Seasonal Prediction (1)
: Introduction/Predictability

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Overview

- Predictability
- Methods

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Climate prediction

Climate

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Climate is what we expect,

Weather is what we get

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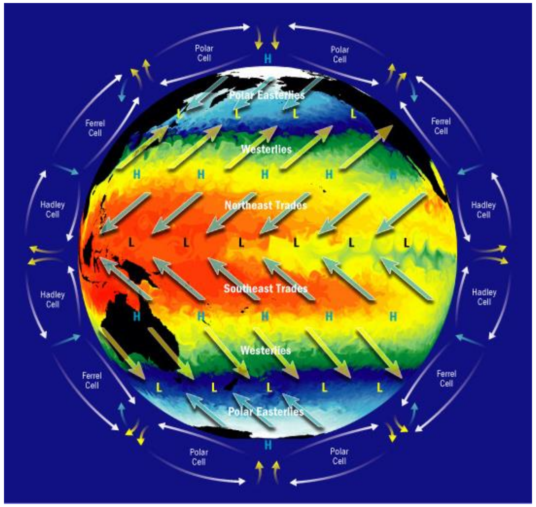
Climate = Expectation



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Climate Change = We need to change our Expectation

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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

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

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

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"

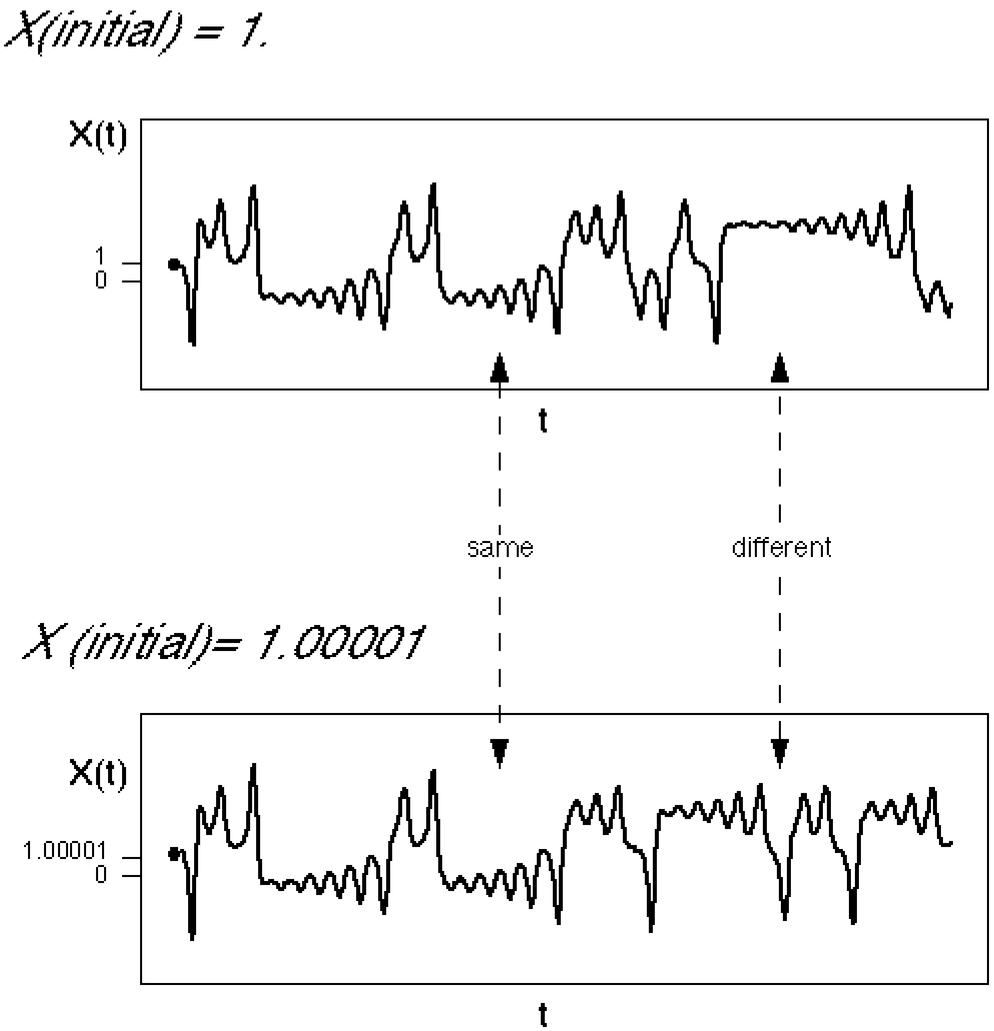
Predictability

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(initial) = 1.
X(t)
1
0
t
same
different
X (initial)= 1.00001
X(t)
1.00001
0
t

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

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Predictability

Depends on *what to predict*

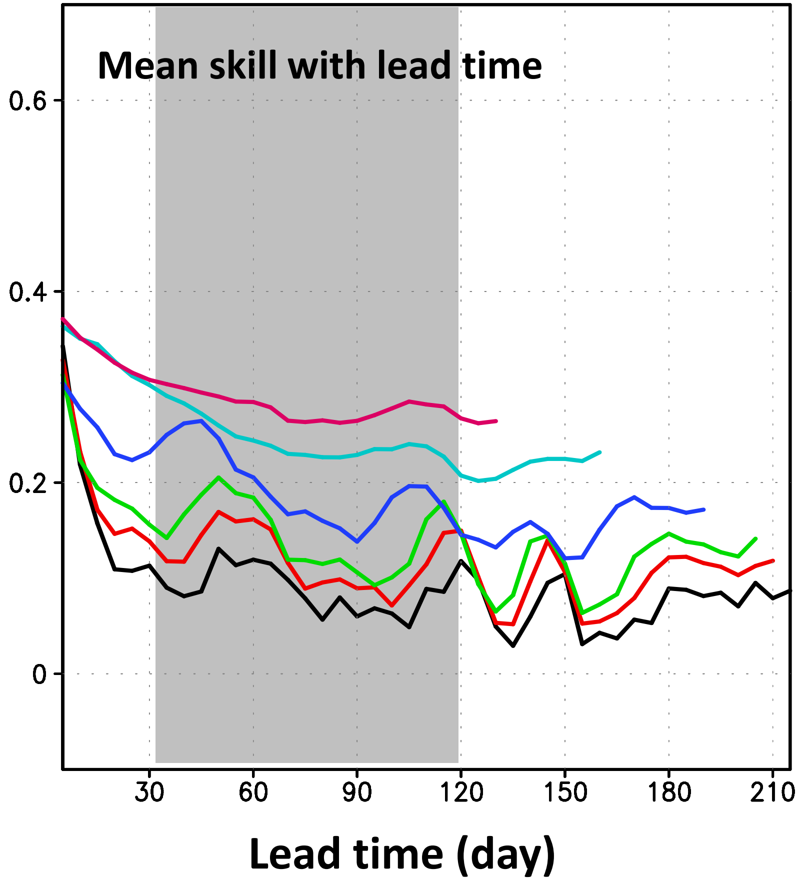
Prediction of

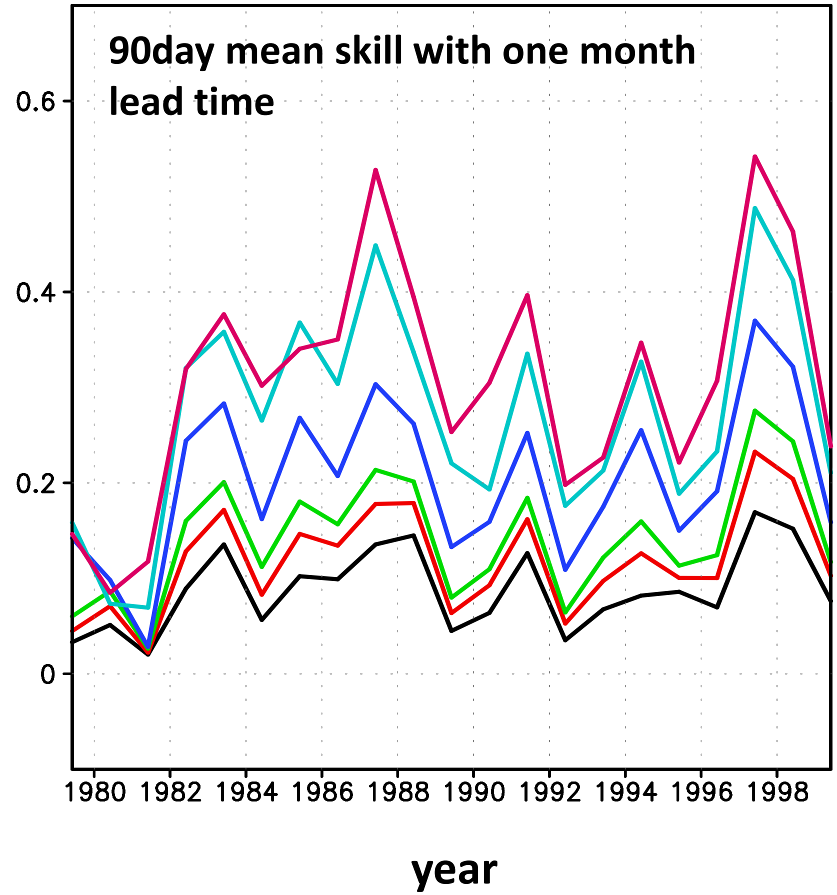
1. *Mean Temperature during a day*
2. *Mean Temperature during a month*
3. *Mean Temperature during a century*

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

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How long?

Time mean of weather

Climate prediction

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Seasonal forecast

Seasonal Prediction

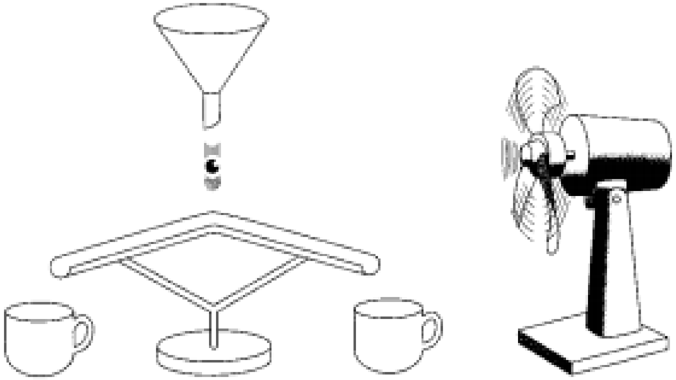
What : state of atmosphere during a season

Condition : Current state, Physical rules, external forcing factor

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

History of Short-term (Seasonal) Climate Prediction

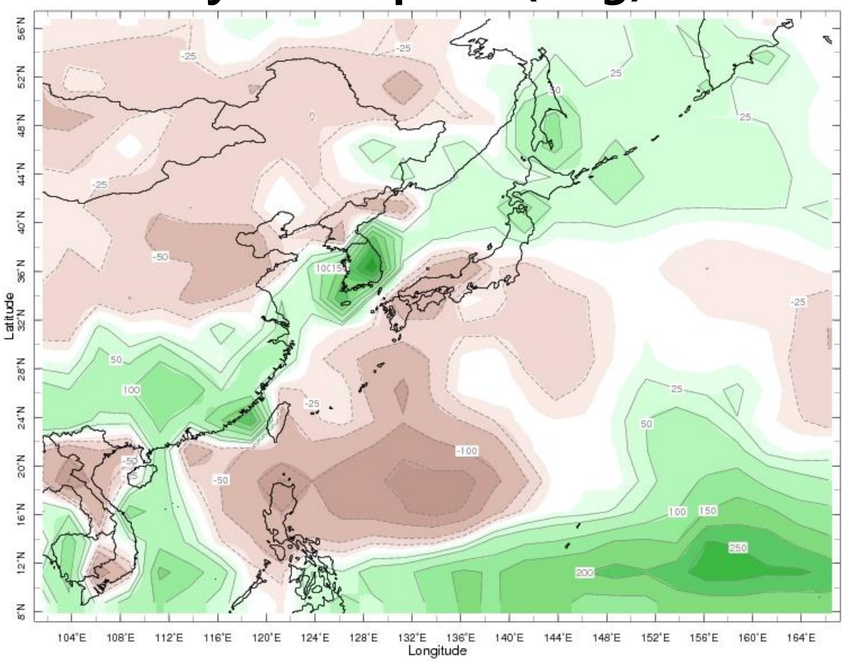
- 1960's : Hypothesis proposed
- 1980's : ENSO prediction + Atm. LFV. (PNA..)
- 1990's : (Experimental) Dyn. Seasonal Fcst.
- 2000's : International collaboration (MIPs)
- 2010's : Operation (GFCS, RCOFs/WMO)



T. Palmer (1998)

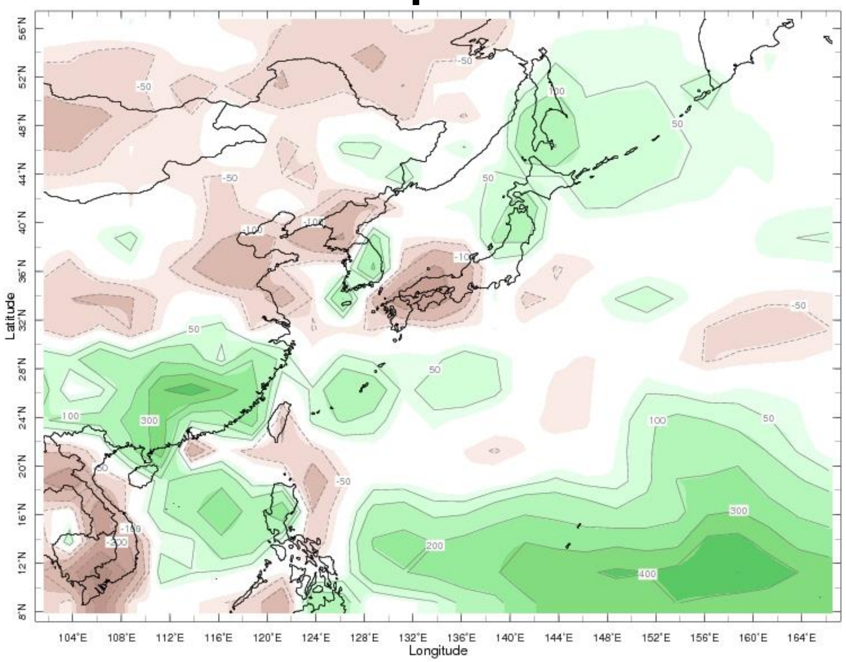
2002 summer rainfall

Monthly mean prec. (Aug)



Aug 2002

Summer mean prec.

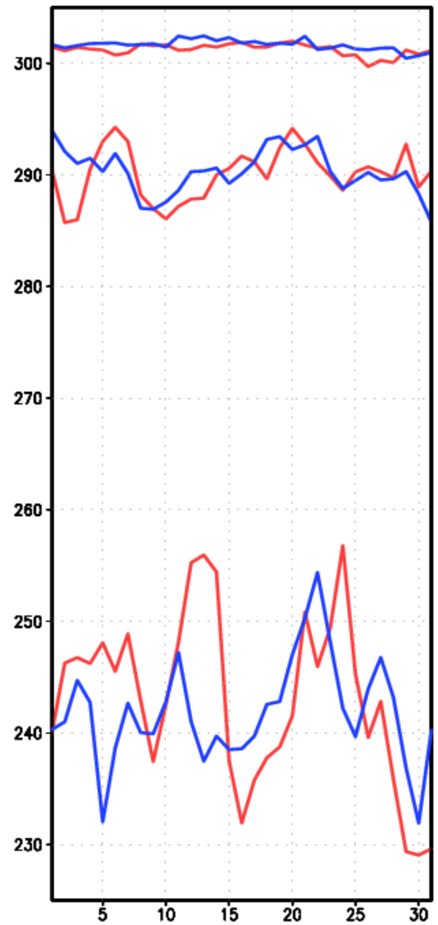


Jun-Aug 2002

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"?**

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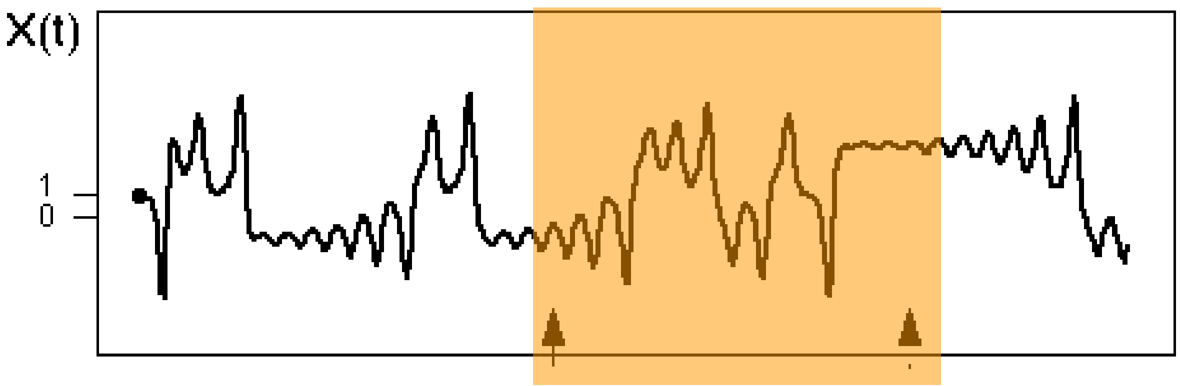
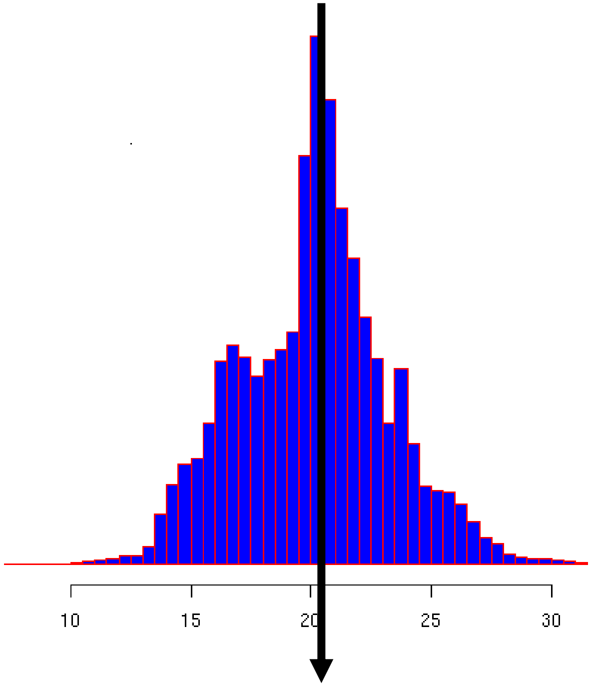
Weather statistics

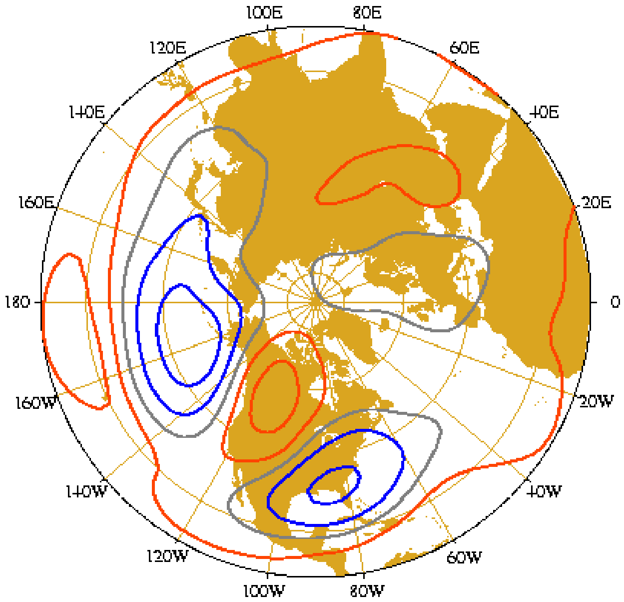




Seasonal mean

Primary seasonal weather statistics : seasonal mean

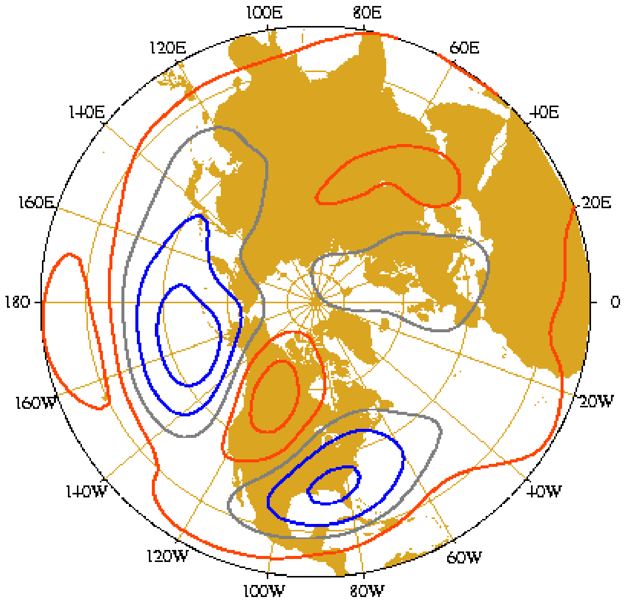
PNA debates



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

$$X = Xs + Xn$$

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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)

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Mechanisms of Variability

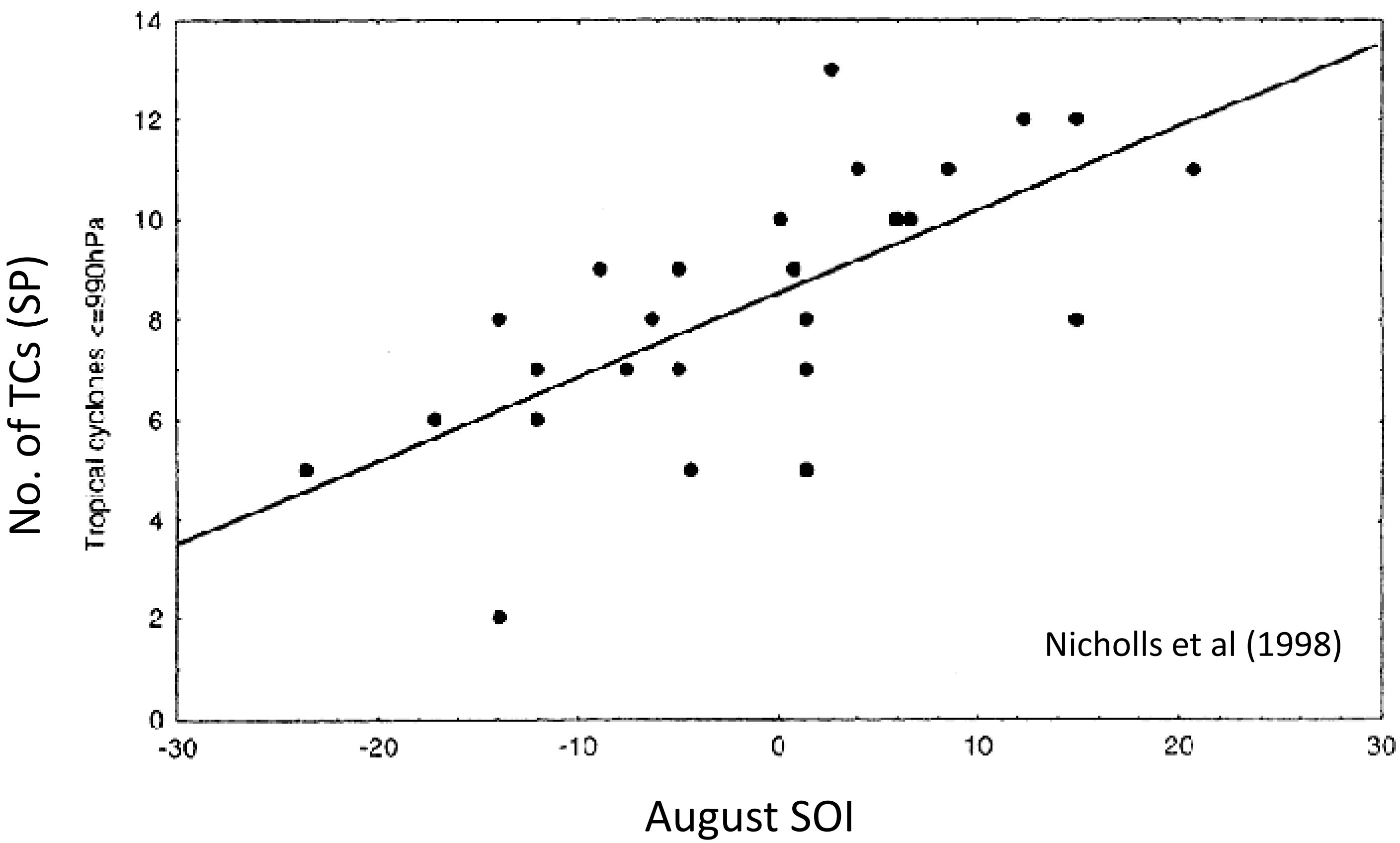
	Internal	External
Weather:	1. Internal Dynamics of Atmosphere	• Boundary Condition of SST, Soil wetness, Snow, Sea ice, etc.
Climate: (seasonal-decadal)	2. Internal Dynamics of Coupled Ocean-Land-Atmospshere	• Solar, Volcanoes
Climate Change:	3. Internal Dynamics of Sun-Earth System	• Human effects: (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

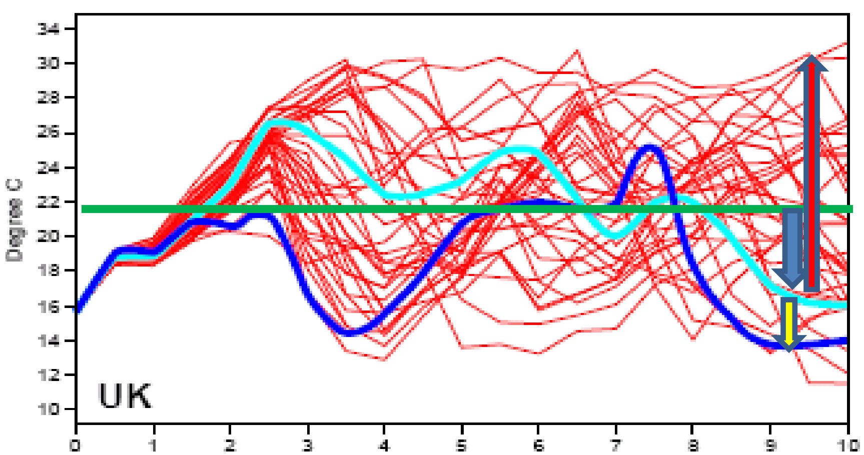
Two scales



No. of TCs (SP)
Tropical cyclones <=990hPa
14
12
10
8
6
4
2
0
-30
-20
-10
0
10
20
30
Nicholls et al (1998)
August SOI

Predictability

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

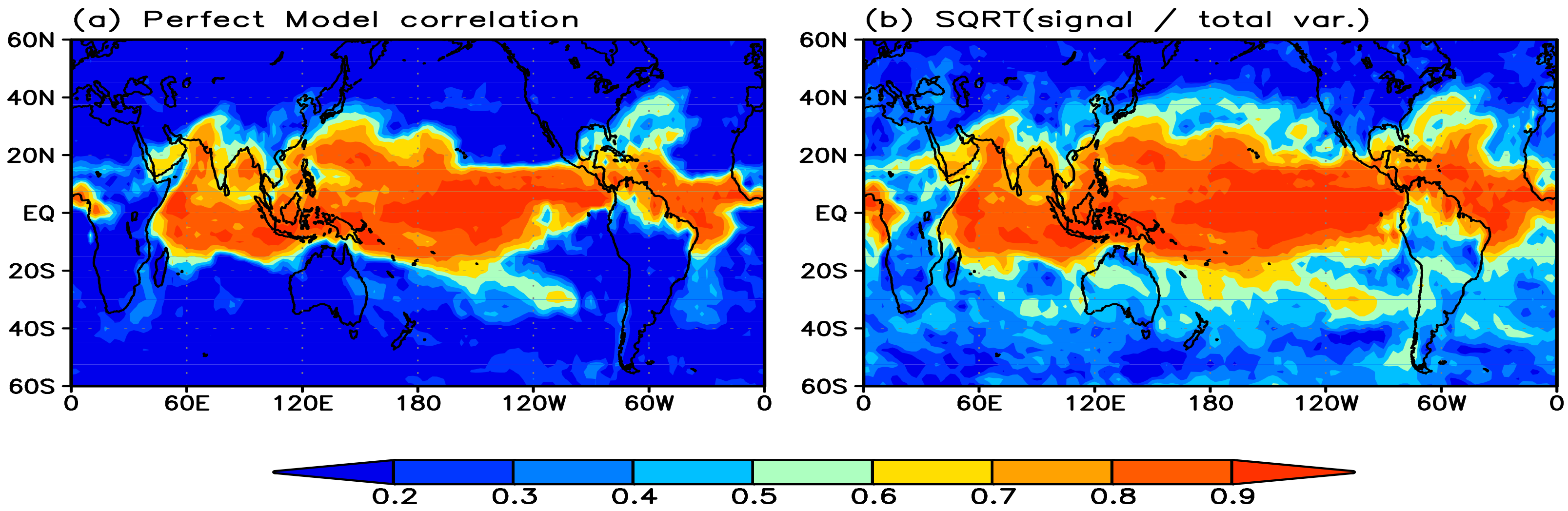


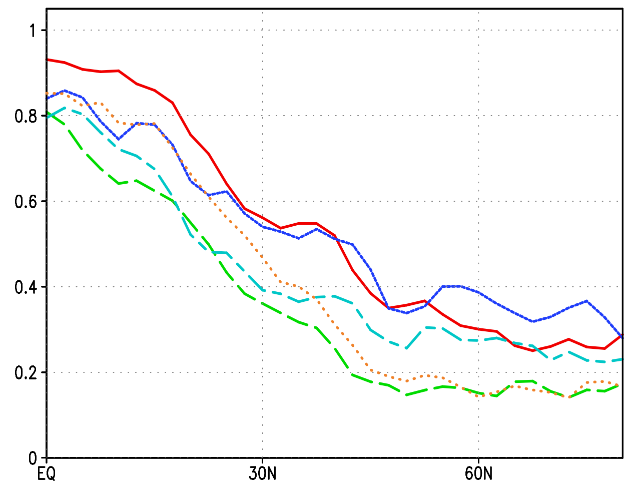
X=Xs+Xn

Xs : ensemble mean

Xn : 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

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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%)

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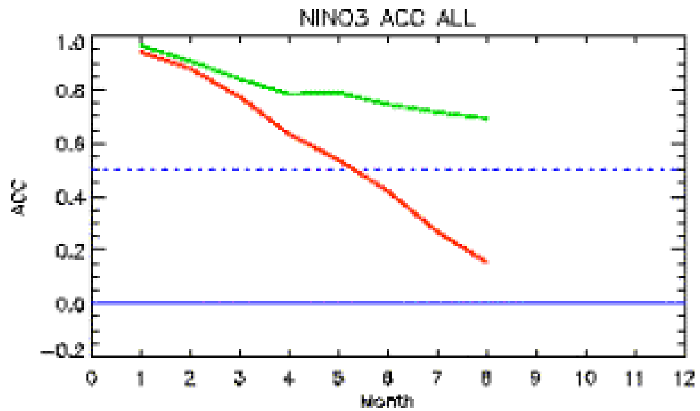
Statistical forecasting

- (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)



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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)

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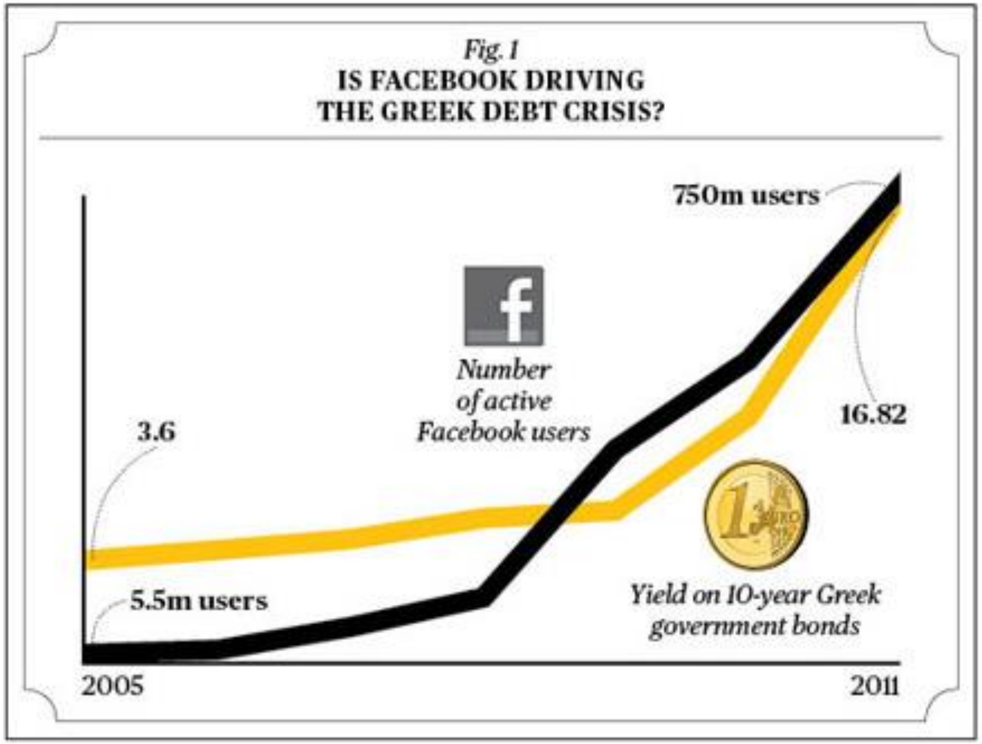
Predict yield of Greek bonds with Facebook users

- Is it appropriate?



From *business week*

If yes, why?

If not, why?

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

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

$$x'(t+1) = a_1 y_1(t) + a_2 y_2(t) b$$

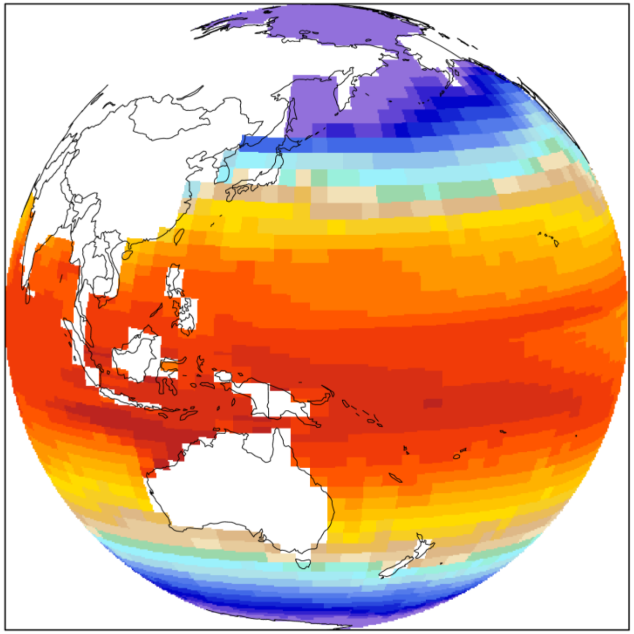
$$x_1'(t+1) + x_2'(t+1) = a_1 y_1(t) + a_2 y_2(t) b$$

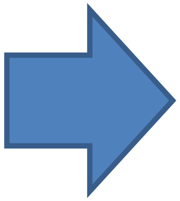
- One to One : often not very satisfactory, limited cases
- One to Multi : easy to overfit (lie)
- Multi to Multi : looks nice but often produce nothing practical

- 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)

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

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

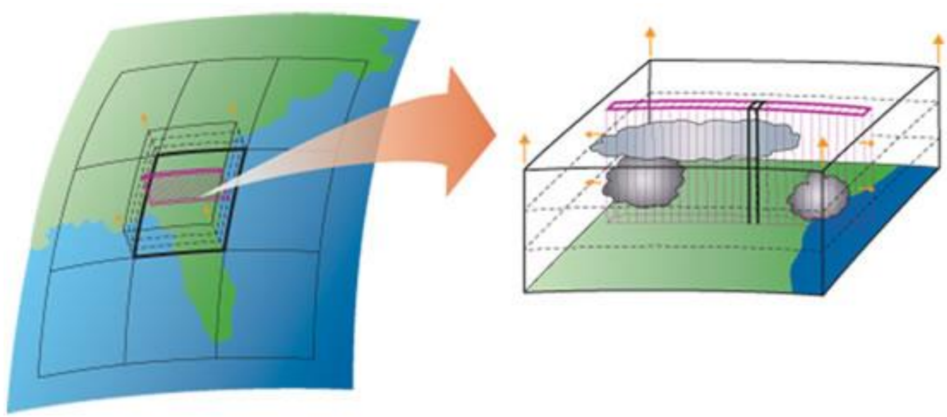
$$\frac{\partial \theta}{\partial t} + \vec{u} \cdot \vec{\nabla}\theta = l$$

⋮

```
//Belehradek functions:
//
//     DTstage(T+9.11)^-2.05
//
    MinTime=pow(T[j]+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 4oC.
//
//b=[ones(1,6)*params.bnaup,ones(1,6)*params.bcop];

    for(k=0;k<6;k++)
        Rfood[k]=(1.-exp(-(F[j]-c)*params.bnaup));
    for(k=6;k<12;k++)
        Rfood[k]=(1.-exp(-(F[j]-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(T[j]/Tc,z);
    //gammaT=0.1; //Overrride temp dependent mortality
```

Numerical modeling

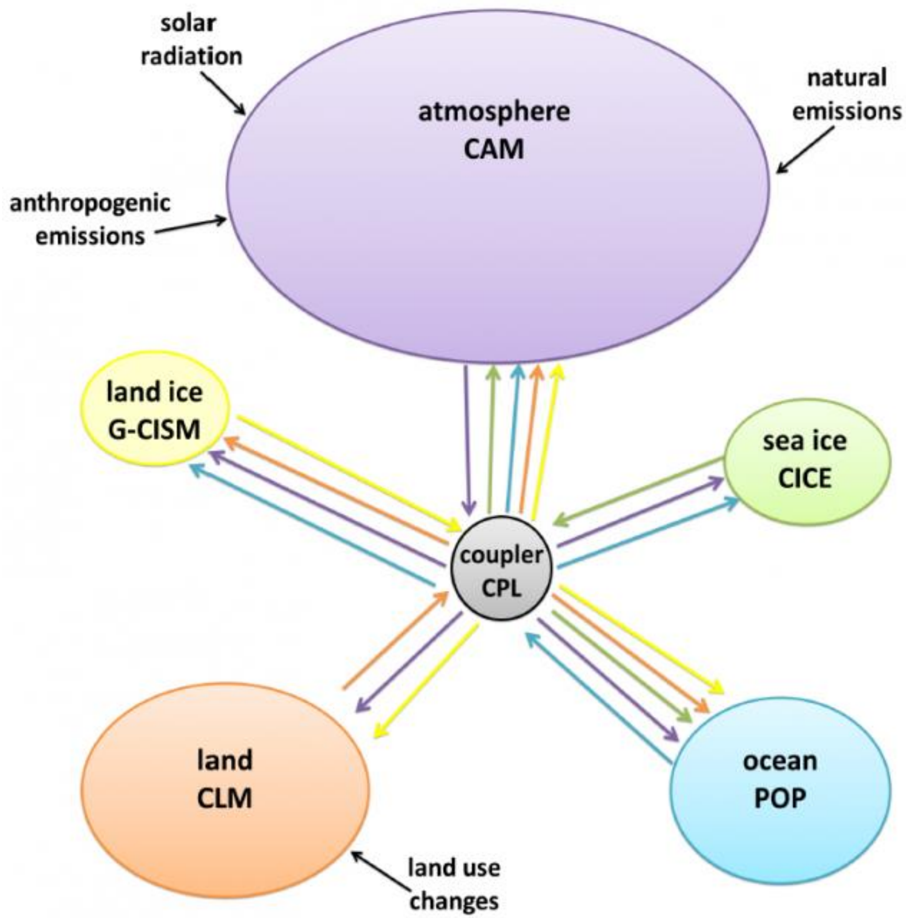
- Issue
  - Digitization (physical variable is continuous, but computer needs digitization”
    - Resolution, subgrid-scale parameterization



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

GCMs

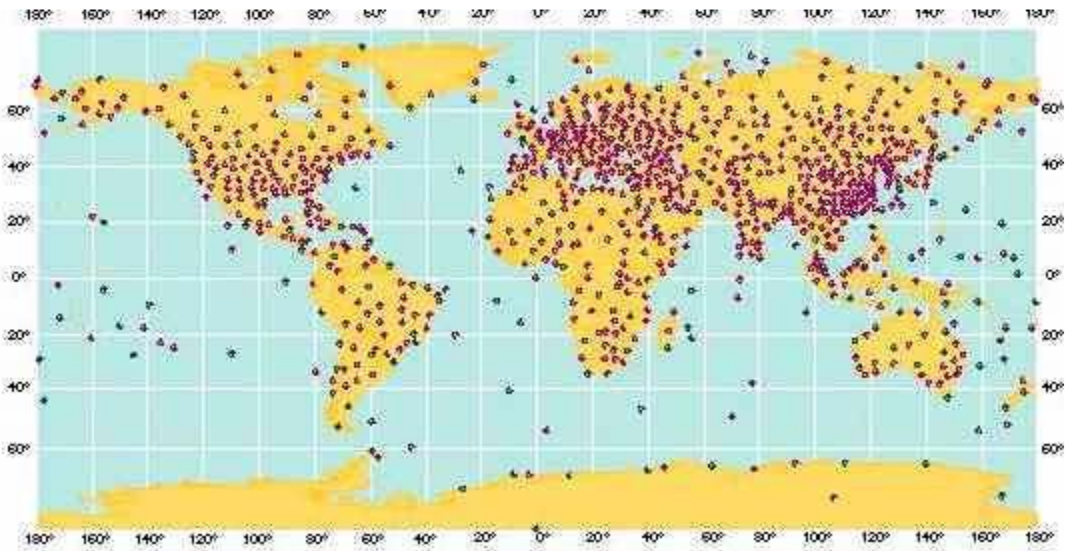
- Coupled GCM
  - Atmosphere
  - Ocean
  - Sea-Ice
  - Land surface
  - Chemistry
  - Biosphere

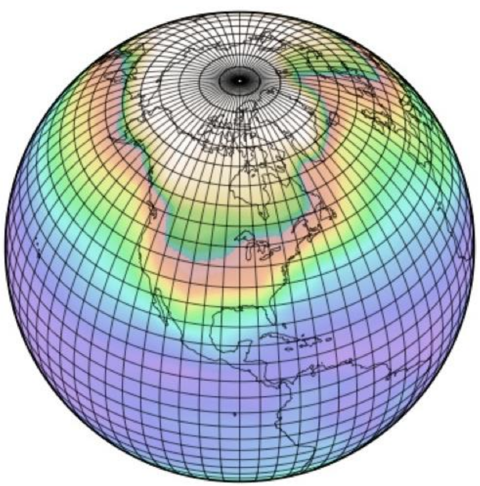


Initialization

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)





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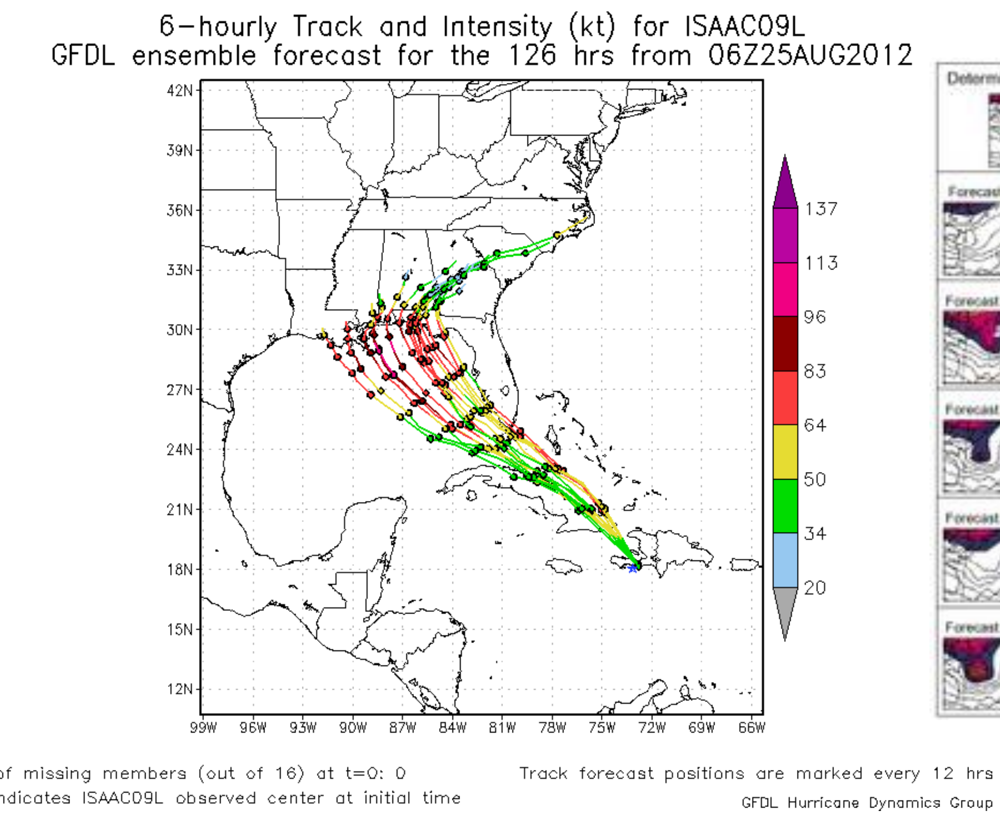
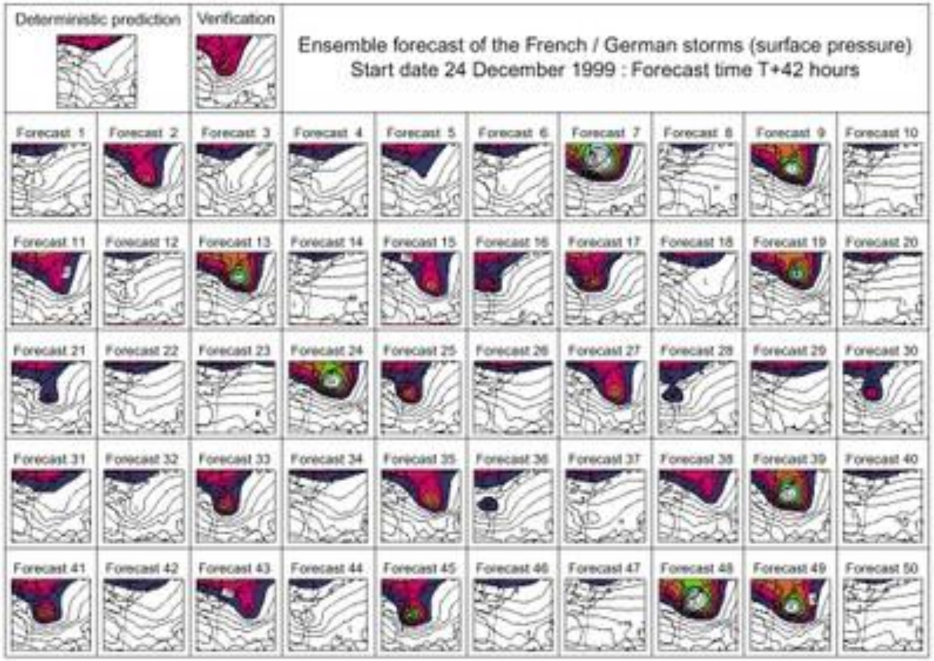
Ensemble Forecasting

- Run many times
  - Starts from slightly different initial conditions





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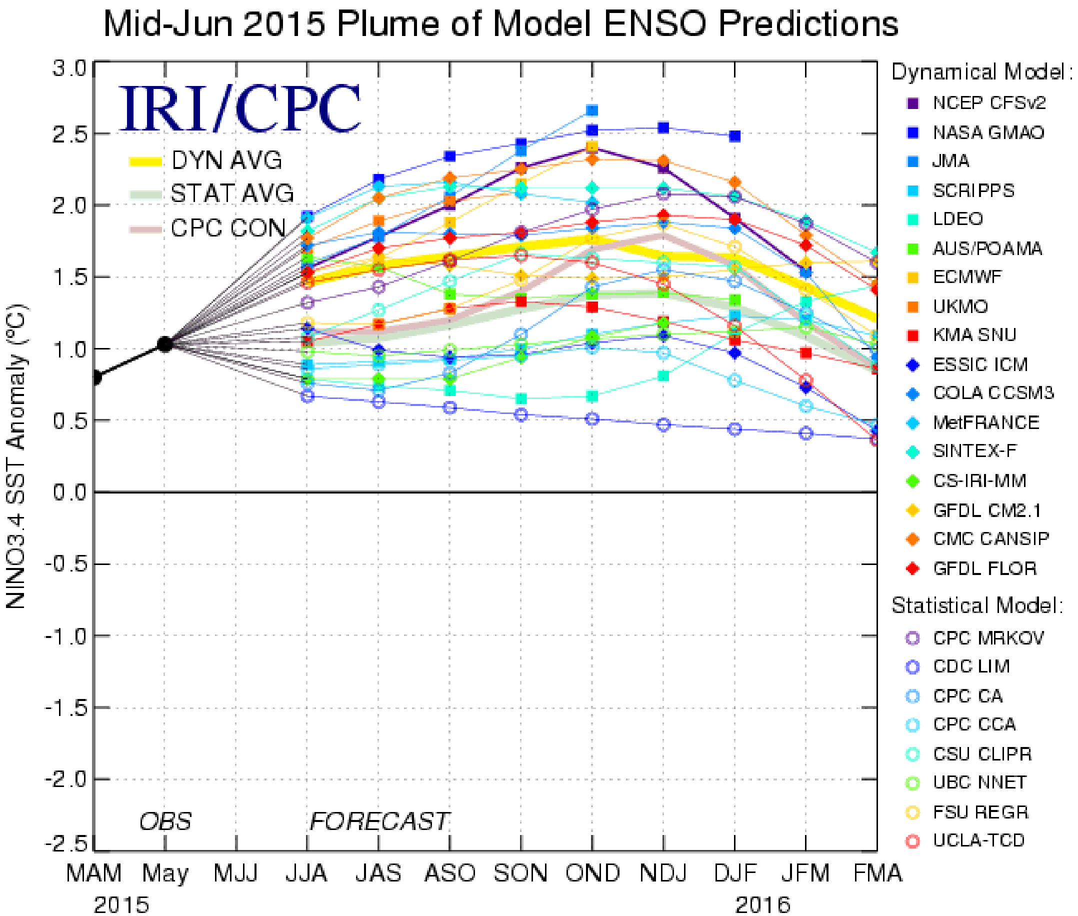
Multi Model Ensemble Forecasting

- Run with many models



Which one??

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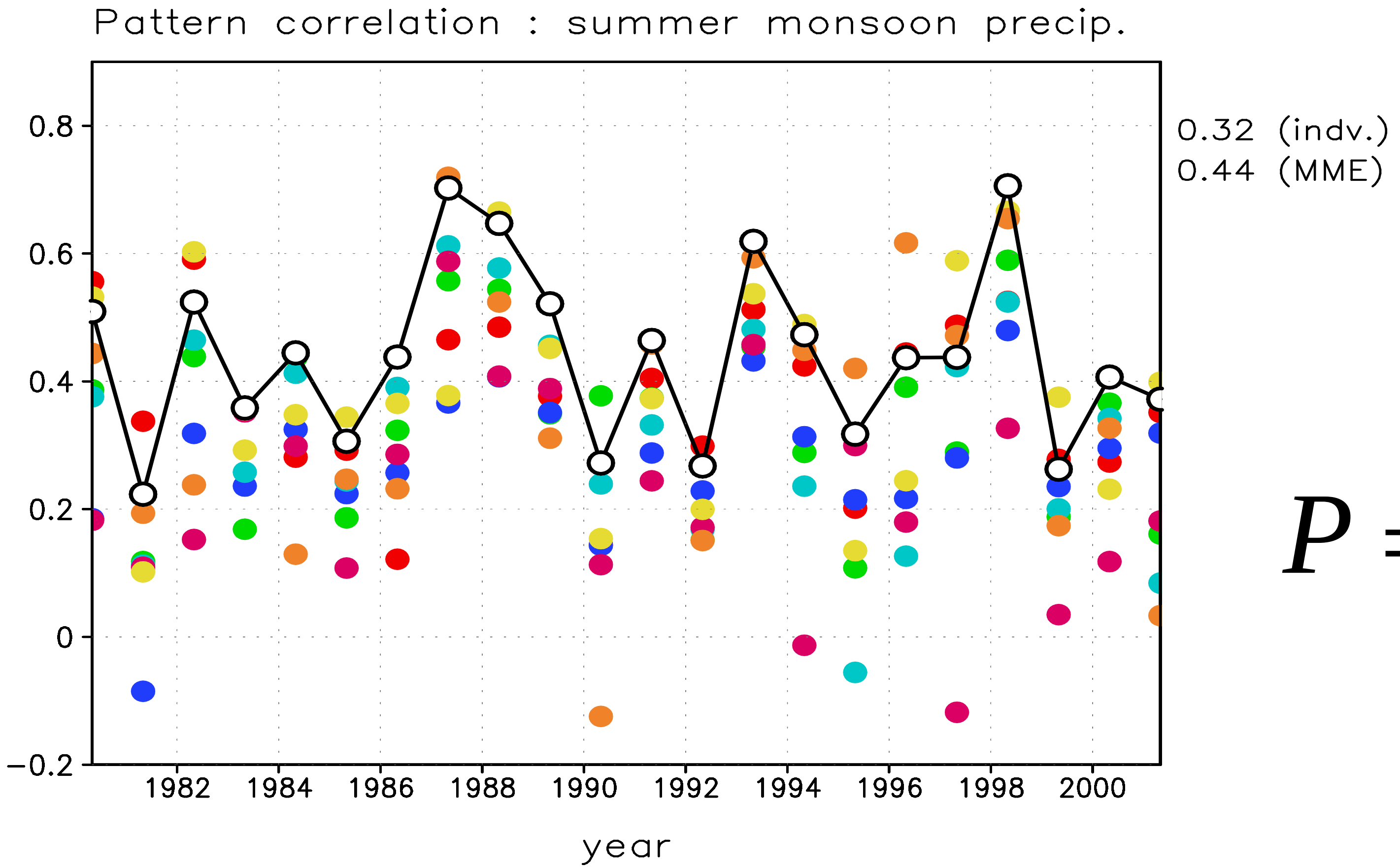
Use all!



$$P = \sum_i a_i F_i$$

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

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

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

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

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

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

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

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

$$Observation: x = x_s + x_n$$

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

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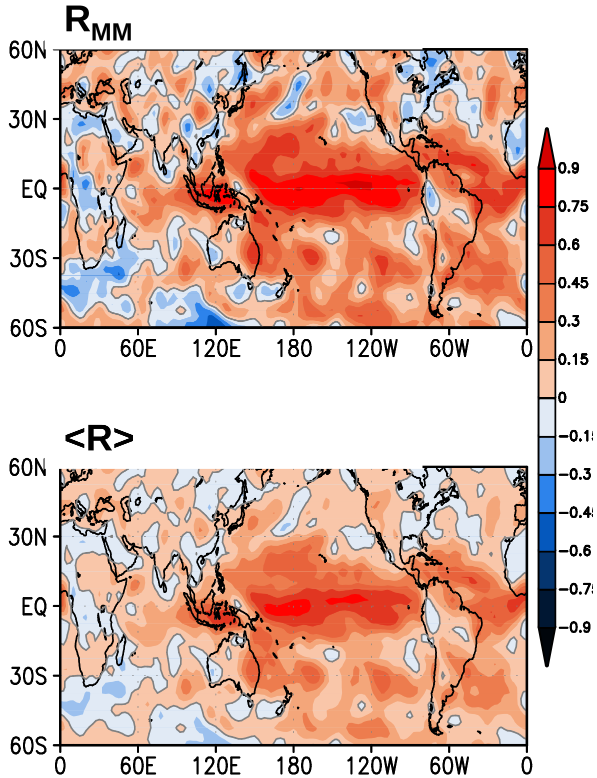
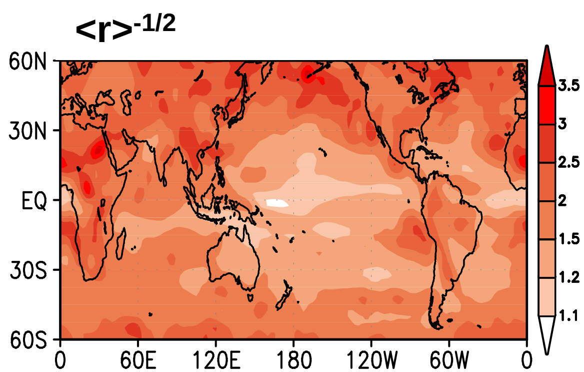
Temporal correlation skill (SUMMER MEAN PRCP)

Multi-model ensemble correlation skill



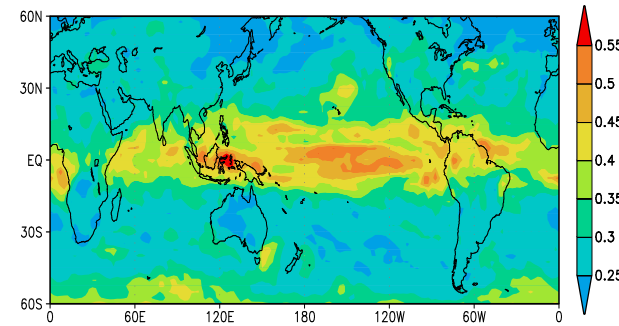
Mean correlation skill of individual models

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)

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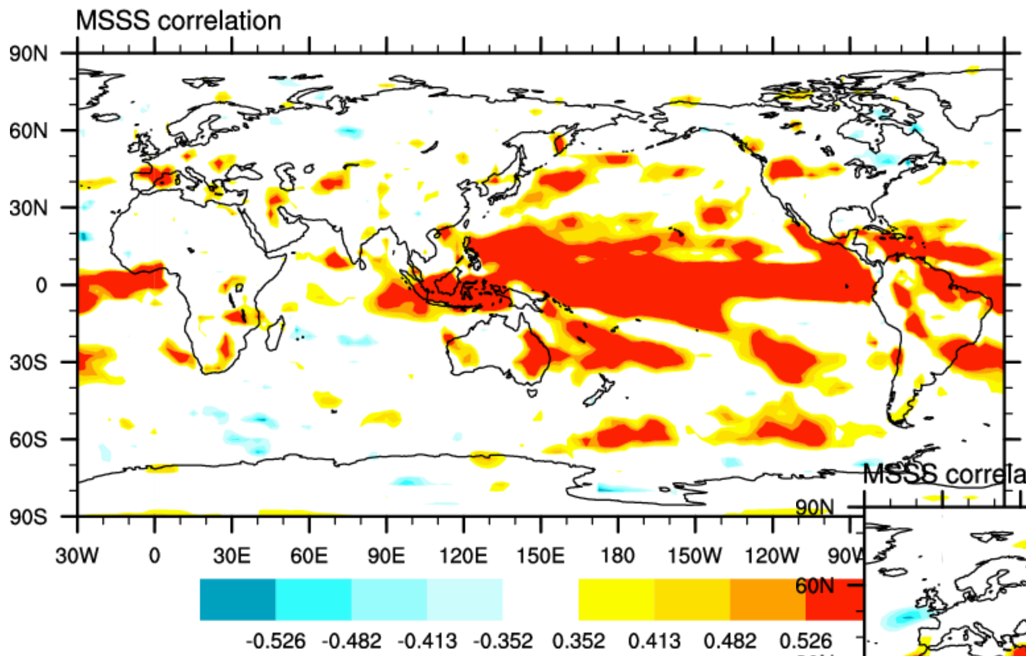
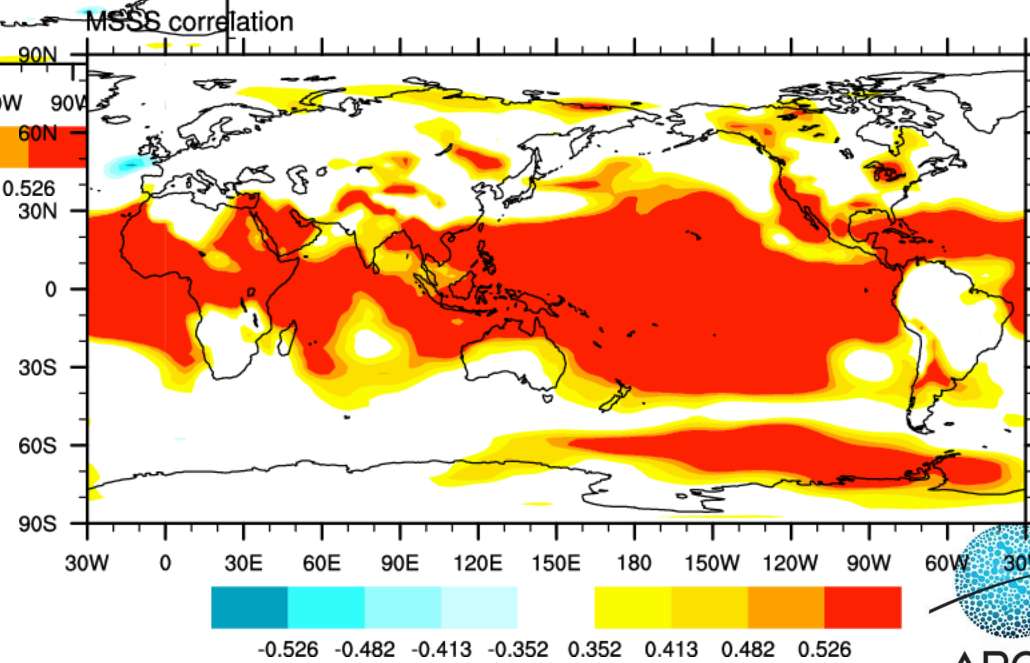
APCC MME (TCC)

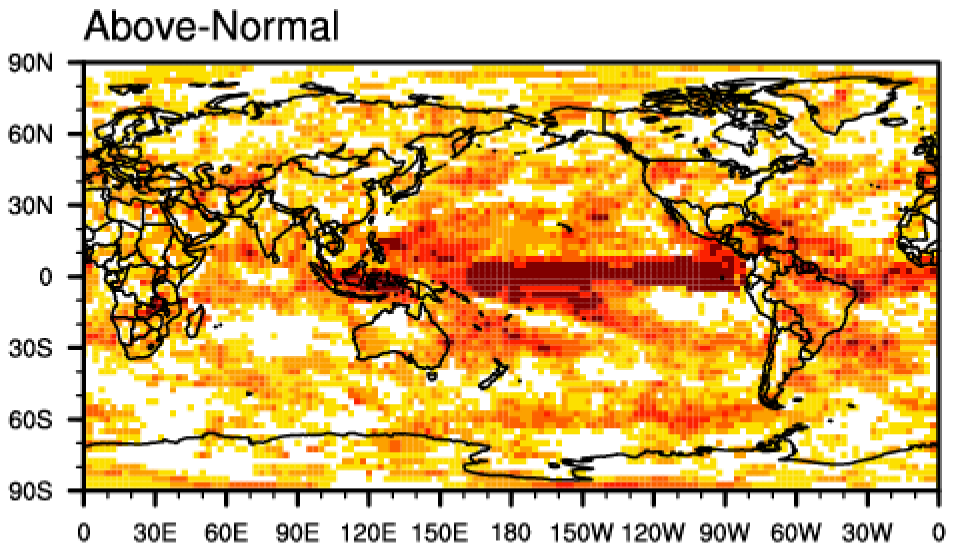


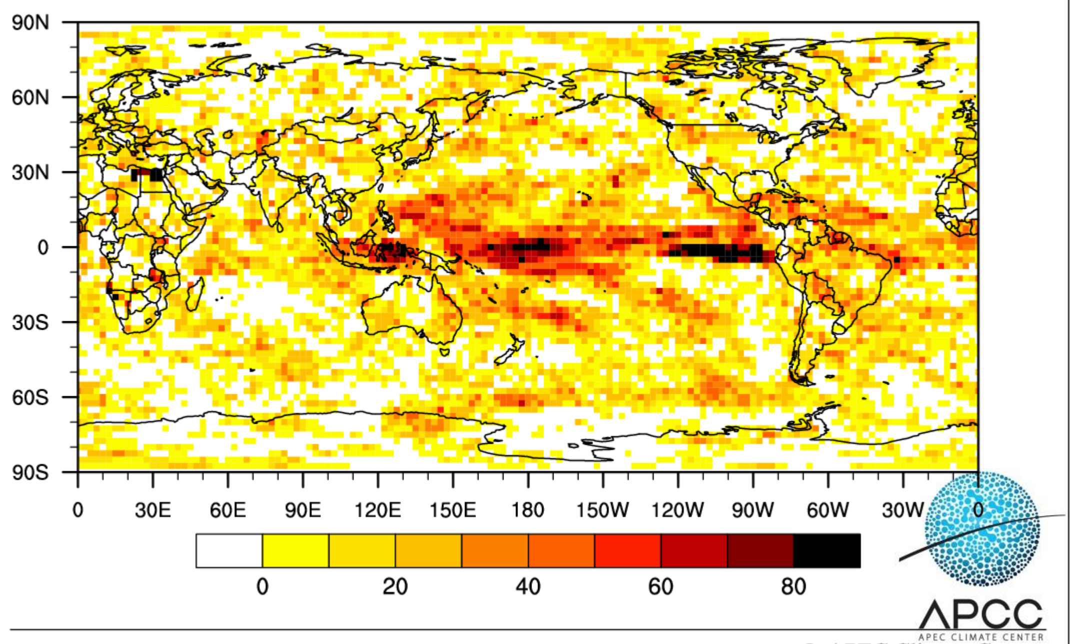
Rainfall (JJA)



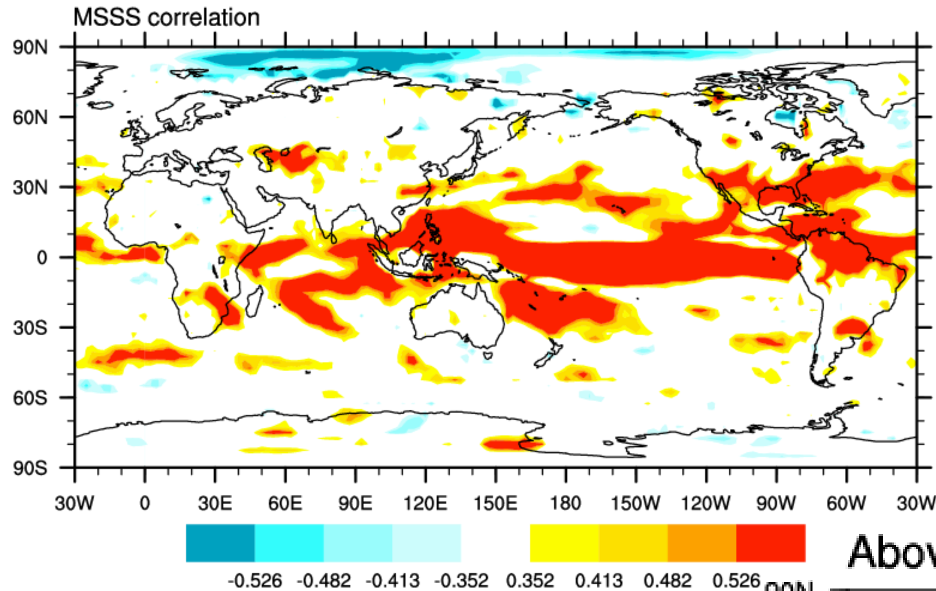
SLP (JJA)

ROC Score : PREC, JJA (1983-2005)



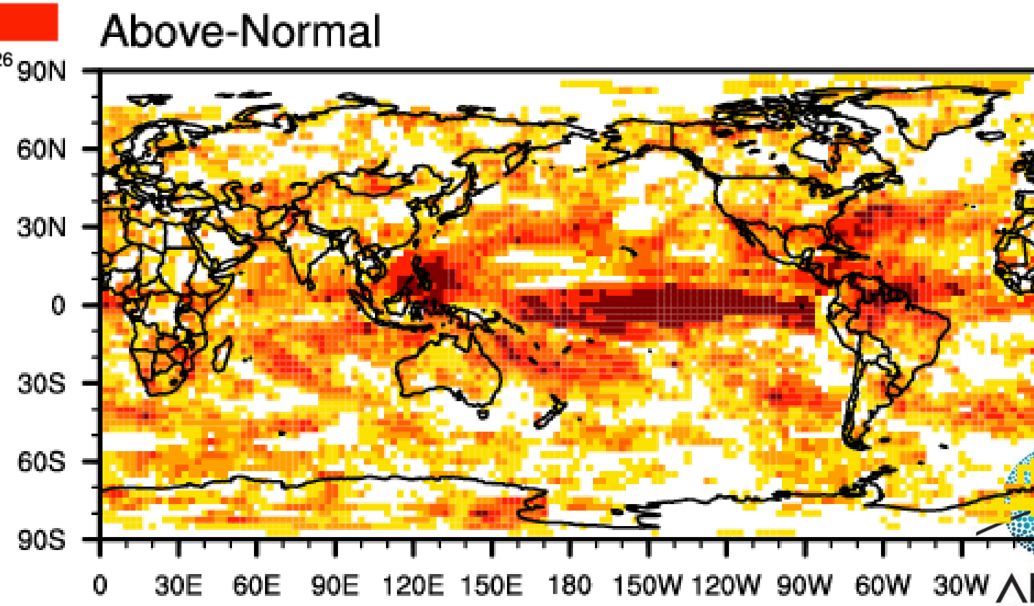


DJF season



Rainfall (JJA)

ROC score
Rainfall (DJF)



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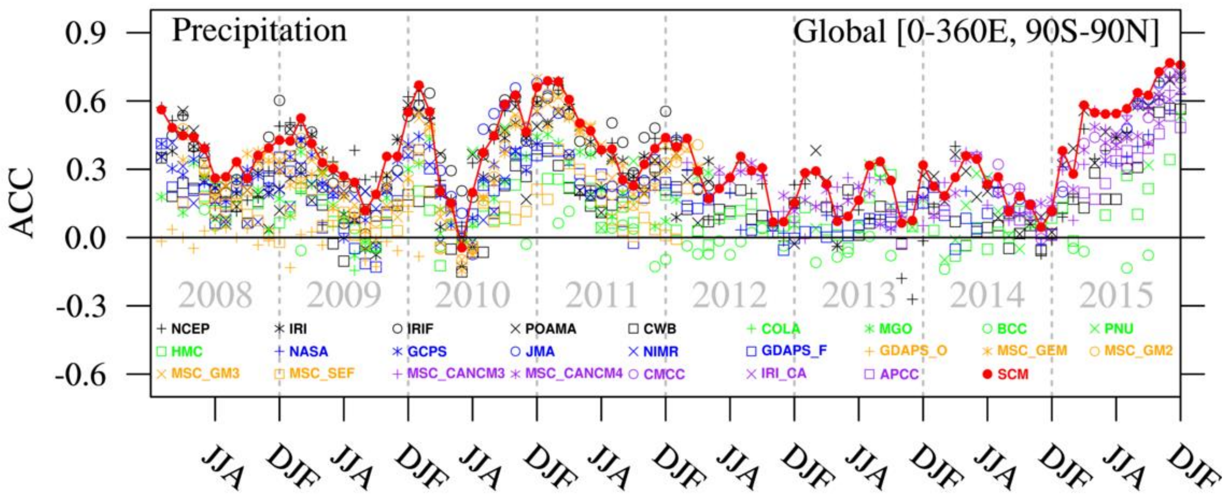
APCC operational forecast



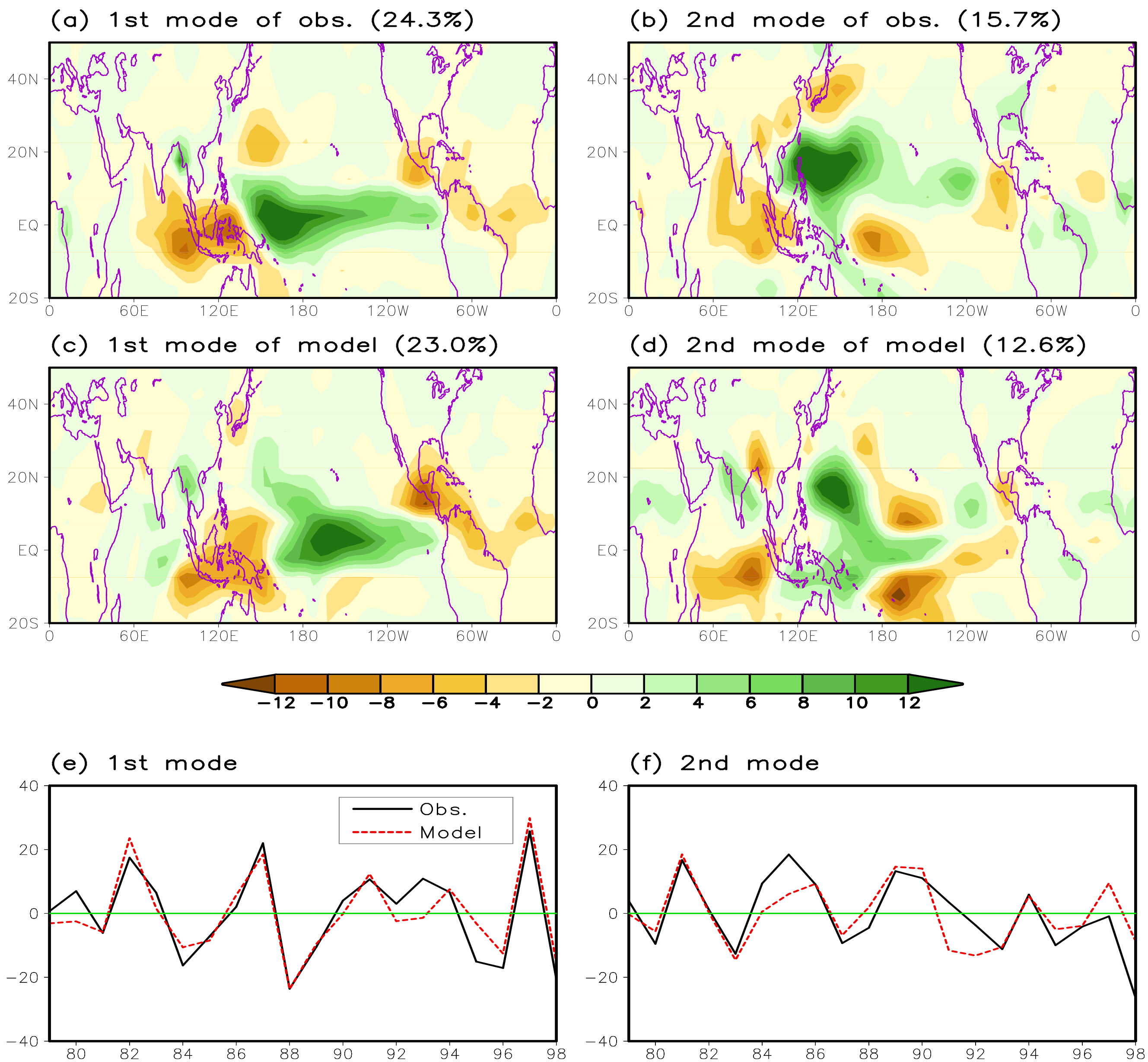
	APCC	WMOLC	ECMWF	NCEP	UKMO	JMA
AN	0.569457	0.541897	0.535531	0.52996	0.528975	0.531497
NN	0.520962	0.521424	0.537661	0.519823	0.524022	0.514656
BN	0.567702	0.533777	0.516511	0.535767	0.516994	0.534244

Realtime rainfall forcast 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

EOFs of Summer Mean Precipitation



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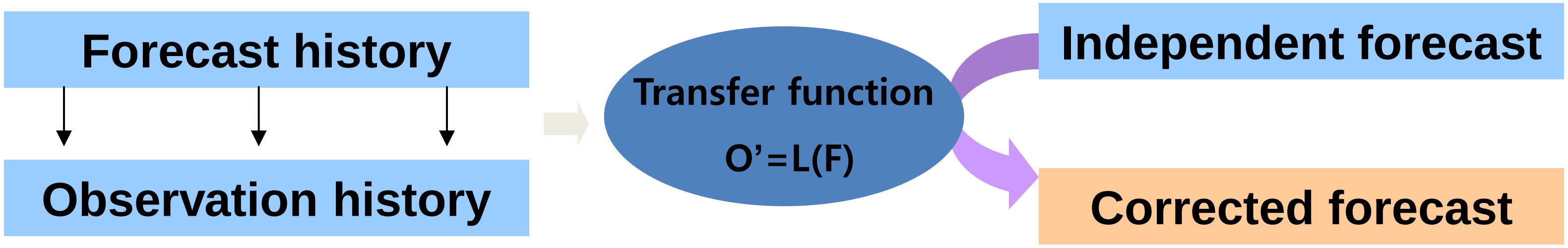
Statstical downscaling : CLIK



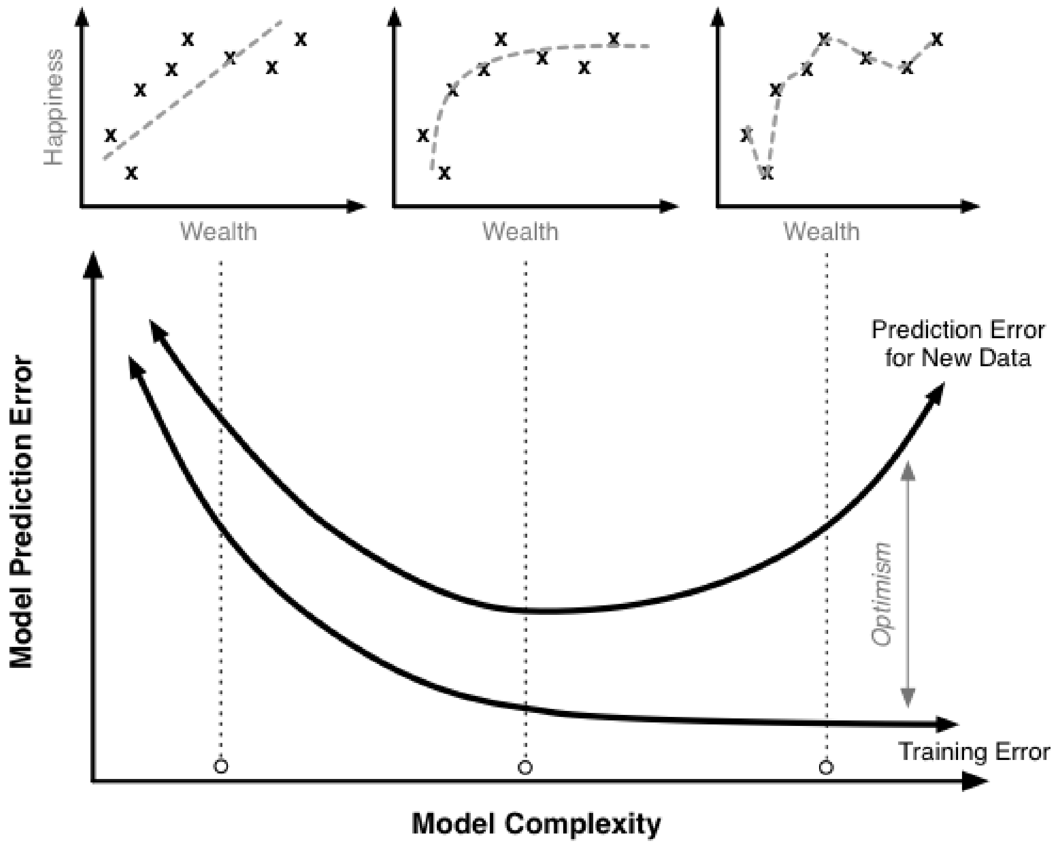
There are many approaches in post-process, All of them share similar assumption. :
Statistics between forecast and observation is stationary

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**



from Scott Fortmann-Roe 2012

If model output is fitted to the unpredictable noise : Overfitting.

What if we remove "noise" in the observation?

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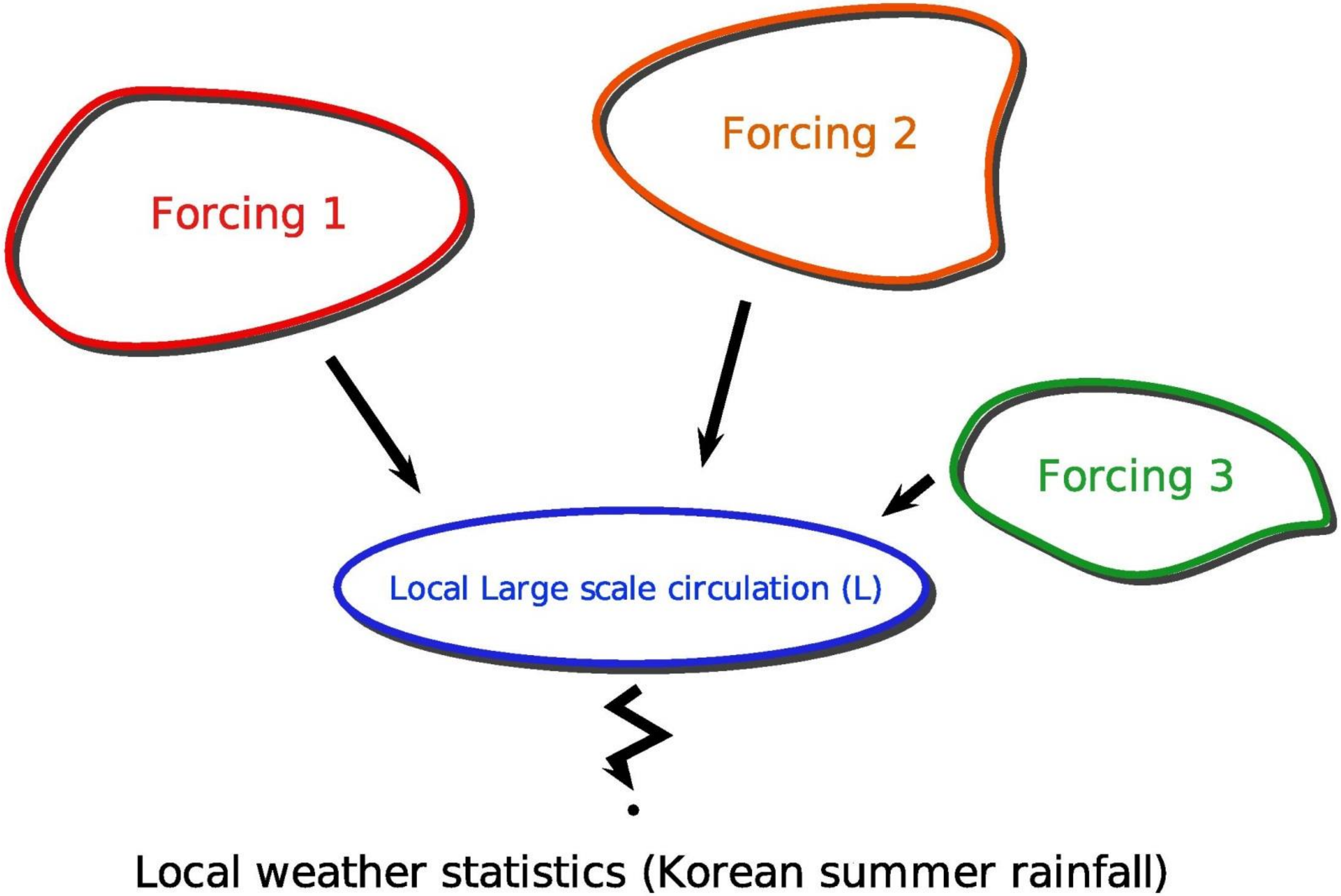
Downscaling (post-process) should be based on

- 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 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

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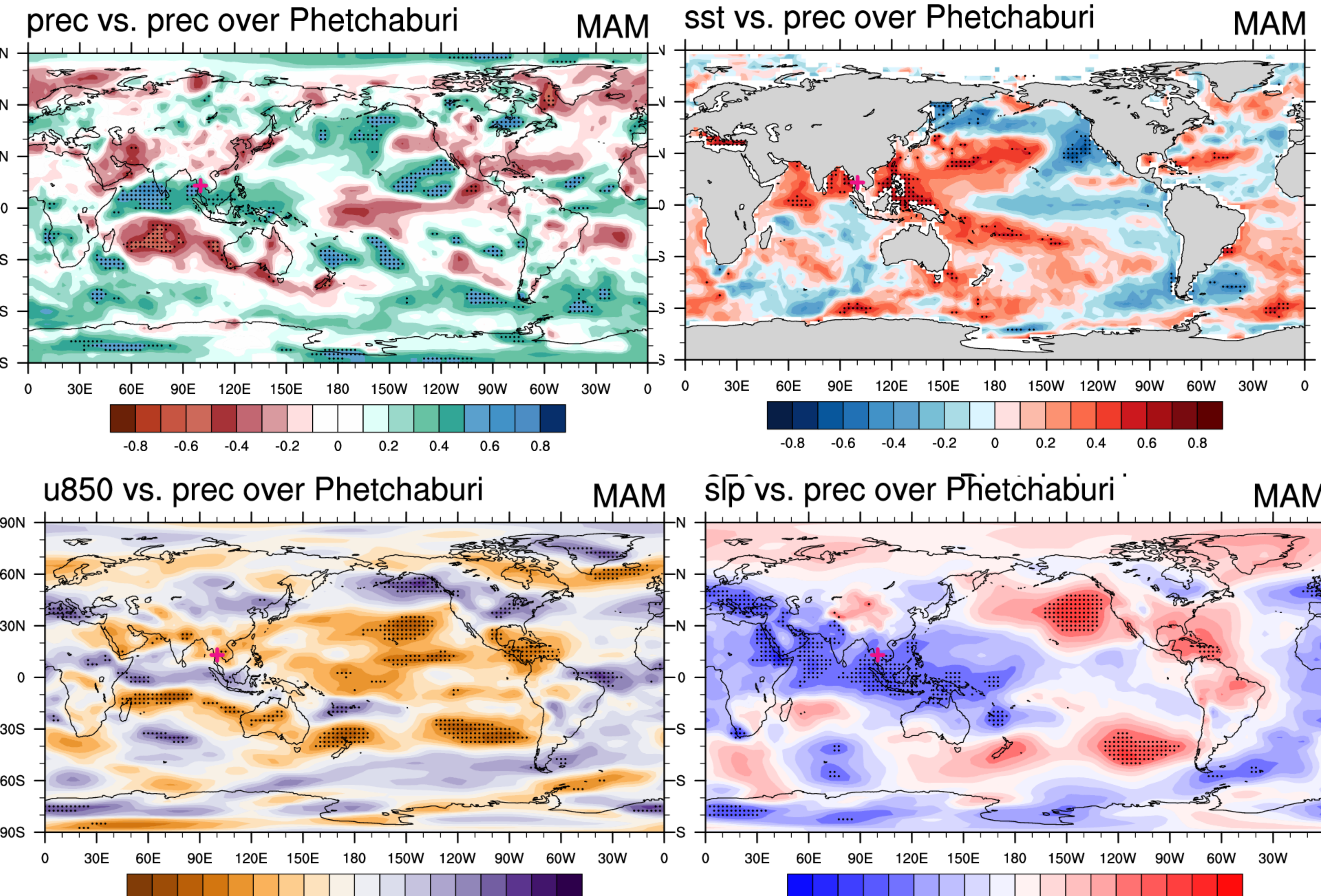
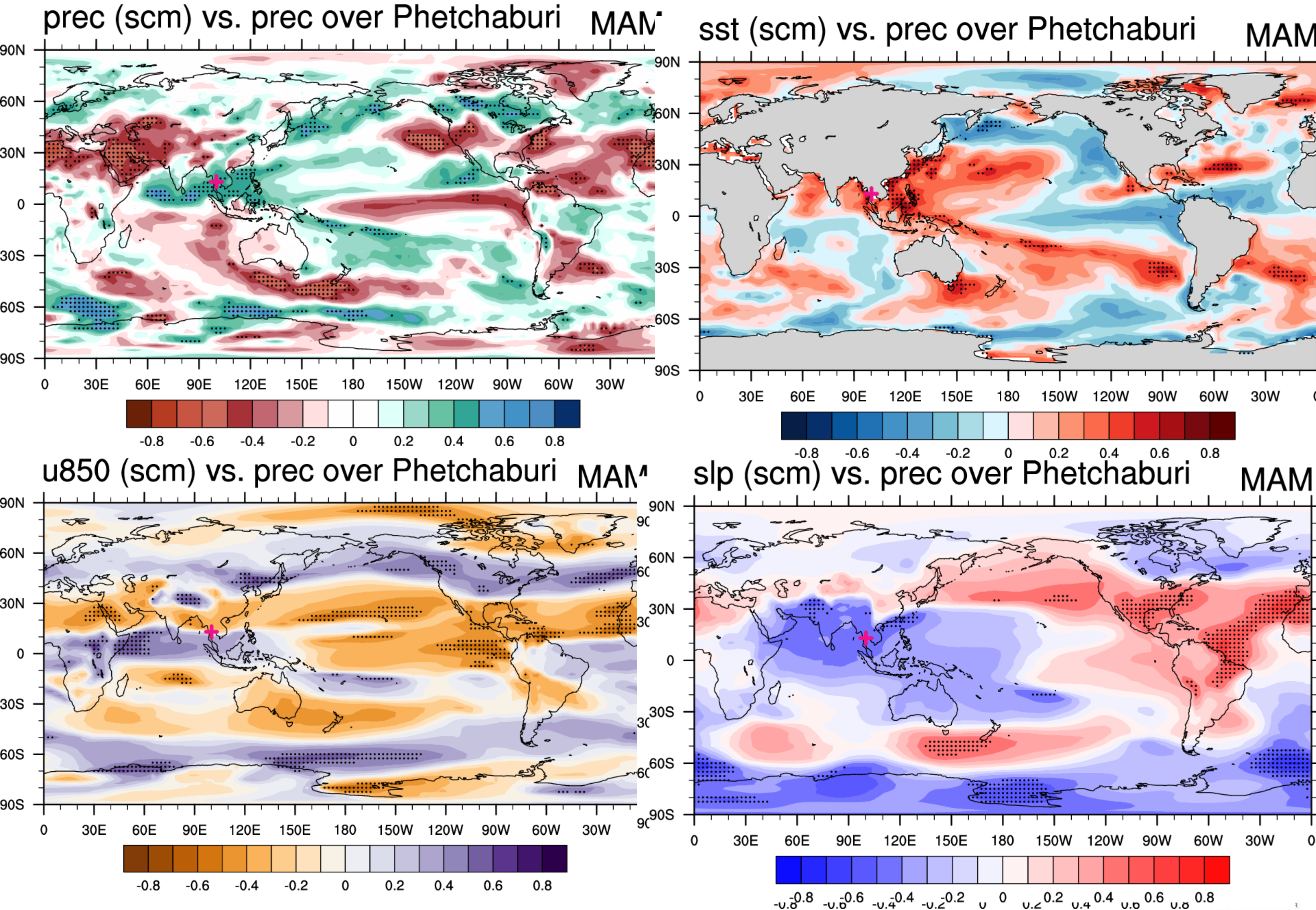
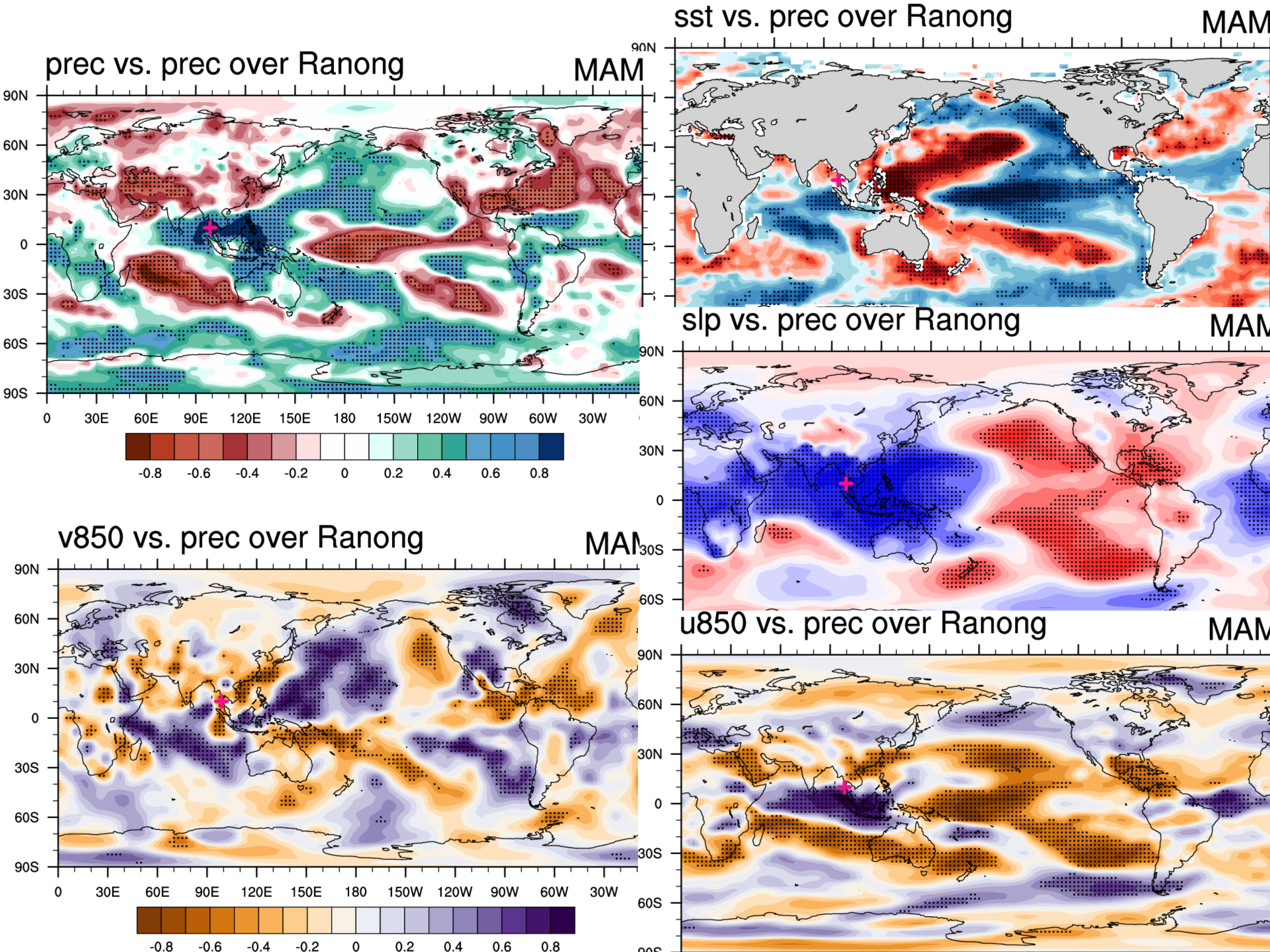
LARGE SCALE PATTERN ASSOCIATED WITH RAINFALL

Local large circulation and Teleconnection

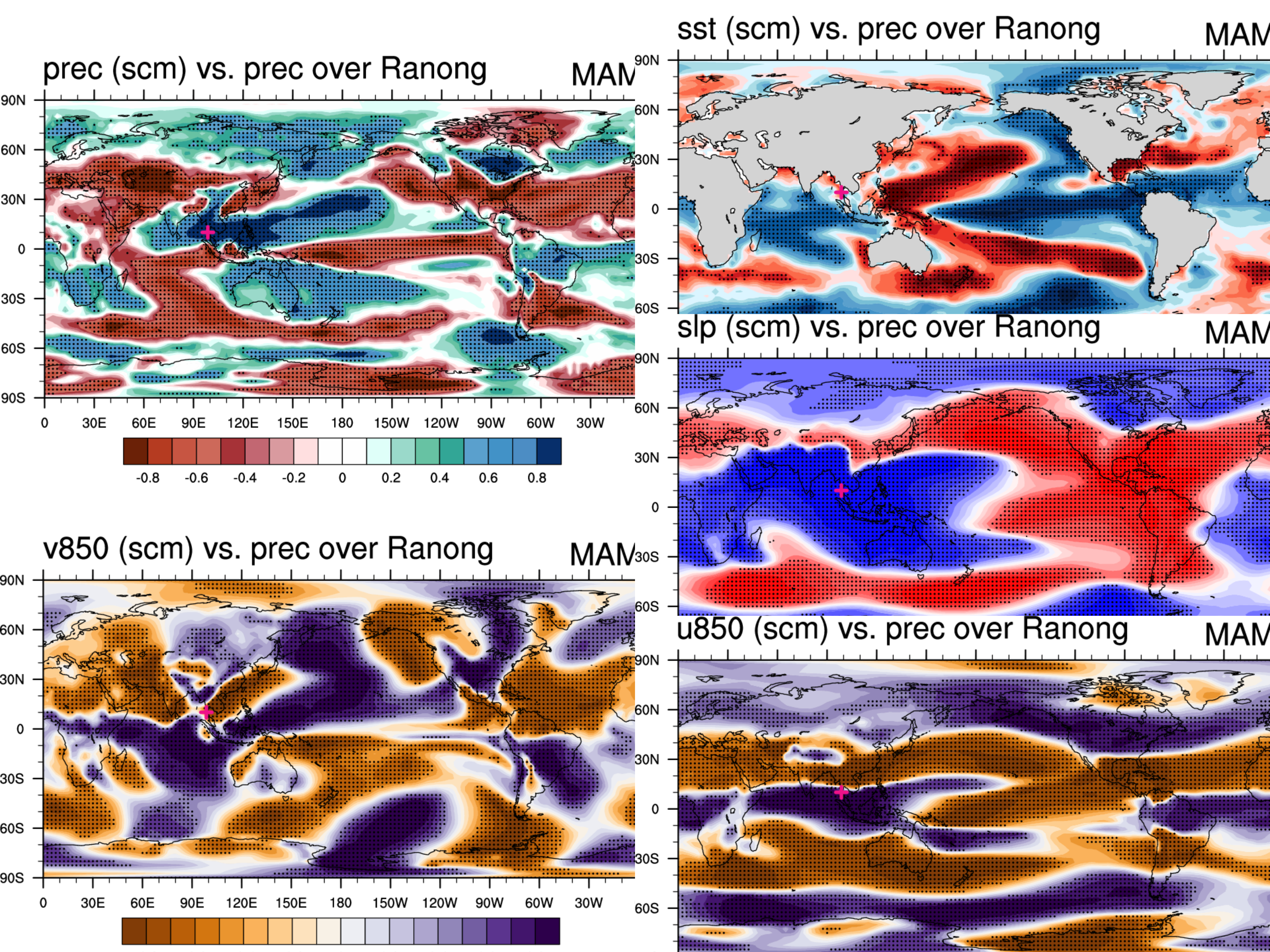
One Point Correlation map with seasonal mean local rainfall with other variables


prec vs. prec over Phetchaburi
MAM
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
0
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
sst vs. prec over Phetchaburi
MAM
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
0
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
u850 vs. prec over Phetchaburi
MAM
90N
60N
30N
0
30S
60S
90S
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
0
slp vs. prec over Phetchaburi
MAM
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W

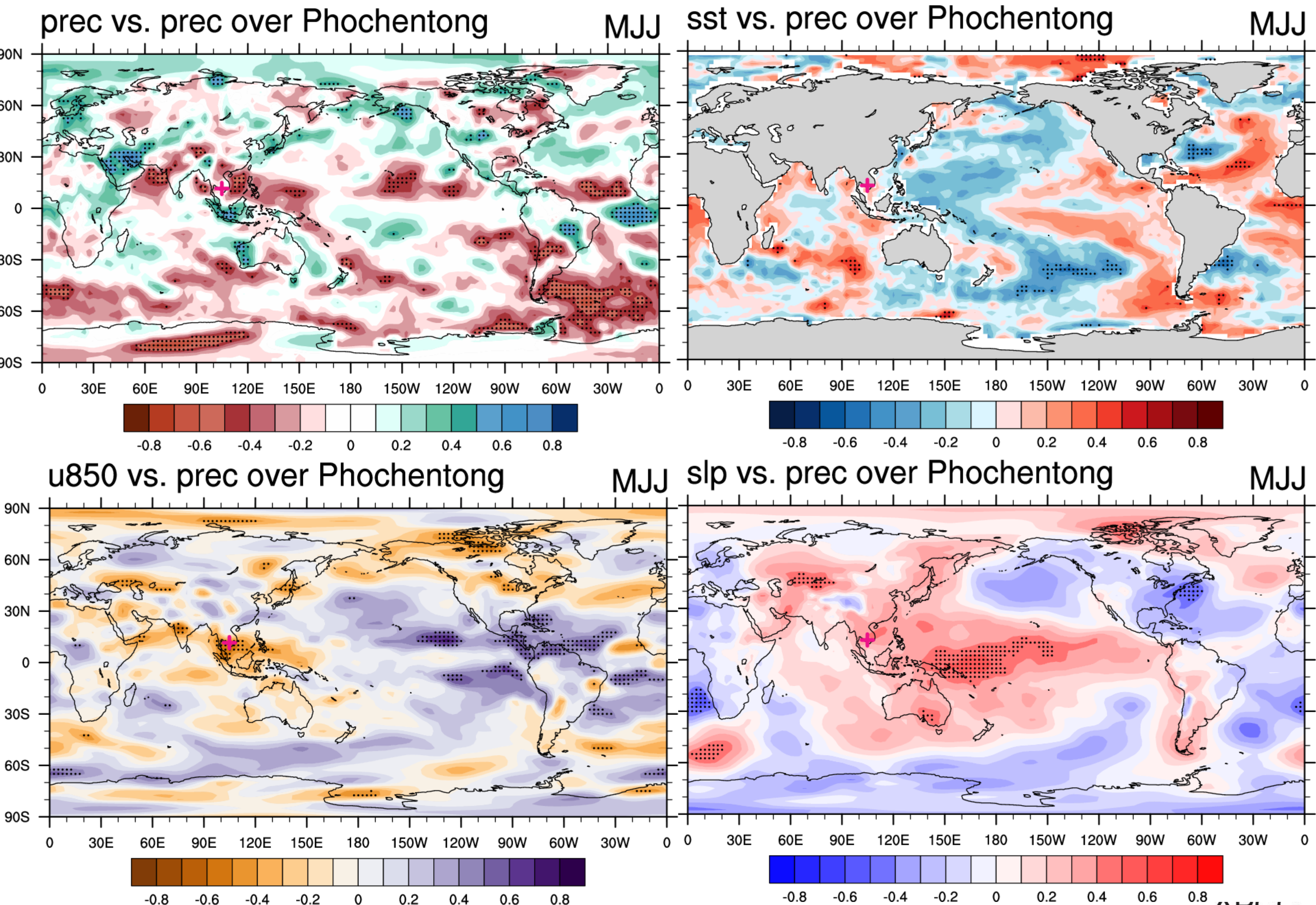

prec (scm) vs. prec over Phetchaburi
MAM
90N 60N 30N 0 30S 60S 90S
0 30E 60E 90E 120E 150E 180 150W 120W 90W 60W 30W
-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8
sst (scm) vs. prec over Phetchaburi
MAM
90N 60N 30N 0 30S 60S 90S
0 30E 60E 90E 120E 150E 180 150W 120W 90W 60W 30W
-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8
u850 (scm) vs. prec over Phetchaburi
MAM
90N 60N 30N 0 30S 60S 90S
0 30E 60E 90E 120E 150E 180 150W 120W 90W 60W 30W
-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8
slp (scm) vs. prec over Phetchaburi
MAM
90N 60N 30N 0 30S 60S 90S
0 30E 60E 90E 120E 150E 180 150W 120W 90W 60W 30W
-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8



prec vs. prec over Ranong
MAM
sst vs. prec over Ranong
MAM
slp vs. prec over Ranong
MAM
v850 vs. prec over Ranong
MAM
u850 vs. prec over Ranong
MAM
90N 60N 30N 0 30S 60S 90S
0 30E 60E 90E 120E 150E 180 150W 120W 90W 60W 30W
-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8



prec (scm) vs. prec over Ranong
MAM
90N
60N
30N
0
30S
60S
90S
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
v850 (scm) vs. prec over Ranong
MAM
sst (scm) vs. prec over Ranong
MAM
slp (scm) vs. prec over Ranong
MAM
u850 (scm) vs. prec over Ranong
MAM



prec vs. prec over Phochentong
MJJ
sst vs. prec over Phochentong
MJJ
u850 vs. prec over Phochentong
MJJ
slp vs. prec over Phochentong
MJJ
90N
60N
30N
0
30S
60S
90S
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
0
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
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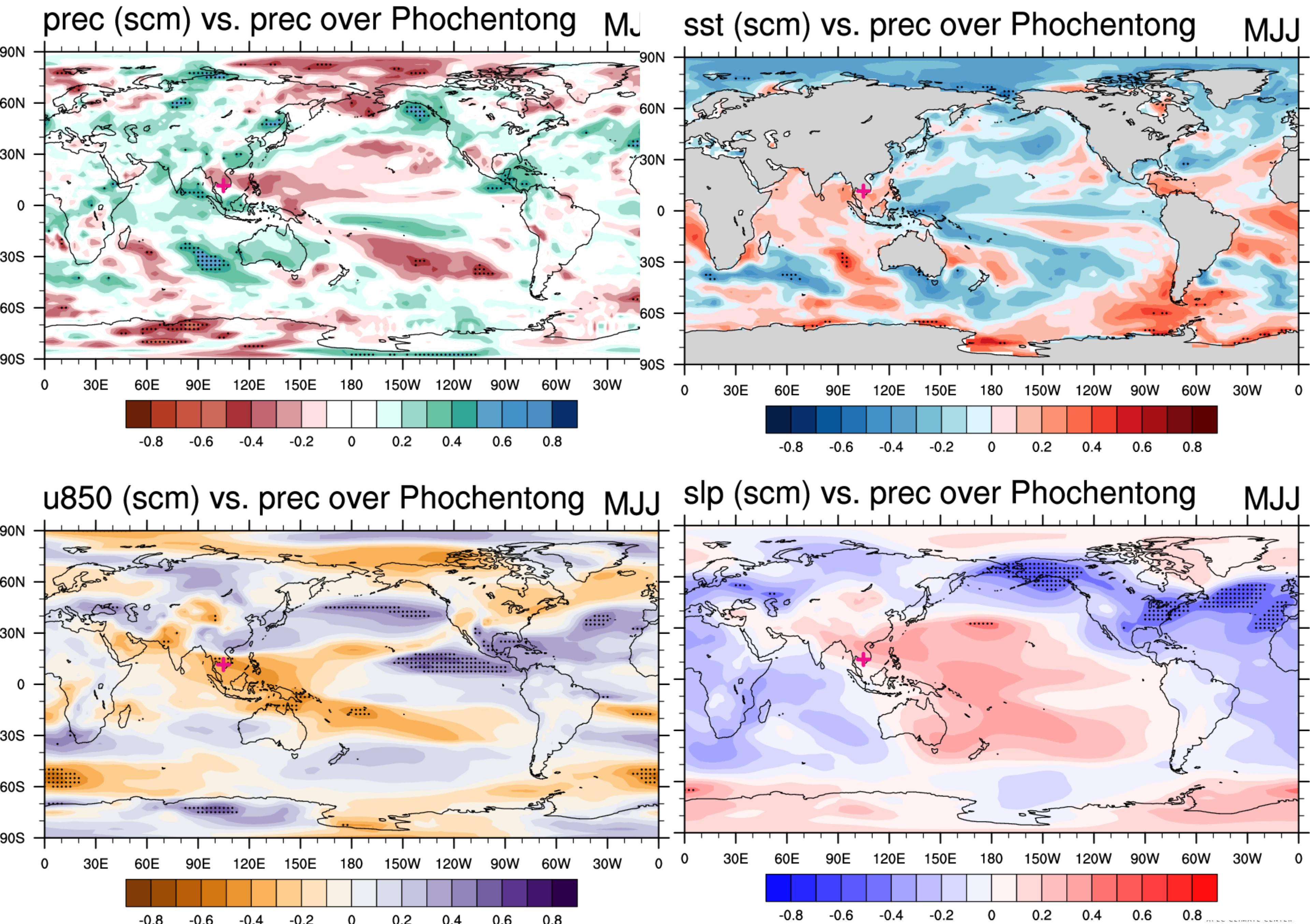
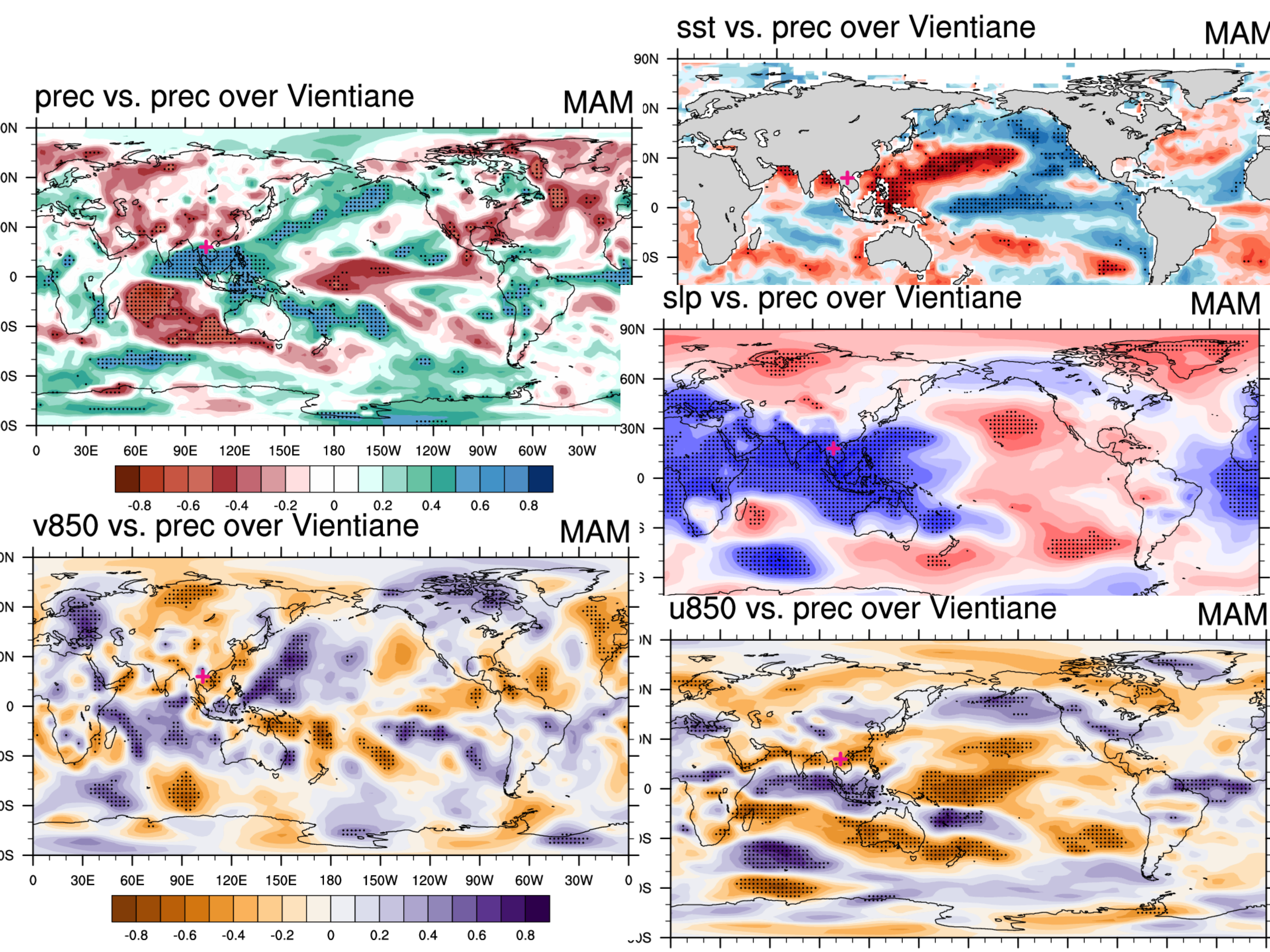
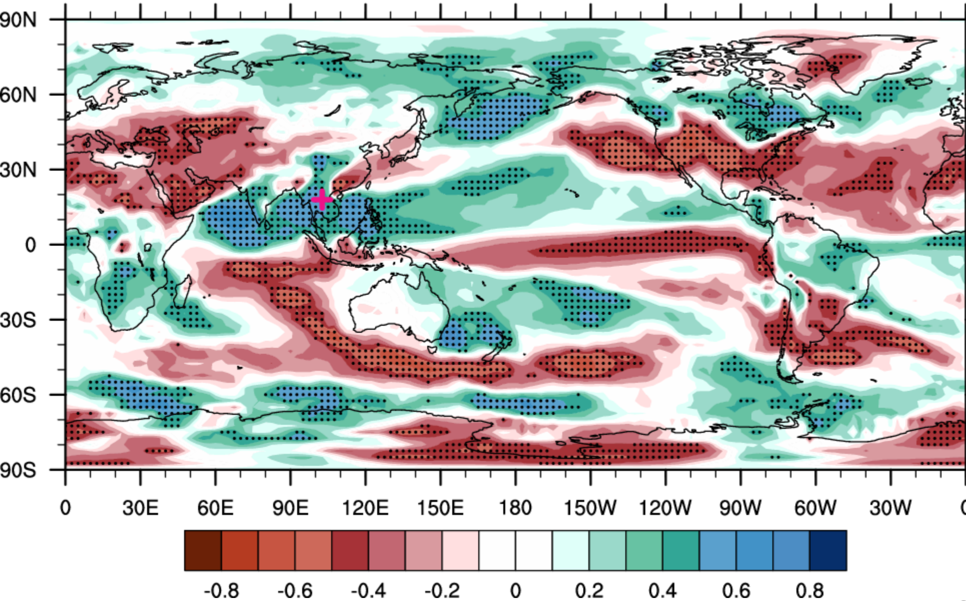

prec (scm) vs. prec over Phochentong MJJ
sst (scm) vs. prec over Phochentong MJJ
u850 (scm) vs. prec over Phochentong MJJ
slp (scm) vs. prec over Phochentong MJJ
90N
60N
30N
0
30S
60S
90S
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8

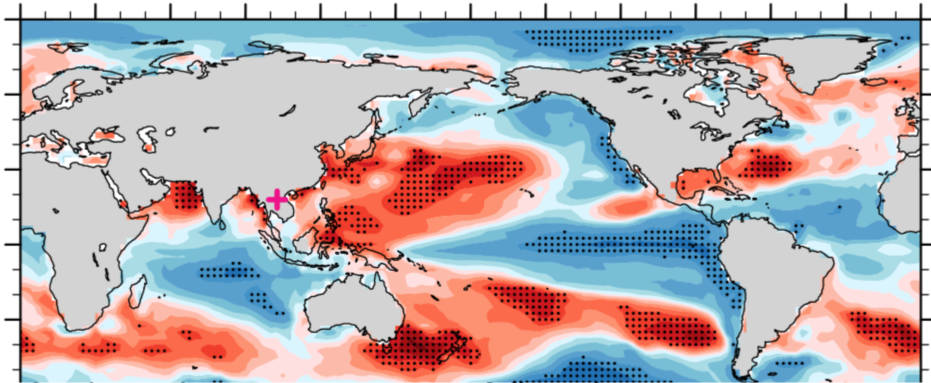


sst vs. prec over Vientiane
MAM
90N
0
prec vs. prec over Vientiane
MAM
0
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
slp vs. prec over Vientiane
MAM
90N
60N
30N
0
v850 vs. prec over Vientiane
MAM
0
0
30E
60E
90E
120E
150E
180
150W
120W
90W
60W
30W
0
-0.8
-0.6
-0.4
-0.2
0
0.2
0.4
0.6
0.8
u850 vs. prec over Vientiane
MAM
0

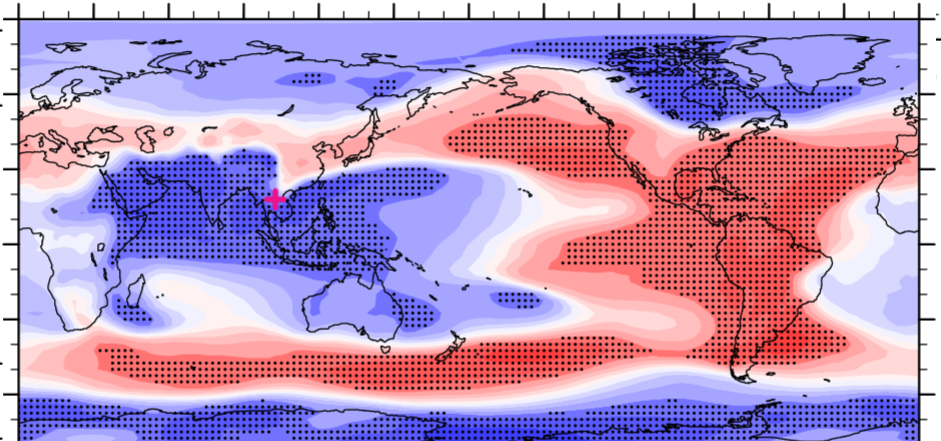
prec (scm) vs. prec over Vientiane MAM



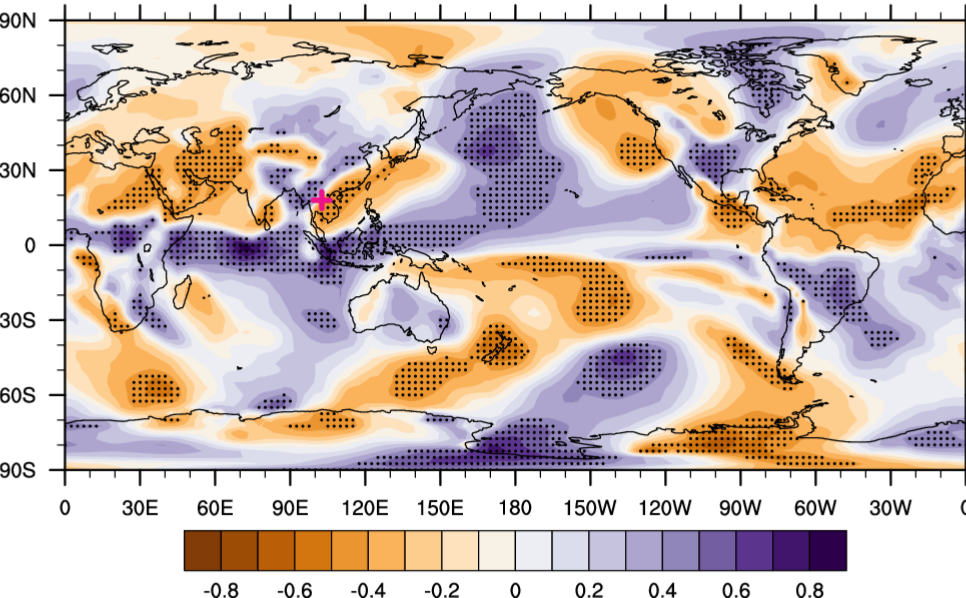
sst (scm) vs. prec over Vientiane MAM



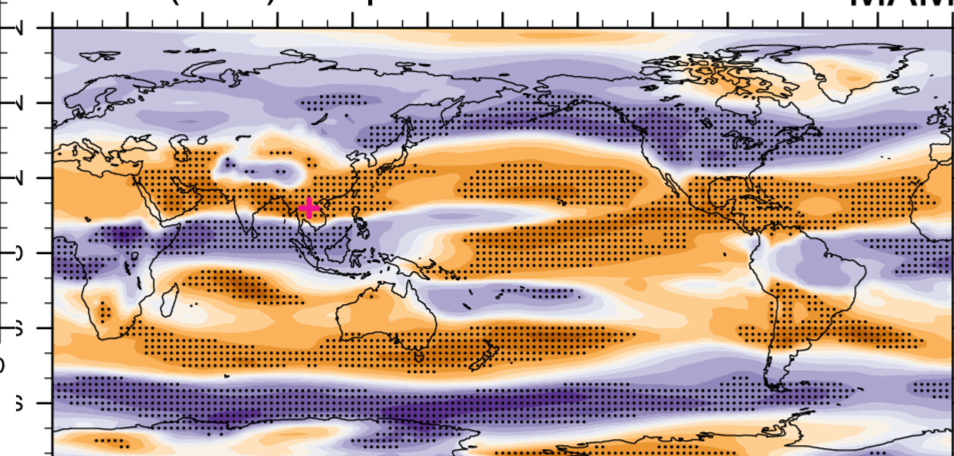
slp (scm) vs. prec over Vientiane MAM



v850 (scm) vs. prec over Vientiane MAM



u850 (scm) vs. prec over Vientiane MAM



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!

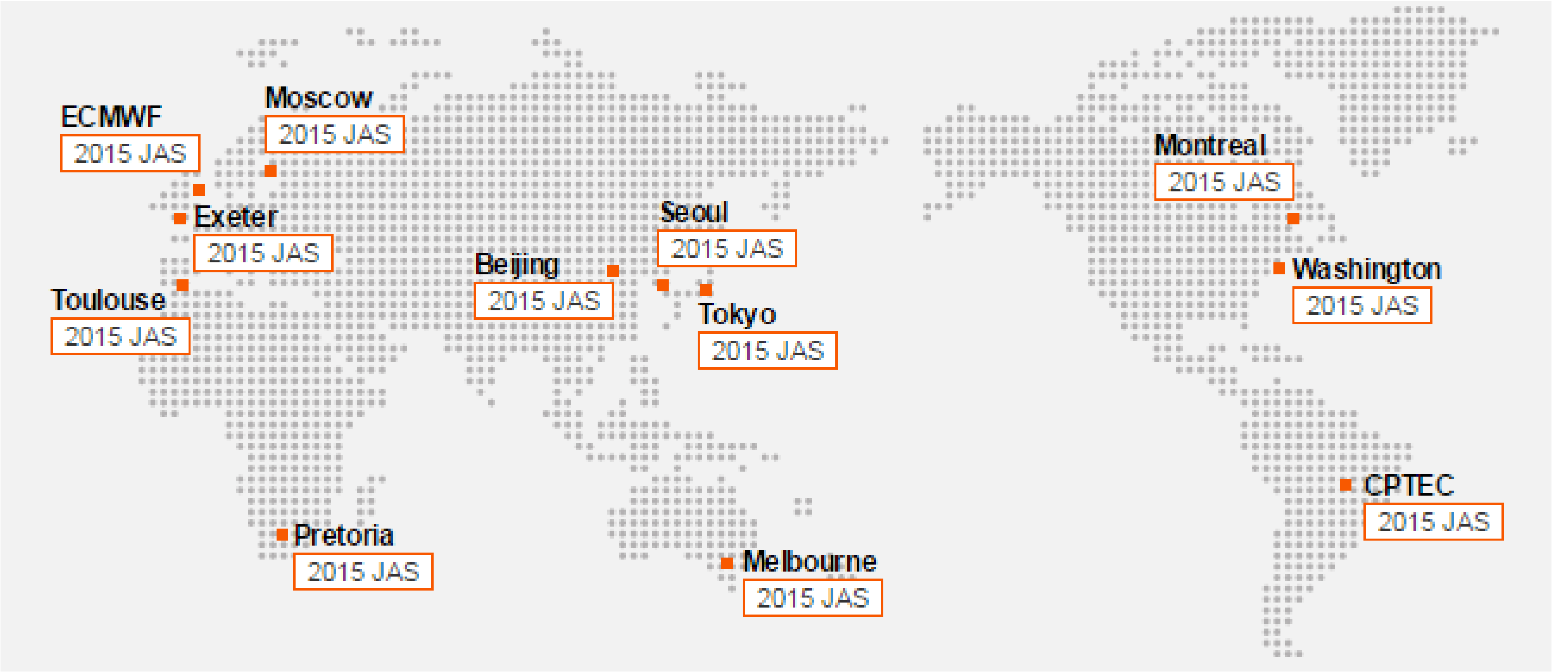
Current observation (monitoring)

- 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 loose 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)

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Cautions

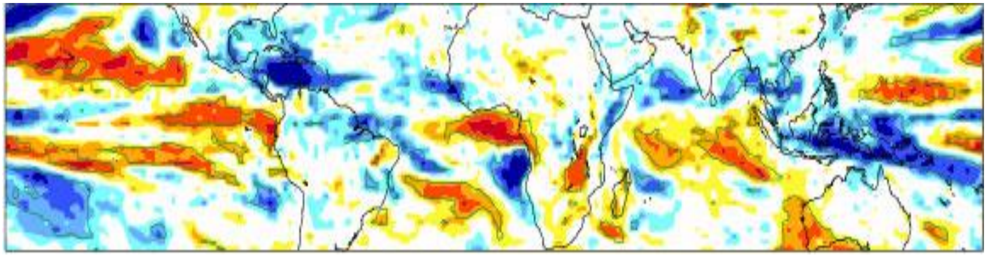
- How reliable our evaluation is?
- Even if you trust them, they can be wrong.
  - One reason to issue “probabilistic forecast”

2006 JJA mean Rainfall forecast

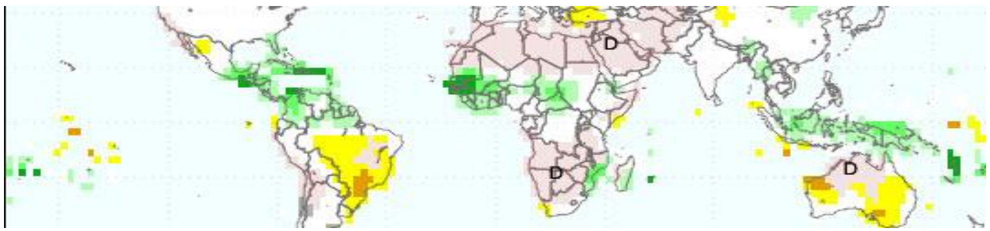
Warm colors : dry

Cool colors : wet

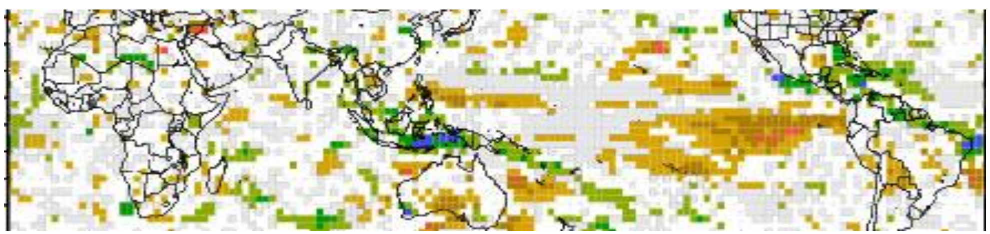
ECMWF



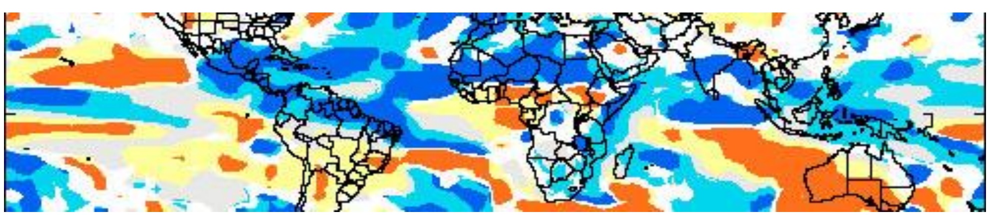
IRI

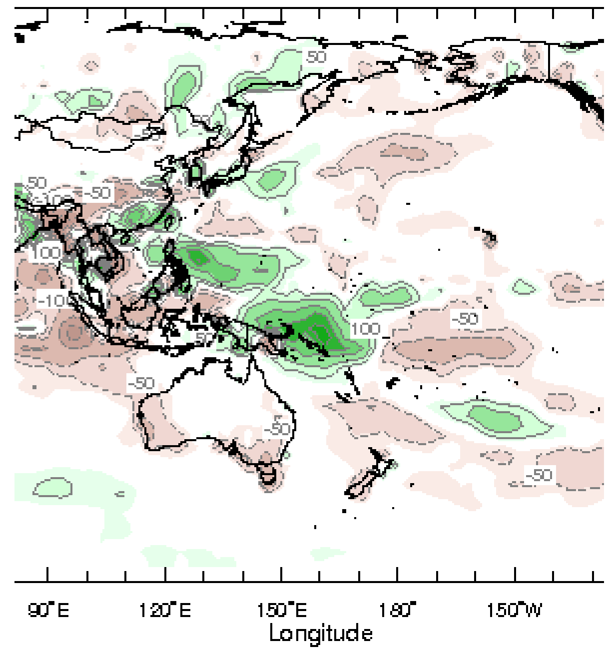


JMA



UKMO

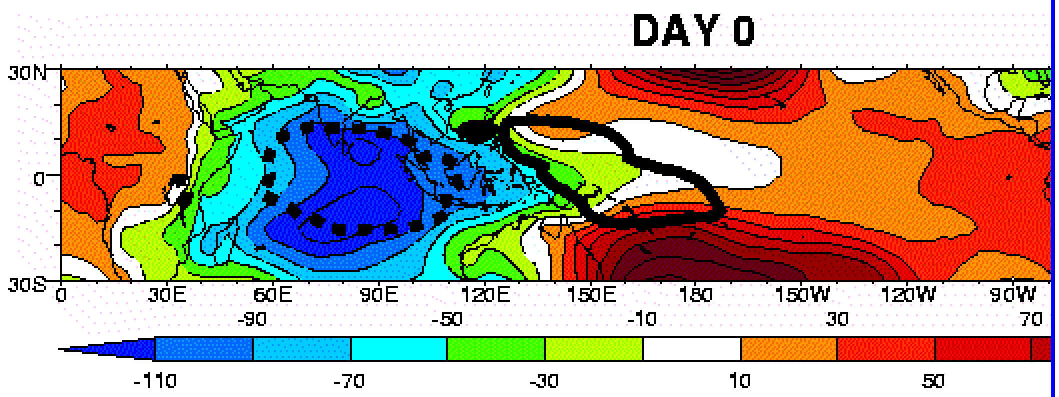


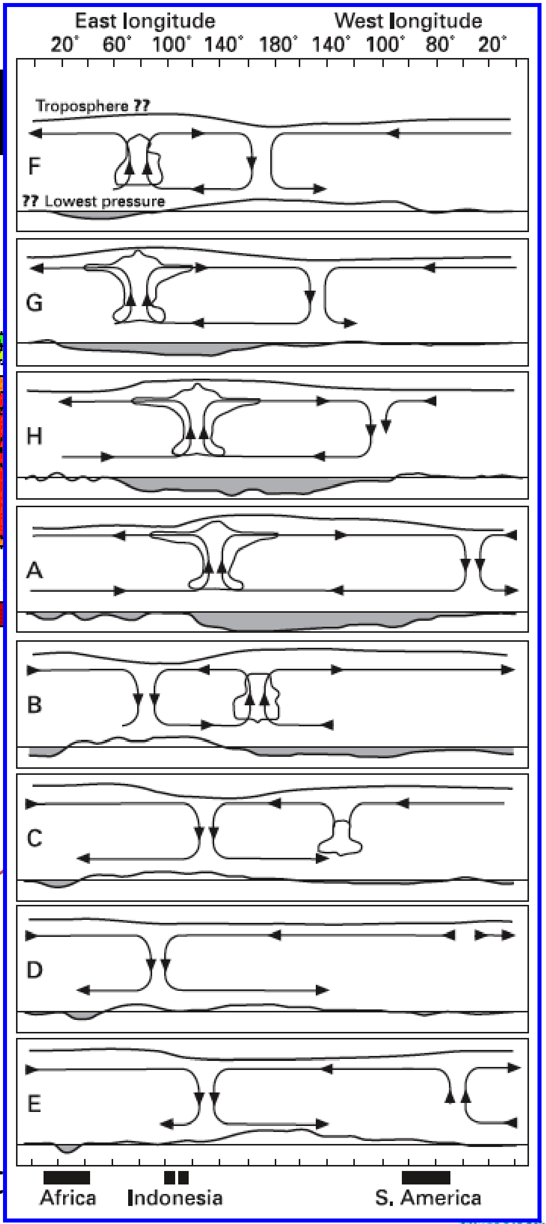


A few more…

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

Madden-Julian Oscill





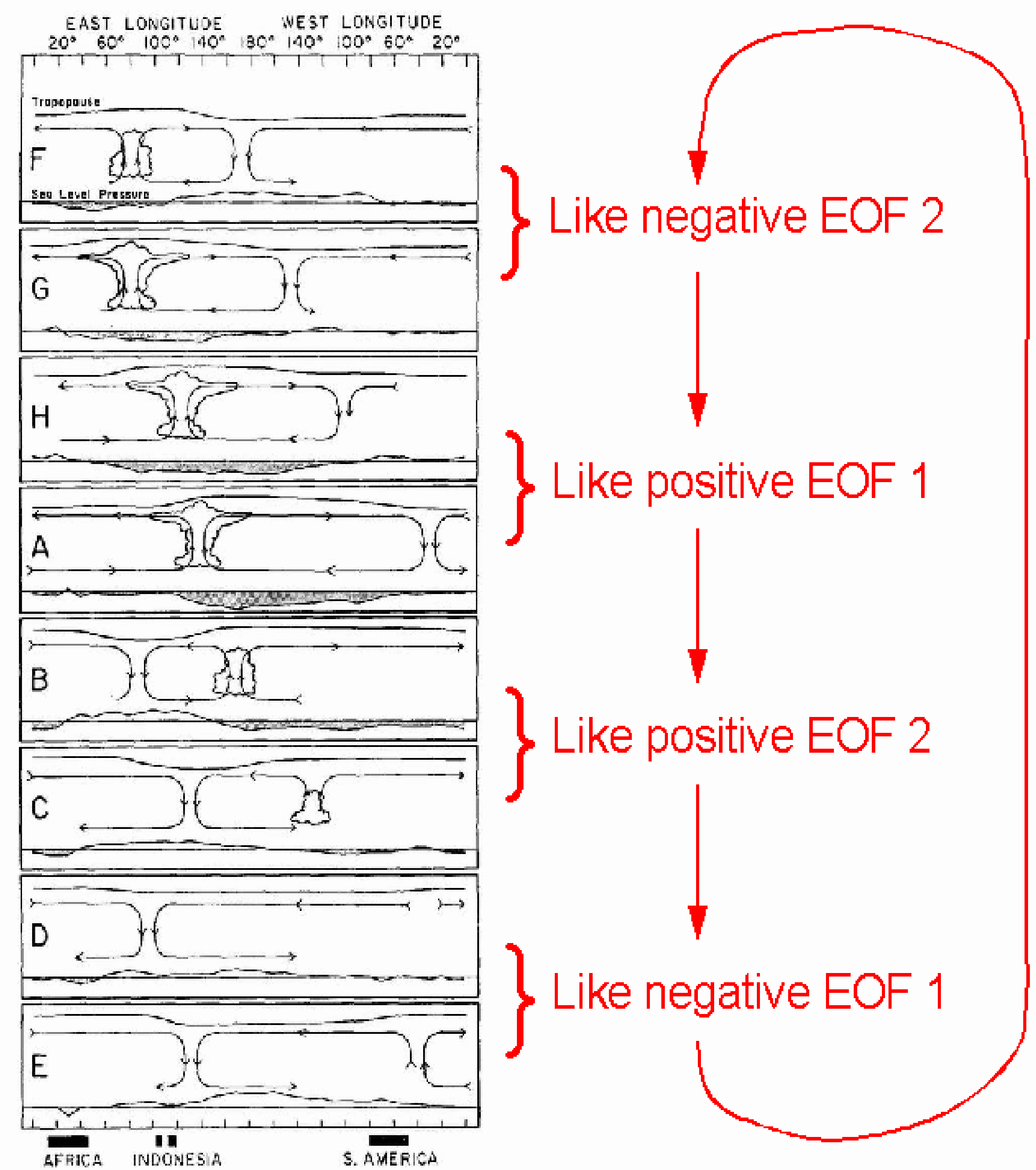
Madden & Julian (1971) : 40-50day oscillation (30-60days ISO)

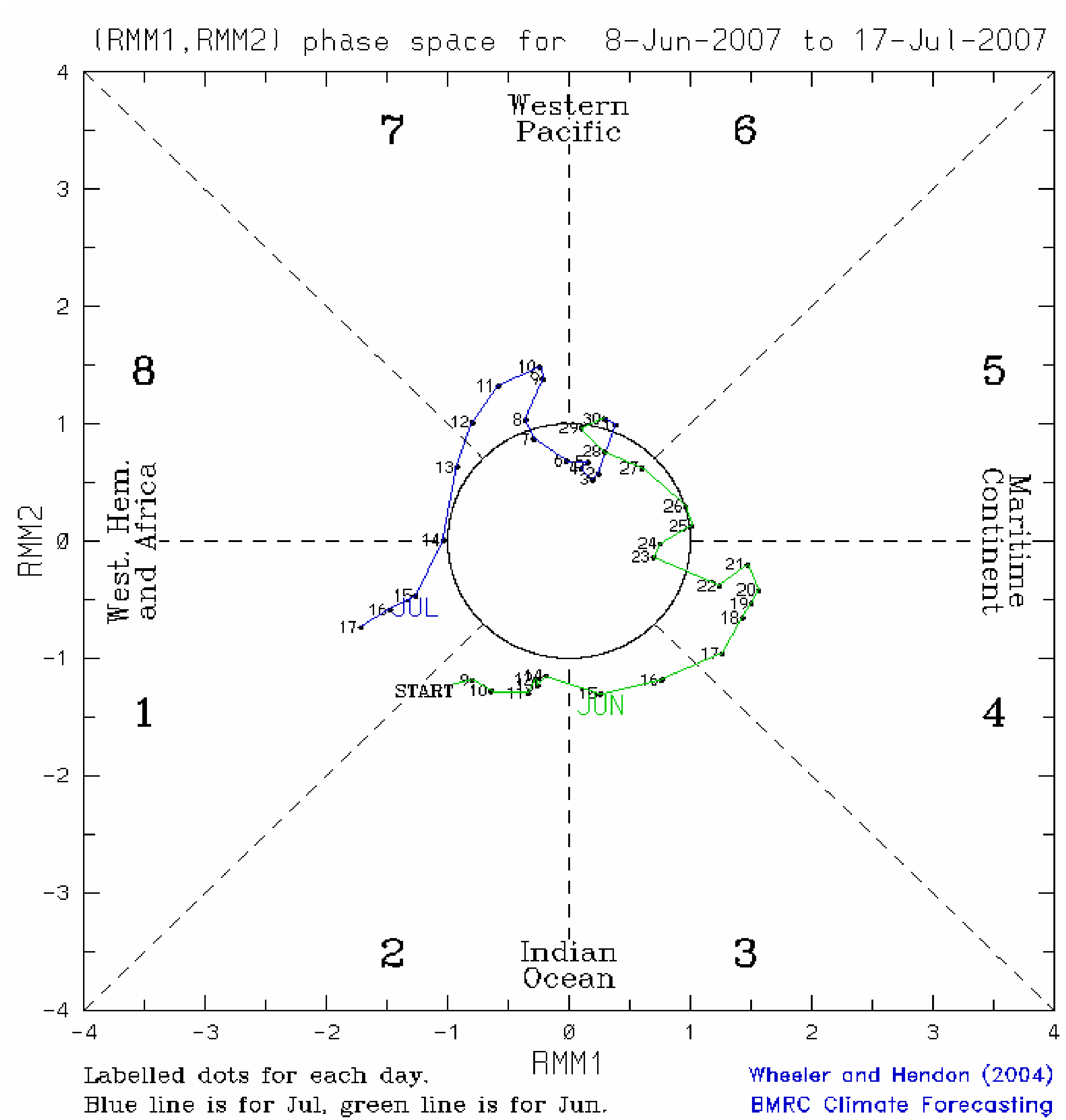
Eastward moving large scale convective anomaly along the equ
baroclinic structure

(precipitation anomaly is predominant in Indo-Pacific sector)

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

Madden and Julian (1972)





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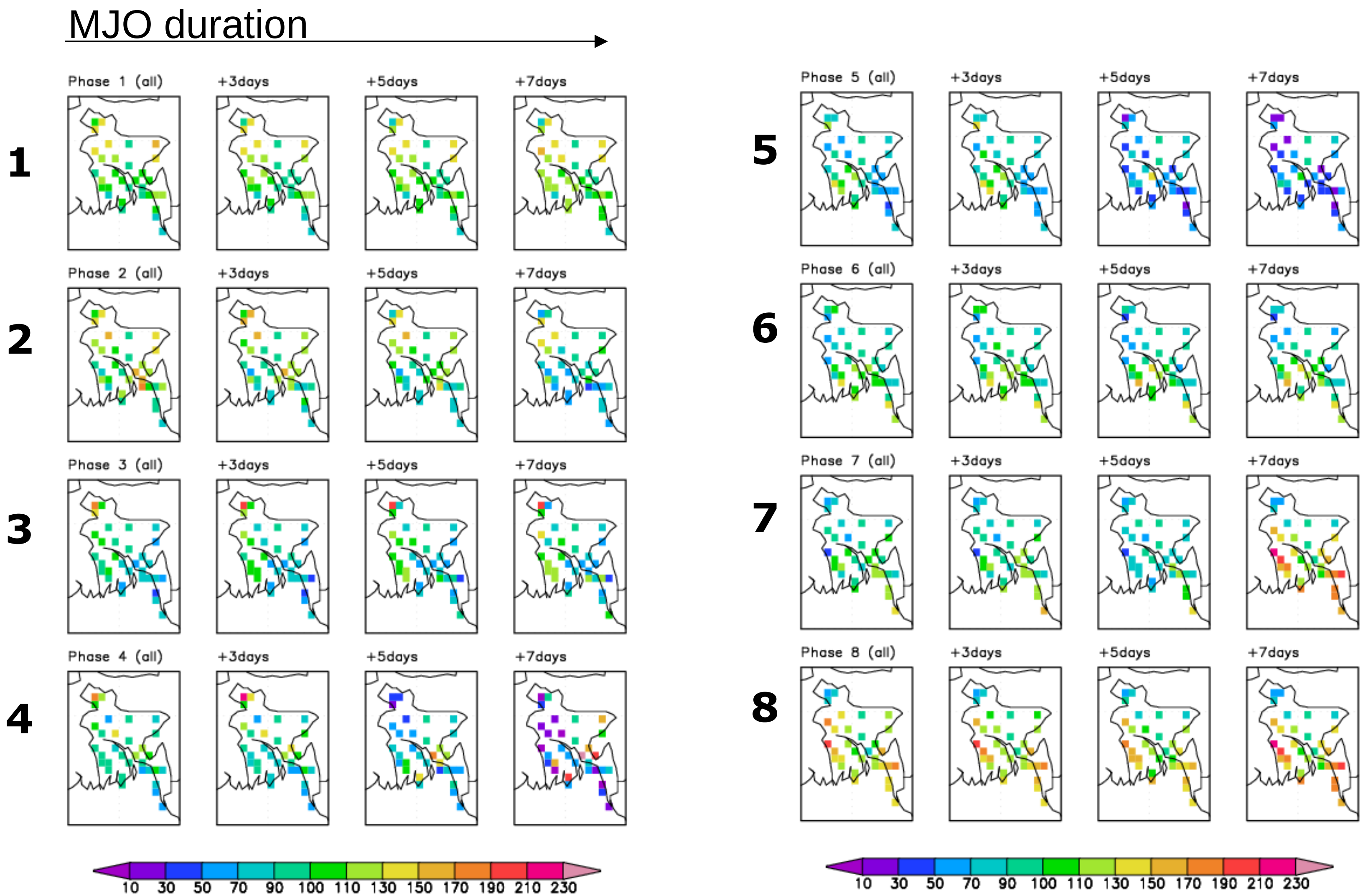
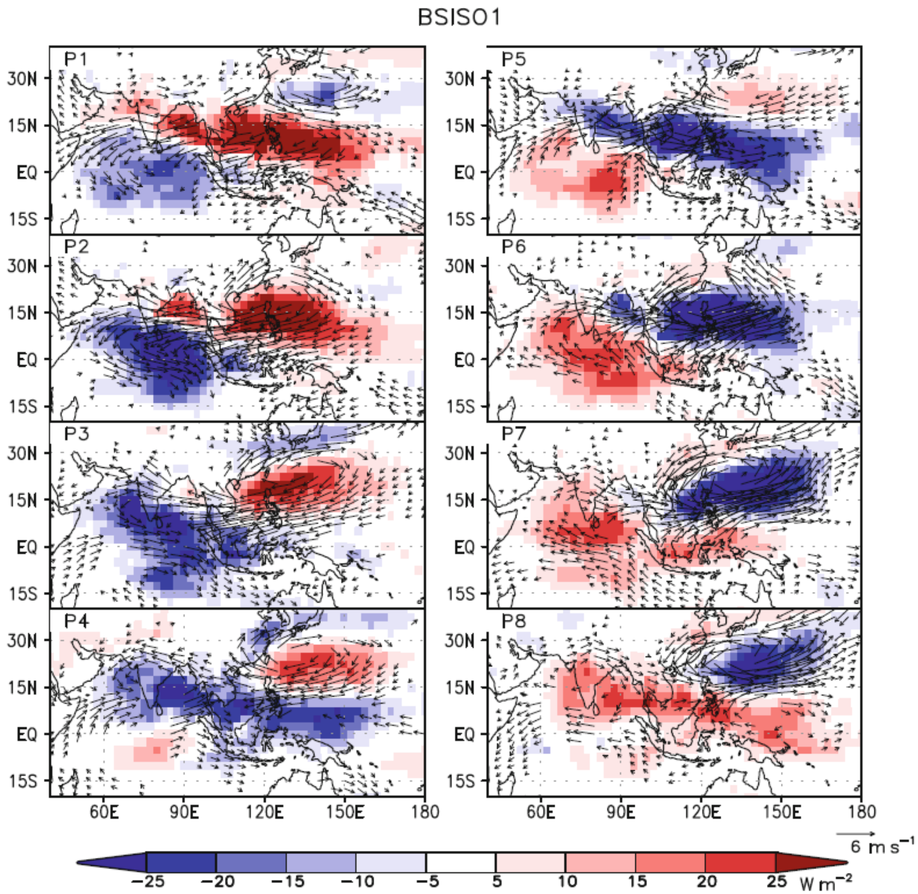
MJO and Bangladesh rainfall (% of climatology)



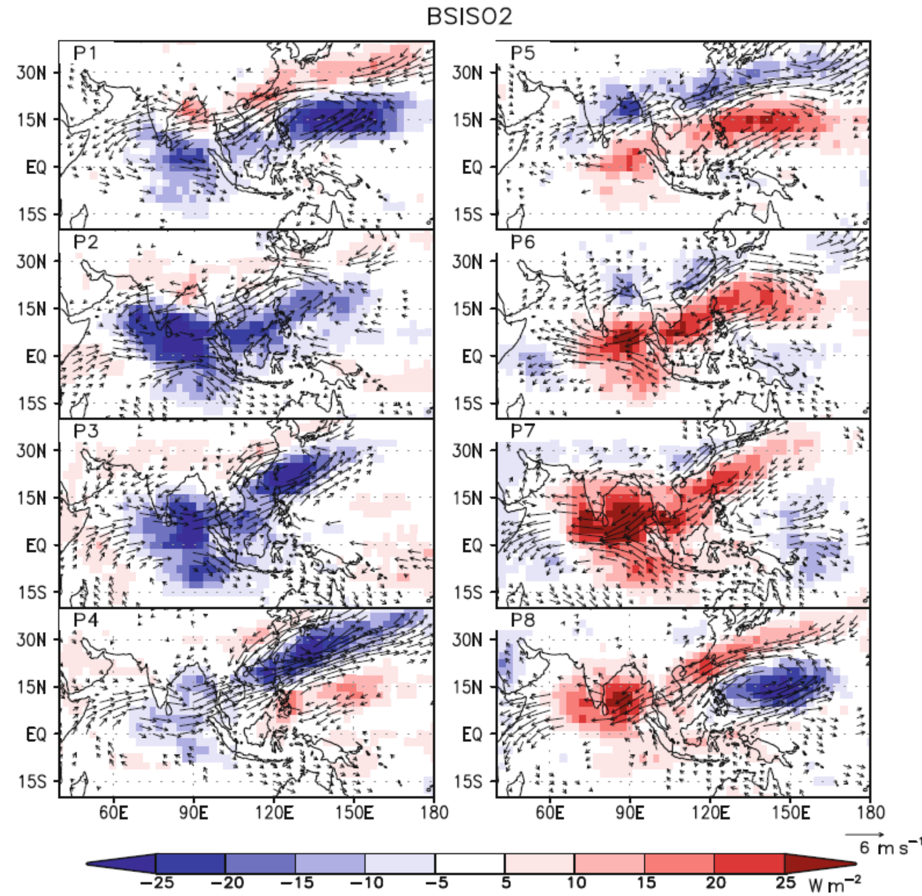
BSISO (Boreal Summer ISO)

www.apcc21.org → Climate Information Services → BSISO

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.

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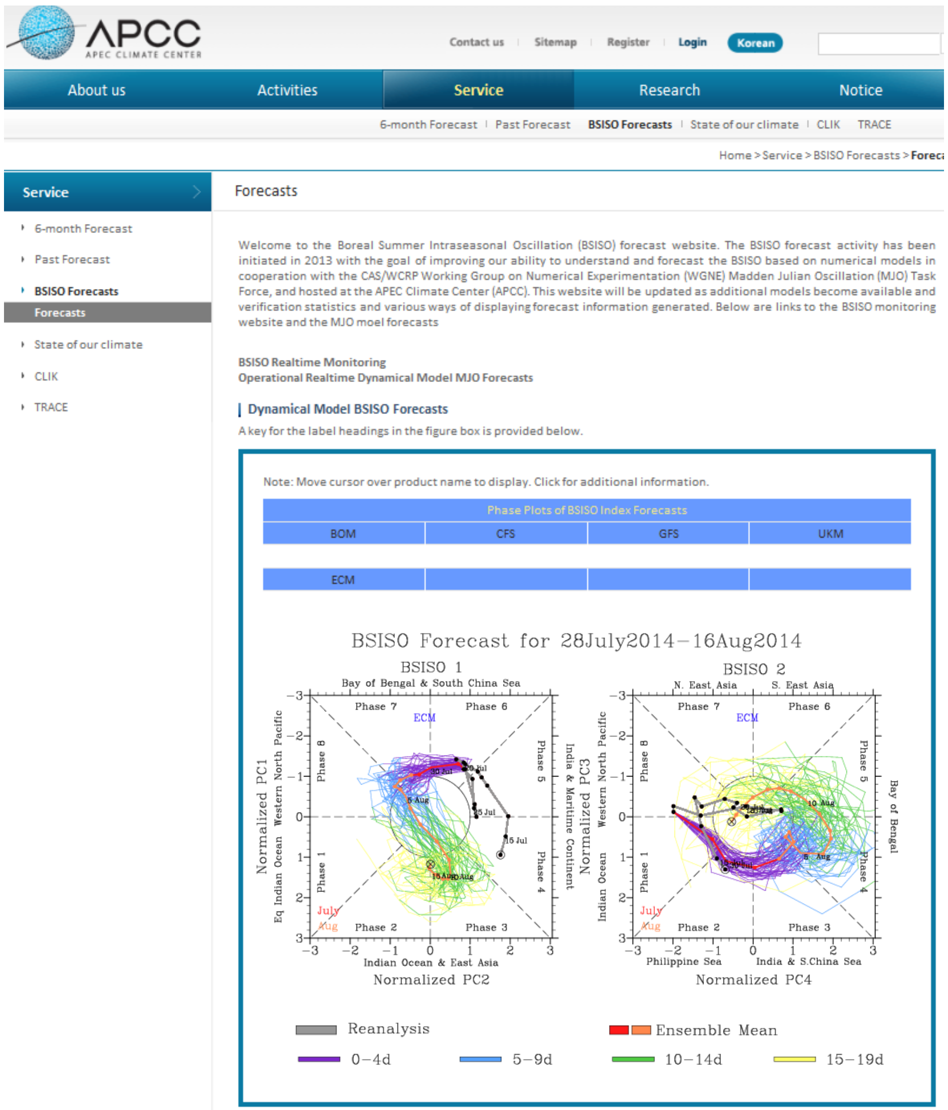
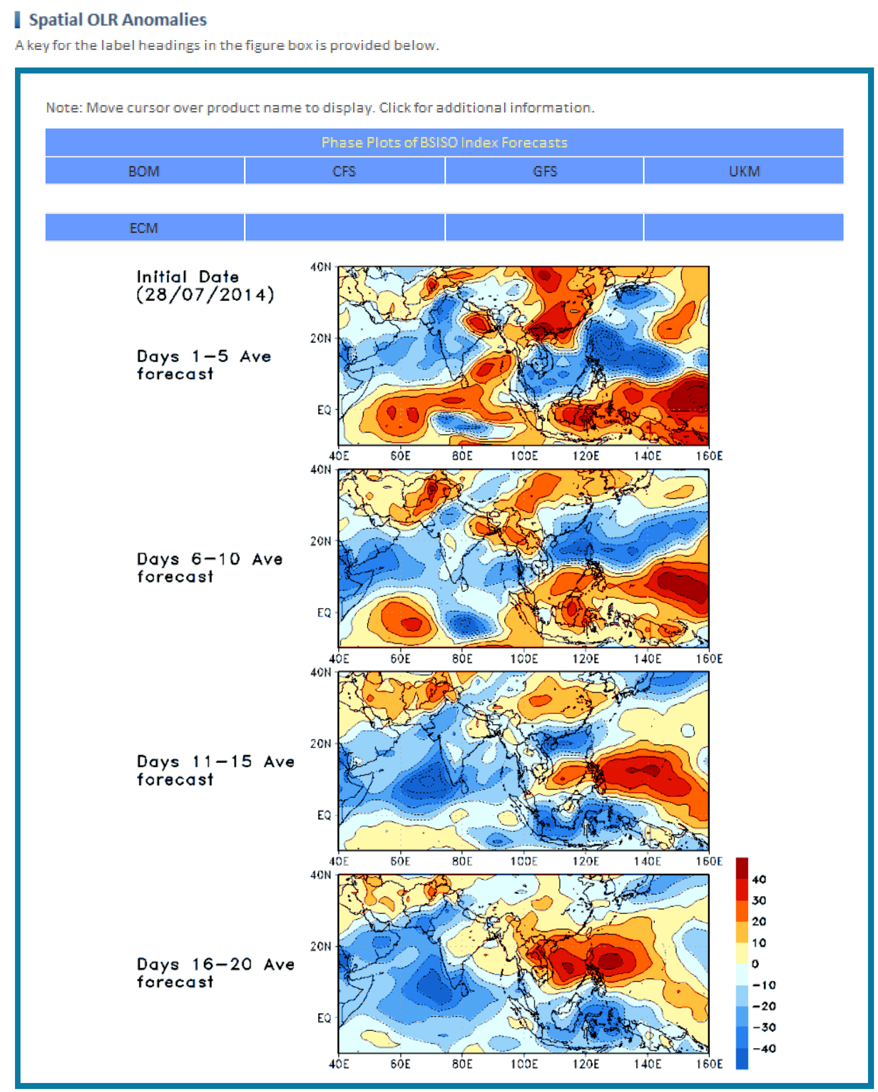
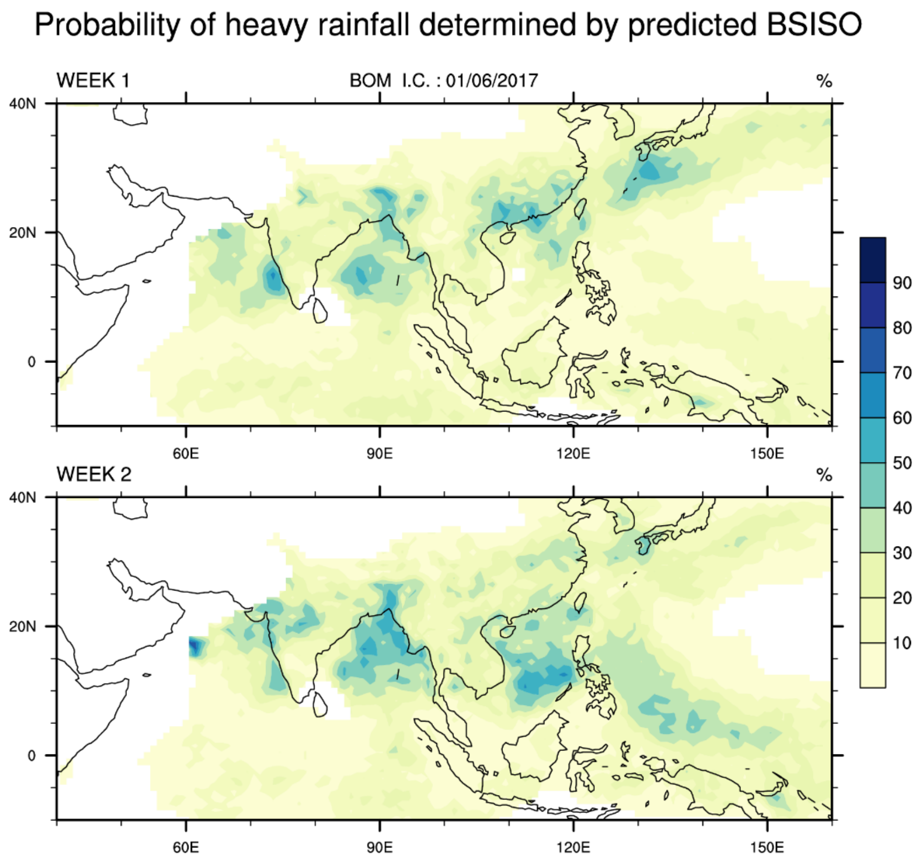
BSISO forecast (May-Oct)

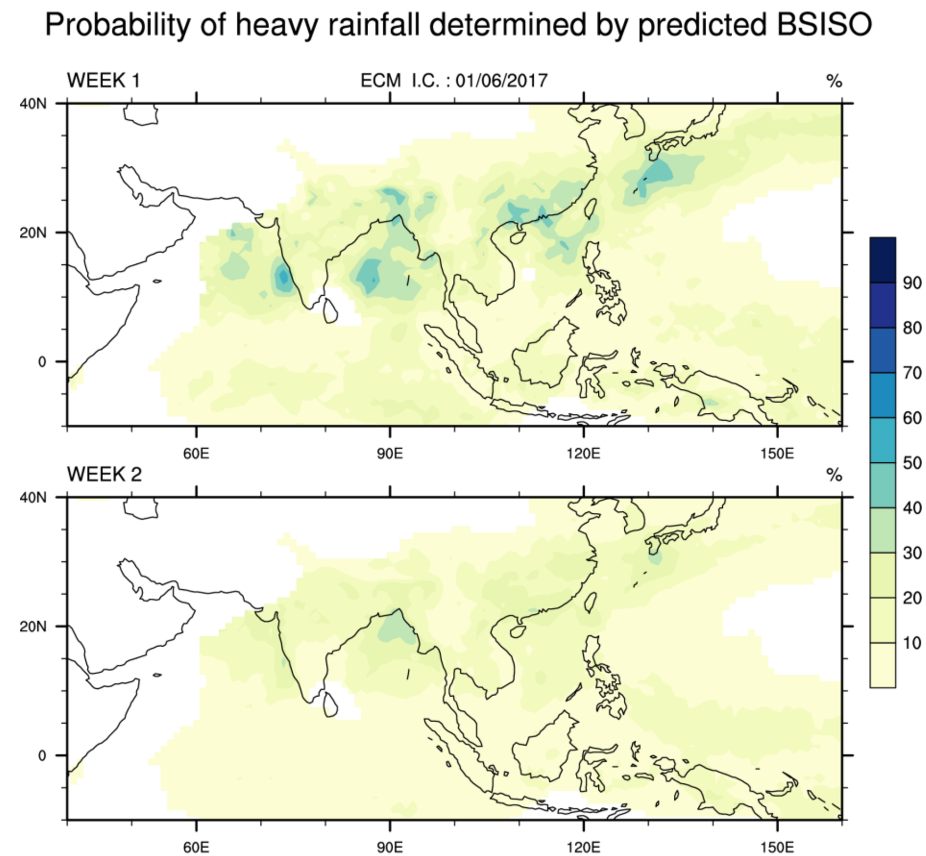




Heavy rainfall probability (BSISO)



Probability of occurrence for heavy rainfall event as defined by daily rainfall exceeding the 90th percentiles value (22.0 mm/day) for Jun. during 1981-2010.



Probability of occurrence for heavy rainfall event as defined by daily rainfall exceeding the 90th percentiles value (22.0 mm/day) for Jun. during 1981-2010.

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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 initiativea?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

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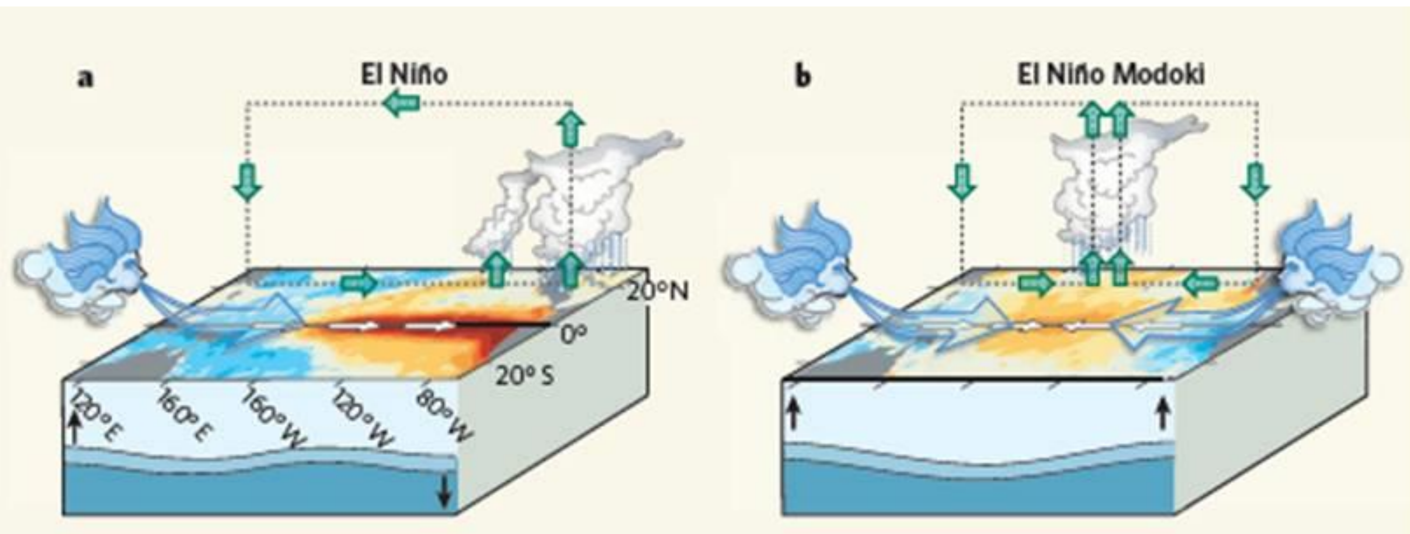
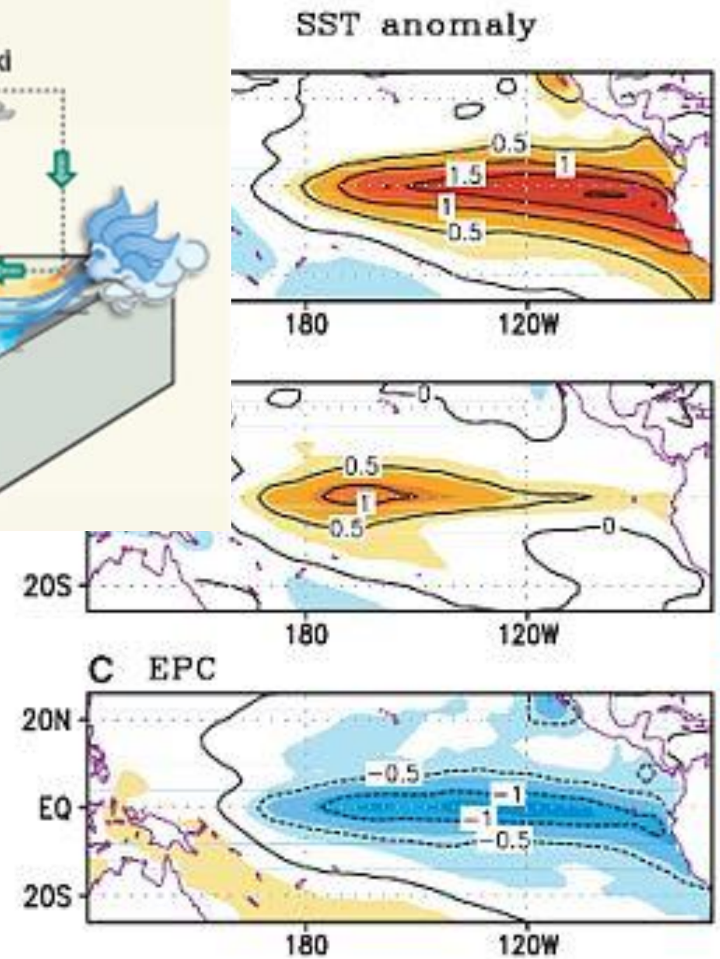
New type of El Nino

El Nino Modoki (Central Pacific El Nino)





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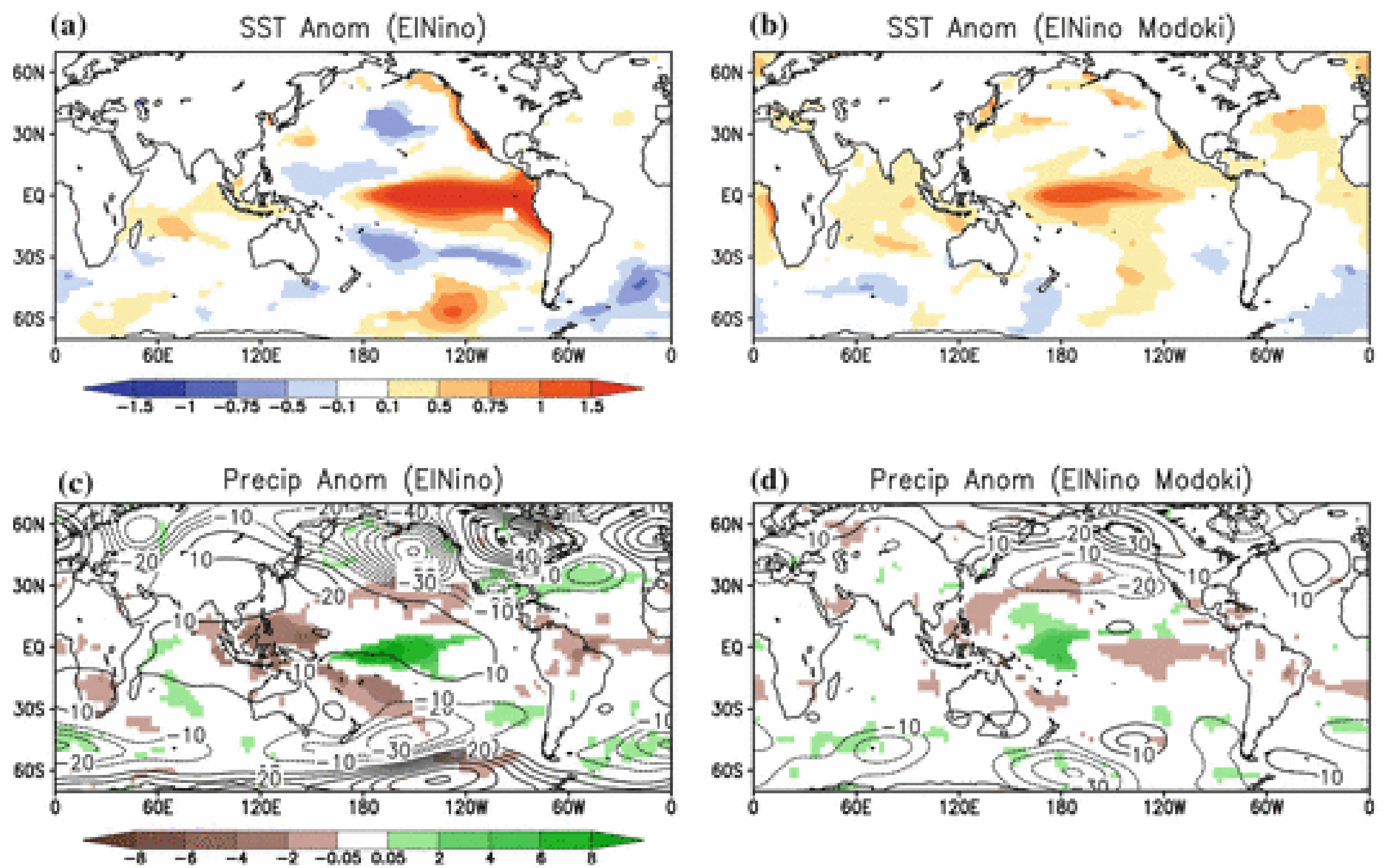

Fig. 2 **a** Composite observed significant SST (°C) anomalies during El Niño years. **b** Same as **a** but for El Niño Modoki years. **c** Composite observed significant precipitation (mm/day; *shaded*) anomalies and 500 hPa Geopotential Height (m; *contours*) anomalies during El Niño Years. **d** Same as **c** but for El Niño Modoki years. All the *shaded* values are significant at 90% using *t* test

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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

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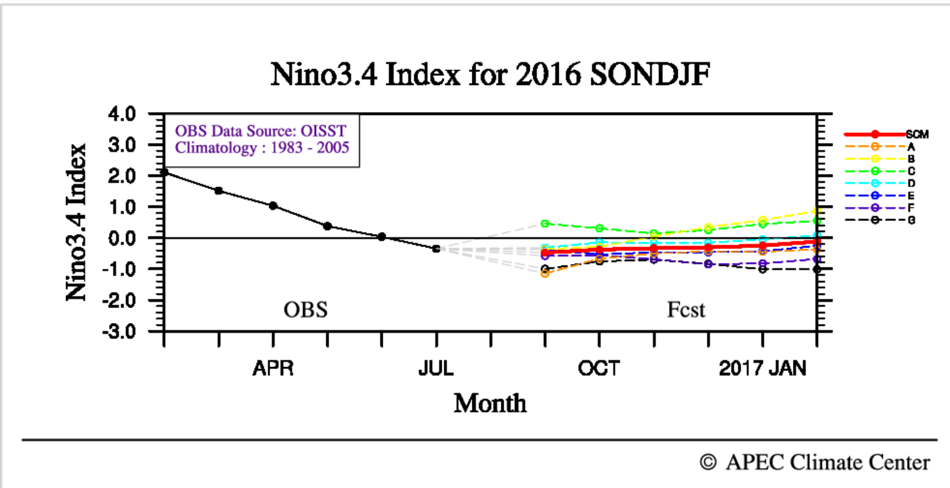
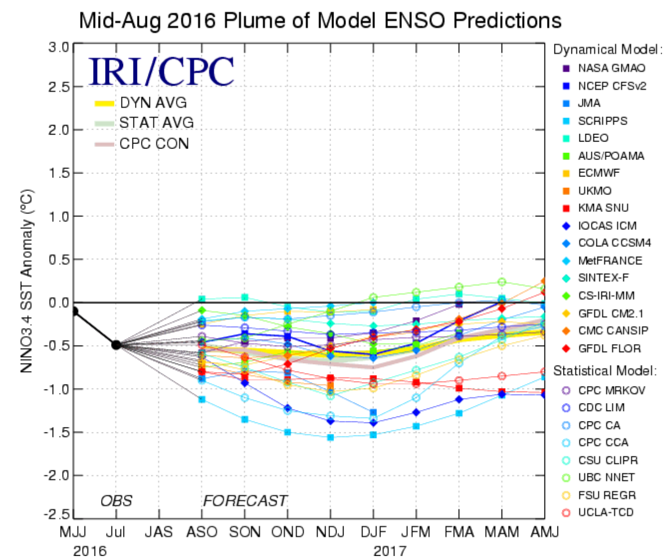
Practices





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???

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Suggestions?
Questions?

Thanks

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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

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Downscaling in CLIK

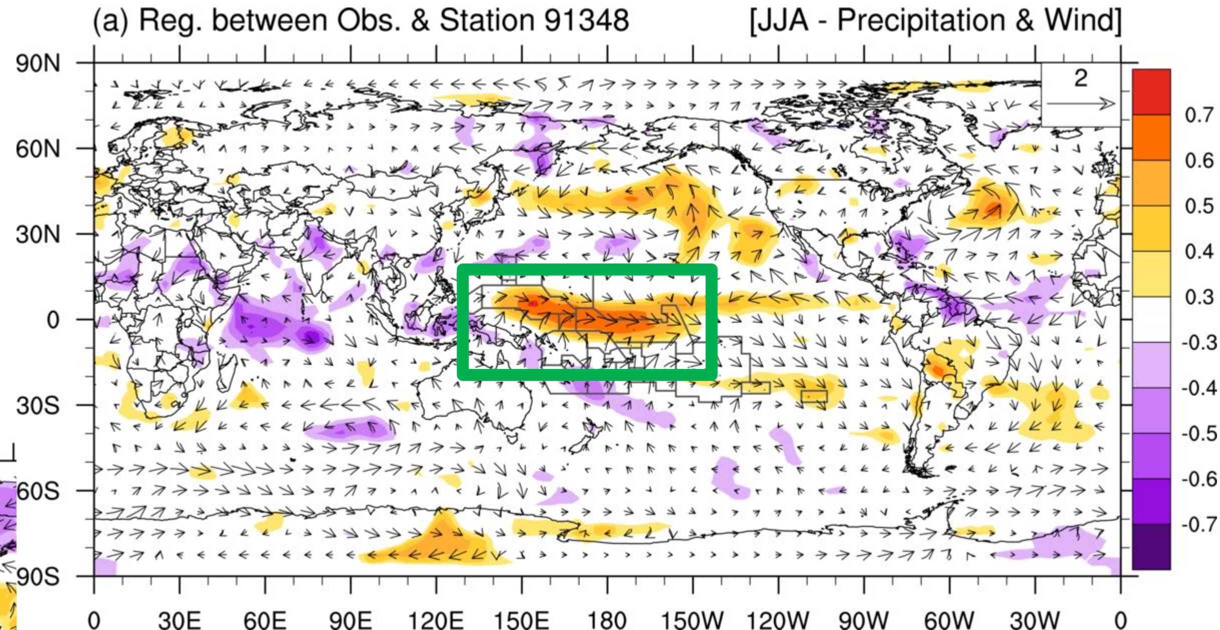
- Use "observed" large scale pattern (X) associated with climate variability at stations
- X needs to be predicted by GCMs to some degree
  - X becomes predictor (user selected area)

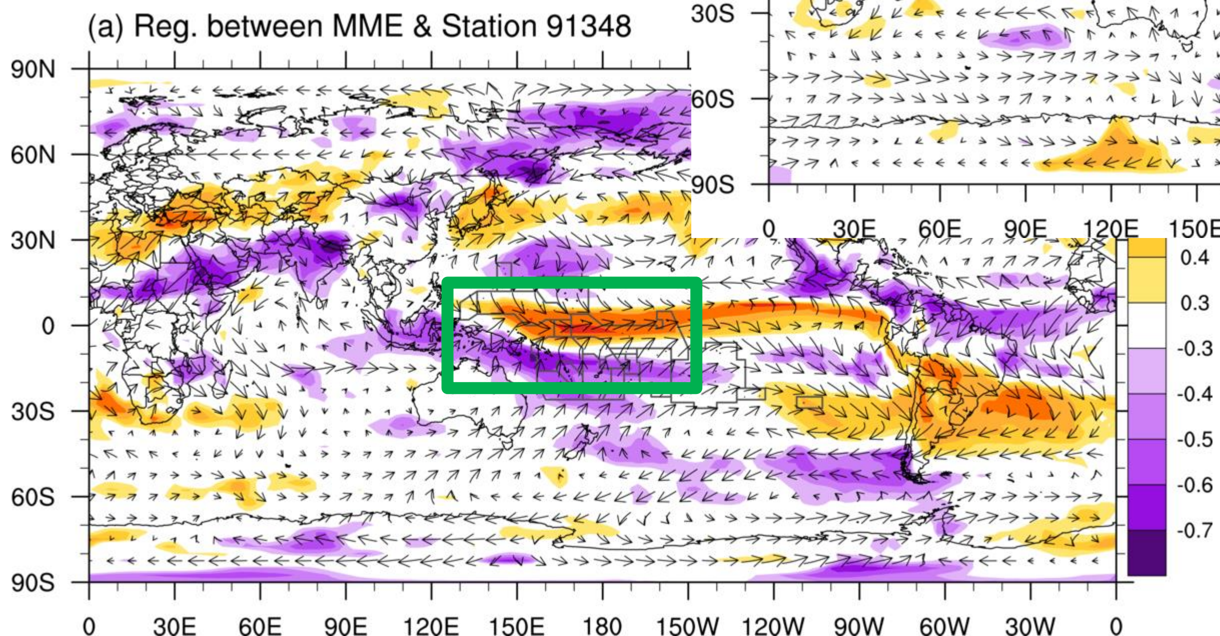
- CLIK does not provide any prior information for selection of predictor (to avoid overfitting)
  - Basic knowledge on Local large scale circulation and associated global teleconnection is necessary

Predictor selection

Meaningful pattern? (hopeful)
: significance score

Station data





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

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The most important thing you need is,

Patience