Reconstructing the Southern Annular Mode over the Common Era

by assimilating drought atlases and a global proxy network.

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Context

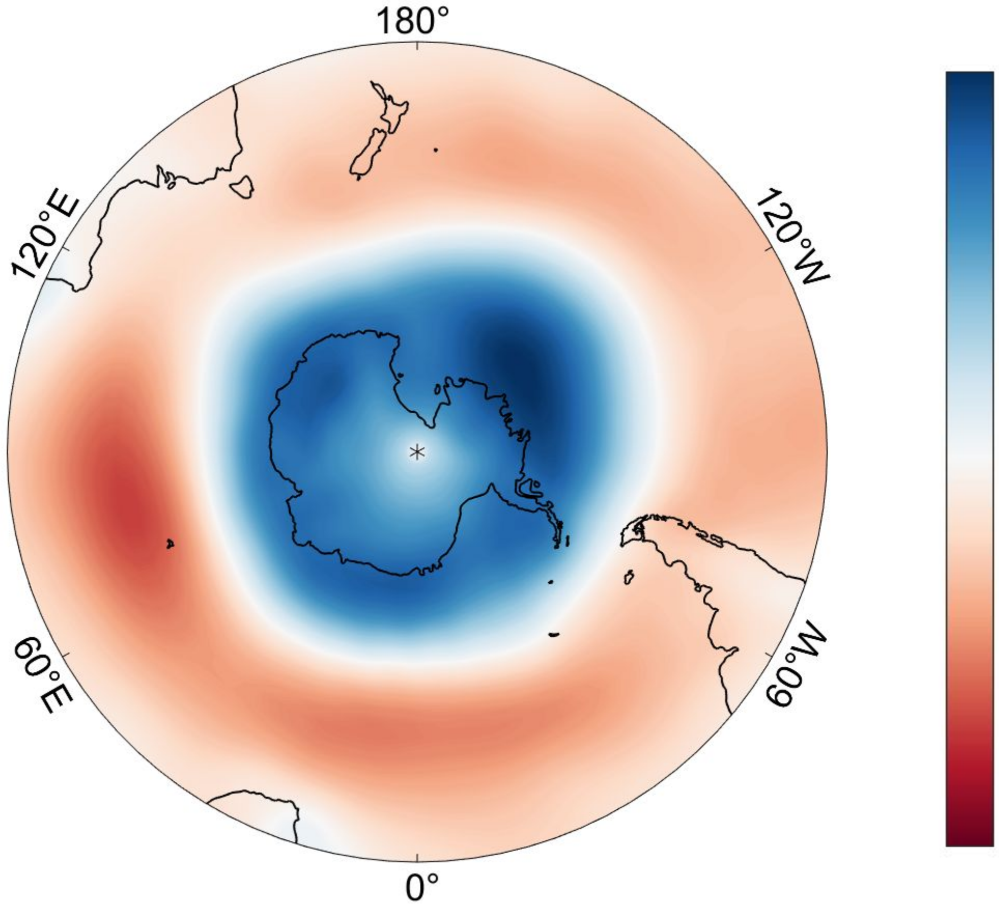
Data Assimilation Method

Results

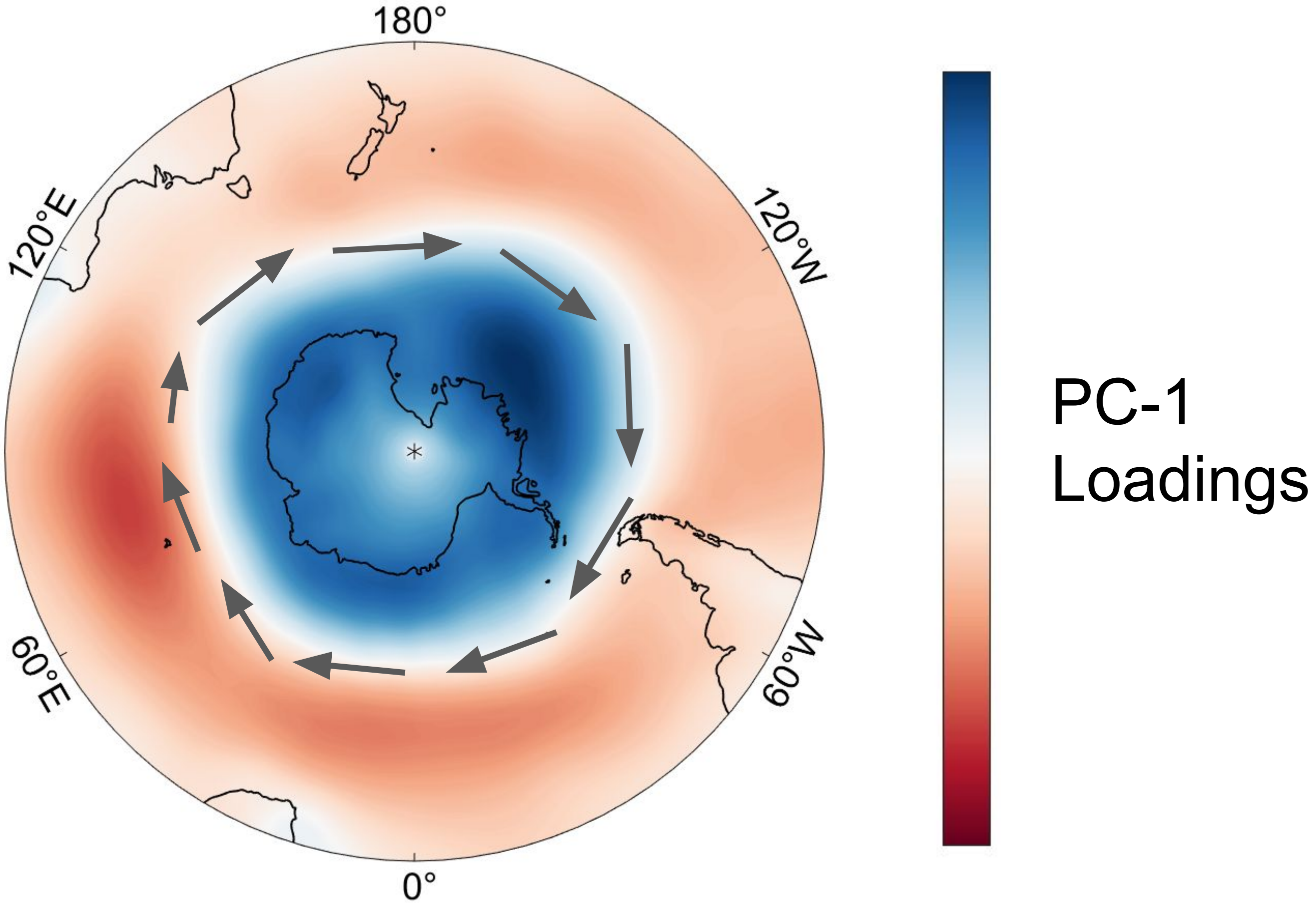
Southern Annular Mode (SAM)

Leading mode of atmospheric variability in the Southern Hemisphere

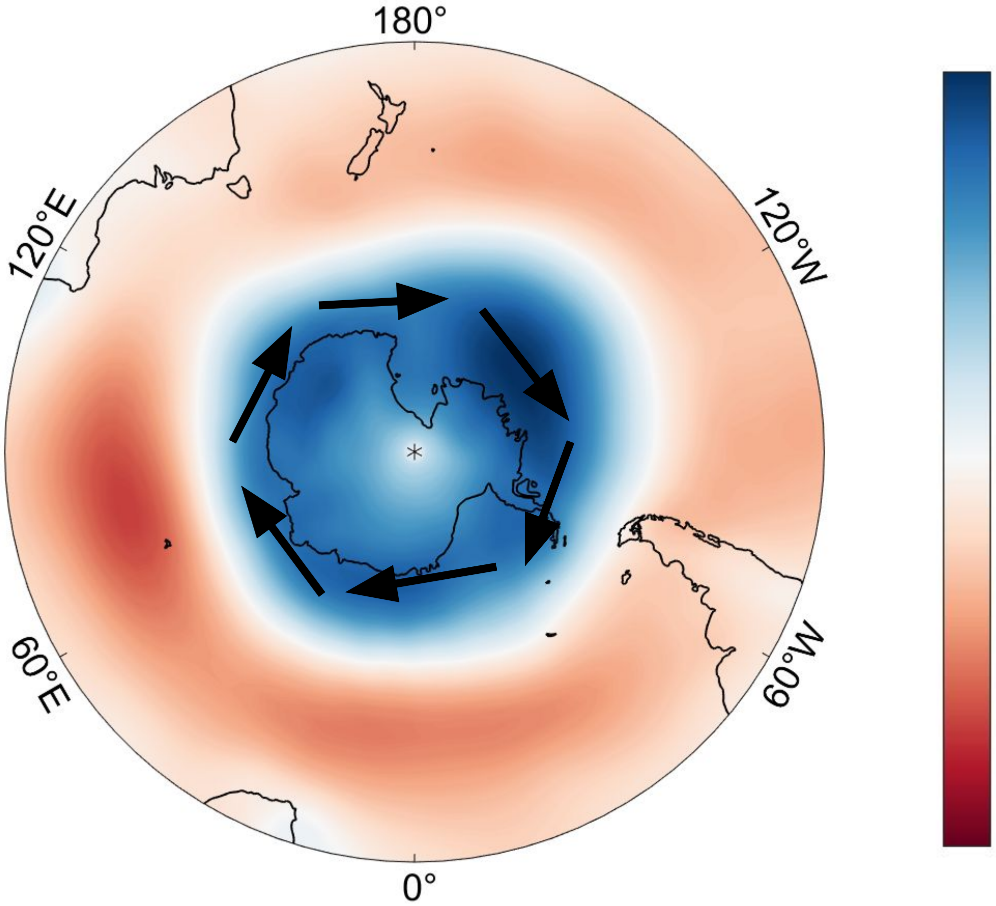
Southern Annular Mode (SAM)



Westerly Winds

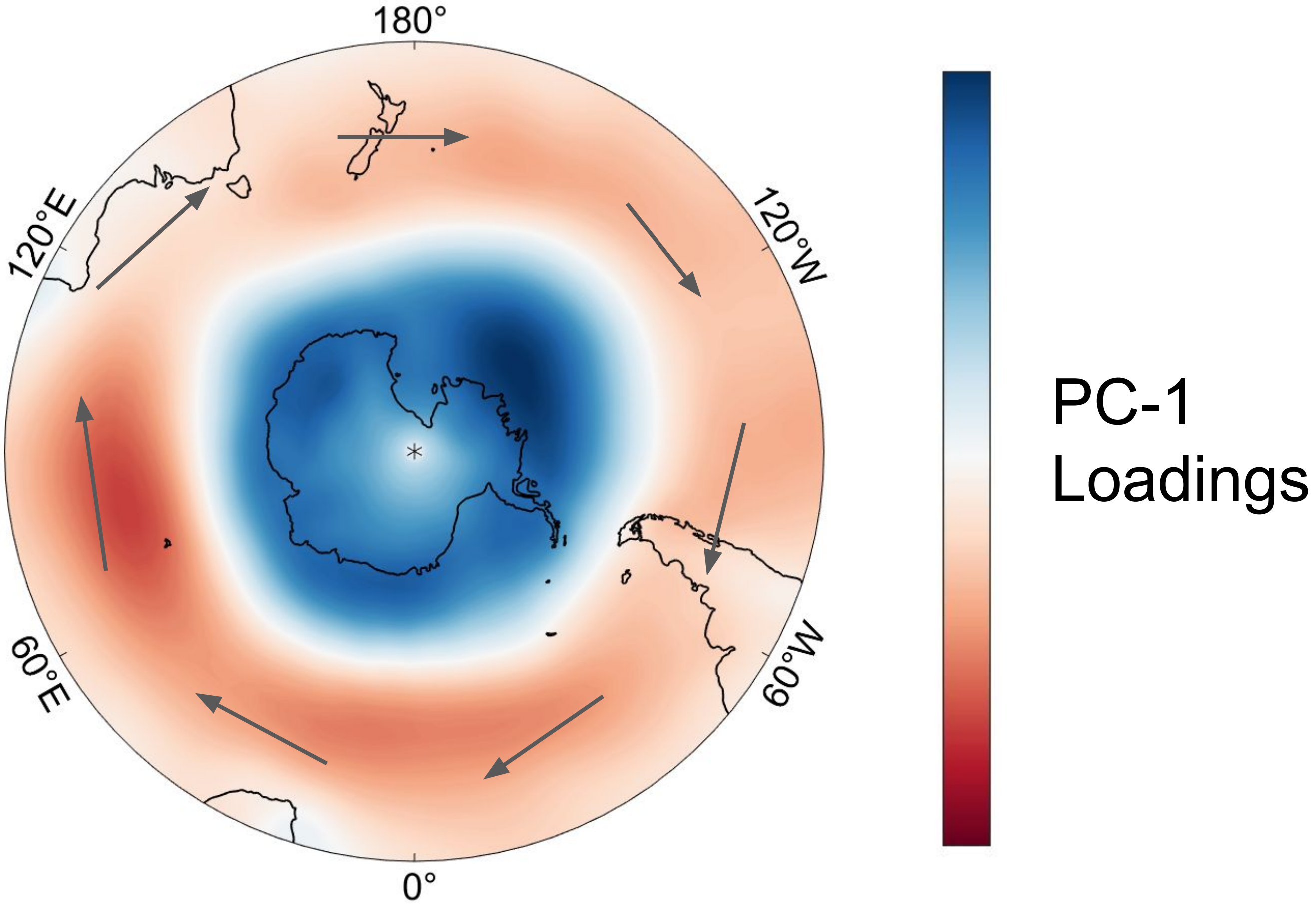


Positive Phase

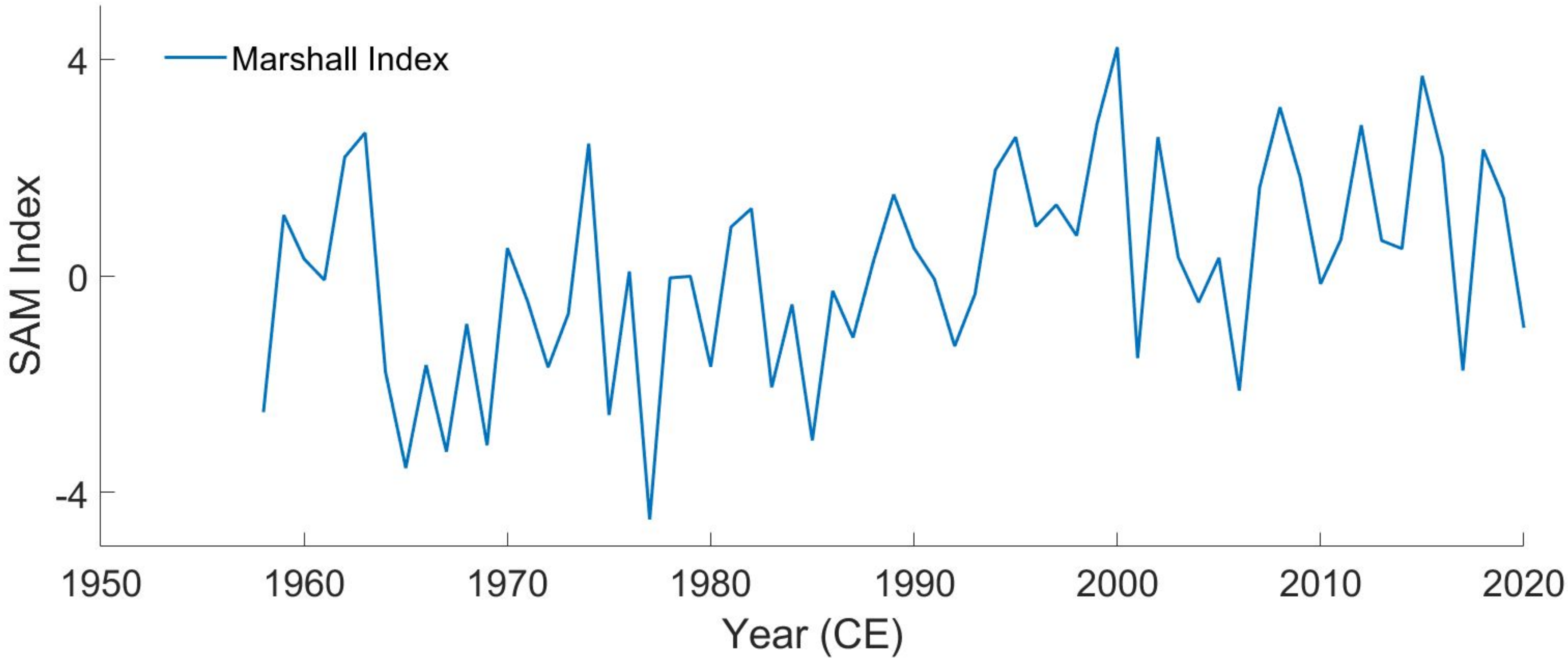


PC-1
Loadings

Negative Phase

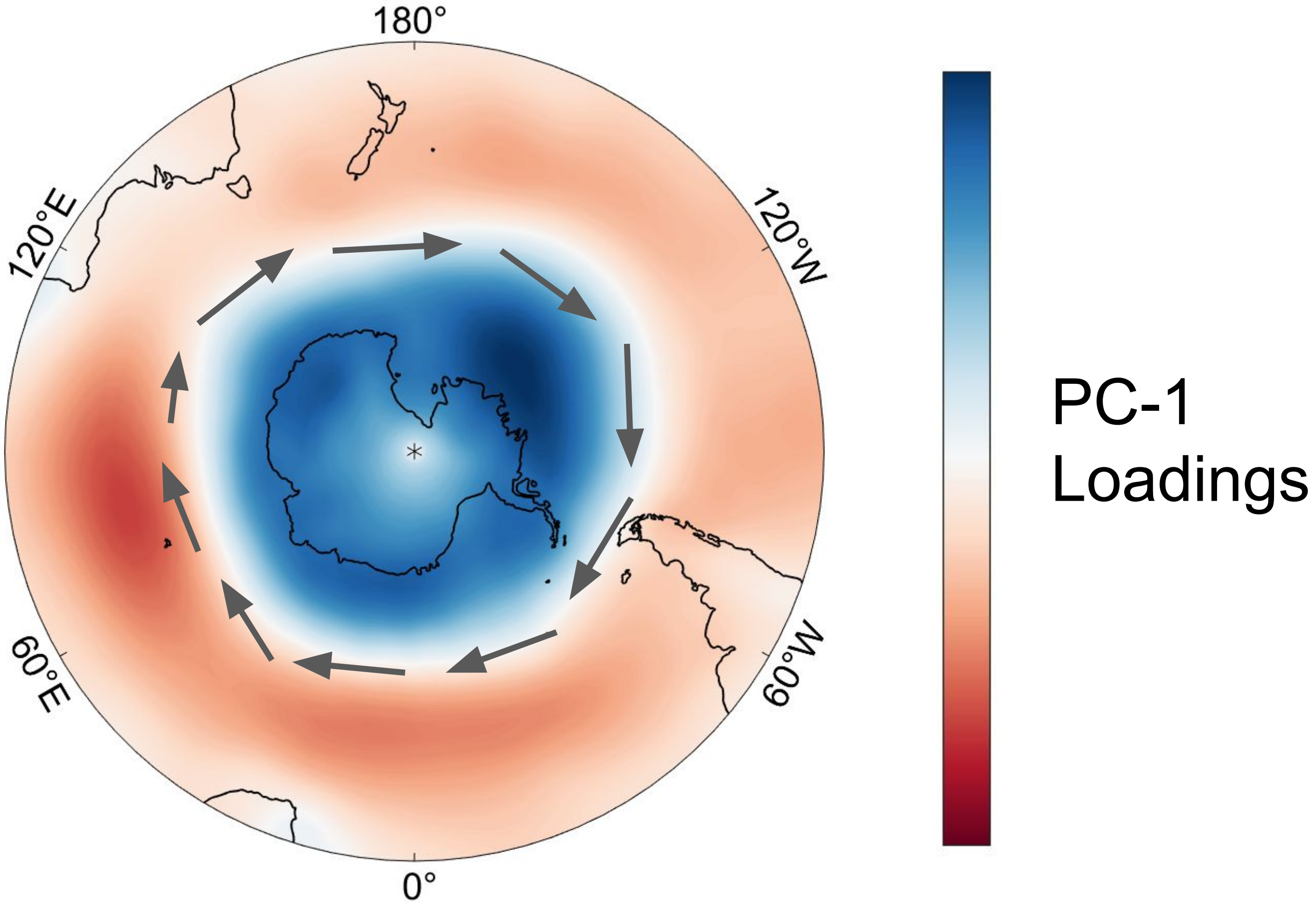


SAM Index



Marshall Index
4
0
-4
SAM Index
1950
1960
1970
1980
1990
2000
2010
2020
Year (CE)

Climate Impacts



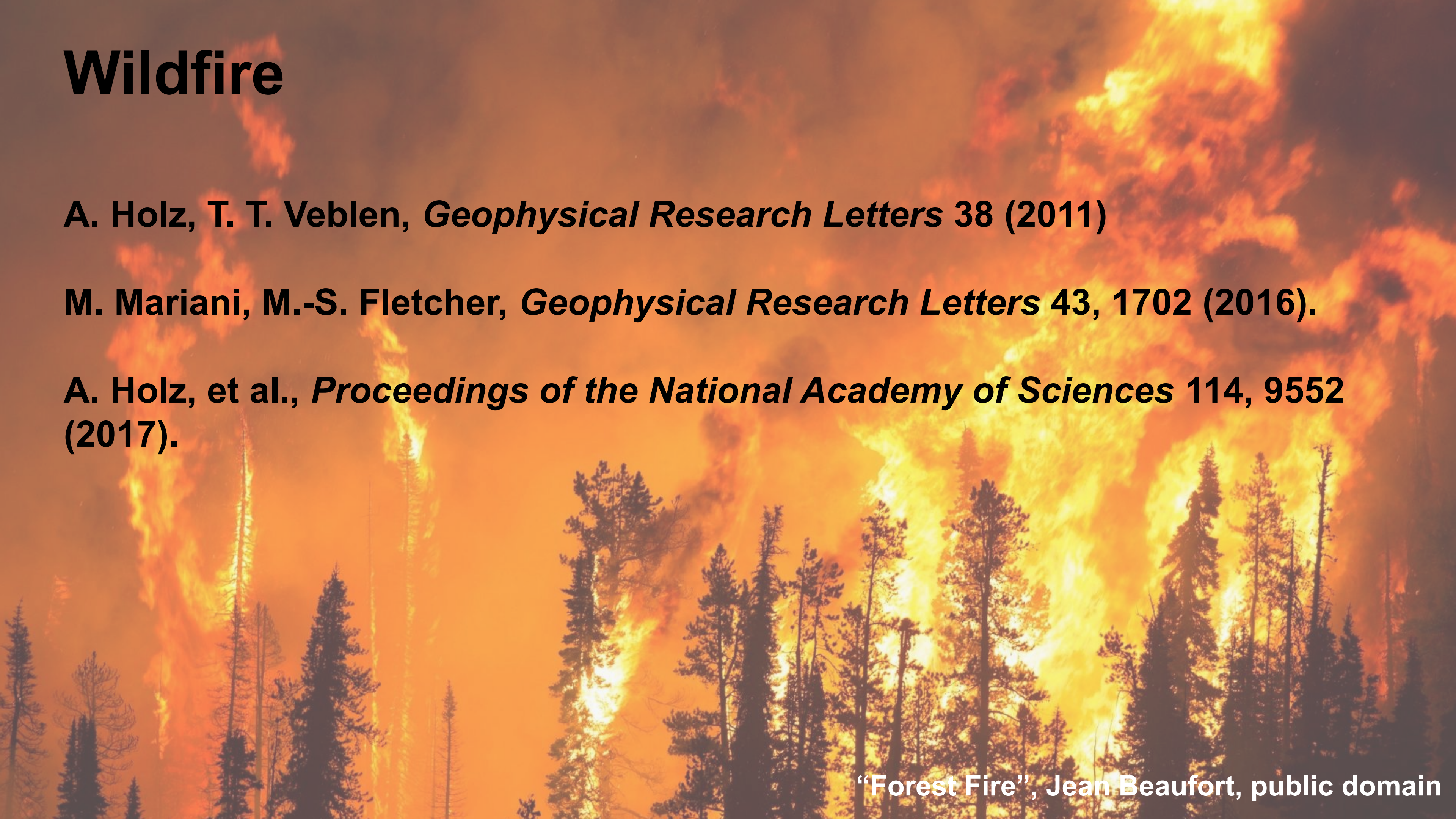
“Forest Fire”, Jean Beaufort, public domain

Wildfire

A. Holz, T. T. Veblen, *Geophysical Research Letters* 38 (2011)

M. Mariani, M.-S. Fletcher, *Geophysical Research Letters* 43, 1702 (2016).

A. Holz, et al., *Proceedings of the National Academy of Sciences* 114, 9552 (2017).



“Forest Fire”, Jean Beaufort, public domain

“Dry soil”, Francesco Ungaro, public domain

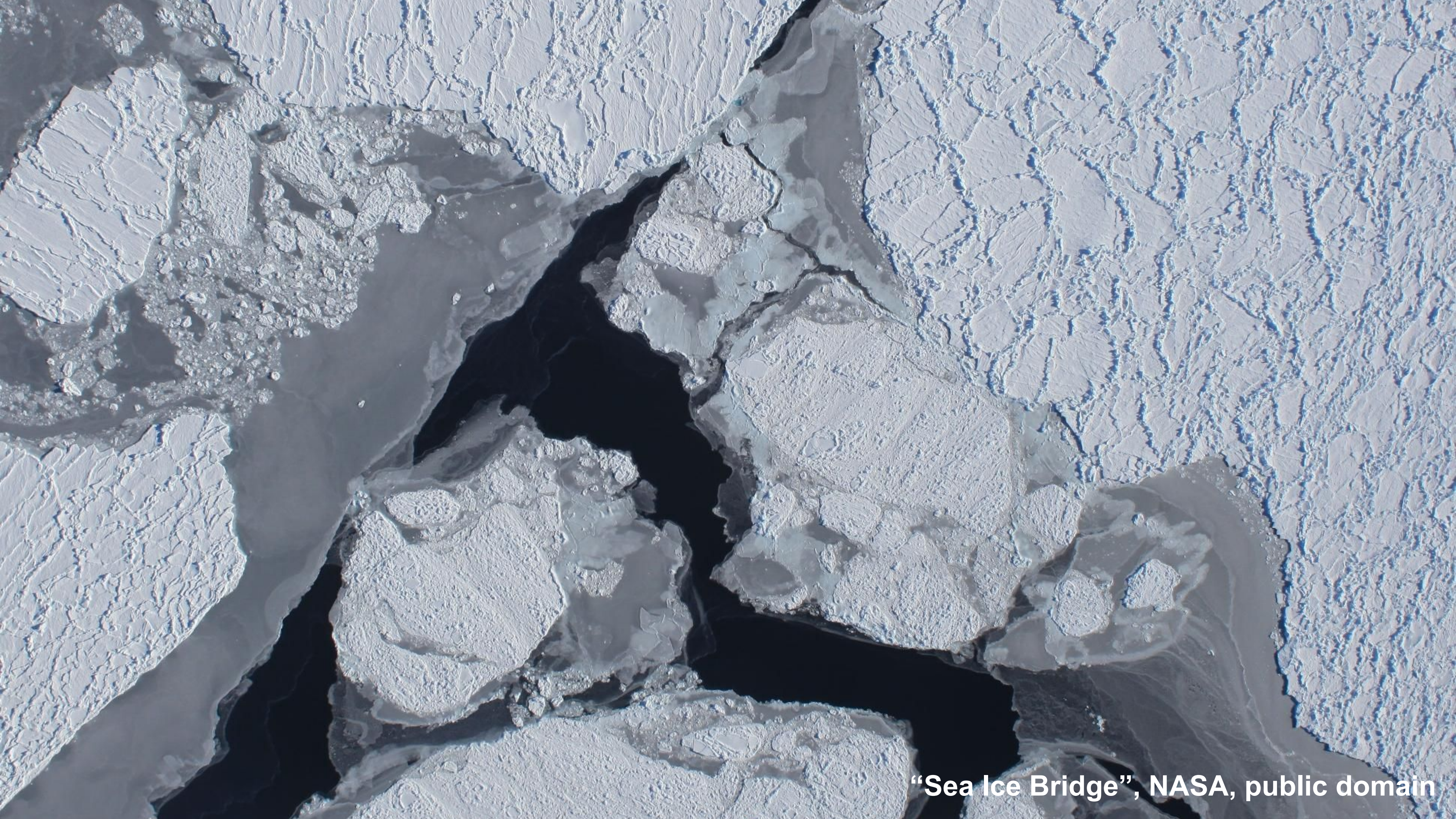
Drought

P. M. Sousa, R. C. Blamey, C. J. Reason, A. M. Ramos, R. M. Trigo, *Environmental Research Letters* 13, 124025 (2018).

D. C. Verdon-Kidd, A. S. Kiem, *Geophysical Research Letters* 36(2009).

W. Cai, P. Van Rensch, S. Borlace, T. Cowan, *Geophysical Research Letters* 38 (2011).

“Dry soil”, Francesco Ungaro, public domain



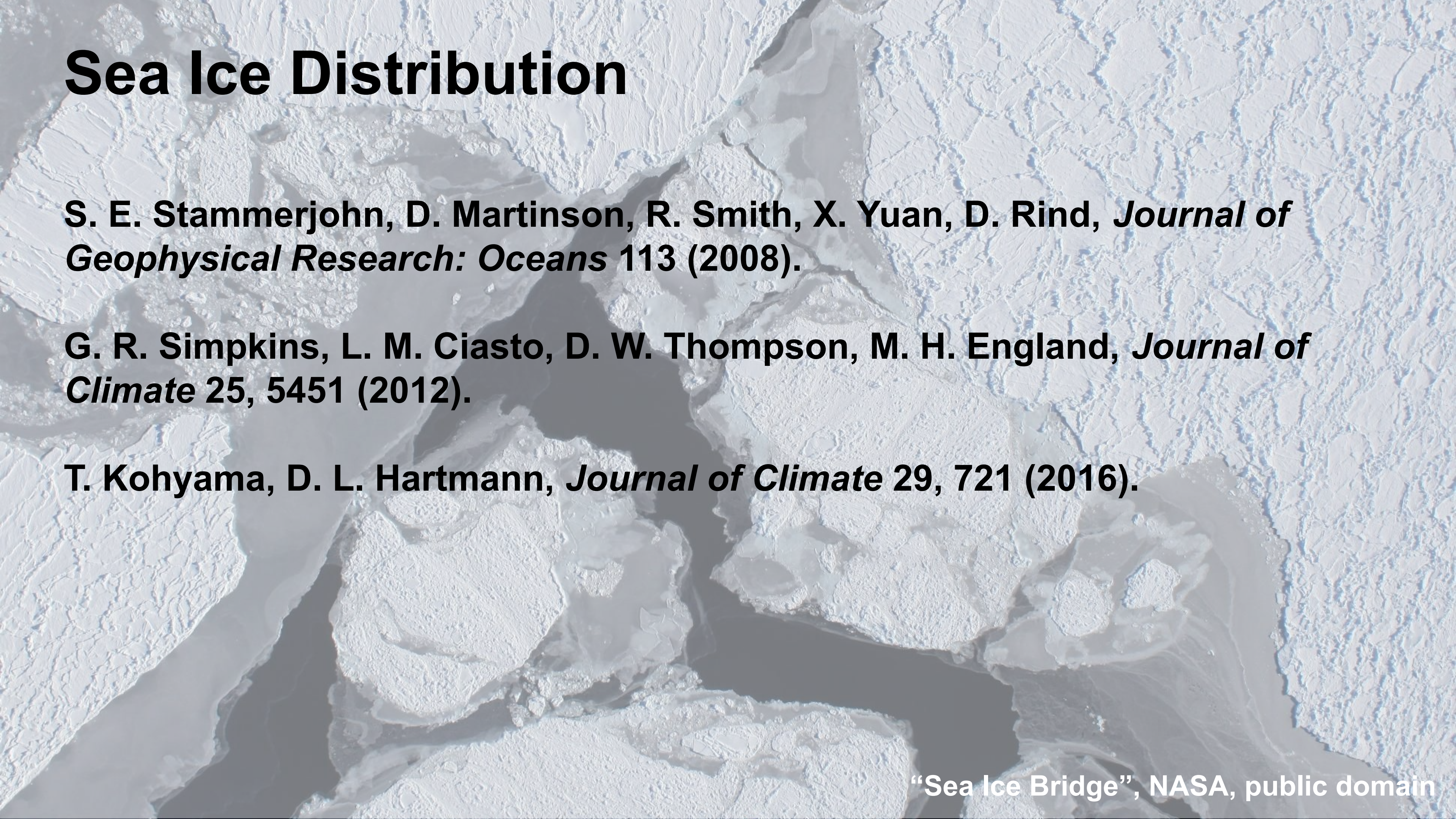
“Sea Ice Bridge”, NASA, public domain

Sea Ice Distribution

S. E. Stammerjohn, D. Martinson, R. Smith, X. Yuan, D. Rind, *Journal of Geophysical Research: Oceans* 113 (2008).

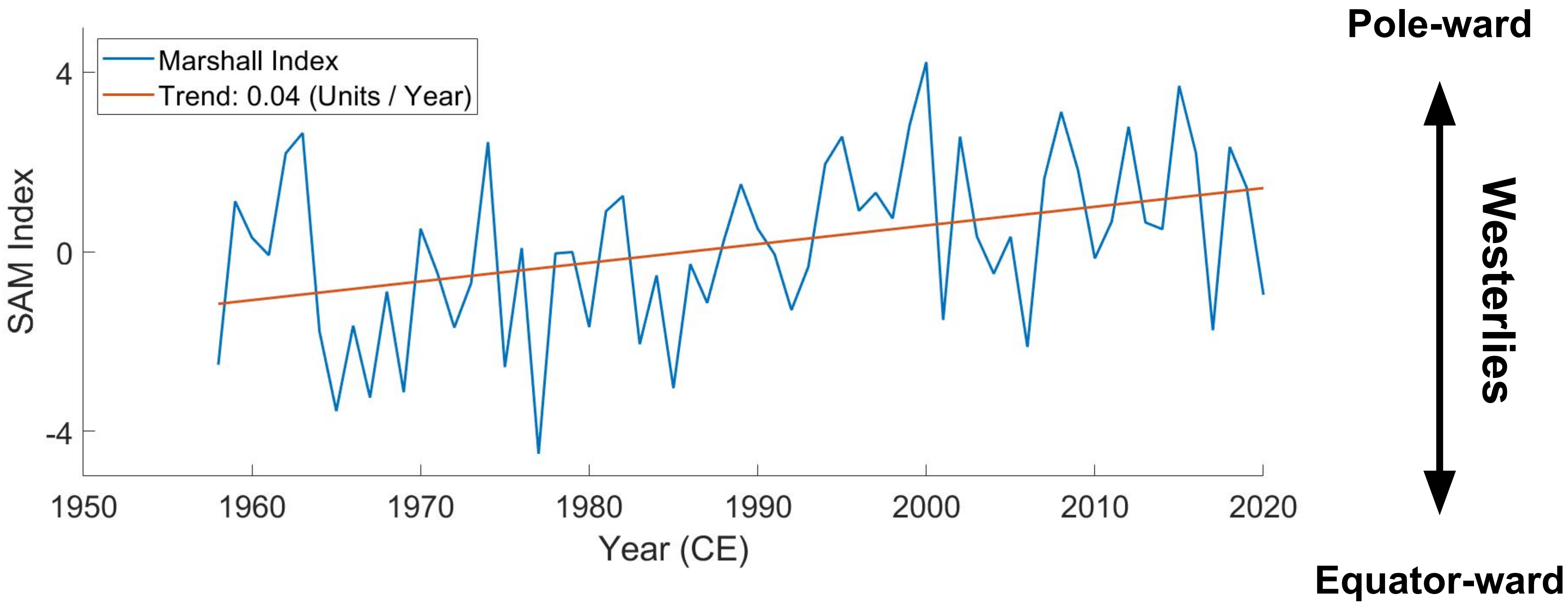
G. R. Simpkins, L. M. Ciasto, D. W. Thompson, M. H. England, *Journal of Climate* 25, 5451 (2012).

T. Kohyama, D. L. Hartmann, *Journal of Climate* 29, 721 (2016).

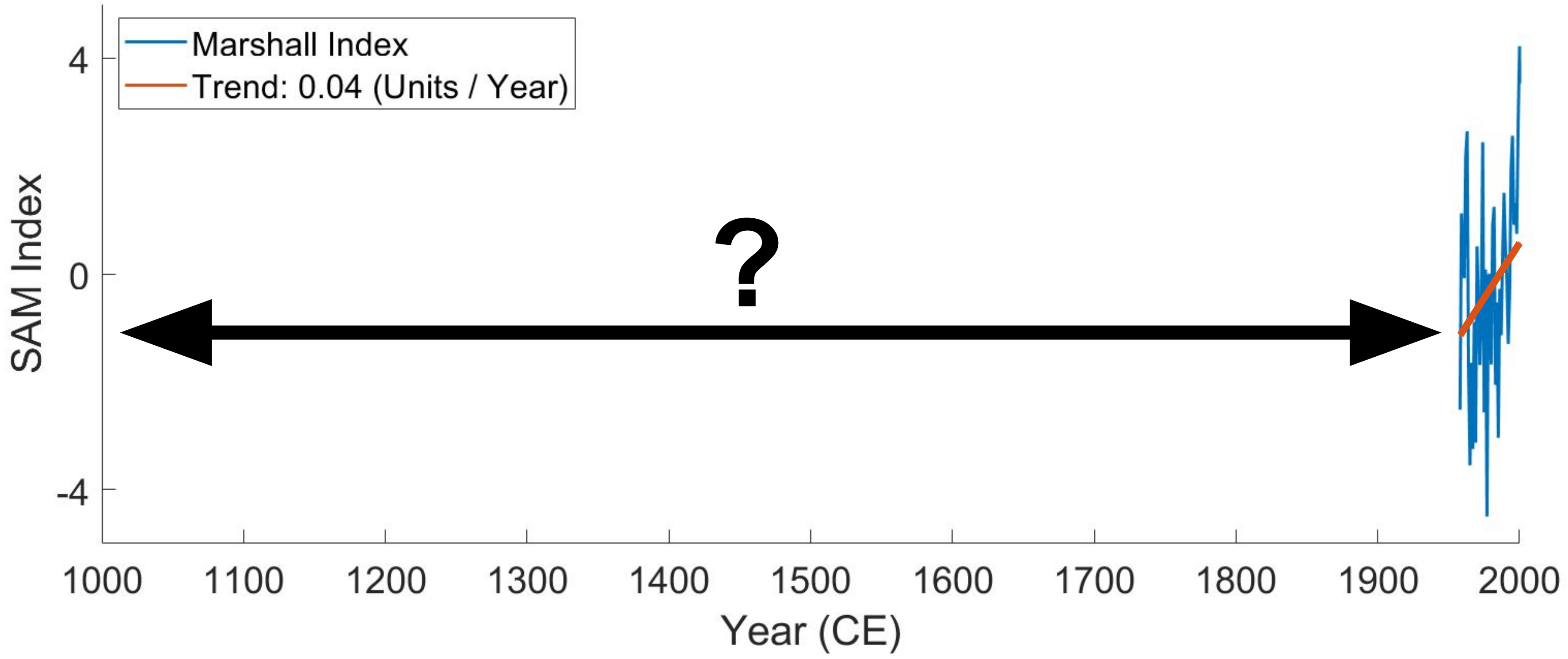


“Sea Ice Bridge”, NASA, public domain

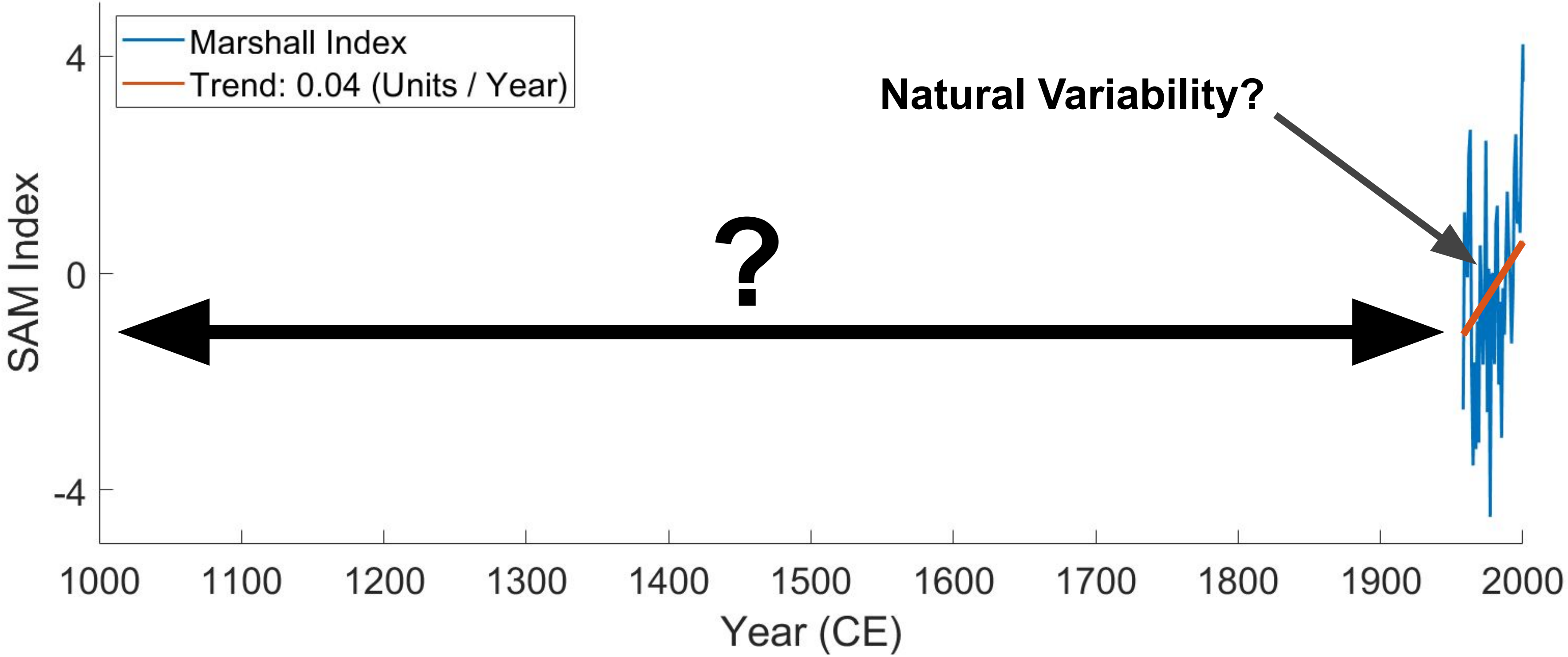
Modern Trend



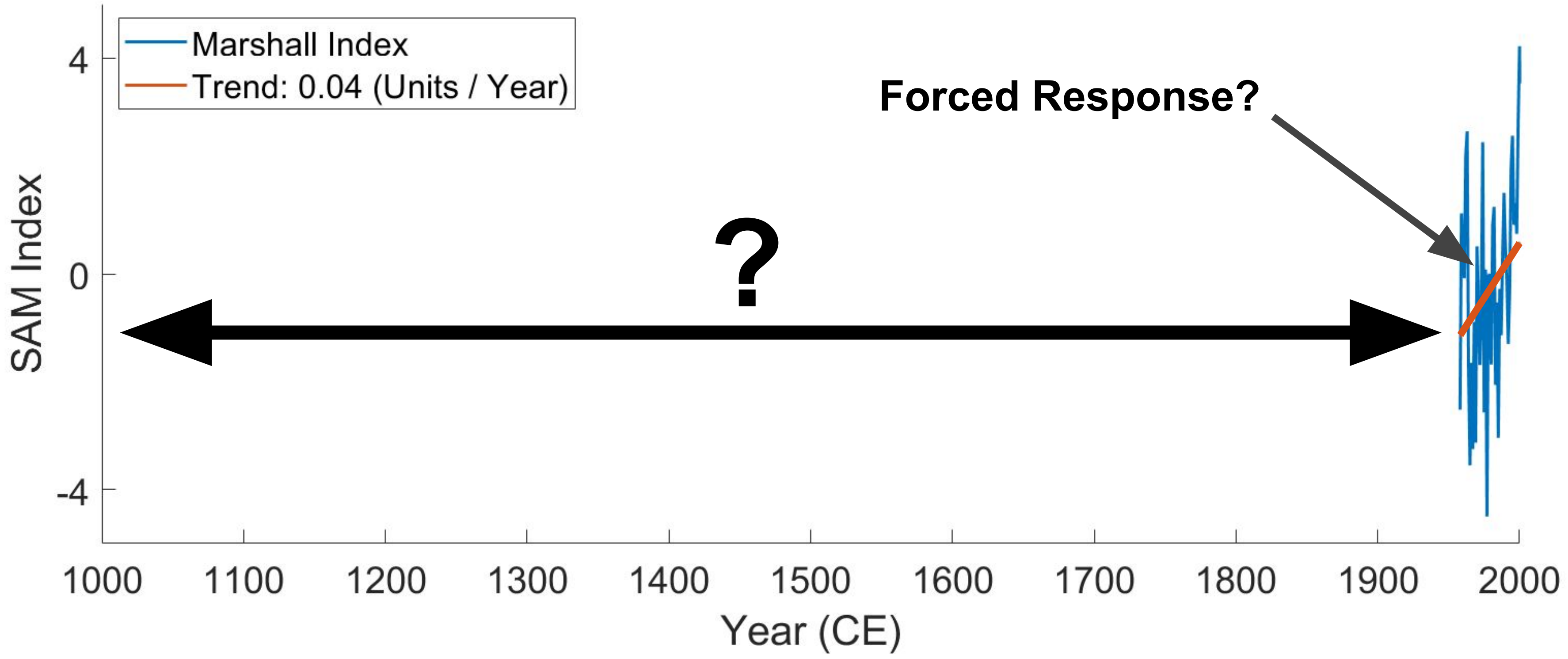
Long-term Context



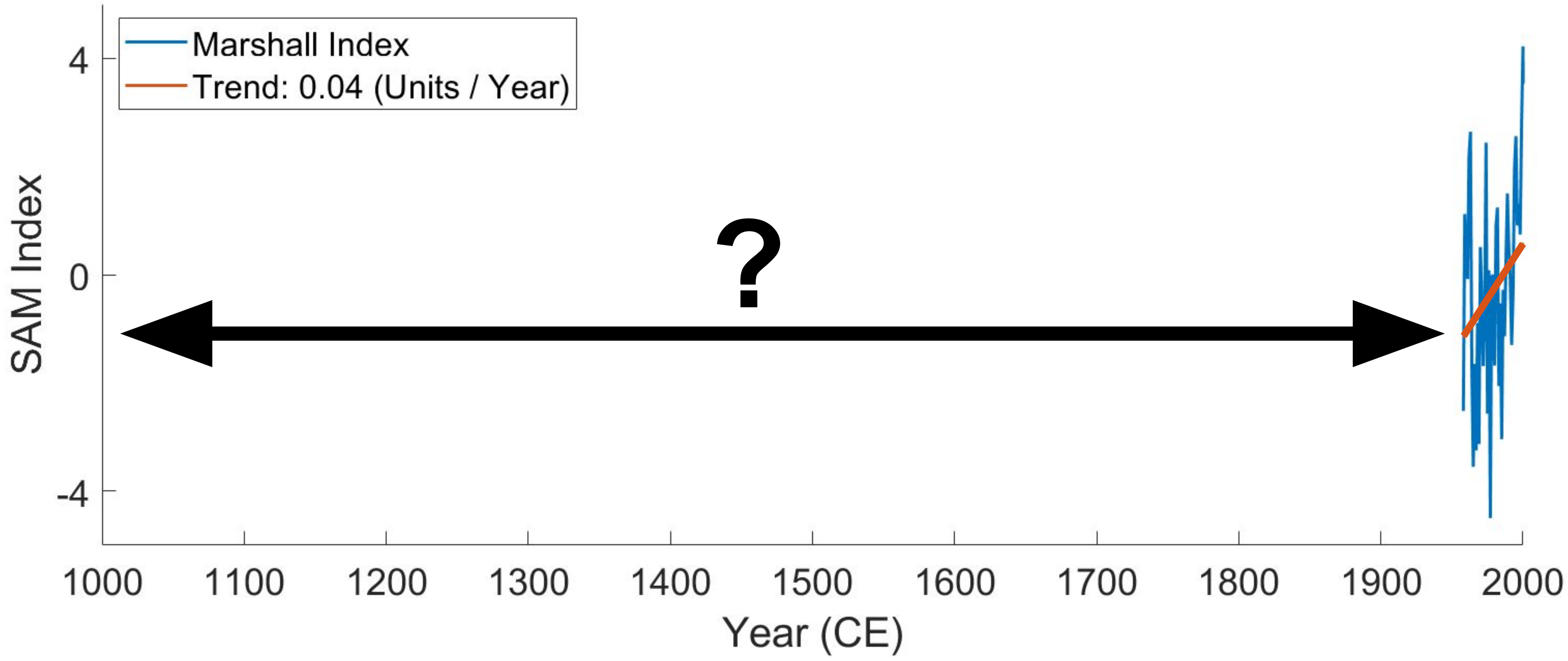
Long-term Context



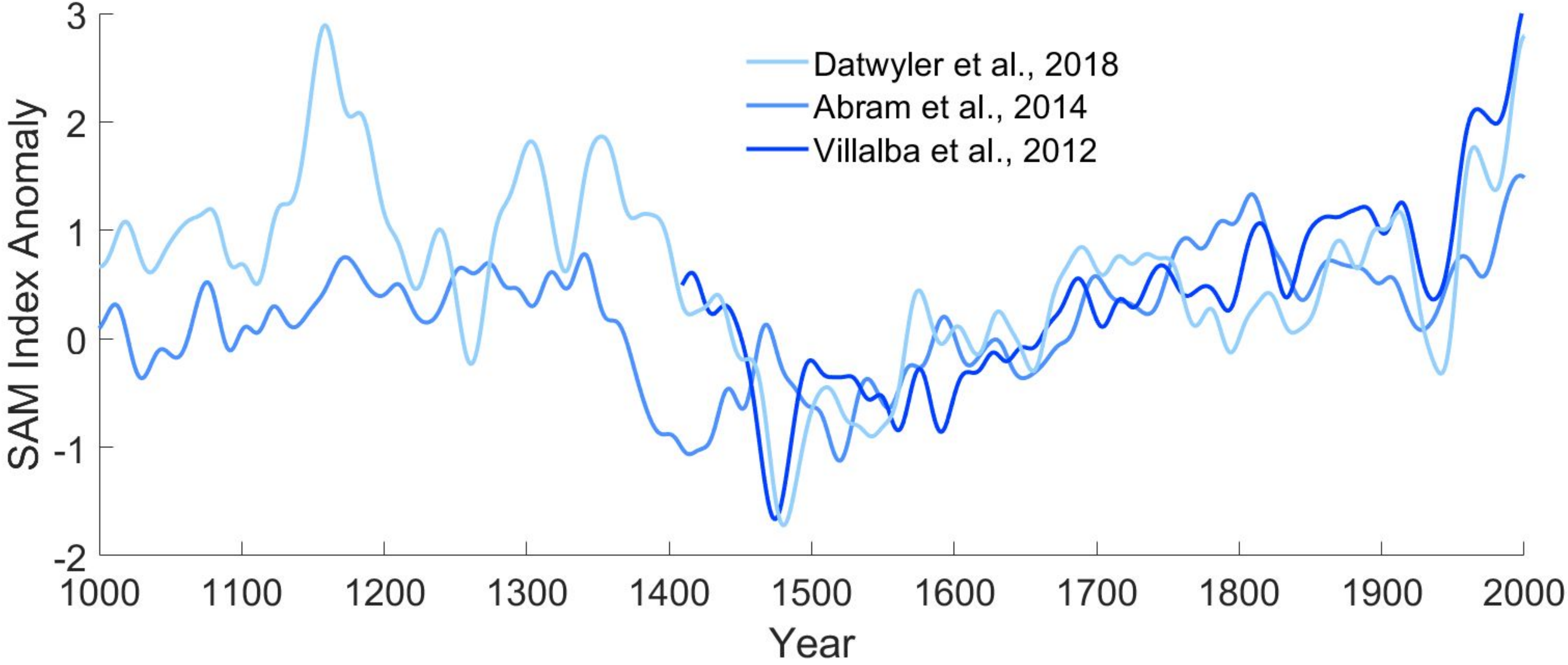
Long-term Context



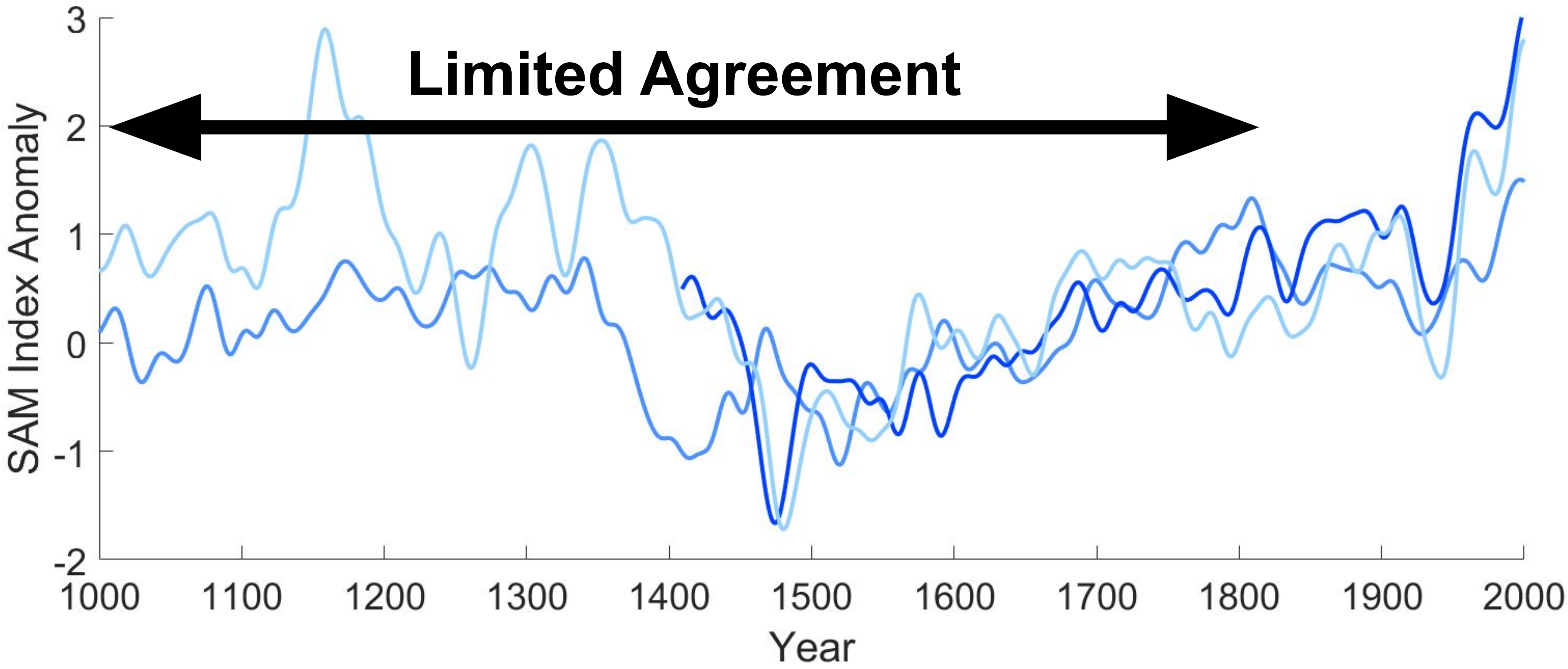
Long-term Context



SAM Reconstructions



SAM Reconstructions



IPCC AR6

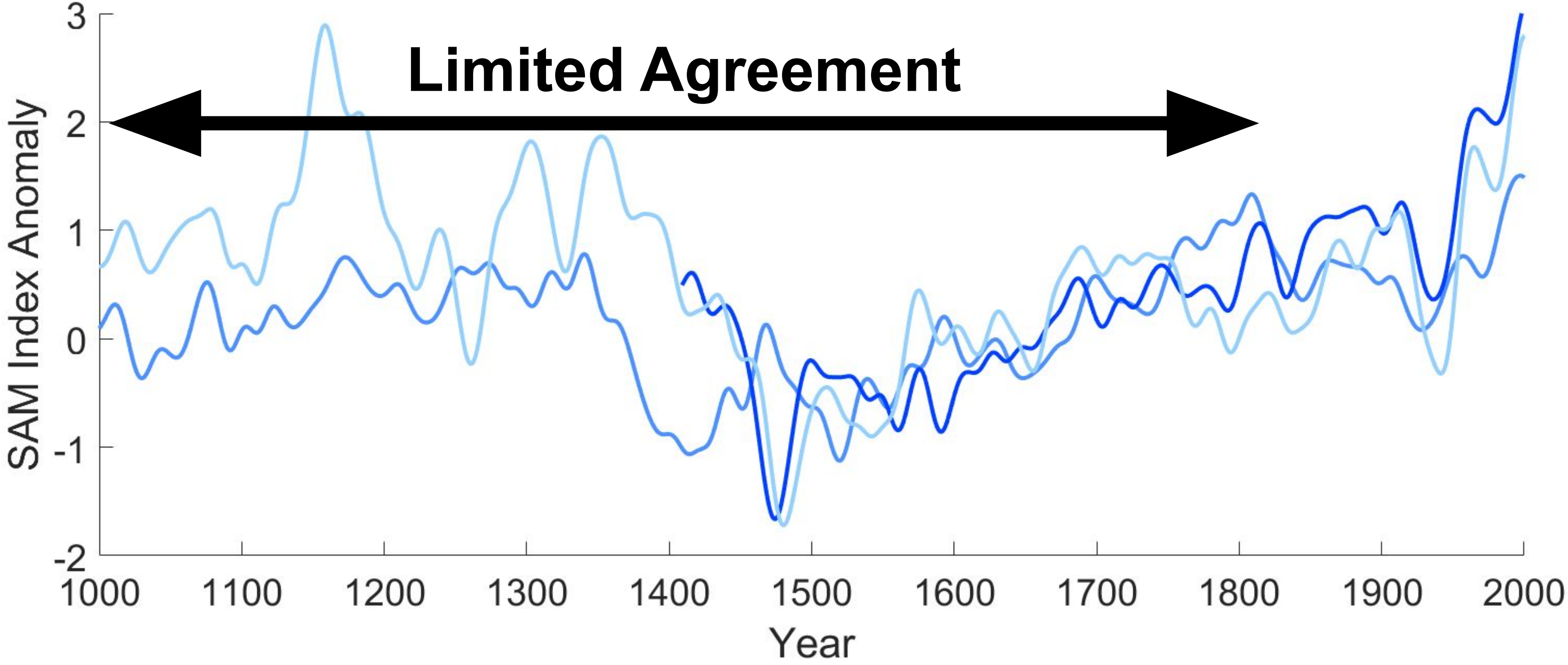
“The recent positive trend in the SAM is *likely* unprecedented in at least the past millennium,

although ***medium confidence* arises due to the differences between proxy records before 1800 CE.”**

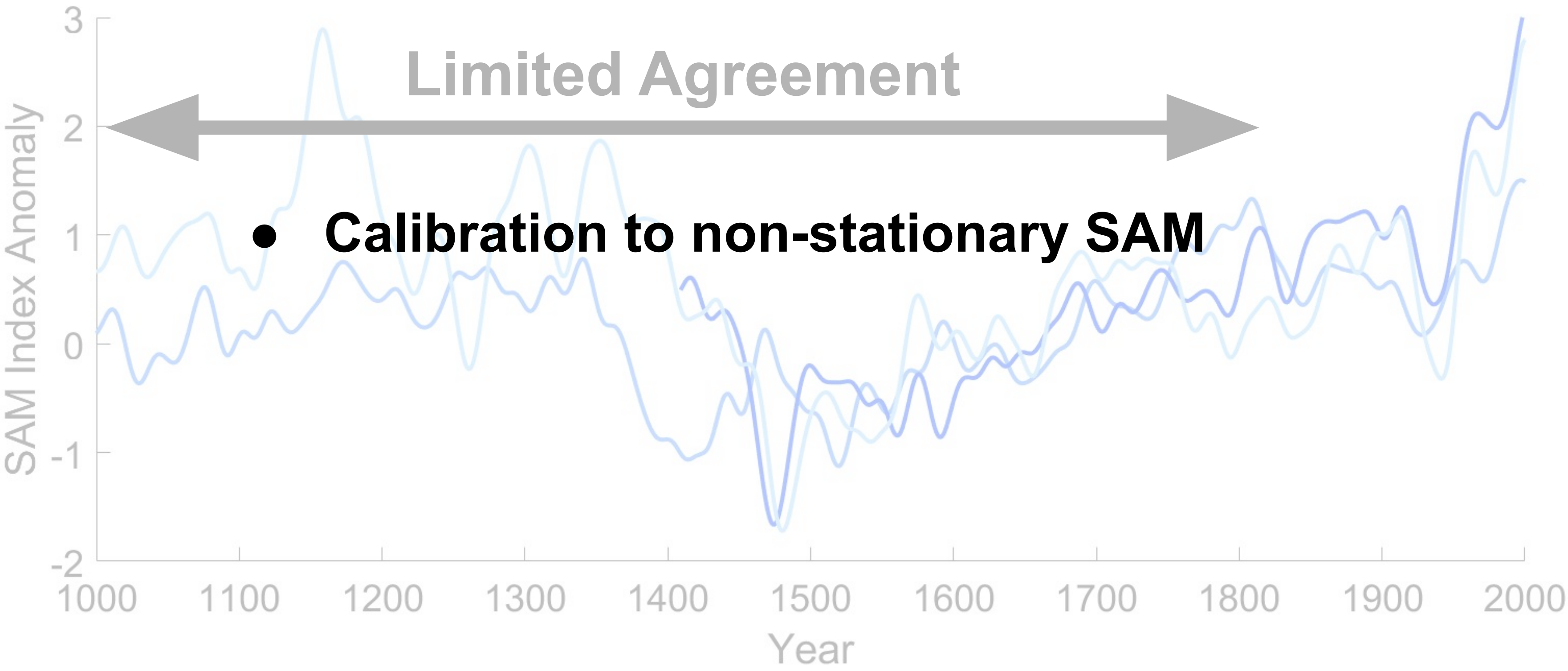
- IPCC AR6, 2021*

* The Physical Science Basis, Chapter 2

SAM Reconstructions

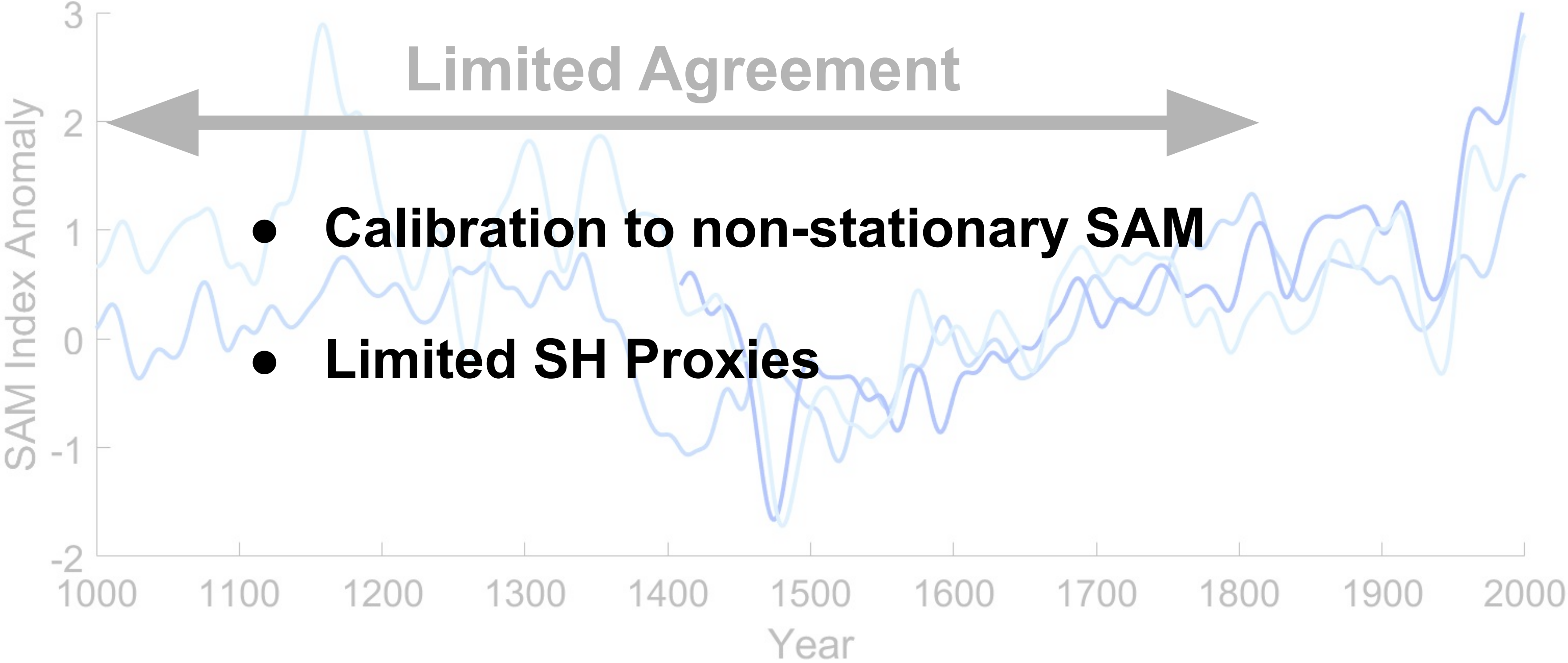


SAM Reconstructions



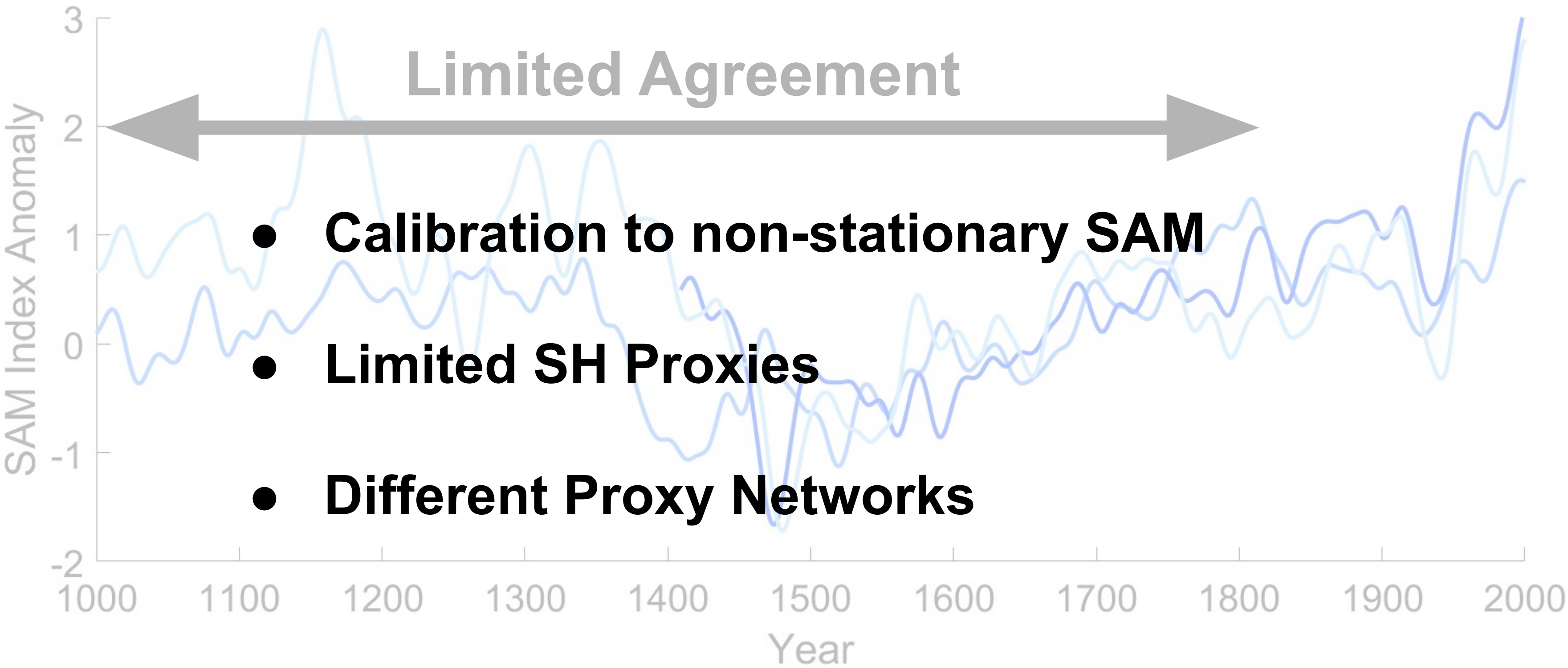
- **Calibration to non-stationary SAM**

SAM Reconstructions



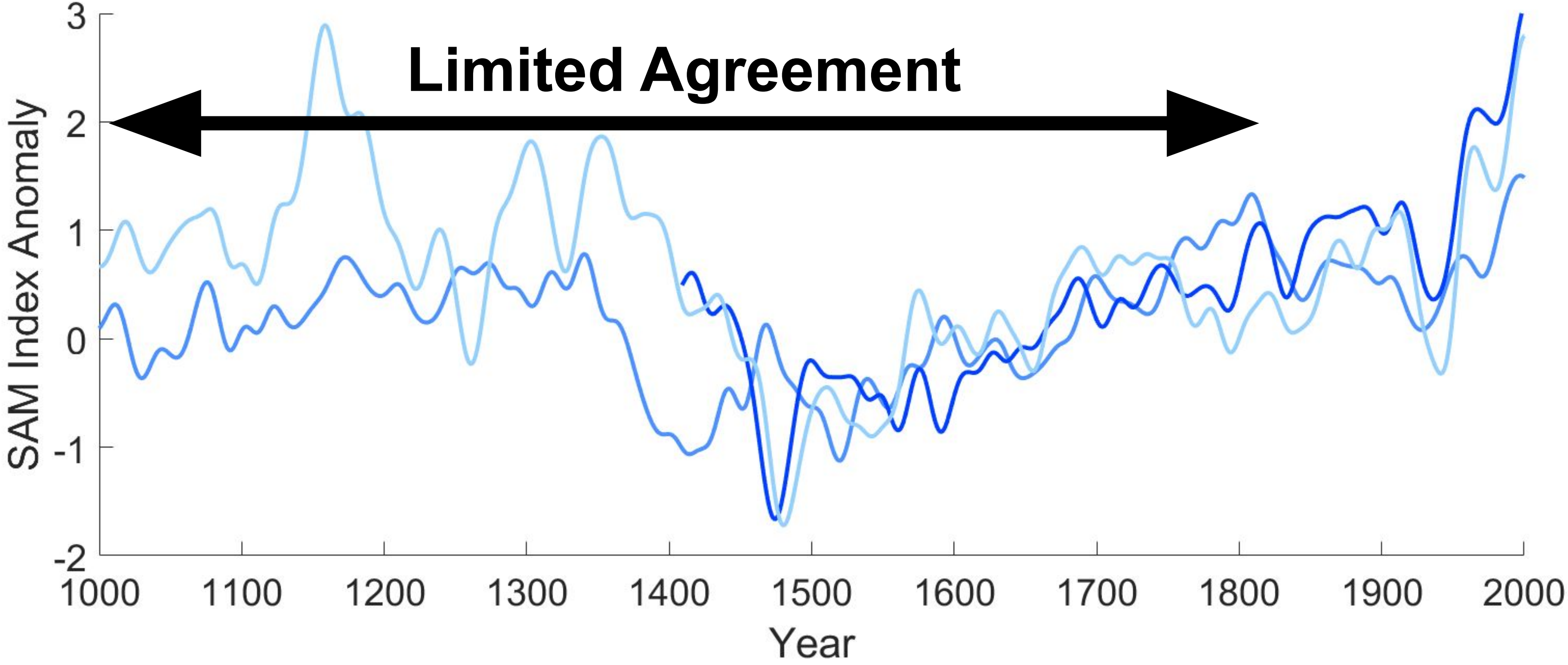
- **Calibration to non-stationary SAM**
- **Limited SH Proxies**

SAM Reconstructions

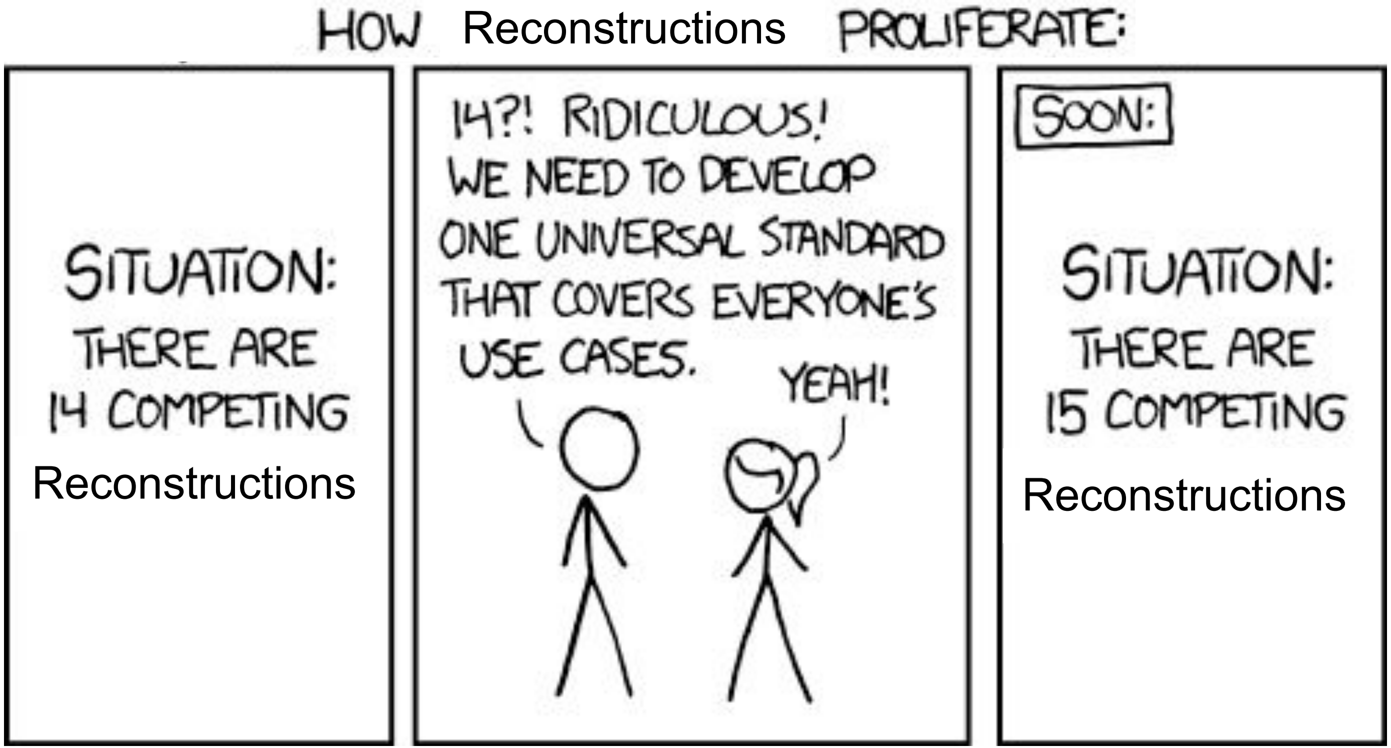


- Calibration to non-stationary SAM
- Limited SH Proxies
- Different Proxy Networks

SAM Reconstructions



A New Reconstruction



Adapted from "Standards" by Randall Munroe on xkcd.com, CC-BY-NC 2.5

Context

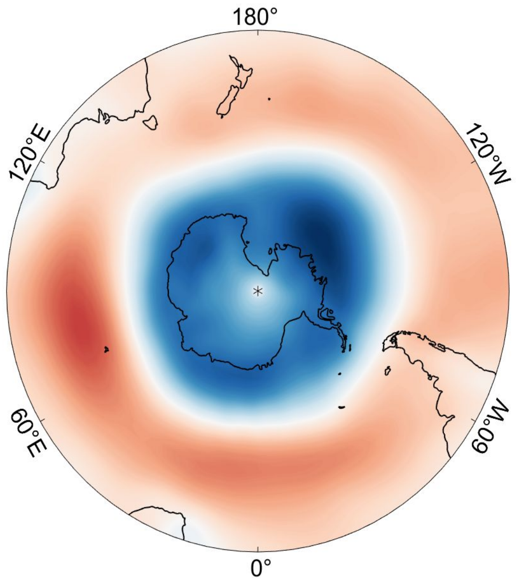
Data Assimilation Method

Results

Data Assimilation

Climate Models

Data Assimilation!



Proxy Records

Climate Reconstructions

New Reconstruction

1. **Data Assimilation**
    a. **Method: Offline ensemble Kalman filter**

- Evensen et al., (1994)
- Oke et al., (2002)
- Evensen et al., (2003)

New Reconstruction

1. **Data Assimilation**
   a. Method: Offline ensemble Kalman filter
   b. **Prior: Stationary multi-model**

- Parsons et al., (2021)
- King et al., (2021)

New Reconstruction

1. **Data Assimilation**
   a. Method: Offline ensemble Kalman filter
   b. Prior: Stationary multi-model
   c. **Proxy system models:**

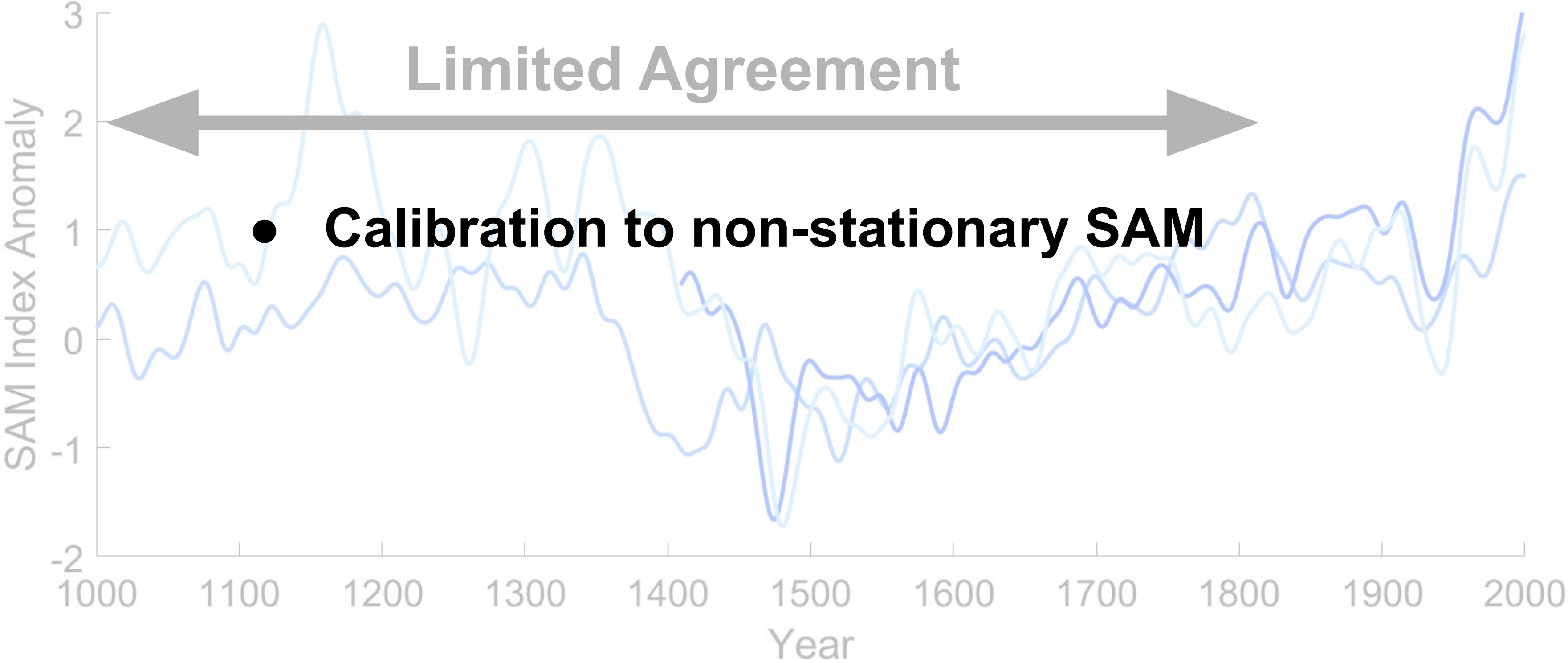
- Evans et al., (2013)

New Reconstruction

A. **Data Assimilation**
    1. Method: Offline ensemble Kalman filter
    2. Prior: Stationary multi-model
    3. Proxy system models:

B. **Advantages**
    1.
    2.
    3.

SAM Reconstructions

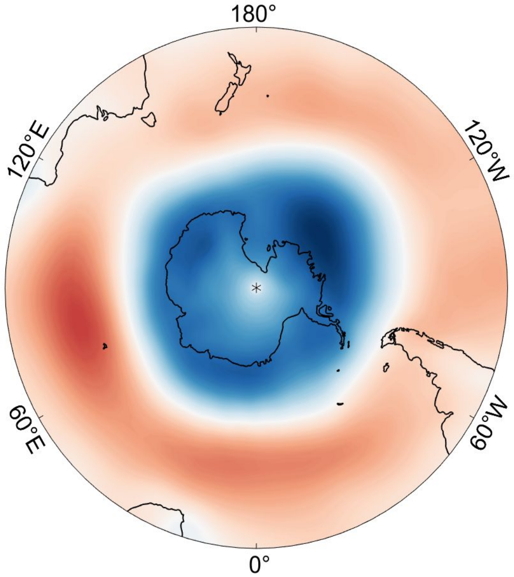


- **Calibration to non-stationary SAM**

Calibration: Existing Reconstructions

Climate Proxies

?



SAM

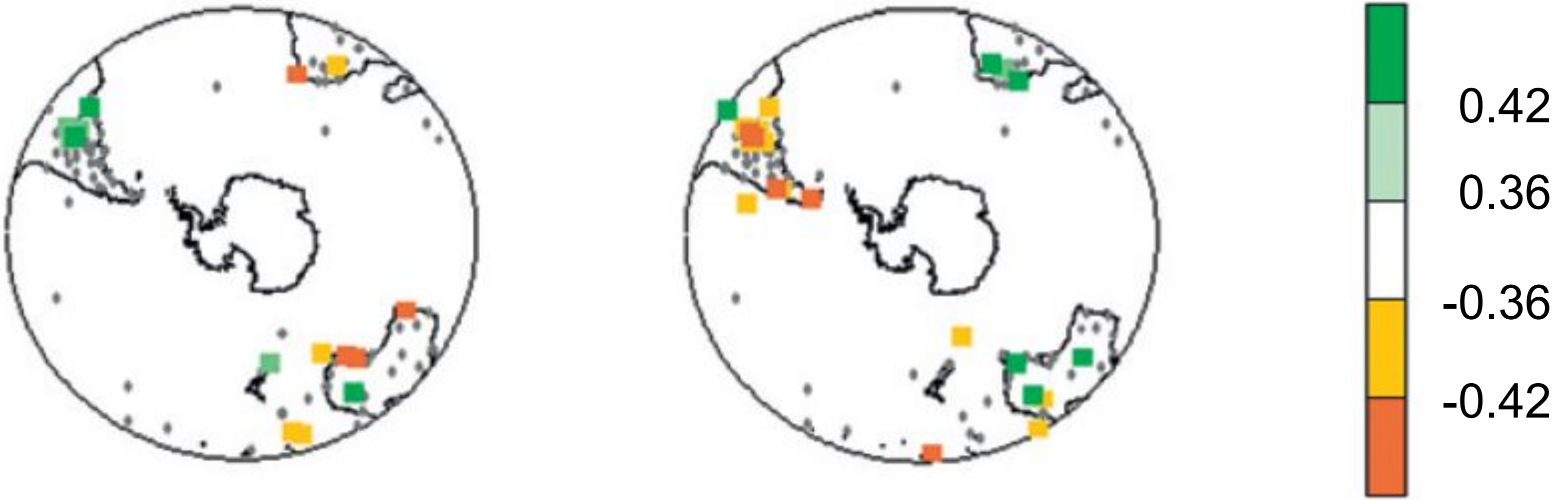
Calibration

Assumption:

Unchanging relationship between SAM and local climate variables.

Non-stationary SAM

Correlation of SAM with Precipitation



1958 - 1979 **1983 - 2004**

Adapted from Silvestri and Vera, (2009). Figure 1ab

Calibration: Data Assimilation

Climate Proxies

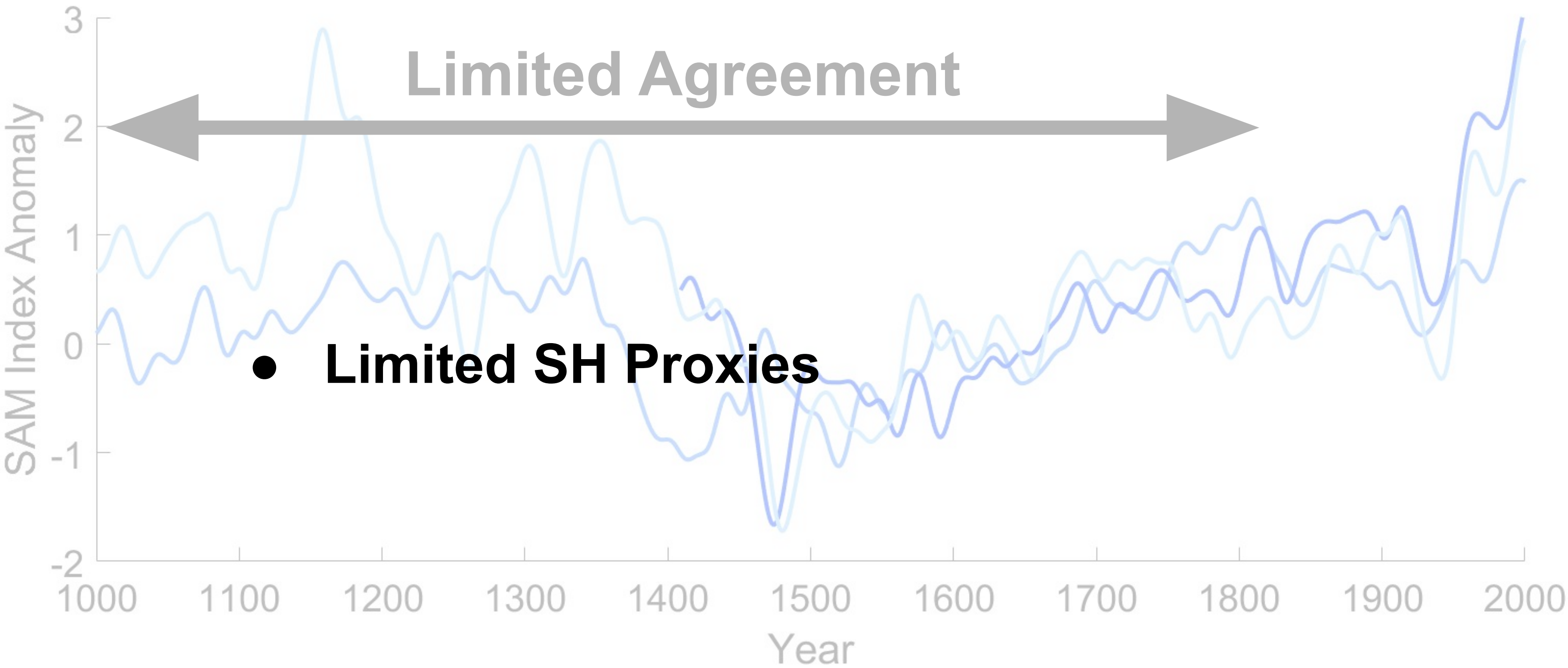
Local Climate Variables

New Reconstruction

A. **Data Assimilation**
   1. Method: Offline ensemble Kalman filter
   2. Prior: Stationary multi-model
   3. Proxy system models:

B. **Advantages**
   1. Not calibrated to SAM index
   2.
   3.

SAM Reconstructions

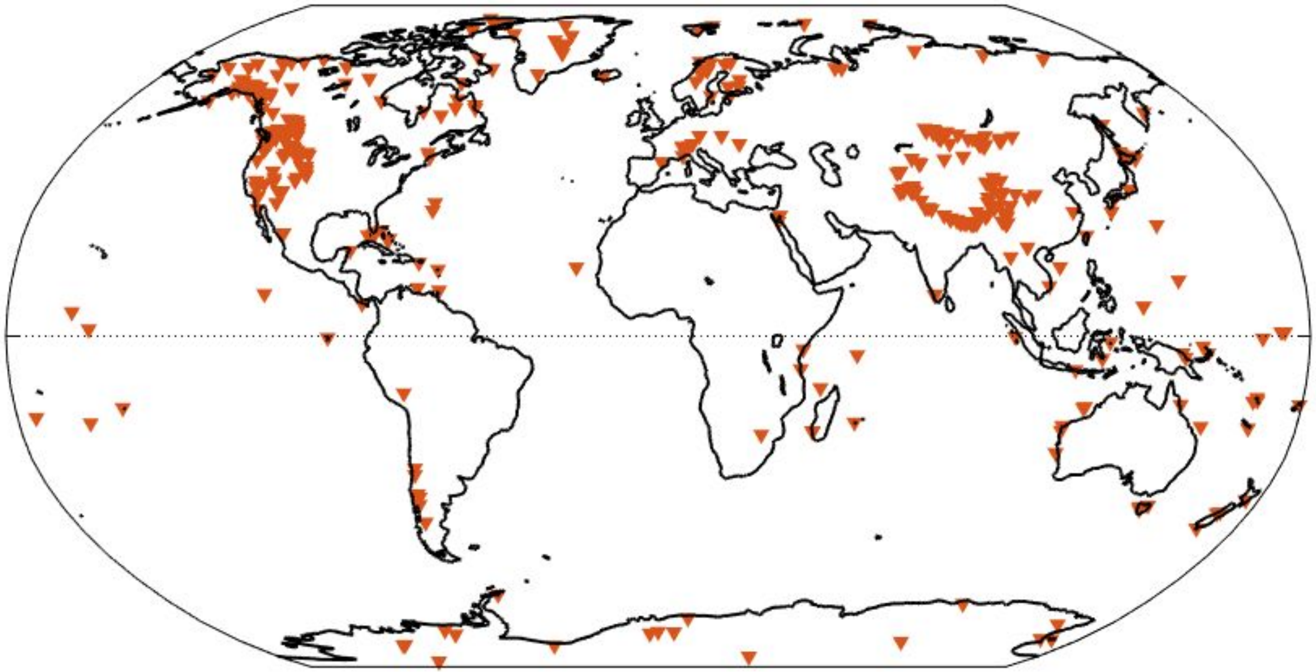


- **Limited SH Proxies**

Fewer SH Proxies

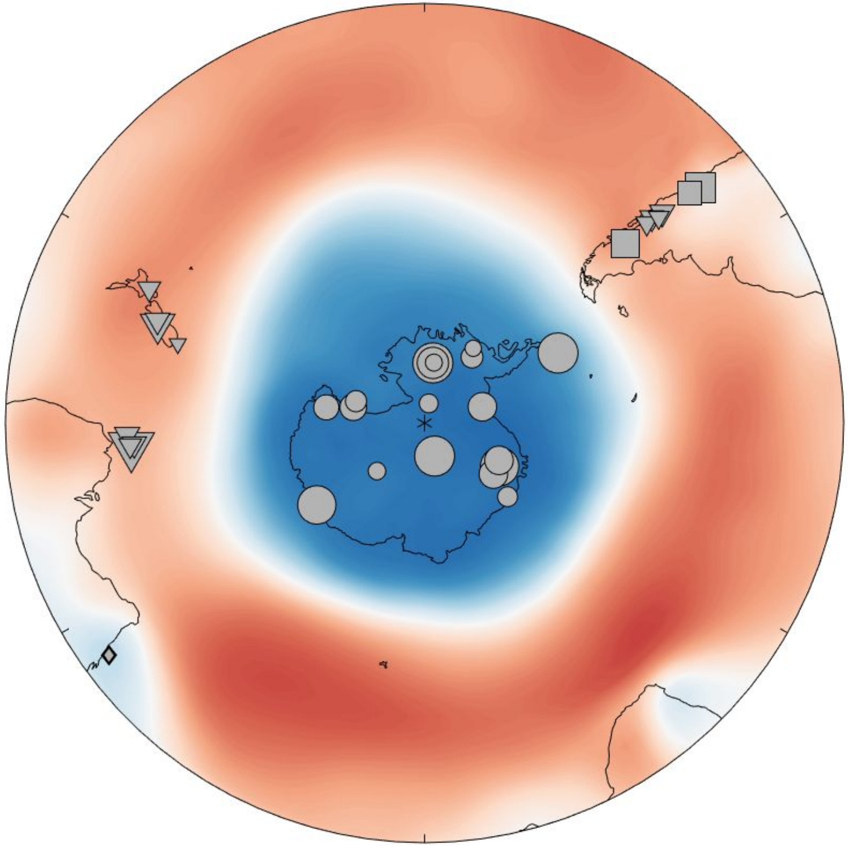
481

89

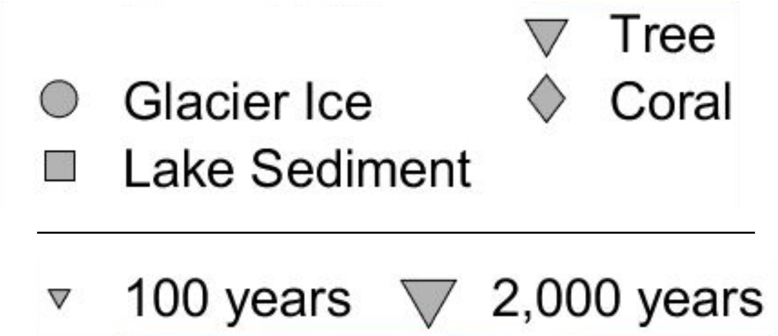


PAGES2k, annual resolution or greater

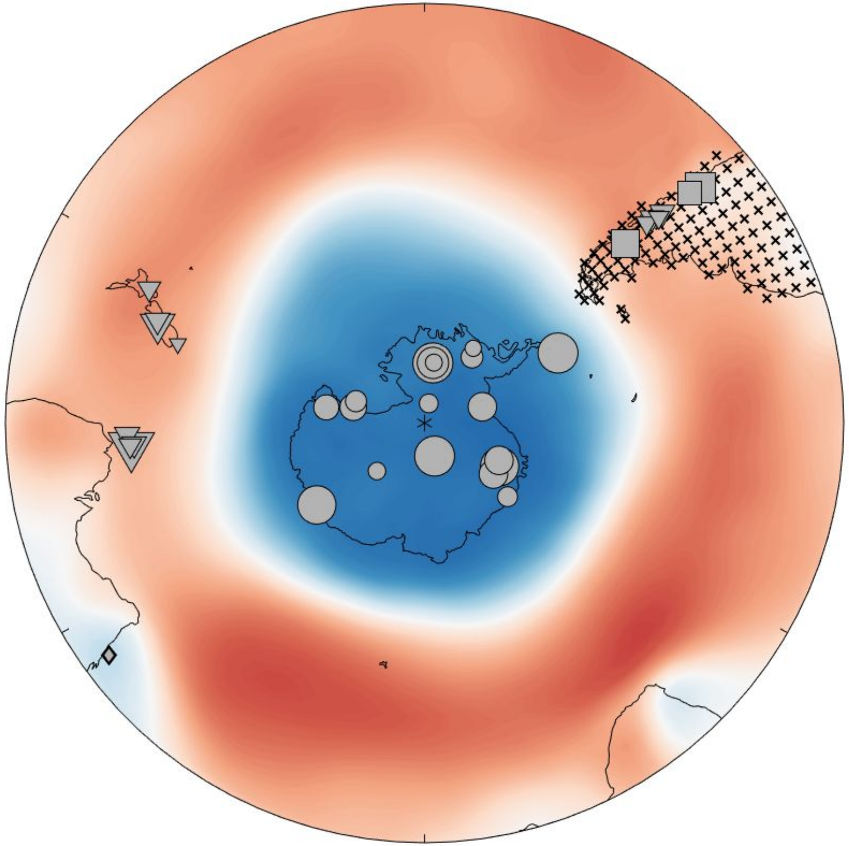
Proxy Network



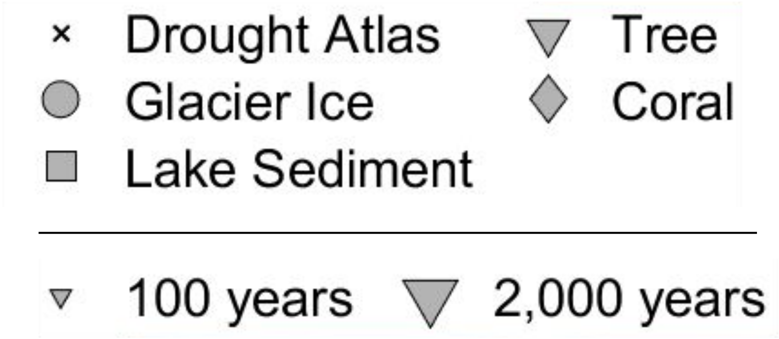
PAGES2k



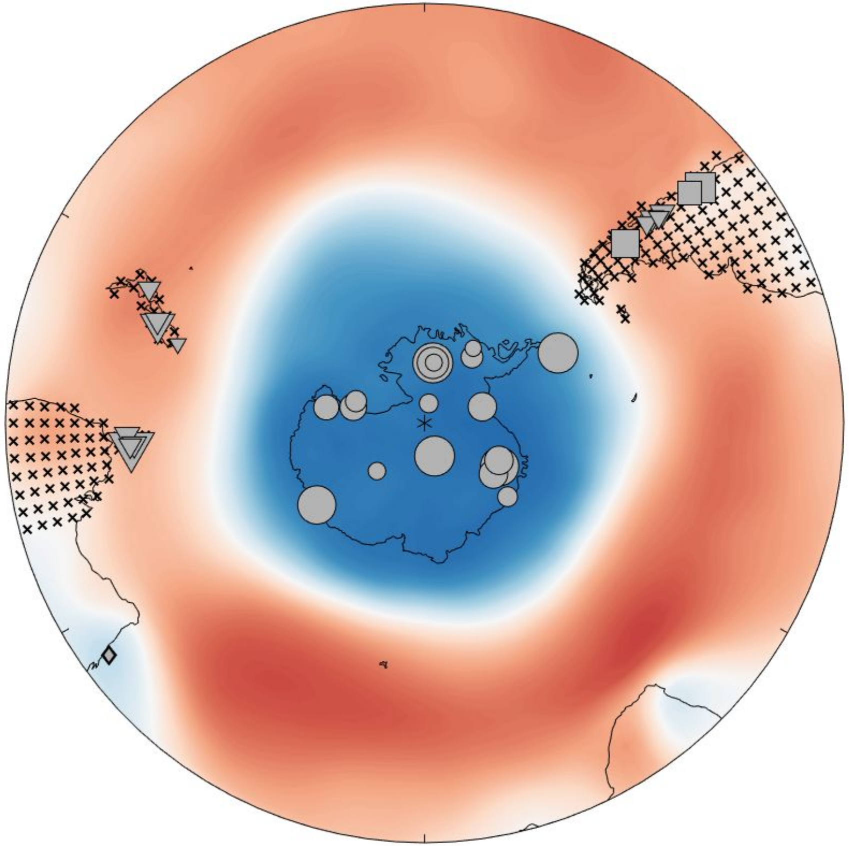
Proxy Network



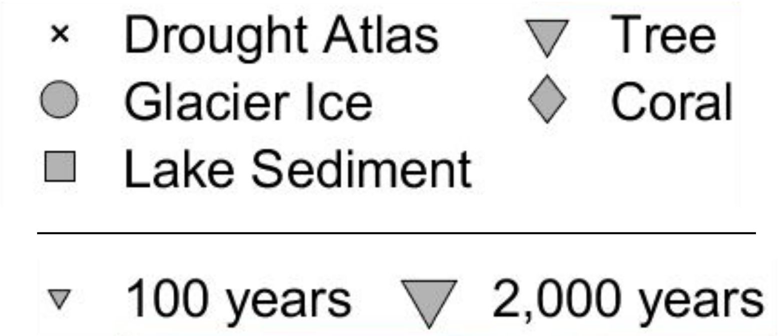
PAGES2k, SADA



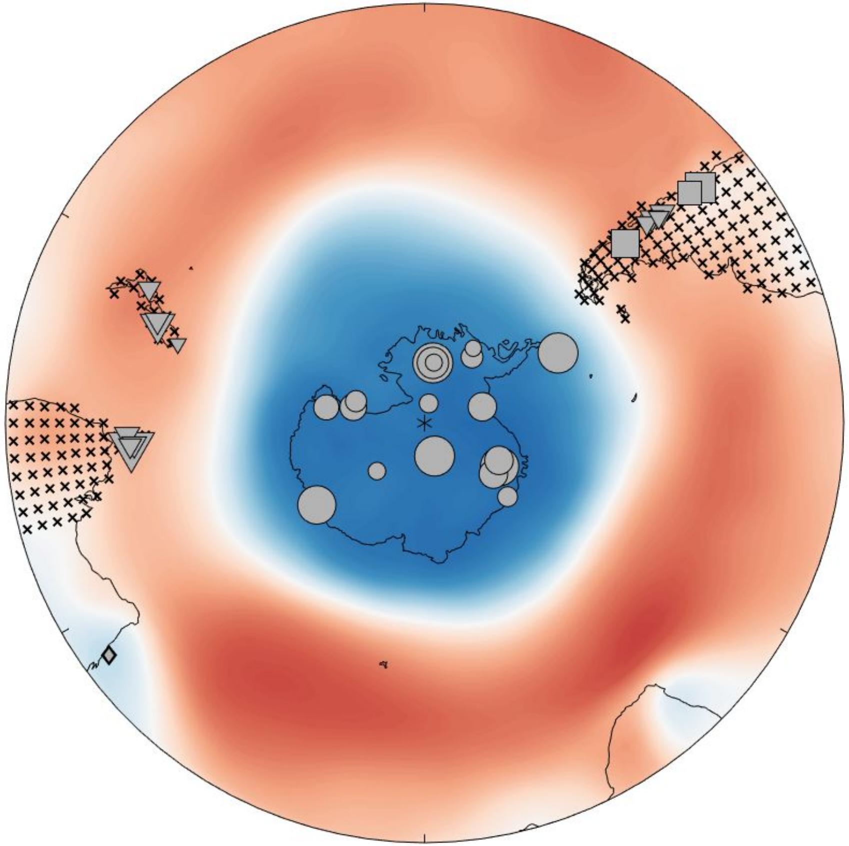
Proxy Network



PAGES2k, SADA, ANZDA



Proxy Network



SADA: 286 records

ANZDA: 177 records

New Reconstruction

A. **Data Assimilation**
   1. Method: Offline ensemble Kalman filter
   2. Prior: Stationary multi-model
   3. Proxy system models:

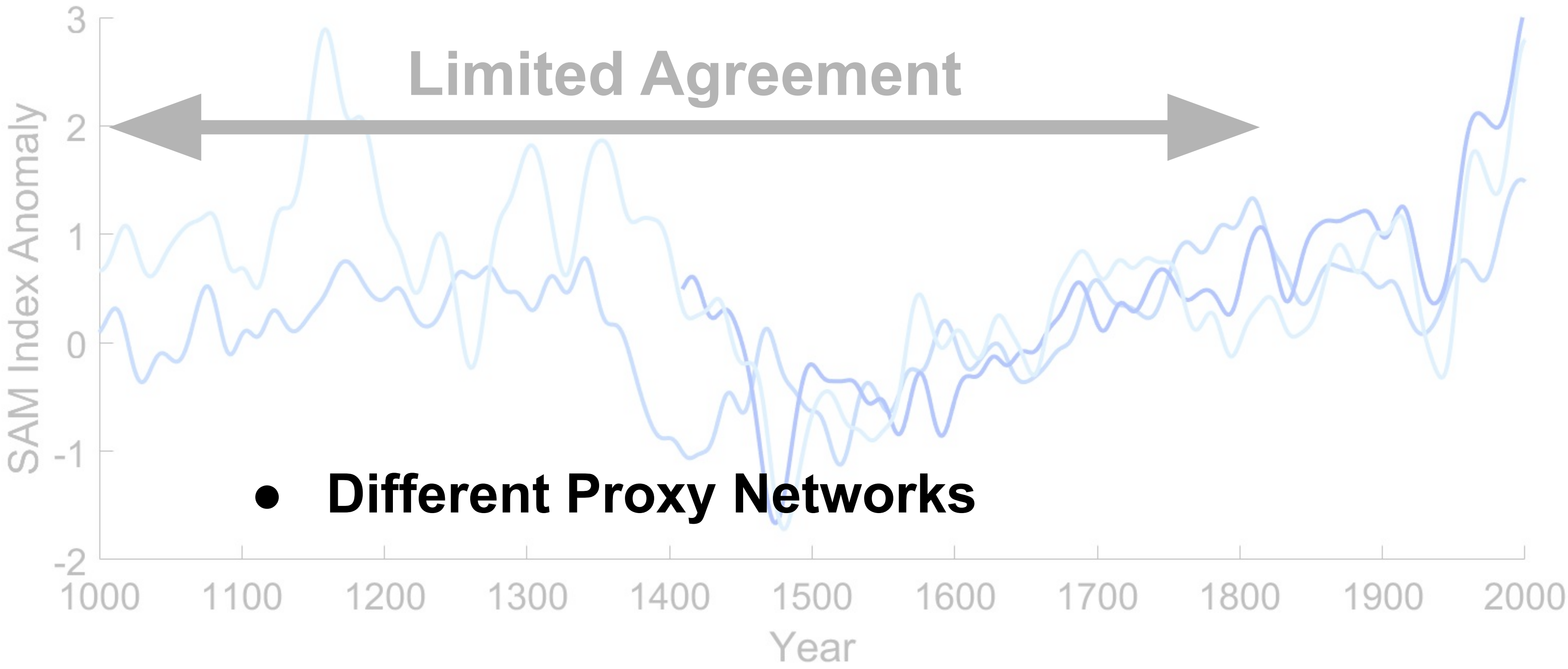
B. **Advantages**
   1. Not calibrated to SAM index
   2. More proxy information (drought atlases)
   3.

New Reconstruction

1. **Data Assimilation**
   a. Method: Offline ensemble Kalman filter
   b. Prior: Stationary multi-model
   c. Proxy system models: Linear seasonal temperature, $\mathbf{PDSI_{Thornthwaite}}$

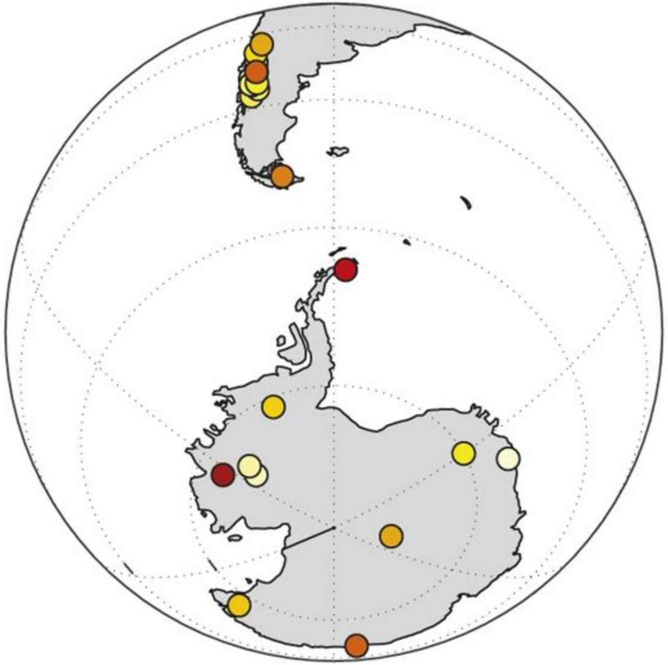
- Evans et al., (2013)

SAM Reconstructions

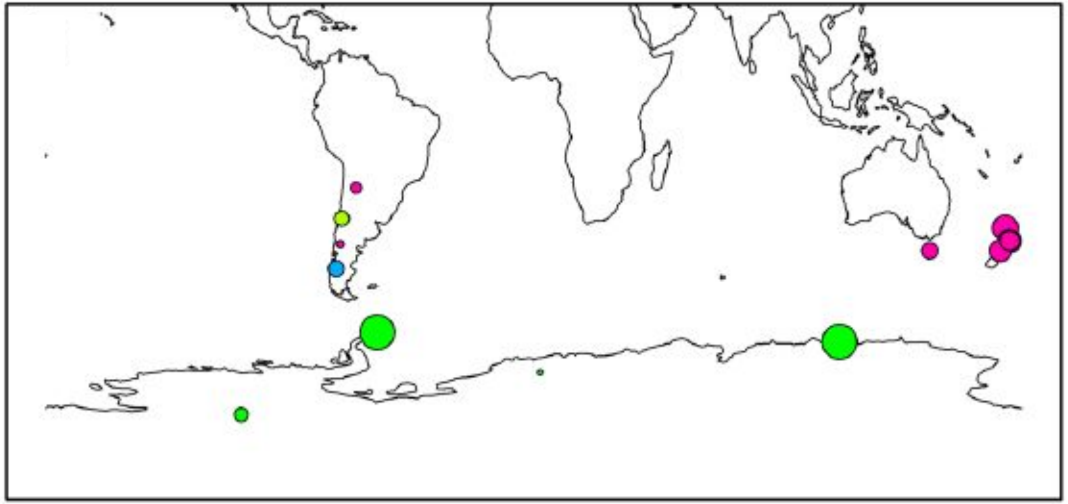


- **Different Proxy Networks**

Different Proxy Networks

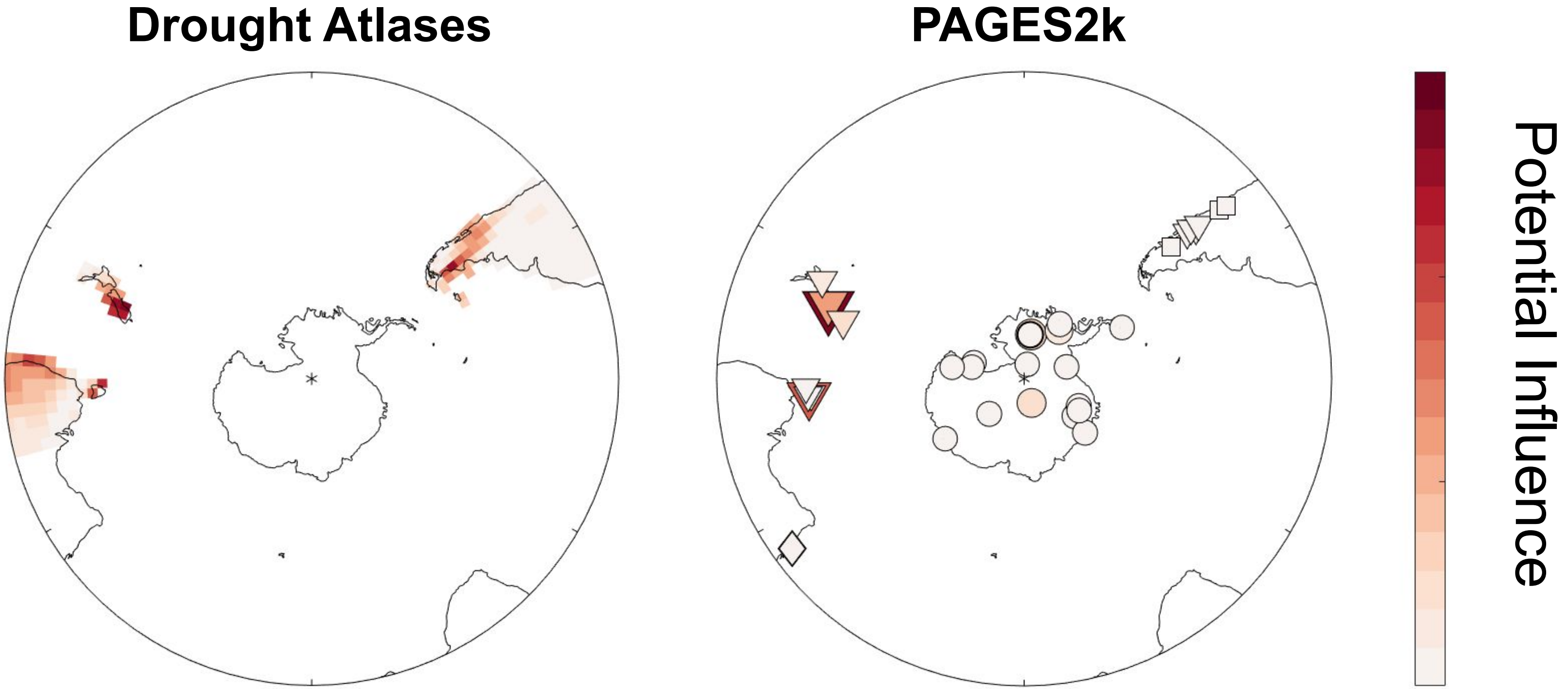


Abram et al., (2014). Figure 2a

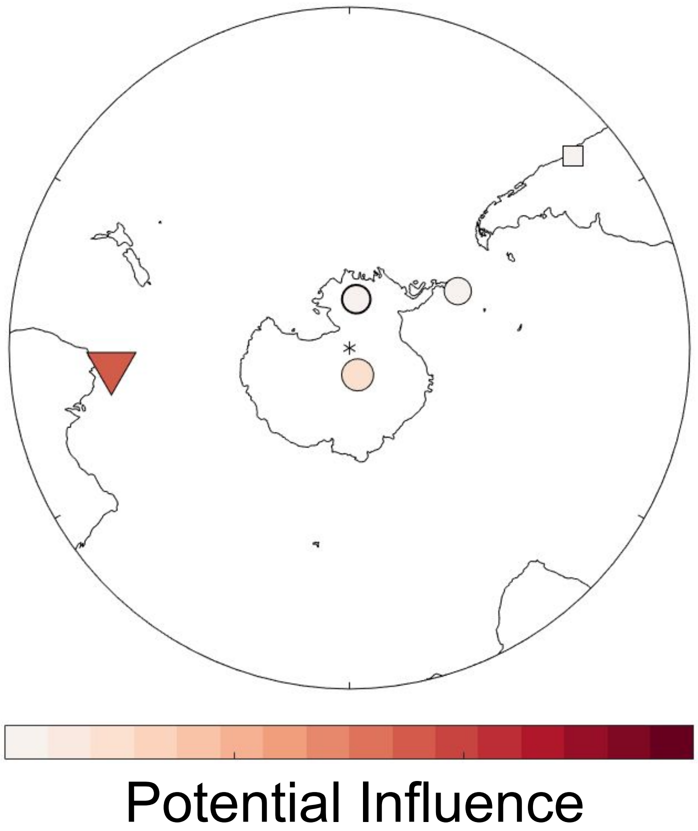


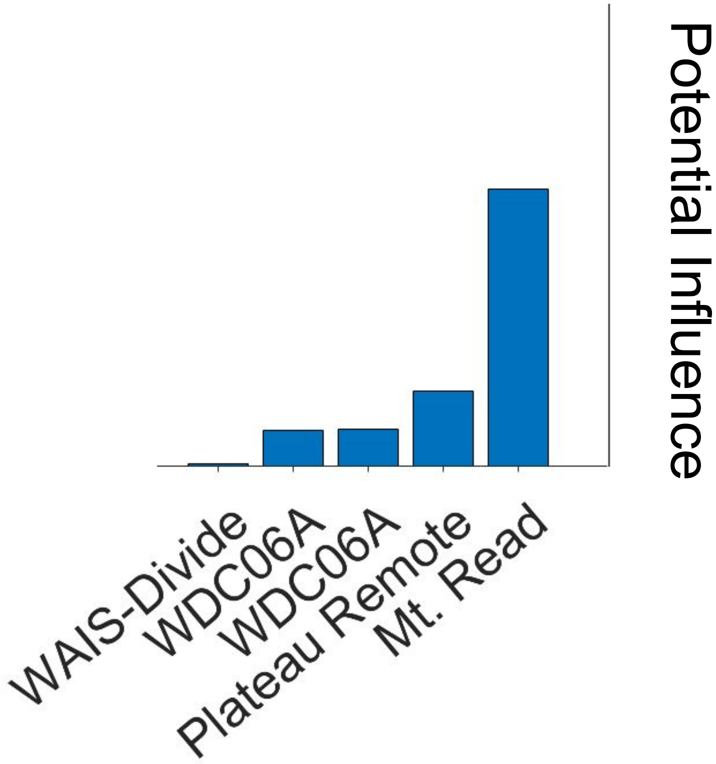
Datwyler et al., (2018). Figure 7b

Optimal Sensor

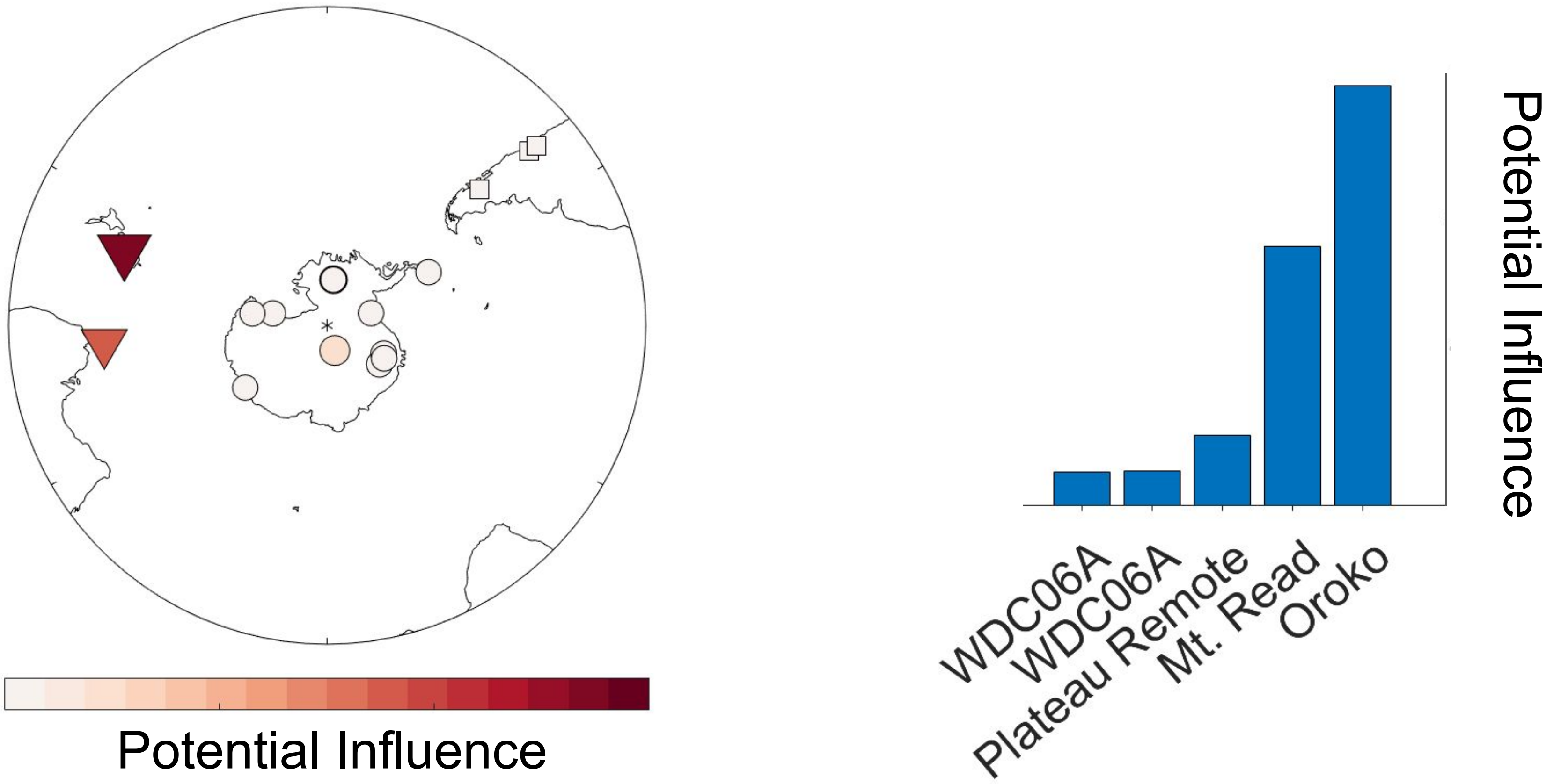


8 - 165 CE


Potential Influence

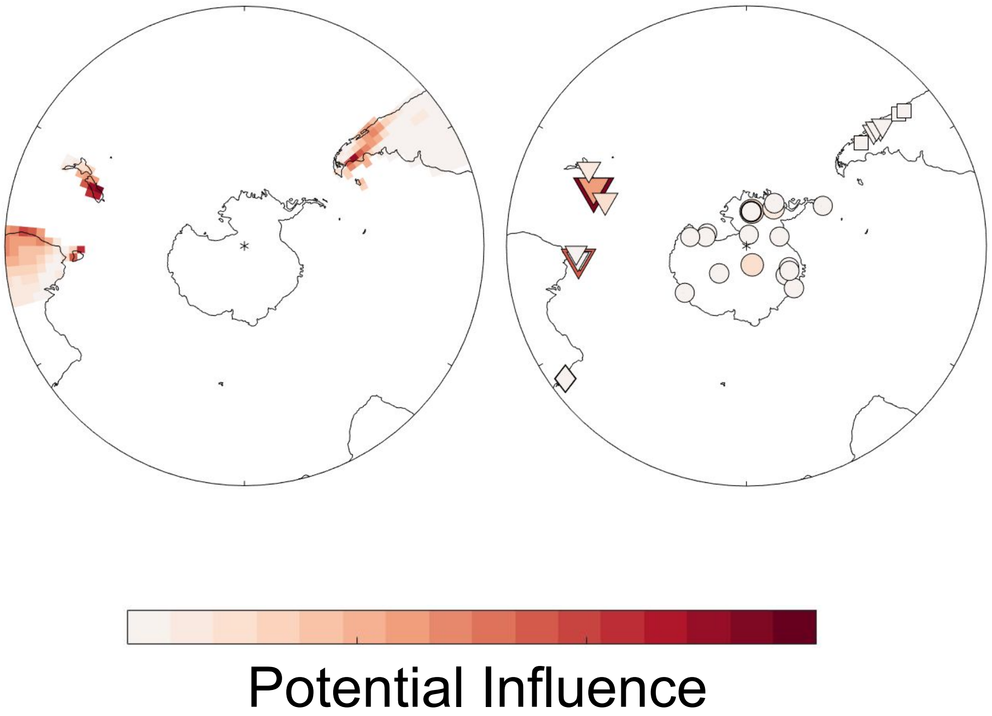

Potential Influence
WAIS-Divide
WDC06A
WDC06A
Plateau Remote
Mt. Read

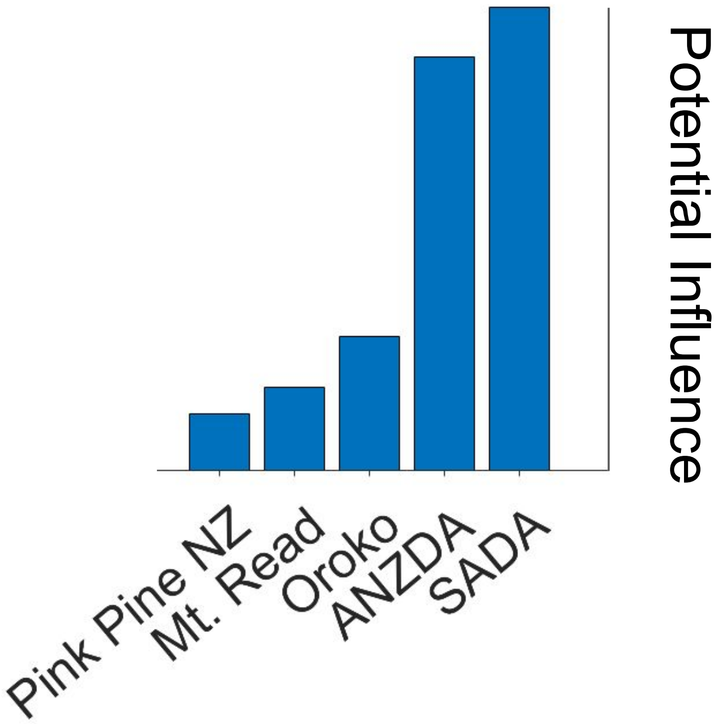
1232 - 1399 CE



Potential Influence
WDC06A
WDC06A
Plateau Remote
Mt. Read
Oroko
Potential Influence

1848 - 1983 CE


Potential Influence


Pink Pine NZ
Mt. Read
Oroko
ANZDA
SADA
Potential Influence

New Reconstruction

A. Data Assimilation

1. Method: Offline ensemble Kalman filter
2. Prior: Stationary multi-model
3. Proxy system models: Linear seasonal temperature, $PDSI_{Thornthwaite}$

B. Advantages

1. Not calibrated to SAM index
2. More proxy information (drought atlases)
3. Assess proxy influence (optimal sensor)

Context

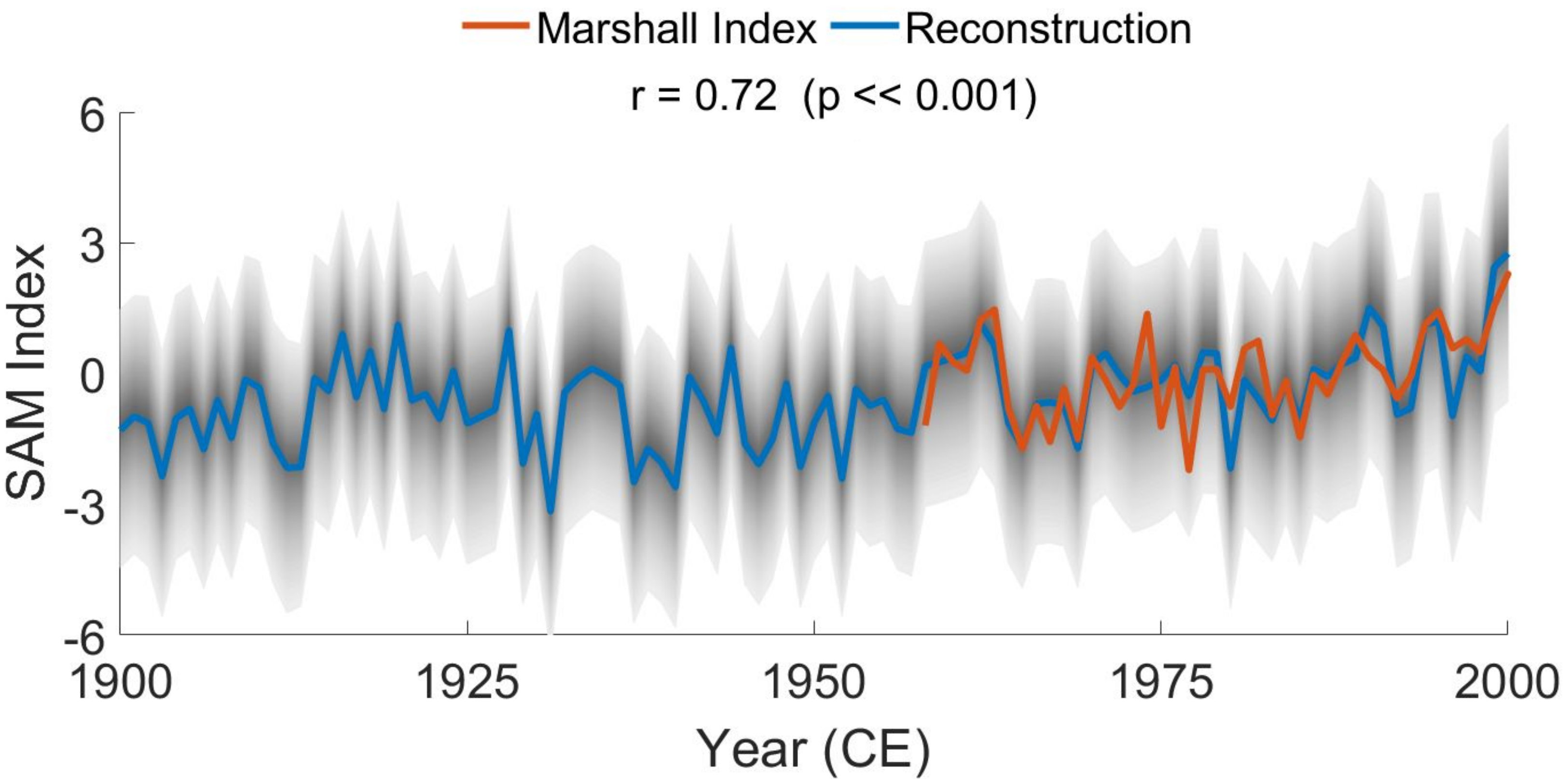
Data Assimilation Method

Results

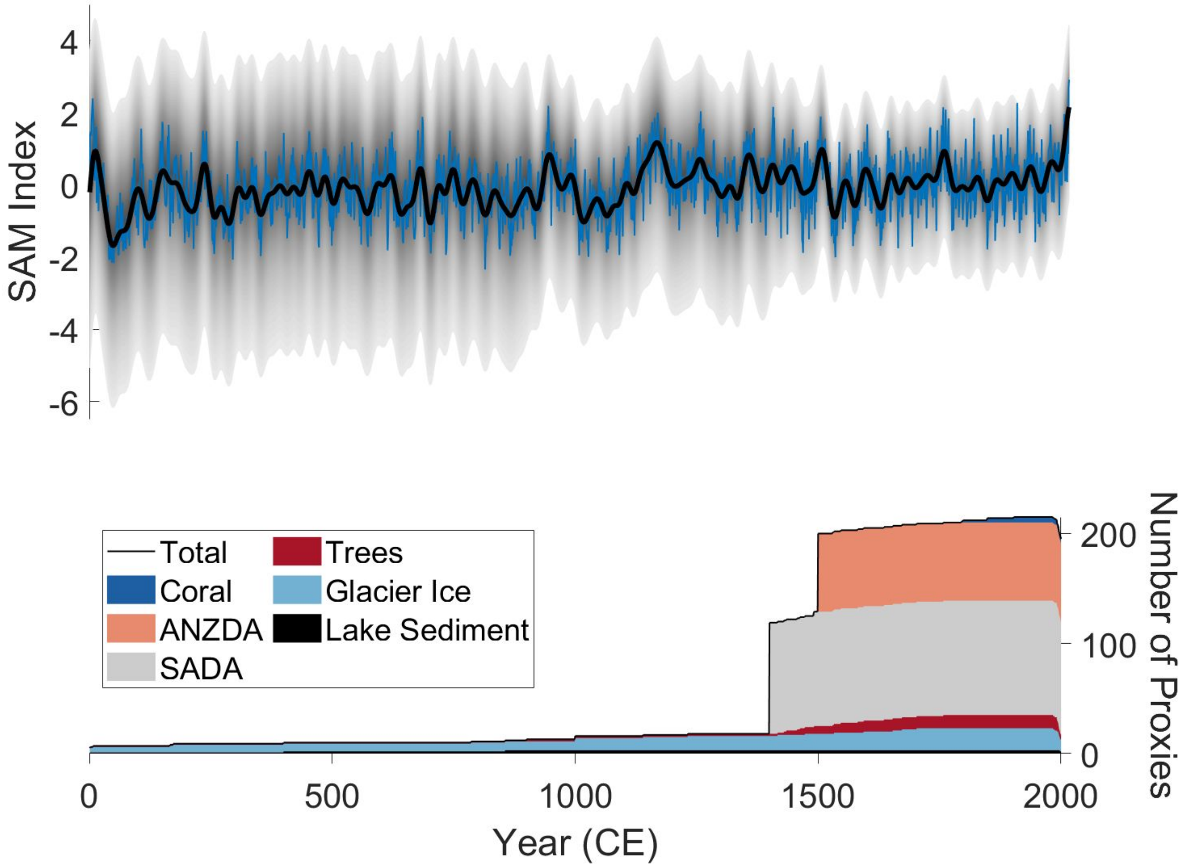
Results

A. New reconstruction

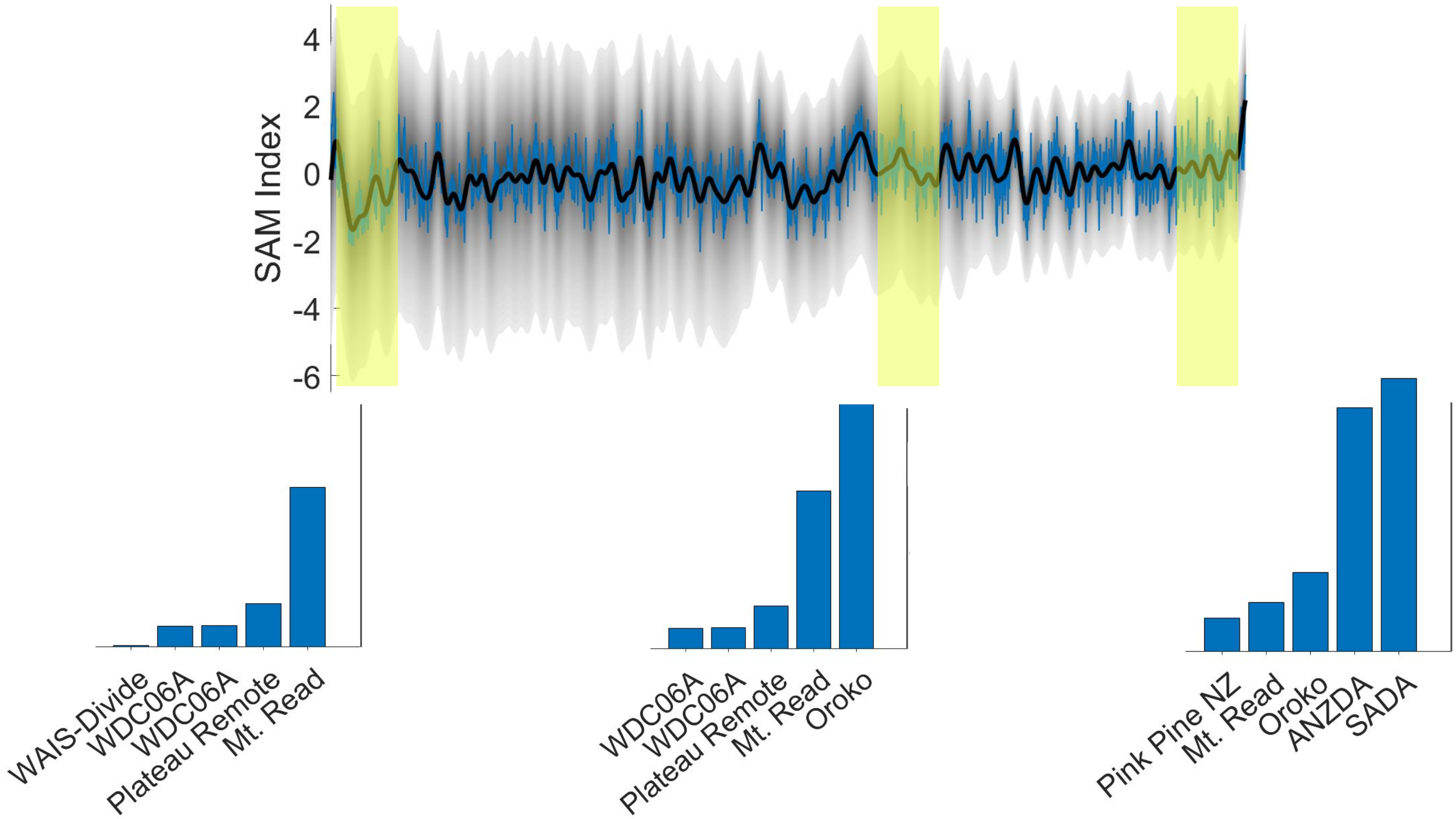
Reconstruction



Reconstruction



Reconstruction



Results

A. New reconstruction

- Correlation with Marshall Index: 0.72 ($p \ll 0.001$)
- Quantified influential proxies

Results

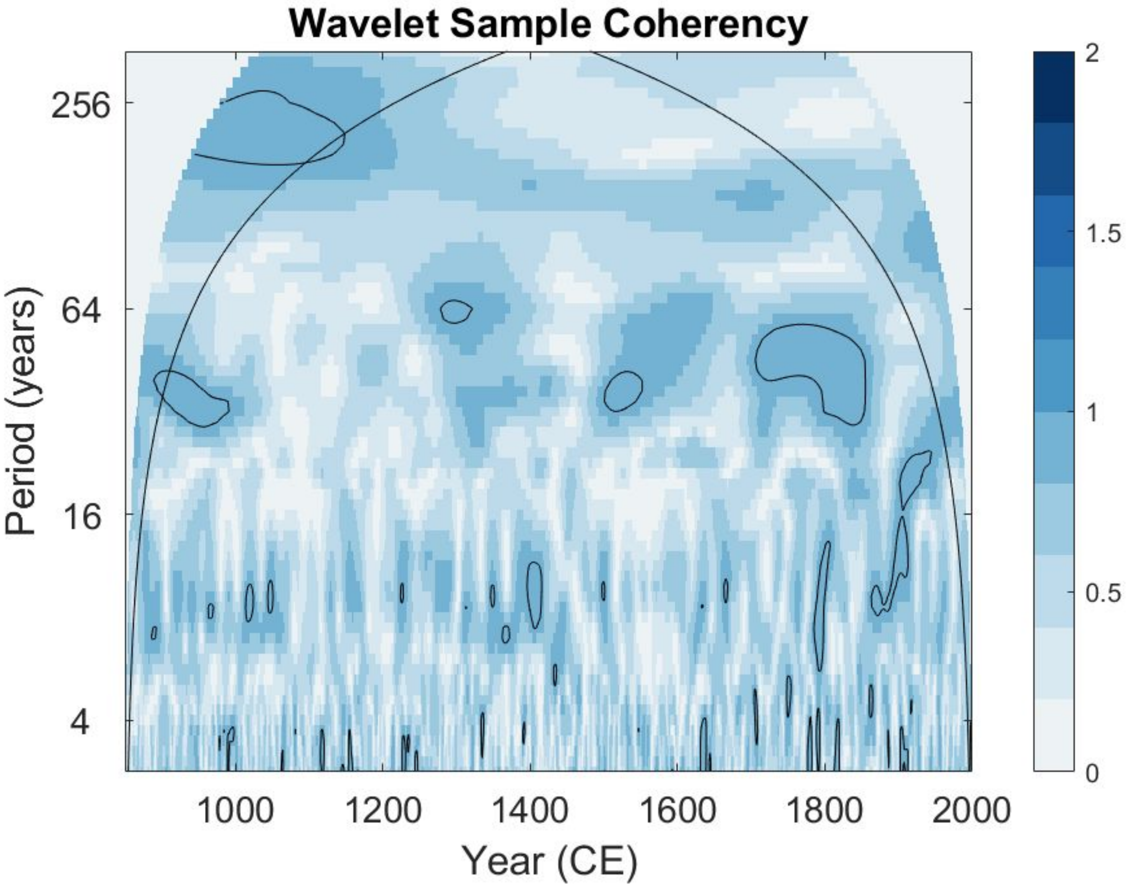
A. New reconstruction

- Correlation with Marshall Index: 0.72 ($p \ll 0.001$)
- Quantified influential proxies

B. Long-term drivers of the SAM

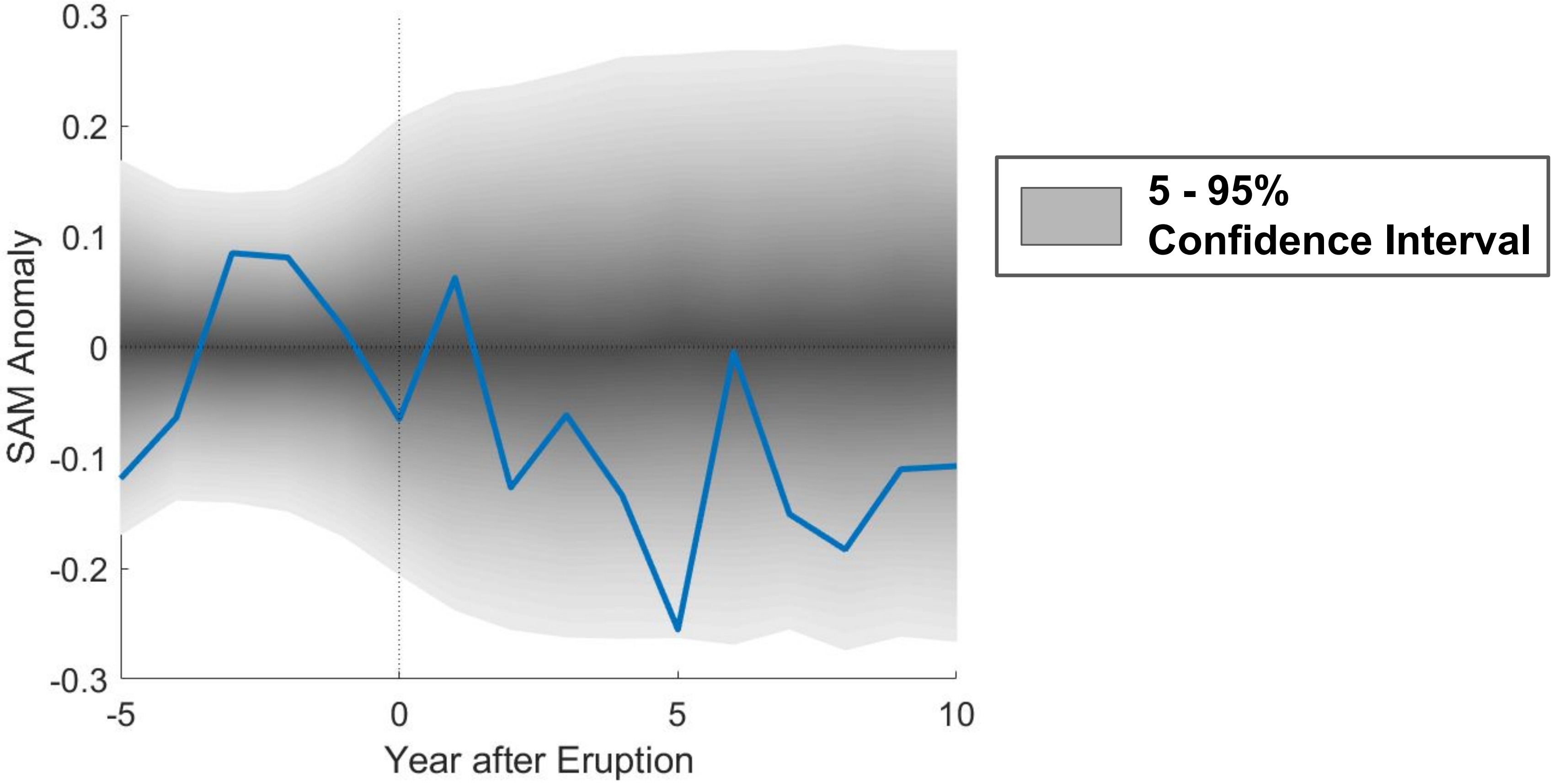
- Internal variability

Solar



Wavelet Sample Coherency
256
64
16
4
Period (years)
1000
1200
1400
1600
1800
2000
Year (CE)
2
1.5
1
0.5
0

Volcanic



0.3
0.2
0.1
0
-0.1
-0.2
-0.3
SAM Anomaly
-5
0
5
10
Year after Eruption
5 - 95%
Confidence Interval

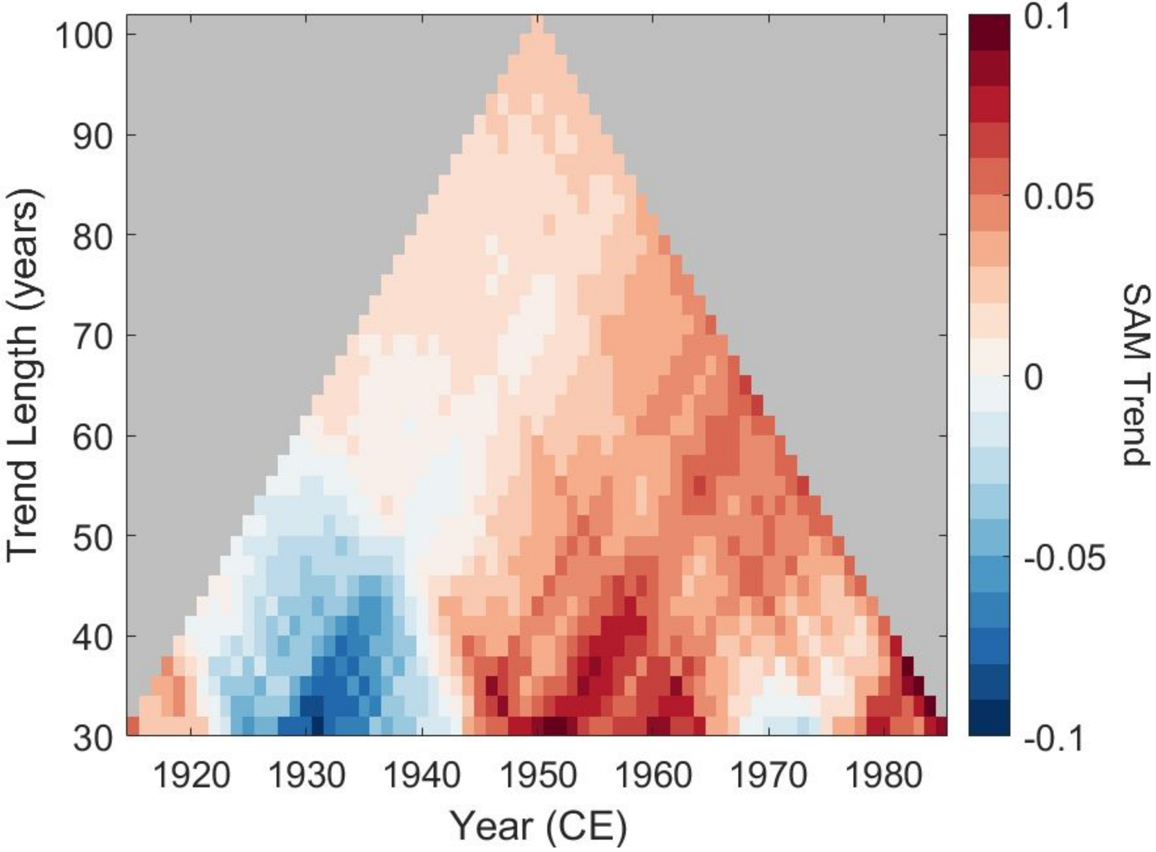
Results

A. **New reconstruction**
   - Correlation with Marshall Index: 0.72 ($p \ll 0.001$)
   - Quantified influential proxies

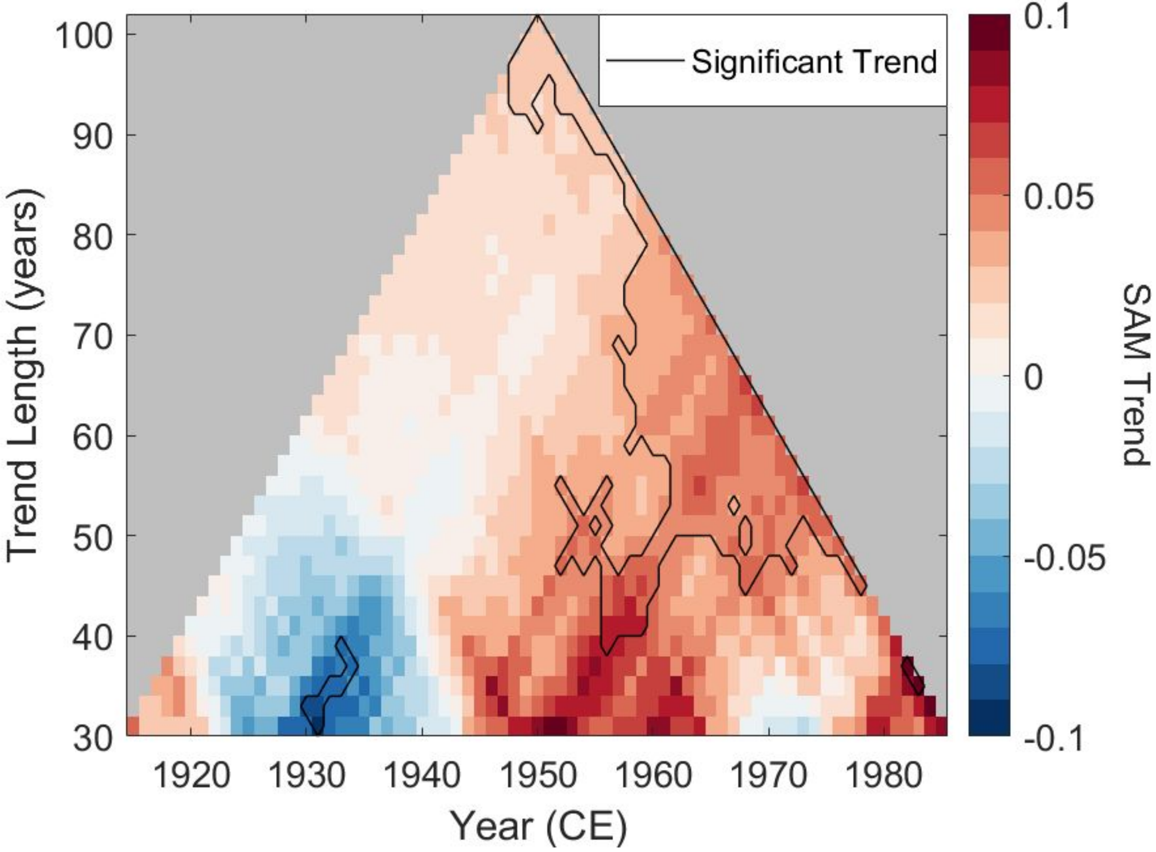
B. **Long-term drivers of the SAM**
   - Internal variability

C. **Modern Trend**

Modern Trends



Modern Trends



Results

A. **New reconstruction**
   - Correlation with Marshall Index: 0.72 ($p \ll 0.001$)
   - Quantified influential proxies

B. **Long-term drivers of the SAM**
   - Internal variability

C. **Modern Trend**
   - Outside range of natural variability (5% - 95%)
   - Significance depends on trend length
   - 45+ years, most recent interval

Caveats and Future Work

A. **Climate model covariance biases**
   - Multi-model ensemble

B. **Proxy system model biases**
   - Reduces proxy influence
   - Develop more sophisticated models

C. **Equally weighted climate models**
   - Models are not fully independent
   - Develop more accurate weighting scheme

DASH Package

DASH Public

A package for paleoclimate data assimilation workflow.

data-assimilation ensemble-kalman-filter paleoclimate particle-filter optimal-sensor

MATLAB

https://github.com/JonKing93/DASH

Thank you!

A. **Coauthors**
- Kevin Anchukaitis
- Kathy Allen
- Tessa Vance
- Amy Hessl

B. **National Science Foundation**
- P2C2 AGS-1803946

References

Abram, N. J., Mulvaney, R., Vimeux, F., Phipps, S. J., Turner, J., & England, M. H. (2014). Evolution of the Southern Annular Mode during the past millennium. *Nature Climate Change*, *4*(7), 564-569.

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