

Decadal Multi-Model Potential Predictability

G.J. Boer

Canadian Centre for Climate Modelling
and Analysis

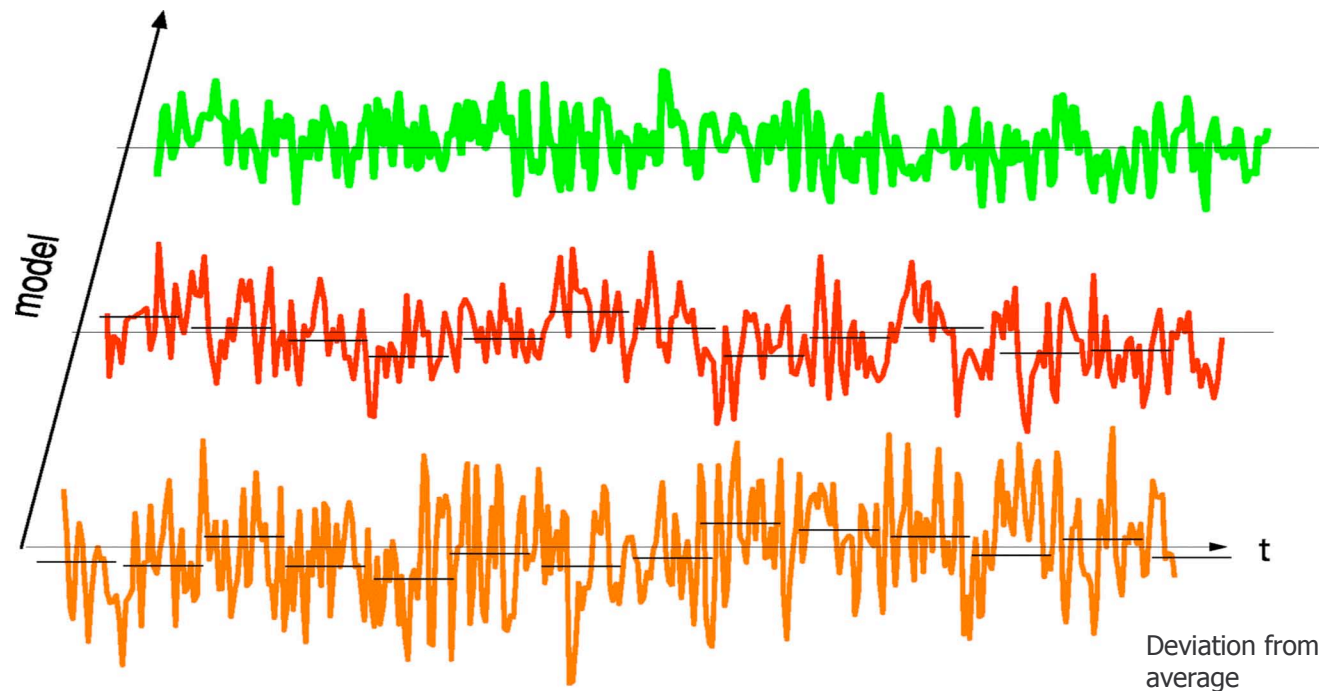
Environment Canada

University of Victoria

Predictability approaches

- *Classical predictability*
 - feature of the physical system
 - measures the rate of separation of initially close states
- *Potential predictability*
 - looks for the existence of deterministic long timescale variability
 - presumes this variability is “potentially” predictable with enough knowledge
 - location and nature of the potential predictability should suggest mechanisms and processes

Internally generated long timescale variability



Deviation from average

$$X(t) = X_{\alpha\bullet} + (X_{\alpha j} - X_{\alpha\bullet})$$

$$S^2 = S_v^2 + S_\varepsilon^2$$

M-year average

Statistical model

- Statistical model is

$$X = \Omega + \nu + \varepsilon$$

with associated variances

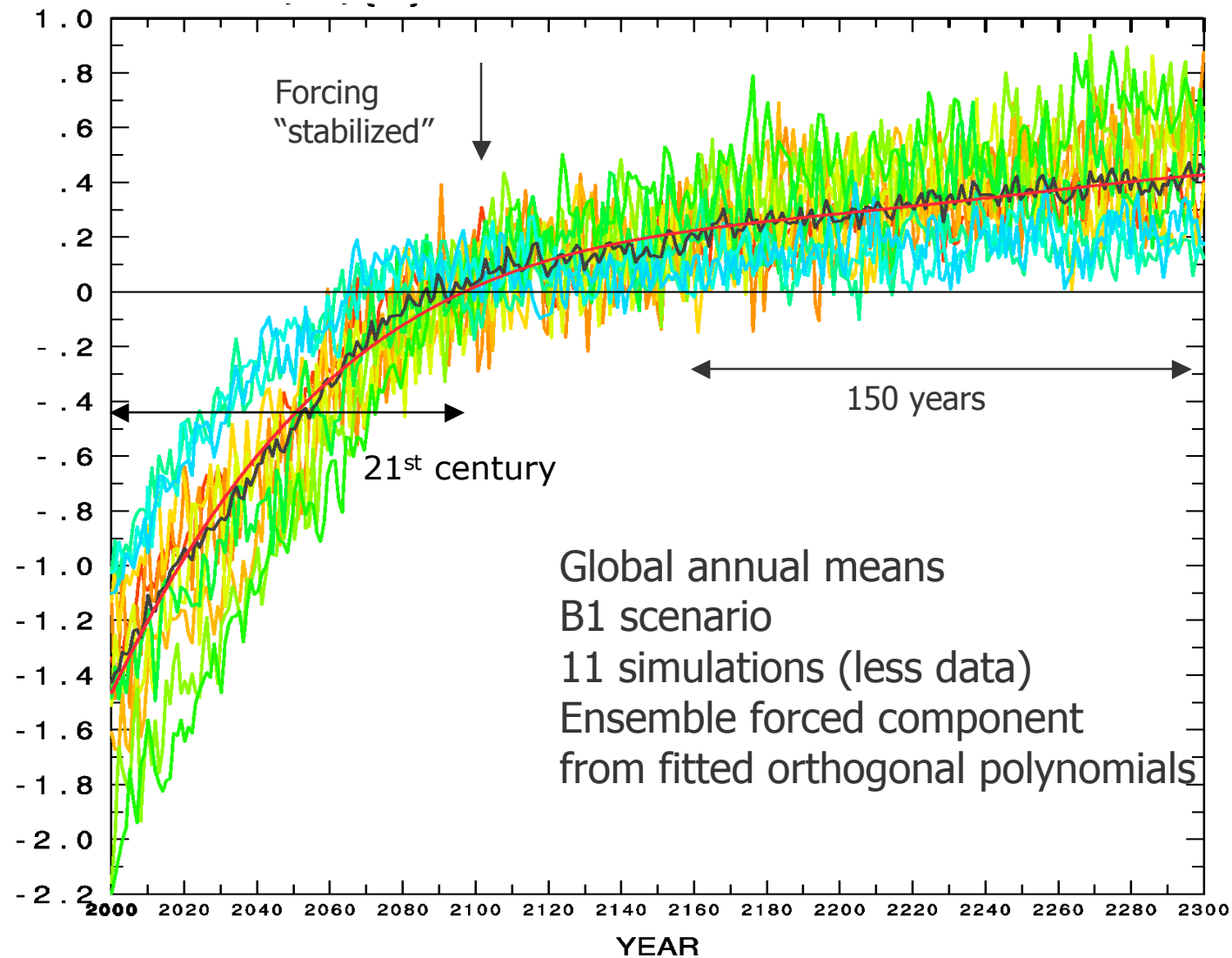
$$\sigma^2 = \sigma_{\Omega}^2 + \sigma_{\nu}^2 + \sigma_{\varepsilon}^2$$

- Ω is long timescale *externally forced* variability (if present)
- ν is long timescale *internally generated* variability
- ε is short timescale *unpredictable "noise"* variability

- Potential predictability variance fraction is

$$p = (\sigma_{\Omega}^2 + \sigma_{\nu}^2) / \sigma^2 = p_{\Omega} + p_{\nu}$$

Forced and internally generated variability



Approach

- Need suitable statistical tests and approaches
- Require lots of “observations” for statistical confidence
- Aim for geographic distribution of the *potential predictability variance fractions* (ppvfv)
- We take a *multi-model ensemble approach* using CMIP3 data (IPCC AR4)

Statistics

$$X_i = X_{(\alpha-1)M + j} = X_{\alpha j}$$

$$\begin{aligned} i &= 1 \dots NM \\ \alpha &= 1 \dots N \\ j &= 1 \dots M \end{aligned}$$

$$X_{\alpha j} = P_{\alpha} + (X_{\alpha \bullet} - P_{\alpha}) + (X_{\alpha \bullet} - X_{\alpha j})$$

$$X_{\alpha \bullet} = \frac{1}{M} \sum_{j=1}^M X_{\alpha j}$$

is M-year average

$$S^2 = S_{\Omega}^2 + S_V^2 + S_{\varepsilon}^2 = \overline{X^2}$$

$$= \overline{P_{\alpha}^2} + \overline{(X_{\alpha \bullet} - P_{\alpha})^2} + \overline{(X_{\alpha \bullet} - X_{\alpha j})^2}$$

$$P_{\alpha} = \sum_{k=1}^K a_k p_k(\alpha)$$

is orthogonal
polynomial fit

Long TS
forced
(if present)

Long TS
internally
generated

Short TS
noise

Statistics are pooled across models in multi-model case

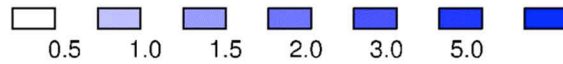
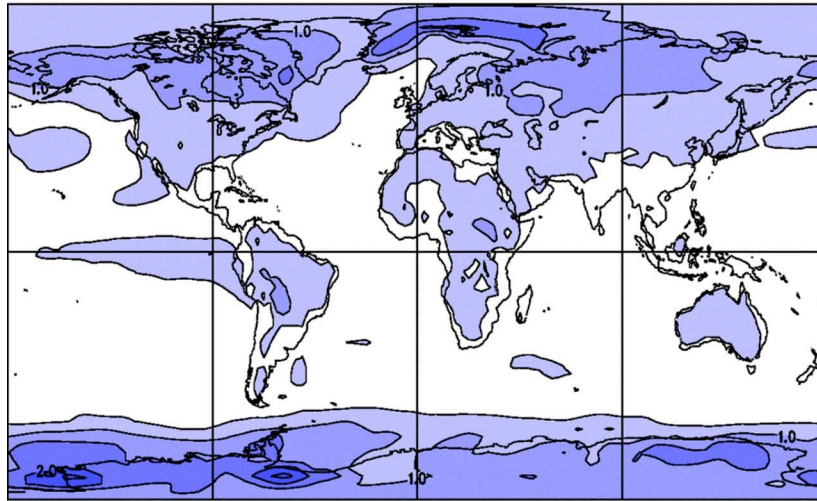
Apply to CMIP3 control climates

- (intended to be) equilibrium climate
- no external forcing – we consider the *internally generated* variability
- Potential predictability:
 - measured as *fraction of variance* $p_v = \sigma_v^2 / \sigma^2$
 - indication of *relative importance*
 - expect low p_v where σ_v^2 is low **or** σ^2 high
- results from 27 models
- simulations lengths from 100 to 1000 years
- we consider surface air temperature and precipitation (the two main climate parameters)
- measures potential predictability in the *model world*

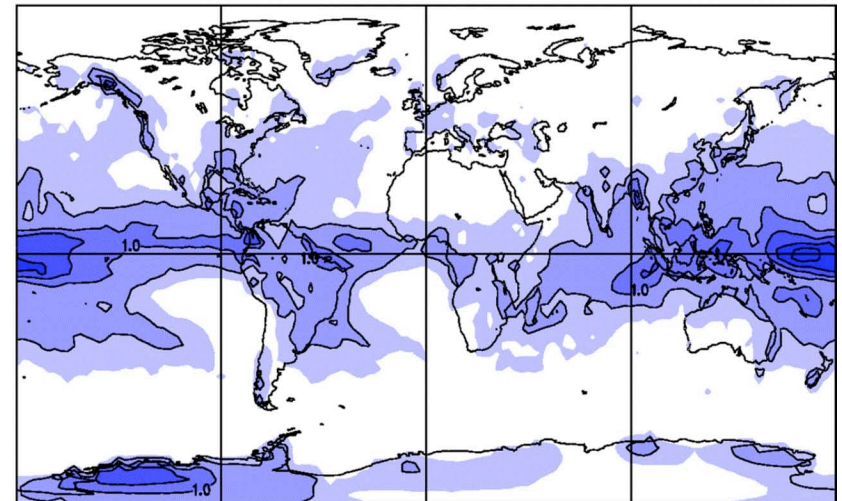
Standard Deviation of annual means

Observation-based

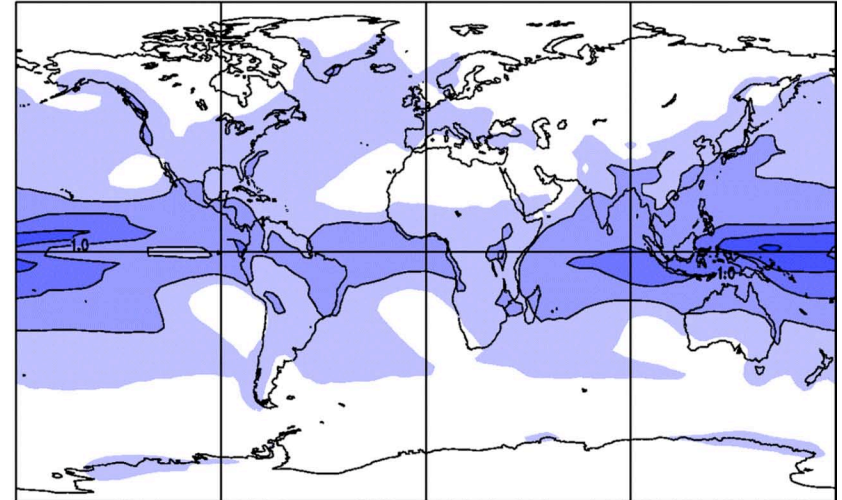
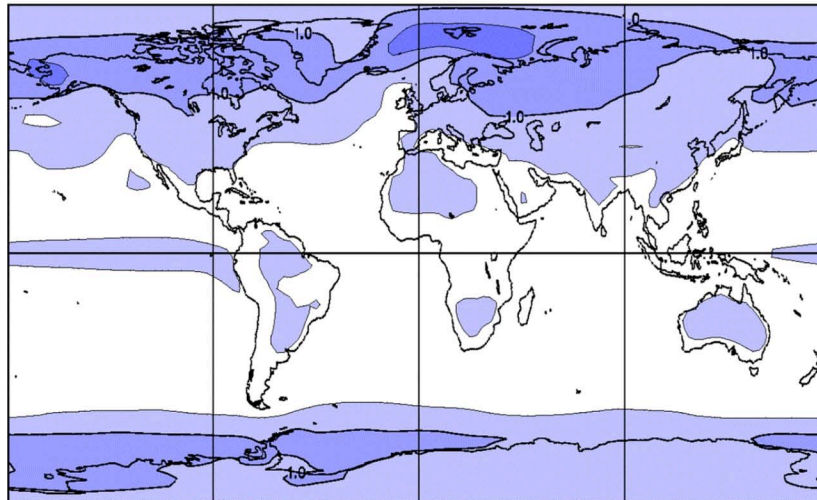
Temperature



Precipitation

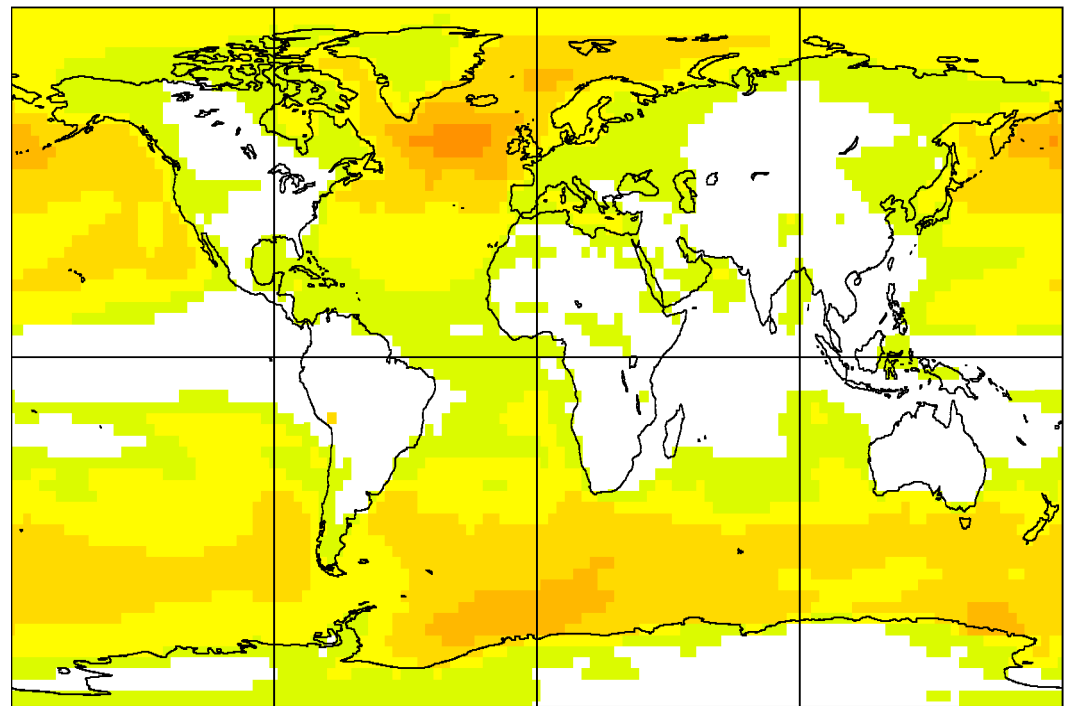


Multi-model ensemble



Temperature: potential predictability variance
fraction $p_v = \sigma_v^2 / \sigma^2$ (%) for *decadal means*

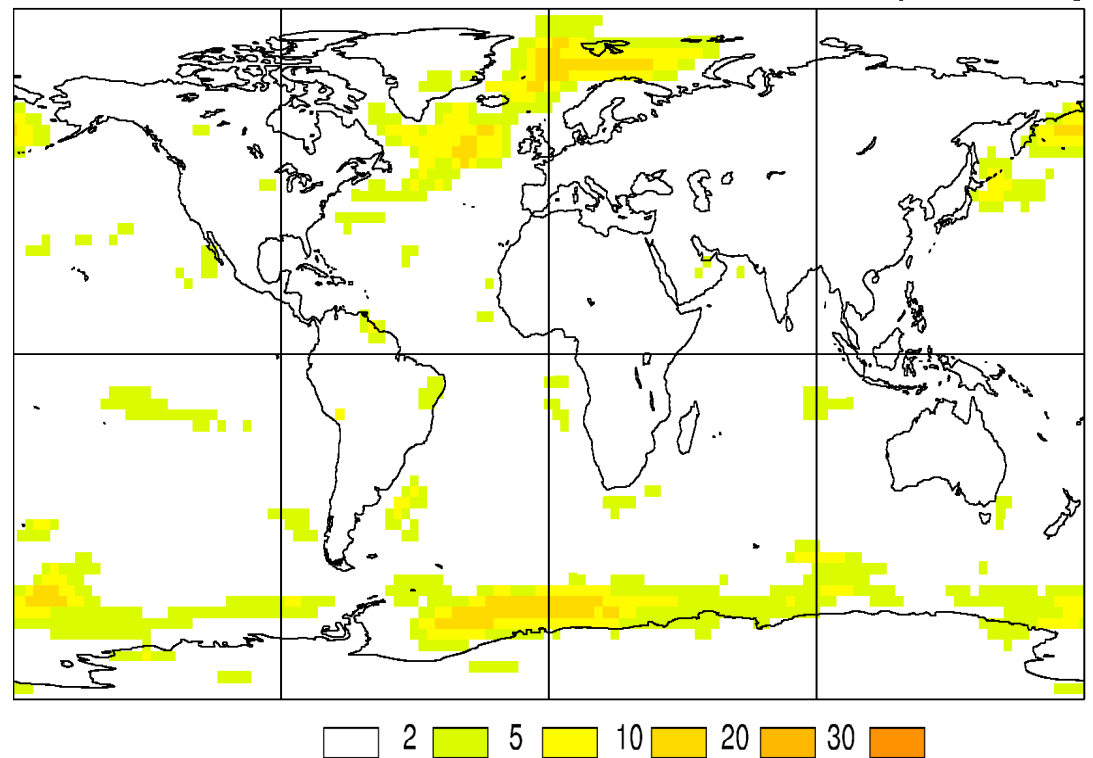
- Ratio of “predictable” to total variance
- MME provides stability of statistics: *ppvf* in white areas <2% and/or not significant at 98% level
- Long timescale predictability found mainly over oceans
- Some incursion into land areas but modest *ppvf* (*denominator* is large)



Control simulations

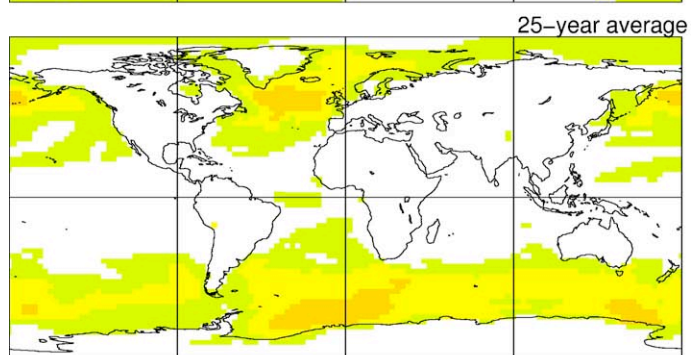
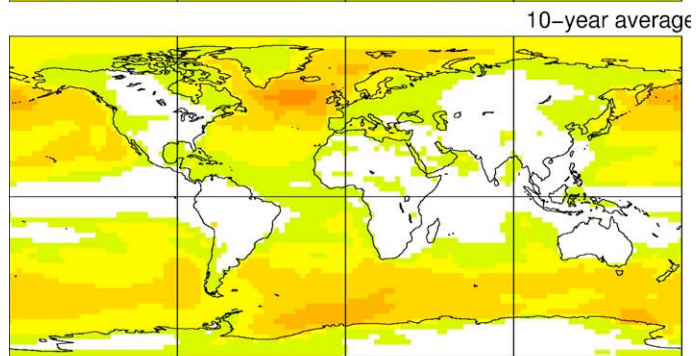
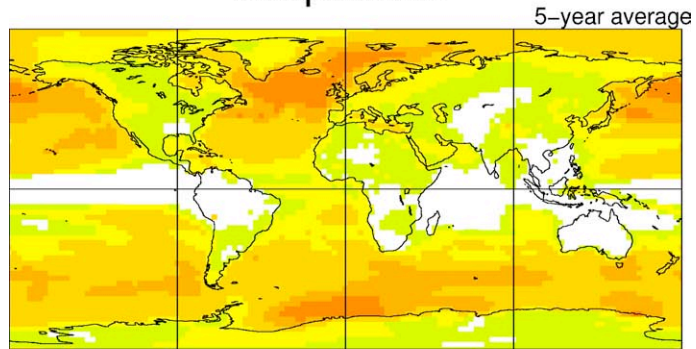
Precipitation: potential predictability variance fraction $p_v = \sigma_v^2 / \sigma^2$ (%) for decadal means

- MME provides “some” significant areas of precipitation
- Much less potentially predictable than temperature
- Little incursion into land areas
- Precipitation predictability a weakened version of temperature predictability at these timescales

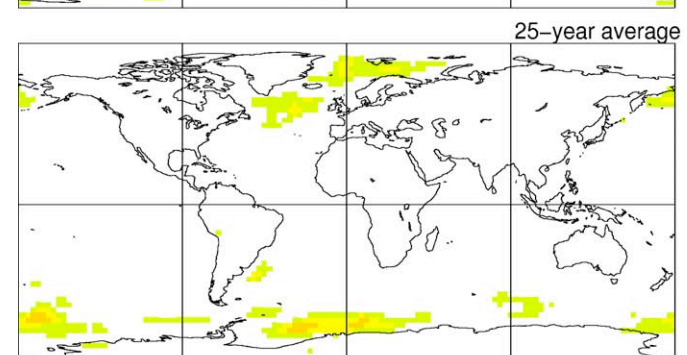
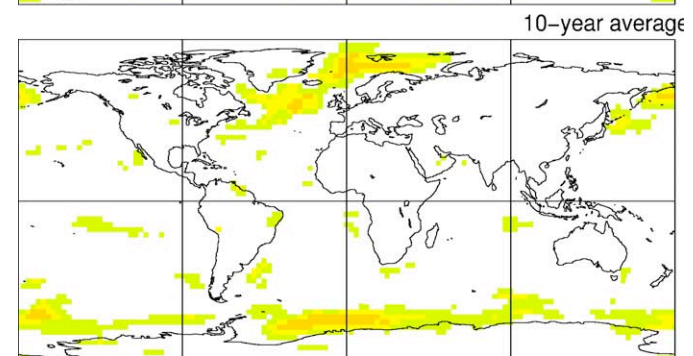
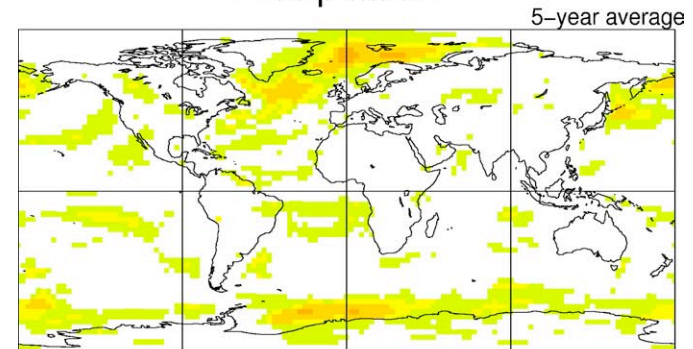


Control simulations

Temperature



Precipitation

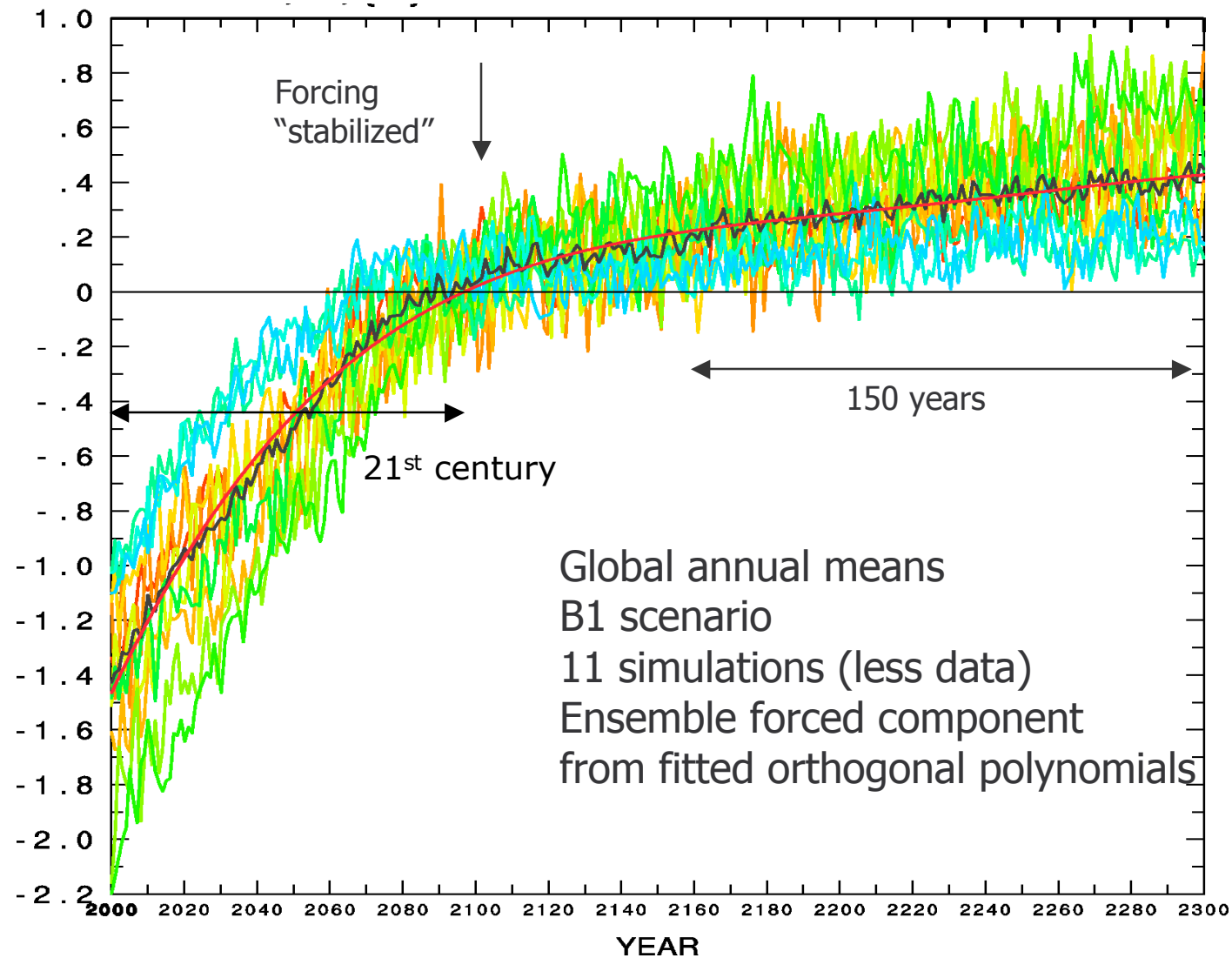


Proposed experiment allows us to consider also “pentades

21st Century potential predictability of forced and internal decadal variability

- B1 Scenario
- period is from 2000 to 2100
- CMIP3 multi-model approach
- only 11 simulations for full data period (up to 2300)
- initial illustrative calculation of variance components

Forced and internally generated variability



Variances

- $X_{ij} = W_{ij} + x_{ij}$
 - i is decade number and j year within it
 - W_{ij} is representation of *forced component* from polynomial fit
 - x_{ij} is variation about forced component
- $Y_{ij} = W_{i.} + x_{i.} + (x_{ij} - x_{i.})$
 - $W_{i.} + x_{i.}$ are contributions to decadal potentially predictable variance from forced and internal variability
 - $(x_{ij} - x_{i.})$ is noise

Multi-model ensemble approach

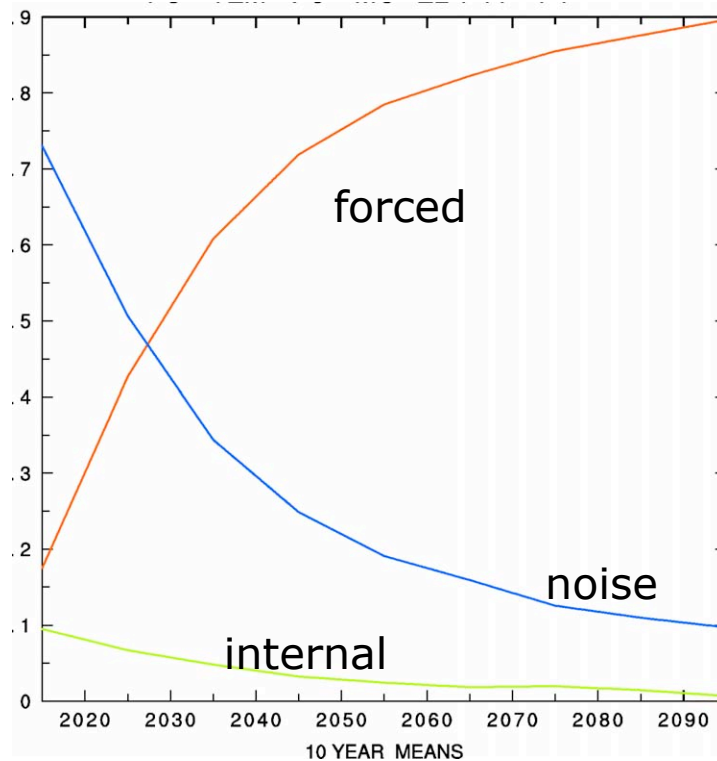
- potential predictability without the statistics (i.e. approximate)
- $\{Y^2\} = \underbrace{\{W_{i.}^2\}}_{\text{forced}} + \underbrace{\{x_{i.}^2\}}_{\text{internal}} + \underbrace{\{(x_{ij} - x_{i.})^2\}}_{\text{noise}}$
 - $\{ \}$ is average over j and over ensemble
- Variance fractions for each decade
$$p_F = \{W_{i.}^2\} / \{Y^2\}$$
$$p_I = \{x_{i.}^2\} / \{Y^2\}$$
$$p = p_F + p_I$$
- forced component dominates for longer prediction times

Predicting for the next decade

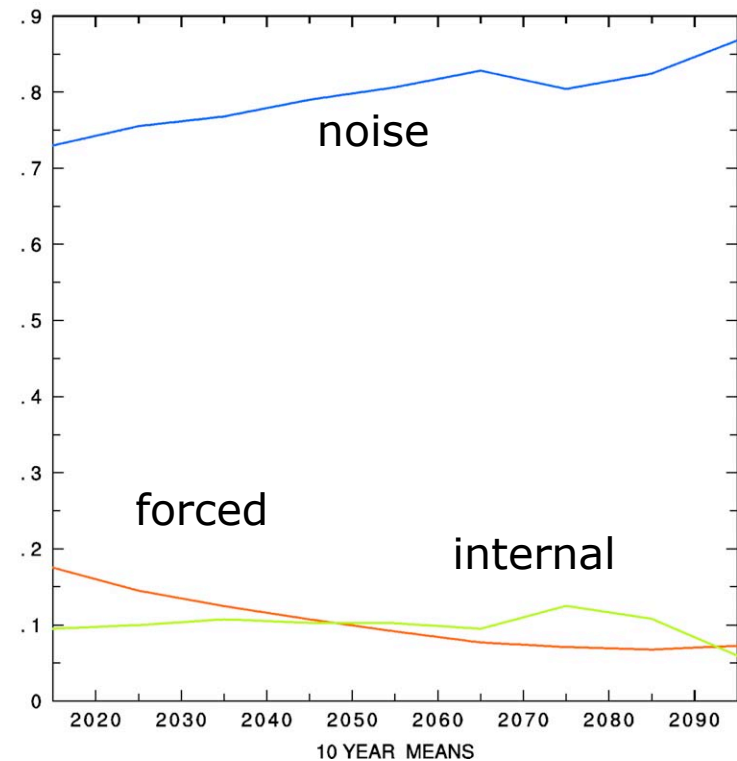
- $Z_{ij} = (W_{i\cdot} - W_{i-1\cdot}) + x_{i\cdot} + (x_{ij} - x_{i\cdot})$
 - take the decadal *change* in forced component to be the forecast information for the next decade
- $\{Z^2\} = \{(W_{i\cdot} - W_{i-1\cdot})^2\} + \{x_{i\cdot}^2\} + \{(x_{ij} - x_{i\cdot})^2\}$
- Variance fractions each decade
$$p_F = \{(W_{i\cdot} - W_{i-1\cdot})^2\} / \{Z^2\}$$
$$p_I = \{x_{i\cdot}^2\} / \{Z^2\}$$
$$p = p_F + p_I$$

Decadal variance *fractions*: Temperature

Multi-decade from end of
20th century

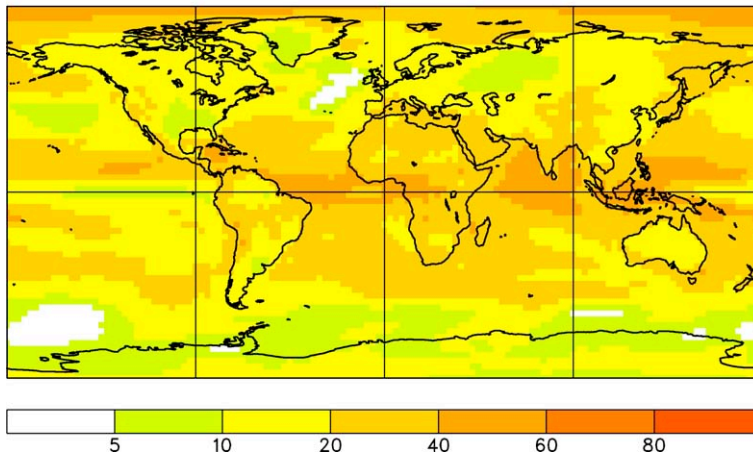


Next decade within the
21st century

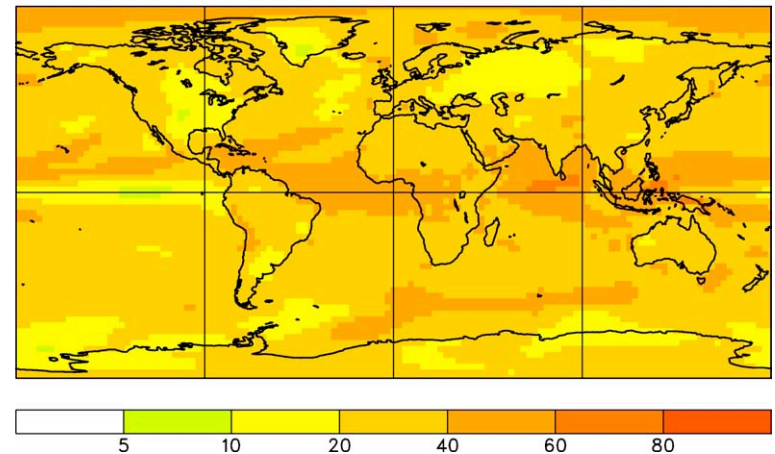


Variance fractions for decade 2010-2020: Temperature

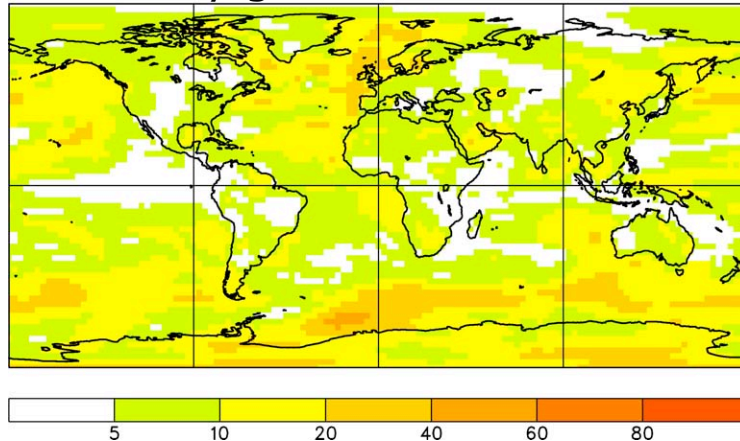
Forced



Net



Internally generated

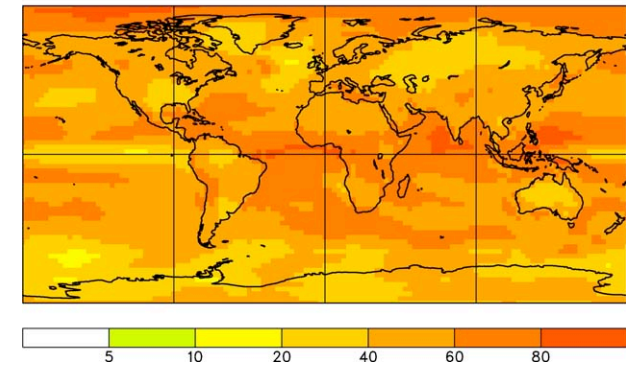


- Although forced component is largest over high latitude land, *fractional* variance component is not
- Internally generated component of "similar" size (and resembles control run results)
- Net fractional decadal variance largest over tropical oceans

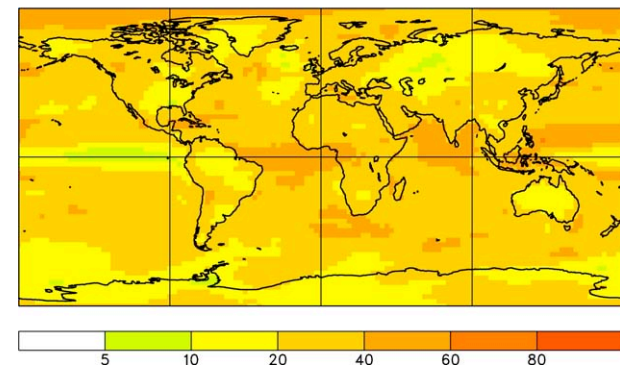
Net variance fractions for decade 2020-2030: Temperature

- early 21st century
- “multi-year” view
 - forced component grows and soon dominates
 - *fractionally* is largest over tropical oceans
 - internal variability diminishes as a *fraction*
- for “next decade” view
 - forced component comparable to internal

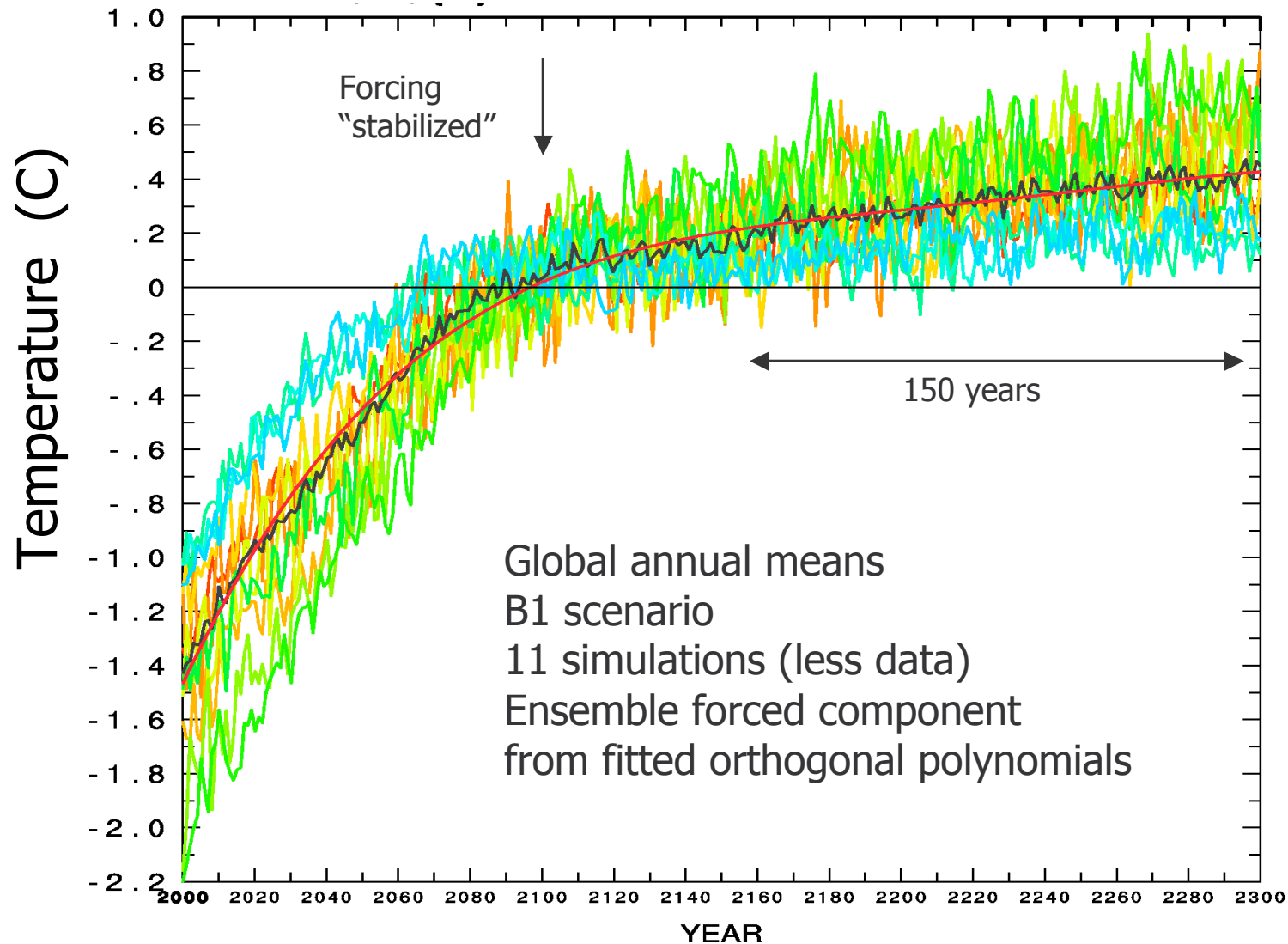
multi-decade



next decade



Potential predictability of internal component for a warmer world

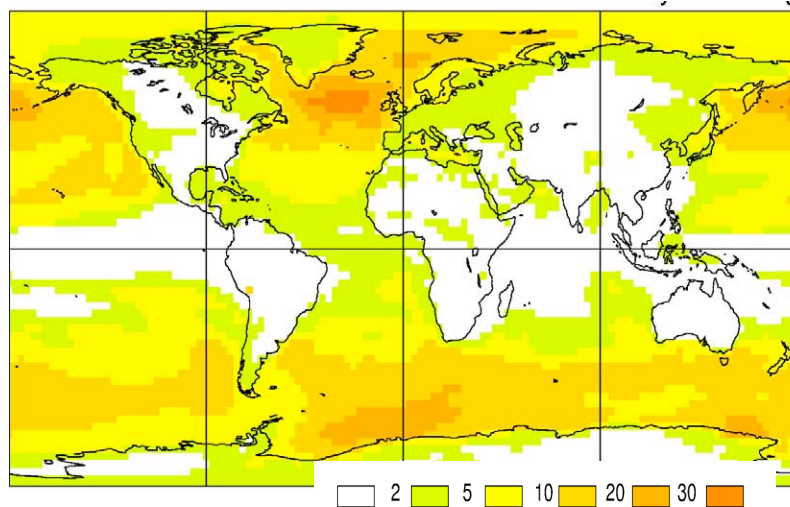


Potential predictability in a warmer world (stabilization case)

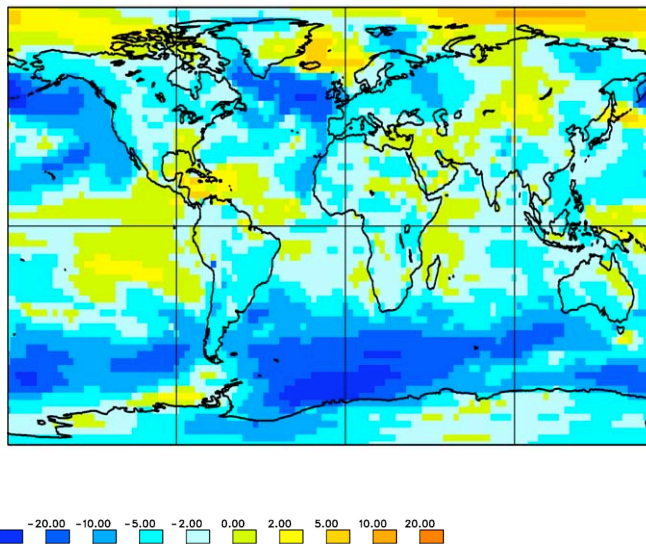
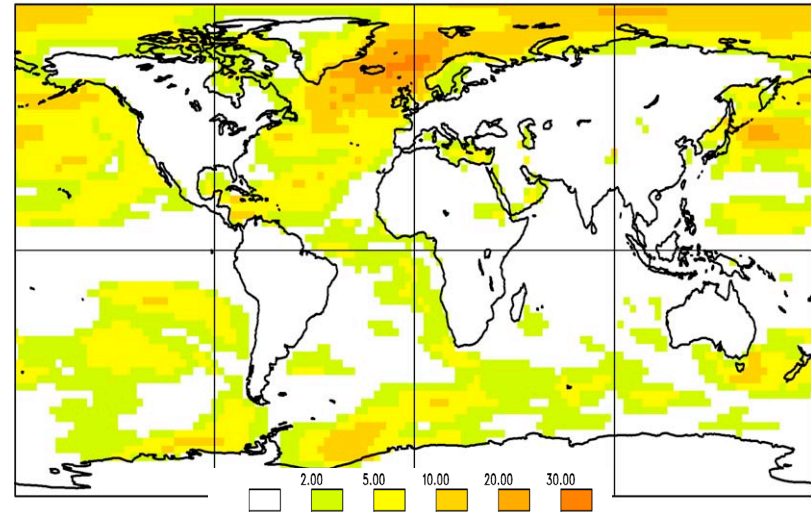
- Consider long timescale *internally generated* natural variability σ_v^2
 - last 150 years of stabilization simulations
 - remove *forced polynomial trend*
 - estimate potential predictability in warmer world
 - estimate change in potential predictability from control case

Decadal potential predictability p_v for *Temperature*

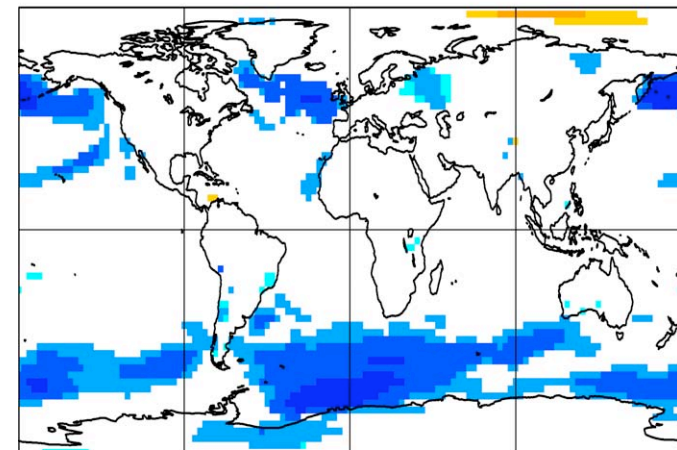
Control simulation



B1 stabilization scenario



Difference in warmer world



Where confidence bands *don't* overlap

MME decadal potential predictability of temperature and precipitation

- model based measure
- potential predictability of the *unforced control* climate
 - fish over folk (especially for precipitation)
 - shorter the better
 - “hot spots” over extratropical oceans for both temperature and precipitation
 - comparatively little potential predictability over land and tropical oceans
 - predictability found for regions/processes where surface connects to deeper ocean

MME decadal potential predictability of temperature and precipitation

- potential predictability in the 21st century
 - adding forced component alters picture
 - tropics become more important
 - forced component soon dominates for *multi-decadal* forecasts
 - forced component and internal component comparable for *next decade* forecast
- potential predictability of unforced variability *decreases* in warmer world

The challenges of potential predictability

- to identify the mechanisms associated with regions of high potential predictability
- to understand the lack of potential predictability over land and, for unforced variability, tropical oceans
- to test potential predictability results by means of (multi-model) prognostic decadal predictions

End of presentation