

Seasonal Prediction (0) : Climate Variability

Jin Ho Yoo
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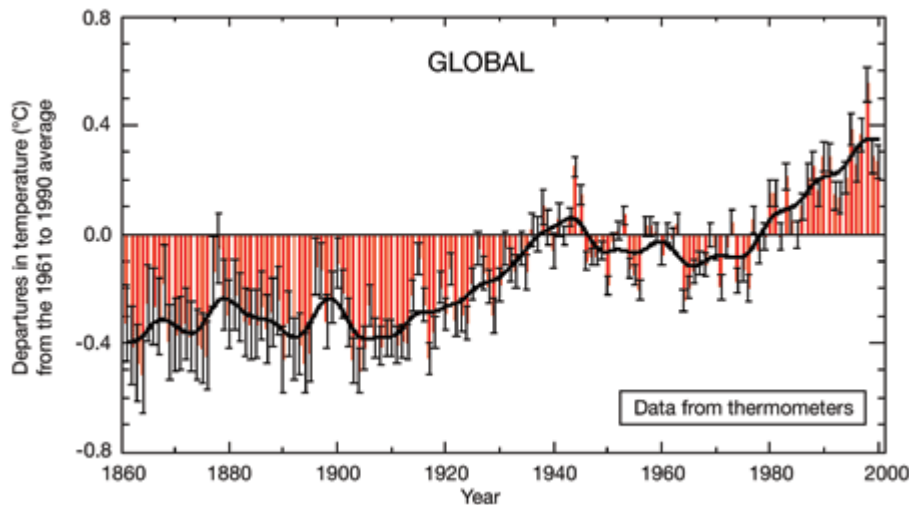
Tell us your “weather” during a season

1. What happens?
2. Irregularity?

Q: What climate phenomena are
you familiar with?

Climate Change

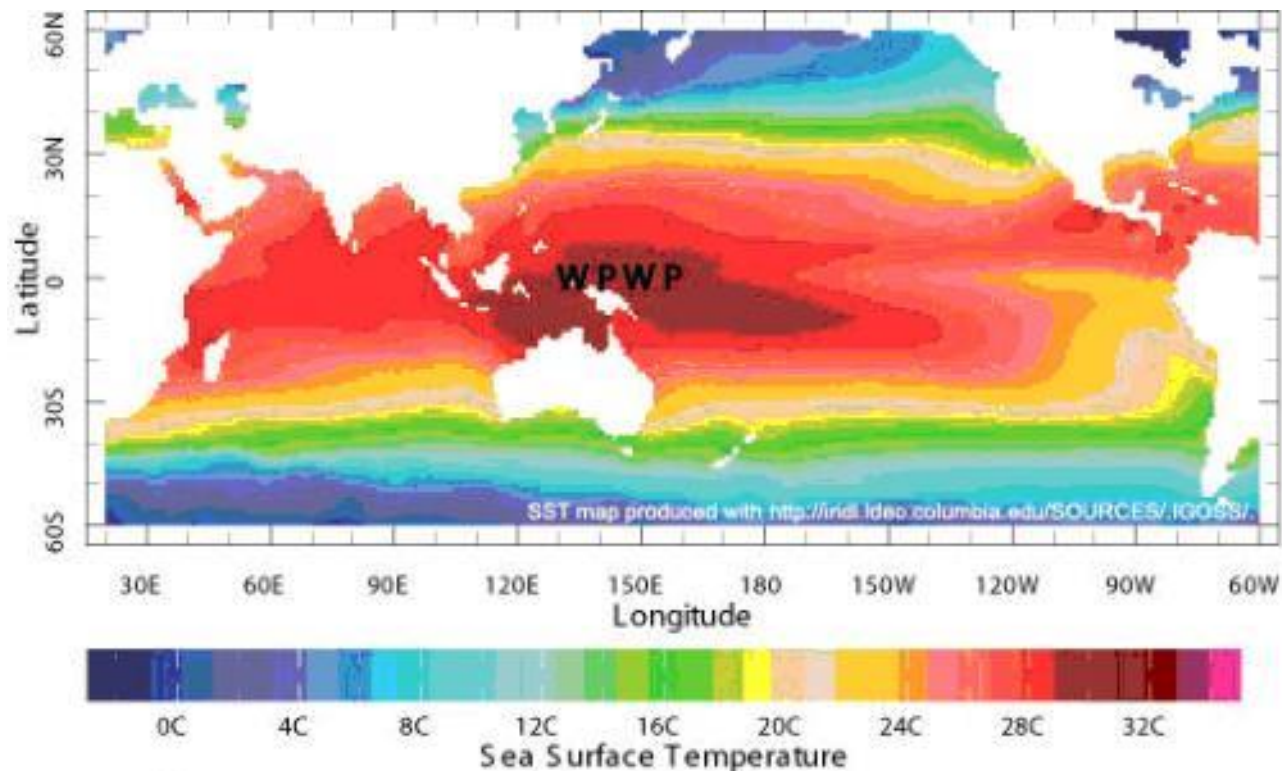
- Changes in our **expectation**



search ID: mlyn203

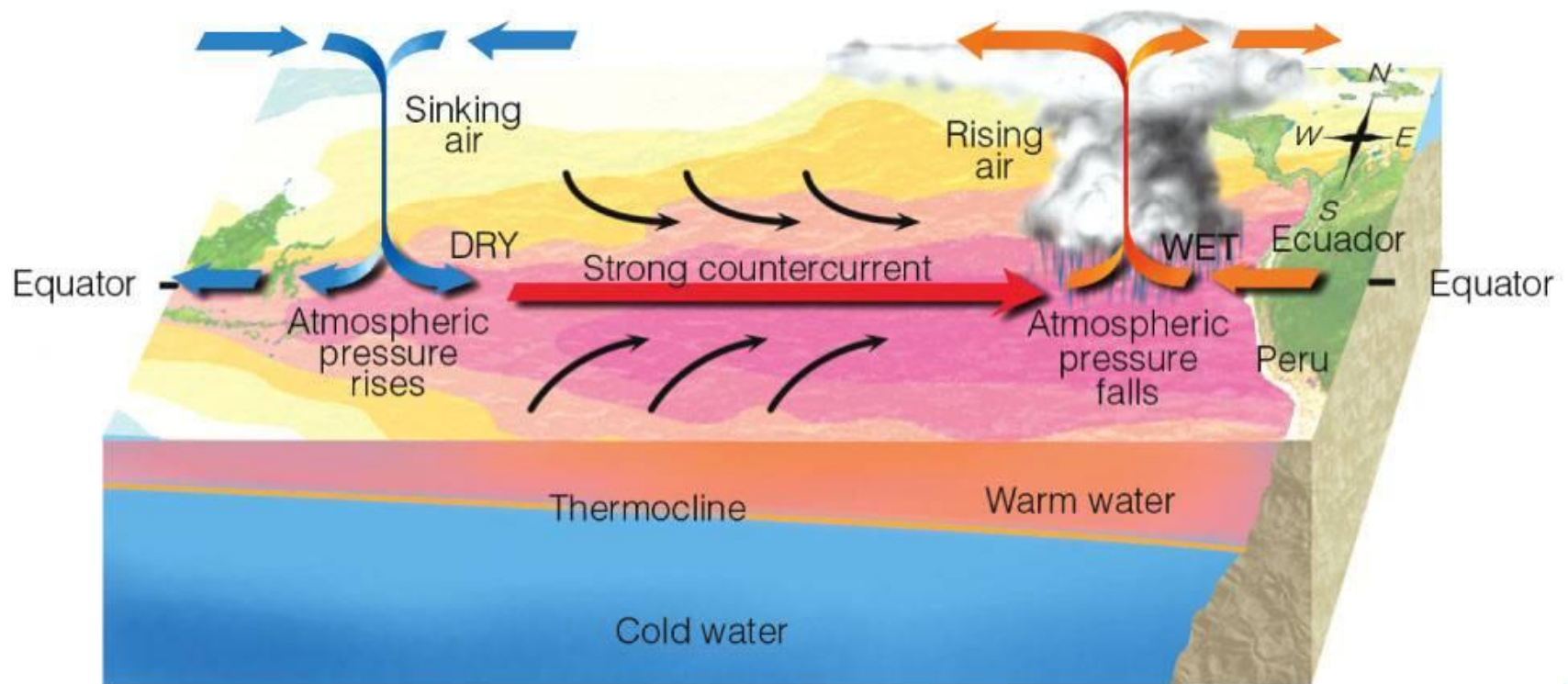
Warm pool

- Climate Engine : remember “mean” feature



El Nino, ENSO

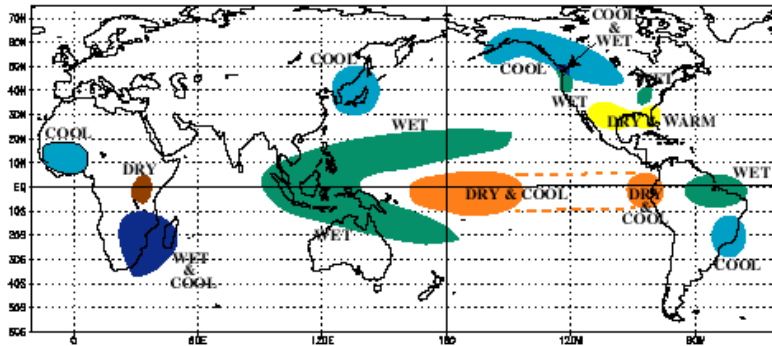
- A Big Ocean Swing



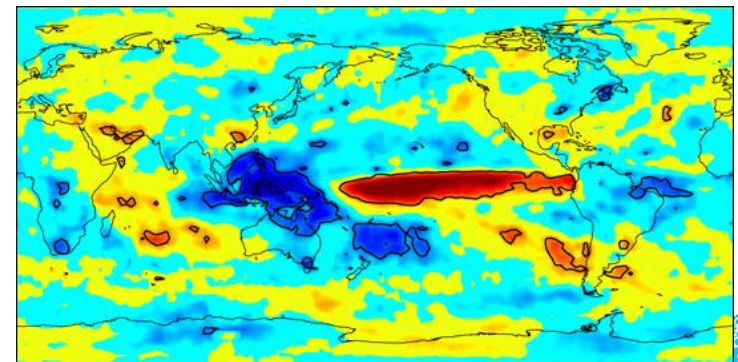
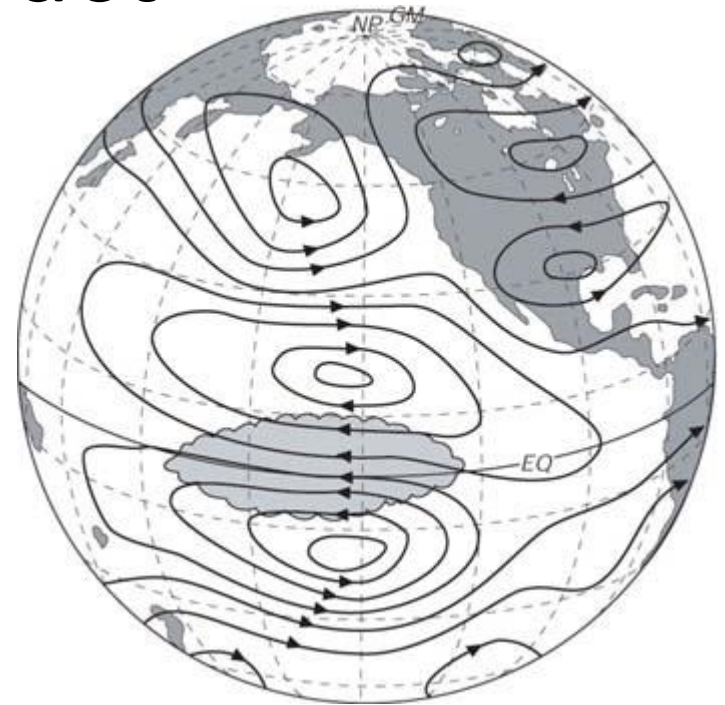
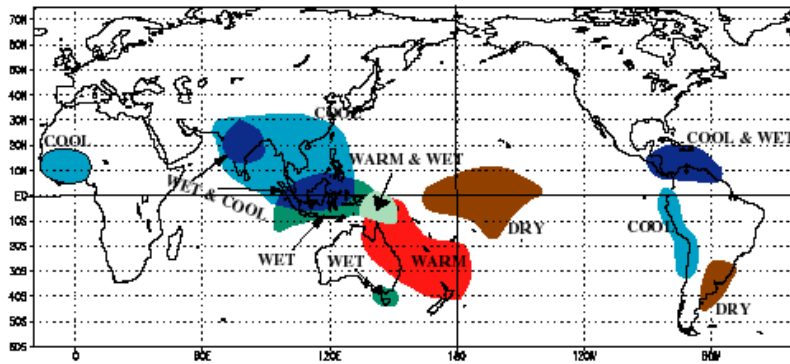
(b) El Niño Conditions

ENSO impact

COLD EPISODE RELATIONSHIPS DECEMBER - FEBRUARY



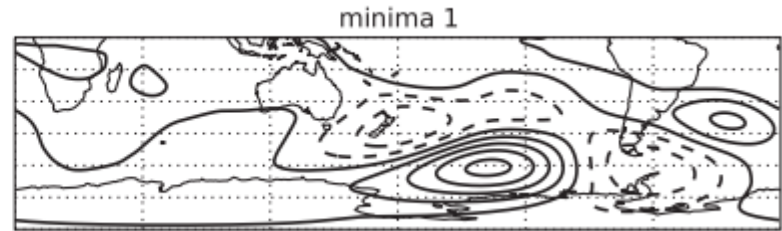
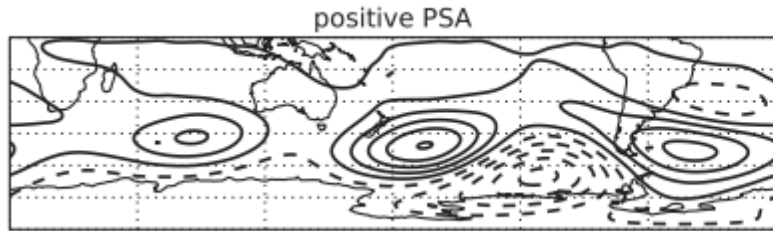
COLD EPISODE RELATIONSHIPS JUNE - AUGUST



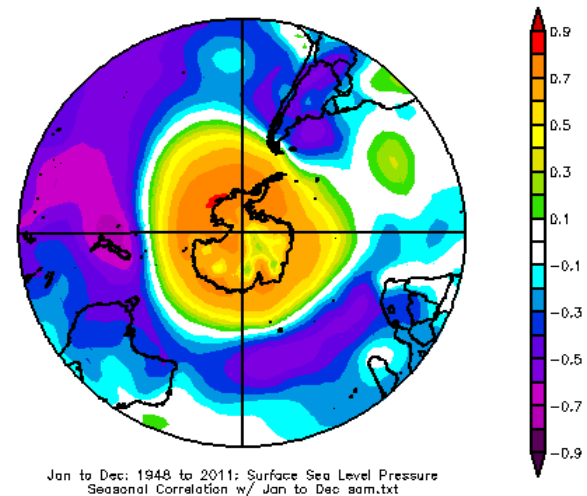
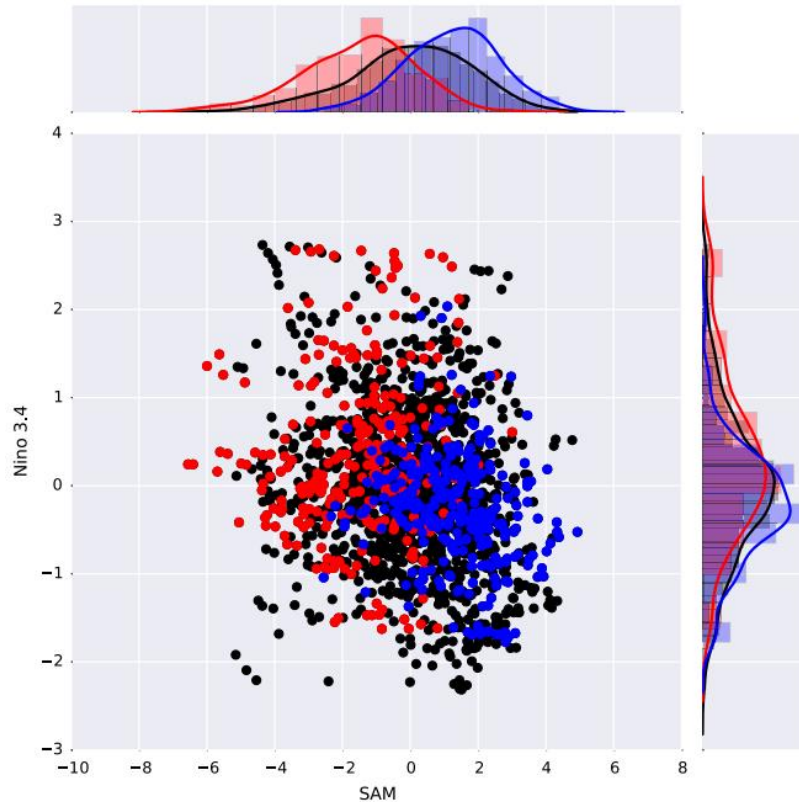
Precipitation (peak El Niño)
 lower than normal normal higher than normal



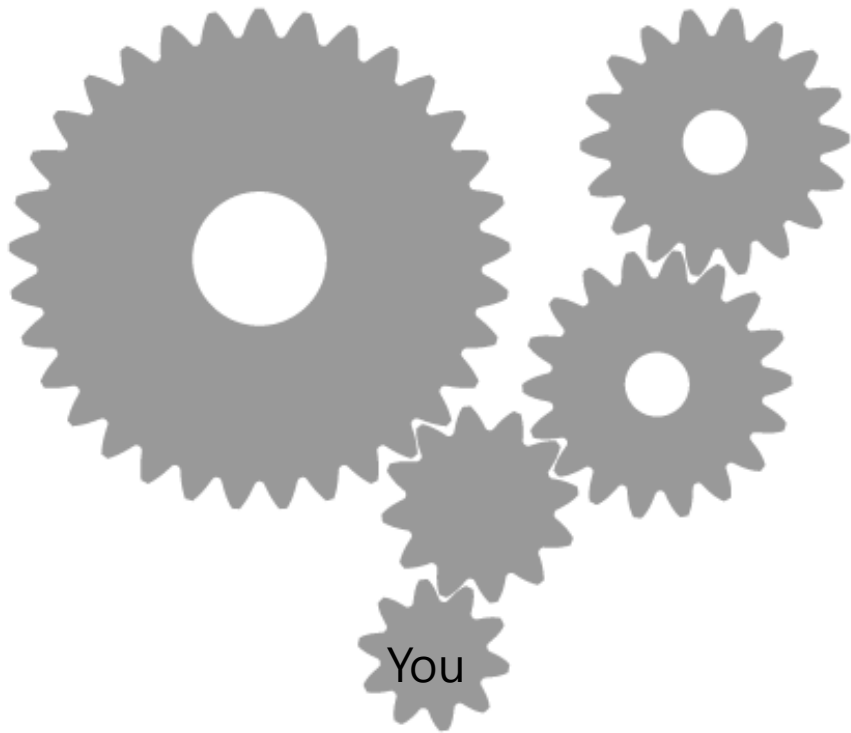
Extratropical LFV (SAM, PSA)



Irving and Simmonds (2016)

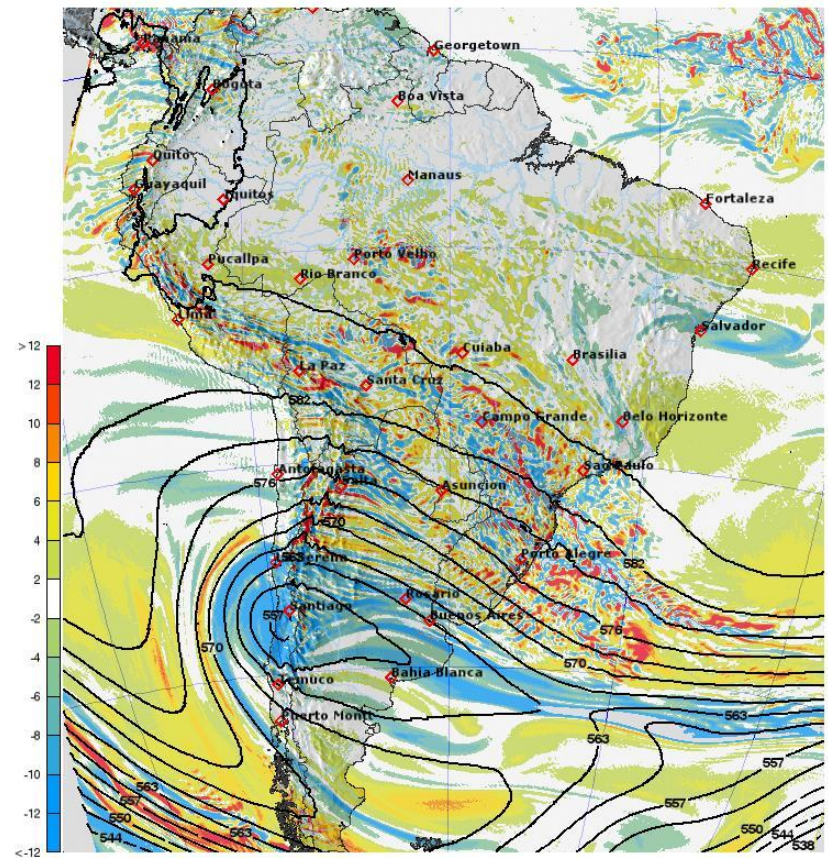


How they change weather?



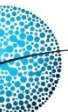
Rel. Vorticity 500 mb & 500 mb Geopotential height (gpm)

Sun 04.09.2016 09:00 ART



www.meteoblue.com

NEMSSA RUN: 03.09.12Z



ICC
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Seasonal Prediction (1) : Introduction/Predictability

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Overview

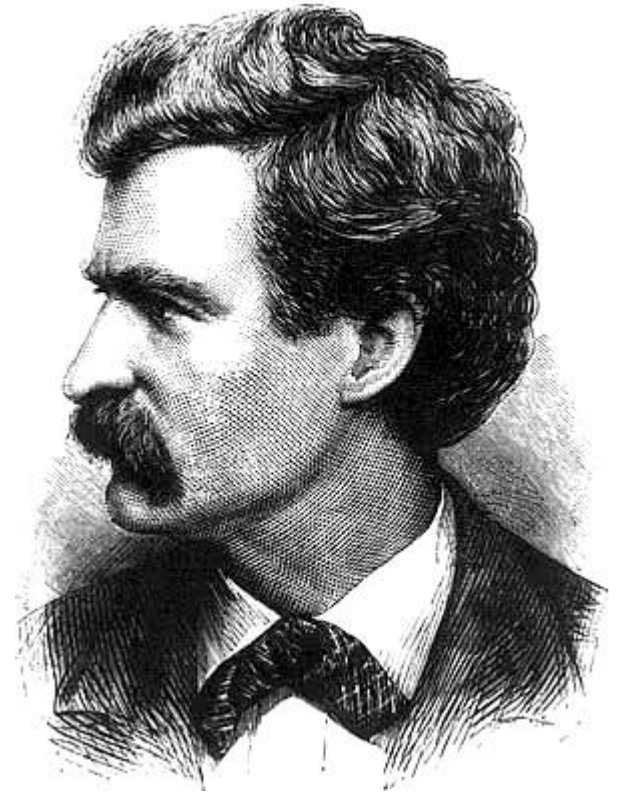
- Predictability
- Methods
- Verification + Downscaling
- Operation

Climate prediction

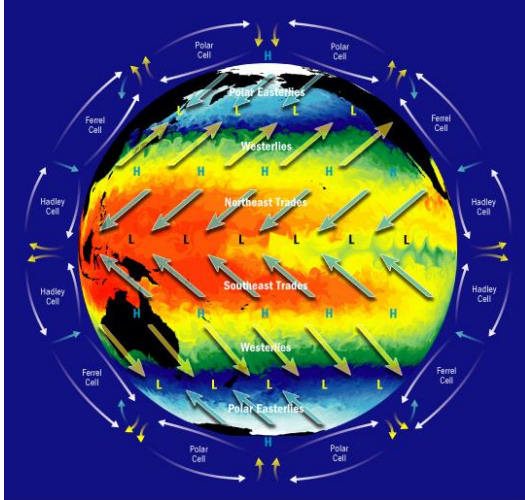
Climate

**Climate is
what we expect,**

**Weather is
what we get**



Climate = Expectation



Climate Change = We need to change our Expectation

Climate prediction = Expectation of Expectation

How uncertain!

Prediction

a rigorous, (often quantitative), statement forecasting **what** will happen **under specific conditions**

Prediction (in Meteorology)

a rigorous, (often quantitative), statement forecasting **what** will happen **under specific conditions**

What : atmospheric state
Conditions??

Atmosphere is dynamical system

$$\frac{d\vec{X}}{dt} = F(\vec{X}, a)$$

$$\vec{X}(t_0 + \tau) = \vec{X}(t_0) + \int_0^\tau F(\vec{X}(t), a(t)) dt$$

Prediction (in Meteorology)

a rigorous, (often quantitative), statement forecasting **what** will happen **under specific conditions**

What : atmospheric state (weather)

Conditions: Current state, Physical rules, external forcing factors

Determinism

$$\frac{d\vec{X}}{dt} = F(\vec{X}, a)$$

Perfect prediction is possible when we have knowledge of all necessary “conditions”

Chaos

Small difference in the initial state cause huge difference later even in the deterministic nonlinear system.

$$\frac{d\vec{X}}{dt} = F(\vec{X}, a)$$

Our knowledge is never perfect!

→ perfect forecast is impossible

How well we can predict?

“Predictability”

Predictability

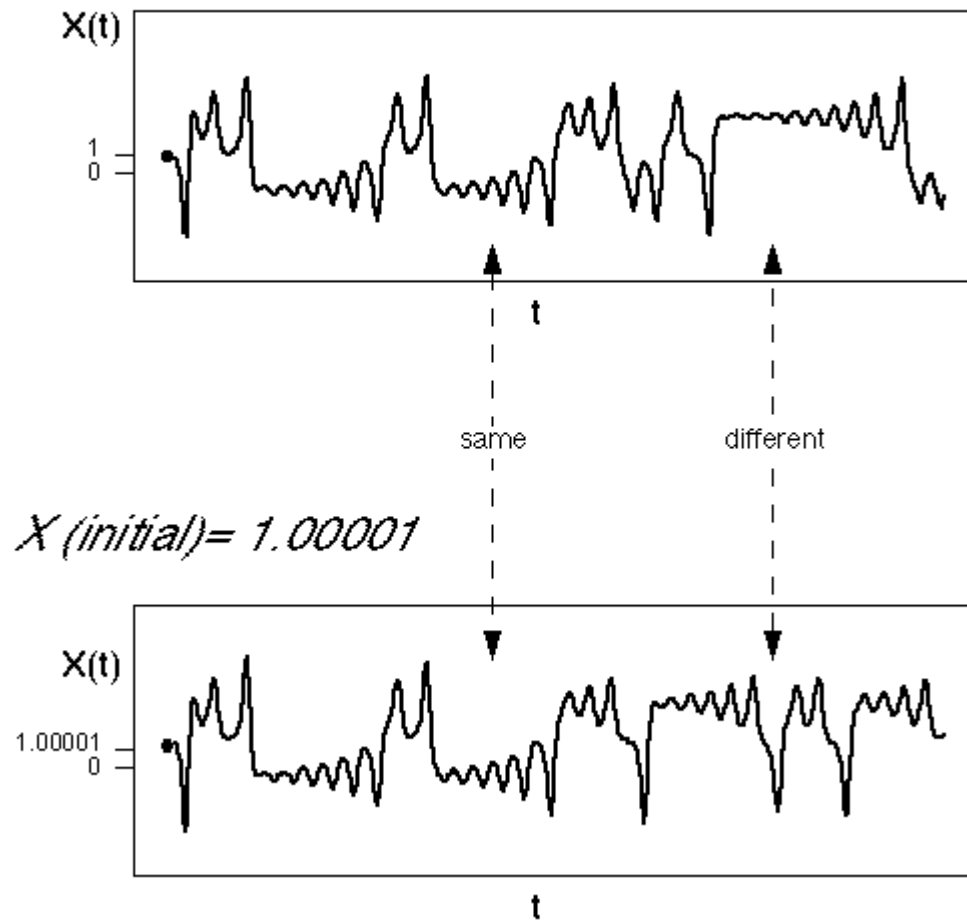
Depends on *what to predict*

Prediction of

1. *Temperature of this room tomorrow*
2. *Temperature of this room in 30days later*
3. *Temperature of this room in 30years later*

Lead time(τ)

$$X(\text{initial}) = 1.$$



Predictability

Depends on *what to predict*

Prediction of

- 1. Temperature of Seoul (Korea)*
- 2. Temperature of Jakarta (Indonesia)*
- 3. Temperature of Villa Las Estrellas (Antarctica)*

Location

Predictability

Depends on *what to predict*

Prediction of

- 1. Temperature*
- 2. rainfall*
- 3. wind speed*

Physical variables

Why Predictability is varying with location/variables

Characteristics of variability is different

- Tropics : weather = local convection (time scale ~ few hours)
- Extratropics : weather = synoptic system (time scale ~ few days)
- Daily rainfall is more chaotic (highly nonlinear) than temperature/pressure

Predictability

Depends on *what to predict*

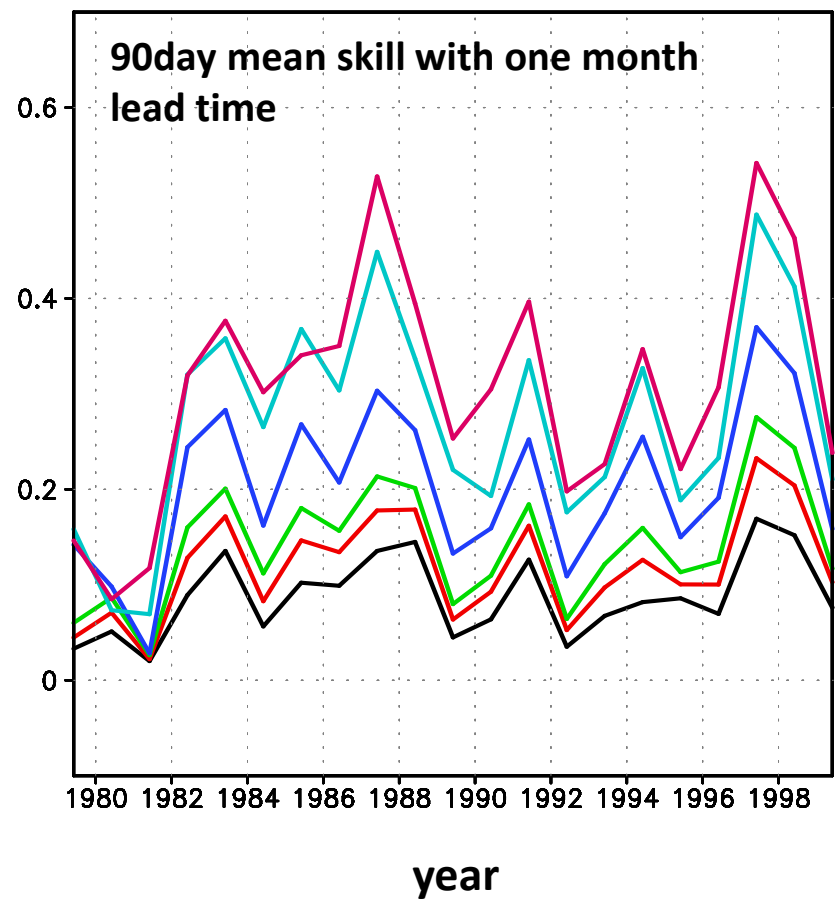
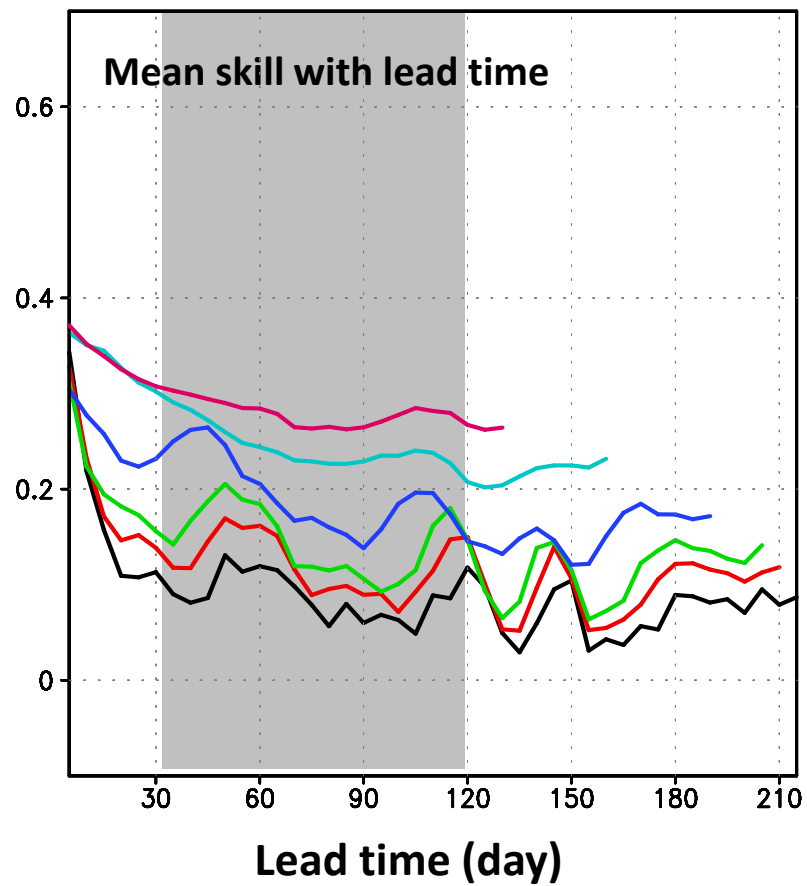
Prediction of

- 1. Mean Temperature during a day*
- 2. Mean Temperature during a month*
- 3. Mean Temperature during a century*

Time scale of predictand

Seasonal mean and Intraseasonal predictability

Global pattern correlation skill of GCPs precipitation forecast (SMIP)



5day, 10day, 15day, 30day, 60day, 90day averaged field

How long?

Time mean of weather

Climate prediction

Seasonal forecast



Seasonal Prediction

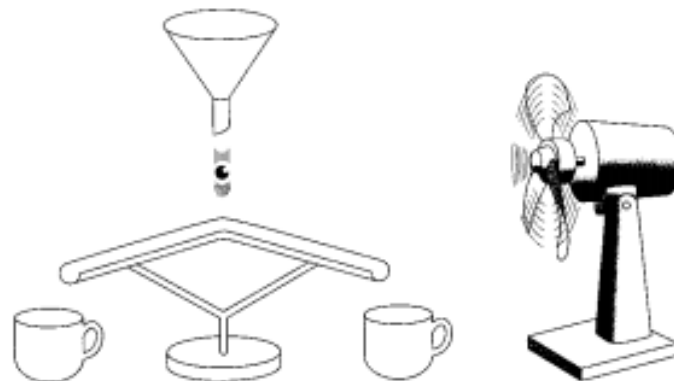
What : state of atmosphere **during a season**

Condition : Current state, Physical rules, external forcing factor

Lead time \sim 1 month (e.g. DJF forecast at Nov)

History of Short-term (Seasonal) Climate Prediction

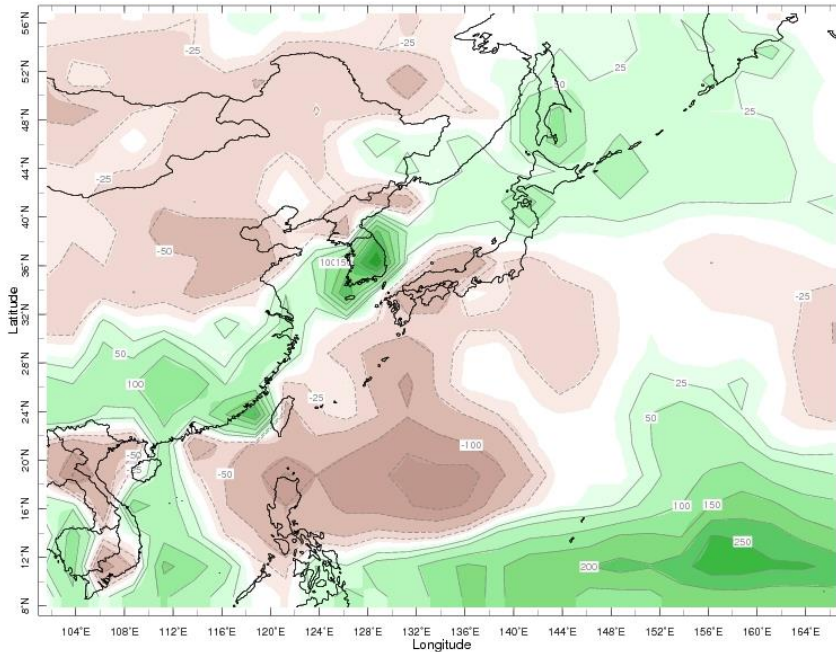
- 1960's : Hypothesis proposed
- 1980's : **ENSO** prediction + Atm. LFV. (PNA..)
- 1990's : (Experimental) Dyn. Seasonal Fcst.
- 2000's : International collaboration (MIPs)
- 2010's : Operation (GFCS, RCOFs/WMO)



T. Palmer (1998)

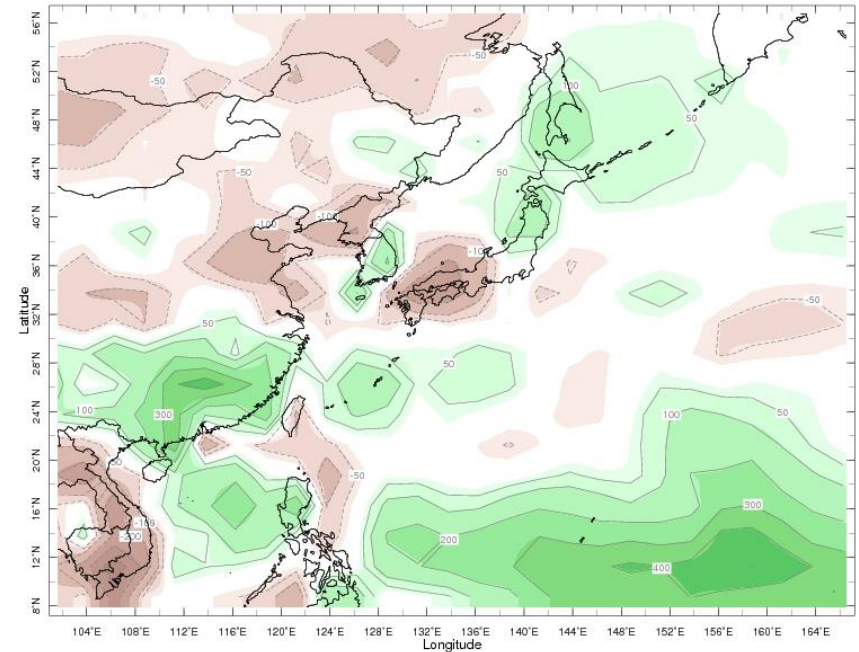
2002 summer rainfall

Monthly mean prec. (Aug)



Aug 2002

Summer mean prec.

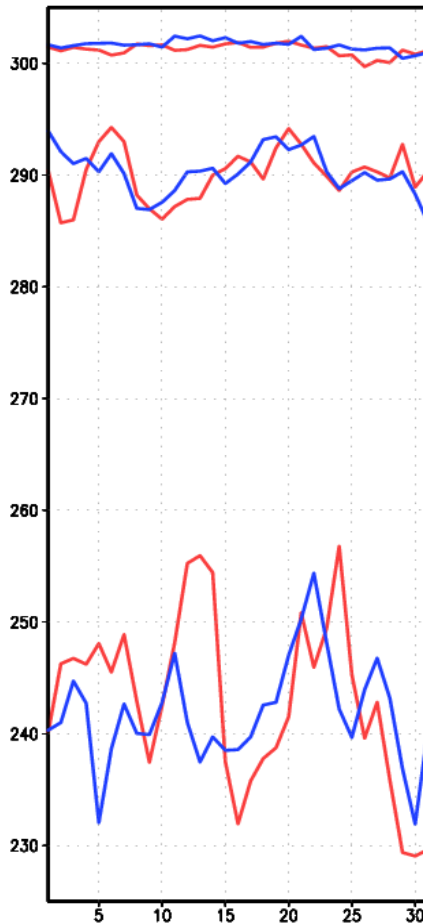


Jun-Aug 2002

Typhoon "RUSA" passed at 8/31 (1000mm a day)

Seasonal forecast

How is the seasonal mean determined?

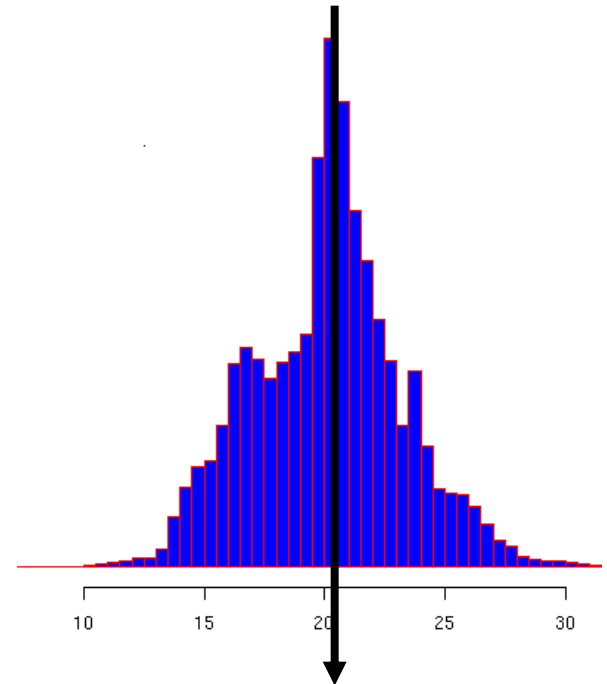
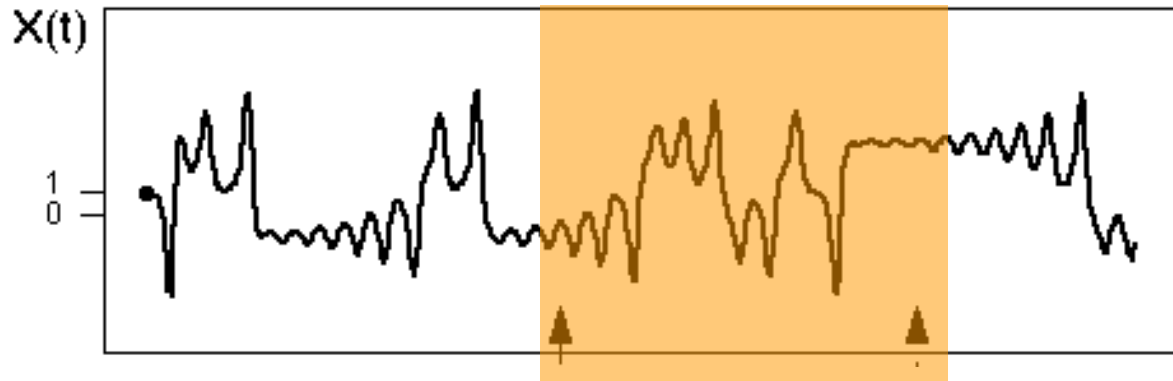


What causes change (variability) of the mean?

- By chance?
- By “something”?



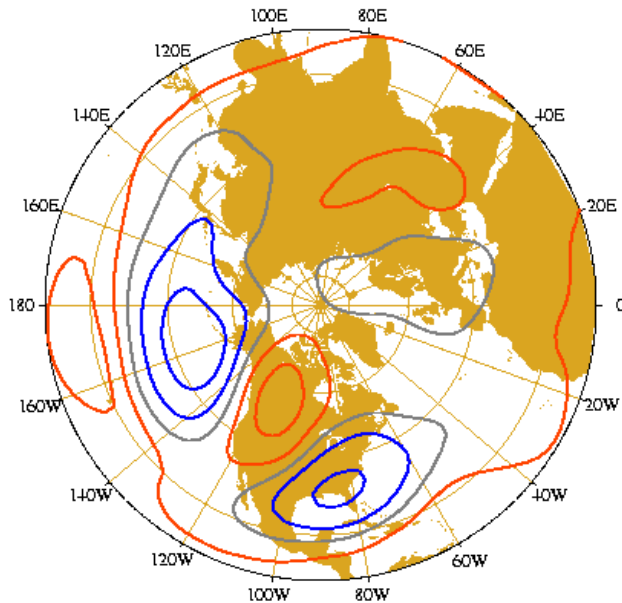
Weather statistics



Primary seasonal weather statistics : seasonal mean

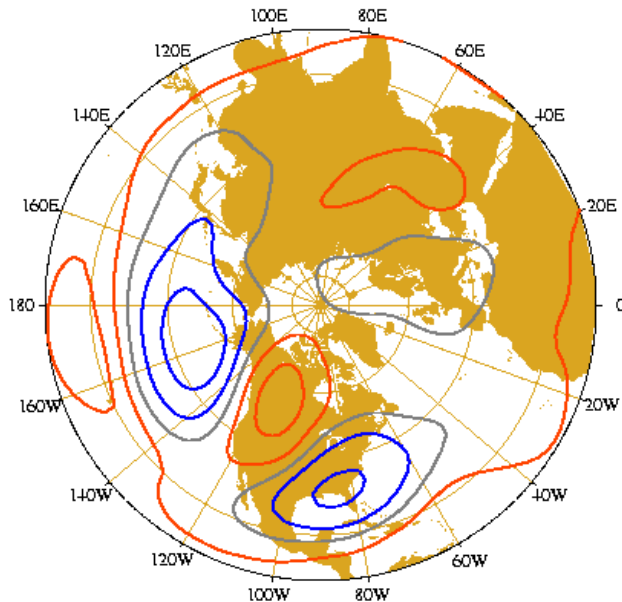
Seasonal mean

PNA debates



1. Forced by El Nino
2. Atmospheric internal variability (random)

PNA debates



1. Forced by El Nino
Predictable (signal)
2. Atmospheric internal
variability (random)
Unpredictable (noise)

Matter of Signal & Noise

$$X = X_s + X_n$$

Potential predictability

Measured by relative magnitude (variance) of signal and noise

Signal >> Noise : more predictable

Signal << Noise : less predictable

Signal in Seasonal prediction

- What is the **Signal**? (How we can “see”?)
 - Tendency of weather that has be physically caused by slow varying processes
- What derives the Signal?
 - External forcing (or interaction)
 - Slow varying processes (ENSO)

Mechanisms of Variability

Internal

External

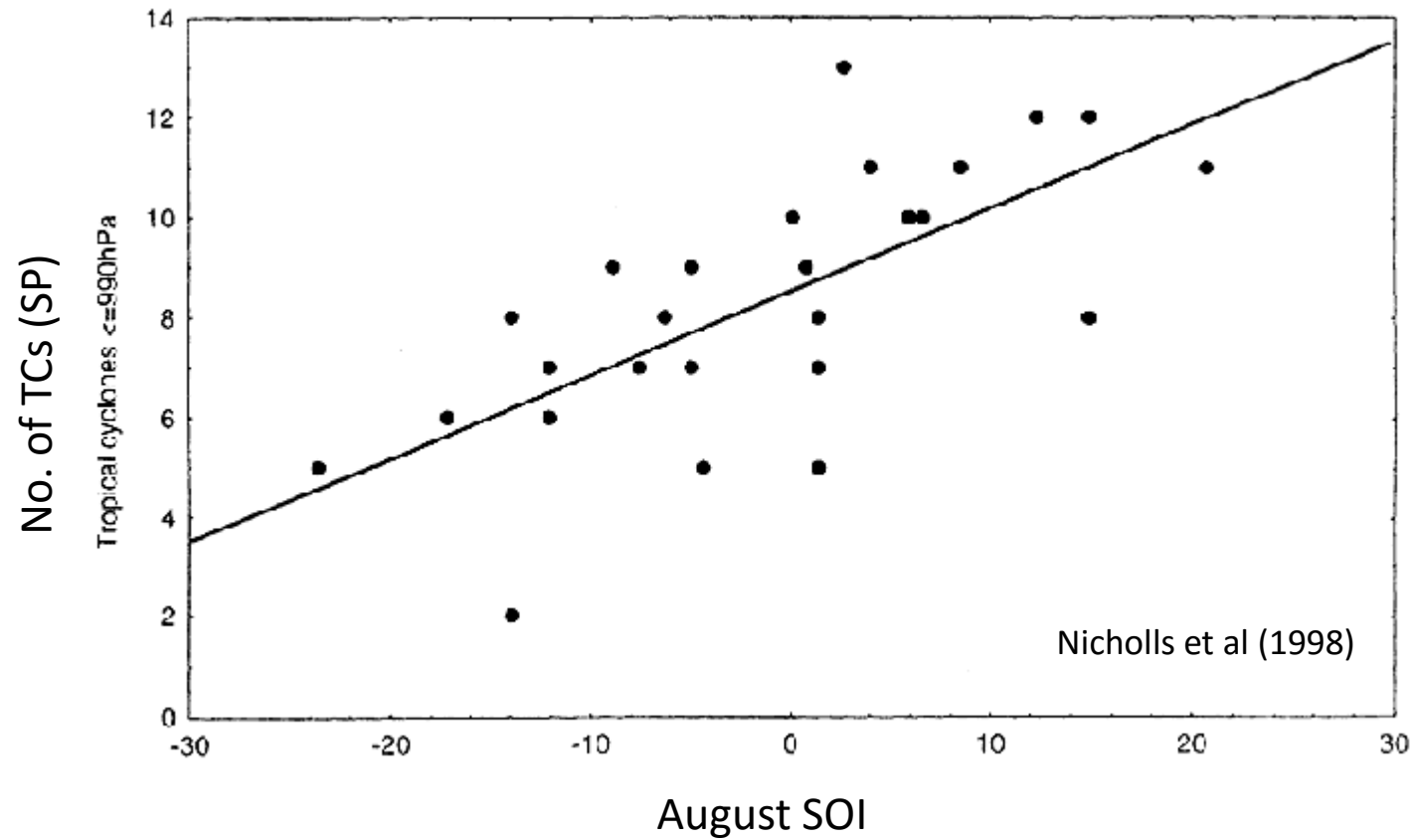
Weather:	1. Internal Dynamics of Atmosphere	<ul style="list-style-type: none">• Boundary Condition of SST, Soil wetness, Snow, Sea ice, etc.
Climate: (seasonal-decadal)	2. Internal Dynamics of Coupled Ocean-Land-Atmosphere	<ul style="list-style-type: none">• Solar, Volcanoes
Climate Change:	3. Internal Dynamics of Sun-Earth System	<ul style="list-style-type: none">• Human effects: (Greenhouse gases, land use changes)

From J. Shukla (2007)

Two scales

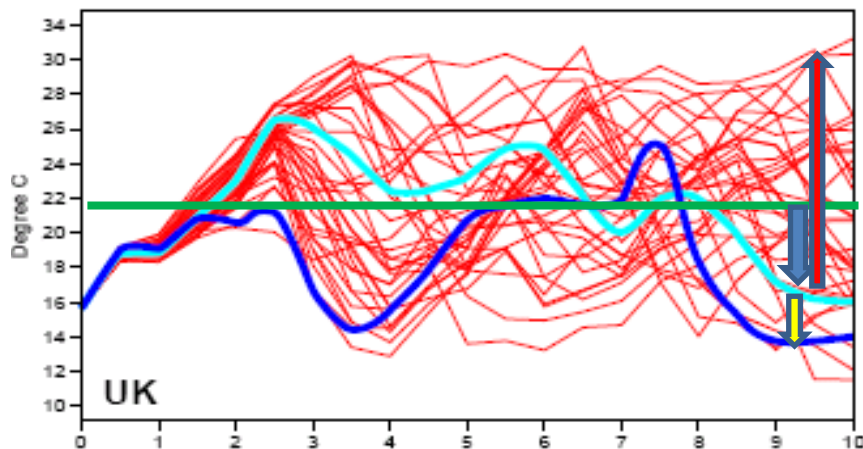
- **Fast and small** scale processes : noise
 - Weather, Tropical cyclone
- **Slow and large** processes : signal
 - Climate, ITCZ, ENSO

Two scales



Predictability

- Relative ratio between signal and noise
- BUT we don't know actual signal
 - Estimation of potential predictability by models
 - Ensemble prediction

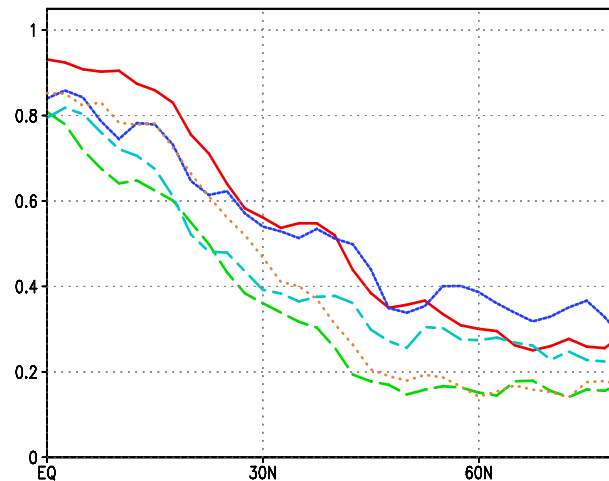
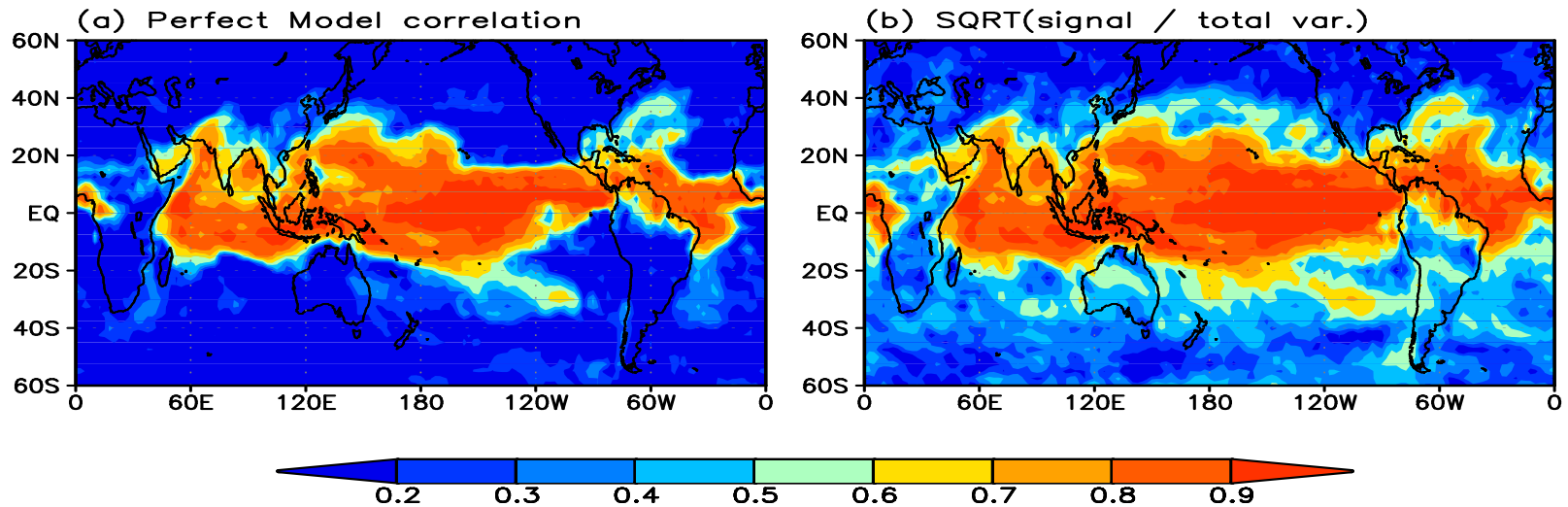


$$X = X_s + X_n$$

X_s : ensemble mean

X_n : deviation from ensemble mean

Estimated potential predictability of rainfall



Potential predictability

- Estimated limit of the predictability given prediction methods (model)
 - Depends on **nature** itself as well as prediction **model**
 - We cannot change the nature but model is our product
 - Potential predictability may be able to be improved (or not) if our model is improved

Seasonal Prediction (2) : Methods

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Methods

- Statistical (Empirical)
 - Use observed relationship of climate system to predict future
 - Linear
- Dynamical
 - Based on “physical law” of climate system and expect to mimic “the memory”
 - Nonlinear

Which one is better?

Statistical

- Simple and cheap
- Based on data
- Data is real thing but do we have enough?

Dynamical

- Complex and expensive
- Based on Law
- Is our understanding accurate?

Statistical forecasting

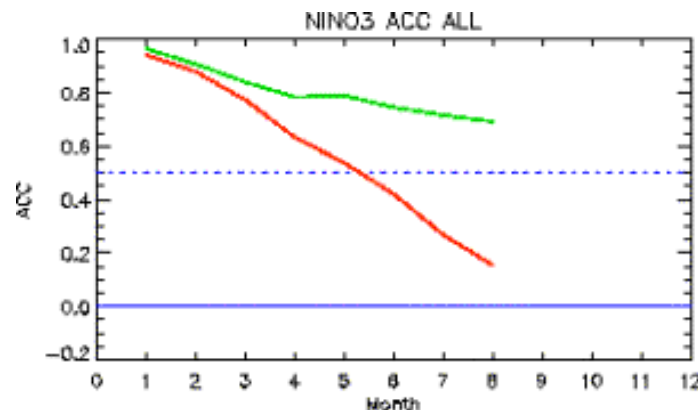
- (0) Climatology

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

- Baseline of seasonal forecasting
- “Nothing particular, Sir.”
- Deterministic forecast
 - Rainfall amount will be similar to 30year average
- Probabilistic forecast
 - Near normal ?
 - I don't know? (33%:33%:33%)

Statistical forecasting

- (1) Persistence $x'(t + 1) = x'(t)$
 - Assume that future will be same as it is now
 - ANOMALY !
 - Often Close to people's expectation
 - Effective when the autocorrelation is large
 - Often used for ENSO forecast (Nino3.4)

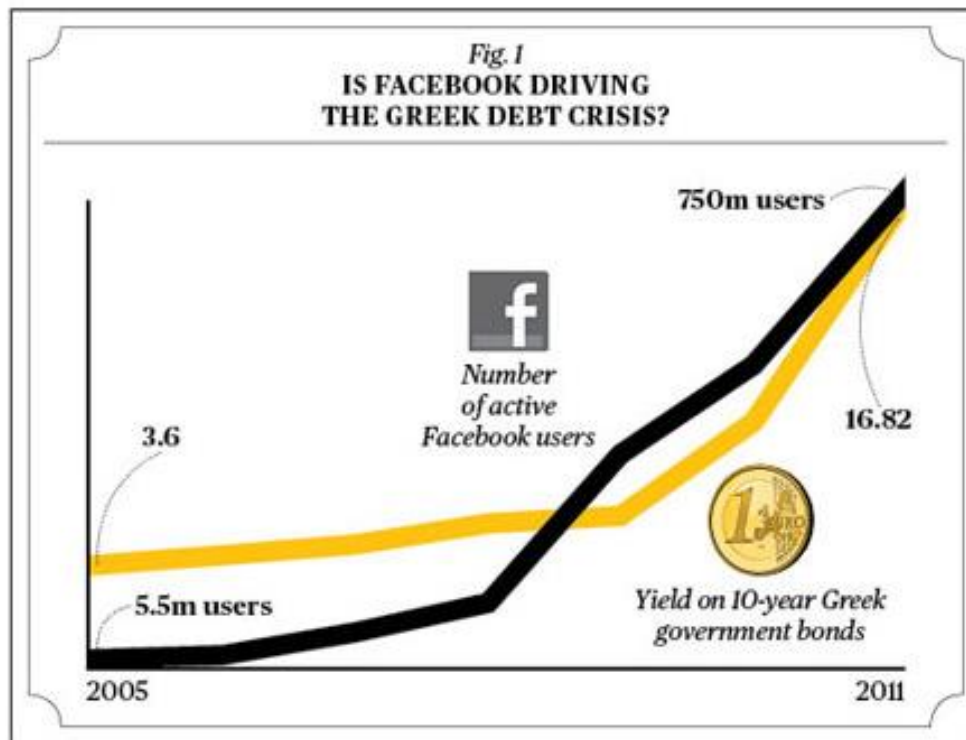


Statistical forecasting

- (2) Regression $x'(t + 1) = ay(t) + b$
 - The most popular method and many variations
 - x : predictand (e.g. rainfall at a station)
 - y : predictor (e.g. NINO3.4 SST)

Predict yield of Greek bonds with Facebook users

- Is it appropriate?



If yes, why?

If not, why?

From *business week*

Regression based forecast

- Question #1

$$x'(t + 1) = ay(t) + b$$

- How to define predictor (y)?
- By definition, predictor should cause some changes in variation of predictand
- Predictand : my mood in the morning
- Predictor?

Regression based forecast

- Question #2

$$x'(t + 1) = ay(t) + b$$

- How to define **a** and **b**?
- your choice. Linear, nonlinear, single, multi....
 - Complex one is not necessarily better.
- Predictand : my mood in the morning
- Predictor :
- a , b?

Regression based forecast

- Question #1 : Predictor selection
 - Should be based on Physical relationship between predictors and predictands
 - Predictor cannot be tiny signal in the seasonal forecast
 - Keep “doubt” on the possibility of selection by chance
 - Selected predictor should be validated with separate data

Regression based forecast

- Question #2 : appropriate Function

$$x'(t + 1) = ay(t) + b$$

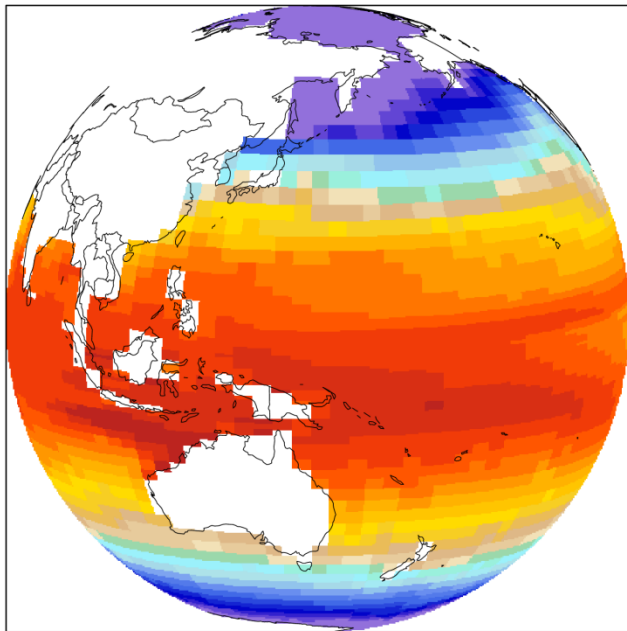
$$x'(t + 1) = a_1y_1(t) + a_2y_2(t)b$$

$$x_1'(t + 1) + x_2'(t + 1) = a_1y_1(t) + a_2y_2(t)b$$

- One to One : often not very satisfactory, limited cases
- One to Multi : easy to overfit (lie)
- Multi to Multi : looks nice but often produce nothing practical
- If they gives similar result, the simpler is the better

Dynamical forecast

- Use GCM : Global Climate Model
 - It used to be called “General Circulation Model”



Dynamical forecast

- Governing Equations
 - Written as computer program code (NWP)

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = \nabla \Phi - 2\Omega \times \mathbf{u} - \frac{1}{\rho} \nabla p + \mathcal{F}$$

$$\frac{\partial \rho}{\partial t} + \bar{\nabla}(\rho \bar{u}) = 0 \quad \Leftrightarrow \quad \frac{D\rho}{Dt} = -\rho \nabla \cdot \mathbf{u}$$

$$\frac{\partial \theta}{\partial t} + \bar{u} \cdot \bar{\nabla} \theta = l$$



```
//Behrdeh functions:
//
//   DTstage(T+9.11)^-2.05
//
MinTime=pow(Tfj]+9.11, -2.05); //Minimum time to advance to stage (in days)
for(k=0;k<numLifeStage;k++)
{
    MaxRate[k]=MinTime*DTstage[k];
    MaxRate[k]=MaxRate[k]*ToSecs; //Convert to seconds
    MaxRate[k]=1.0/MaxRate[k]; //Convert to rate
}

//Parameters for Ivlev functions controlling food dependence
//
//   R=a[1-exp[-b*(food-c)]]--development rate (days^-1)
//
// But, idea is that temp sets max growth rate, and food tells us how close
// we get to the max. In this sense, a=1 (Campbell figured an absolute
// rate, we're essentially normalizing his rates by rate at 40C.
//
//b=[ones(1,6)*params.bnaup,ones(1,6)*params.bcop];

for(k=0;k<6;k++)
    Rfood[k]=[1.-exp[-(Ffj]-c)*params.bnaup]];
for(k=6;k<12;k++)
    Rfood[k]=[1.-exp[-(Ffj]-c)*params.bcop]];

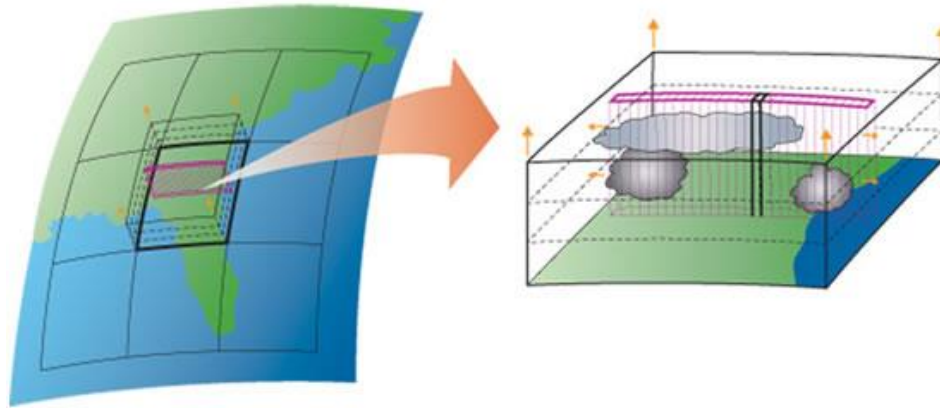
//Multiply Rfood by MaxRate to get the actual rate.
for(k=0;k<12;k++)
    R[k]=MaxRate[k]*Rfood[k];

R[12]=0.; //adults don't molt

//M[k]=mortality rate for stage k at node j
//
gammaT=gamma0*(1.-gamma0)*pow(Tfj]/Tc,z);
//gammaT=0.1; //Override temp dependent mortality
```

Numerical modeling

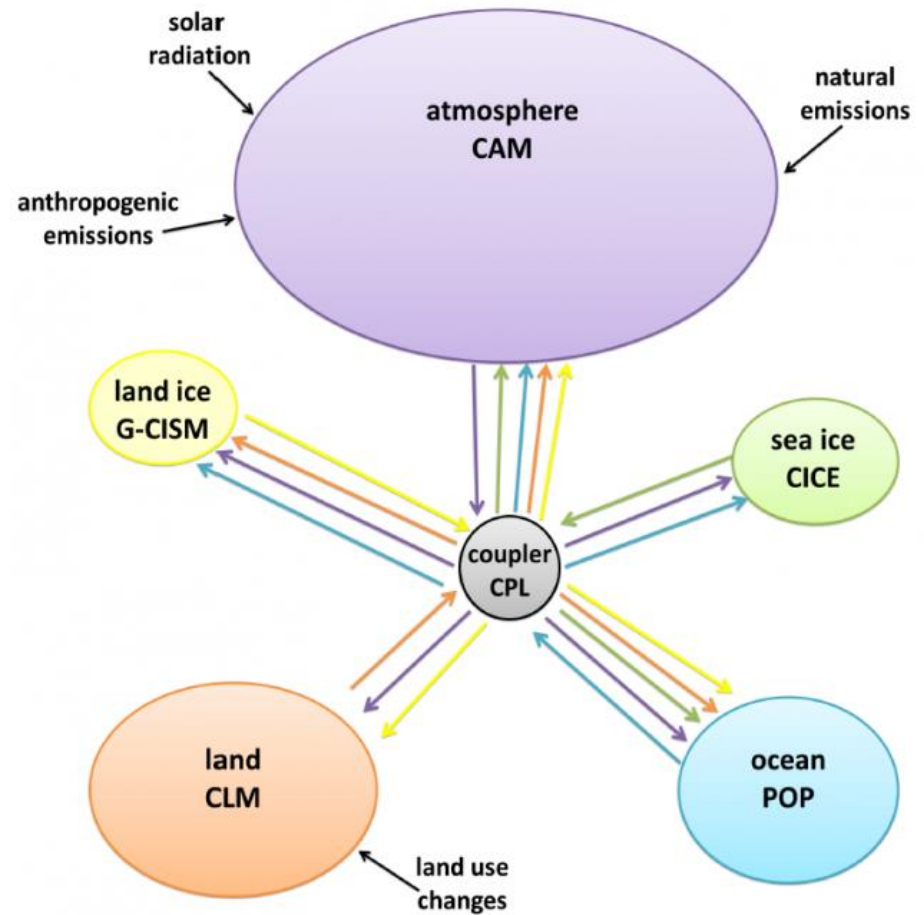
- Issue
 - Digitization (physical variable is continuous, but computer needs digitization”
 - Resolution, subgrid-scale parameterization



- Unknown processes, tunable parameters
- Initialization (for forecasting)

GCMs

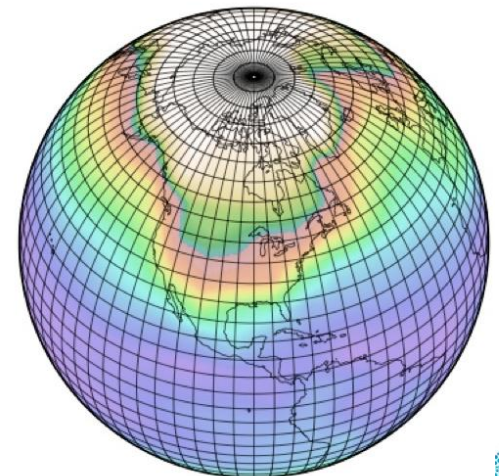
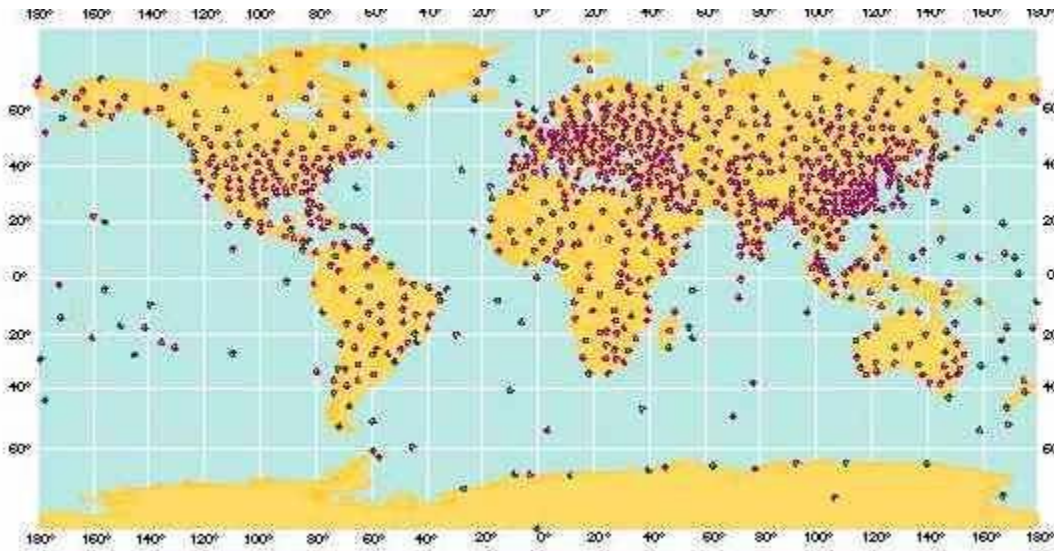
- Coupled GCM
 - Atmosphere
 - Ocean
 - Sea-Ice
 - Land surface
 - Chemistry
 - Biosphere



Initialization

Estimating Current status of climate system

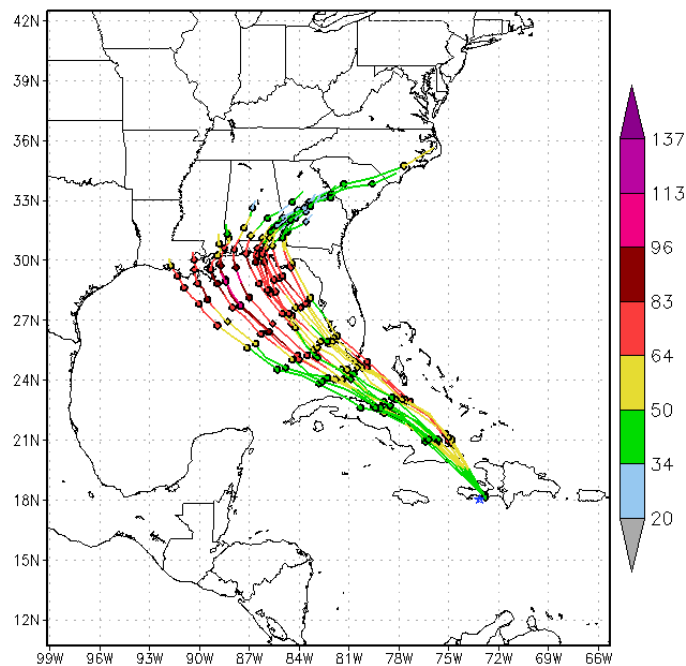
- Preparing the beginning climate state of GCM with available observation
 - Balance between Wrong GCM vs Wrong OBS.
 - Balance between components (Atm, Ocn)



Ensemble Forecasting

- Run many times
 - Starts from slightly different initial conditions

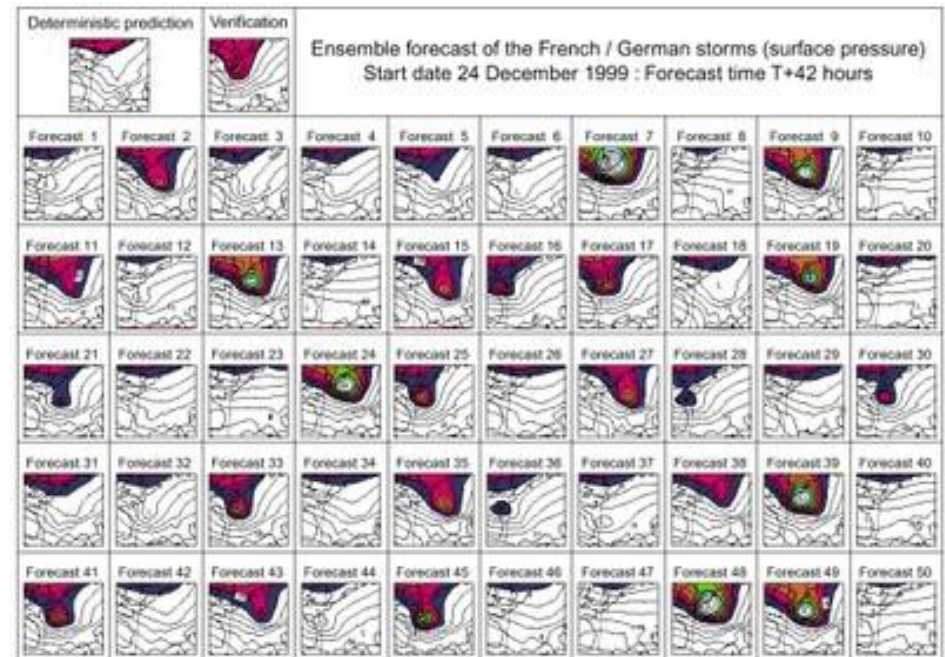
6-hourly Track and Intensity (kt) for ISAAC09L
GFDL ensemble forecast for the 126 hrs from 06Z25AUG2012



of missing members (out of 16) at t=0: 0
indicates ISAAC09L observed center at initial time

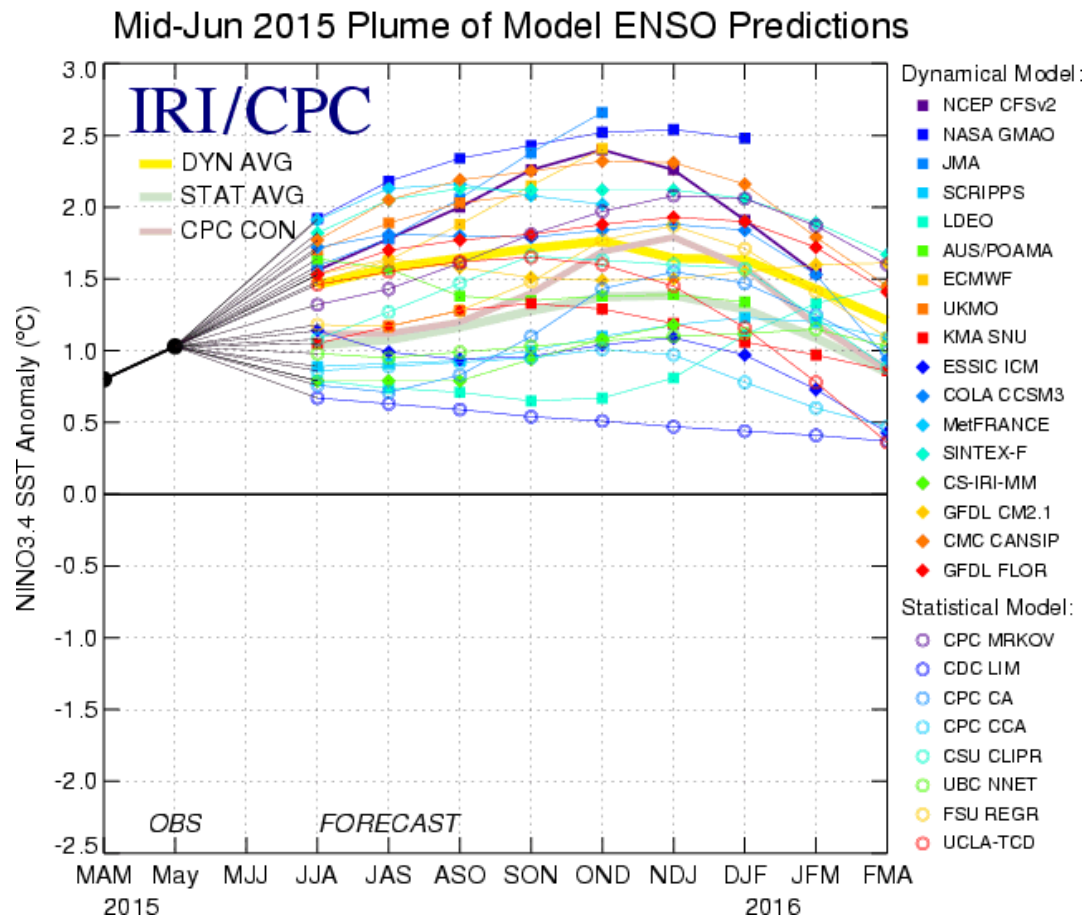
Track forecast positions are marked every 12 hrs

GFDL Hurricane Dynamics Group



Multi Model Ensemble Forecasting

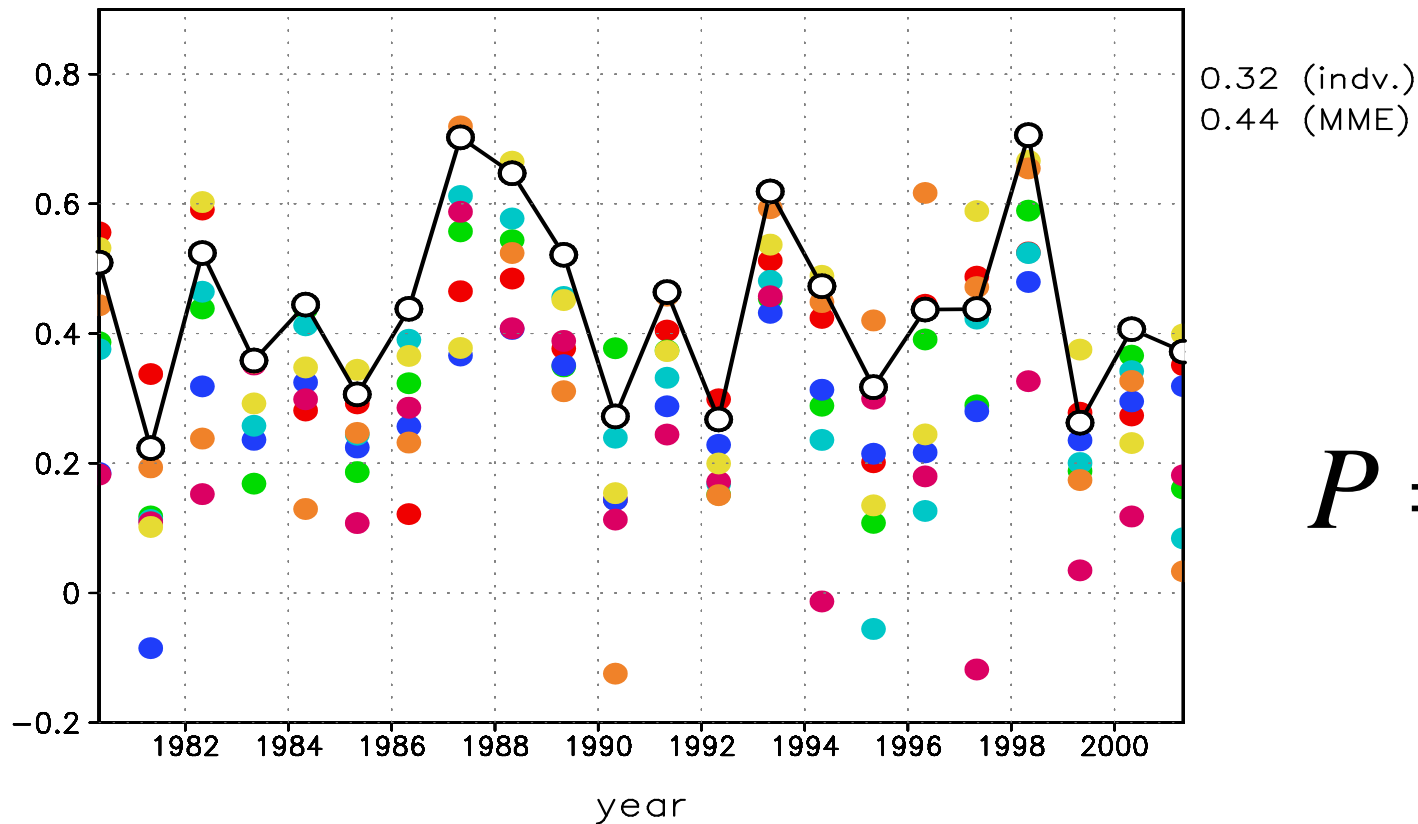
- Run with many models



Which one??

Use all!

Pattern correlation : summer monsoon precip.



$$P = \sum_i a_i F_i$$

Predictability of Multi Model Ensemble

Correlation skill of a single model

$$R_i = \frac{\overline{xy_i}}{\sqrt{V(x)V(y_i)}}$$

Correlation skill of MME

$$\langle y \rangle = 1/M \sum_{i=1}^M y_i$$

$$R_{MM} = \frac{\overline{x\langle y \rangle}}{\sqrt{V(x)V(\langle y \rangle)}} = \frac{1}{M} \sum_{i=1}^M \left(R_i \sqrt{\frac{V(y_i)}{V(\langle y \rangle)}} \right) = \langle R \rangle \sqrt{\frac{\langle V(y) \rangle}{V(\langle y \rangle)}}$$

$$\langle R \rangle = \frac{1}{M} \sum_i R_i$$

$$V(\langle y \rangle) = \langle V_{Single} \rangle - \frac{M-1}{M} \langle V(y_n) \rangle - \frac{M-1}{M} \langle (V(e) - C(e)) \rangle$$

$$R_{MM} = \frac{\langle R \rangle}{\sqrt{V(\langle y \rangle)}} = \frac{\langle R \rangle}{\sqrt{\langle r \rangle}}$$

$$\langle r \rangle = \frac{1}{M^2} \sum_i \sum_j \frac{\overline{y_i y_j}}{V}$$

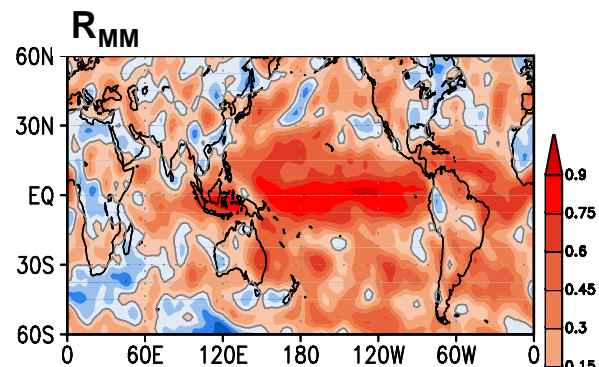
$$E_{MM} = \langle V_{Single} \rangle (1 + \langle r \rangle - 2\langle R \rangle)$$

Observation : $x = x_s + x_n$

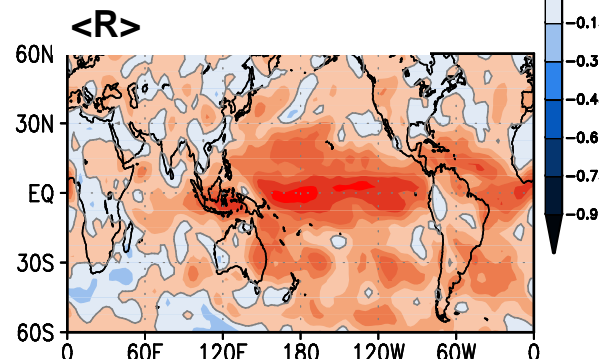
Forecast : $y = y_s + y_n = x_s + e + y_n$

Temporal correlation skill (SUMMER MEAN PRCP)

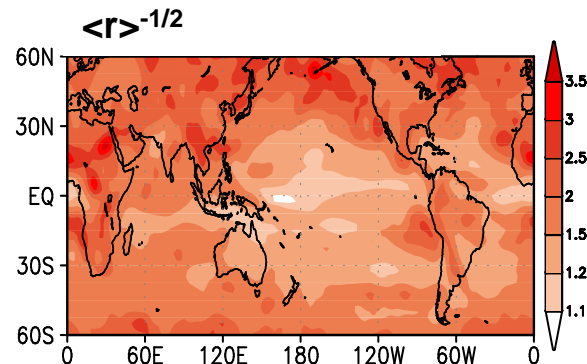
Multi-model ensemble correlation skill



Mean correlation skill of individual models

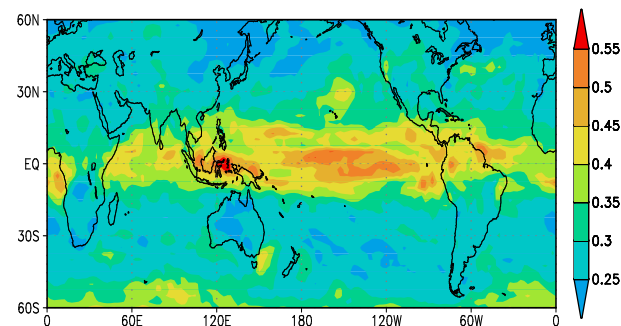


Inflation factor of correlation skill by multi-model ensemble



$$V(\langle y \rangle) = V_{Single} - \frac{M-1}{M} \langle V(y_n) \rangle - \frac{M-1}{M} \langle V(e) - C(e) \rangle$$

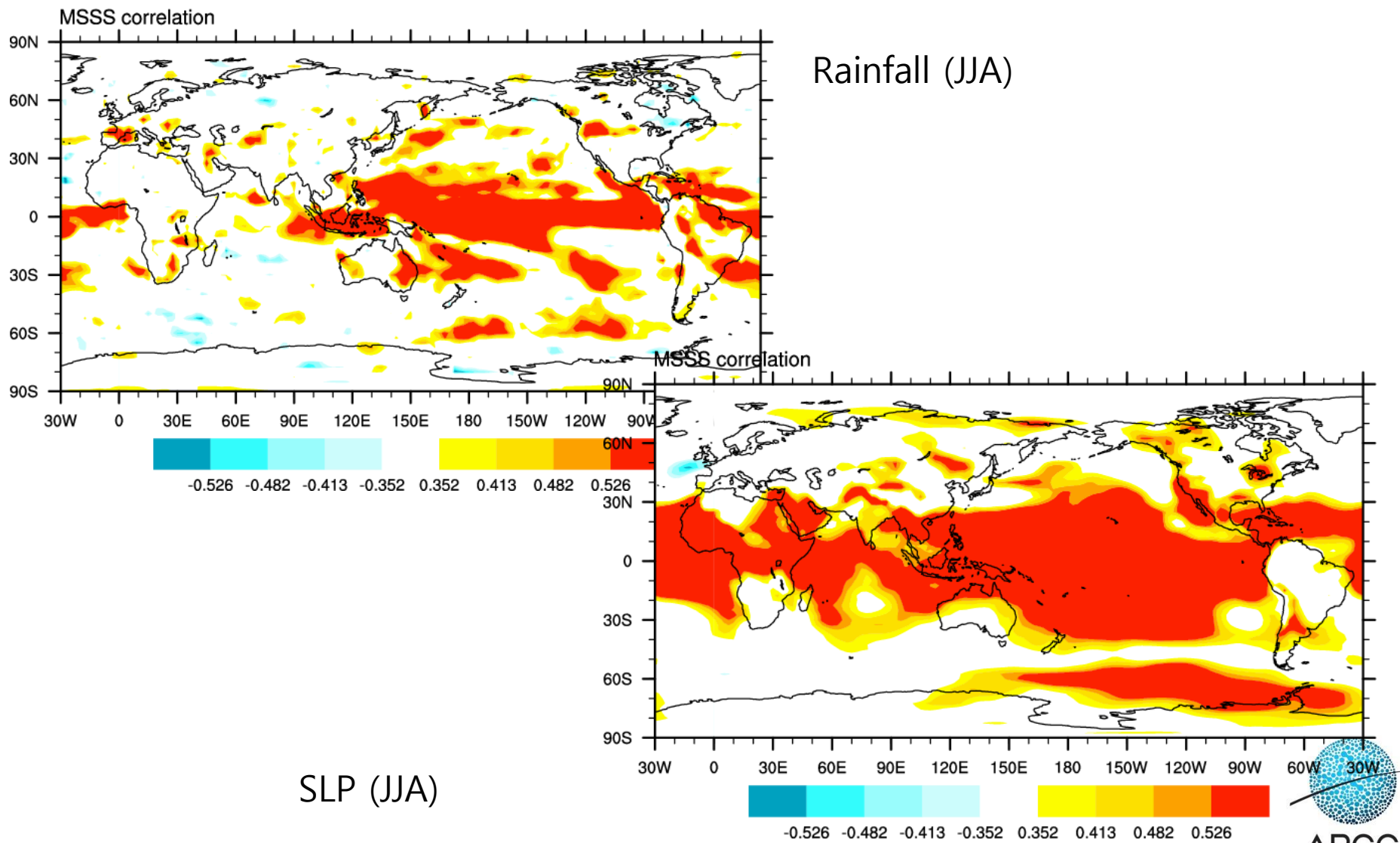
Contribution of systematic error (conditional) cancellation



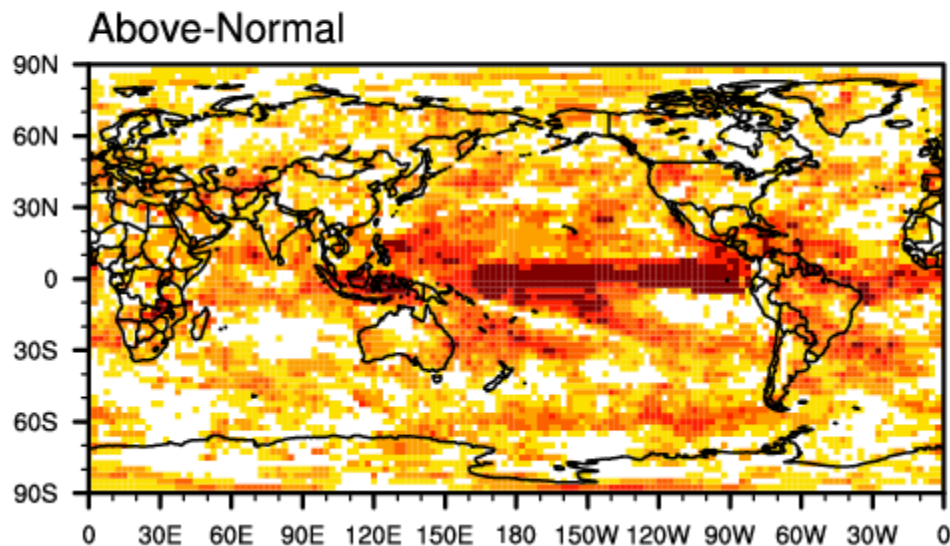
$$R_{MM} = \frac{\langle R \rangle}{\sqrt{V(\langle y \rangle)}} = \frac{\langle R \rangle}{\sqrt{\langle r \rangle}}$$

Independent and good models :
Best forecast result (on average)

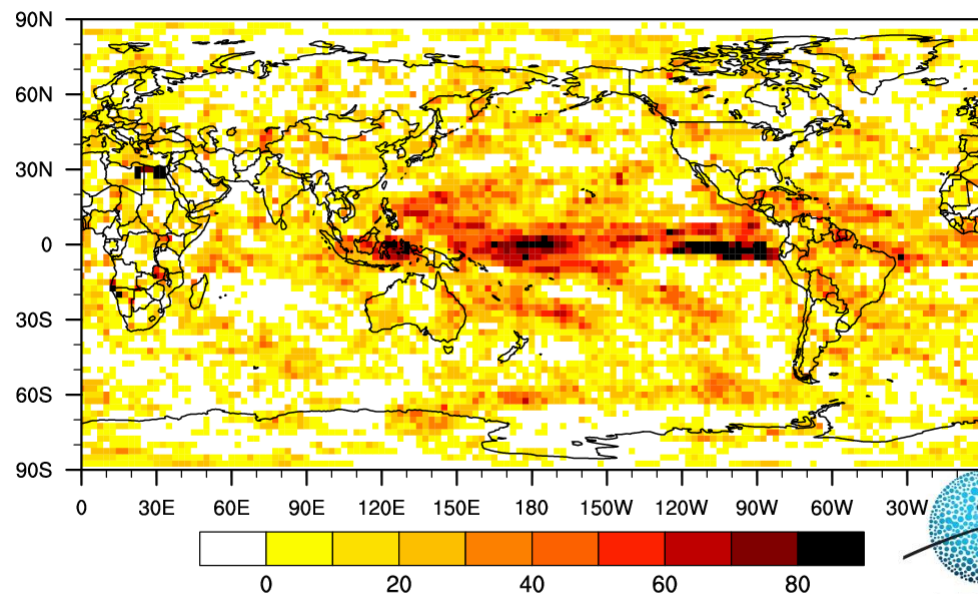
APCC MME (TCC)



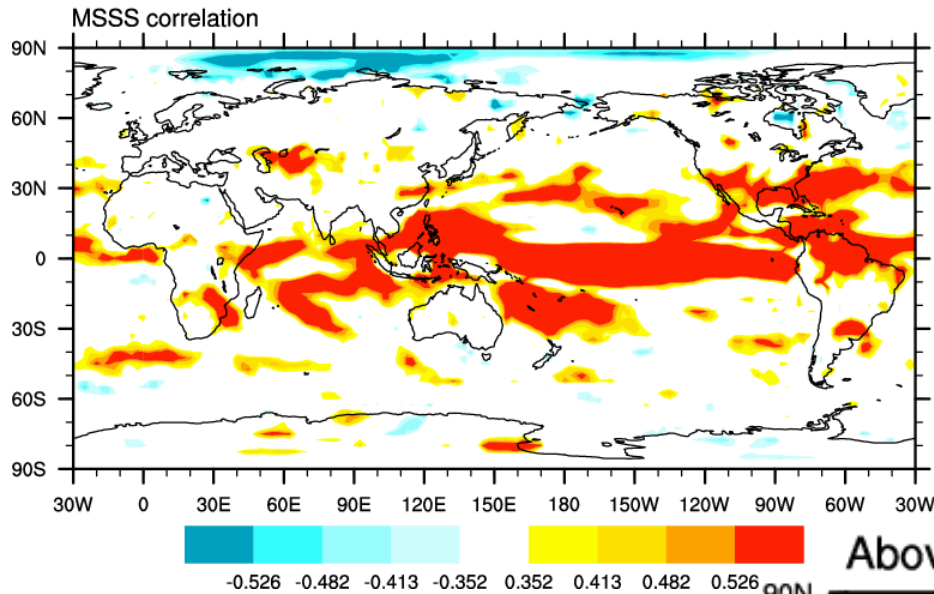
ROC Score : PREC, JJA (1983-2005)



Heidke Skill Score : PREC, JJA (1983-2005)



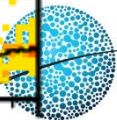
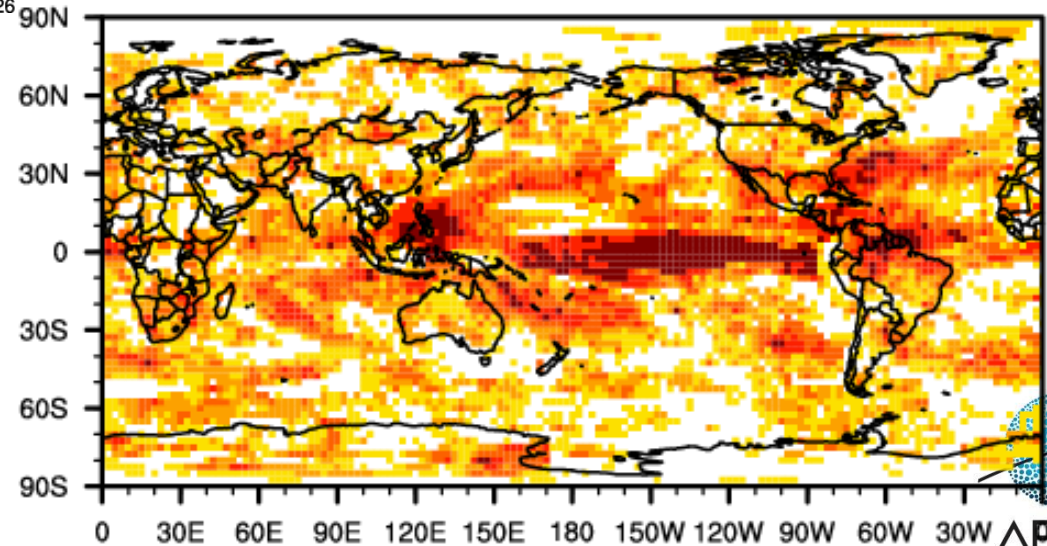
DJF season



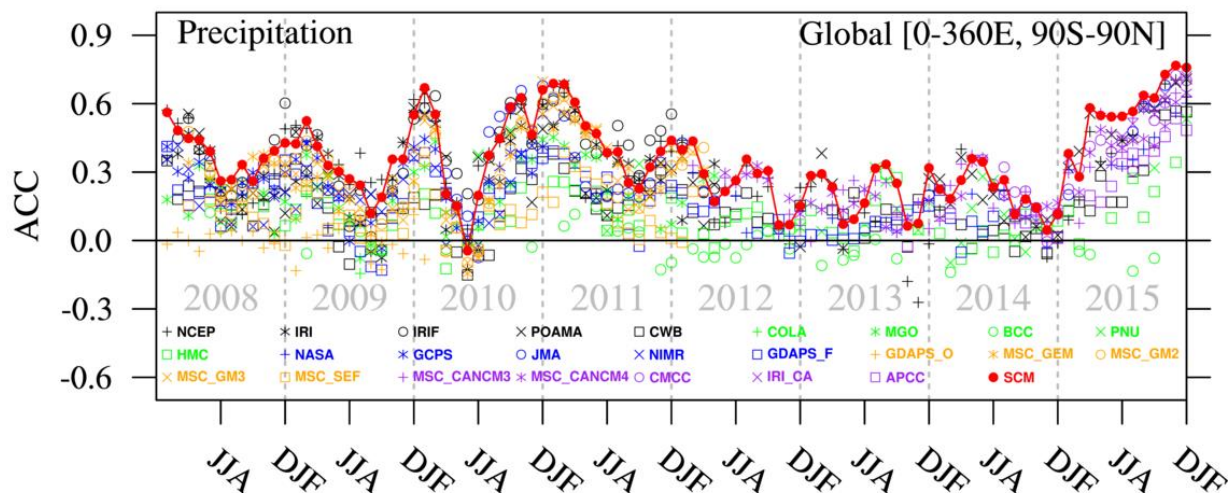
Rainfall (JJA)

ROC score
Rainfall (DJF)

Above-Normal



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

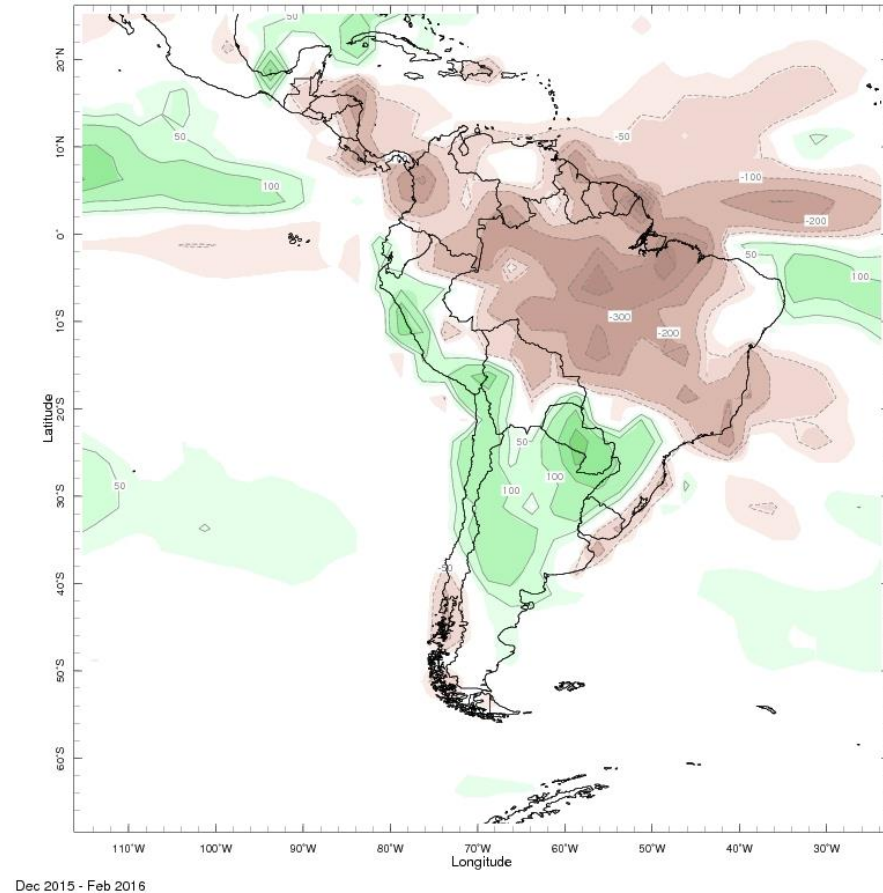
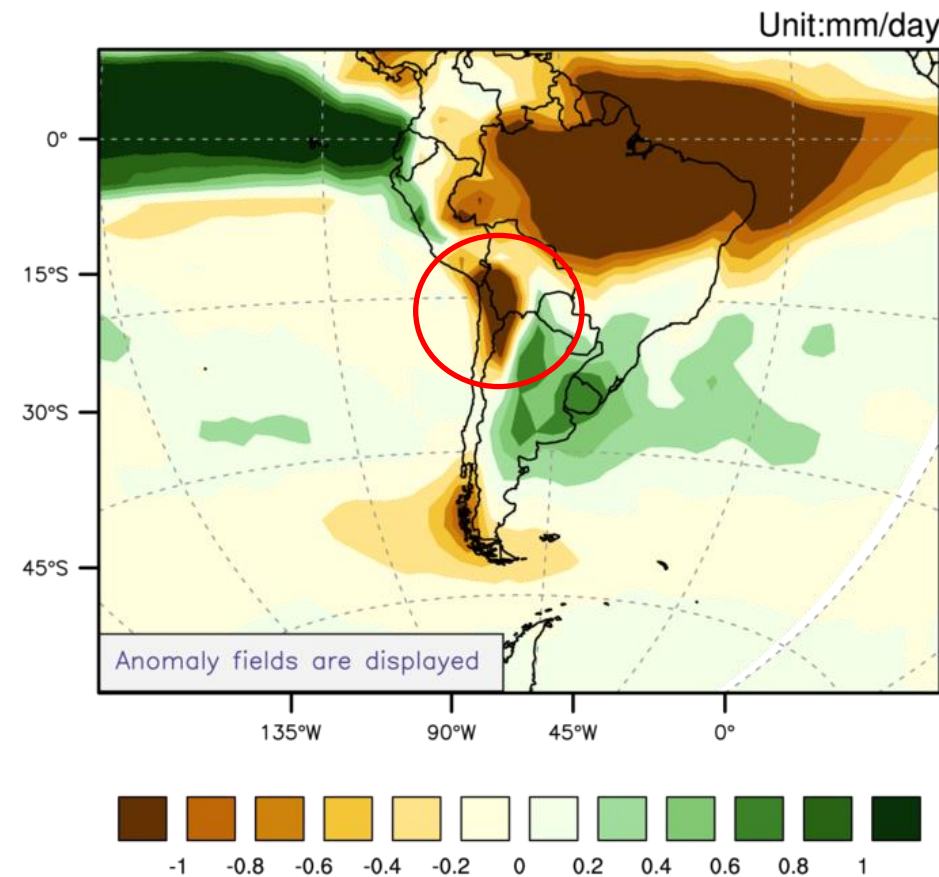
Realtime **rainfall** forecast for last 4 years (12-15)
 ROC score : Perfect = 1, Meaningless(no skill) =0.5,

Even with MME,

- Still many region in the world, predictability is low
- Any room for further improvement?
 - Post process

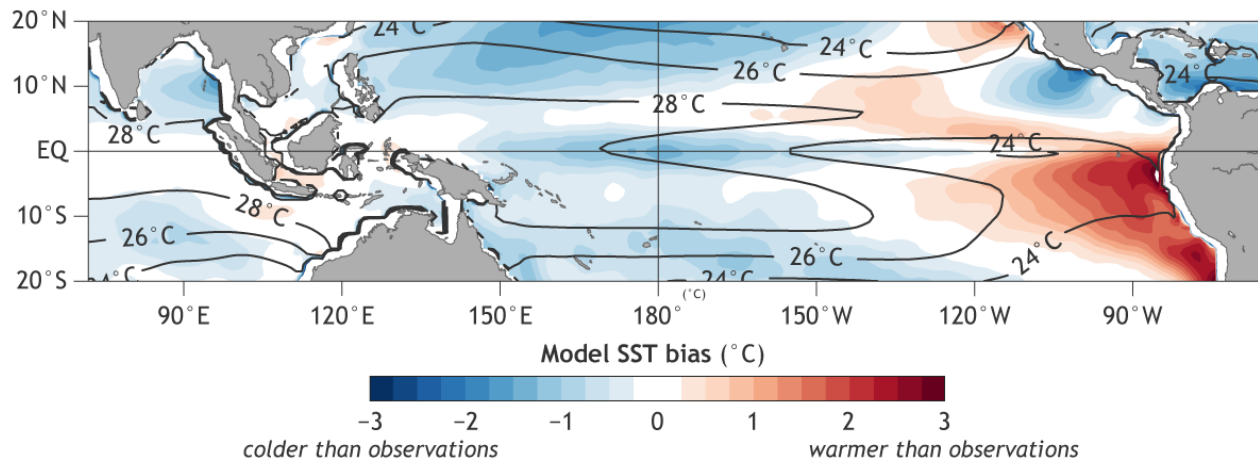
Challenges in South America

Precipitation for December 2015-February 2016



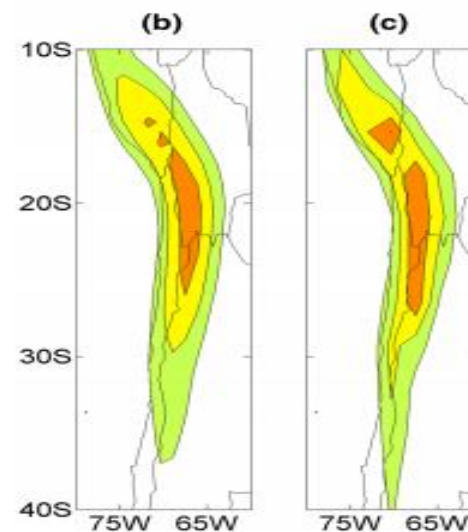
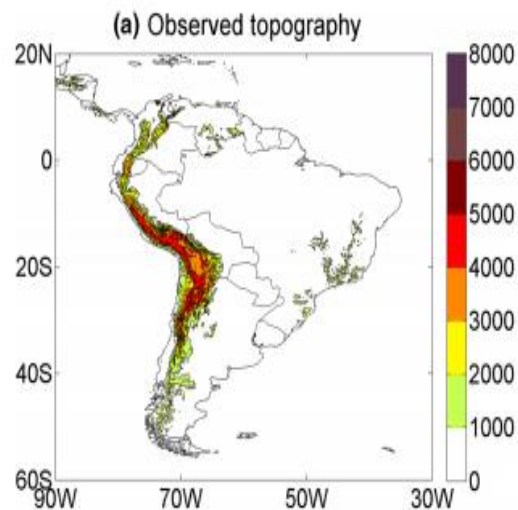
Models have biases

Modeled sea surface temperature and bias



SST

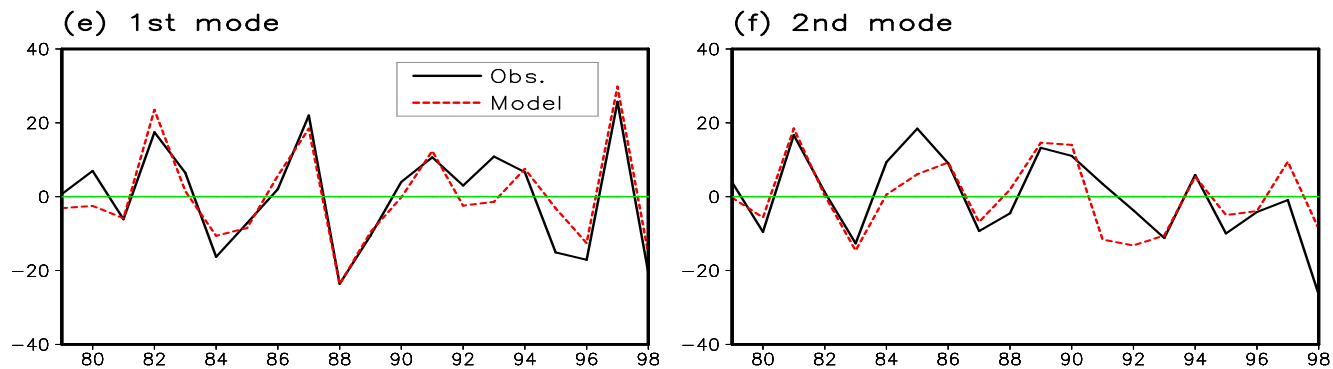
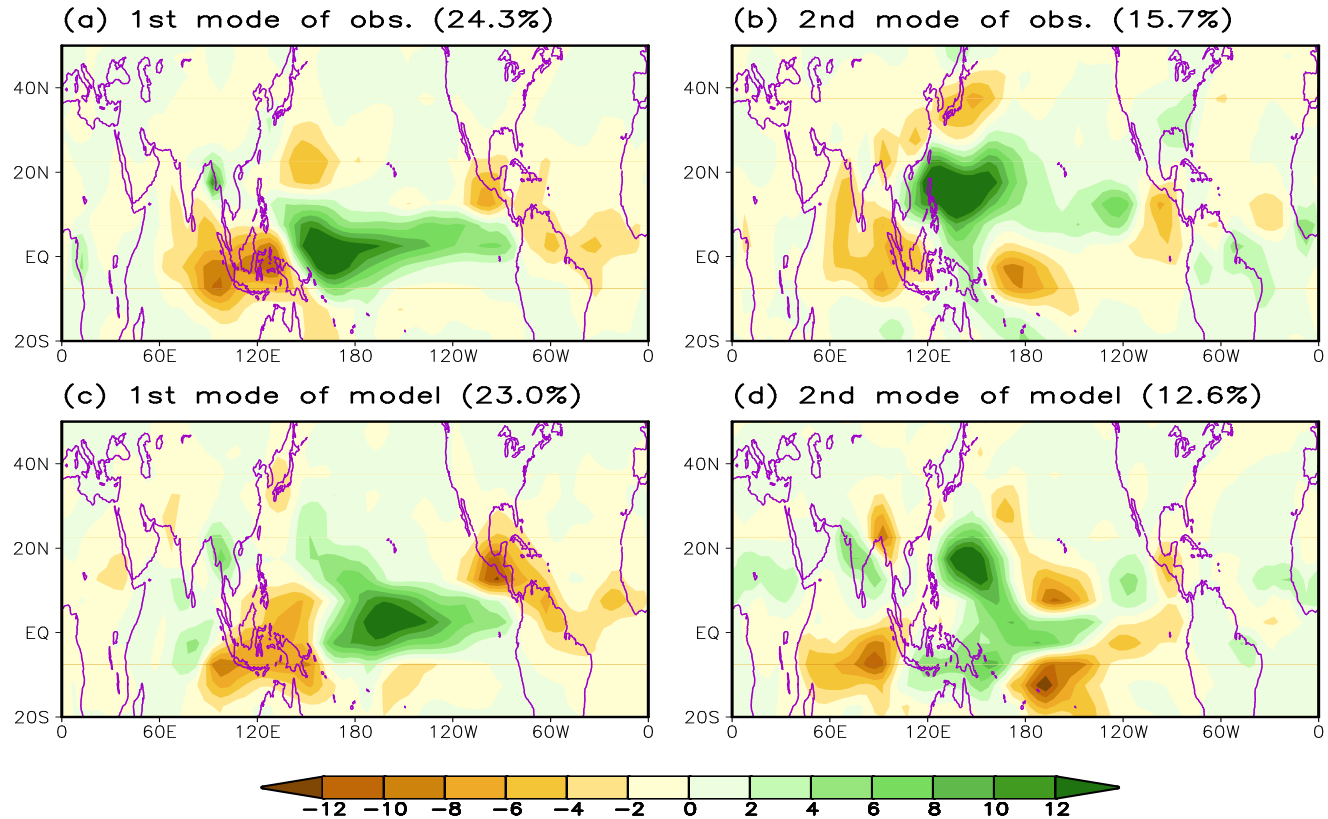
NOAA Climate.gov



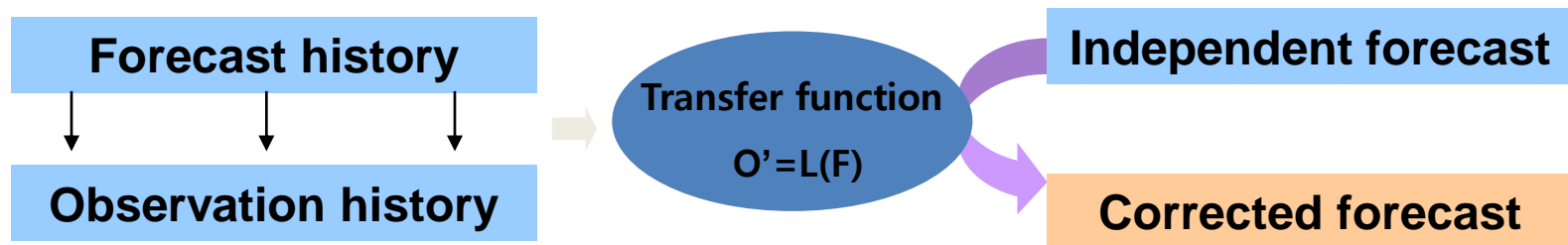
Orography

Saurrel et al. (2015)

EOFs of Summer Mean Precipitation



Statistical downscaling : CLIK



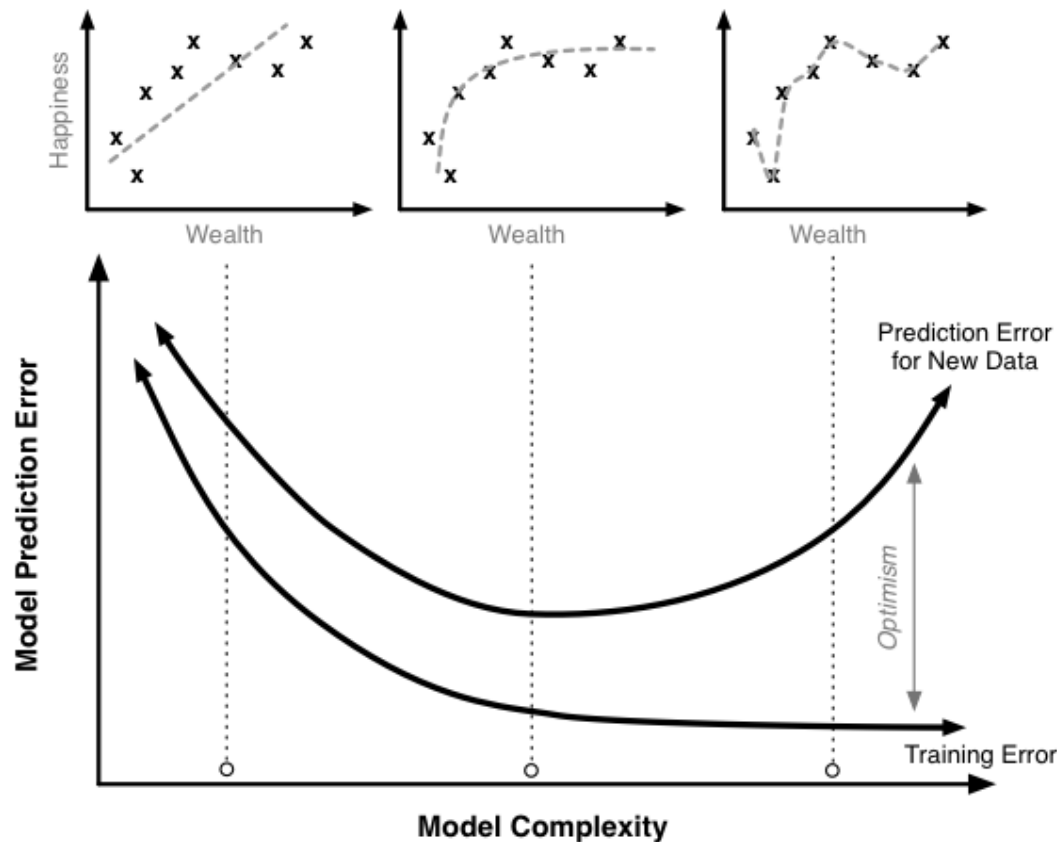
There are many approaches in post-process, All of them share similar assumption. :
Statistics between forecast and observation is stationary

If statistics is not stationary, post-process will not work in independent forecast

Thus, statistical stability is a rule of thumb in the statistical post-process (avoiding overfitting)

Weakness : overfitting

■ Consider potential predictability



If model output is fitted to the unpredictable noise : Overfitting.
What if we remove “noise” in the observation?

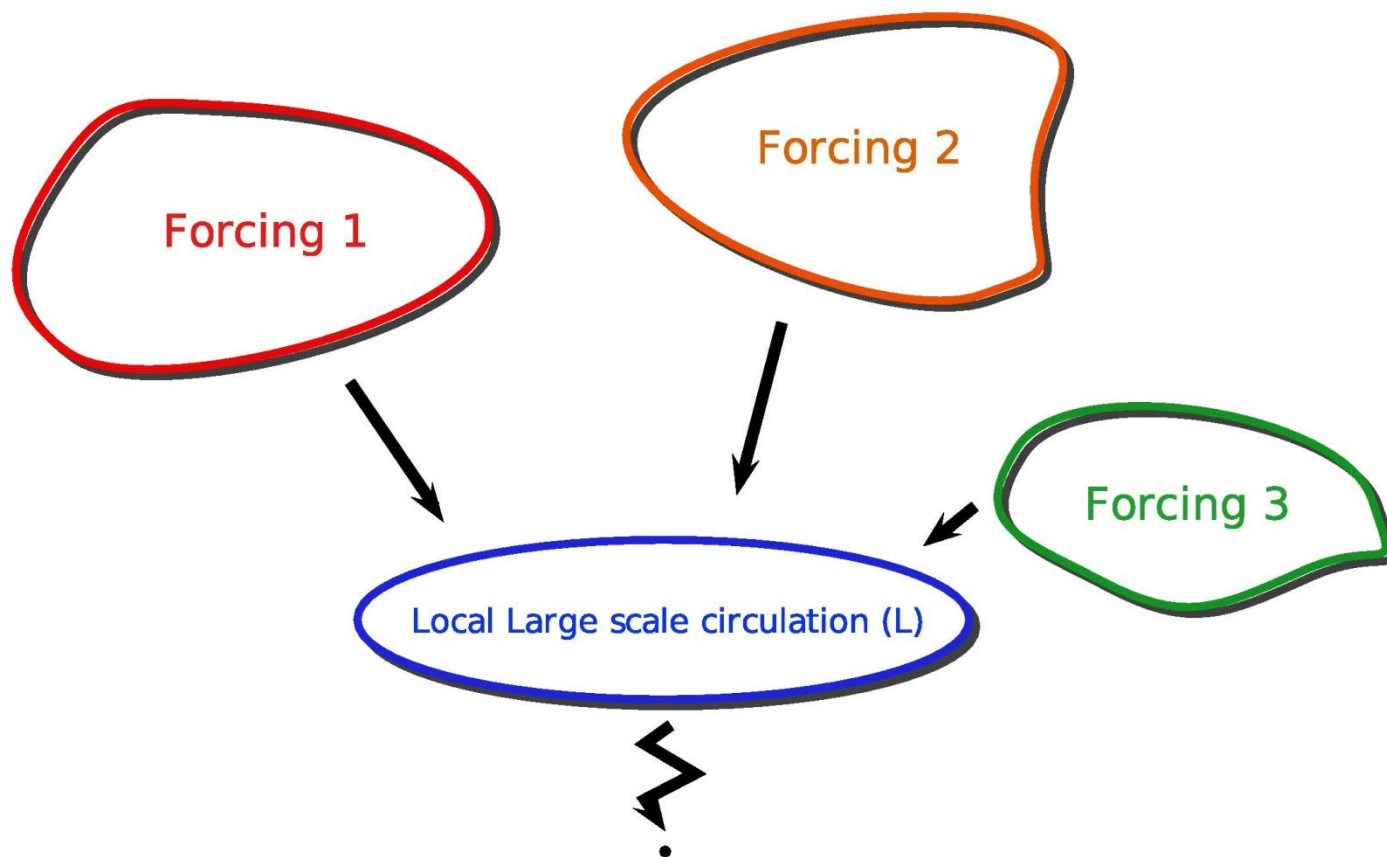
from Scott Fortmann-Roe 2012

Downscaling (post-process) should be based on

- Physical understanding of;
 1. What weather event/system consists of your seasonal climate (LOCAL, predictand)
 2. What external (slow varying factor) controls the weather system (GLOBAL, predictor)

And, whether model is able to predict 1 or 2

Local large scale circulation



Local weather statistics (Korean summer rainfall)

Local large scale circulation

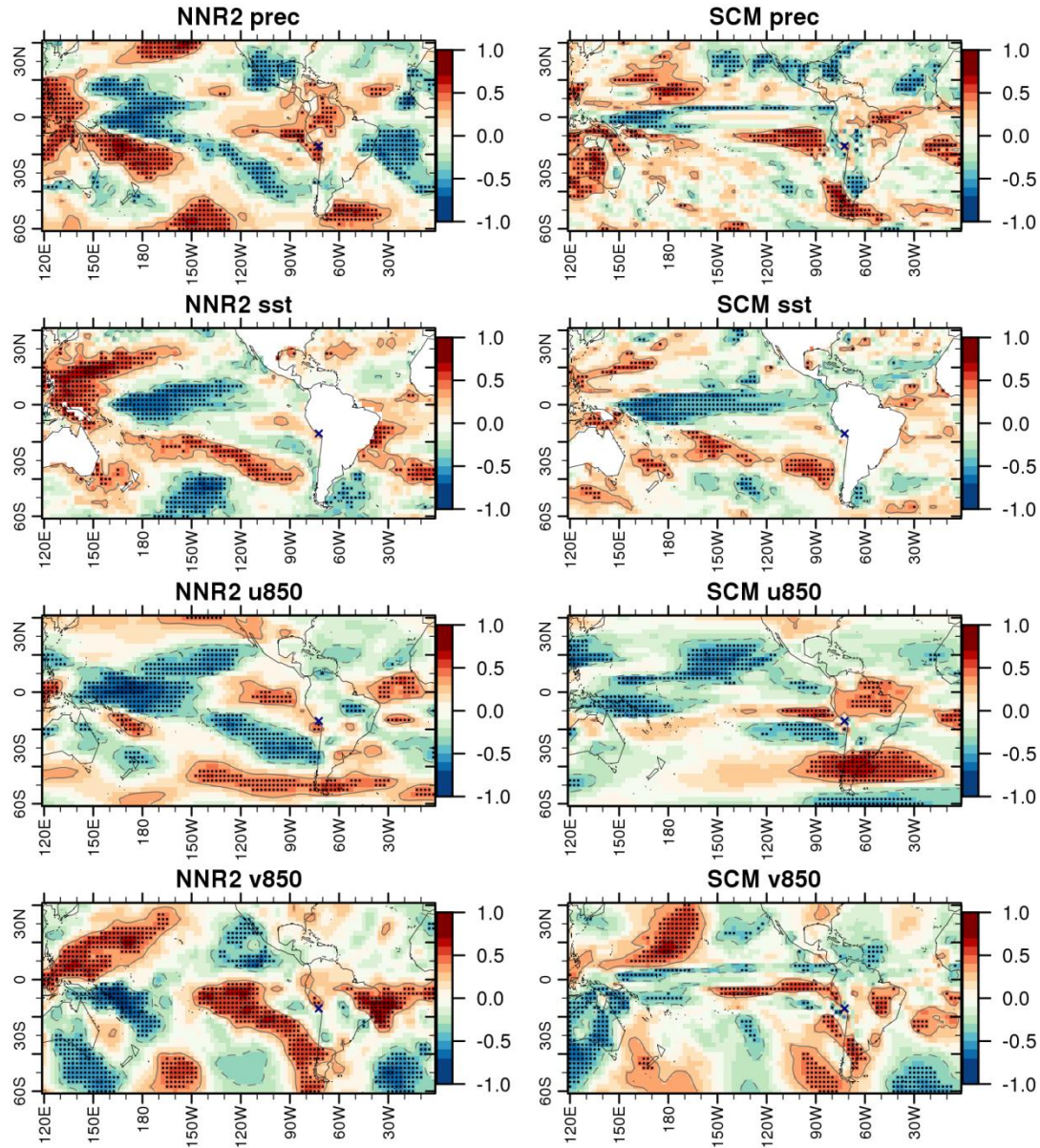
- Local climate (i.e. seasonal mean) is defined by how weather behaved during a season (statistics)
- Therefore, understanding weather behavior is the first step of seasonal forecast (often ignored..)
- In many cases, local large scale pattern that directly affect local weather is visible in seasonal time scale
 - Question is whether we can predict that large scale pattern directly or via teleconnection

LARGE SCALE PATTERN ASSOCIATED WITH RAINFALL

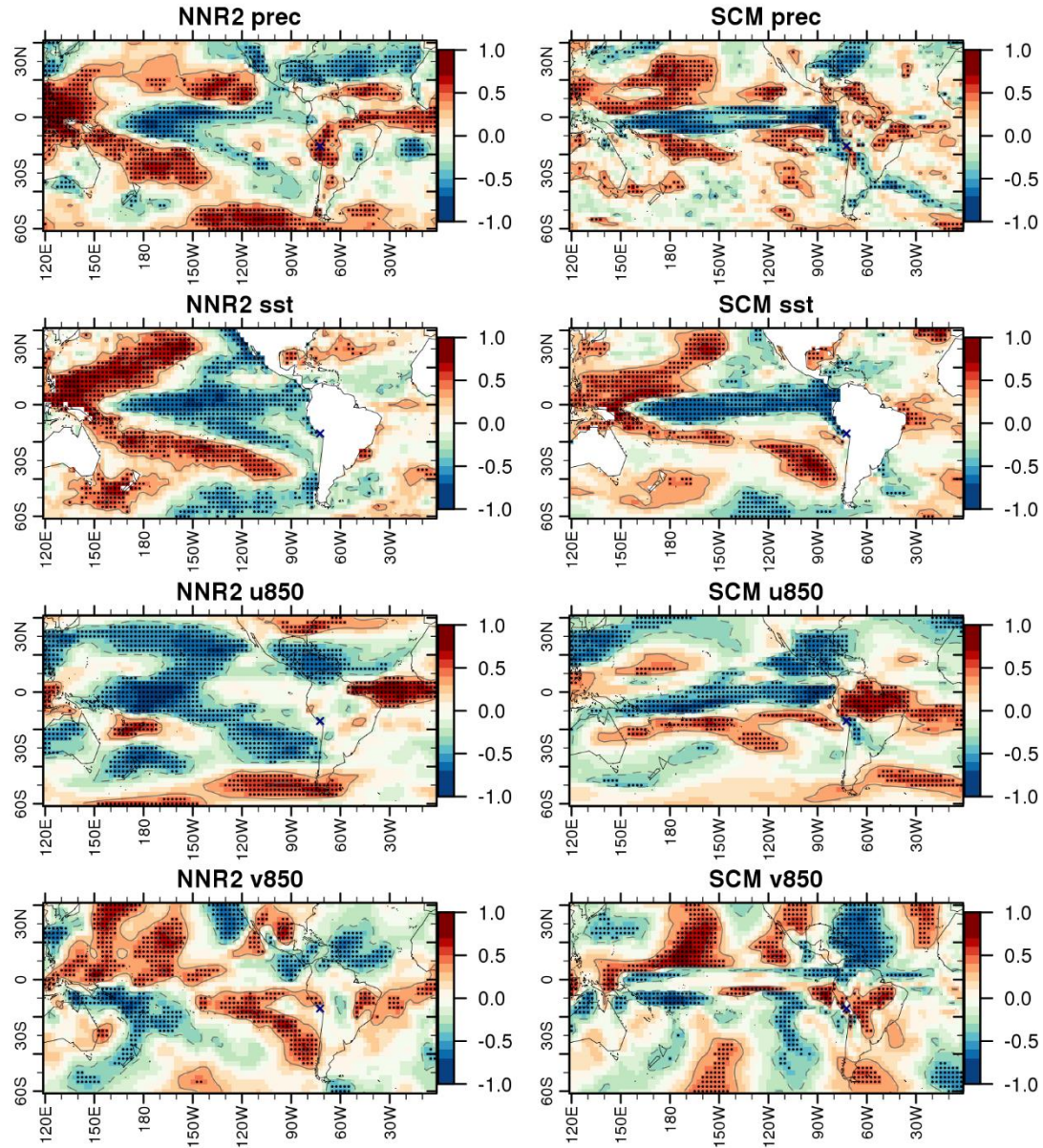
Local large circulation and Teleconnection

One Point Correlation map with seasonal mean local rainfall with other variables

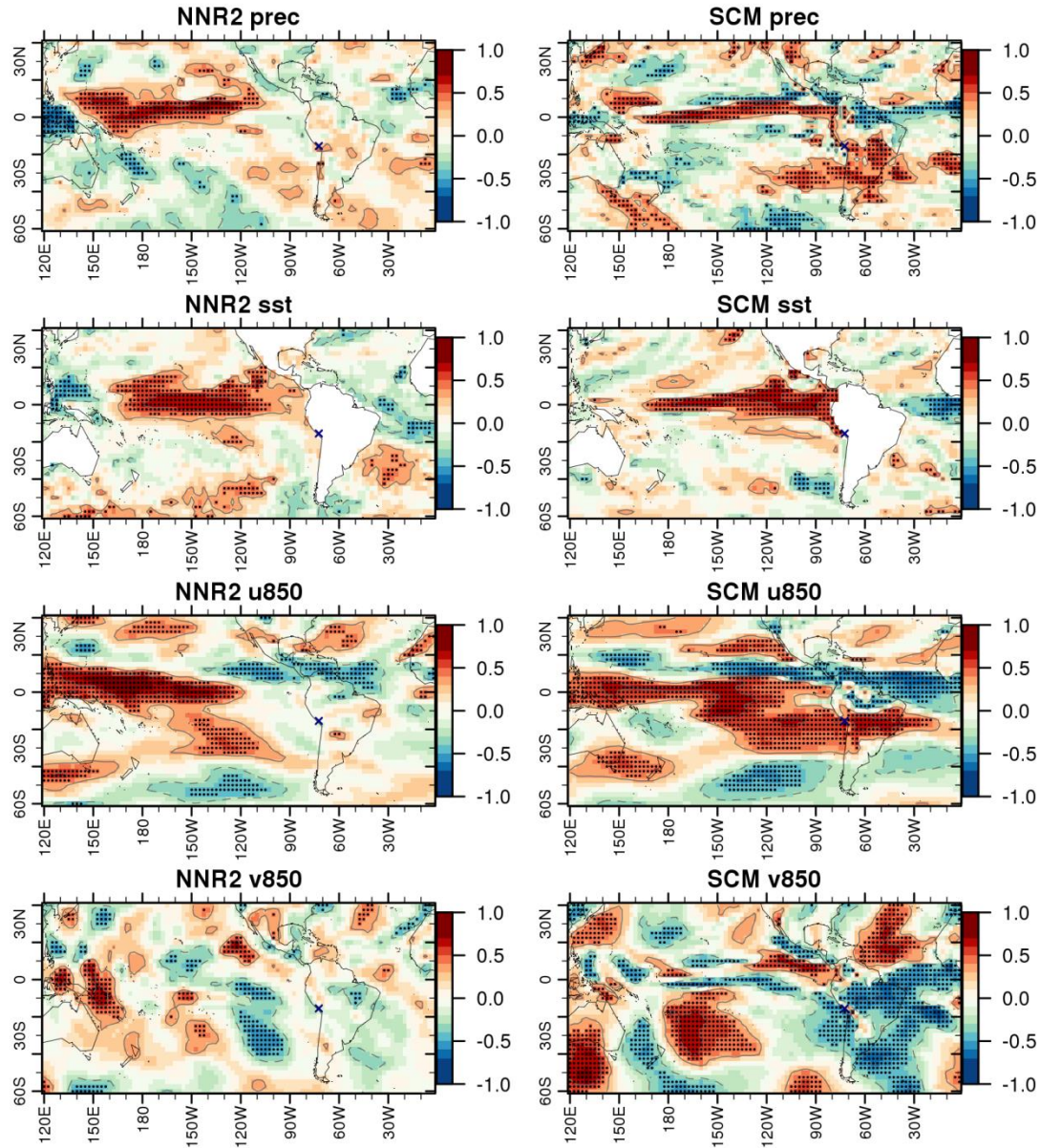
andahua [DJF]



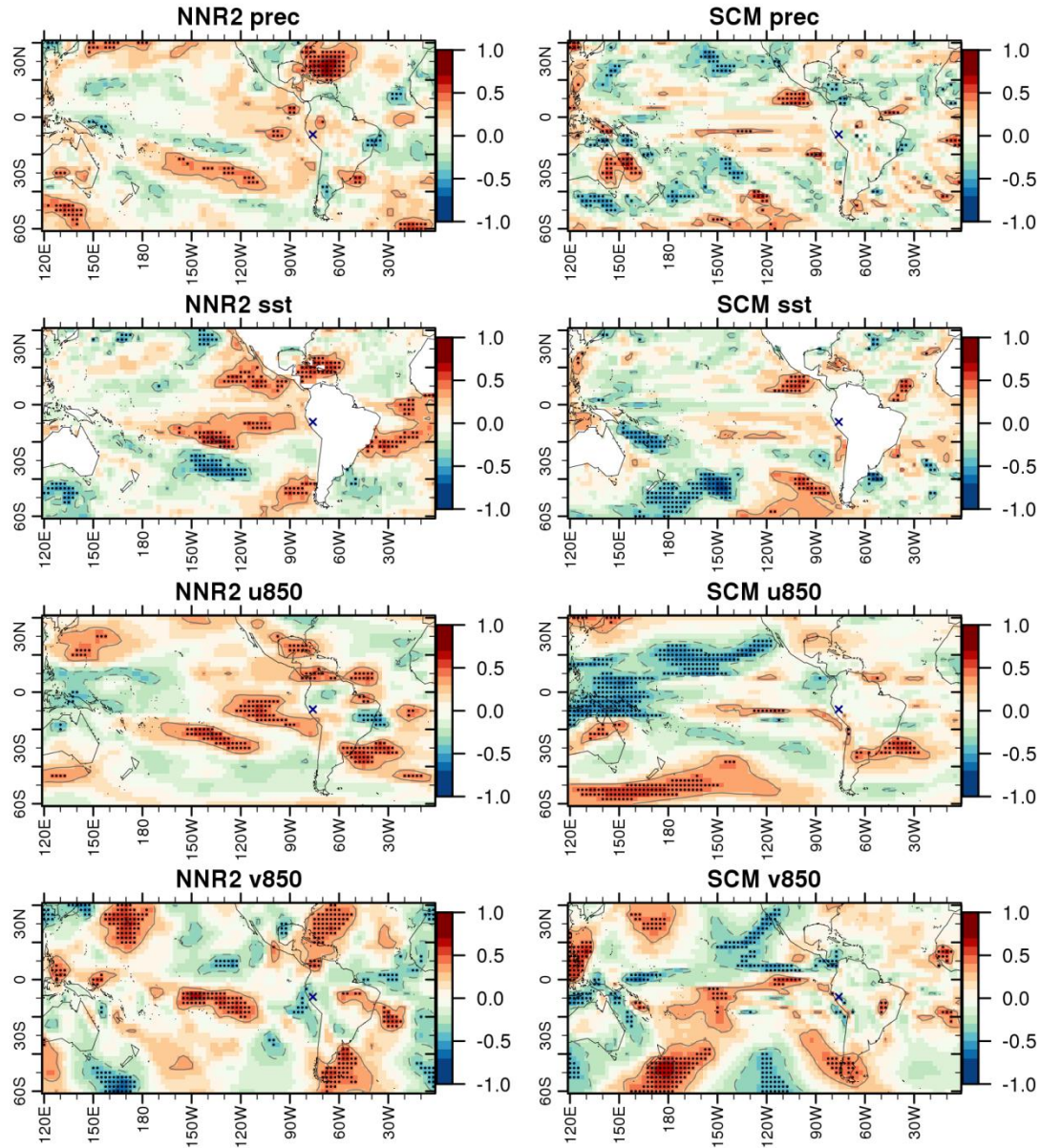
andahua [FMA]



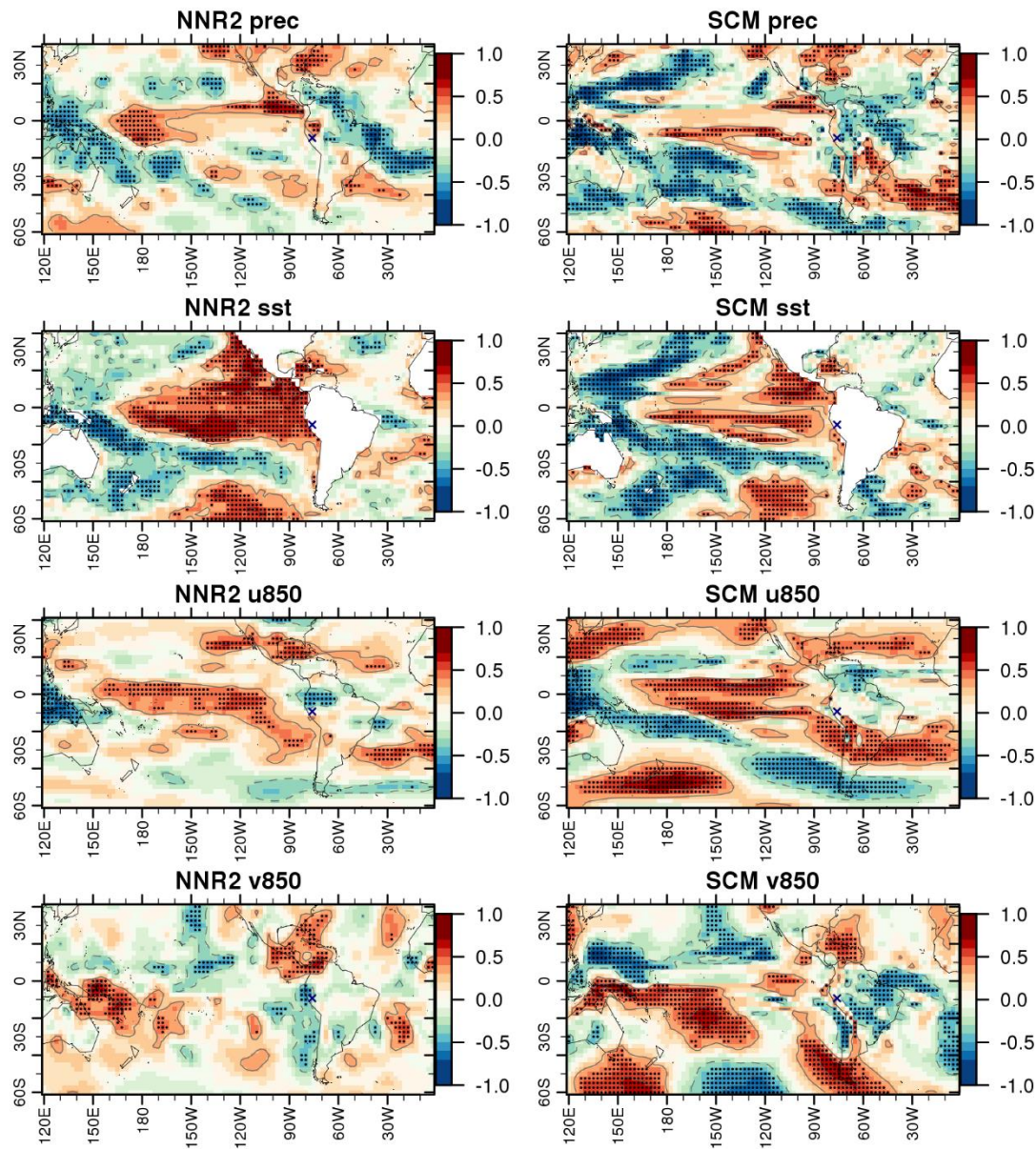
andahua [JJA]



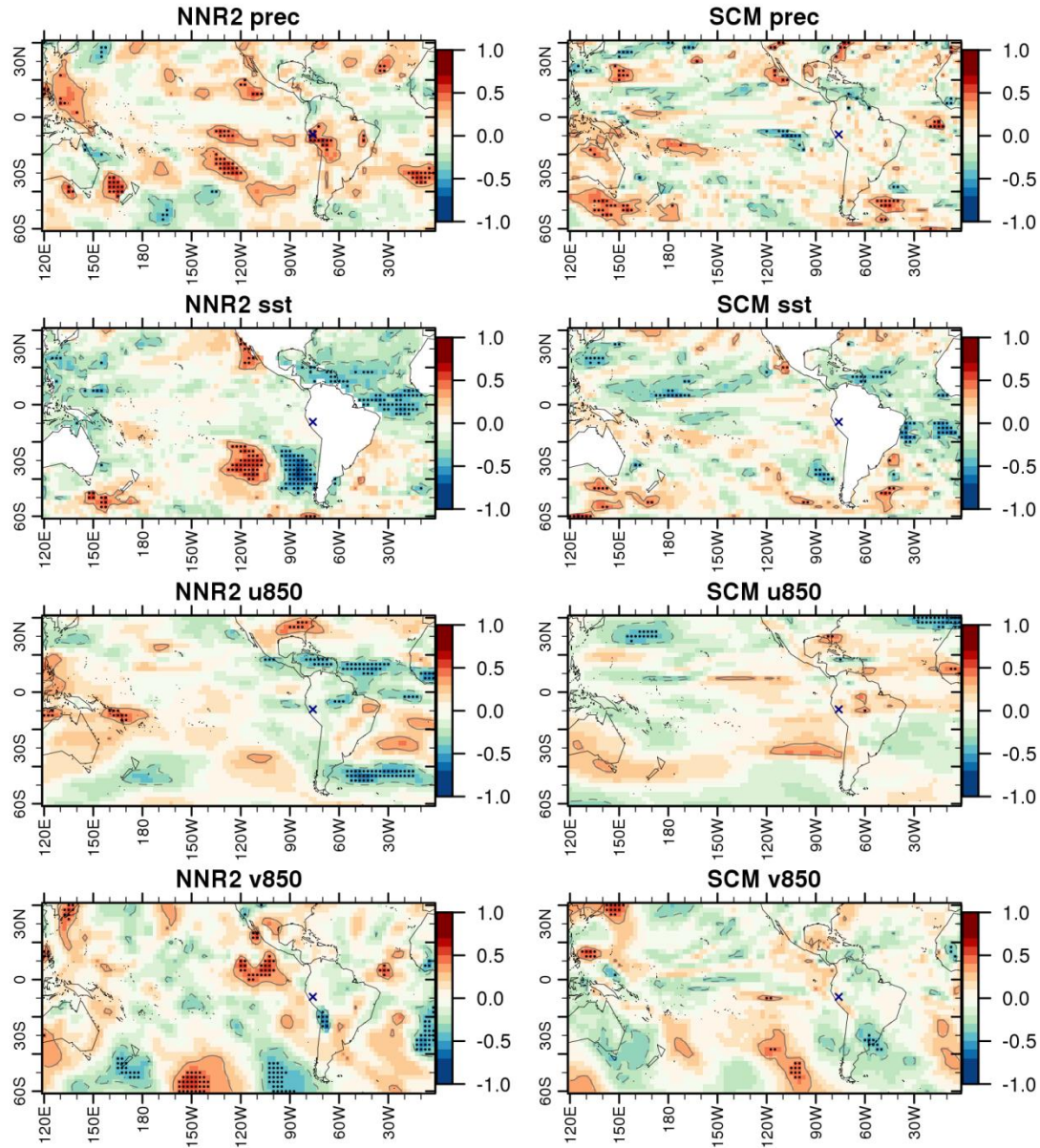
Tingo Maria [DJF]



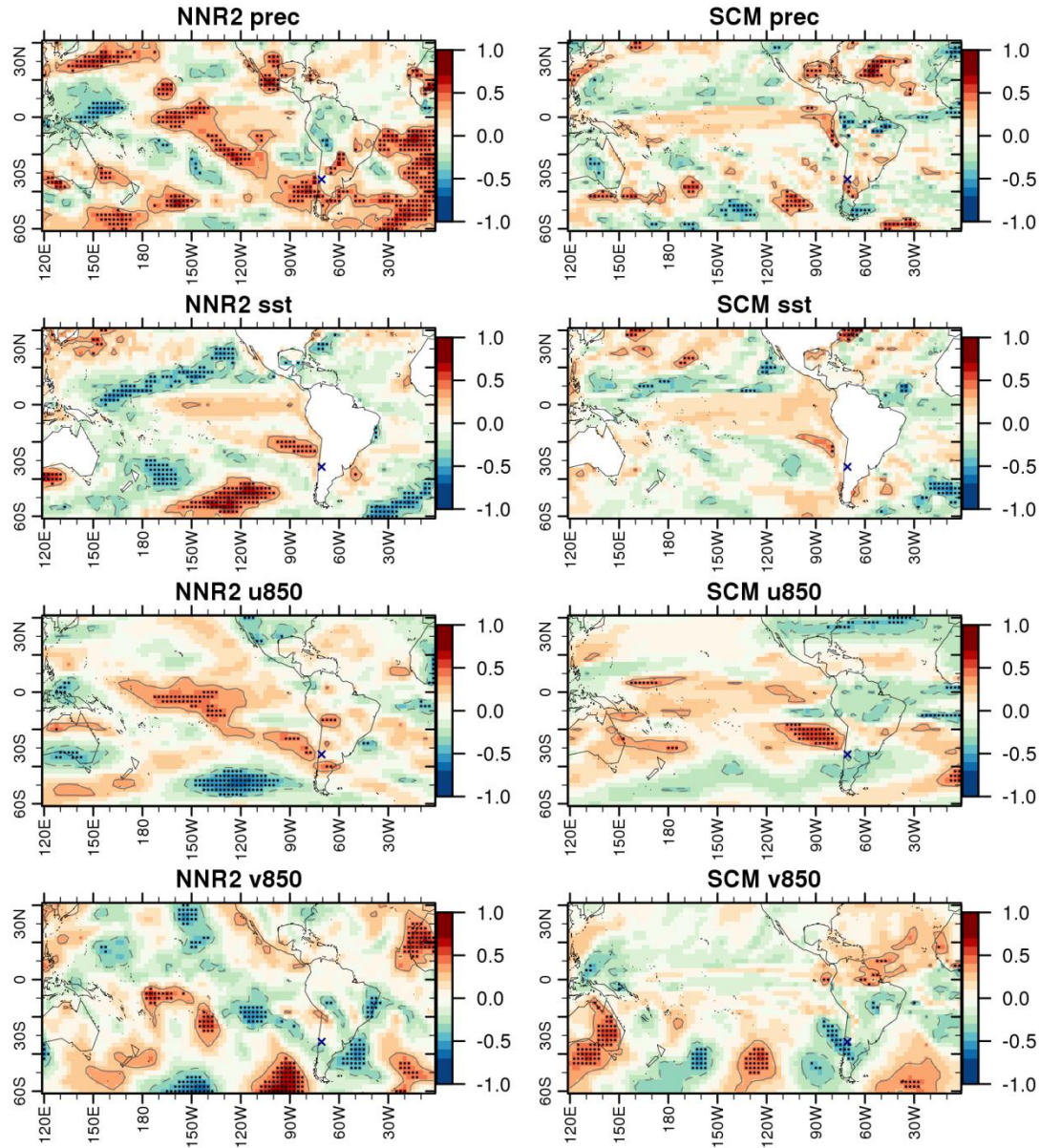
Tingo Maria [OND]



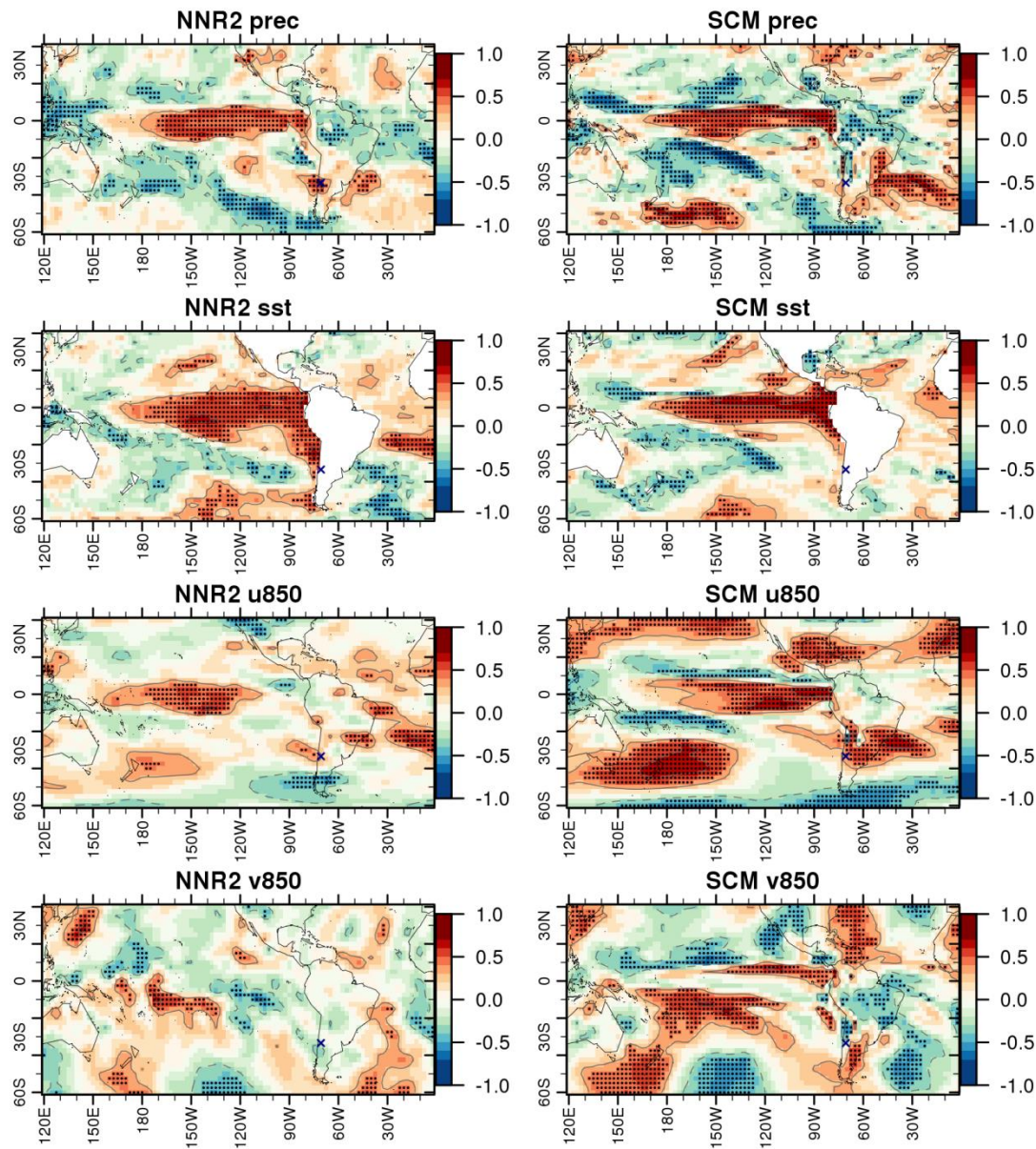
Tingo Maria [JJA]



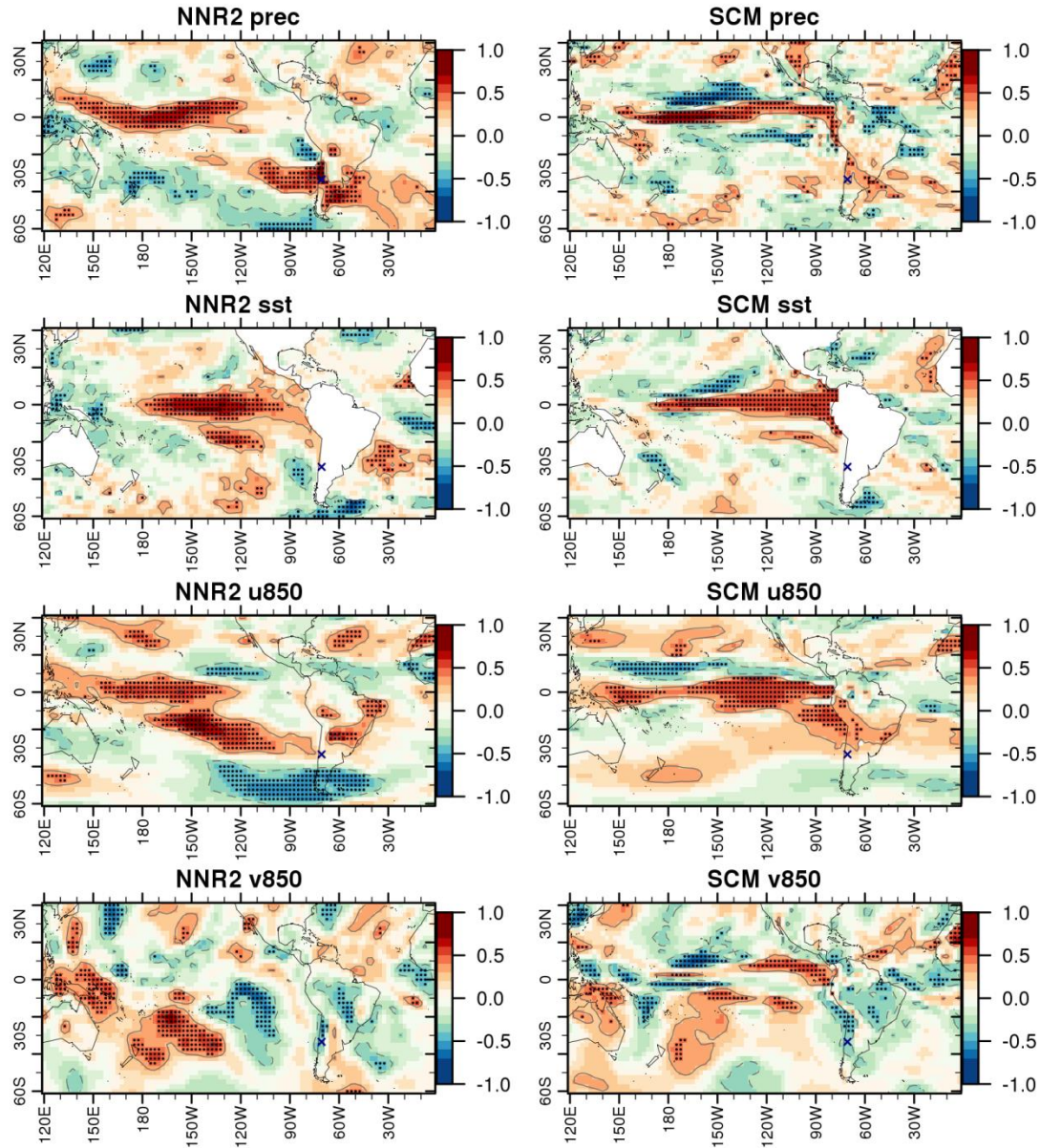
Quinta_Normal_santiago [DJF]



Quinta_Normal_santiago [OND]

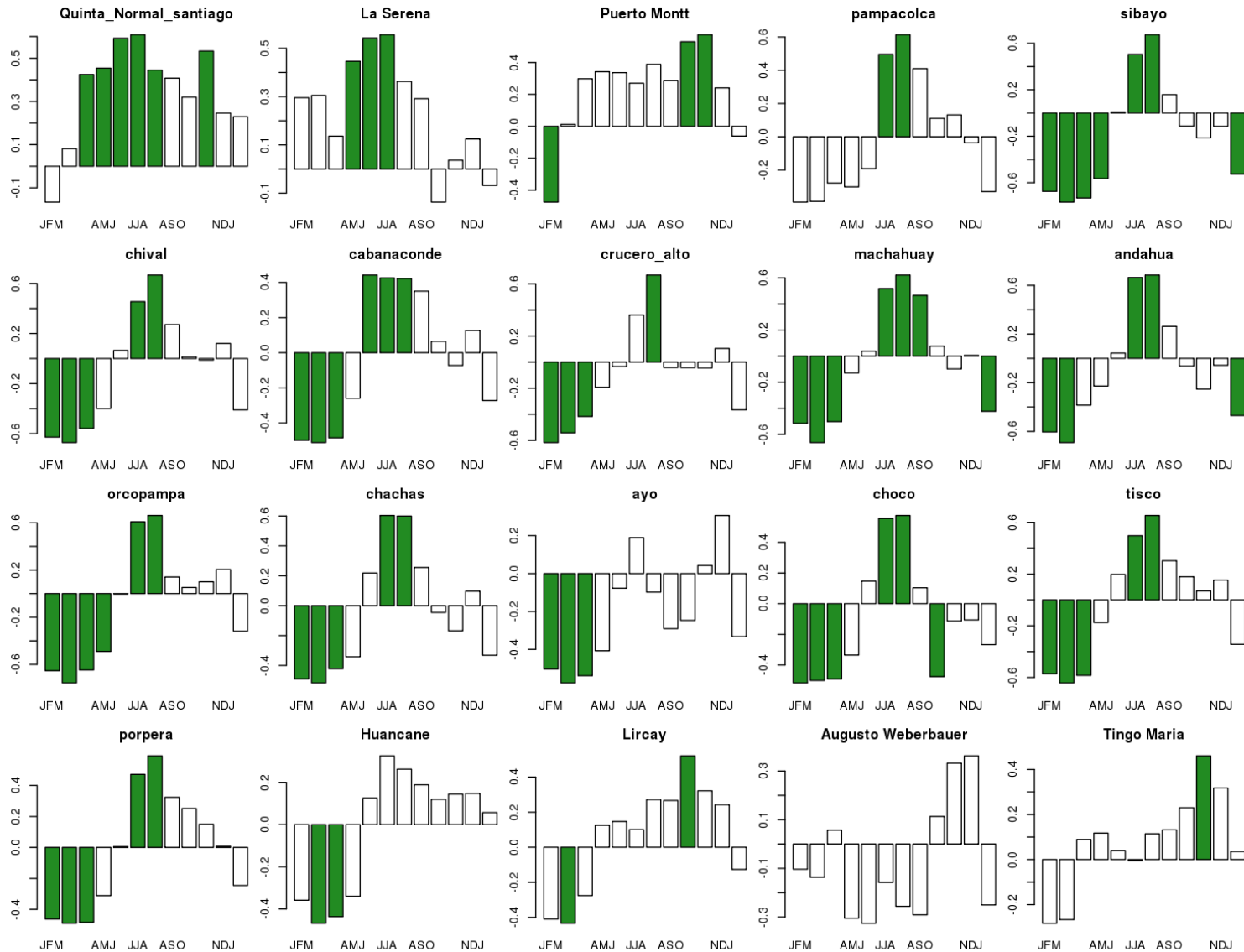


Quinta_Normal_santiago [JJA]



RELATIONSHIP WITH ENSO

corr. btw. station prec. and ONI



Seasonal Prediction (3) : Evaluation and Downscaling

Jin Ho Yoo
APEC Climate Center

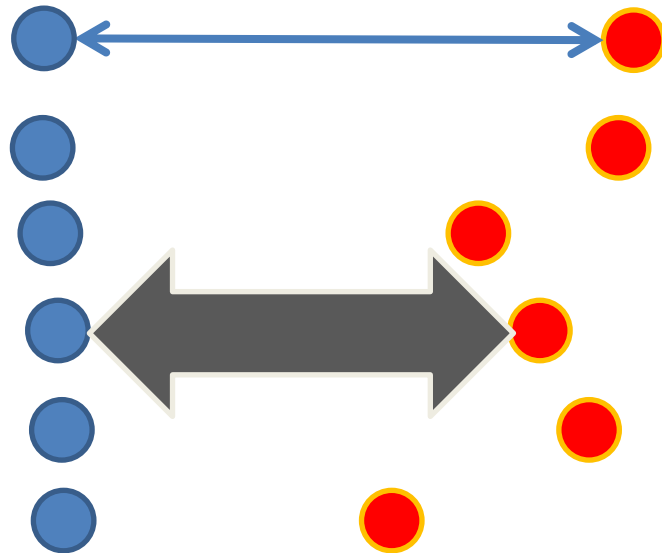
How GOOD?

- Evaluation of forecast : verification



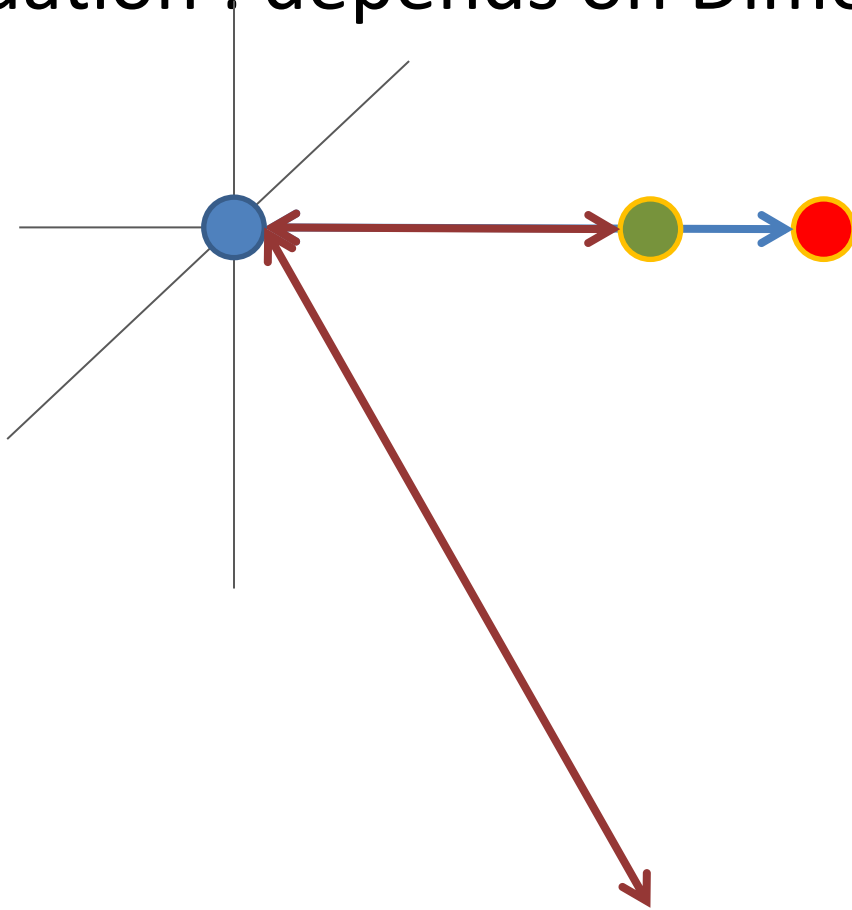
Verification

- Evaluation : measure of closeness



Verification

- Evaluation : depends on Dimension/Viewpoint



Deterministic forecast

- Various measures

- MSE (Mean Square Error), RMSE (Root MSE)

- $MSE = \frac{1}{N} \sum_i (F_i - O_i)^2$

- ACC (Anomaly correlation, Pattern), TCC (Temporal correlation)

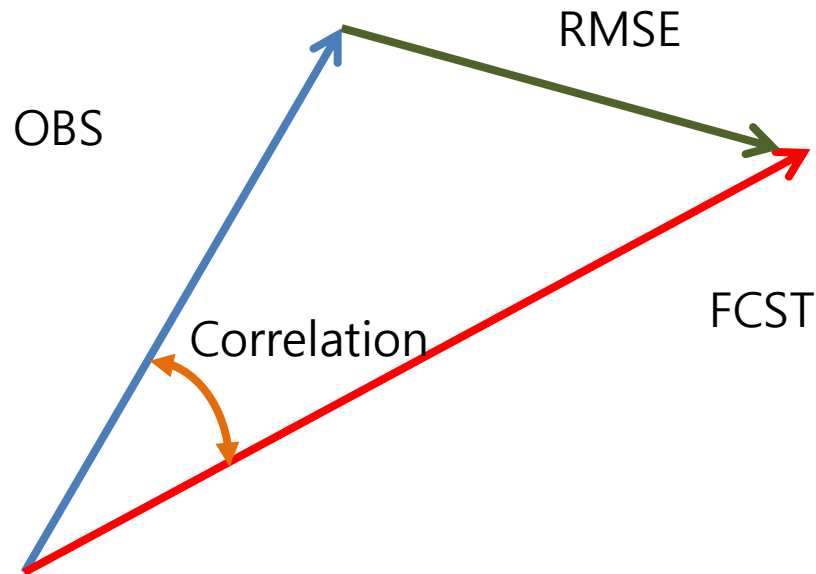
- MSSS (Mean Square Skill Score)

- Conventional form of skill score

- $1 - \frac{E}{E_c}$, E : error/penalty, E_c : error of reference forecast

Verification

- Evaluation : depends on Dimension/Viewpoint

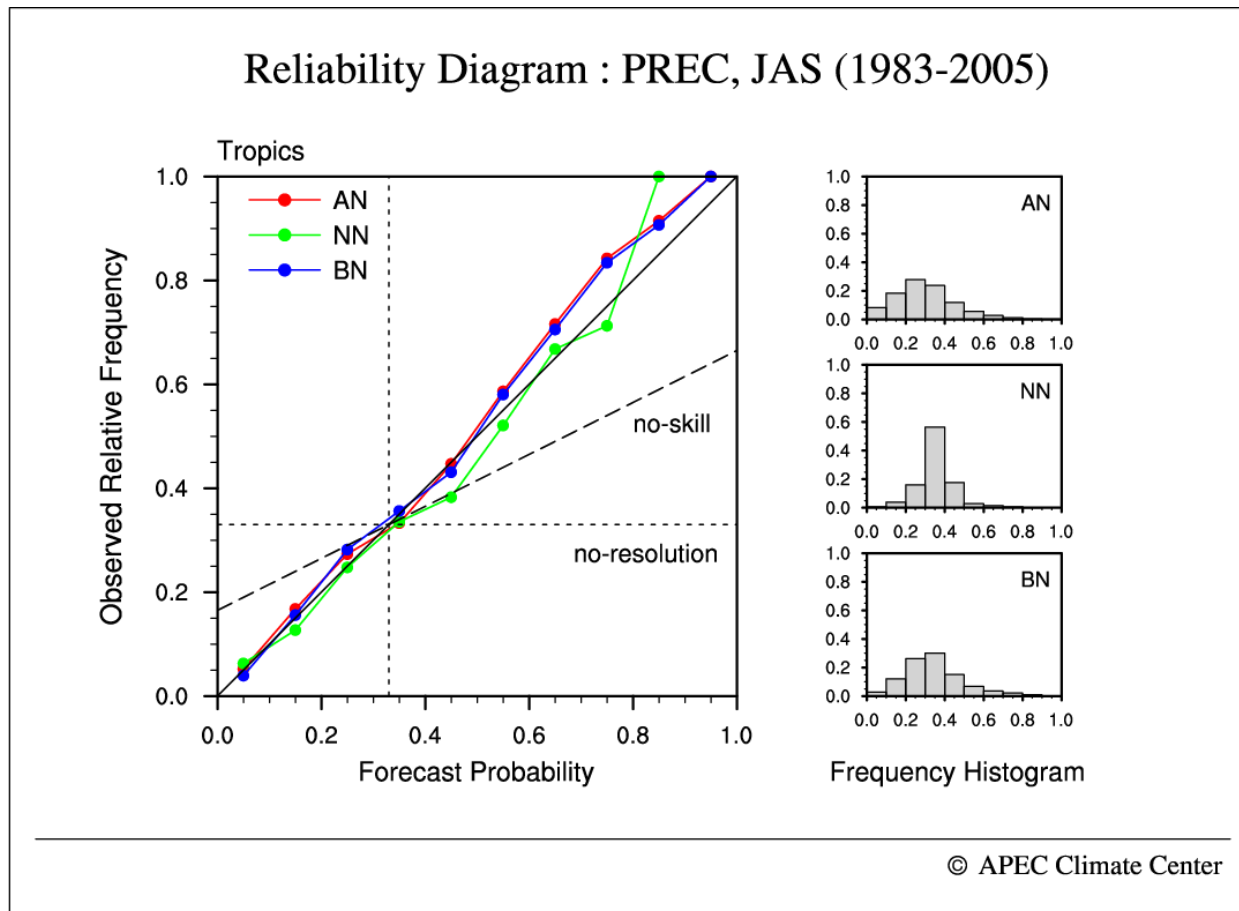


Probabilistic forecast

- Brier score (Brier Skill Score)
 - MSE of prob. forecast
 - $BS = \frac{1}{N} \sum_i (F_i - O_i)^2$, $F=1/0$, $O=1/0$
 - BSS (Brier skill score)
 - $1 - \frac{BS}{BS_c}$,
 - E : error/penalty, E_c : error of reference forecast

Probabilistic forecast (Categorical)

- Reliability curve



Probabilistic forecast (Categorical)

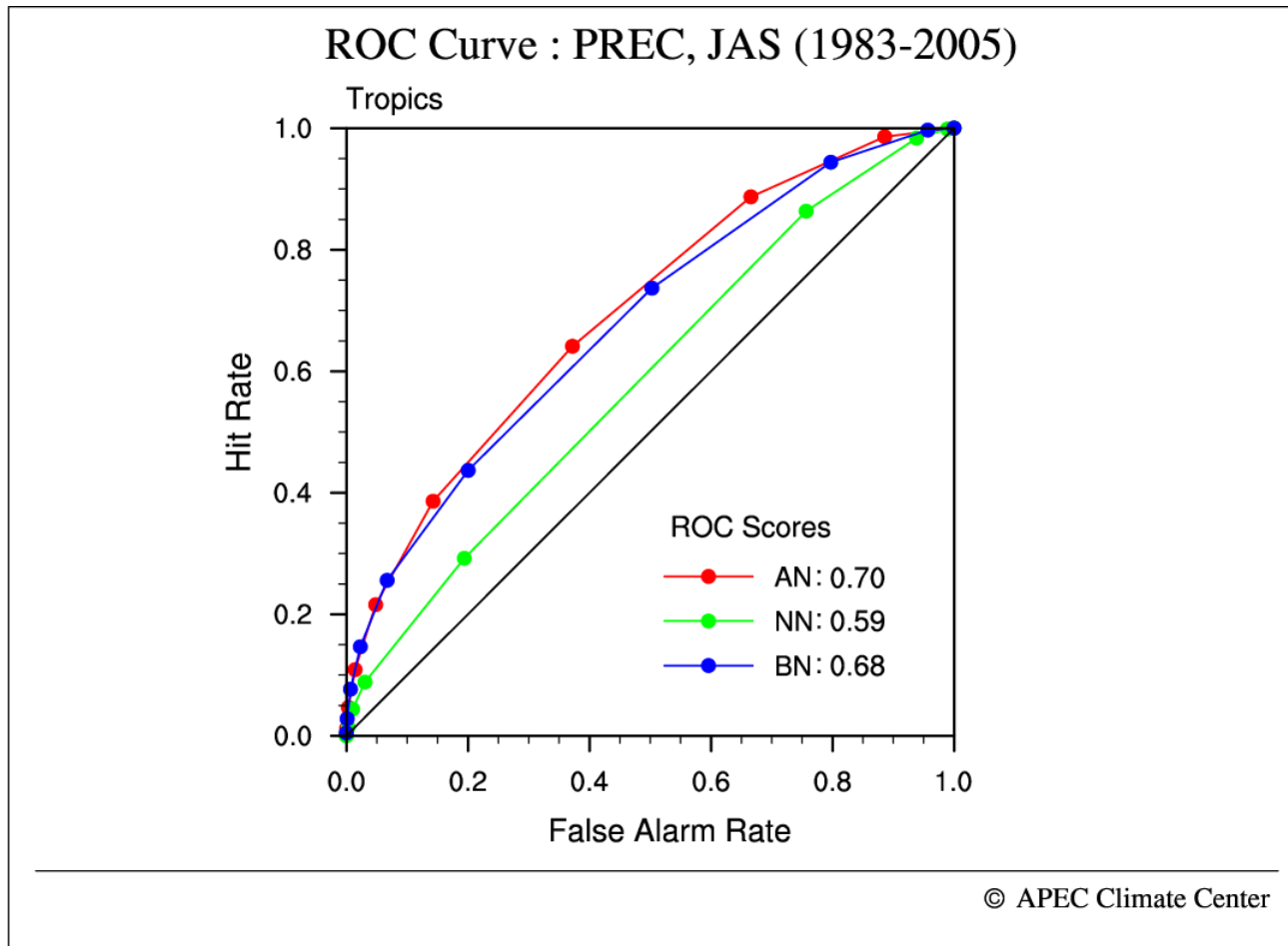
- ROC (Relative Operating Characteristics)

F \ O	Yes	No
	Yes	No
Yes	Hit (H)	False Alarm (F)
No	Miss (M)	Correct Rejection (C)

- $HR \text{ (Hit rate)} = H/(H+M)$
- $FAR \text{ (False Alarm rate)} = F/(F+C)$
 - Good forecast : $HR \uparrow, FAR \uparrow$

Probabilistic forecast (Categorical)

- ROC (Relative Operating Characteristics)



Probabilistic forecast (Categorical)

- HSS (Heidke Skill Score)

F \ O	Yes	No
	Yes	No
Yes	Hit (H)	False Alarm (F)
No	Miss (M)	Correct Rejection (C)

- $$HSS = \frac{(score - score\ by\ chance)}{(perfect\ score - score\ by\ chance)}$$

$$\frac{\{(h+c)/n - [(h+f)(h+m) + (f+c)(m+c)]/n^2\}}{\{1 - [(h+f)(h+m) + (f+c)(m+c)]/n^2\}}$$

Forecast economic value

$$V = \frac{E_{cli} - E_{fore}}{E_{cli} - E_{per}}$$

V=1 : perfect forecast

V=0 : climatological forecast

E_{fore} : Expected expense of forecast

E_{per} : Expected expense of perfect forecast

E_{cli} : Expected expense of climatological forecast

- When the forecast is **perfect**, $f = m = 0$. and $h = \bar{o}$. Then, $E_{per} = hC = \bar{o}C$
- When the forecast is **climatology**. The only one kind of action will be kept.

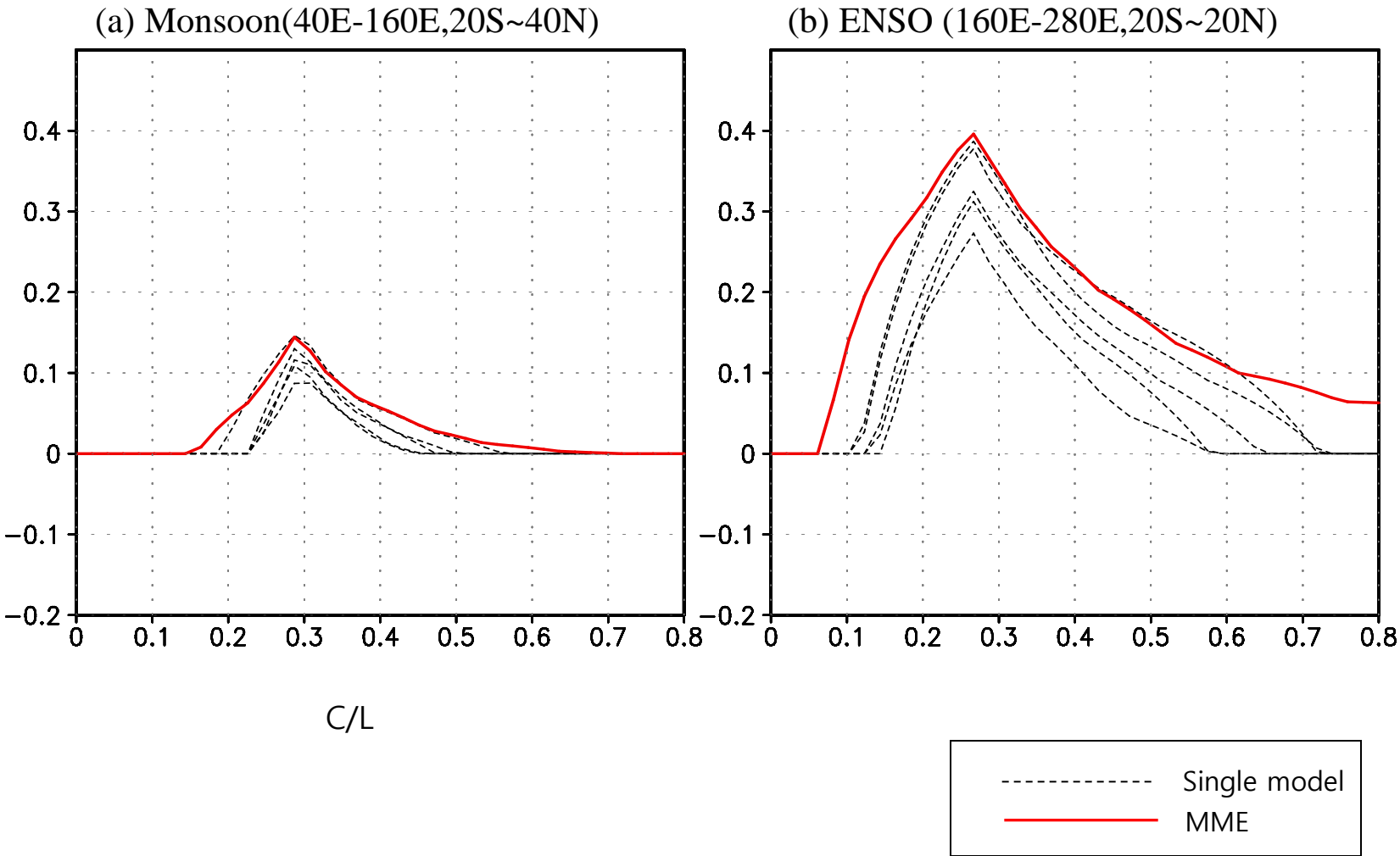
If Yes : $E = (h+f)C = C$, otherwise $E = mL = \bar{o}L$. If decision maker is rational, he/she will choose action of low expense. Thus, $E_{cli} = \min(C, \bar{o}L)$

		Observation (real event)	
		Yes	No
Forecast (action)	Yes	Hit (h) Cost	False alarm (f) Cost
	No	Miss (m) Loss	Correct rejection (c) 0

$$E_{fore} = (h + f)C + mL$$

$$V = \frac{\min(\frac{C}{L}, \bar{o}) - (h + f)\frac{C}{L} - m}{\min(\frac{C}{L}, \bar{o}) - \frac{C}{L}\bar{o}}$$

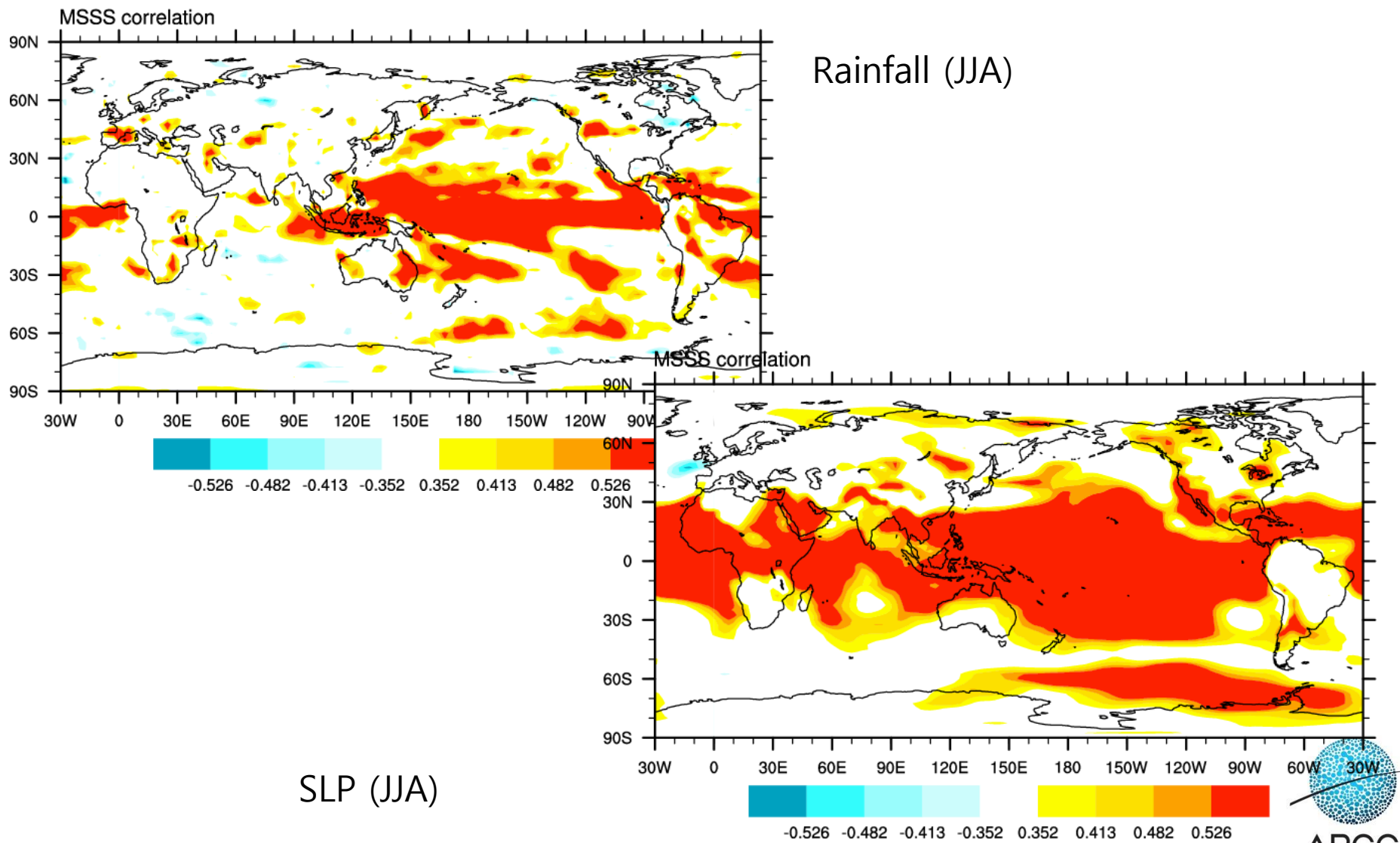
Value of Probabilistic forecast (Above normal) : GCMs



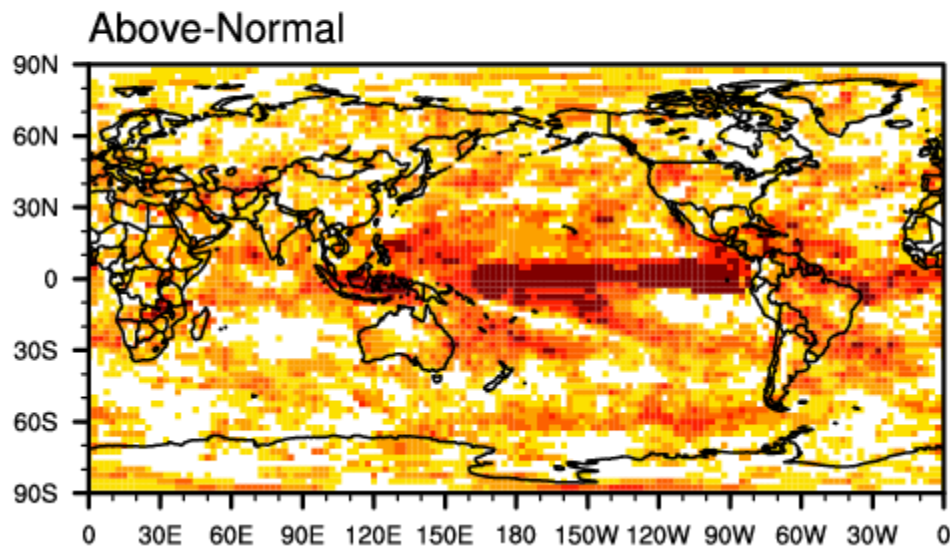
Forecast verification

- There are numerous ways
- Can be chosen by “what” do you want to see
- If not clear, use popular one.
- Difficulties in “translating” meteorological skill score into Public wording.
- Let’s see some results!

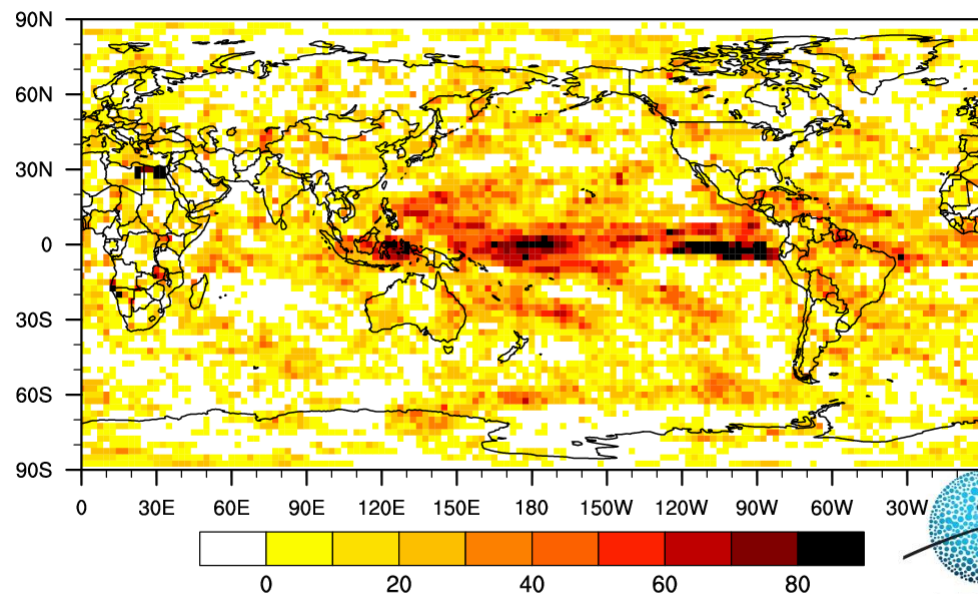
APCC MME (TCC)



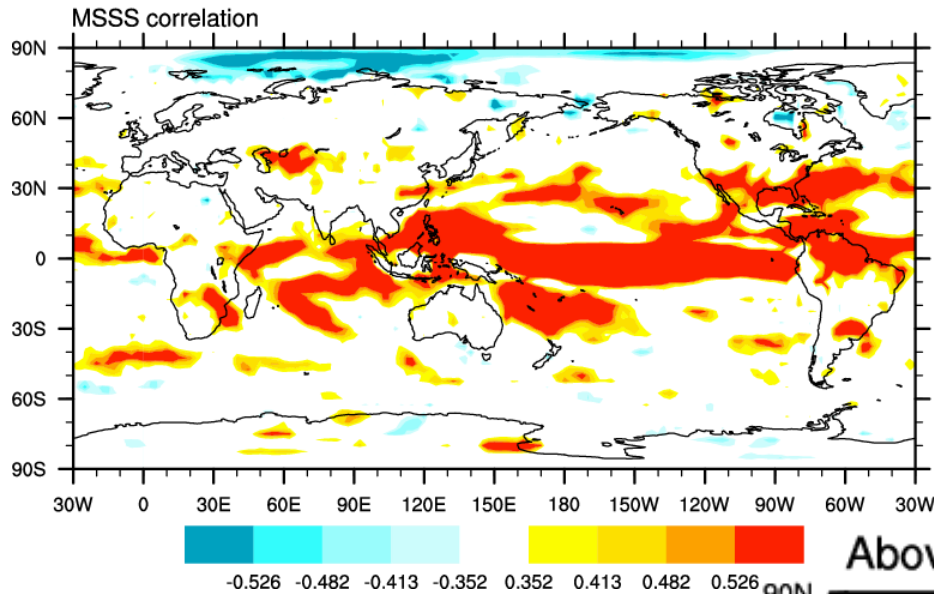
ROC Score : PREC, JJA (1983-2005)



Heidke Skill Score : PREC, JJA (1983-2005)

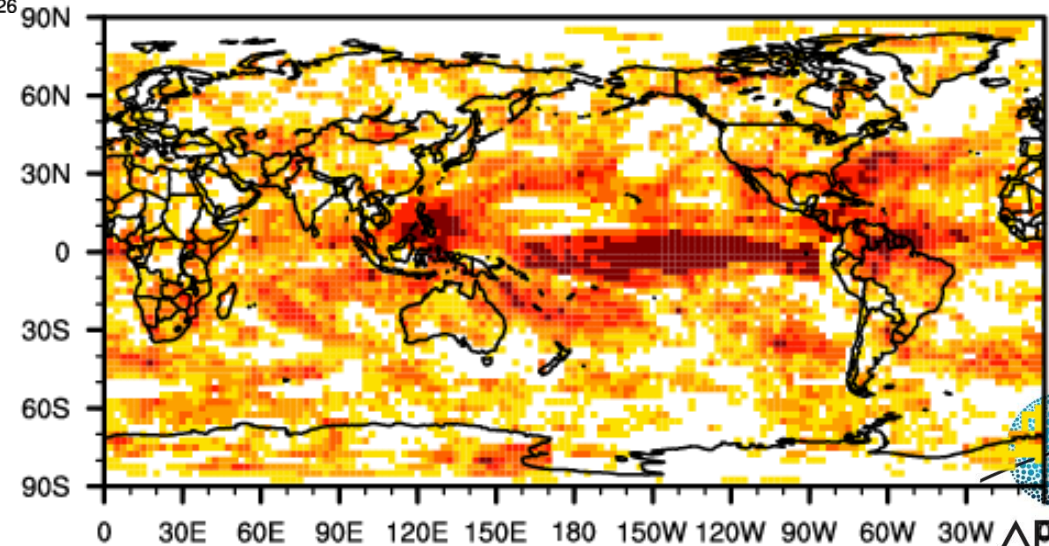


DJF season



Rainfall (JJA)

Above-Normal



R = 0.5, how good it is?

- Explaining 25% variance (R^2)
- A single verification score cannot tell everything.
- Multi aspect evaluation is necessary
- User oriented verification would be useful
- “this man can run fast, how good he is?”
- BUT we can “COMPARE” diff. things

Seasonal Prediction (4) : Operation and discussion

Jin Ho Yoo
APEC Climate Center



What we do?

- Collecting data and information
- Combine them
- Make a draft (preliminary decision)
- Consultation (discussion)
- Issue!

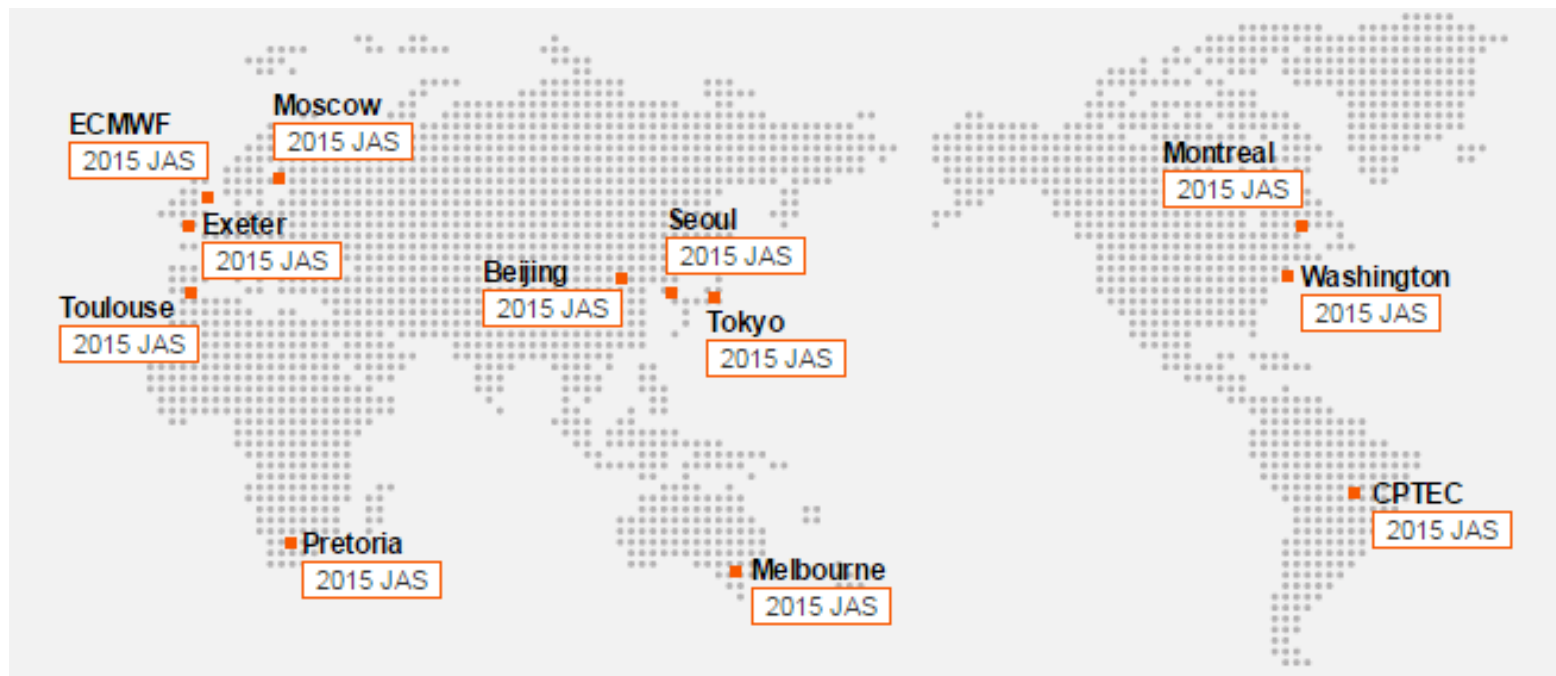
Current observation (monitoring)

- ENSO
 - WMO El Nino Update
- IOD
- ISO
 - CPC MJO page, APCC BSISO page

Why we monitor (analyze) current climate state?

Global forecast Information

- Dynamical Seasonal Prediction
 - GPCs, [WMO LC_LRF](#), [APCC](#), IRI, NMME



www.wmolc.org : only open to WMO members

Monitoring & Forecast information

- More maps are not always helpful unless they are **DIGESTED** properly
- It is known that **Multi Model Ensemble** tends to produce better forecast than a single model but it can lose regional details (maybe because of this, general skill is high)
- At best, all the information is merely explain large scale feature

Combining information

- If you can trust one thing, that is enough
- If you have different information with similar reliability, trust both
 - Are they **Independent**?
- If you can distinguish good and better information (but they are different), combine them with weight
- If you don't have any idea on the reliability, treat them similarly (they are all 'state-of-art' information)

Cautions

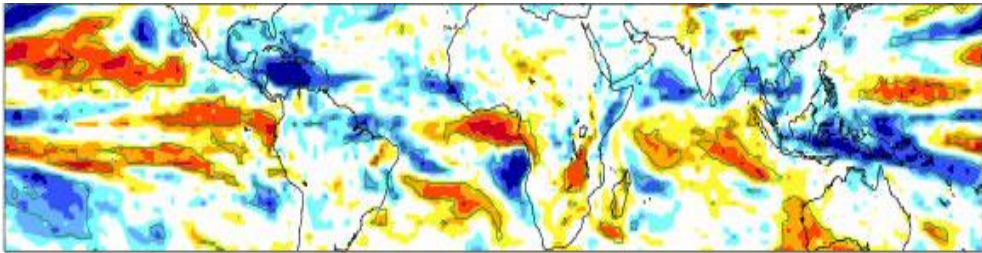
- How reliable our evaluation is?
- Even if you trust them, they can be wrong.
 - One reason to issue “probabilistic forecast”

2006 JJA mean Rainfall forecast

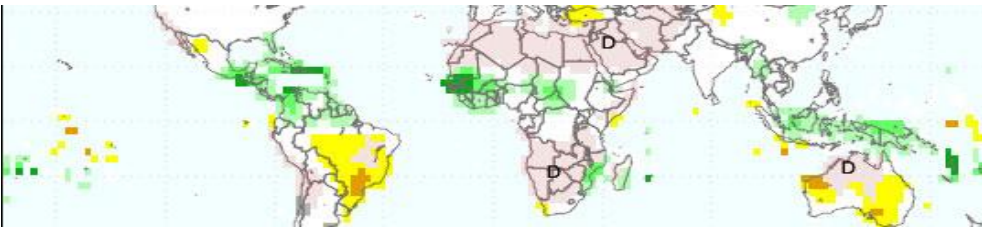
Warm colors : **dry**

Cool colors : **wet**

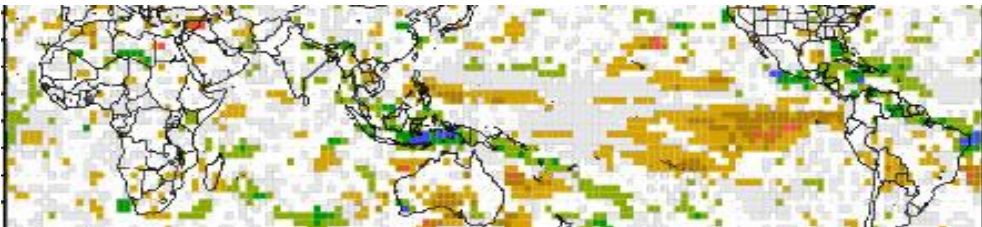
ECMWF



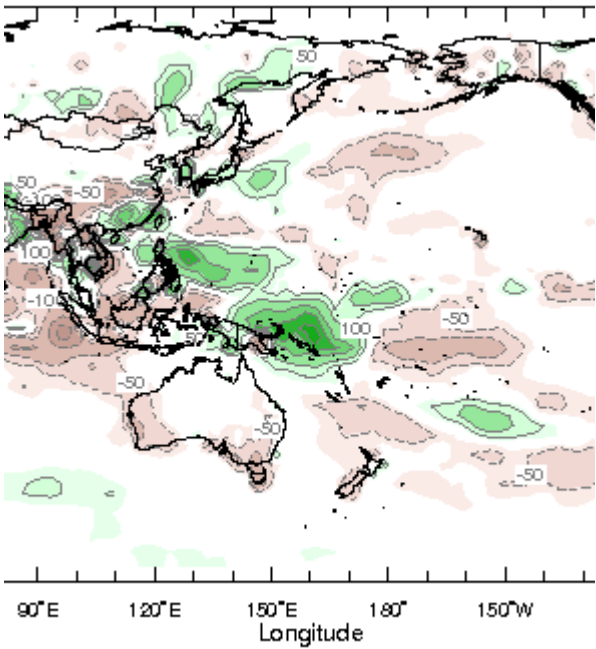
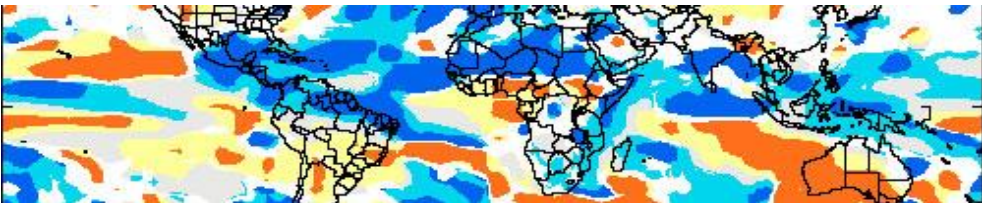
IRI



JMA



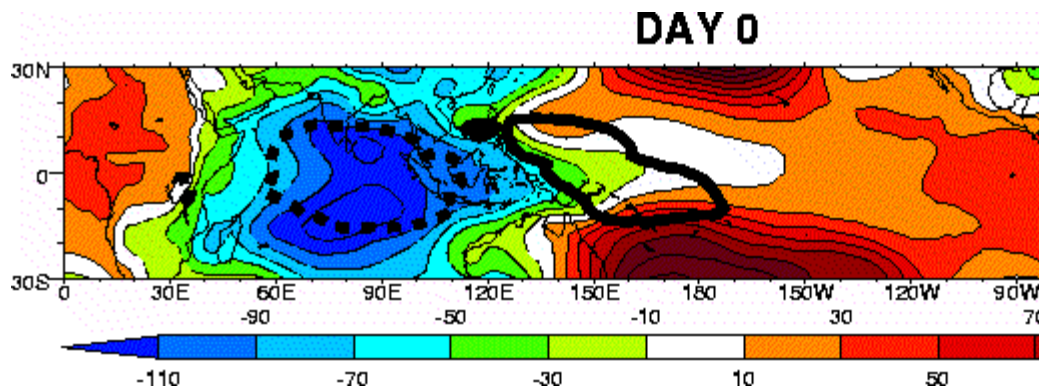
UKMO



A few more...

- Subseasonal information (MJO...)
- A new type of El Nino (El Nino Modoki)
- Way forward

Madden-Julian Oscillation

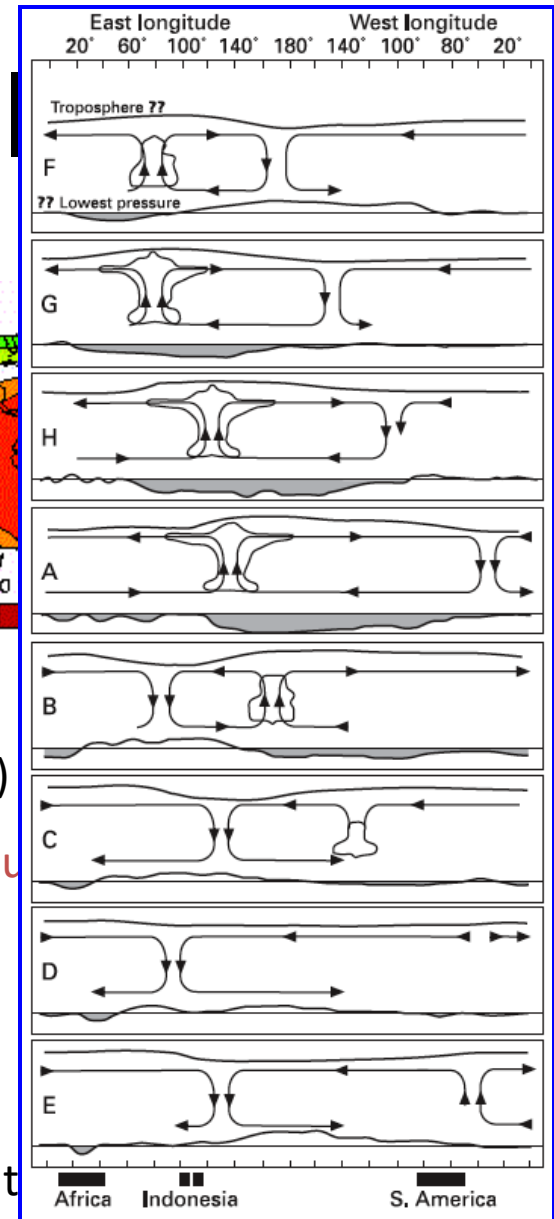


Madden & Julian (1971) : 40-50day oscillation (30-60days ISO)

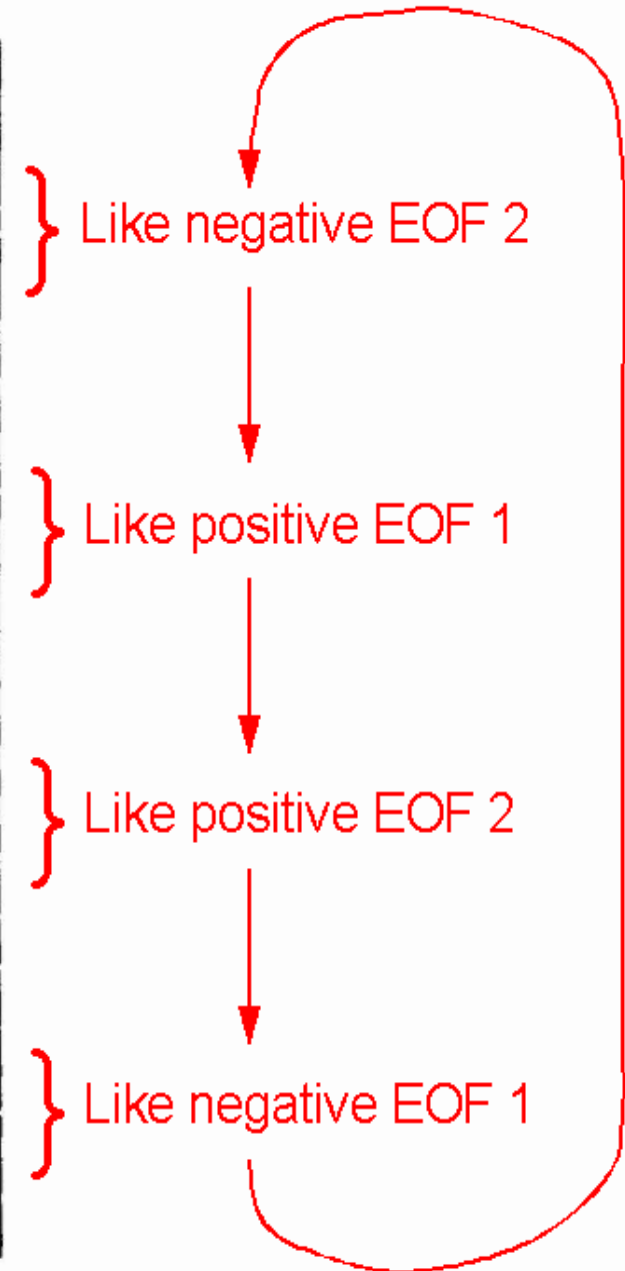
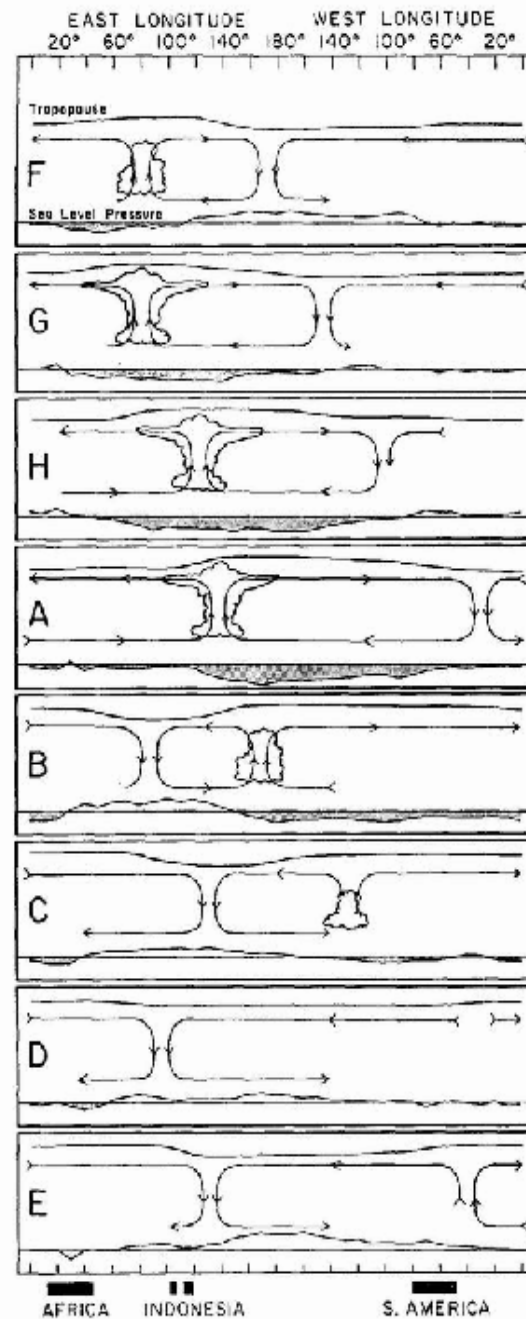
Eastward moving large scale convective anomaly along the equator
baroclinic structure

(precipitation anomaly is predominant in Indo-Pacific sector)

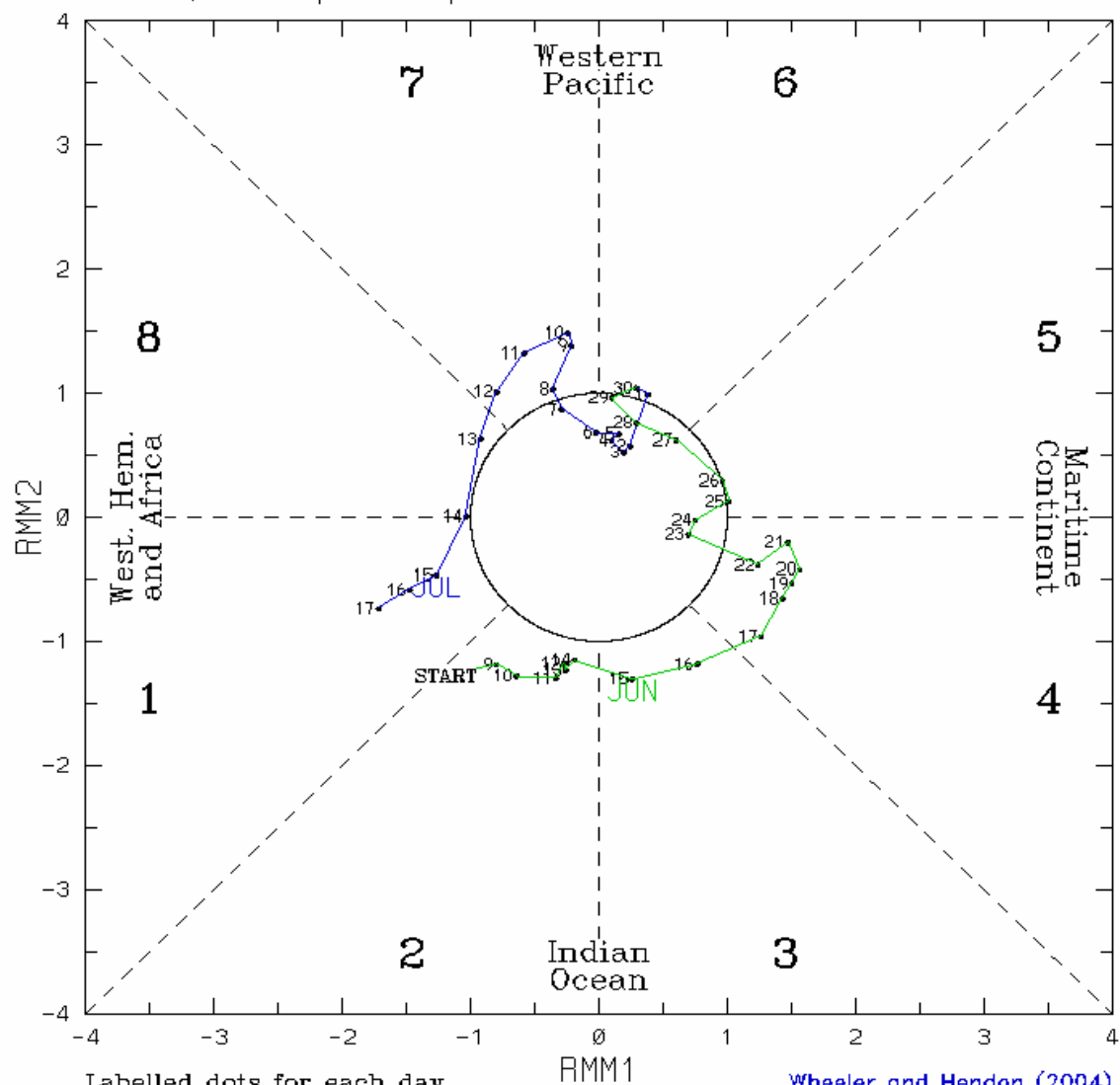
It can be a predictability source of extended range forecast in the



Madden and Julian (1972)



(RMM1,RMM2) phase space for 8-Jun-2007 to 17-Jul-2007



Labelled dots for each day.

Blue line is for Jul, green line is for Jun.

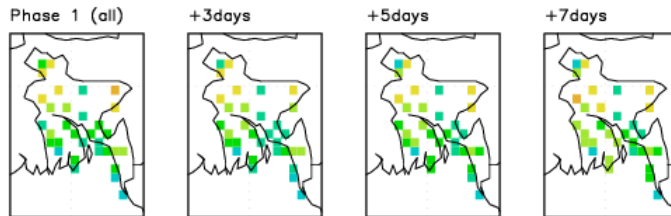
Wheeler and Hendon (2004)

BMRC Climate Forecasting

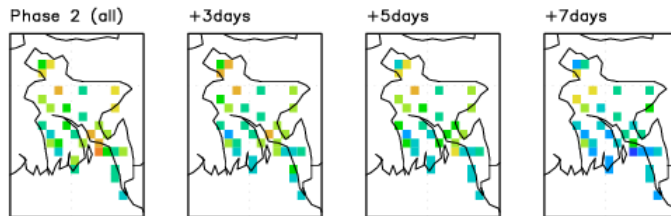
MJO and Bangladesh rainfall (% of climatology)

MJO duration →

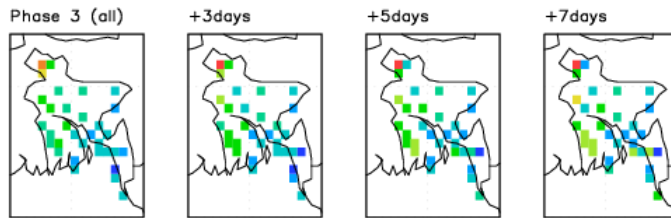
1



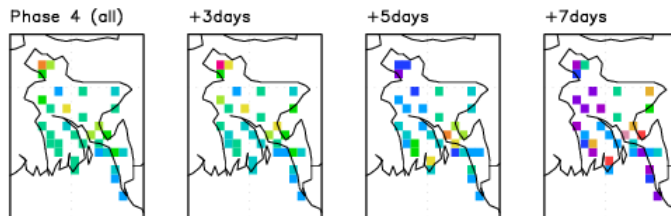
2



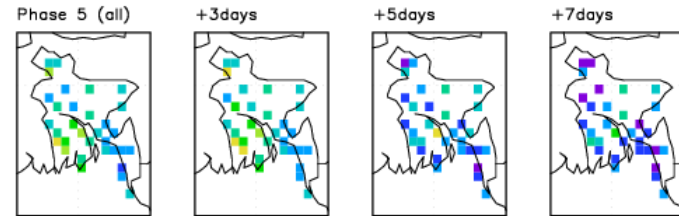
3



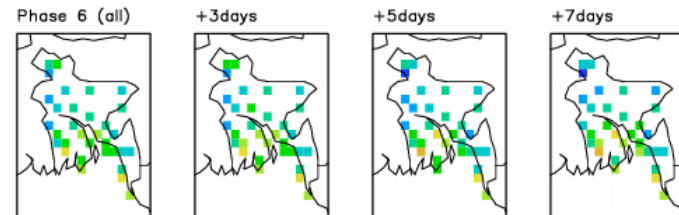
4



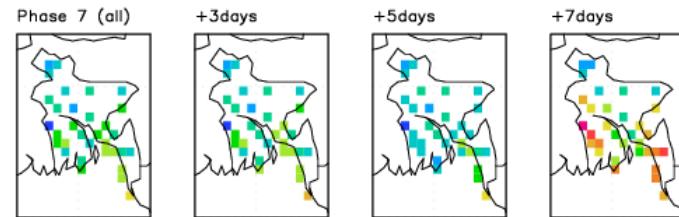
5



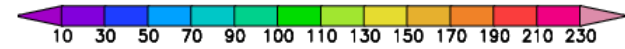
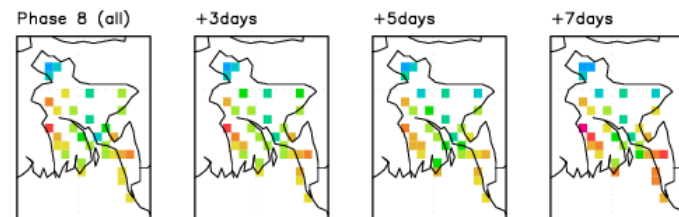
6



7

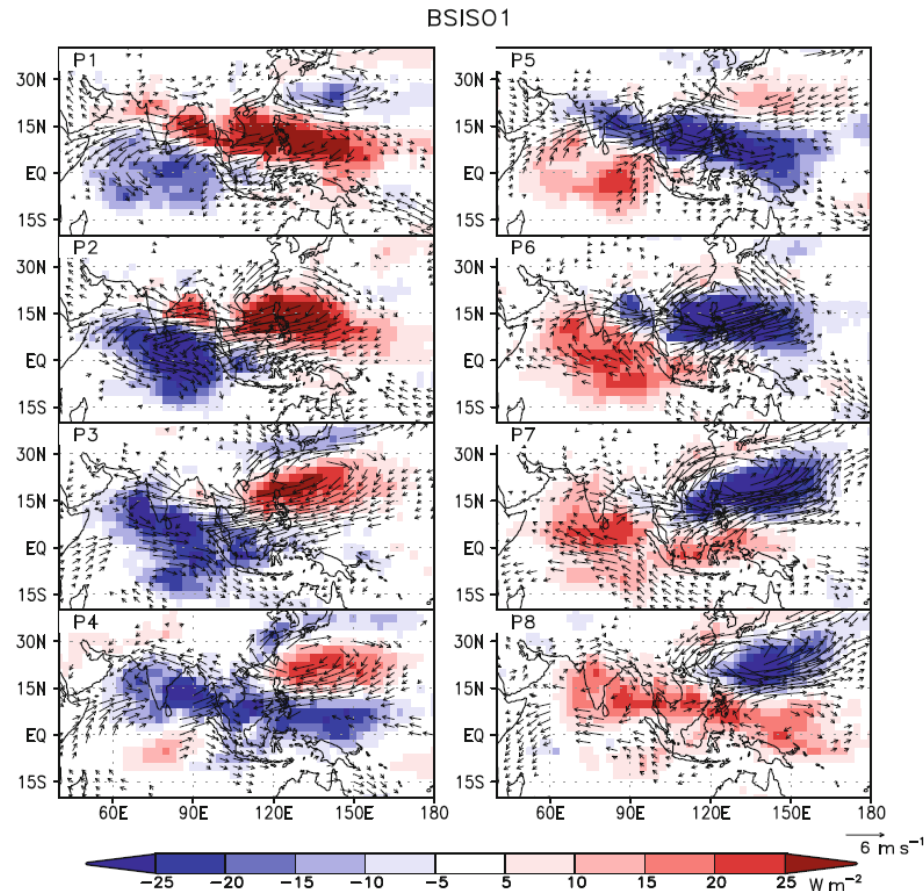


8

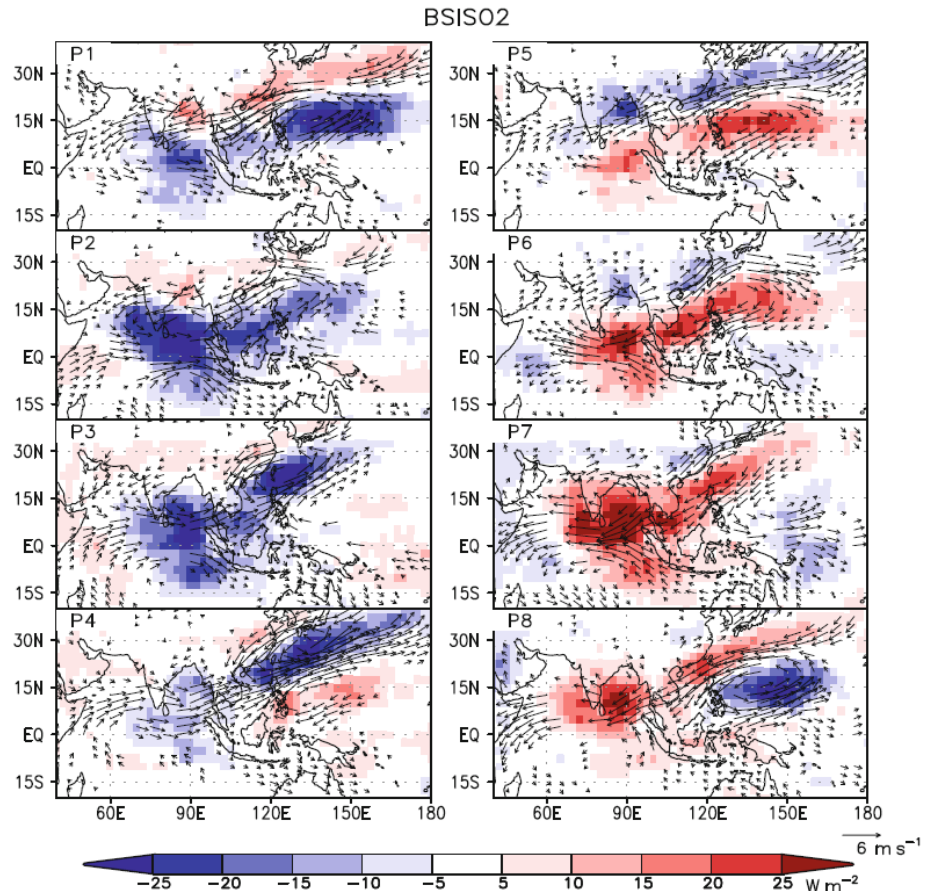


BSISO (Boreal Summer ISO)

The canonical northward propagating component

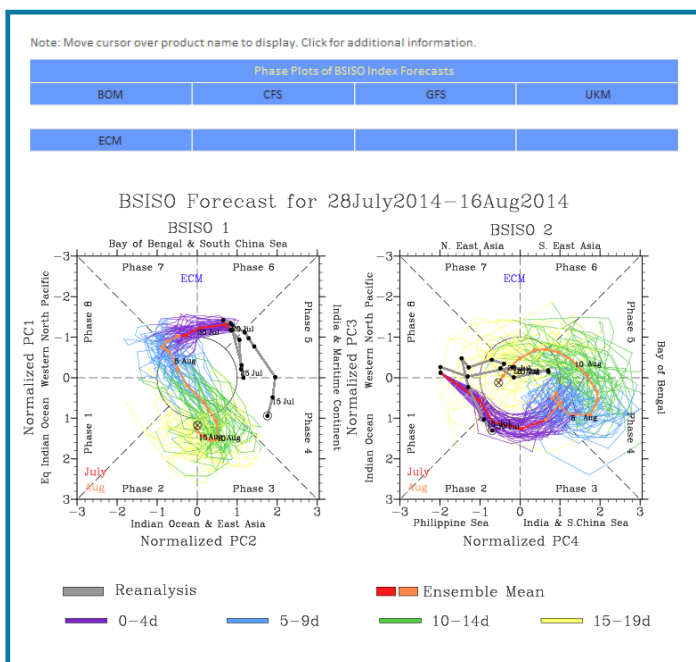
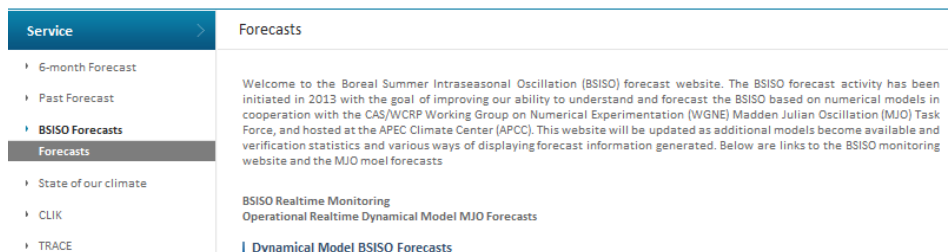
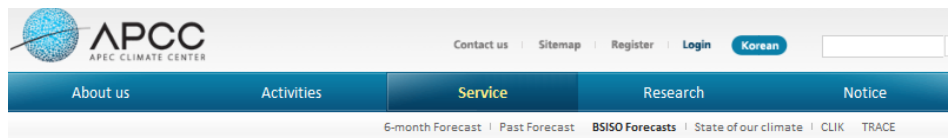


The AMS pre-monsoon and onset component



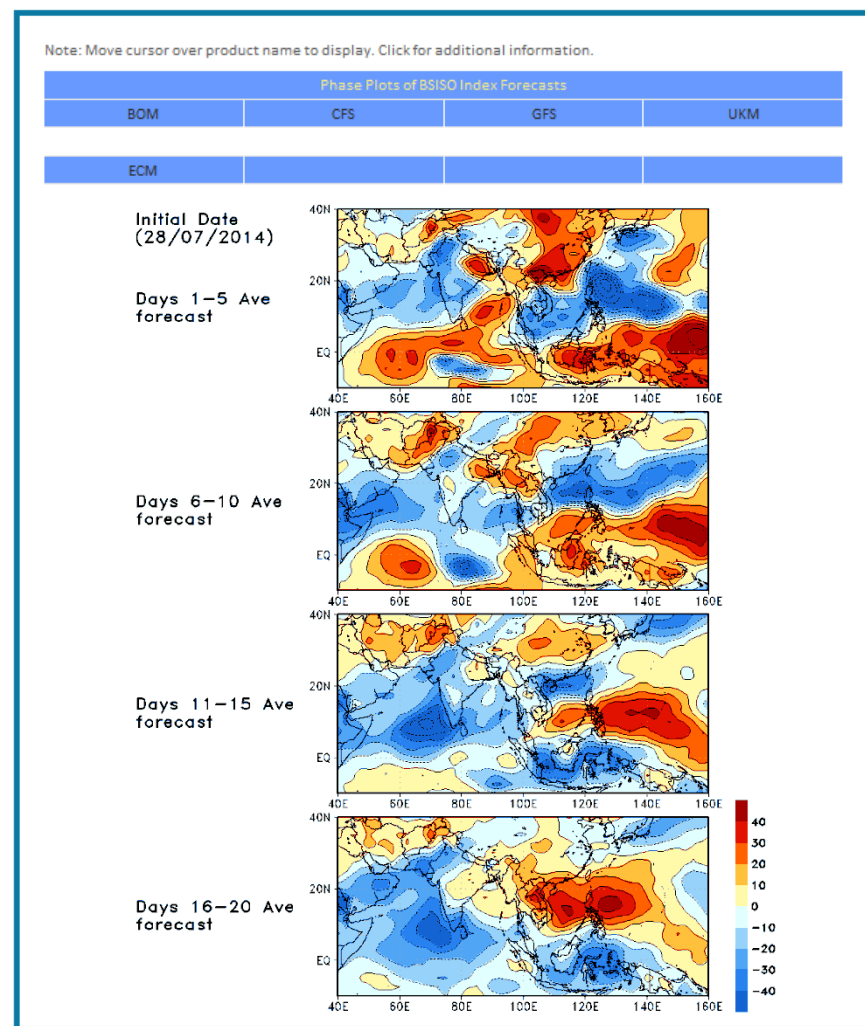
Lee, J.-Y., B. Wang, M. C. Wheeler, X. Fu, D.E. Waliser, and I.-S. Kang, 2013: Real-time multivariate indices for the boreal summer intraseasonal oscillation over the Asian summer monsoon region. *Clim. Dyn.*, 40, 493-509.

BSISO forecast (May-Oct)



Spatial OLR Anomalies

A key for the label headings in the figure box is provided below.



S2S project

Subseasonal to seasonal (15-60days)

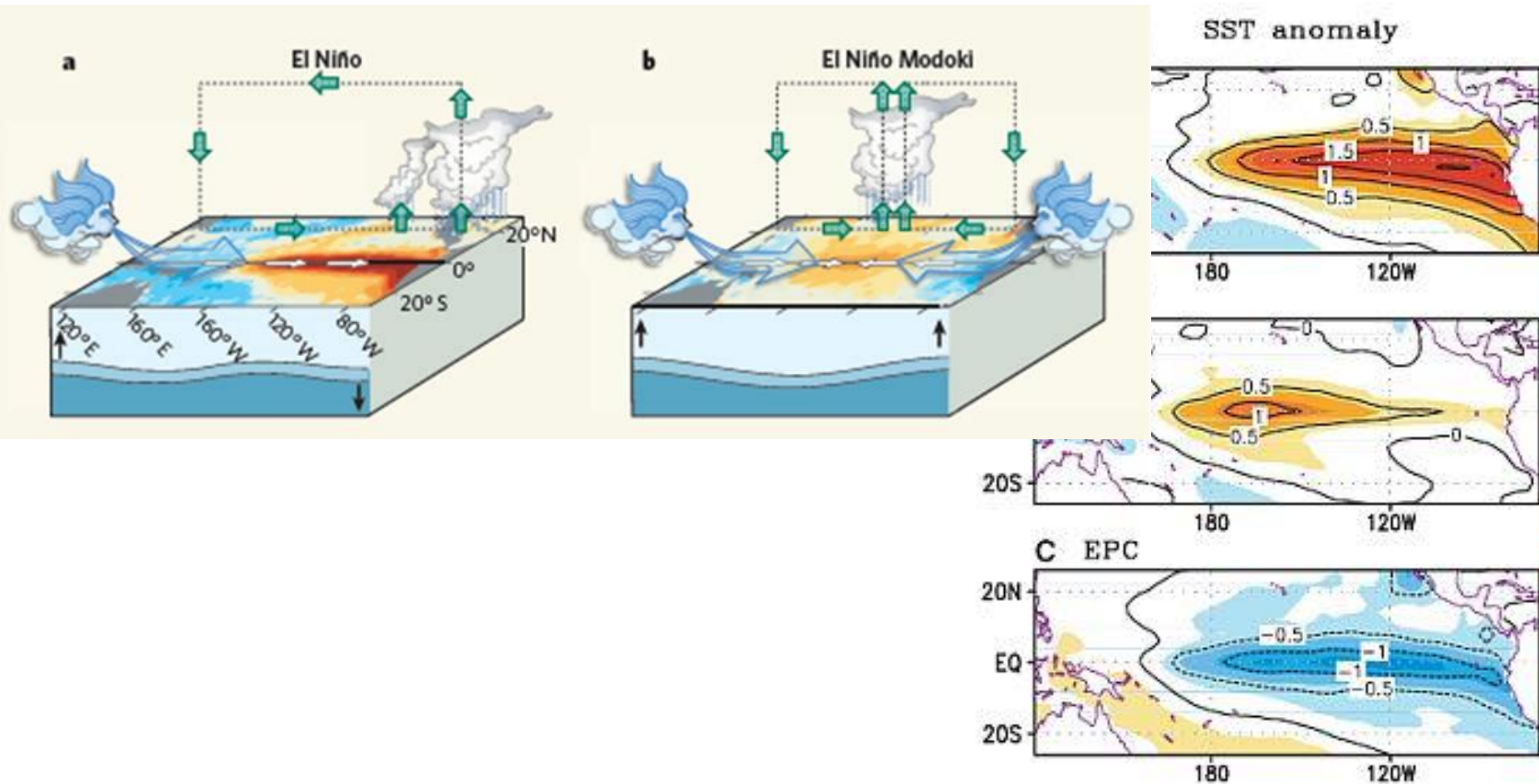
Objectives

- 1.To improve forecast skill and understanding on the subseasonal to seasonal timescale with special emphasis on high-impact weather events
- 2.To promote the initiative's uptake by operational centres and exploitation by the applications community
- 3.To capitalize on the expertise of the weather and climate research communities to address issues of importance to the Global Framework for Climate Services

<http://s2sprediction.net>

New type of El Nino

El Nino Modoki (Central Pacific El Nino)



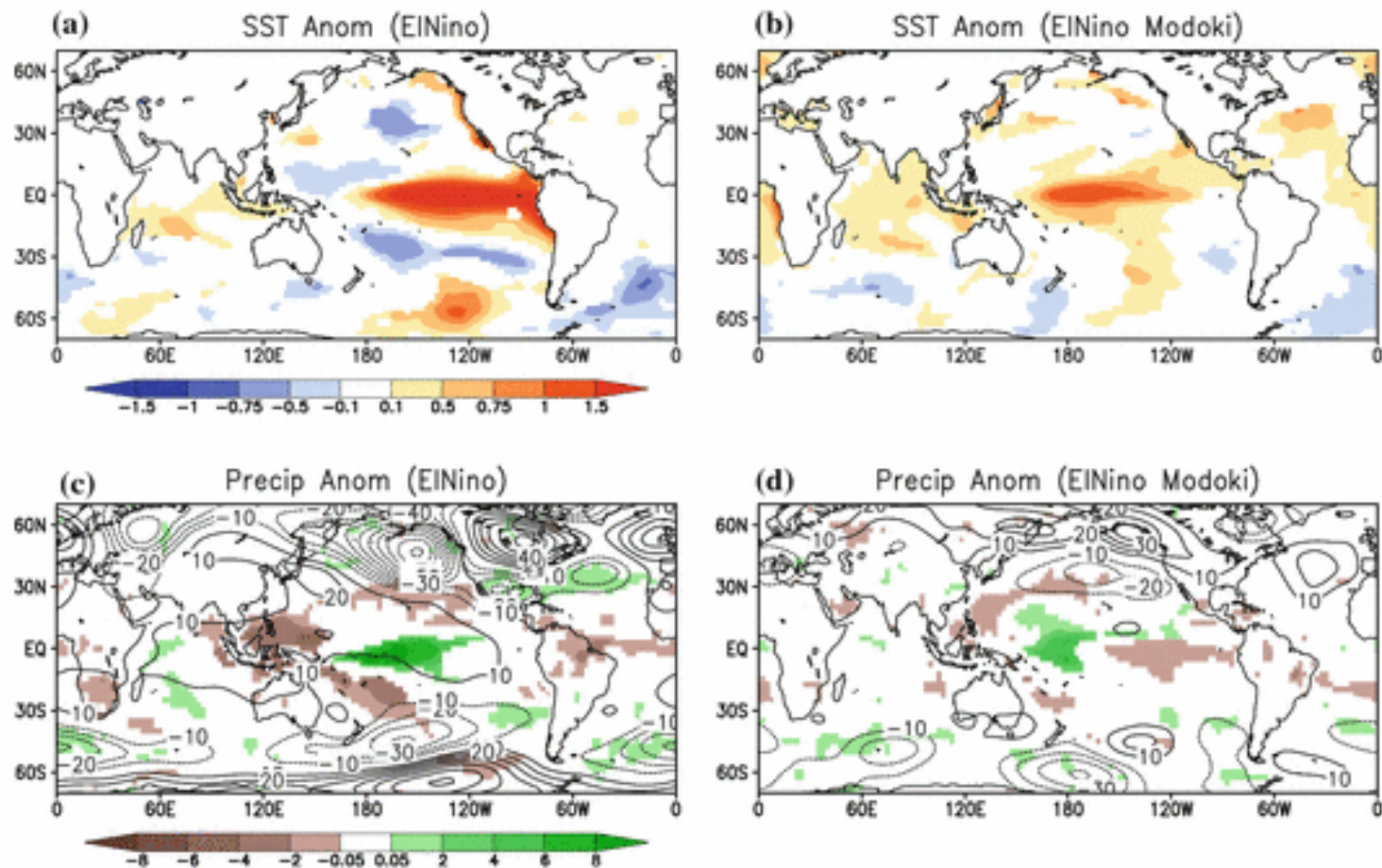


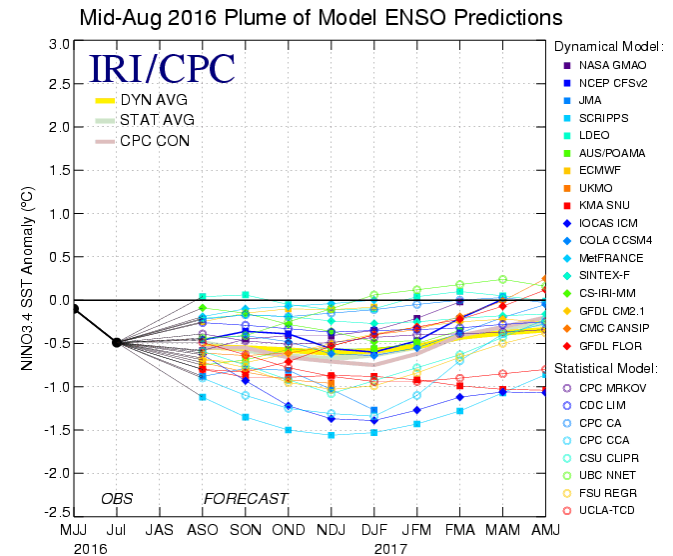
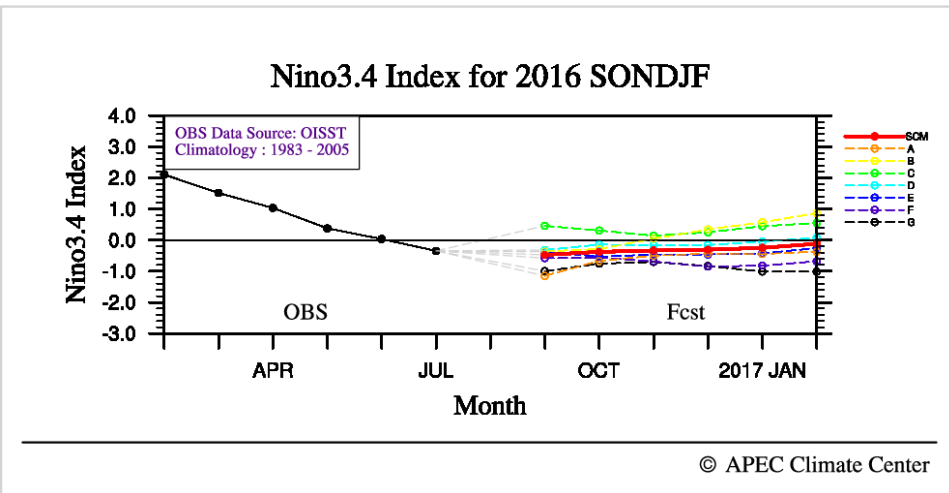
Fig. 2 a Composite observed significant SST ($^{\circ}\text{C}$) anomalies during El Niño years. b Same as a but for El Niño Modoki years. c Composite observed significant precipitation (mm/day; shaded)

anomalies and 500 hPa Geopotential Height (m; contours) anomalies during El Niño Years. d Same as c but for El Niño Modoki years. All the shaded values are significant at 90% using t test

So, what are you going to do with **CLIK**

- Hope you were able to find suitable predictors for your locations
 - **Yes** : produce forecast
 - **No** : try more! (It is important to understand the large scale circulation that affects local weather)
- Once you've got a forecast, you need to combine them with other informations.
 - It's an area of “**art**” at this moment

Practices



1. What is the most visible (potentially important) climate fluctuation : La Nina
2. Is the La Nina coming? Is it going to be Strong?
3. How did La Nina change the weather statistics before?
4. What is the forecast from MODELS? How much are they reliable? If it is not reliable, do we have “calibrated” forecast from them??
5. What is your conclusion???

Suggestions?
Questions?

Thanks



Note that,

- YOU should have an “Guess field” that is associated with your seasonal mean climate variability (“positive SST over certain region causes more rainfall at our station”)
- Model should be able to mimic that physical relationship even with some error
- CLIK will work if you can find a predictor satisfying above two thing

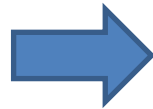
Downscaling in CLIK

- Use “observed” large scale pattern (X) associated with climate variability at stations
- X needs to be predicted by GCMs to some degree
 - X becomes predictor (user selected area)
- CLIK does not provide any prior information for selection of predictor (to avoid overfitting)
 - Basic knowledge on Local large scale circulation and associated global teleconnection is necessary

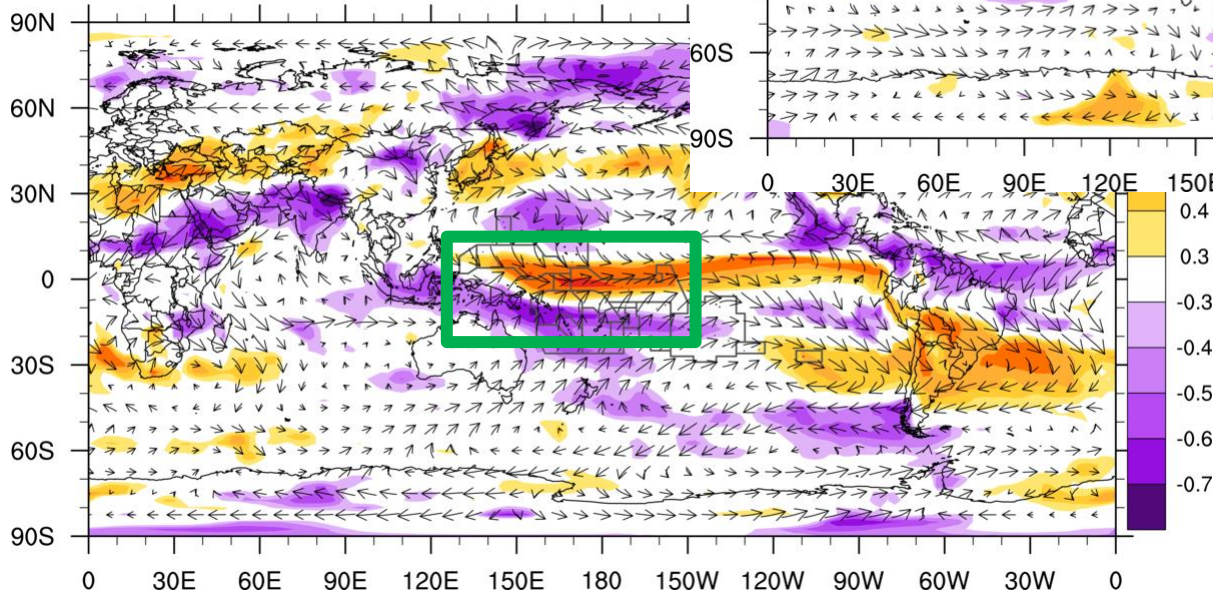
Predictor selection

Meaningful pattern? (hopeful)
: significance score

Station data

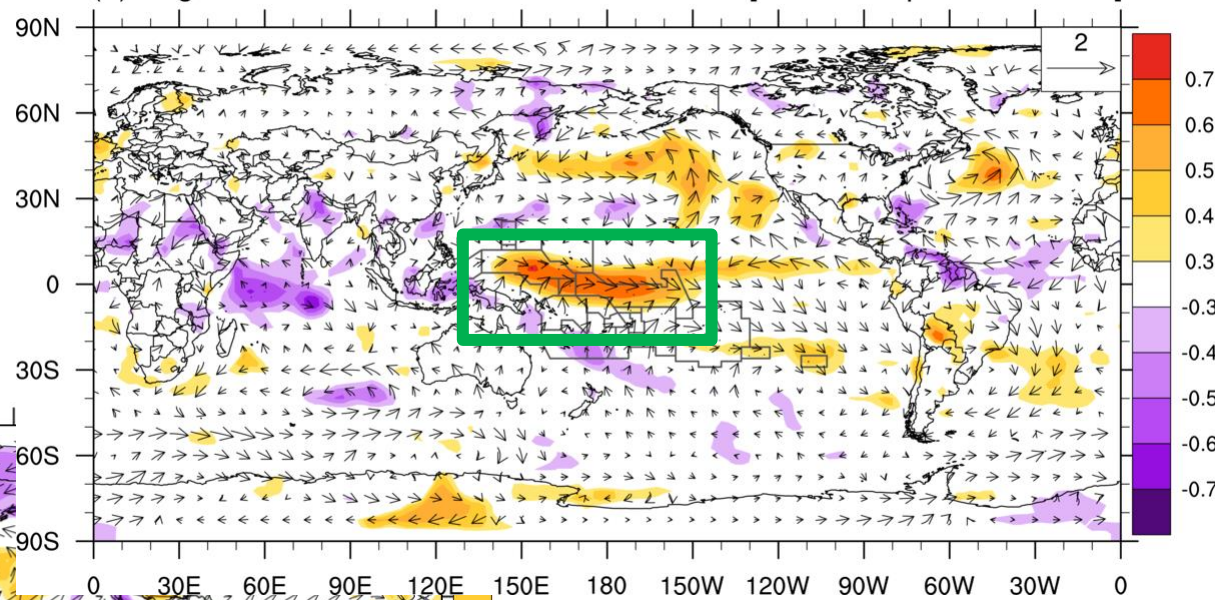


(a) Reg. between MME & Station 91348



(a) Reg. between Obs. & Station 91348

[JJA - Precipitation & Wind]



Consistency between obs.
and GCMs (good)
: pattern score

The most important thing you need is,

Patience

