

Observations, Reanalyses and Ice Cores: Early Results from a Synthesis of Antarctic Climate

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The Transantarctic Mountains, A. Huerta, 2003



The “Big” Picture

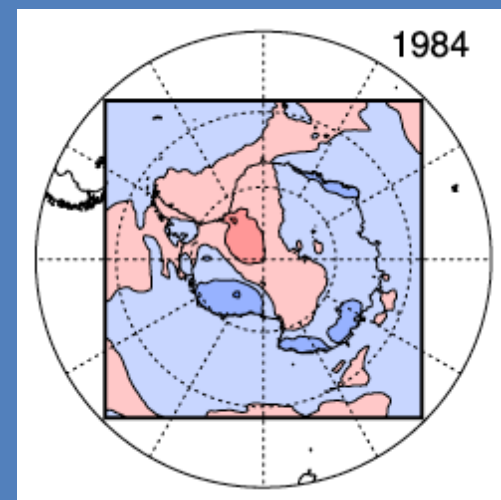
- Create a new picture of (West) Antarctic climate, recent and “paleo”, using:
 - Data
 - Polar MM5 1979-2002 (modeling)
 - READER observations (meteorology)
 - Ice cores (paleoclimate)
 - Artificial neural network (ANN) techniques
- Pilot study underway...

Methods Overview

- Self-organizing maps (SOMs) to summarize variability in meteorology
 - Definitely for model data
 - Possibly for READER observations
- Neural networks to
 - Relate SOM results to ice cores and build ice core-based meteorological reconstructions
 - Possibly fill gaps in observational records

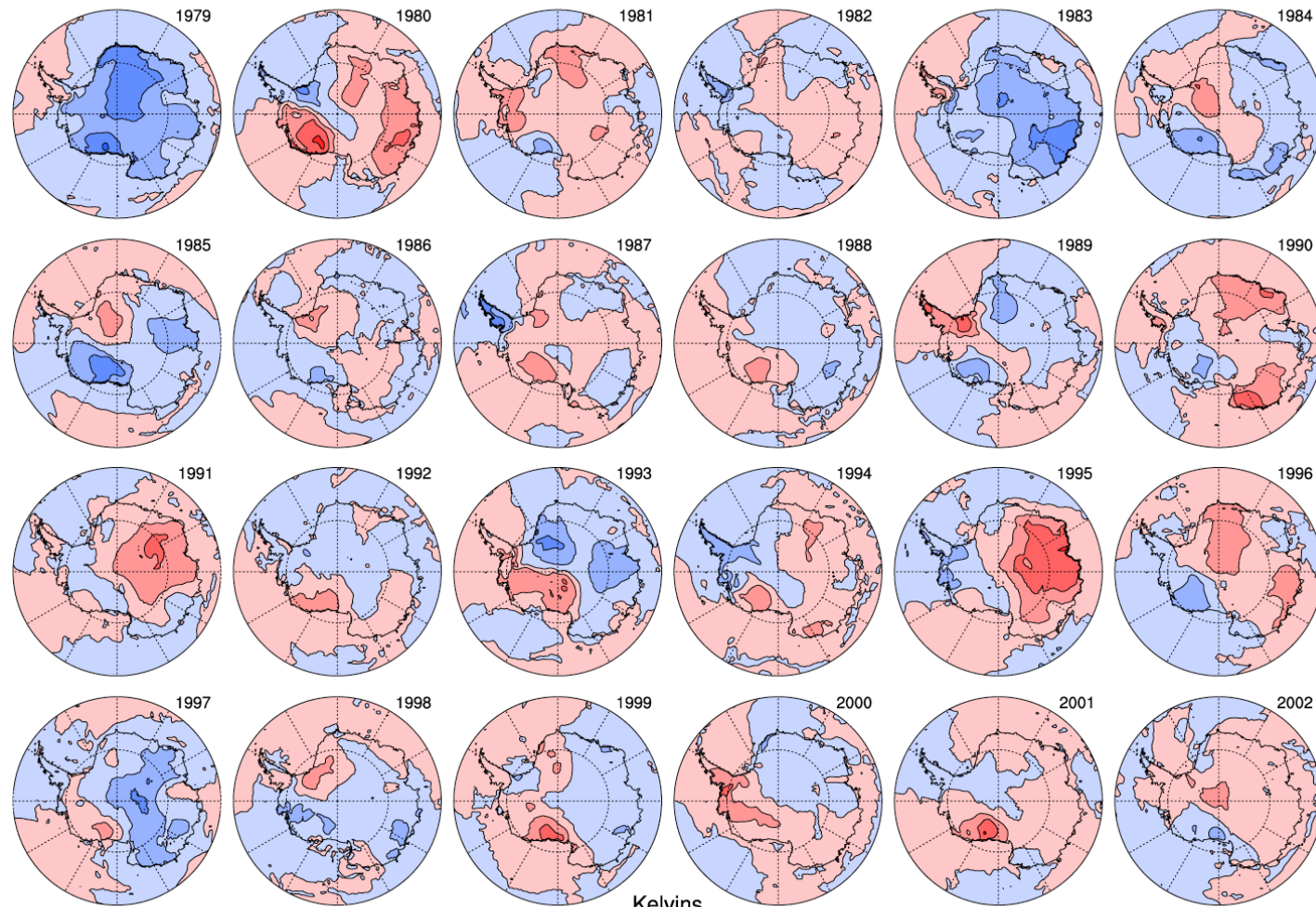
The Model Dataset

- Andy Monaghan's ~24-year Polar MM5
 - Externally driven by ERA-40, 6-hourly
 - Jan 1979 - Aug 2002
 - 60 km grid
- Pilot Study
 - Daily July T-2m (744 days)
 - Standard SOM-based analyses
 - Relate to READER observations



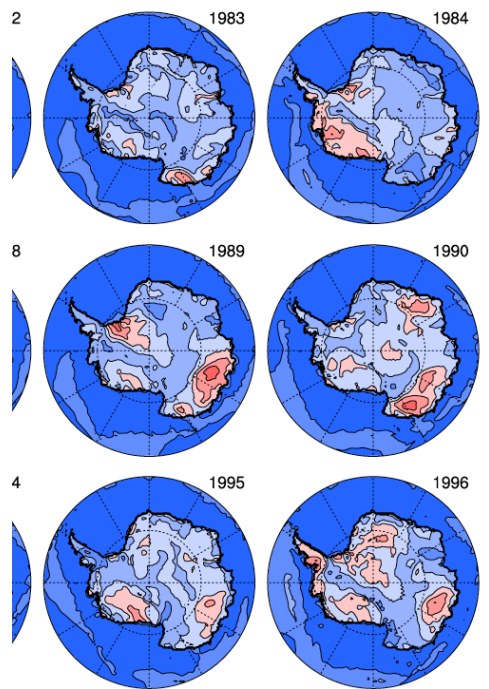
Grid Domain

Monthly Mean T2m, July 1979-2002 (Gridpoint Anomalies)

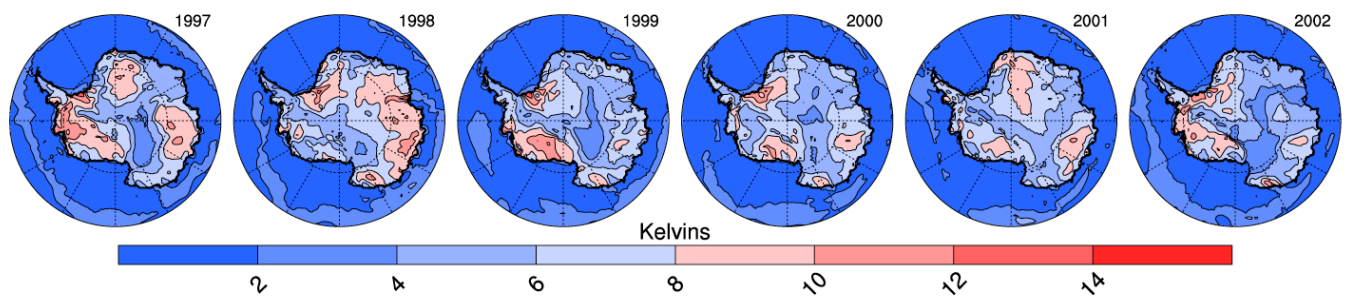


← Monthly Mean

1979-2002



Monthly Standard Deviation →



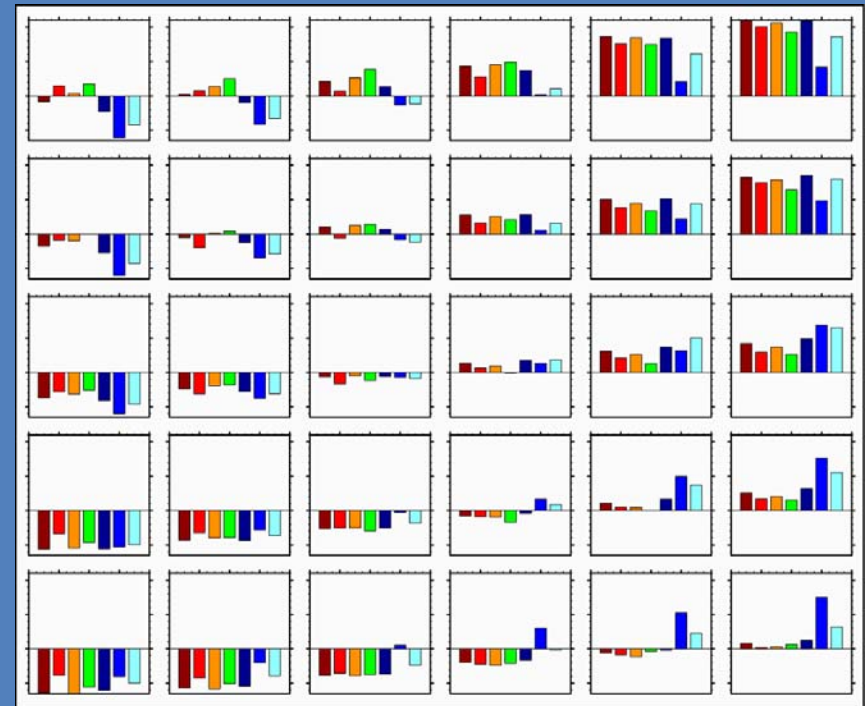
Self-organizing Maps (SOMs)

- 1) Concise summary of data variability expressed as a user-defined number of generalized patterns

Patterns arranged in a grid by their relative similarity

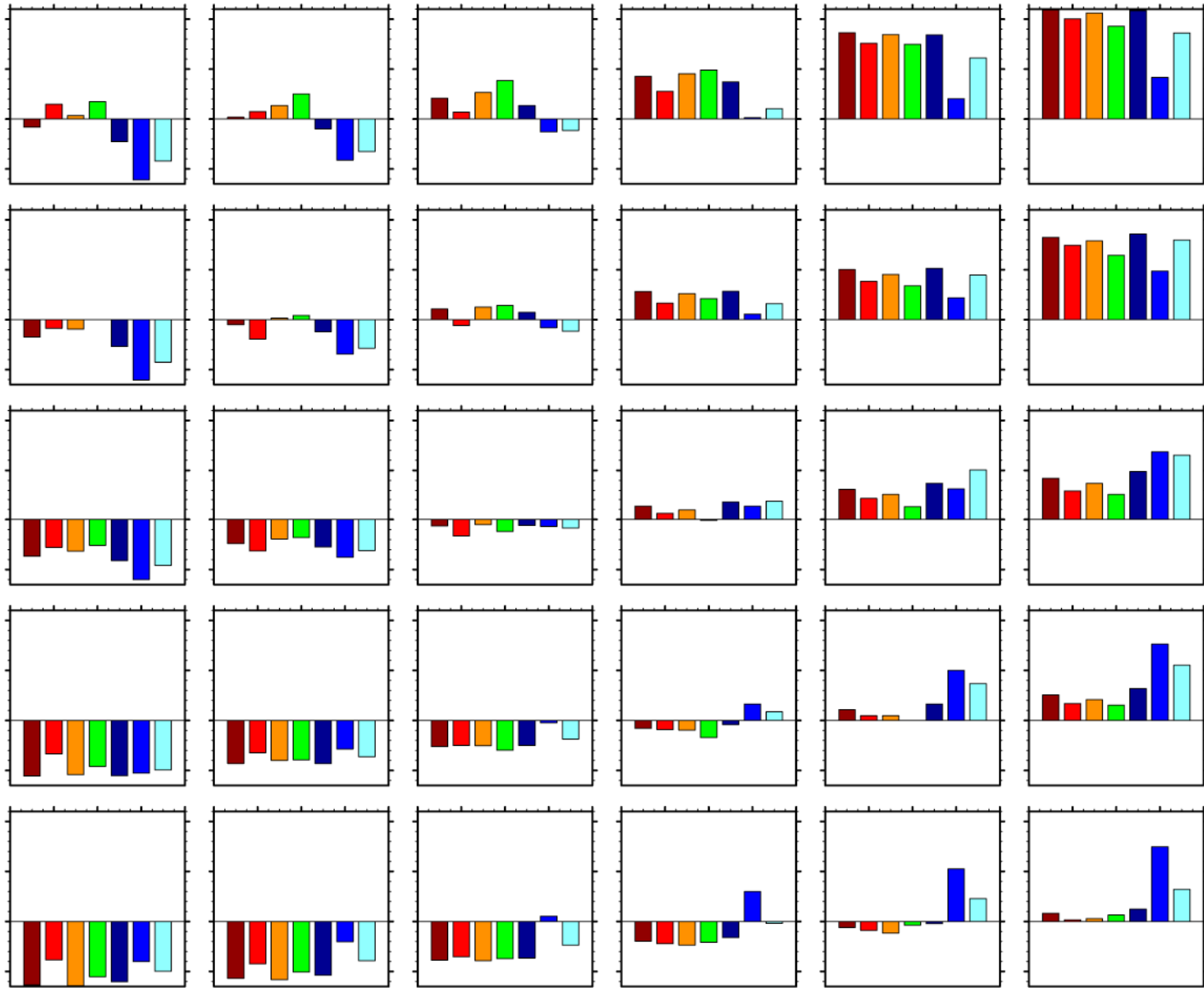
Patterns tend to change smoothly across rows/columns

A projection (mapping) from the multidimensional input space to the 2-D pattern space



Holocene ice core chemistry

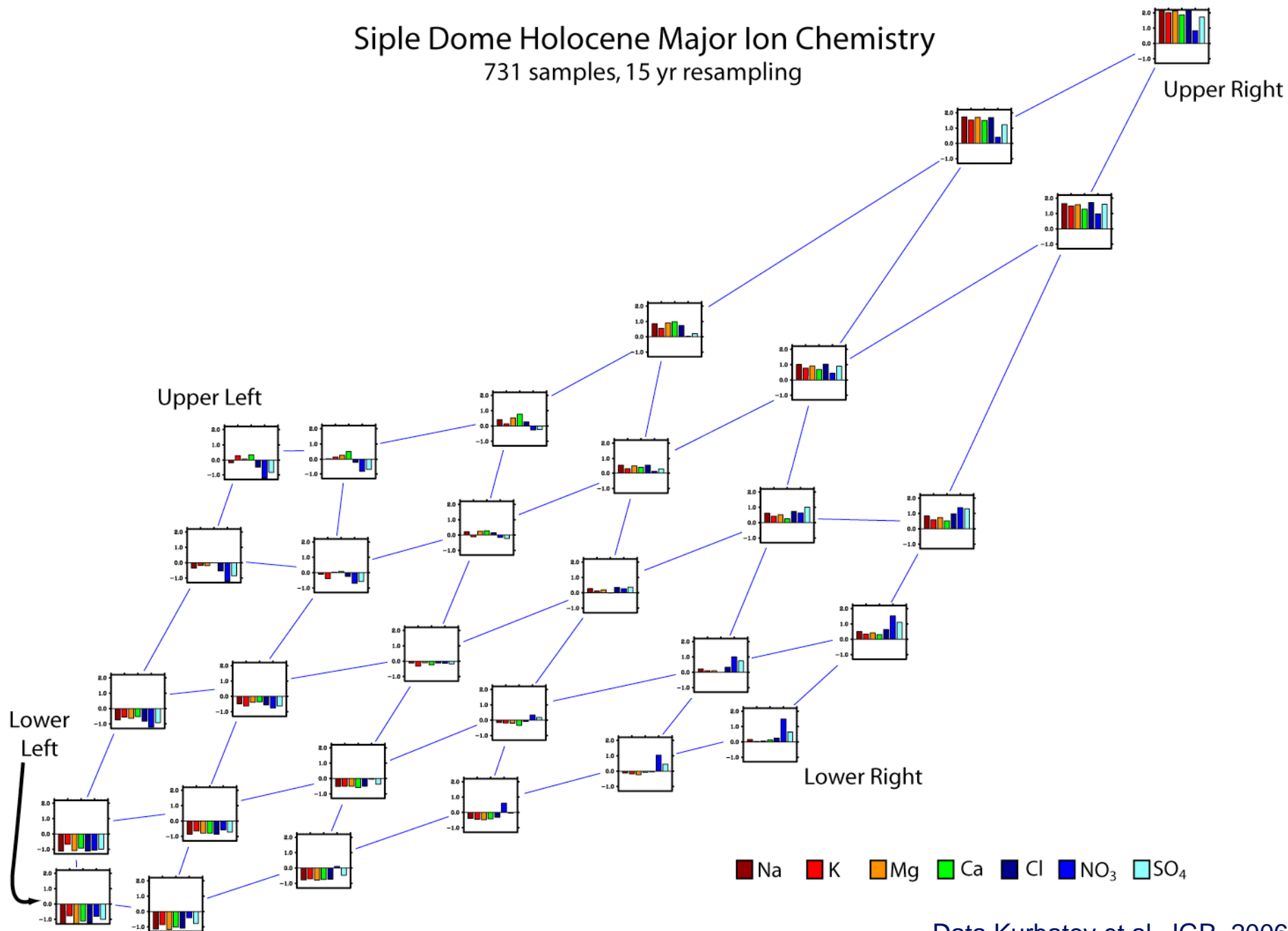
Siple Dome Holocene Major Ion Chemistry



■ Na ■ K ■ Mg ■ Ca ■ Cl ■ NO₃ ■ SO₄

Siple Dome Holocene Major Ion Chemistry

731 samples, 15 yr resampling



Self-organizing Maps (SOMs)

2) Also used for classifying multivariate data and studying its temporal behavior

Each input record matches one pattern most closely

Records matching the same pattern have it in common

Basis for frequency, transition and trajectory maps

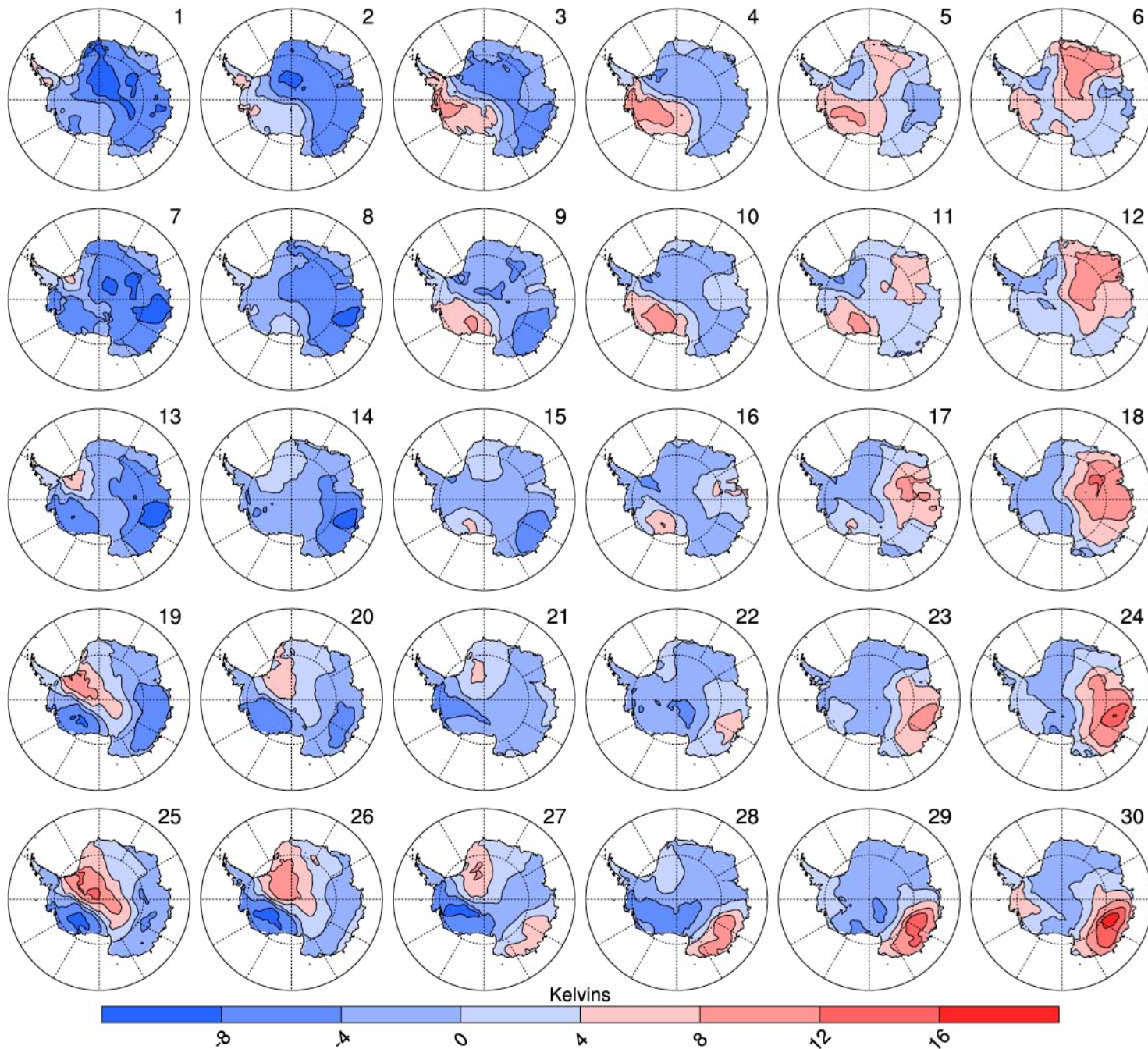
11	6	5	4	2	5
1	5	2	3	2	4
4	5	5	3	5	5
4	6	3	4	3	5
9	5	5	3	6	5

A Frequency Map

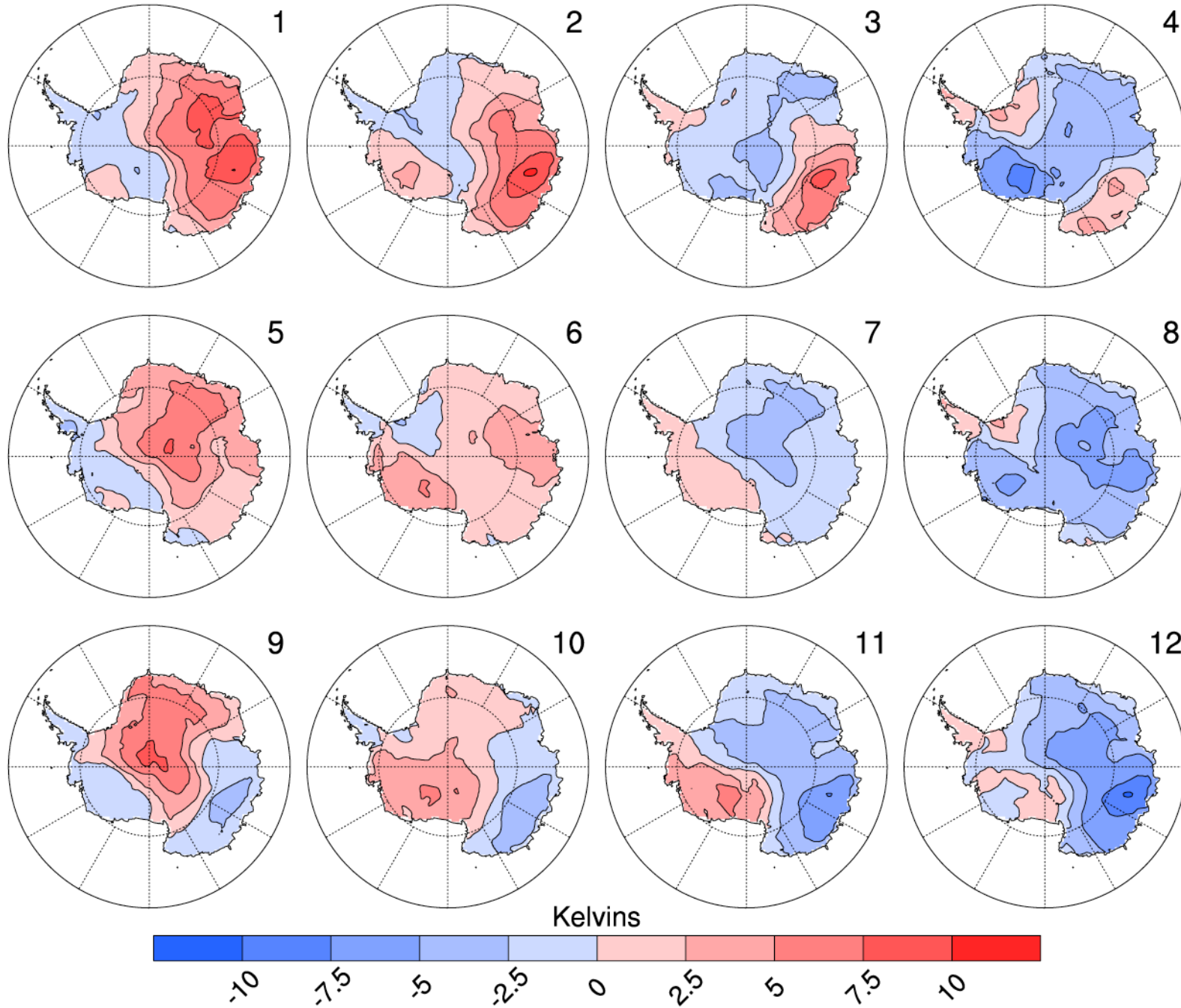
Early Results

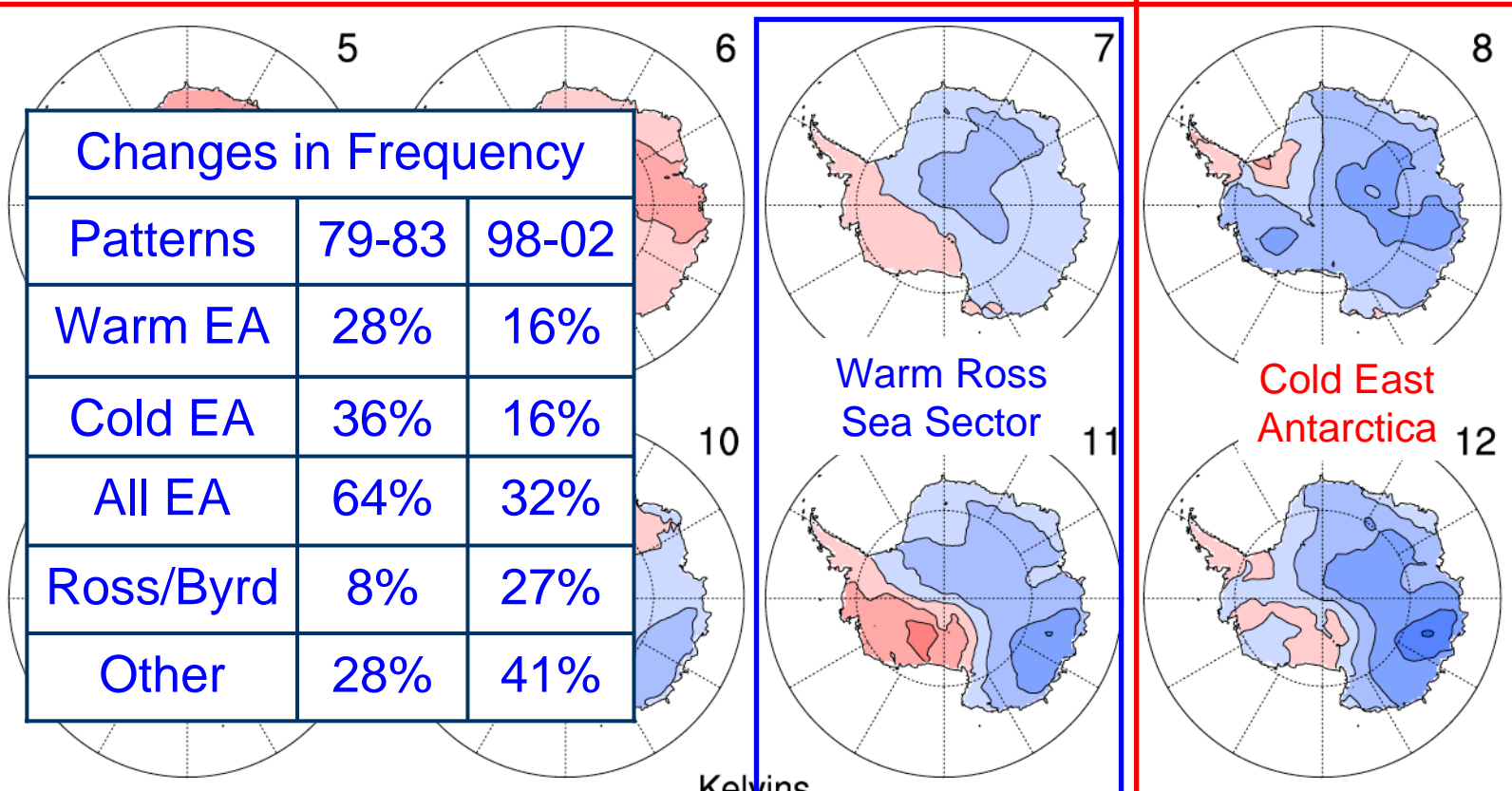
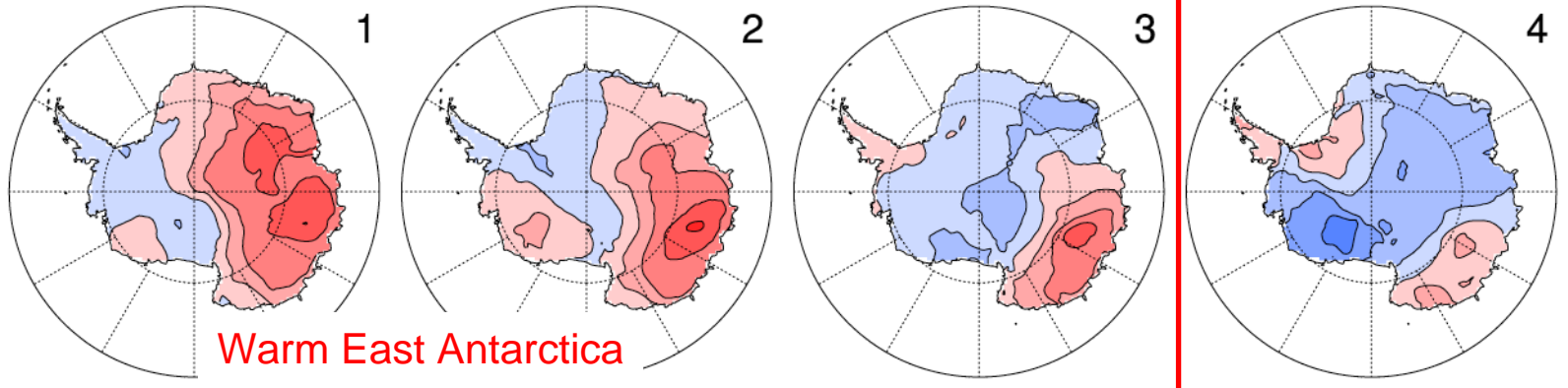
- Patterns and Frequency Changes
- A “Preferred” Transition Path?
- Comparison to READER Observations

6x5 SOM of Daily Temperature (Anomalies)

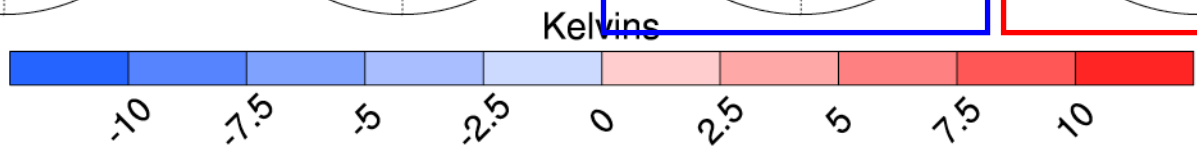


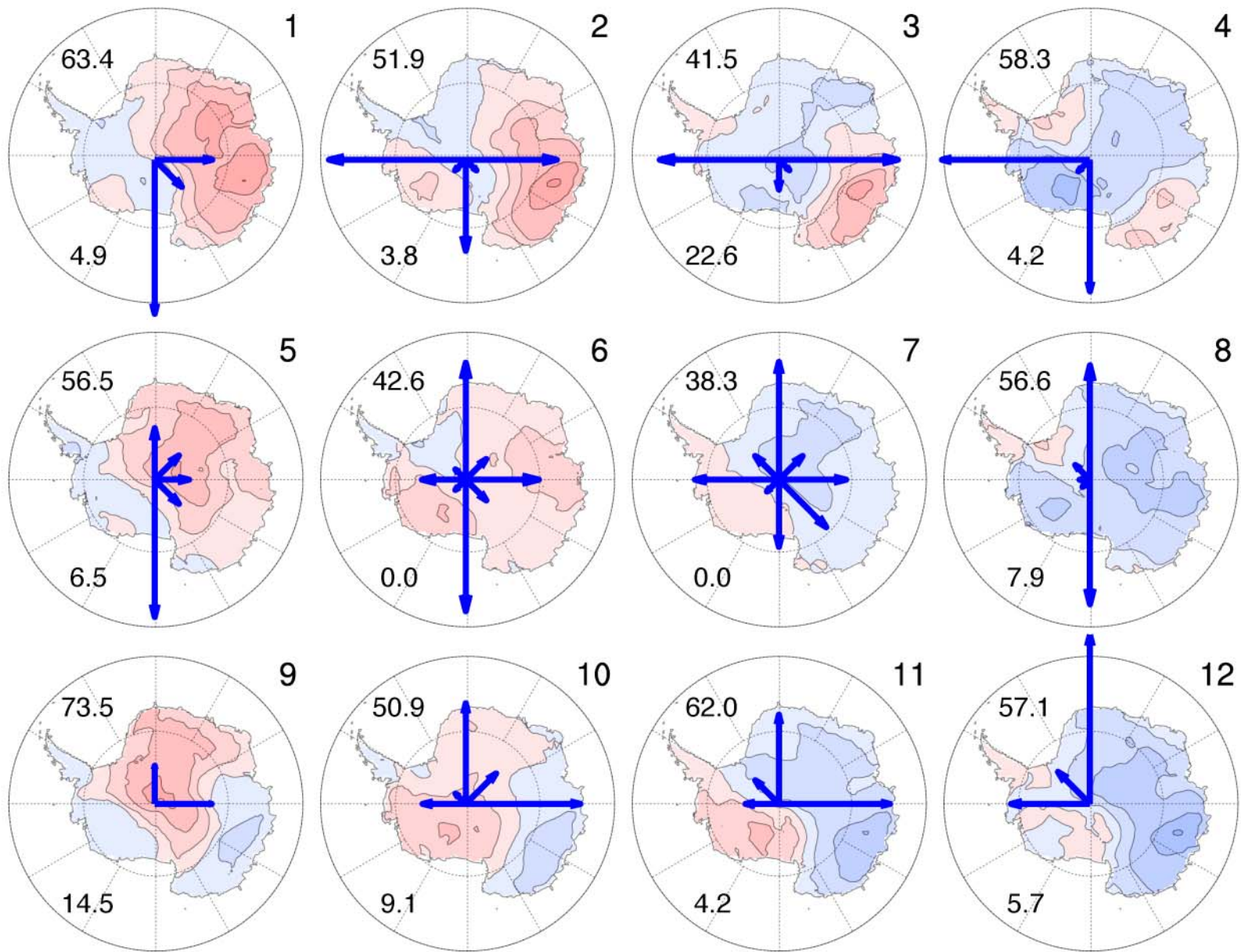
4x3 SOM of Daily Temperature (Anomalies)



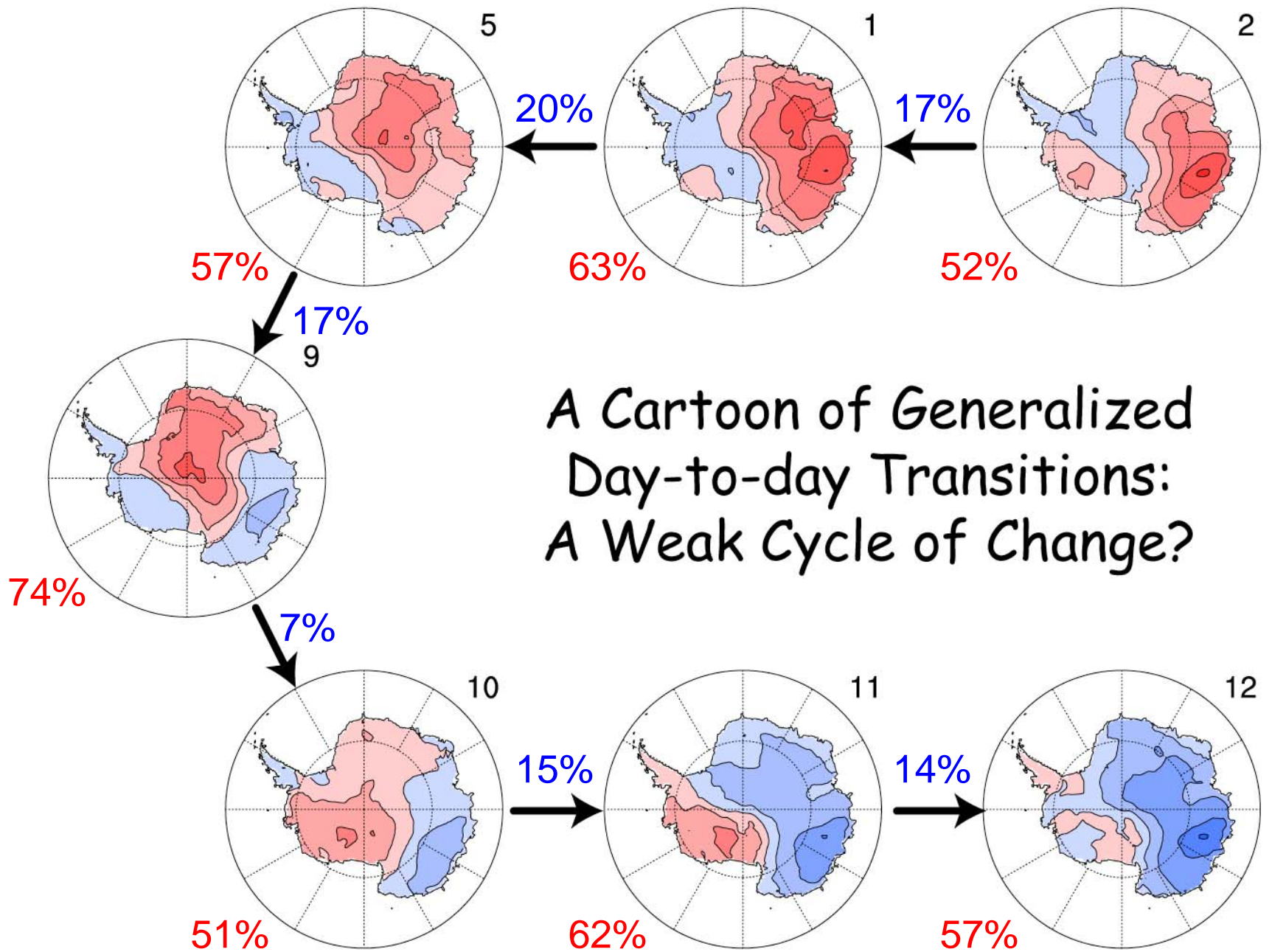


Changes in Frequency		
Patterns	79-83	98-02
Warm EA	28%	16%
Cold EA	36%	16%
All EA	64%	32%
Ross/Byrd	8%	27%
Other	28%	41%



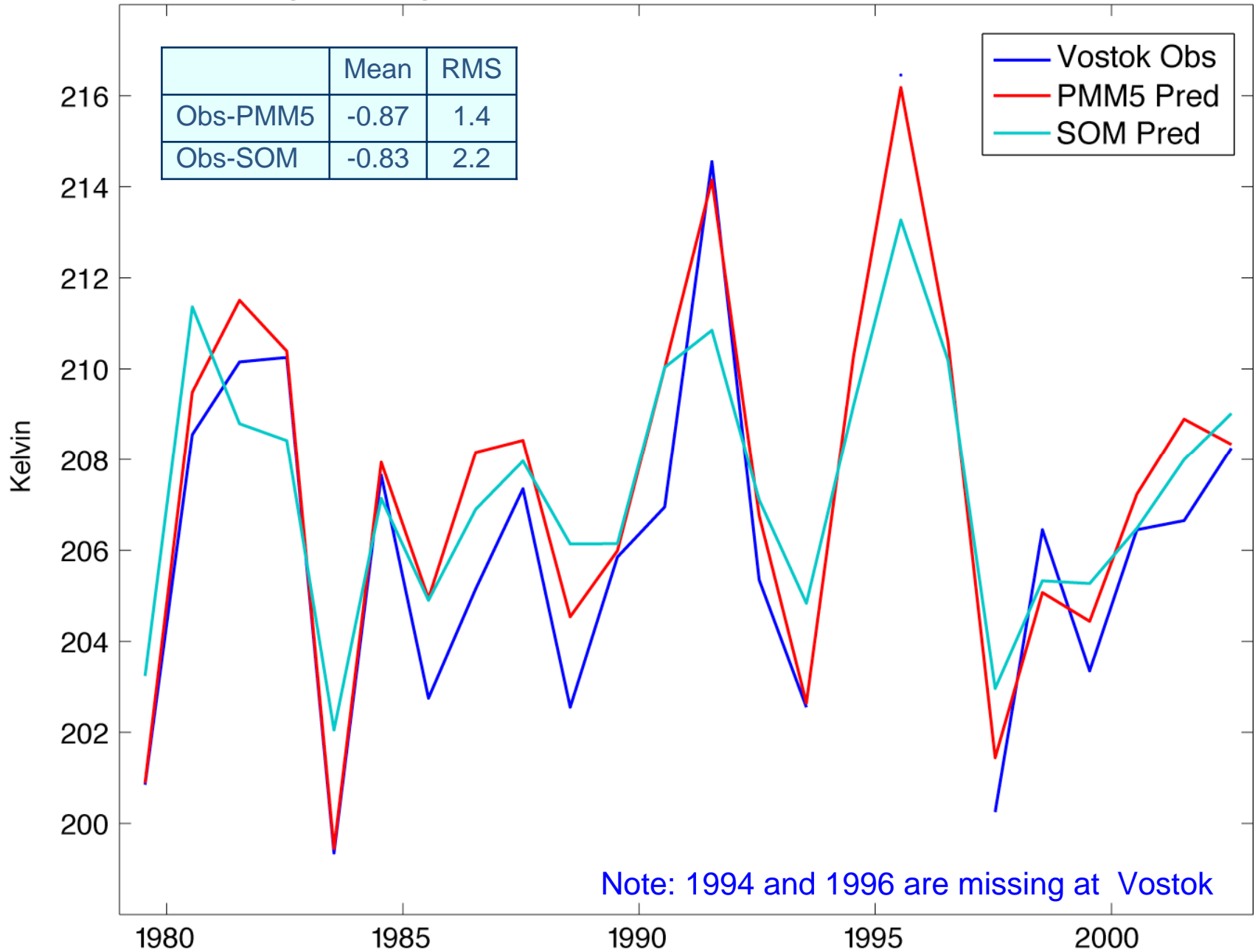


Day-to-day Transitions



A Cartoon of Generalized
Day-to-day Transitions:
A Weak Cycle of Change?

Vostok July Monthly T2m Observations and Predictions: 1979–2002

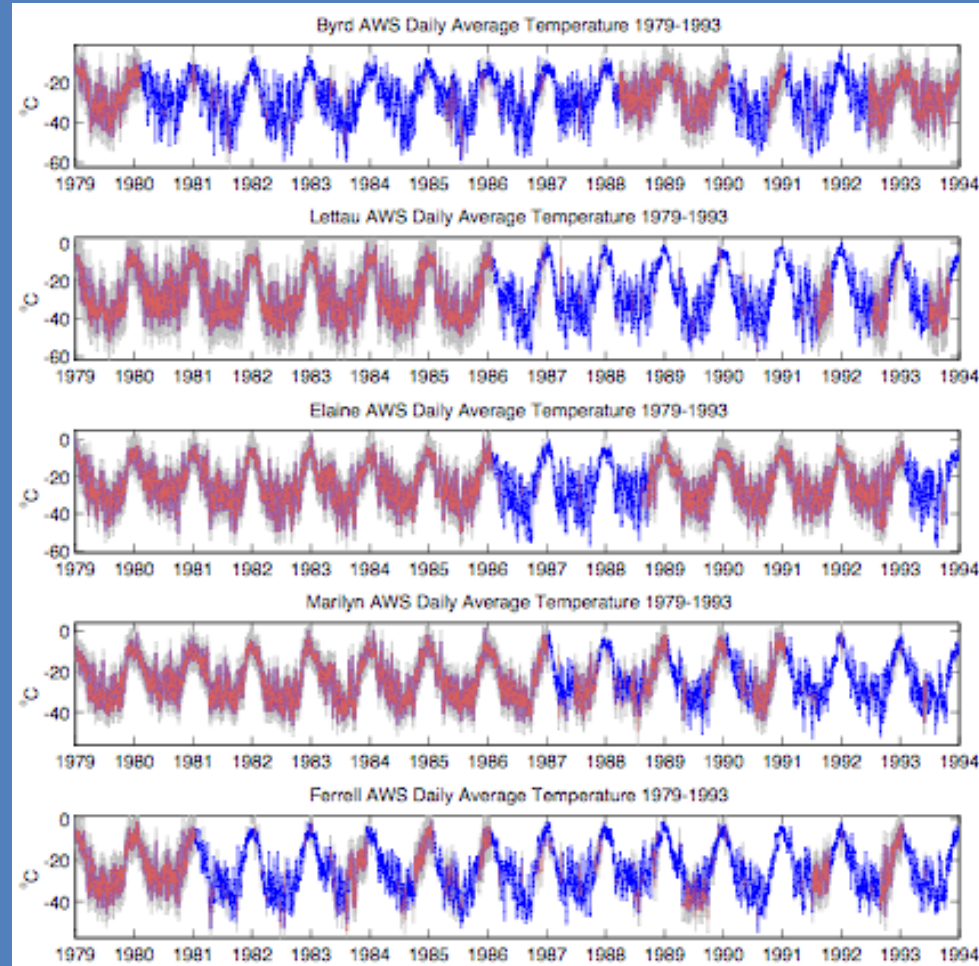


Future Work

- Expand PMM5 analyses to other vars and longer timescales (more “climate”...)
- Explore READER “fill-ins”
- Ice core-based reconstructions

READER “Fill in”

- Train a NN to predict observations using some other data always available (e.g., ERA-40)
- Use trained NN to predict what would have been observed
- Limited to period of the external data

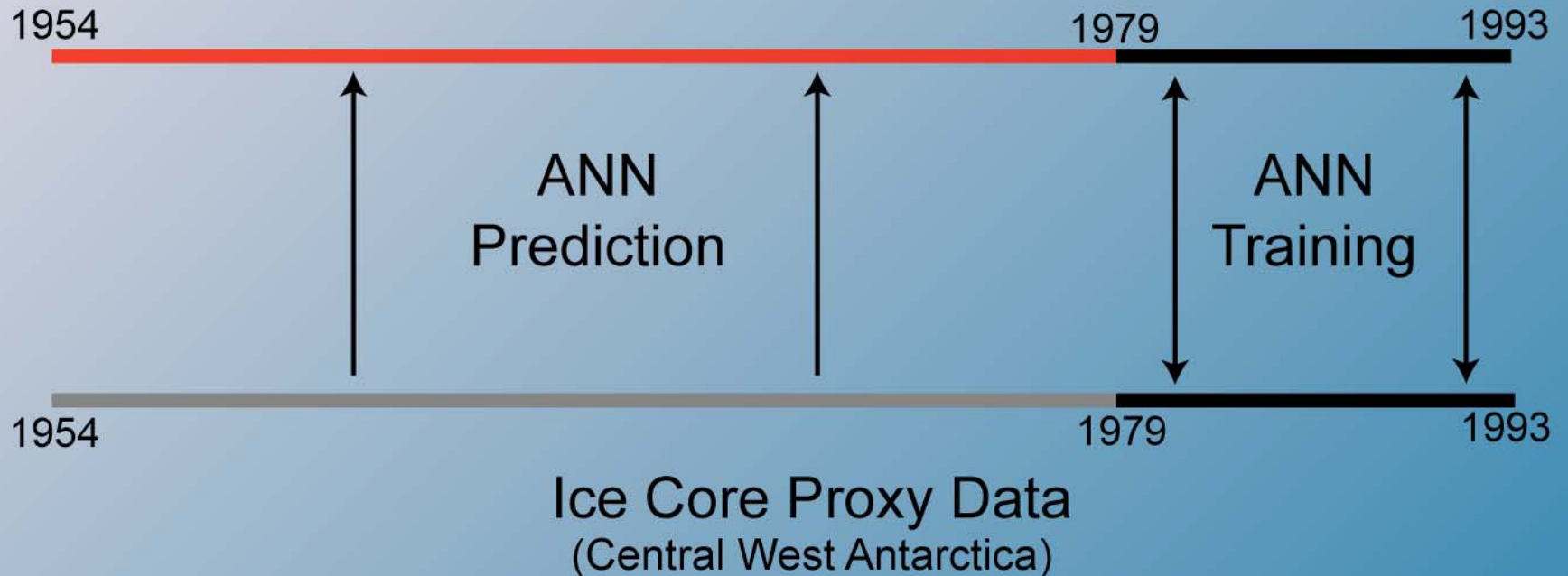


AWS Reconstructions
Reusch and Alley, 2002, 2004

Atmospheric Circulation Data (From ERA-15)

Predicted
SOM Patterns

Analyzed
SOM Patterns



Uniformitarian Extrapolation

Calibration

Conclusions

- Pilot has shown value of SOMs in analysis of PMM5 temperature at daily scale
- One READER comparison looks good
- Still much to do!

