

Skill assessment of the CSIRO multi-year Climate Analysis Forecast Ensemble (CAFE) system

CSIRO decadal climate forecasting project

Dougie Squire, James Risbey, Carly Tozer, Thomas Moore, James Munroe

The CSIRO CLIMATE ANALYSIS FORECAST ENSEMBLE system

- New project to understand and improve predictability on multi-year time scales
- Use a variant of the GFDL CM2.1 ocean (MOM5) - atmosphere (AM2) - land (LM2) - sea ice (SIS) model
- Focus on internal variability

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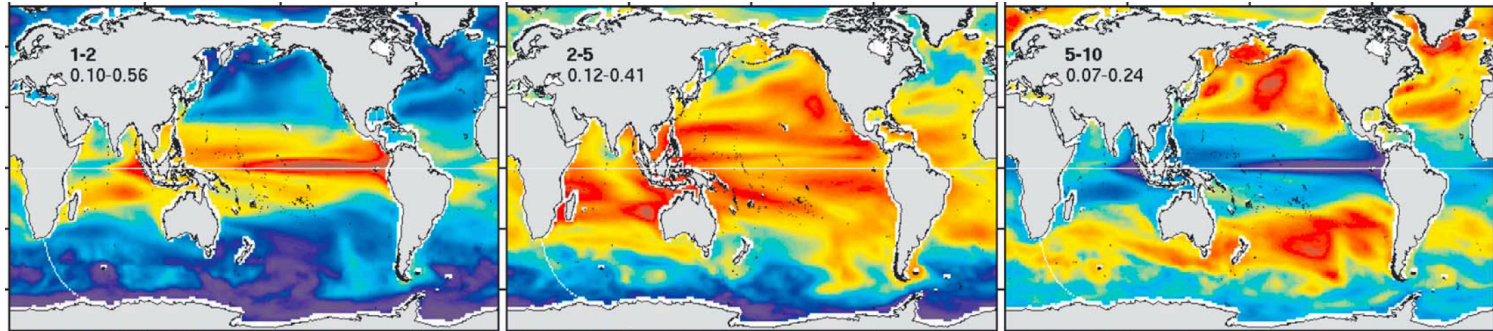
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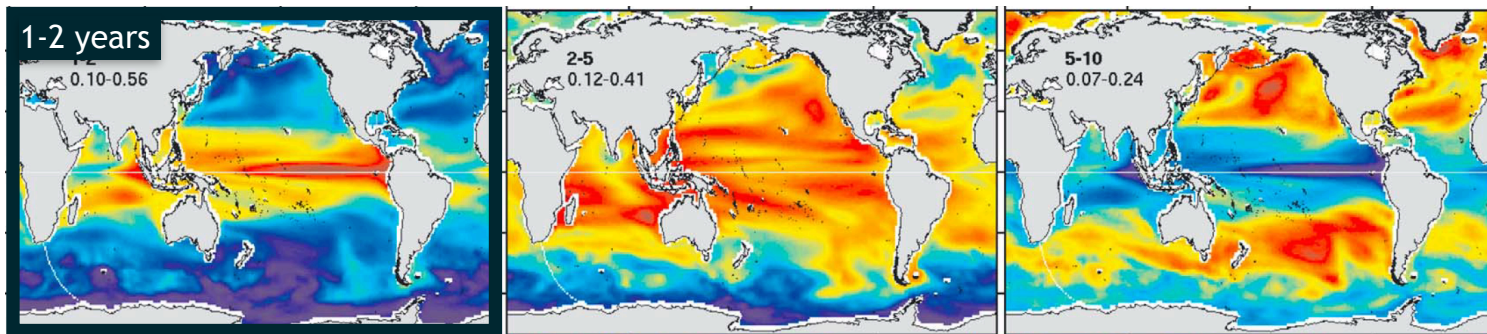


Fractional in-band variances of SLA

Monselesan et al. 2015 GRL

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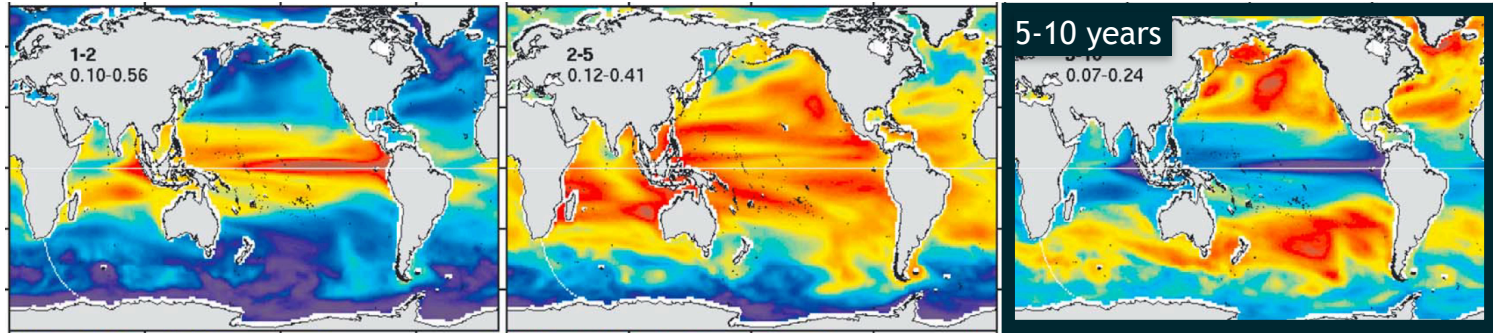


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doppyo diagnostics/verification software

- Leverage emerging efforts towards best practices in big data and reproducibility
 + James Munroe
- Towards a community effort
- Dataset/filetype agnostic

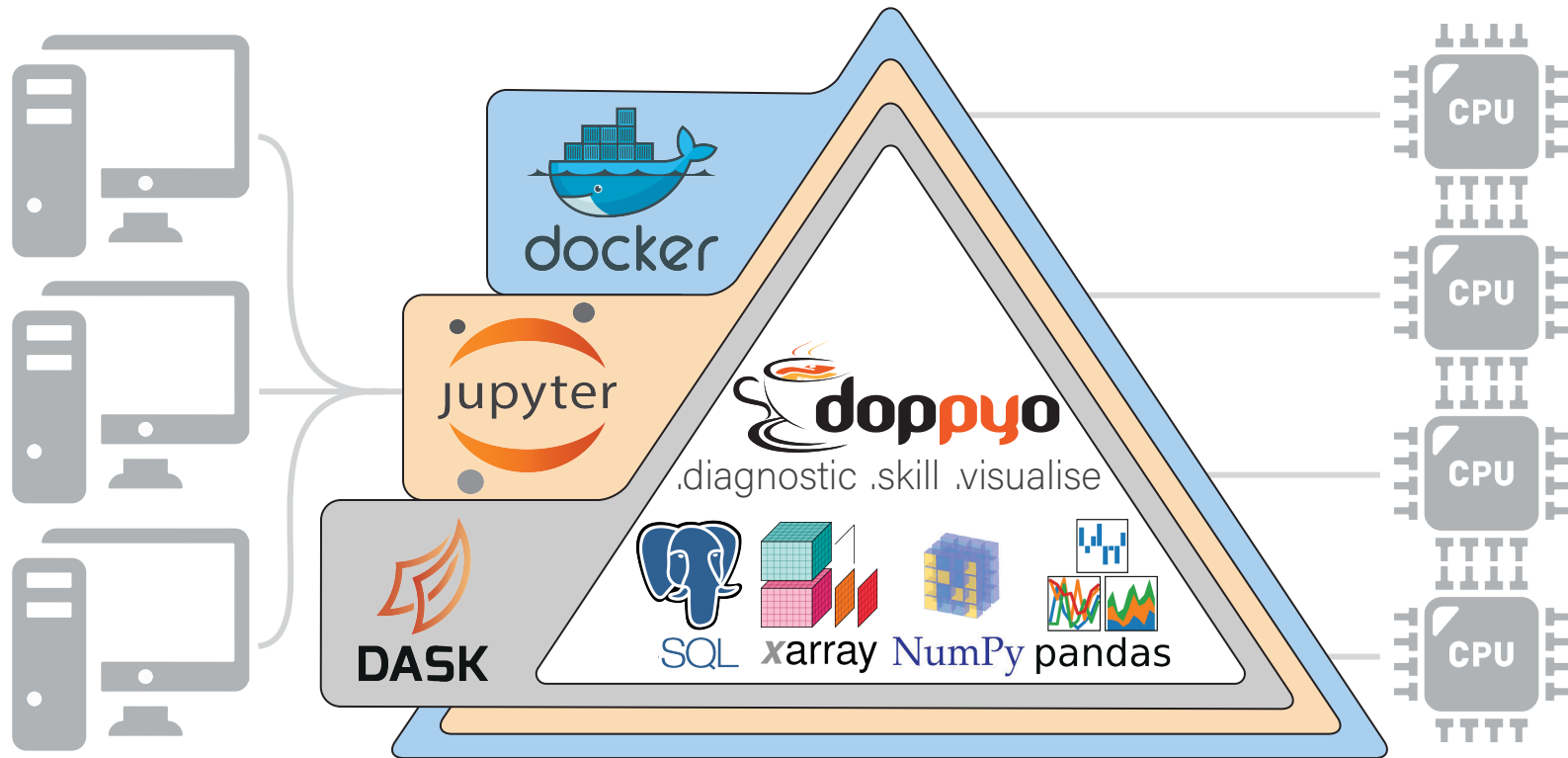
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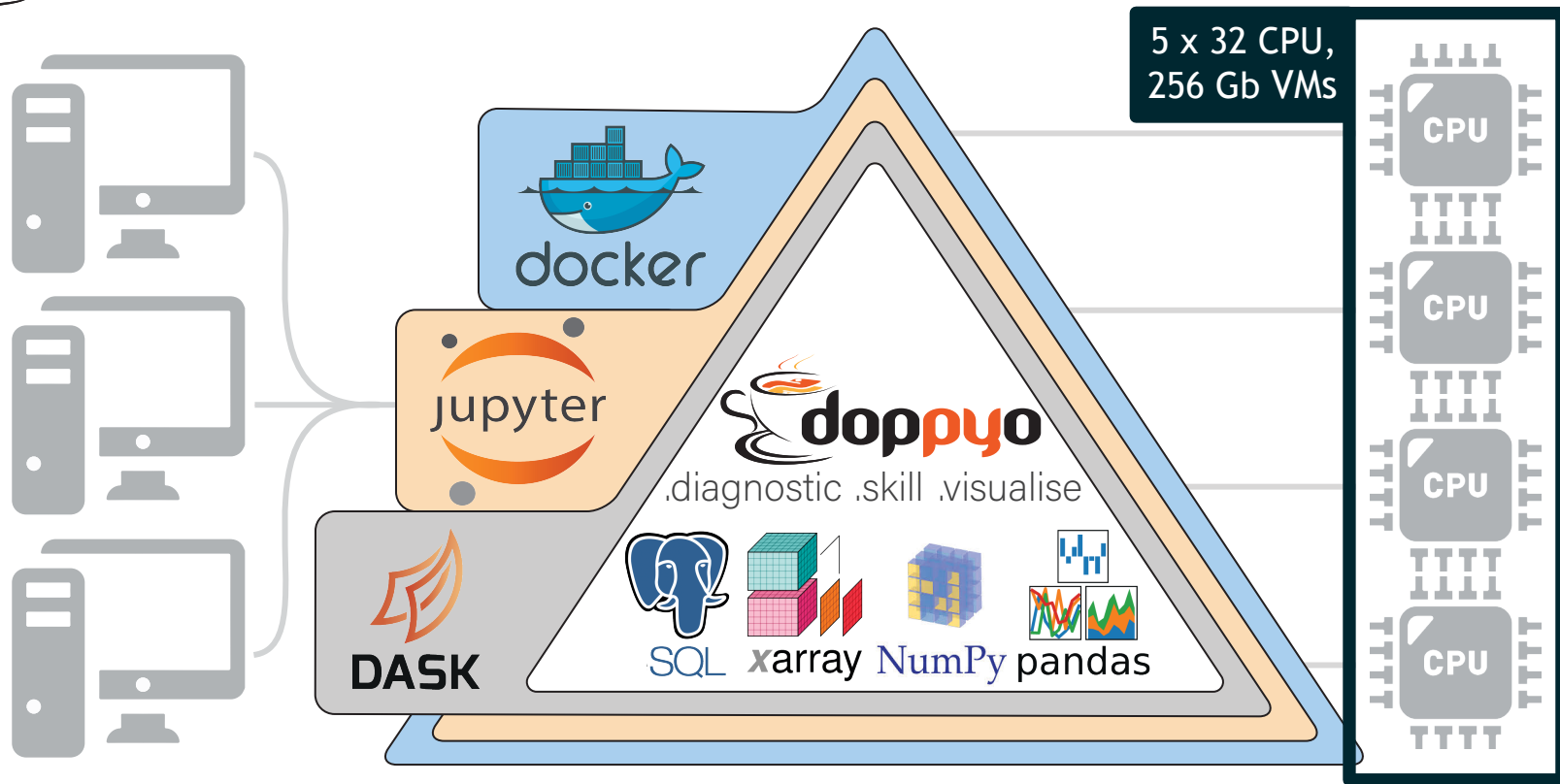
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 **PANGEo** + James Munroe
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doppyo diagnostics/verification software



doppyo diagnostics/verification software



The image shows a JupyterLab interface with a notebook on the left and a sidebar on the right. The notebook contains the following sections and code:

Import required packages

```
In [ ]: import dopypy
import dask
import distributed
import xarray as xr
import numpy as np
```

Initialise dask client

```
In [ ]: client = distributed.Client('tcp://oa-32-cdc.nexus.csiro.au:8786')
client
```

Query database for JRA55 files and load them lazily

```
In [ ]: %%time
gh, temp, u, v, omega = dopypy.utils.load_ncfiles(dataset='jra55', variables=['gh', 'temp', 'u', 'v', 'w'])
```

Lazily compute the total eddy available potential energy

```
In [ ]: %%time
energetics = dopypy.diagnostic.compute_atmos_energy_cycle(temp, u, v, omega, gh, terms='Pe')
```

Compute seasonal means of total eddy available potential energy

```
In [ ]: %%time
```

The sidebar on the right shows a "Favorites" section with a single entry: "Dask: Status" with a circular icon containing the letter "D".

CAFE-f1 hindcasts (today's data)

- 2-year, 11-member hindcasts started monthly over 2002-2016
- Only ocean observations assimilated
- Bred-vector-initialised on sub-surface ocean temperature isosurface corresponding to high in-band variance on 1-2 month time scales
- Mean bias corrected (Stockdale 1997 MWR)

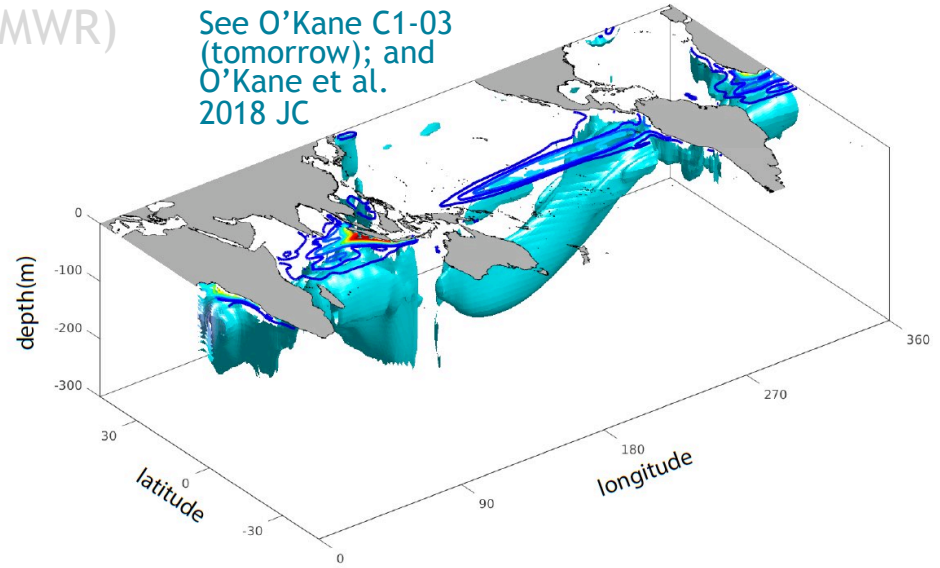
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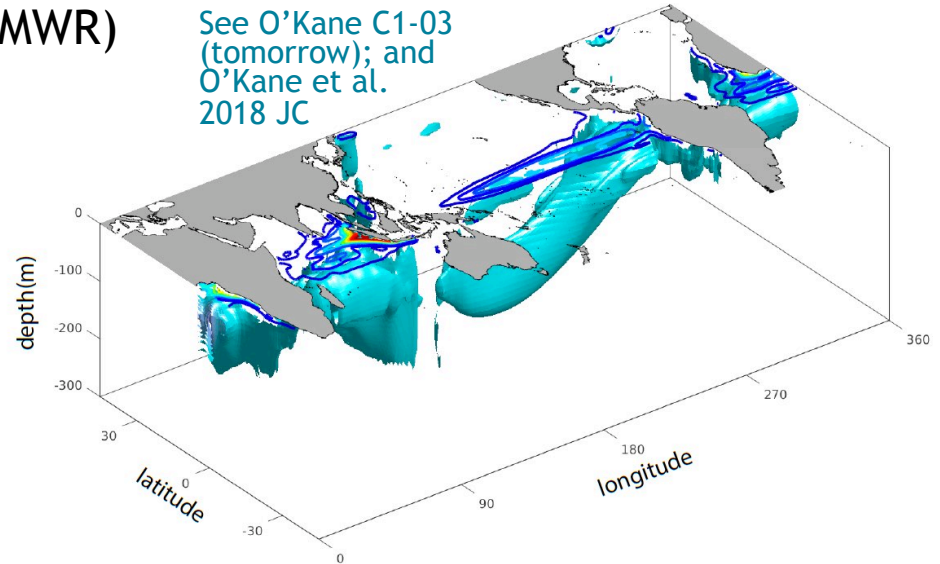
See O'Kane C1-03 (tomorrow); and O'Kane et al. 2018 JC



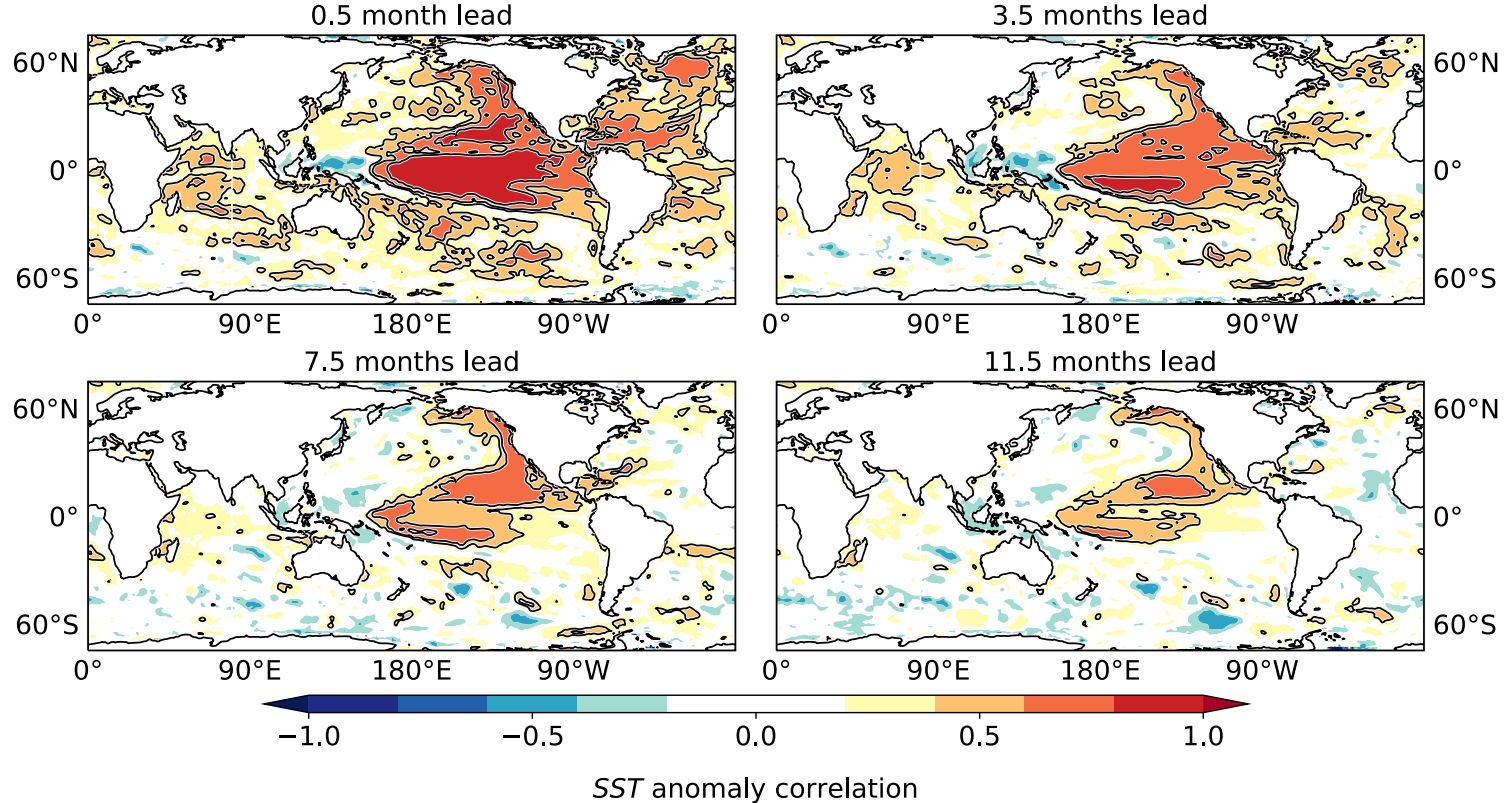
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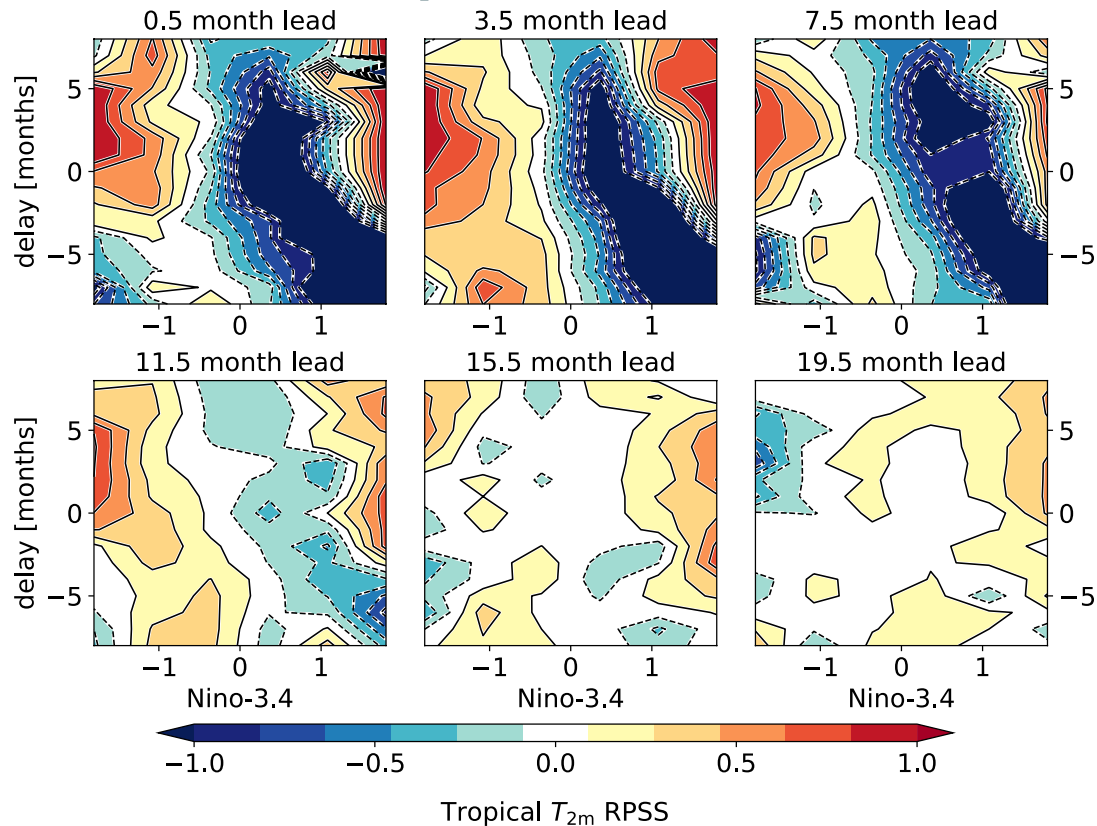
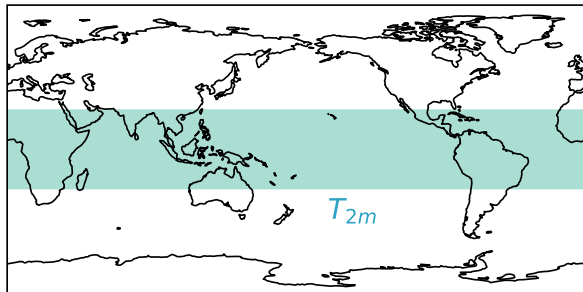
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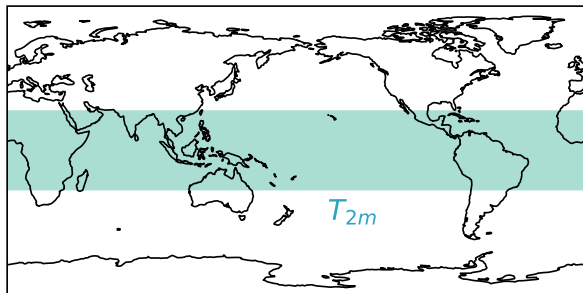
Temporal anomaly correlations of monthly SST



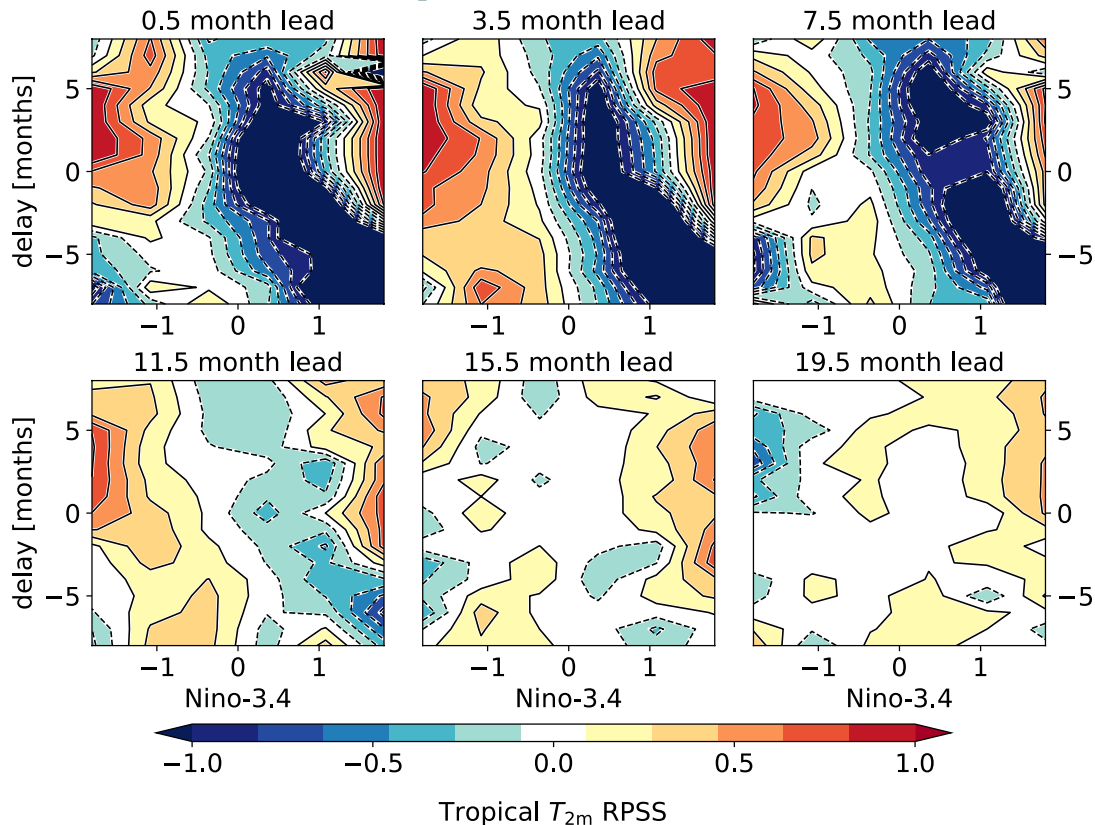
Ranked probability skill score of tropical T_{2m}



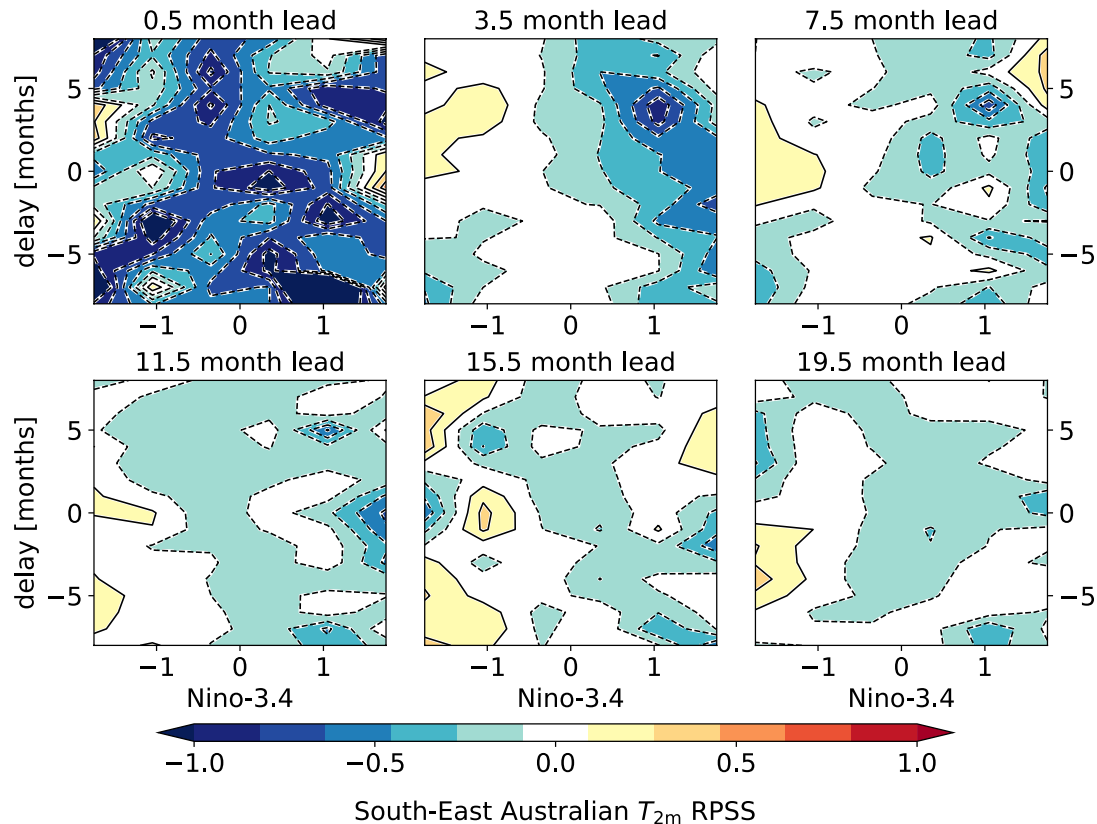
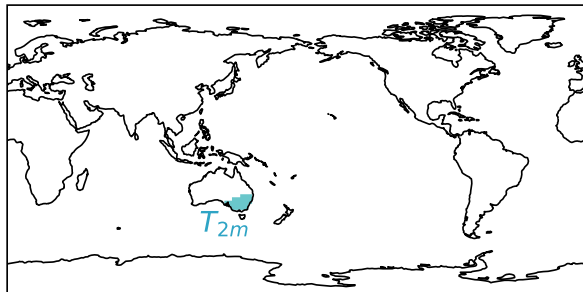
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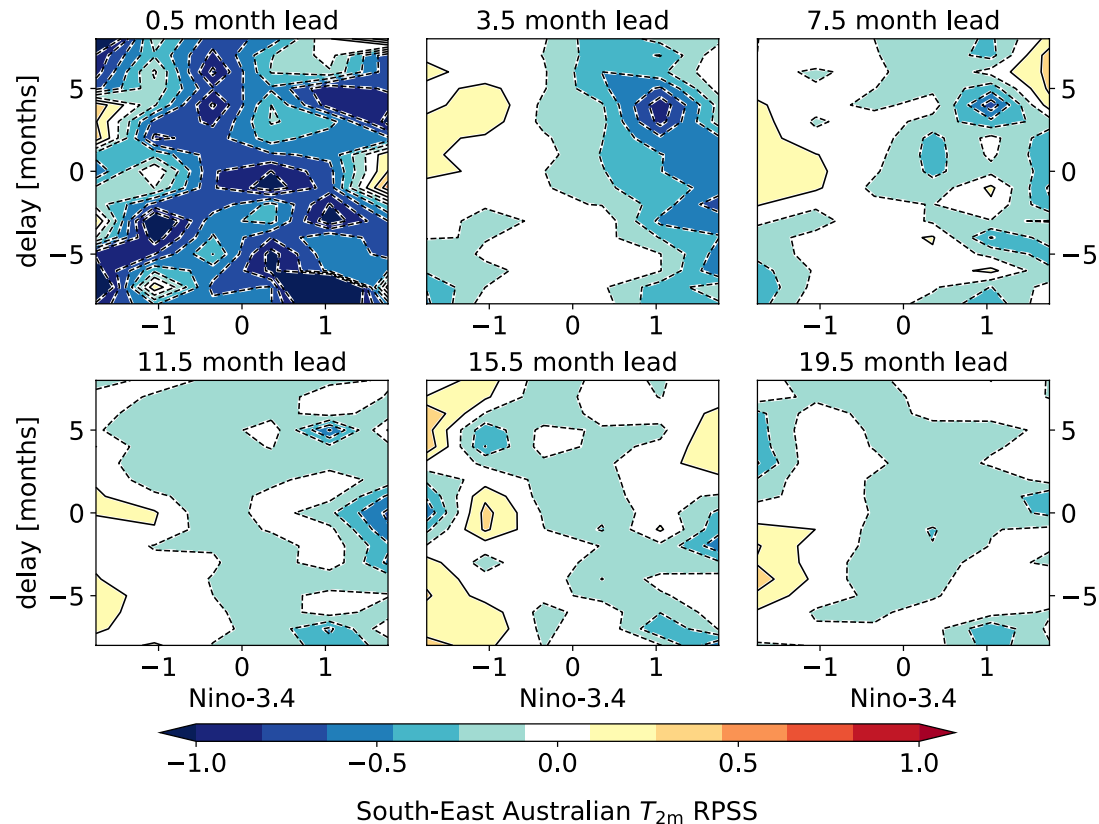
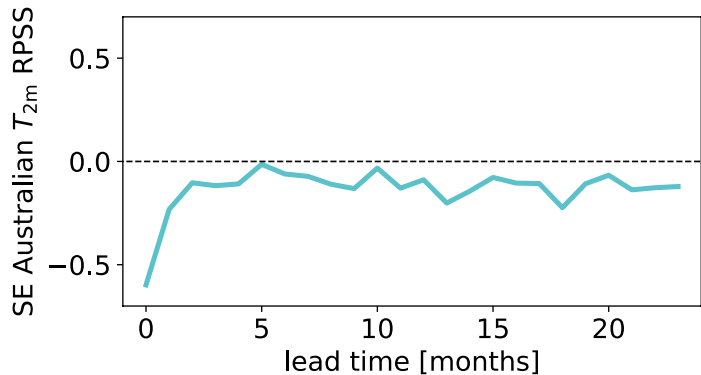
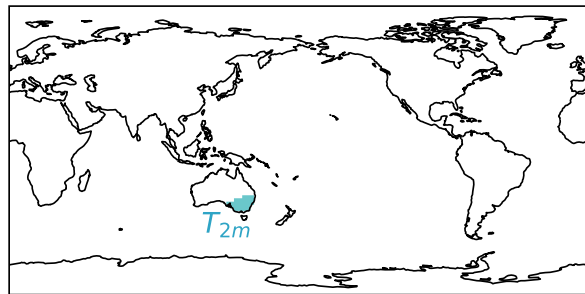
➤ Forecast skill is strongly related to ENSO variability



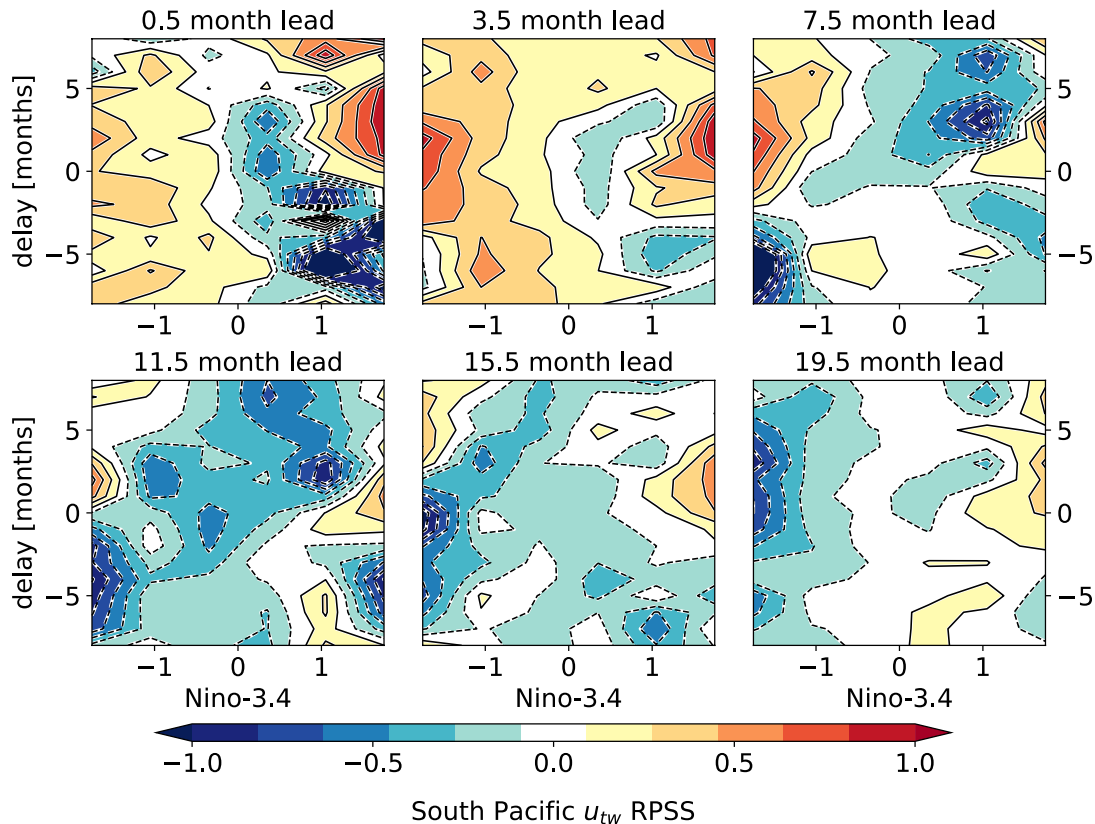
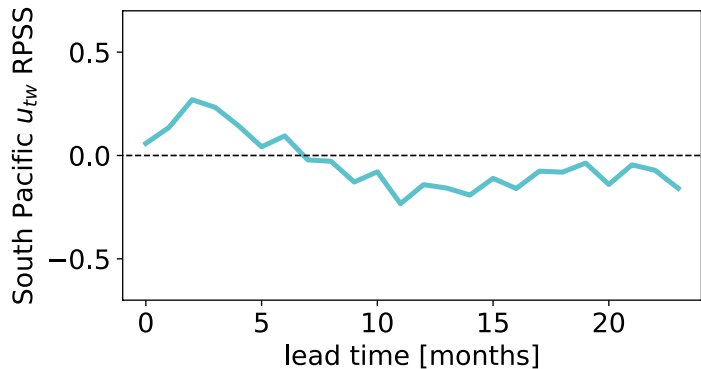
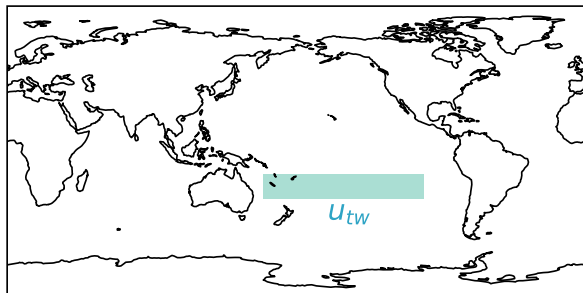
Ranked probability skill score of SE Australian T_{2m}



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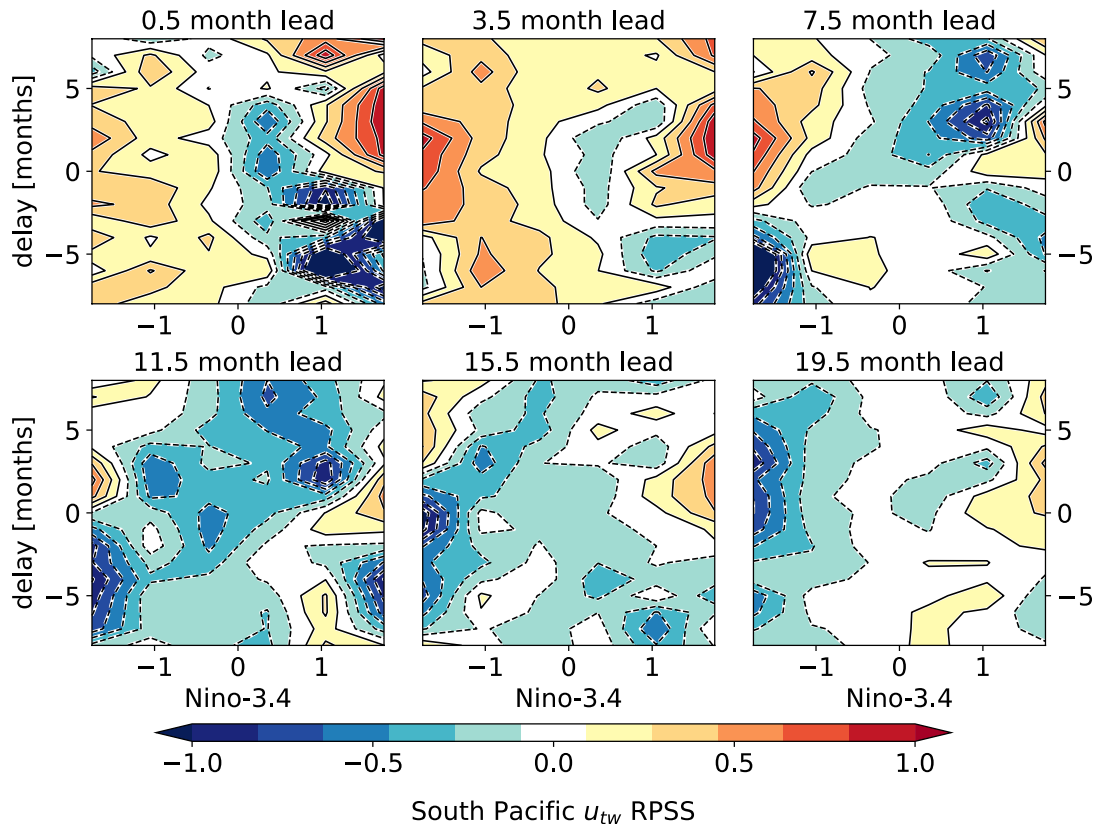
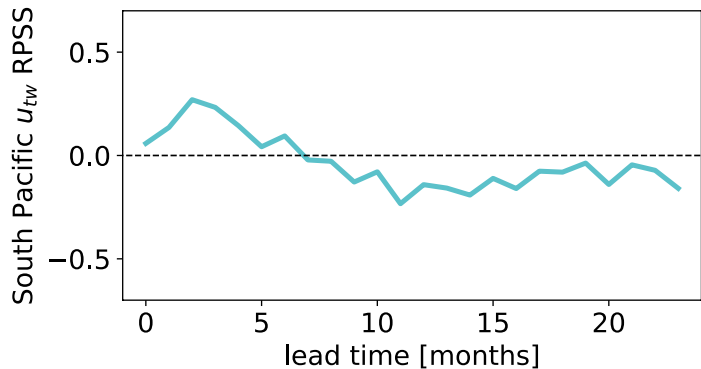
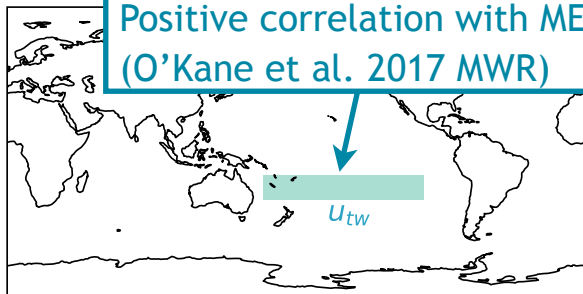


Ranked probability skill score of thermal wind




Ranked probability skill score of thermal wind

Positive correlation with MEI
(O’Kane et al. 2017 MWR)



Conclusions

-  **doppyo** is a diagnostics/verification software package that we are building
- Early CAFE system hindcasts indicate comparable skill to other systems
- Prediction skill in Australia is strongly tied to the tropical ocean (ENSO) and to the CAFE system's ability to simulate relevant teleconnection processes

Contact: Dougie.Squire@csiro.au

Free-running model diagnostics

Longitudinal wave activity flux at 500hPa

H500 EOFs

DJF, CAFE

DJF, JRA55

EOF 1 (SAM), CAFE

EOF 1 (SAM), JRA55

JJA, CAFE

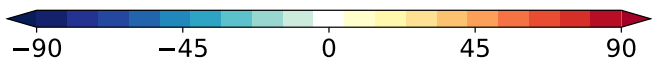
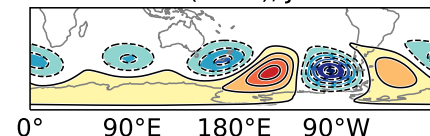
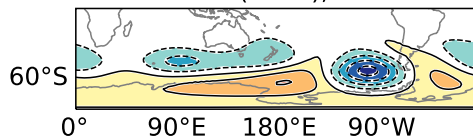
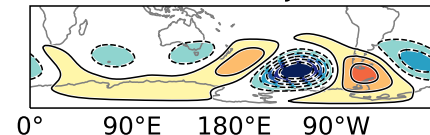
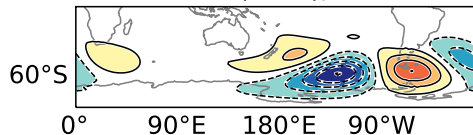
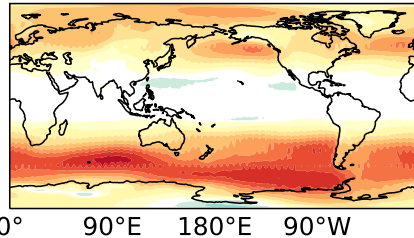
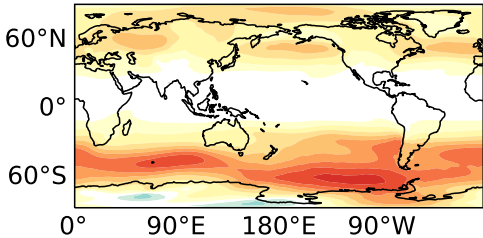
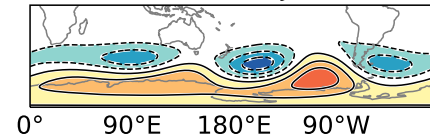
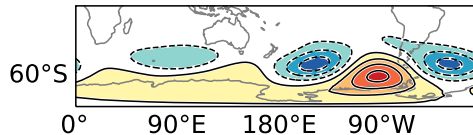
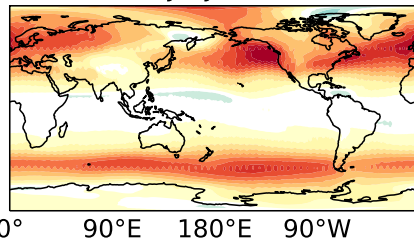
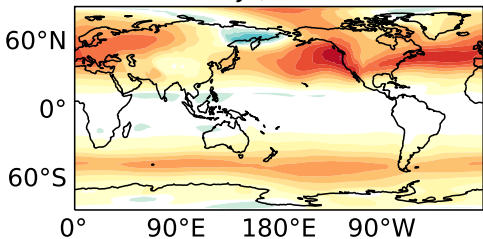
JJA, JRA55

EOF 2 (PSA1), CAFE

EOF 2 (PSA2), JRA55

EOF 3 (PSA2), CAFE

EOF 3 (PSA2), JRA55

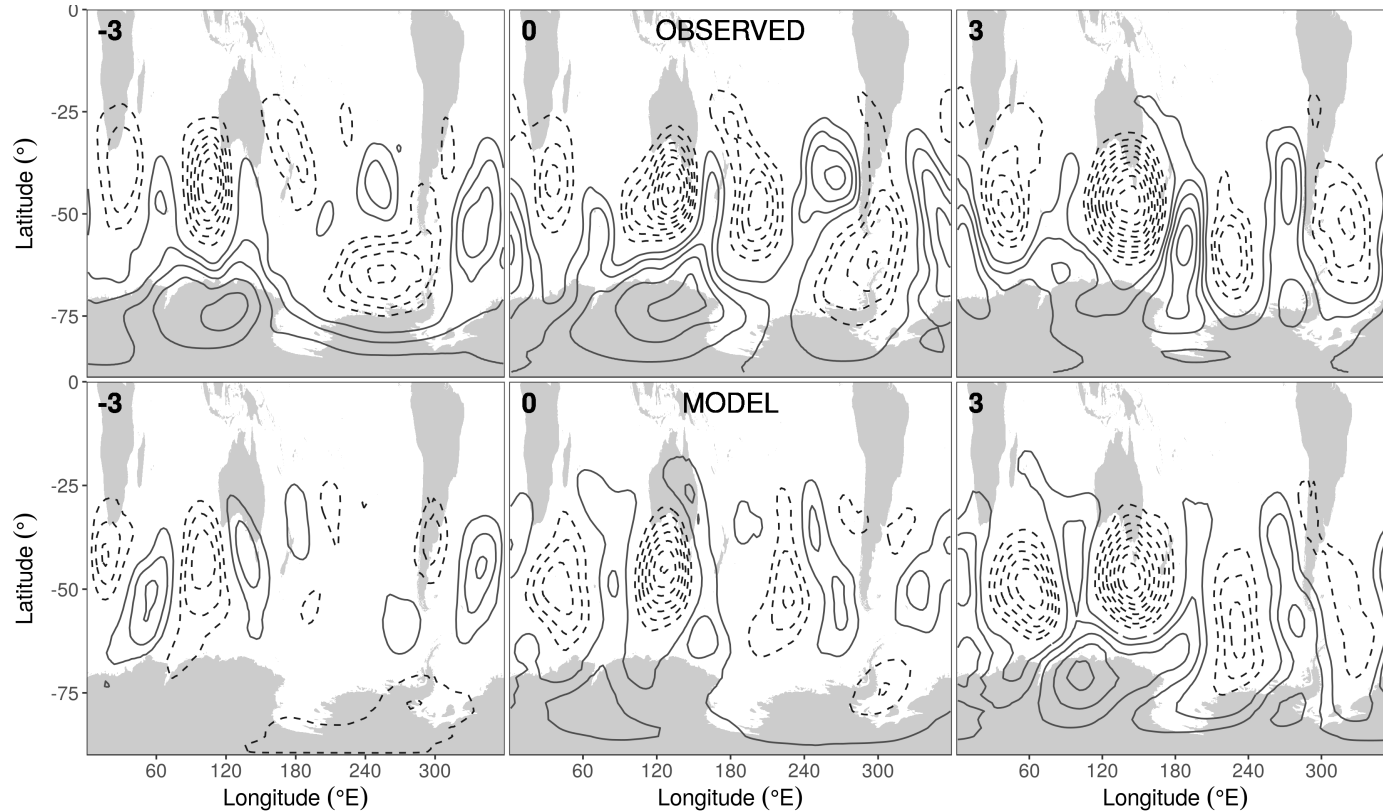


Longitudinal wave activity flux [m^2 / s^2]



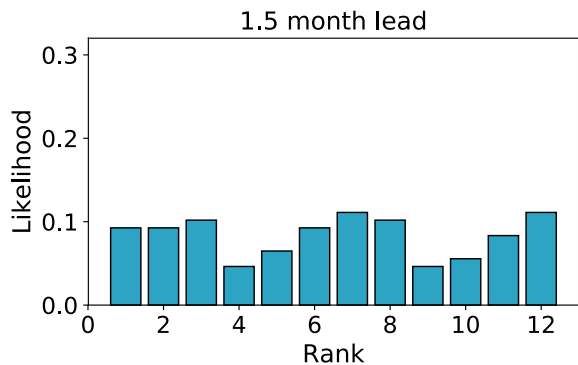
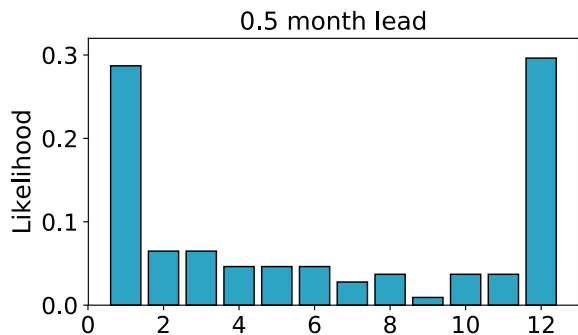
EOFs of 500 hPa height anomalies

h500 anomaly composites for heavy Tasmanian rainfall

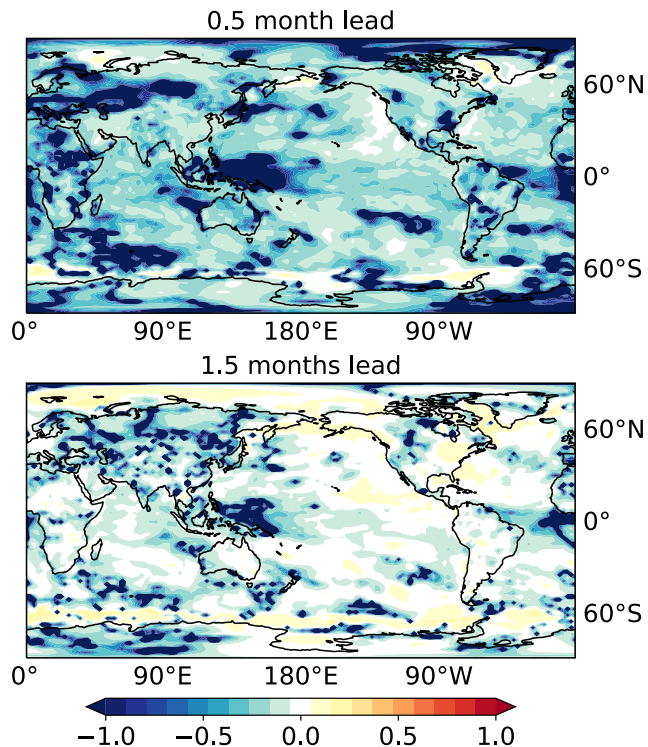


Ensemble spread metrics

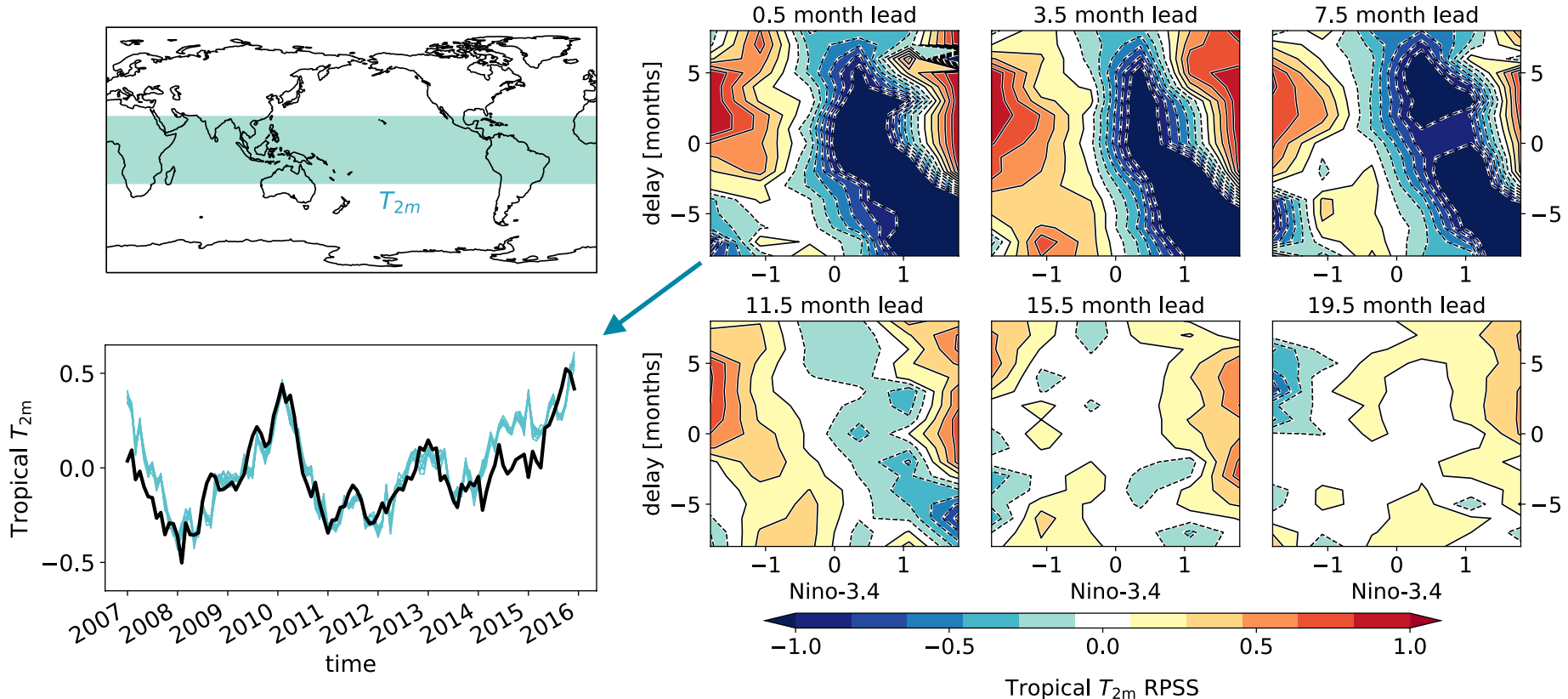
Talagrand of SE Australian T_{2m}



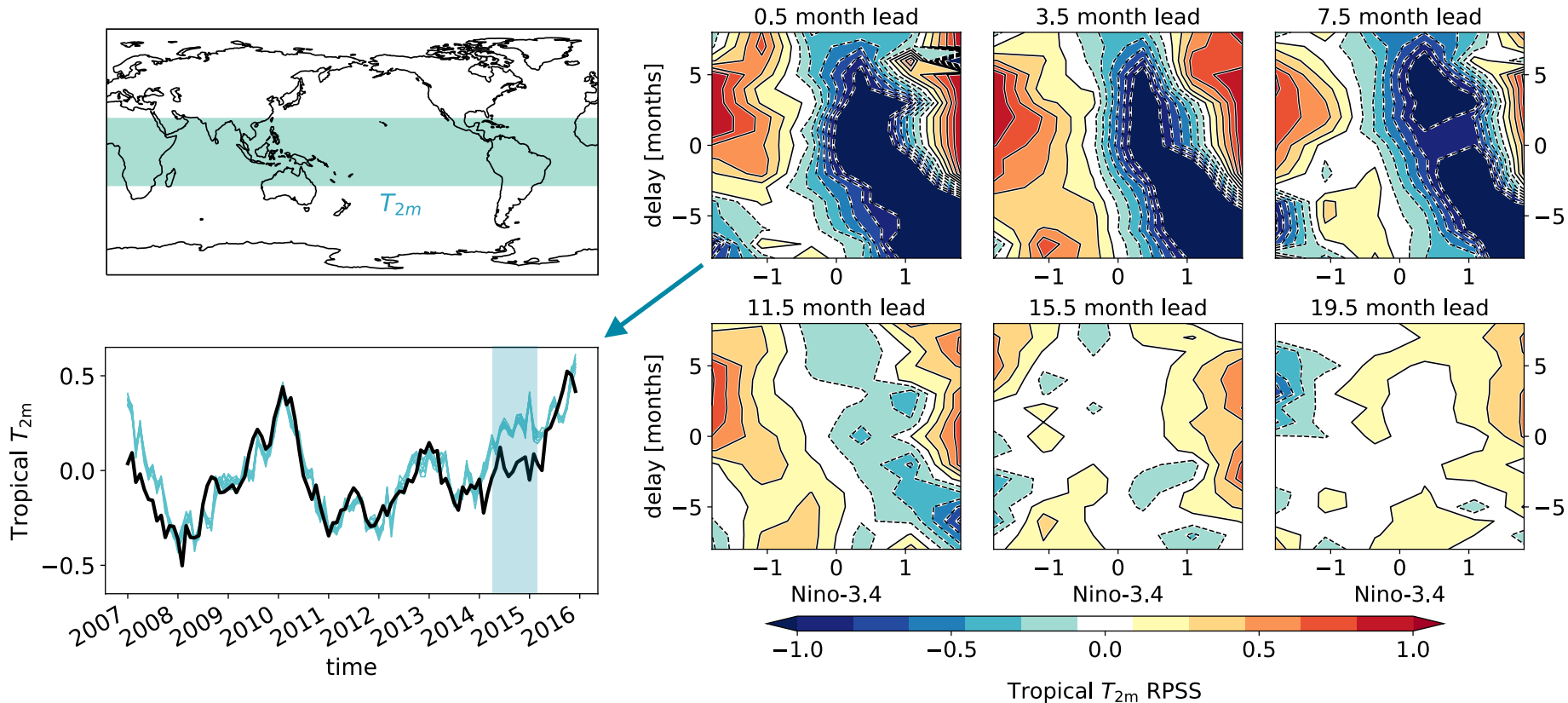
Goddard et al. ensemble spread metric, T_{2m}



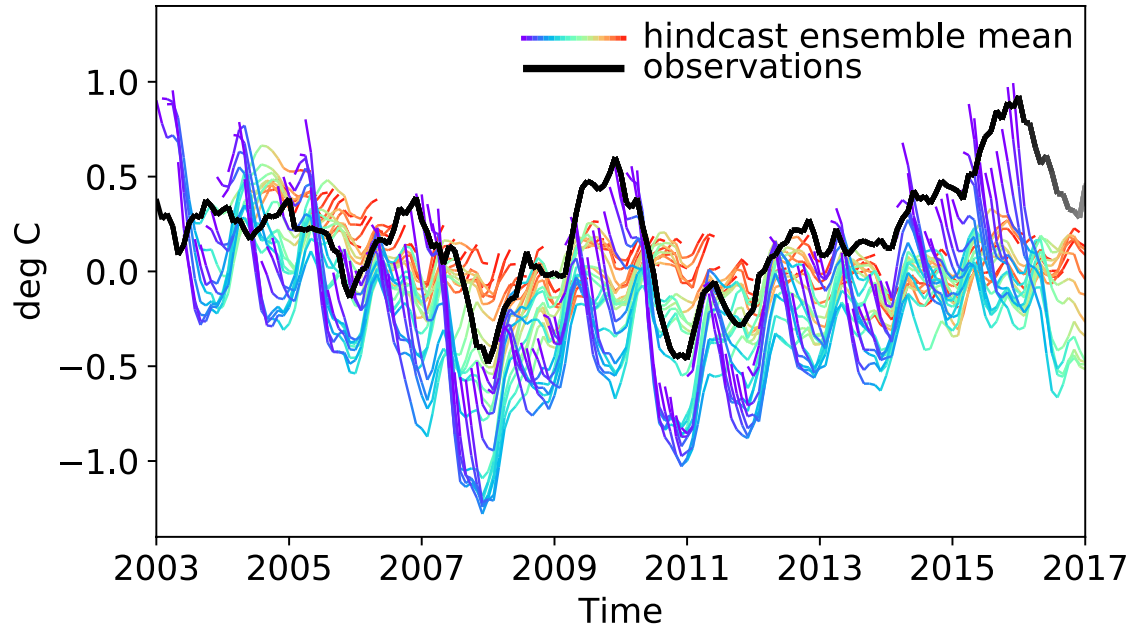
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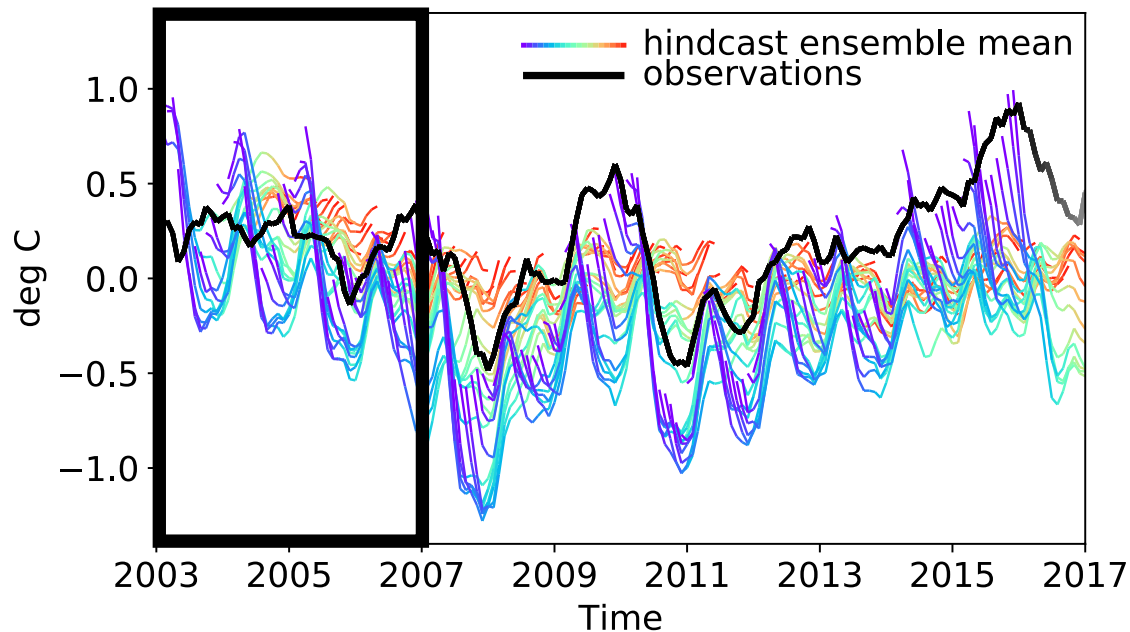
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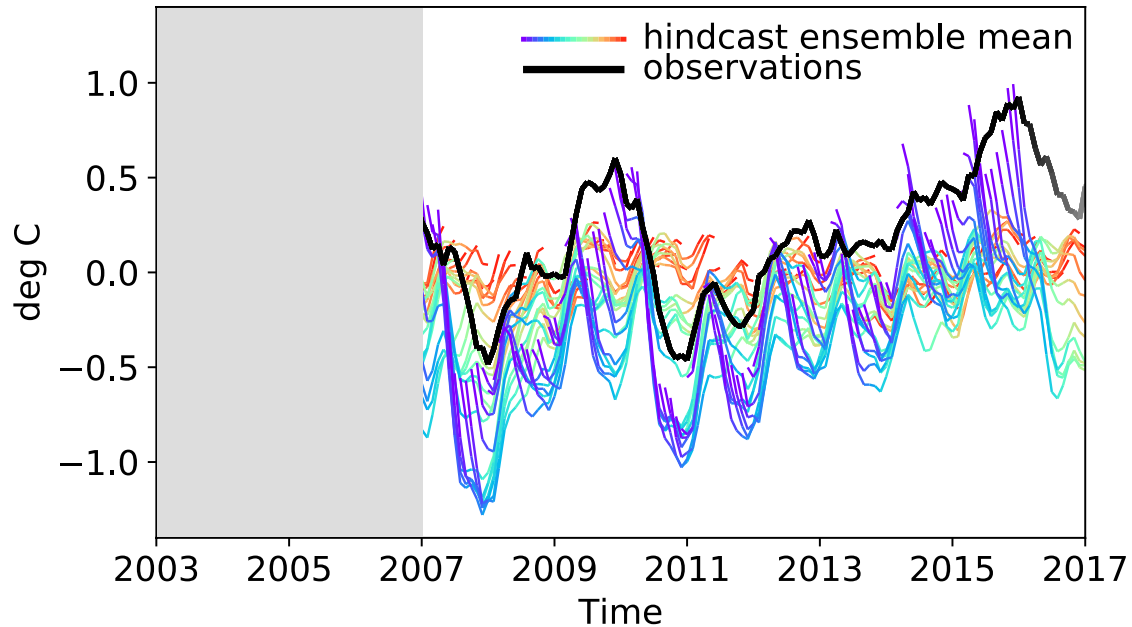
Uncorrected tropical Pacific SST



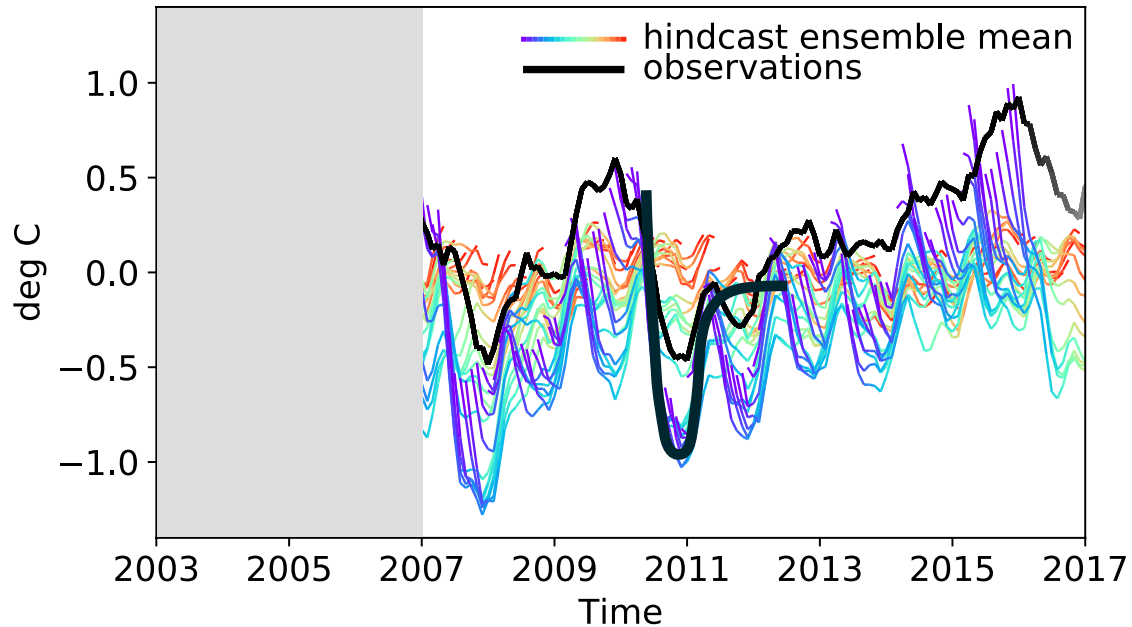
Uncorrected tropical Pacific SST



Uncorrected tropical Pacific SST



Uncorrected tropical Pacific SST



Uncorrected tropical Pacific SST

