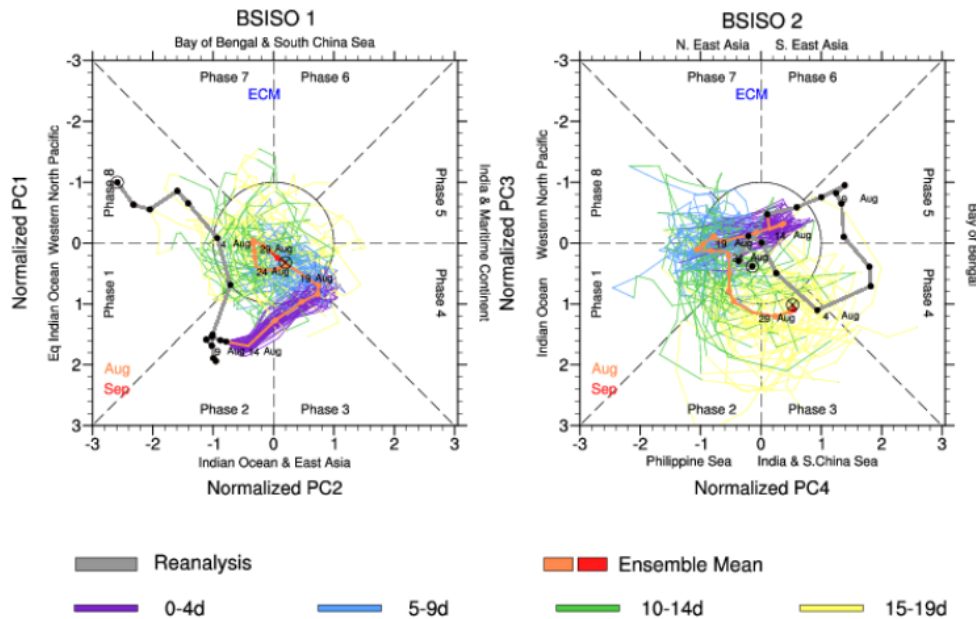
- Forecast

Phase Diagram	Spatial OLR Anomalies	Heavy Rainfall Probability	Verification	Participation
---------------	-----------------------	----------------------------	--------------	---------------

Phase Plots of BSISO Index Forecasts

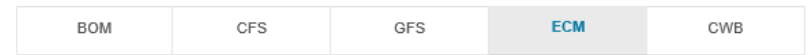
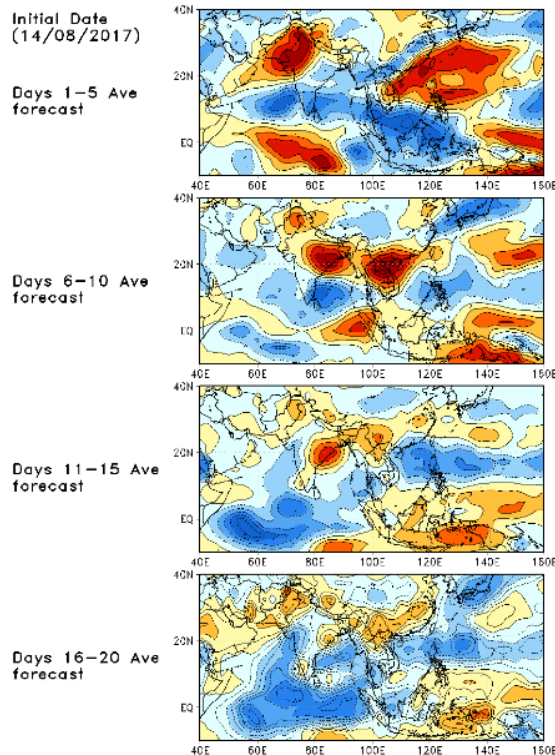
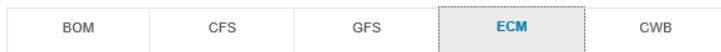
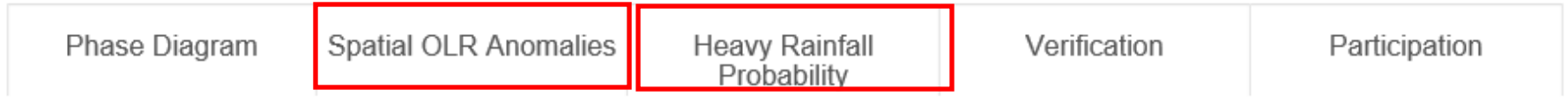
BOM	CFS	GFS	ECM	CWB
-----	-----	-----	-----	-----

BSISO Forecast for 14Aug2017-2Sep2017

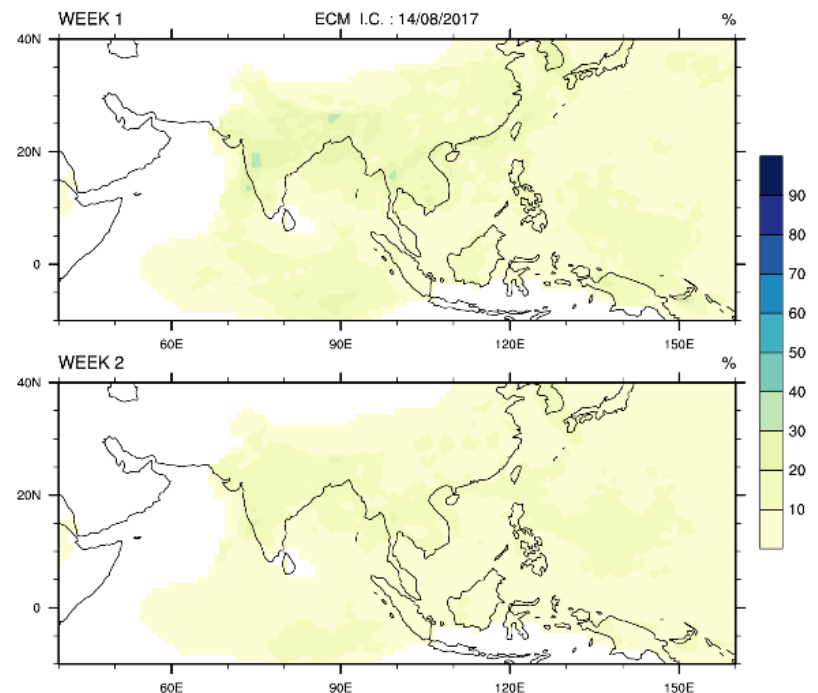


BSISO forecast

- Forecast



Probability of heavy rainfall determined by predicted BSISO



Probability of occurrence for heavy rainfall event as defined by daily rainfall exceeding the 90th percentiles value (22.6 mm/day) for Aug. during 1981-2010.

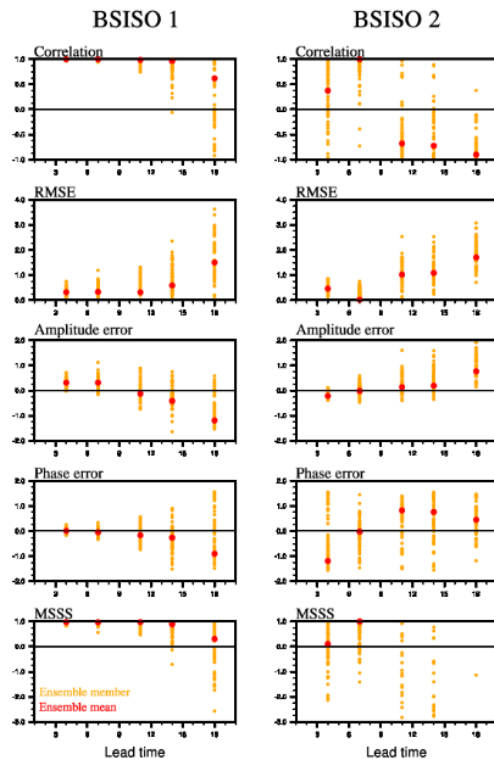
BSISO forecast

- Forecast

Phase Diagram	Spatial OLR Anomalies	Heavy Rainfall Probability	Verification	Participation
---------------	-----------------------	----------------------------	--------------	---------------



BSISO verification for 13Aug 2017(ECM)



- Correlation coefficient**

Perfect score: 1

Skill in forecasting the phase of the BSISO

- Root Mean Square Error (RMSE)**

Perfect score: 0

Errors in both phase & amplitude

- Phase amplitude**

Perfect score: 0

Relative amplitude difference btw observation and forecast

Postive: forecast amplitude is larger than the observed

- Phase error**

Perfect score: 0

Positive: phase speed of the forecast is faster than that of the observation

- Mean square skill score (MSSS)**

Perfect score: 1

Relative level of skill of forecast compared to a climatological forecast that predict no BSISO signal

THANK YOU