

Forced and internally generated 21st century decadal “potential predictability”

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Long timescale climate predictability

- “early days” of *coupled* climate change simulations
- other’s interest in *forced signal*
- then, signal “contaminated” with internally generated natural variability “*noise*”
- but perhaps long timescale variability is *predictable*
- when forcing is weak one can investigate predictability of this *internally generated* component
- but for 21st century we want predictability of both *forced* and *internal components*

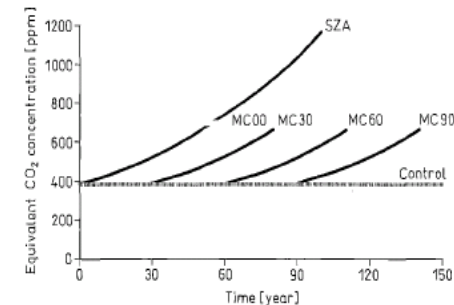
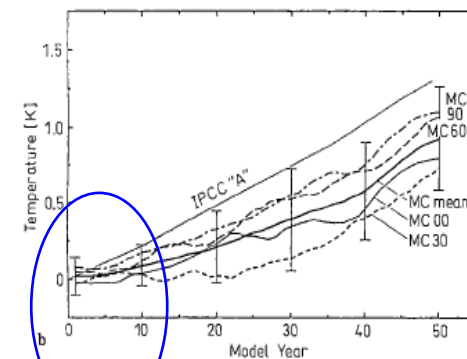


Fig. 1. Schematic diagram of the “Monte Carlo” climate forecasts



Monte Carlo climate change forecasts with a global coupled ocean-atmosphere model

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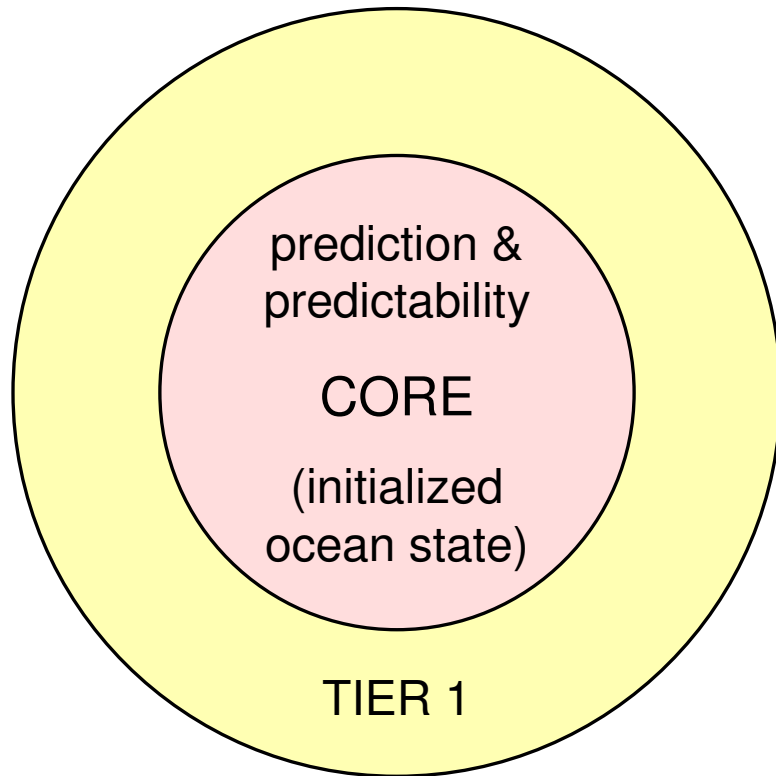
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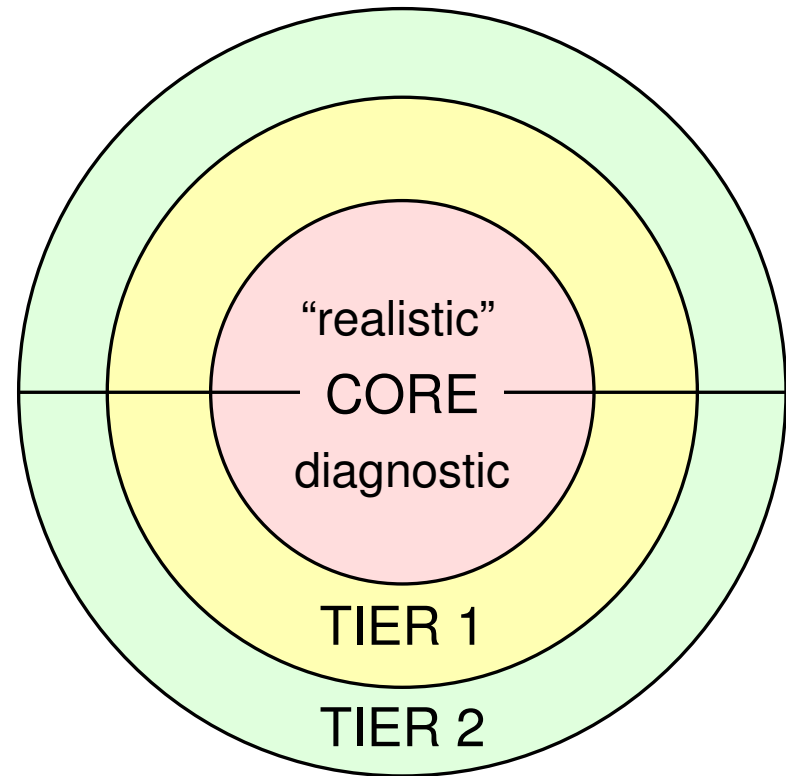
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CMIP5 Experiment Design

“Near-Term”
(decadal)



“Long-Term”
(century & longer)



Decadal predictability and prediction

- Appeals to “long timescale” processes
 - *externally forced* (GHG+A, volcanoes, solar,)
 - *internally generated*
 - oceanic mechanisms (AMO, SO, ...)
 - coupled processes
 - PDO, AMO, NPMO, PGO, ENSO...
 - modulation of “atmospheric” modes (PNA, NAO, NAM, SAM,)
 - atmospheric processes (QBO, ...)

How do we determine the *predictability* of the system on decadal timescales?

- Prognostic perfect model predictability studies
 - Griffies and Bryan (1997)
 - *Boer (2000)*
 - Collins (2002)
 - Collins et al. (2006)
 - Latif et al., (2006)
 - and others
- Diagnostic potential predictability studies
 - Boer (2000, 2004)
 - Pohlmann et al. (2004)
 - Predicate (2004...)
 - Boer and Lambert (2008)
 - *Boer (2010)*
 - and others
- Investigations of forecast skill
 - Smith et al. (2008)
 - Keenlyside et al. (2008)
 - Pohlmann et al. (2009)
 - *Merryfield et al. (2010)*
 - CMIP5 (2010+)

Potential predictability: internally generated component

- *decadal, diagnostic, multi-model*

- Model control runs (CMIP3) - no external forcing

- Annual means of variable X are expressed as

$$X = \mu + v + \varepsilon$$

- μ is the long-term mean

- v is the long timescale *internally generated* component

- ε is the short timescale *unpredictable "noise"* component

- Associated variances are

$$\sigma^2 = \sigma_v^2 + \sigma_\varepsilon^2$$

- Potential predictability variance fraction (*ppvf*) is

$$p = \sigma_v^2 / \sigma^2$$

Internally generated multi-model potential predictability

- Potential predictability variance fraction

$$p = \sigma_v^2 / \sigma^2 = \sigma_v^2 / (\sigma_v^2 + \sigma_\varepsilon^2)$$

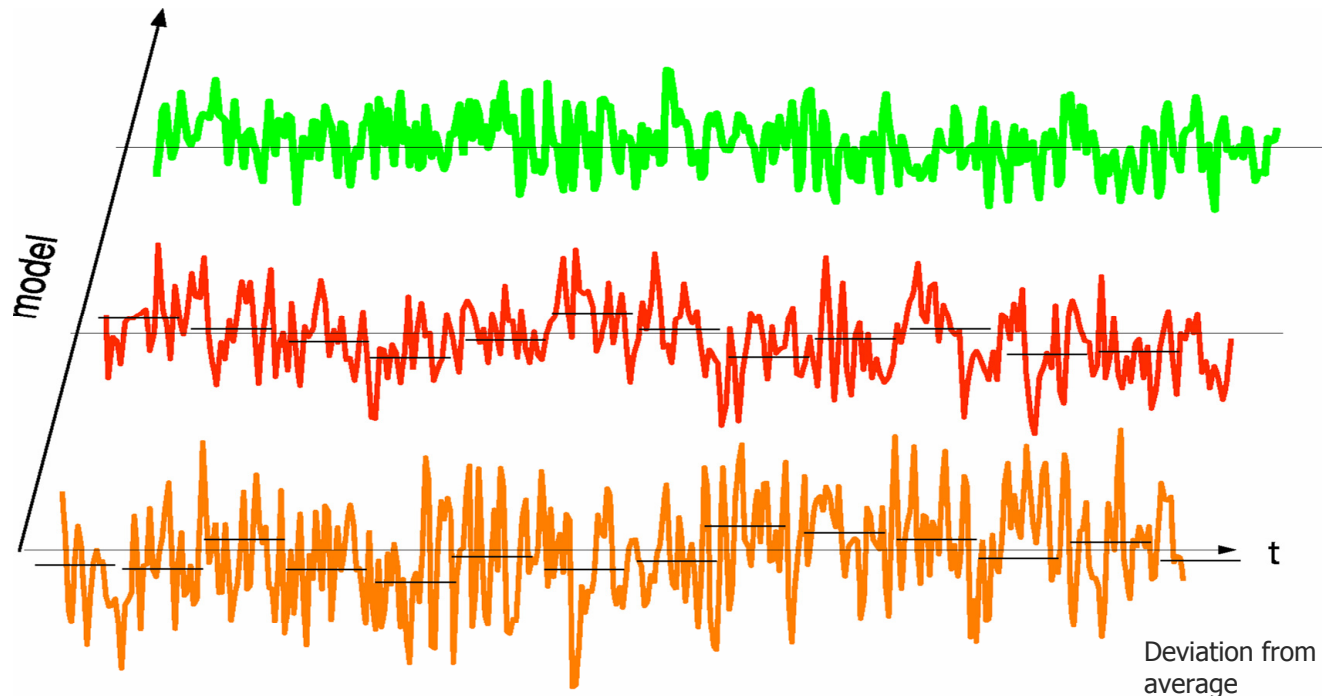
- in terms of a signal to noise measure

$$\gamma = \sigma_v^2 / \sigma_\varepsilon^2$$

$$p = \gamma / (1 + \gamma)$$

- p is small if signal is *small* or if noise is *large*
 - $0 < p < 1$
 - *not only existence of signal, however small, but its relative magnitude*
 - *a variance measure (not a correlation like measure)*

Internally generated long timescale potential predictability



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

$$\sigma^2 = \sigma_v^2 + \sigma_\varepsilon^2$$

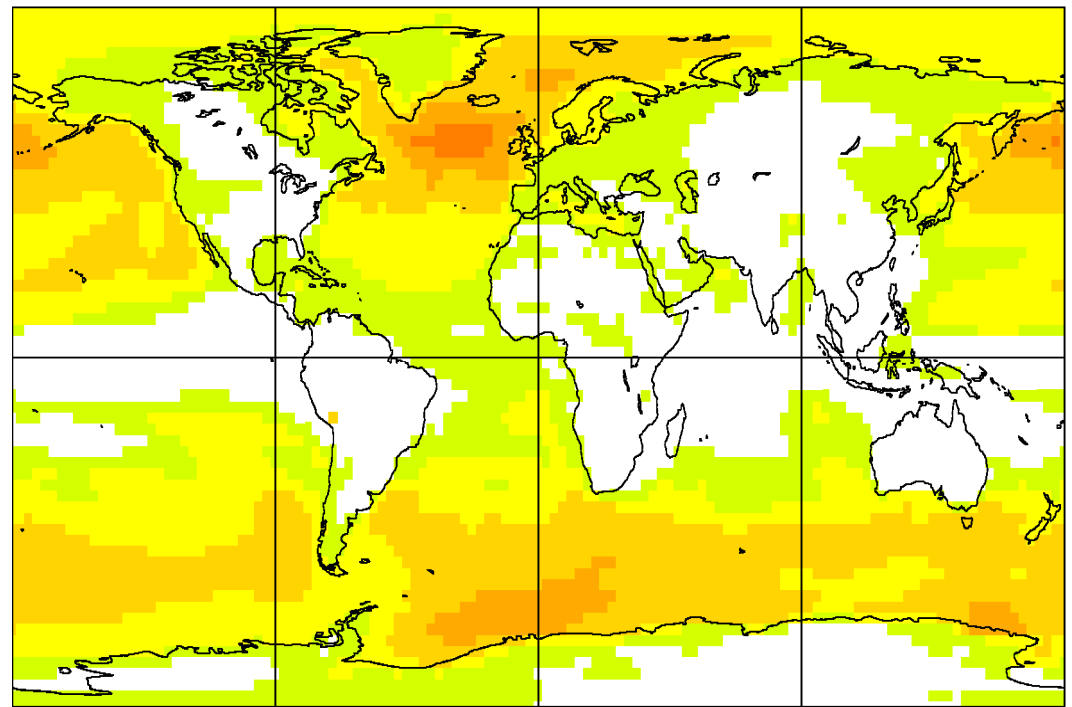
Deviation from average
 $X_{\alpha j} - X_{\alpha\bullet}$
 M-year average
 $X_{\alpha\bullet}$

Virtues of multi-model approach

- the “multi-model” is generally the “best model”
 - no individual model “best” in all regards
 - the “ n -best” models differ with criterion used
 - pooled climate statistics (means, variances, covariances) generally closer to observed
 - applied to seasonal forecasting
 - applied to climate change (Chapter 10, AR4)
- increased the amount of data for statistical stability

Temperature: potential predictability of *internally generated* variability $p_v = \sigma_v^2 / \sigma^2$ (%) for *decadal means* (CMIP3 multi-model control runs)

- Ratio of long timescale 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)

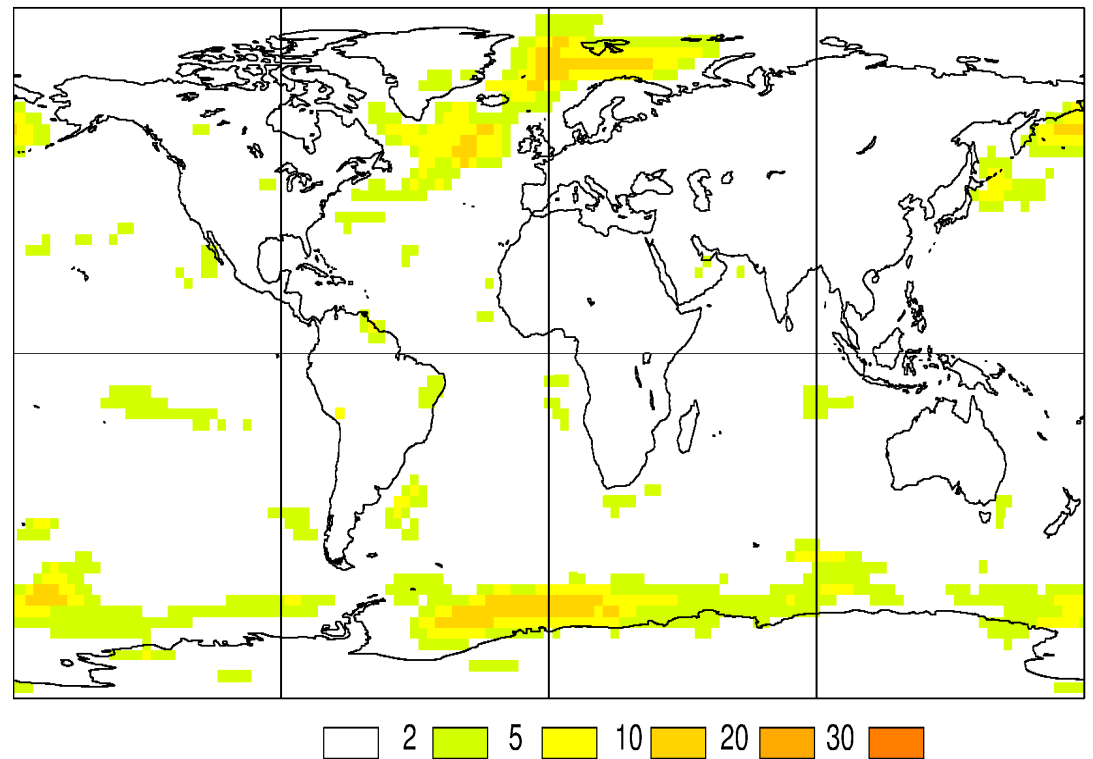


2 5 10 20 30

Control simulations

Precipitation: potential predictability of internally generated variability $\rho_v = \sigma_v^2 / \sigma^2$ (%) 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

21st Century decadal potential predictability

- Variable *now has forced component*

$$X = \mu + \Omega + v + \varepsilon$$

with associated variances

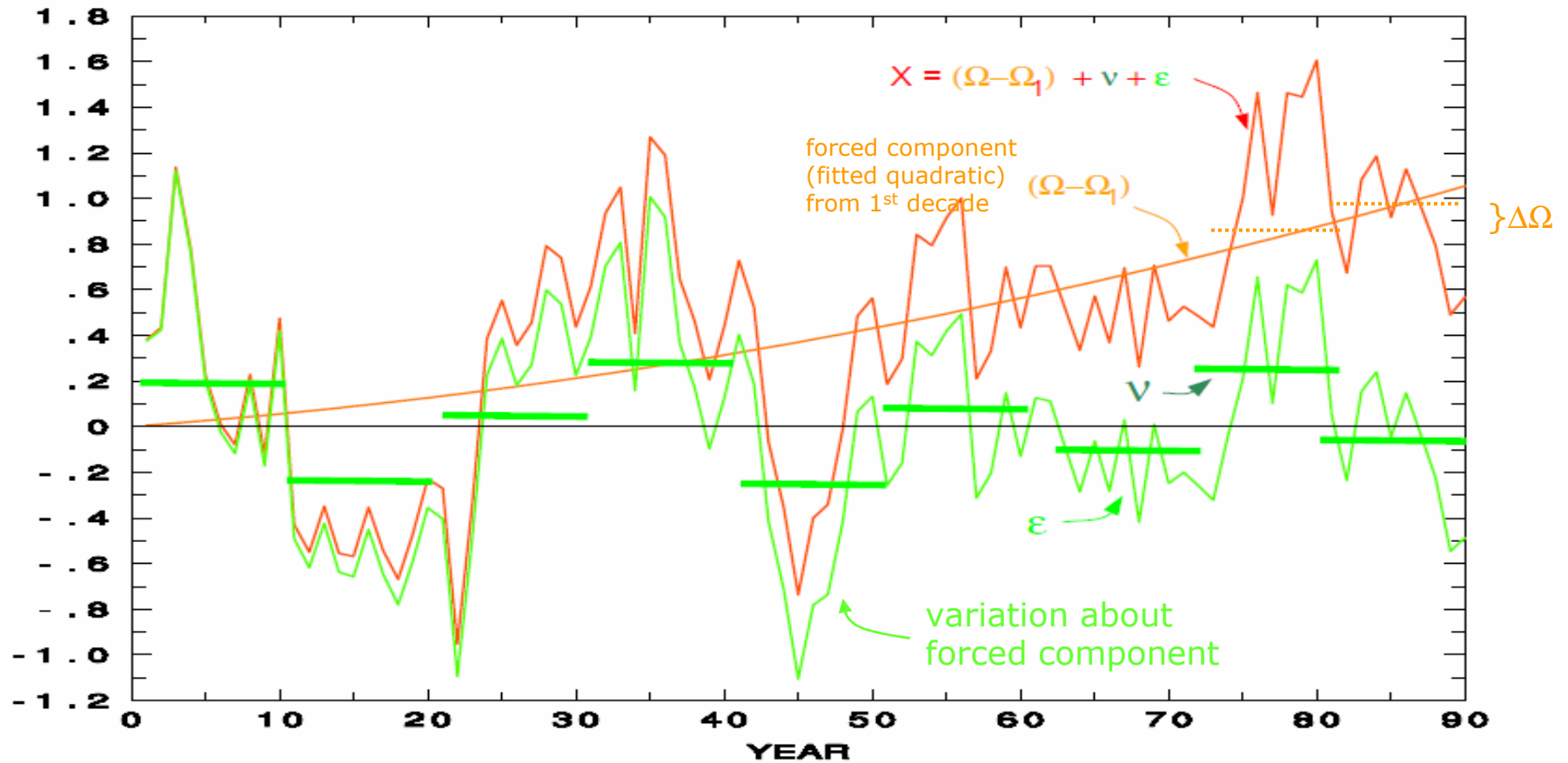
$$\sigma^2 = \sigma^2_{\Omega} + \sigma^2_v + \sigma^2_{\varepsilon}$$

- Ω is long timescale *externally forced* variability
 - obtained by fitting 2nd order orthogonal polynomial
- v is long timescale *internally generated* variability
- ε is short timescale *unpredictable "noise"* variability
- statistics pooled across models
- Potential predictability variance fraction now has two components

$$p = (\sigma^2_{\Omega} + \sigma^2_v) / \sigma^2 = p_{\Omega} + p_v$$

21st century temperature at a point

- forced component from 1st decade



$$\sigma_1^2 = \sigma_{\Omega_1}^2 + \sigma_v^2 + \sigma_\varepsilon^2 \quad \text{multi-decade}$$

$$\sigma_\Delta^2 = \sigma_{\Delta\Omega}^2 + \sigma_v^2 + \sigma_\varepsilon^2 \quad \text{next-decade}$$

Forced component

- Potential predictability variance fraction

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

- *multi-decade* view of forced contribution
 - difference from 1st decade

$$\sigma_{\Omega 1}^2 = (\Omega_k - \Omega_1)^2$$

$$p = (\sigma_{\Omega 1}^2 + \sigma_v^2) / (\sigma_{\Omega 1}^2 + \sigma_v^2 + \sigma_{\varepsilon}^2) = p_{\Omega 1} + p_{v1}$$

- *next -decade* view of forced contribution
 - difference from previous decade

$$\sigma_{\Delta\Omega}^2 = (\Omega_k - \Omega_{k-1})^2$$

$$p = (\sigma_{\Delta\Omega}^2 + \sigma_v^2) / (\sigma_{\Delta\Omega}^2 + \sigma_v^2 + \sigma_{\varepsilon}^2) = p_{\Delta\Omega} + p_{\Delta v}$$

- both numerator and denominator differ so p components differ depending on treatment of forced component

Estimate statistics from sample variances

$$\hat{\sigma}_{\varepsilon}^2 = \frac{m}{m-1} S_{\varepsilon}^2$$

$$\hat{\sigma}_{\nu}^2 = \frac{n}{n-(b+1)} S_{\nu}^2 - \frac{S_{\varepsilon}^2}{m-1}$$

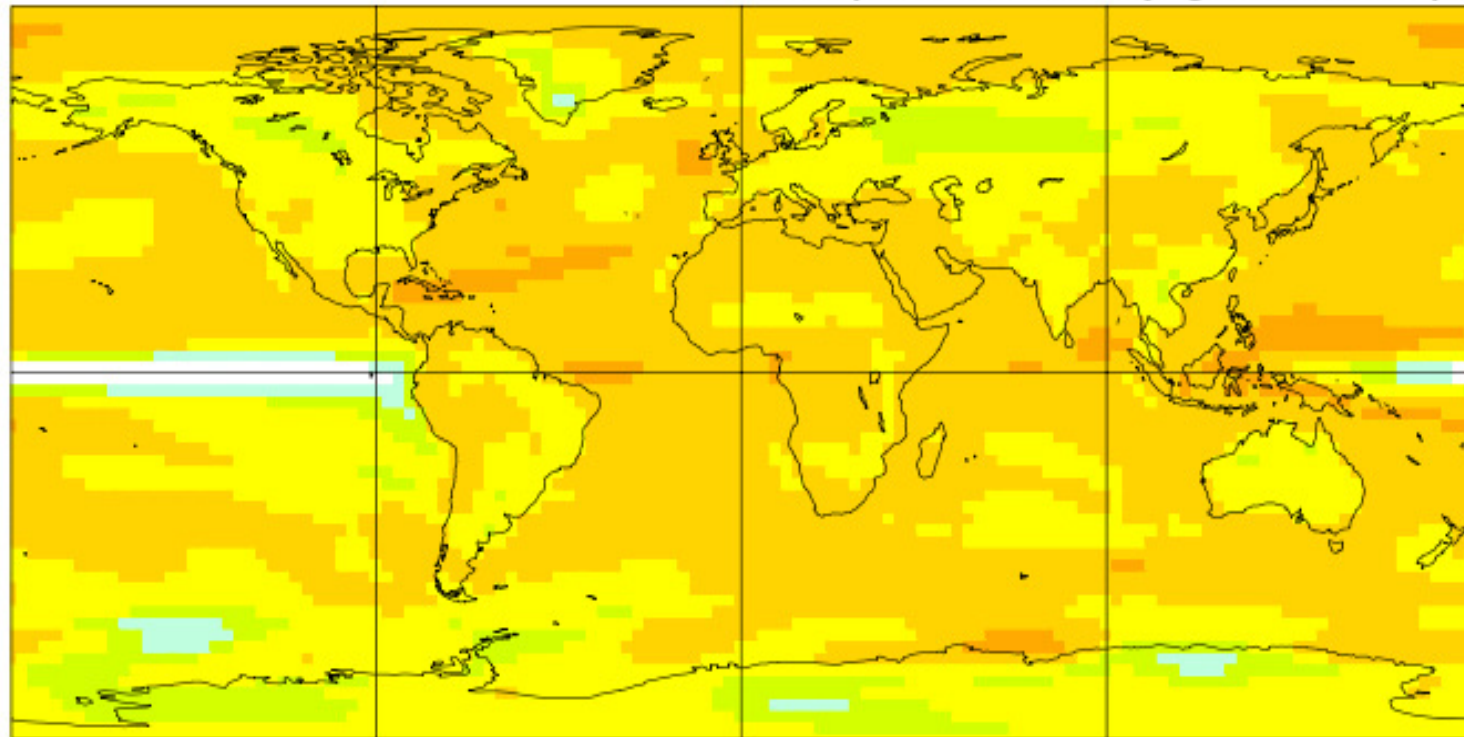
$$\hat{\sigma}_{\Omega 1}^2 = S_{\Omega 1}^2 - \frac{d_1}{n-(b+1)} S_{\nu}^2$$

$$\hat{\sigma}_{\Delta \Omega}^2 = S_{\Delta \Omega}^2 - \frac{d_{\Delta}}{n-(b+1)} S_{\nu}^2$$

- S^2 are sample variances pooled across models
- $m = 10$ years in a decade; $n = 10$ decades in 21st century
- b, d 's arise from the fitting polynomial for the forced component
- decadal sample variance is discounted by part of noise variance
- decadal forced variance discounted by part of decadal variance

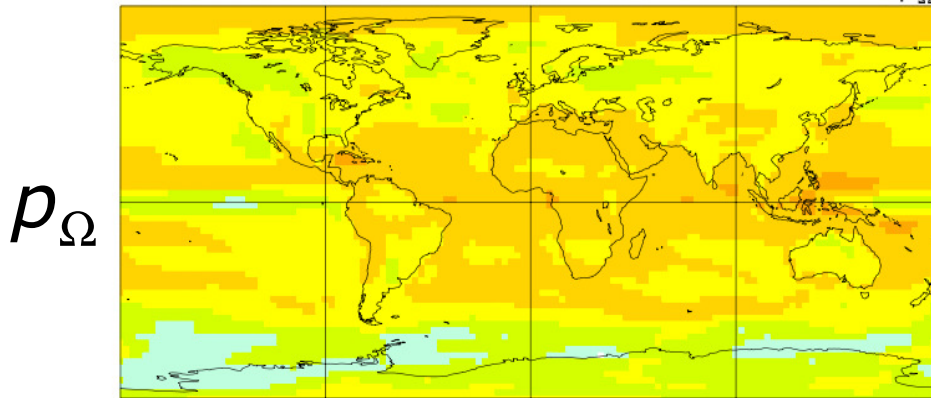
*Net potential predictability of temperature for
2010-20 (or “next decade” result generally)*

Forced plus internally generated p

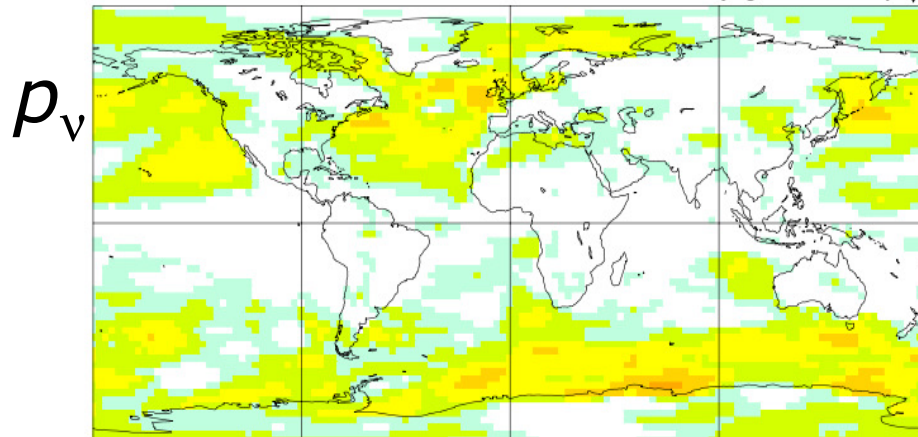


Forced component of potential predictability for temperature and forced component of temperature change for 2010-20

Potential predictability variance fractions: 2010–20
Forced p_{Ω}



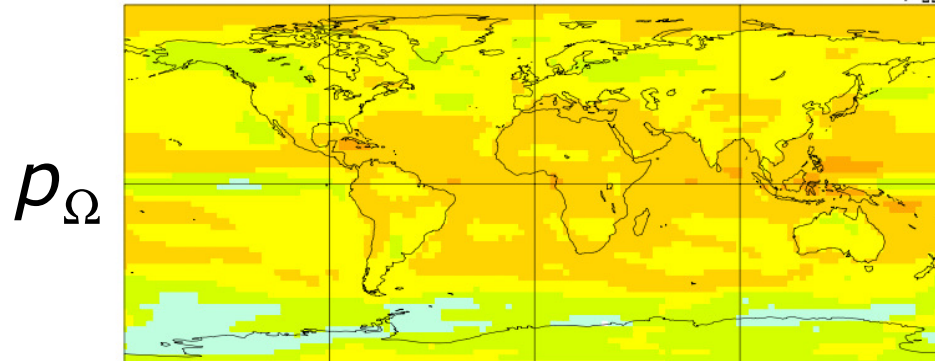
Internally generated p_{ν}



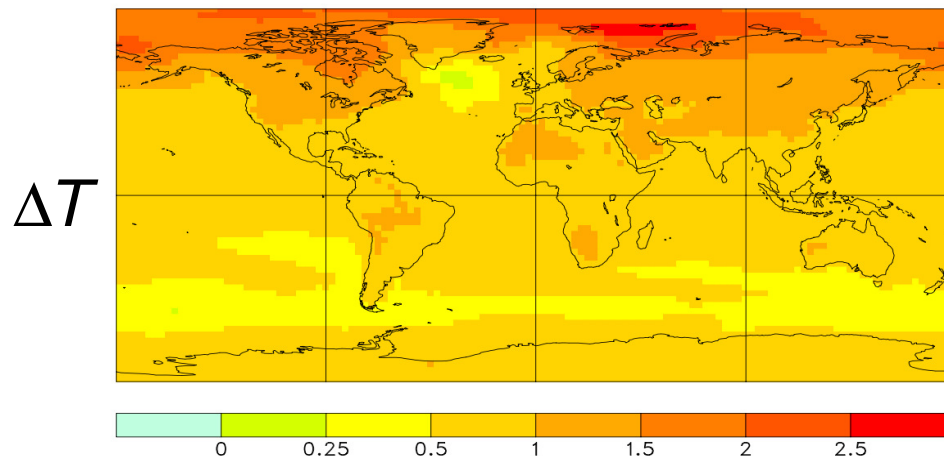
- percentage of total variance over decade
 - associated with forced component
 - associated with internal variability
- p_{Ω} and p_{ν} tend to be inverses of one another so $p = p_{\Omega} + p_{\nu}$ is more uniform than either

Forced component of potential predictability for temperature and forced component of temperature change for 2010-20

Potential predictability variance fractions: 2010–20
Forced p_{Ω}



Forced component of temperature change (C)
from 2000–10 to 2040–50.

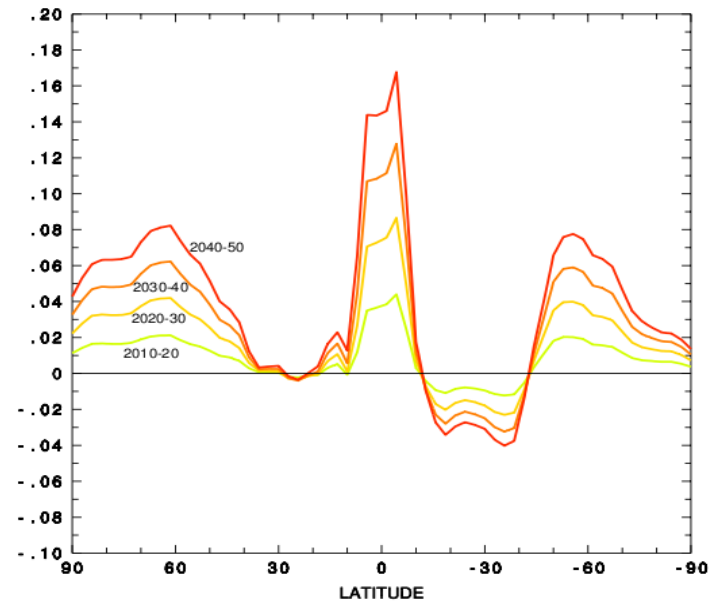


- forced p_{Ω} differs in pattern from forced ΔT
- $p_{\Omega} = \sigma_{\Omega}^2 / (\sigma_{\Omega}^2 + \sigma_v^2 + \sigma_{\epsilon}^2)$
 - for $\sigma_{\Omega}^2 \Rightarrow \Delta T^2$
- noise variance σ_{ϵ}^2 in the denominator discounts ΔT over northern land

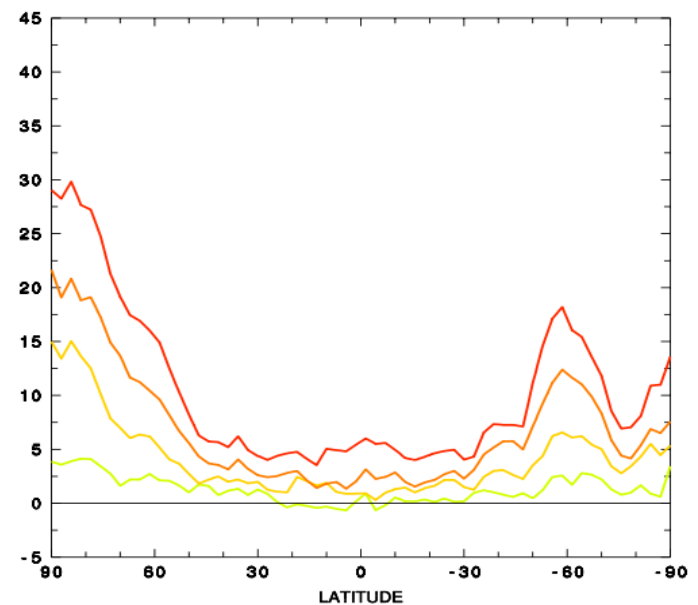
- Precipitation

- *forced component* dominates
- *no* potential predictability for *internally generated* component
- *next-decade* p_{Ω} is small (noise is large)
- *multi-decade* p_{Ω} depends on growing *forced* component

Forced component precipitation
difference from 2000-10

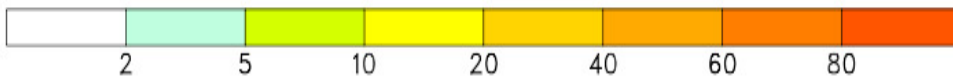
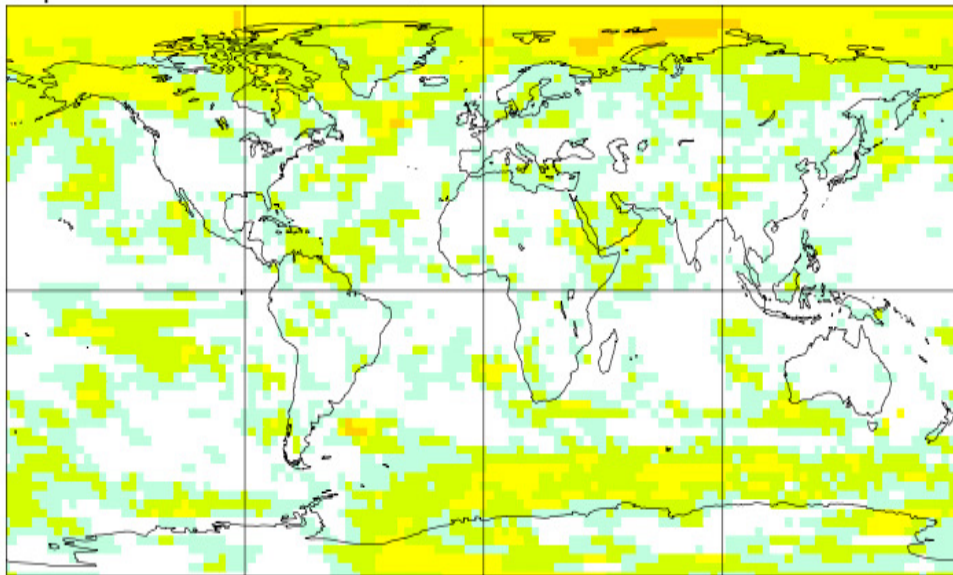


The net *multi-decadal* potential predictability p_l



Potential predictability of precipitation

Potential predictability of precipitation: 2020-30 (i.e. for *second* decade)



- *due to forced component*
- noise variance for precipitation is large
- internally generated p_v is small as a result
- only *multi-decade* $p_{\Omega 1}$ contributes and then only modestly

The challenges and caveats of potential predictability studies

- to identify the mechanisms associated with regions/modes of predictability
- to assess “potential” vs “actual” predictability
- to investigate predictive *skill* of both forced and internally generated variability

end of presentation