# Finding maximum predictable patterns in a S2S model over summer East Asia By

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- > Using APT finding maximum predictable patterns
  - Region of interest
  - Forecast time of interest
- > Exploring four S2S models: CNRM, ECMF, CFSv2, JMA
- > Reforecasts (hindcasts) up to 45 days

|                  | ECMWF    | JMA | CNRM     | CFSv2    |
|------------------|----------|-----|----------|----------|
|                  |          |     |          |          |
| Ensemble members | 10       | 10  | 9        | 3        |
| Ocean coupling   | <b>√</b> | X   | <b>√</b> | <b>\</b> |

(Vitar et al 2017) BAMS)

## APT (1) Average Predictability Time

Standard measure of predictability

$$P(\tau) = \frac{\sigma_{\infty}^2 - \sigma_{\tau}^2}{\sigma_{\infty}^2}$$

- $\sigma_{\infty}^2$  climatological variance; ensemble spread of long-term mean
- $\sigma_{\tau}^2$  ensemble forecast spread at lead time  $\tau$

### $P(\tau)$ varies from nearly **1** to **0** as $\tau$ increases and $\sigma_{\tau}^2 = \sigma_{\infty}^2$

- When the forecast spread equals the climatological spread
- In other words, a forecast is no better than a random guess from the climatologies.

(Jia et al 2015 Jclim; Delsole & Tippett 2009 JAS)

## APT (2) Average Predictability Time

#### Define APT

- √ Characteristic timescale of a (climate) system
  - When a system has strong damping, it has shorter memory and hence less predictable
- ✓ Independent of forecast lead times
  - average/integrate over all lead times
- √ Consistent with the common e-folding (half life) timescale
  - multiply by 2

APT=: 
$$2\sum_{\tau=1}^{\infty} \frac{\sigma_{\infty}^2 - \sigma_{\tau}^2}{\sigma_{\infty}^2}$$

(Jia et al 2011 Jclim; Delsole & Tippett 2009 JAS)

## Maximize APT (3) Eigenvalue Problem

- Maximizing APT
  - A system described by linear stochastic dynamics

$$\frac{dW^{obs}}{dt} = \mathbf{A}W^{obs} + \mathbf{F}\xi$$

 $\boldsymbol{W}$ : state vector

A : dynamics matrix

**F**: forcing matrix

ξ : white noise

- Lower and upper bounds for predictability of such a system
- Maximizing APT leads to a generalized eigenvalue problem;

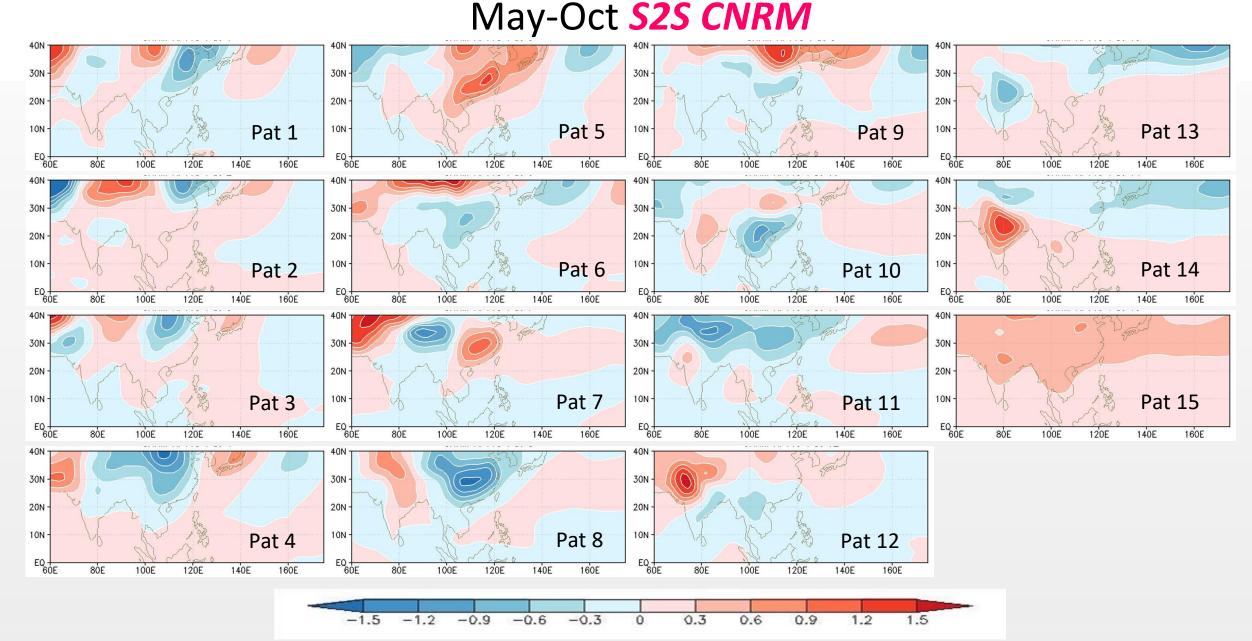
$$\sigma_{\tau}^2 = \boldsymbol{q}^T \Sigma_{\tau} \boldsymbol{q}; \quad \sigma_{\infty}^2 = \boldsymbol{q}^T \Sigma_{\infty} \boldsymbol{q}$$

$$2\sum_{\tau=1}^{\infty}(\Sigma_{\infty}-\Sigma_{\tau})\boldsymbol{q}=\boxed{2}\Sigma_{\infty}\boldsymbol{q}$$

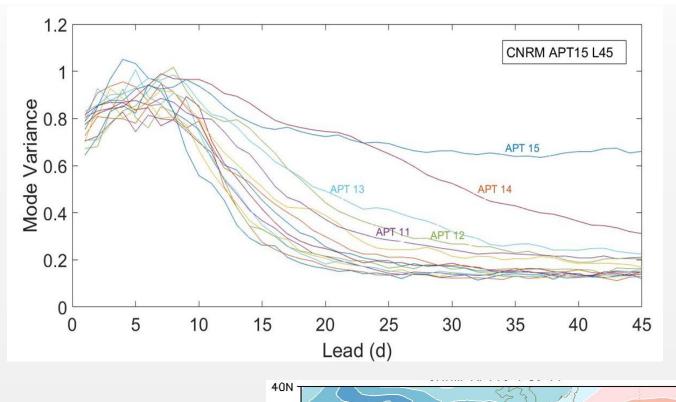
q: projection vector
 such that q<sup>T</sup>W
 maximizes APT

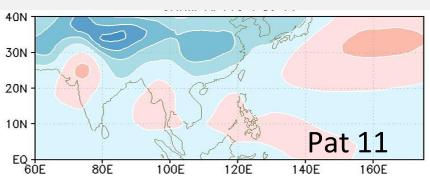
(Tippett & Chang 2003 Tellus; Jia et al 2011 Jclim)

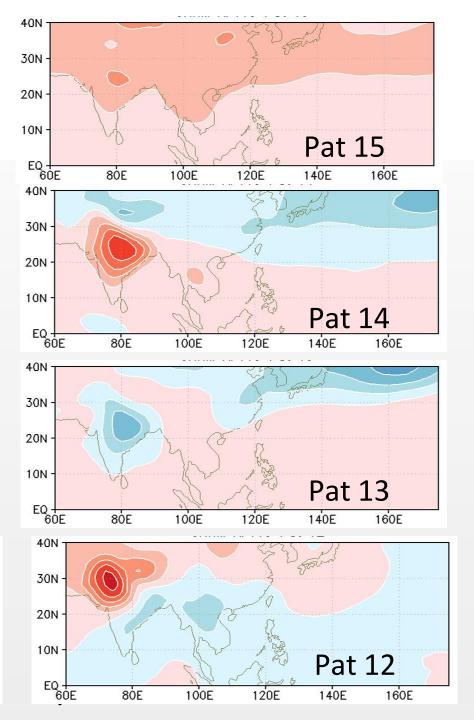
## Maximum Predictable Modes Indo Pacific T2m



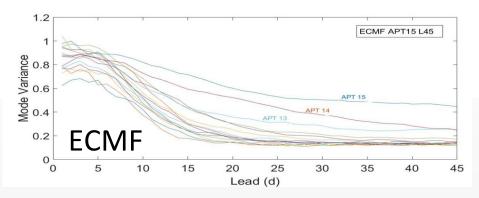
## Persistent Signals APT15 Lead = 45 *S2S CNRM*

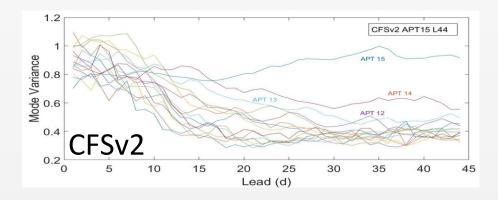


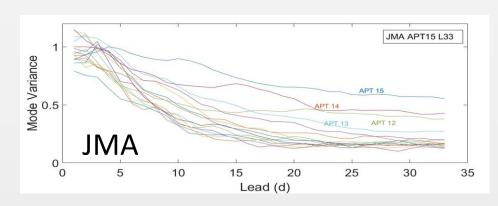


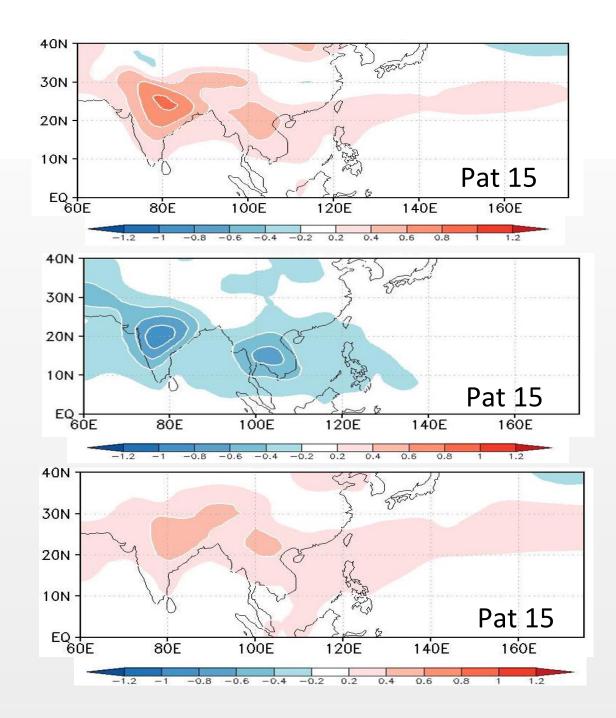


### Persistent Signals APT15

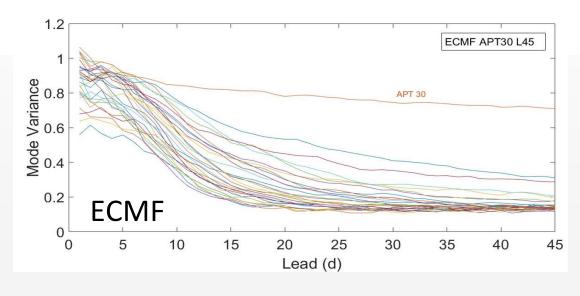


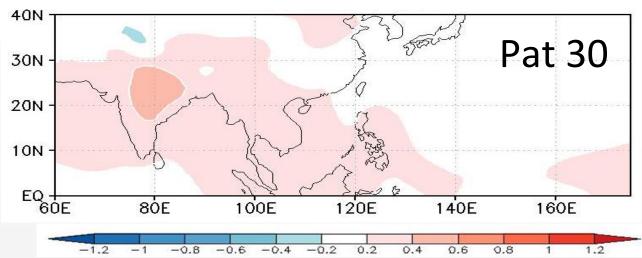


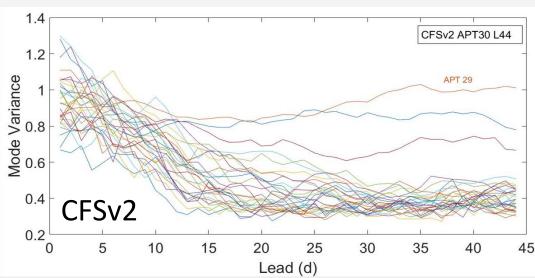


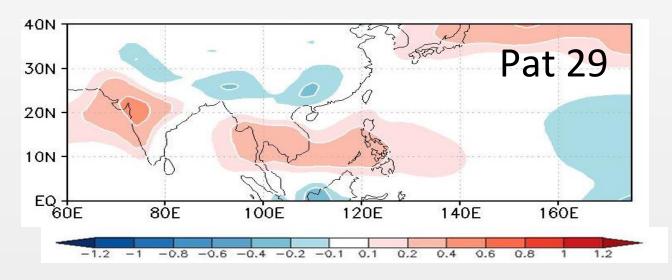


### Persistent Signals APT30



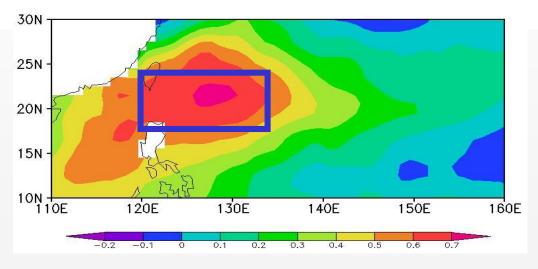


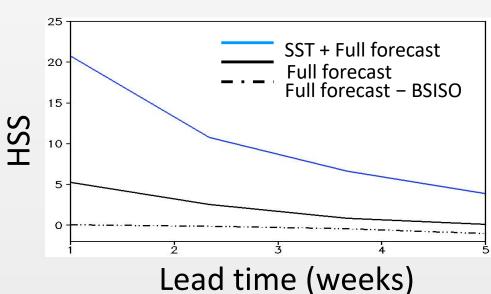


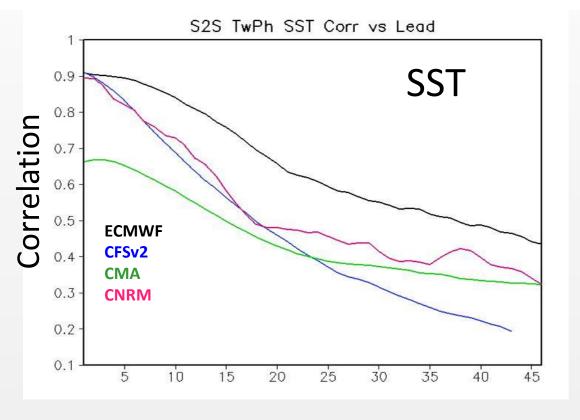


## Reconstruction of Subseasonal Index

#### Tw-Ph SST Index







Lead time (days)