







Climate-Mode Initialization for Decadal Predictions

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Existing data assimilation and initialization practices



with "native" data assimilation product observations are directly assimilated into a prediction system

DATA ASSIMILATION:

gridded ocean, OI, EnKF, 3D-VAR, 4D-VAR,...

INITIALIZATION: full field

Existing data assimilation and initialization practices



Relation to initialization shock

- Mulholland et al 2015: an effect of using non-native un-coupled reanalyses on initialization shocks (in the form of spurious waves, transports, flux exchange) in short-term predictions:
- Dynamical imbalances between
 - initial state and prediction system
 - initial state (ocean) and initial state (atmosphere)
 - can lead to initial shocks and eventually to loss of prediction skill



Attempt to make initial states coming from "non-native" reanalysis to be compatible with the prediction system



Aim

- ightarrow improve prediction skill
- → eliminate initialization shock between the ocean initial state and the prediction system based on MPI-ESM-LR (recent studies on initial shock for MPI-ESM-LR: Pohlmann et al 2016, Kroeger et al 2017)

Approach

→ reshape observed modes of variability (ORAS4) to fit model modes (MPI-ESM-LR)

In numerical weather predictions:

- Normal modes: obtaining correctly balanced initial state by filtering initial conditions to deal with interior-gravity waves: e.g., Williamson (1976), Ballish (1981) and many, many others
- Temporal filtering: using EOFs to filter out the components that are difficult to predict: Branstator et al. (1992)
- Error growth rates: *e.g.*, *Boer* (1984)

- In seasonal predictions: coupled modes initialization: *Kirtman et al (COLA, 2005-2008)*
- In decadal predictions: avoid introducing anomalies that are out of the model internal variability range: Volpi et al 2016, phase initialization: SPECS (Caian et al)

Climate-mode initialization recipe based on MPI-ESM



EOF-modes: weighting choice



EOF-modes: truncation level choice



What do we initialize with?

Standard deviation of initial temperature state @ different depth levels from the November 1st snapshots



Surface temperature correlation skill and skill score: MODE-INIT versus ANOM-INIT









-1-.9-.8-.7-.6-.5-.4-.3-.2-.10 .1 .2 .3 .4 .5 .6 .7 .8 .9 1

Skill Score, ly1

LEAD YEAR 1

Further investigations of the Pacific Ocean:

- ENSO skill
- zonal momentum balance



LEAD YEARS 2-5

Further investigations of the Atlantic Ocean:

- climate indices (NASPG, AMOC@25N) 🗡
- Gulf Stream path
- trend?

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Zonal momentum balance of the upper equatorial Pacific

between pressure gradient force and zonal wind stress in the first month after initialization



Summary for MODE-INIT: an attempt to make initial states coming from "non-native" reanalysis to be compatible with the prediction system

Aim improve prediction skill

by eliminating initialization shock

Performance as compared to anomaly initialization improves temperature skill in the tropical Pacific for short time scales improves zonal momentum balance in the Eq.Pacific

has "mixed success" in the North Atlantic

Approach

reshape observed modes of variability (ORAS4) to fit model modes (MPI-ESM-LR)

Opportunities to develop further:

- regional EOF-modes instead of global ones
- larger EOF space
- different truncation (EOFs with long time scales)

Thank you very much for your attention!

Surface temperature skill for lead years 2-5

Comparison of MiKlip methods (including MODE-INIT) with anomaly initialization

ANOM-INIT correlation skill & & Correlation difference to ANOM-INIT ANOM-INIT MSESS w.r.t. climatology & MSESS w.r.t. ANOM-INIT



Polkova et al 2018 (submitted to JAMES)