A Simple SST Index of Internally-generated Atlantic Multi-decadal Variability (iAMV) that is Robust to Climate Change

Clara Deser NCAR/CGD Climate Analysis Section

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Low-pass filtered North Atlantic [0-60N, 80W-0E] SST anomaly timeseries

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Removal of background climate change:

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- Address using 'perfect model' testbeds (i.e., Large Ensembles).

MPI Grand Ensemble

SST regression maps onto AMV Index (20-year running means)











Ensemble mean of 40 individual member regression maps.













Remove the *pattern* associated with G(t) (Zhang et al., 2019; Qin et al. 2020)



180

120E

60E

6ÓW

12⁰W

Ensemble mean of 40 individual member regression maps.



Is there a better approach?

Remove the *pattern* associated with G(t) (Zhang et al., 2019; Qin et al. 2020) **"Residual method"**

30N 20N 10N 10S 20S 30S 50S 60S 70S 6ÓF 12'0E 12'0W 6ÓW 180 -0.92 Forced Trend (- spatial mean) 70N 60N 50N 40N 30N 20N 10N 10S 20S 30S 40S 50S 60S 70S 6ÓW 120E 60E 180 120W

Ensemble mean of 40 individual member regression maps.















Other Model Large Ensembles (30 members of each)

CMIP5 Multi-Model Large Ensemble Archive Deser et al (2020, *Nature Climate Change*) US CLIVAR WG on LEs









Summary: Defining Internally-generated AMV in a Changing Climate Summary: Defining Internally-generated AMV in a Changing Climate

 Subtracting global-mean SST(t) (i.e., Trenberth & Shea method) works well under current and near-term climate change in most CMIP5 model Large Ensembles, but fails thereafter due to aliasing of the global warming pattern. Summary: Defining Internally-generated AMV in a Changing Climate

- Subtracting global-mean SST(t) (i.e., Trenberth & Shea method) works well under current and near-term climate change in most CMIP5 model Large Ensembles, but fails thereafter due to aliasing of the global warming pattern.
- Removing the global warming pattern via spatial regression onto GSST(t) (i.e., Zhang et al. 2019; residual method) yields robust results (r > 0.8) throughout the 21st century.

Extra



Ensemble mean of 40 individual member regression maps.

1950-2020

Truth: [- Ensemble Mean]



AMV (Residual Method) Pattern Correlations with EM (1950-2020)



r(obs, EM) lies within 5th-95th % spread of model r(i, EM) for all models

(ERSSTV5)

MPI Grand Ensemble: Forced SST Trend



Forced SST* trend 1955-2095





Forced SST* trend 1955-2015

