

A linear Analysis of AMOC Dynamics in GFDL-ESM2M and CCSM4

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Introduction:

- The representation of the Atlantic Meridional Overturning Circulation (AMOC) differs greatly among individual coupled General Circulation Models (GCMs). Physical reasons for these differences are still poorly understood.
- We consider two such GCMs, ESM2M and CCSM4, and their simulation of the AMOC as defined by the anomalous streamfunction minus the Ekman component (Ψ_{NoEk} : see depth-latitude slices below). For this study, we use annually averaged model output.
- In particular, we investigate what role, if any, surface fresh water flux (**FW**: see latitude-longitude plots below) and surface heat flux (**HF**) in the same geographical area as **FW** (see plots below), play in the interannual evolution of the AMOC.

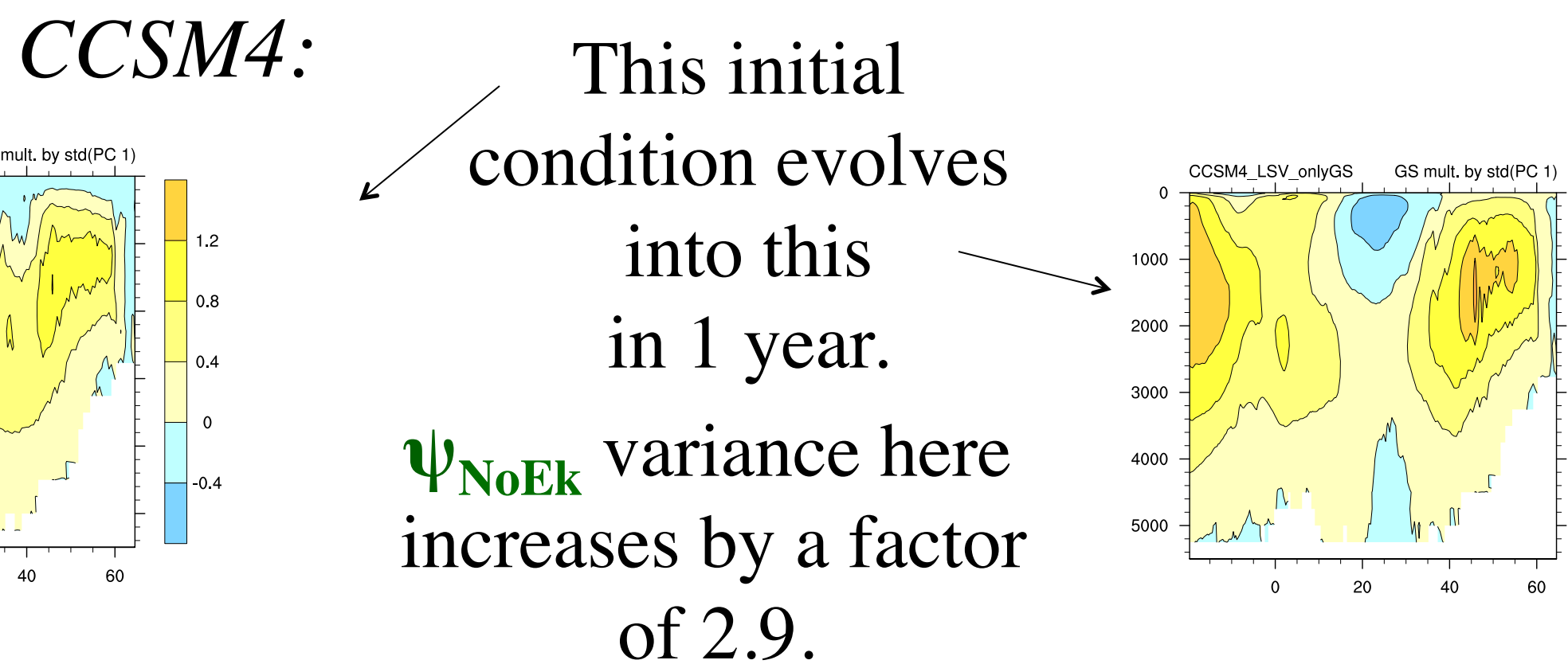
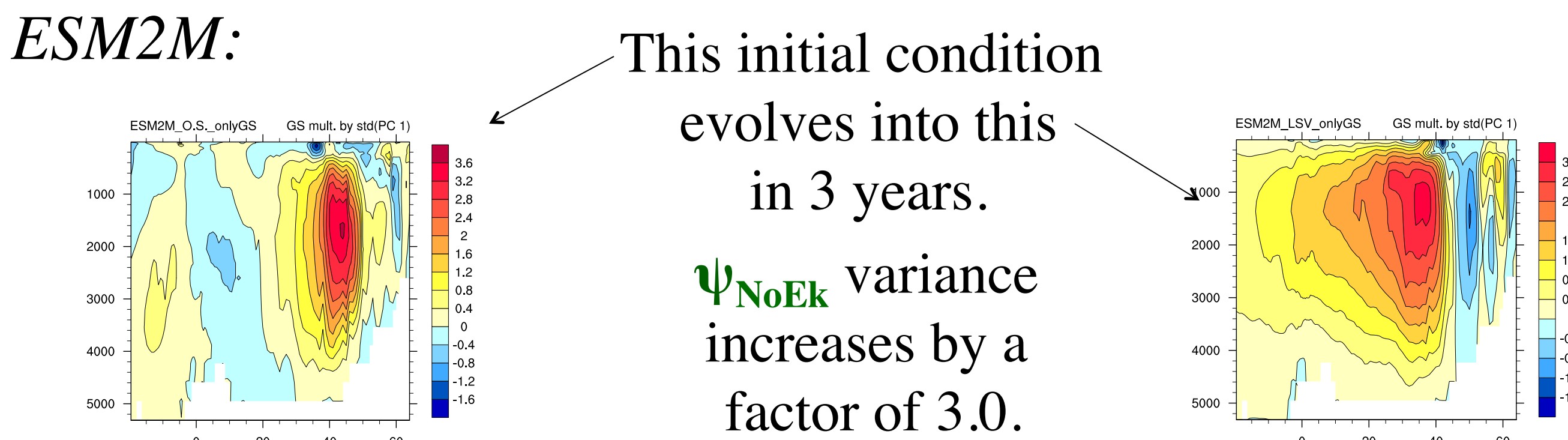
Methodology:

For each model, we consider two classes of state vector $\mathbf{x}$, consisting either of the 10 leading Principal Components (PCs) of Ψ_{NoEk} alone, or of the 10 leading PCs of each field, (Ψ_{NoEk} , **FW**, **HF**), normalized so that the 30 time series are about the same size but the relative variance of the PCs within any single field is preserved. We apply *Linear Inverse Modeling* (LIM) to these fields. That is, we use lagged and contemporaneous statistics to derive the best fit linear model of the form

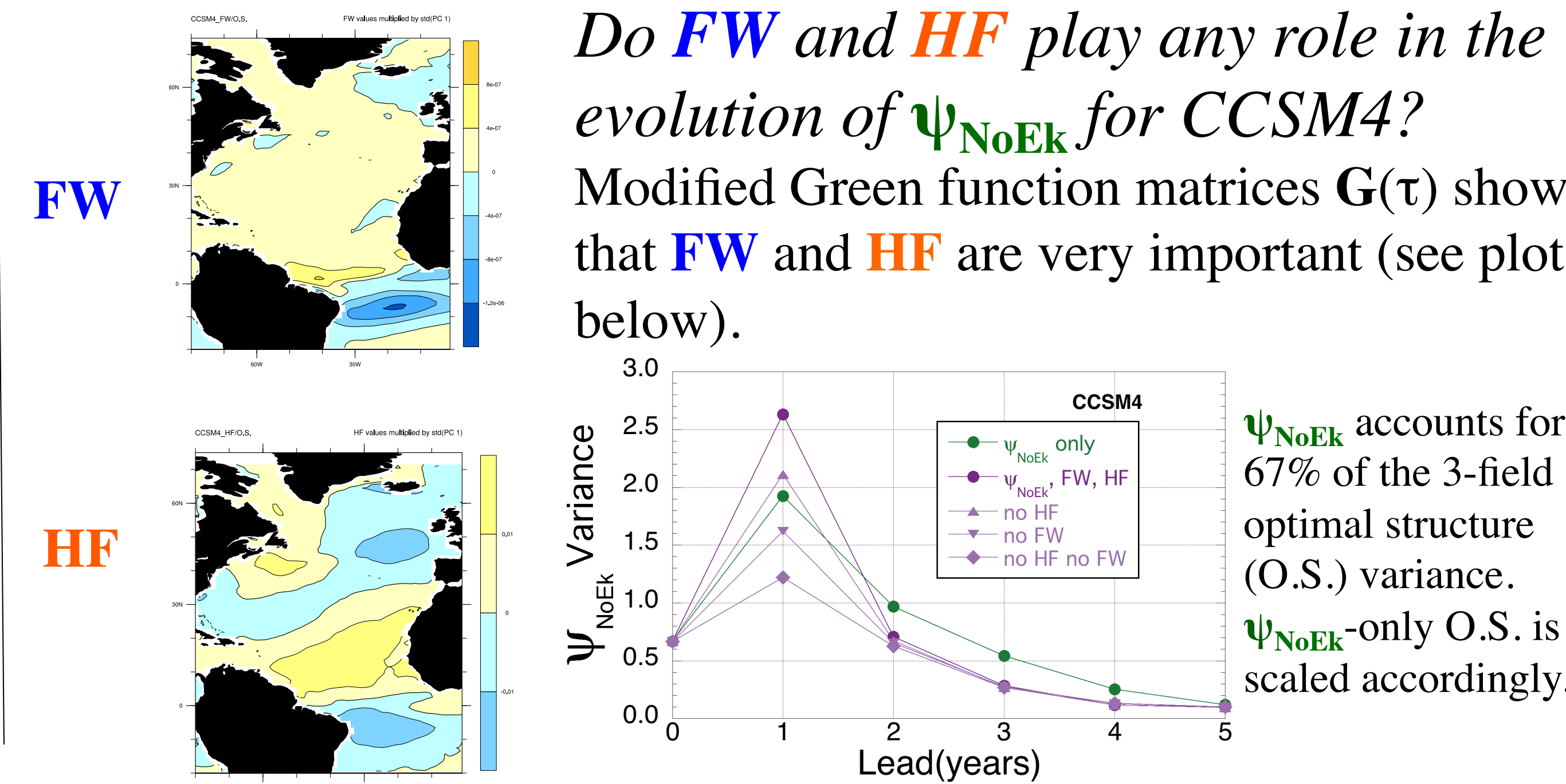
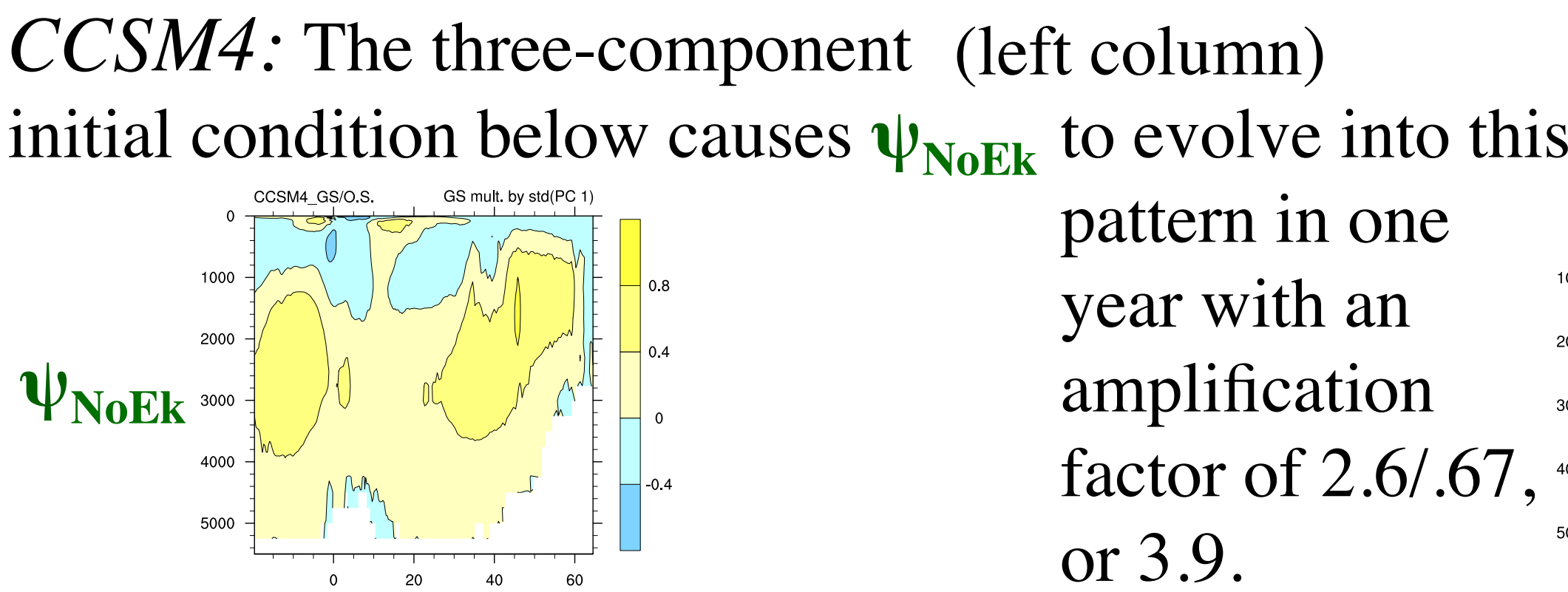
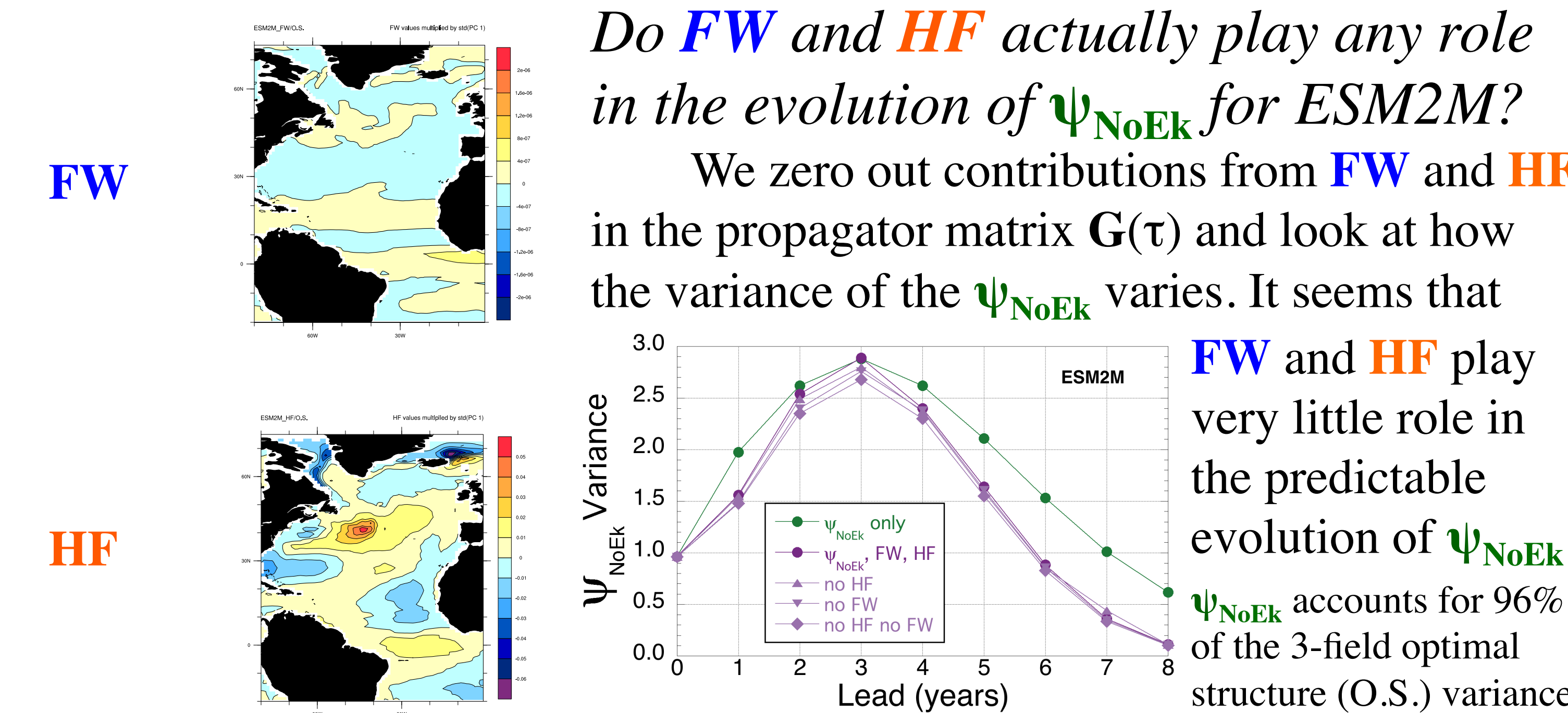
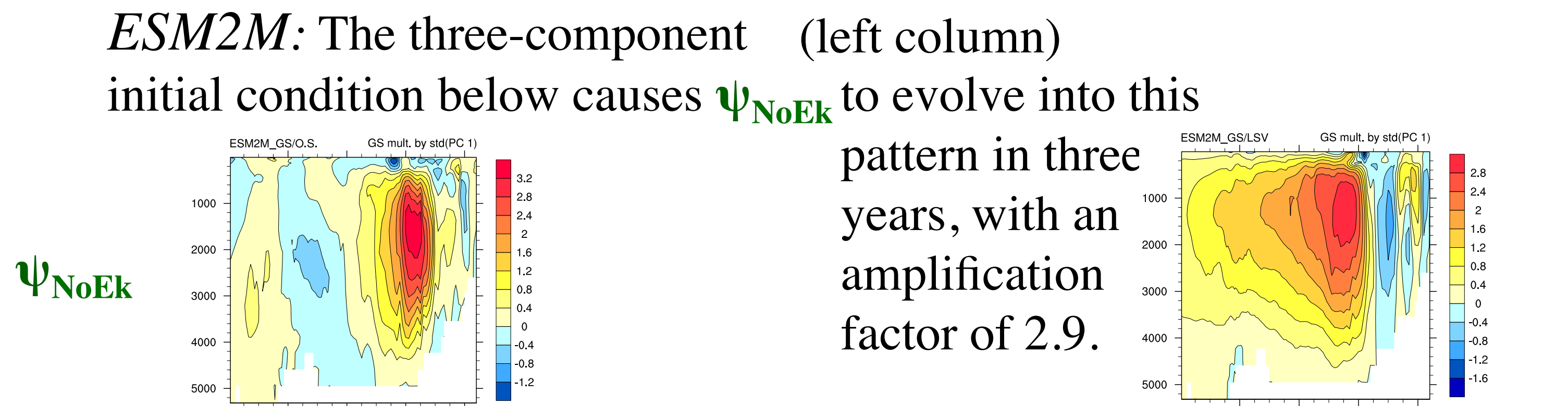
$$\frac{d\mathbf{x}}{dt} = \mathbf{L}\mathbf{x} + \boldsymbol{\xi} \quad \text{where } \mathbf{L} \text{ is a constant matrix and } \boldsymbol{\xi} \text{ is a vector of white noise. With this } \textit{ansatz}, \text{ the most probable prediction of } \mathbf{x}(t+\tau) \text{ given } \mathbf{x}(t) \text{ is } \mathbf{G}(\tau) \mathbf{x}(t),$$

where $\mathbf{G}(\tau) = \exp(\mathbf{L}\tau)$. Further, the initial condition evolving during time τ into the pattern having the largest possible amplitude of, say, Ψ_{NoEk} , is the leading eigenvector of $\mathbf{G}^T(\tau)\mathbf{W}\mathbf{G}(\tau)$, called the *optimal structure*. $\mathbf{W}$ is a weighting matrix ensuring that the amplitude of Ψ_{NoEk} is what is maximized. If $\mathbf{x}$ consists of Ψ_{NoEk} only, then $\mathbf{W}$ is the identity matrix.

Comparison of ESM2M and CCSM4 when $\mathbf{x}$ consists only of Ψ_{NoEk} :

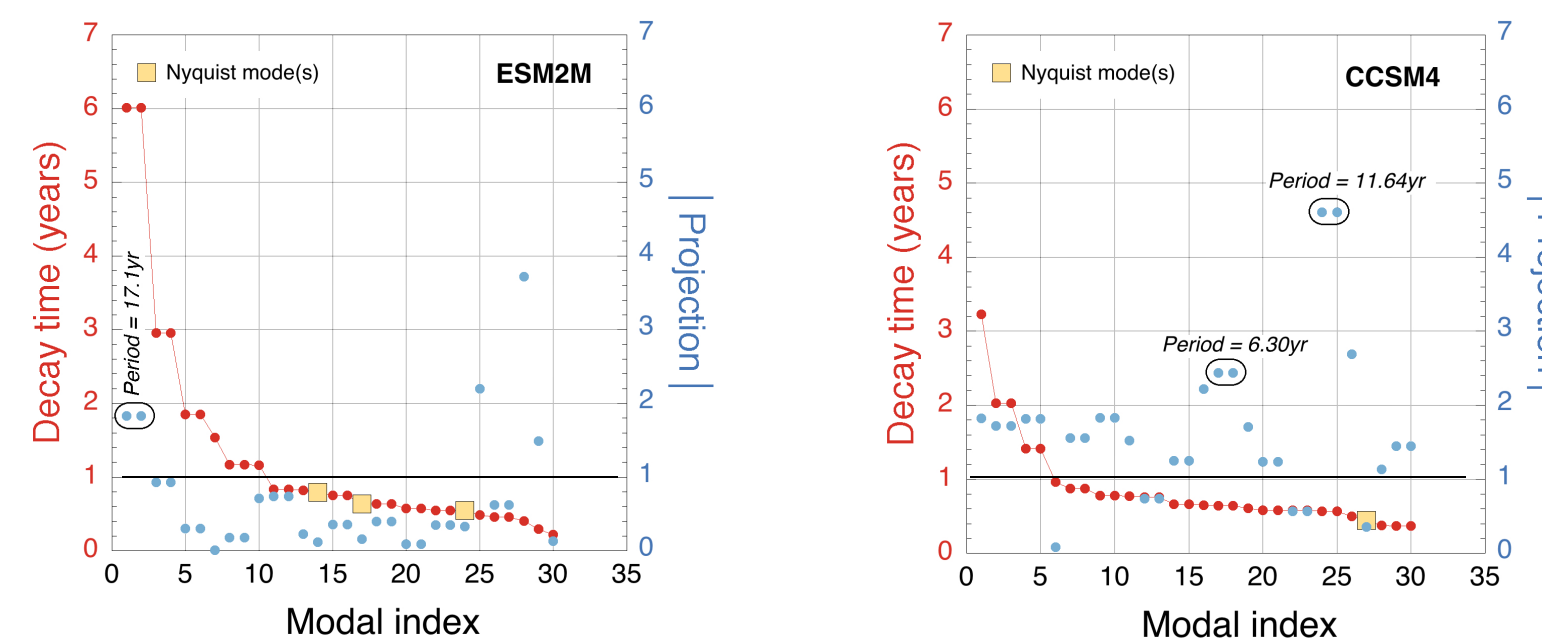


Comparison of ESM2M and CCSM4 when $\mathbf{x}$ consists of (Ψ_{NoEk} , **FW**, **HF**):



Conclusions

- Evolution of Ψ_{NoEk} in ESM2M appears to be due to an internal oscillation. Only 5 eigenvectors of $\mathbf{L}$ (i.e., “modes”), including a resolved (decay time > 1yr) oscillating pair, project significantly (> 1) on the Ψ_{NoEk} optimal structure. The 3-year optimal growth interval is consistent with an oscillation of 12-17 years.



- Evolution of Ψ_{NoEk} in CCSM4 appears to depend heavily on interactions with **FW** and **HF**. The Ψ_{NoEk} optimal structure significantly projects onto a large number of modes, most of them with decay times less than the output resolution (1yr).
- Unresolved (and, therefore, highly biased) modes project most strongly onto the Ψ_{NoEk} optimal structure. We defer analysis of modes themselves to subsequent work with monthly model output.
- These results are consistent with Transfer Function Analysis (talk with D. MacMartin).

Future work

- Develop verification method based on data
- Use monthly output to identify stochastic forcing
- Analyze other GCMs