

Optimizing the Antarctic Observing Network: Motivation and Theory

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Sponsor: National Science Foundation through Award # ANT-1043090

Motivation: Why Objective Network Design?

Measurements are often sited subjectively

- Specific needs dictate location
 - E.g. support transportation (airports; bases)
- Often serve other, unintended, purposes
 - E.g. weather forecasting & climate monitoring
 - And we want redundancy and fault tolerance

Works fine when resources are abundant

- Distributed costs across agencies (networks of networks)
- Installation, maintenance, and data distribution costs low

Motivation: Why Objective Network Design?

Objective network design informs on resource allocation

- How do we make the best weather forecasts?
- How do we best monitor climate?
- Can use along with other info, but good to know

Elements of an objective approach

- Performance measures (“metrics”)
 - need to know how to quantify what you want from a network
- Constraints (access, funds, existing stations, etc.)
- A robust algorithm
 - Theory and interpretation in this talk; main results in next talk

Why Antarctica?

1. Surface measurements are sparse
 - access is difficult
 - expensive to install & maintain
2. Impact is potentially large
 - maximum benefit for fixed investment
 - better weather forecasts & climate monitoring



U Wisconsin AMRC/SSEC

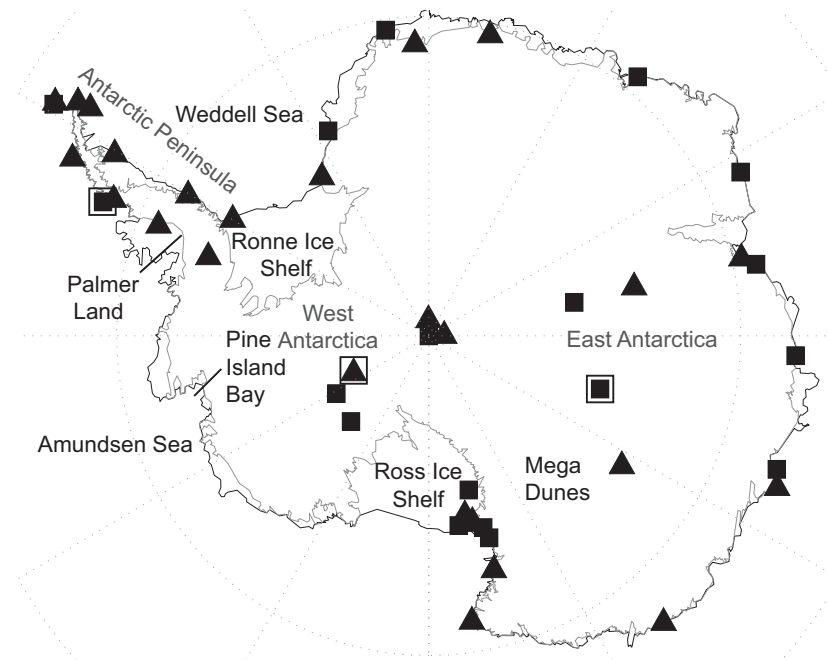
Project goals

- develop techniques to optimize network performance
- test methods on real (OSE) and simulated (OSSE) data
- Recommendations if performance metrics are supplied

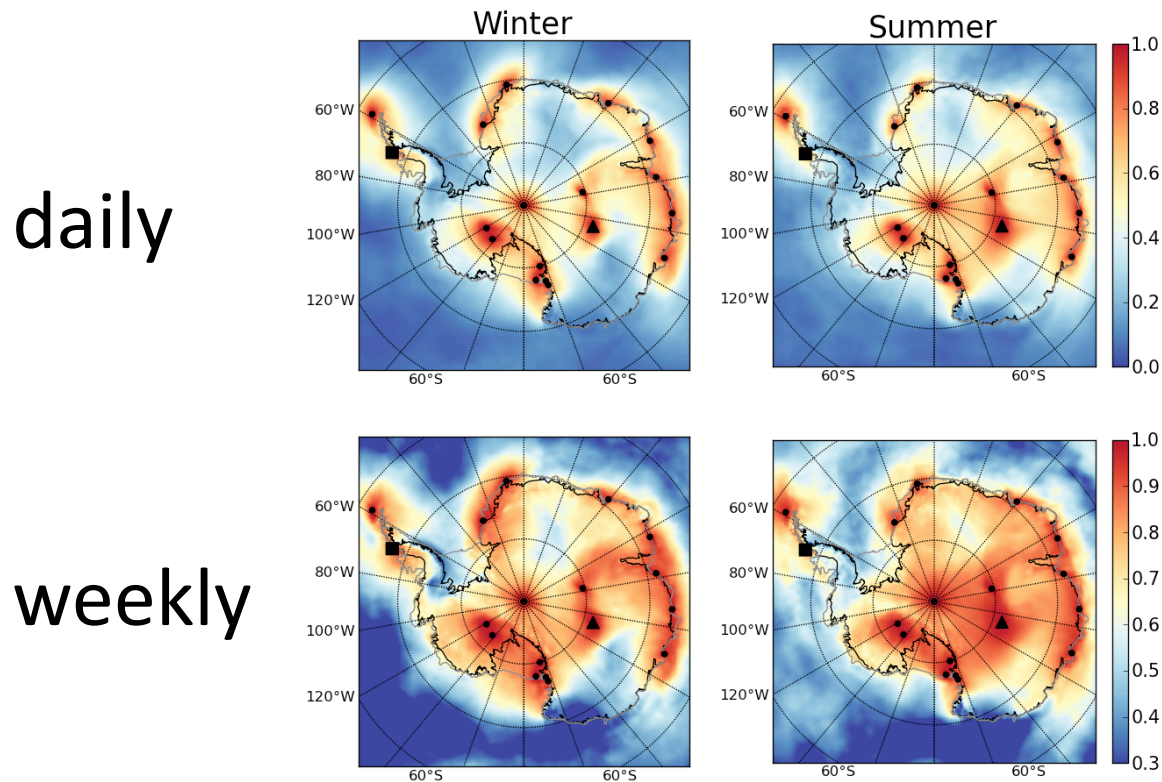
Performance of the Current Network

Bumbaco et al. (2014)

- AWS stations that report regularly
 - 75% and 90% (squares)
 - 2008-2012
- AMPS gridded data
- Two main goals
 - Variance explained
 - Spatial correlations



Performance of the Current Network

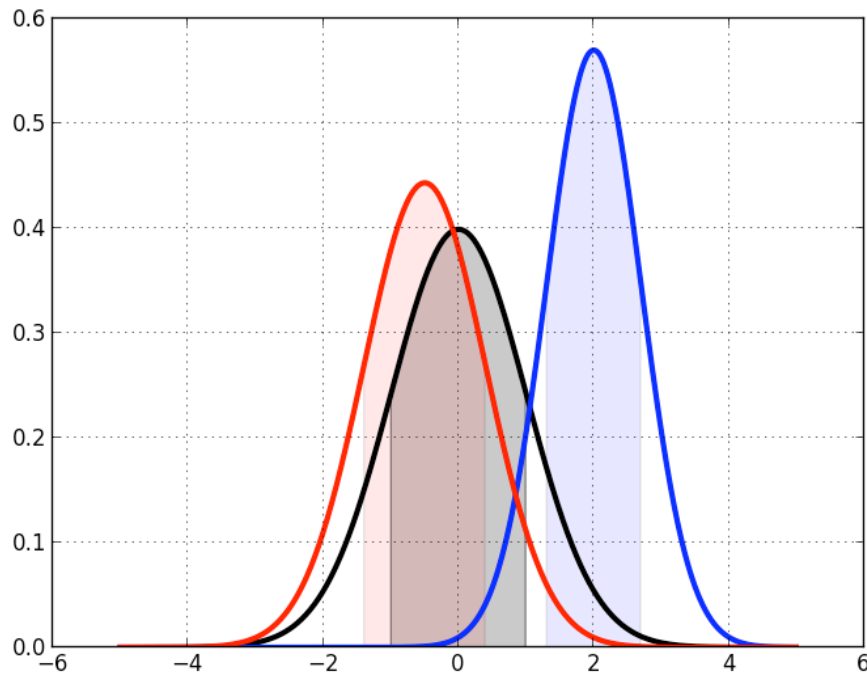


Fraction of temperature variance explained by 90% stations

- Large gaps, even for weekly averages
- Analysis only; doesn't consider forecast performance

How to pick a station location?

Find the point having the largest impact on a performance measure.



Metric distributions for:

- **Reference (e.g. climatology)**
- **Location #1**
- **Location #2**

Pick location #2 because it reduces variance (uncertainty) the most over the reference.

Theory I: Scalar Performance Metric J

Given: statistics for J and the field of interest (x) from a reference source

Approximate changes to J :
$$\delta J \doteq \left[\frac{\partial J}{\partial \mathbf{x}_0} \right]^T \delta \mathbf{x}_0$$

The change in the variance of J is given by:

$$\delta \sigma^2 = \left[\frac{\partial J}{\partial \mathbf{x}_0} \right]^T (\mathbf{B}_{i-1} - \mathbf{B}_i) \left[\frac{\partial J}{\partial \mathbf{x}_0} \right]$$

\mathbf{B}_i is a known function of \mathbf{B}_{i-1} and, **for a single point**, this simplifies to:

$$\delta \sigma^2 = \frac{[\text{cov}(J, x_j)]^2}{\text{var}(x_j) + r_i^2}$$

The Algorithm

$$\delta\sigma^2 = \frac{[\text{cov}(J, x_j)]^2}{\text{var}(x_j) + r_i^2}$$

1. Evaluate $\delta\sigma^2$ for all potential locations
 - Pick location with largest $\delta\sigma^2$
 - Based on ensemble data drawn from AMPS grids
2. Remove impact of an observation at this location
 - Update the ensemble
3. Go back to 1, and look for the next location
 - Stop when noise dominates (choice no different than random)

Repeat many times (10,000+) to account for uncertainty.

Key to interpretation: spatial correlation length scale

- Some locations have information on large regions

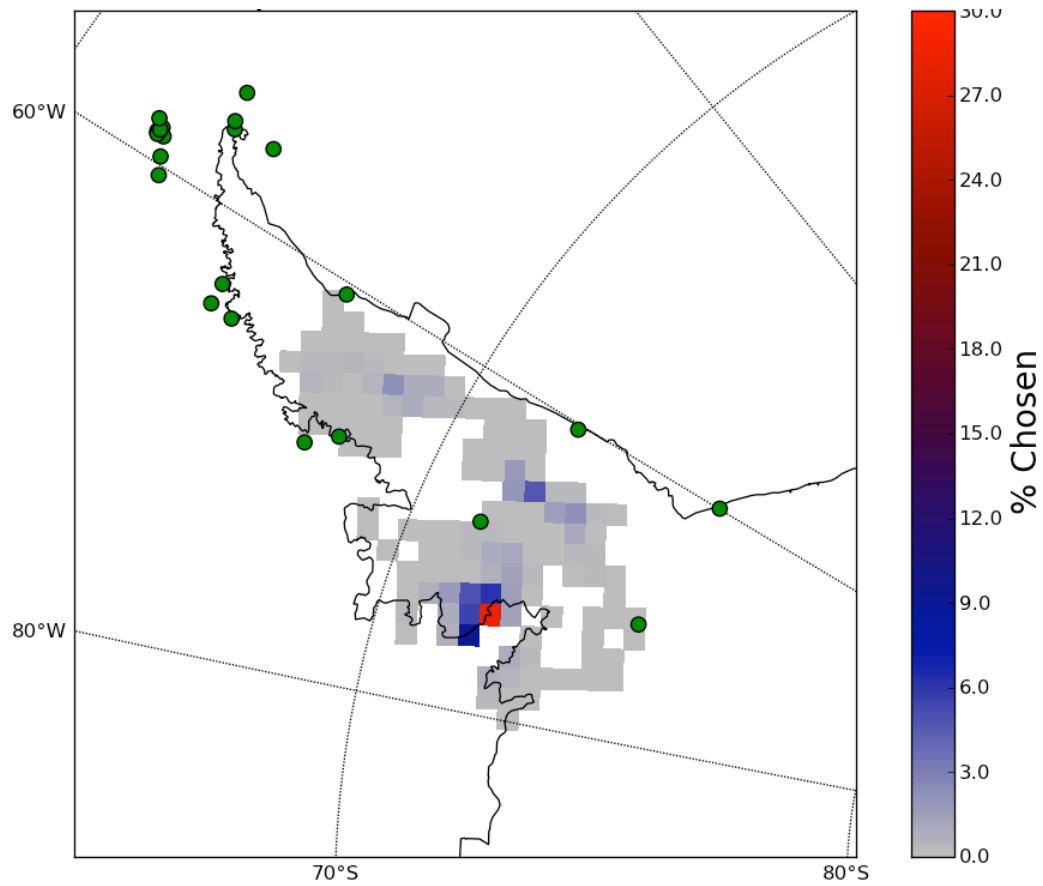
Results for Spatial Averages

J: spatial average temperature for three locations

- Peninsula
- West Antarctica
- East Antarctica

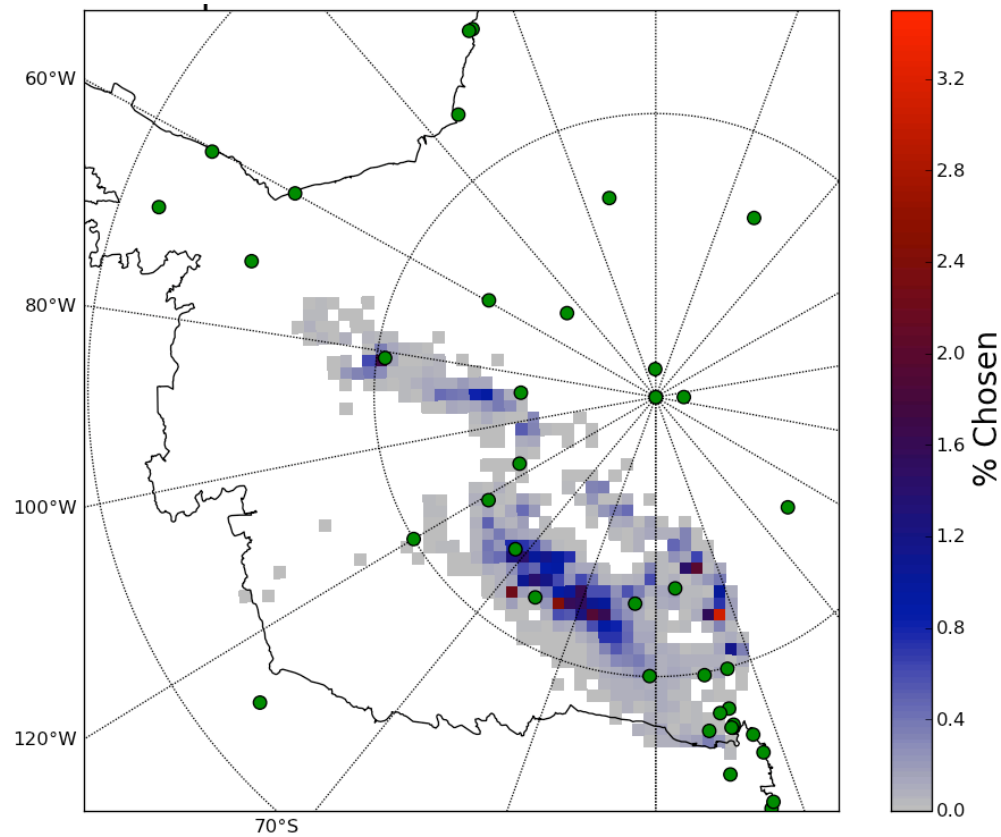
Find first station only (next talk will expand on this)

Peninsula First Station



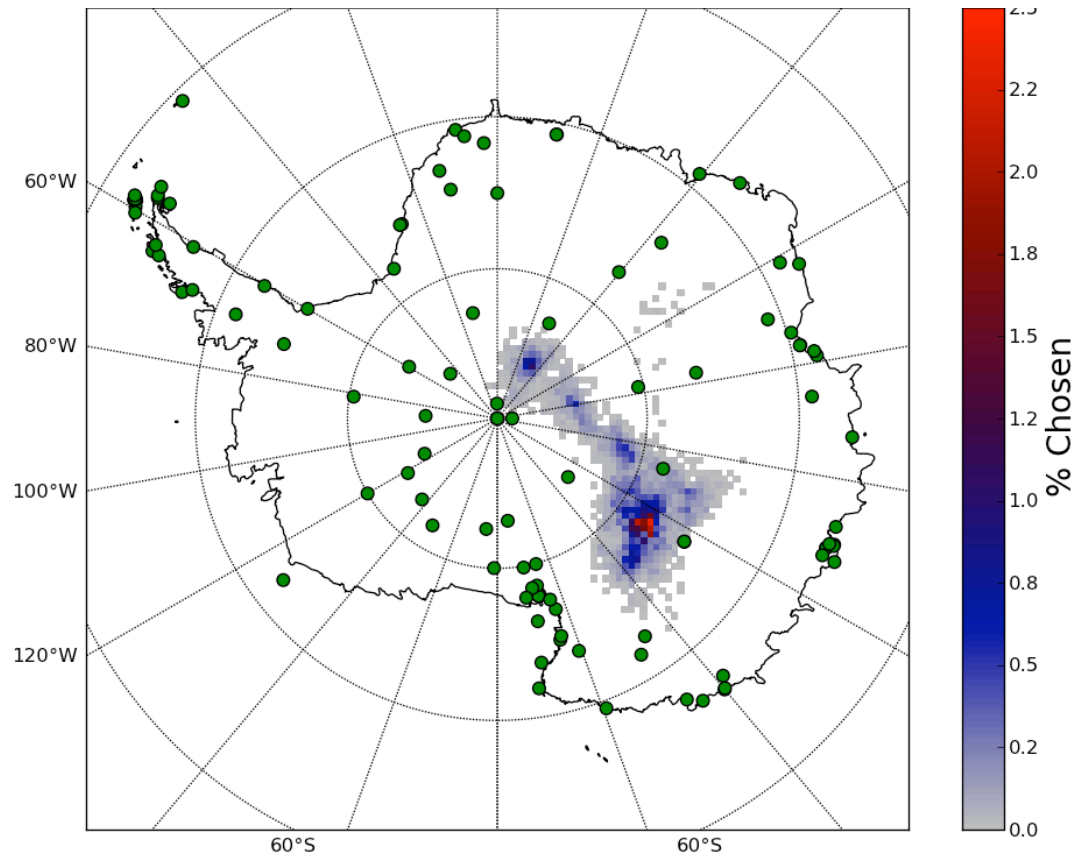
~Palmer Land

West Antarctica First Station



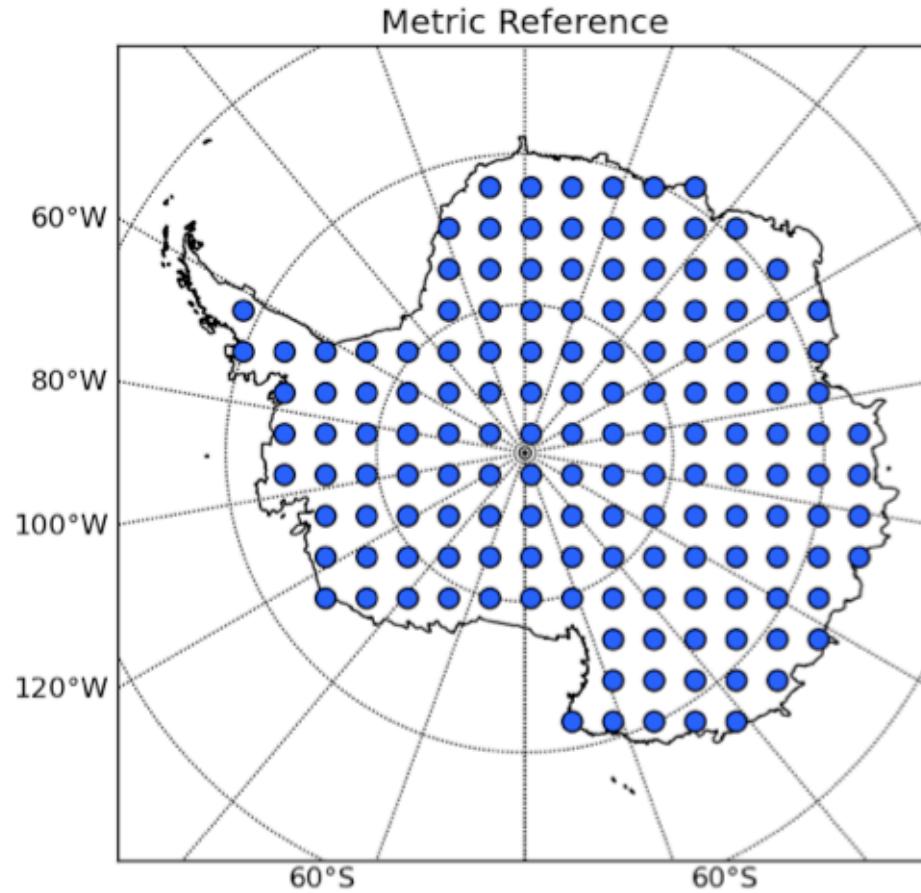
Along flanks of the Transantarctic mountains

East Antarctica First Station



MegaDunes area. Even more sharply defined when sampling error is treated more carefully.

Theory II: Optimize Over Many Performance Measures



Theory II: Many Performance Measures

- Similar to previous, but now $\mathbf{J} = [J_1, J_2, \dots, J_m]^T$
 - find stations that optimize ALL J s simultaneously

- Approximate changes to \mathbf{J} :

$$\delta\mathbf{J} = \mathbf{J}(\mathbf{x} + \delta\mathbf{x}) - \mathbf{J}(\mathbf{x}) \approx [D\mathbf{J}(\mathbf{x})]^T \delta\mathbf{x}$$

- The change in the **covariance** of \mathbf{J} is given by:

$$\delta\Sigma = D\mathbf{J}(\mathbf{x})^T (\mathbf{B}_{i-1} - \mathbf{B}_i) D\mathbf{J}(\mathbf{x})$$

- As for the scalar case, for a single point:

$$\delta\Sigma = [\text{cov}(\mathbf{J}, x_j) \text{cov}(\mathbf{J}, x_j)^T] [\text{var}(x_j) + r_i^2]^{-1}$$

- Summary measure of $\delta\Sigma$: trace (“covariance trace” method)

Theory II: Many Performance Measures

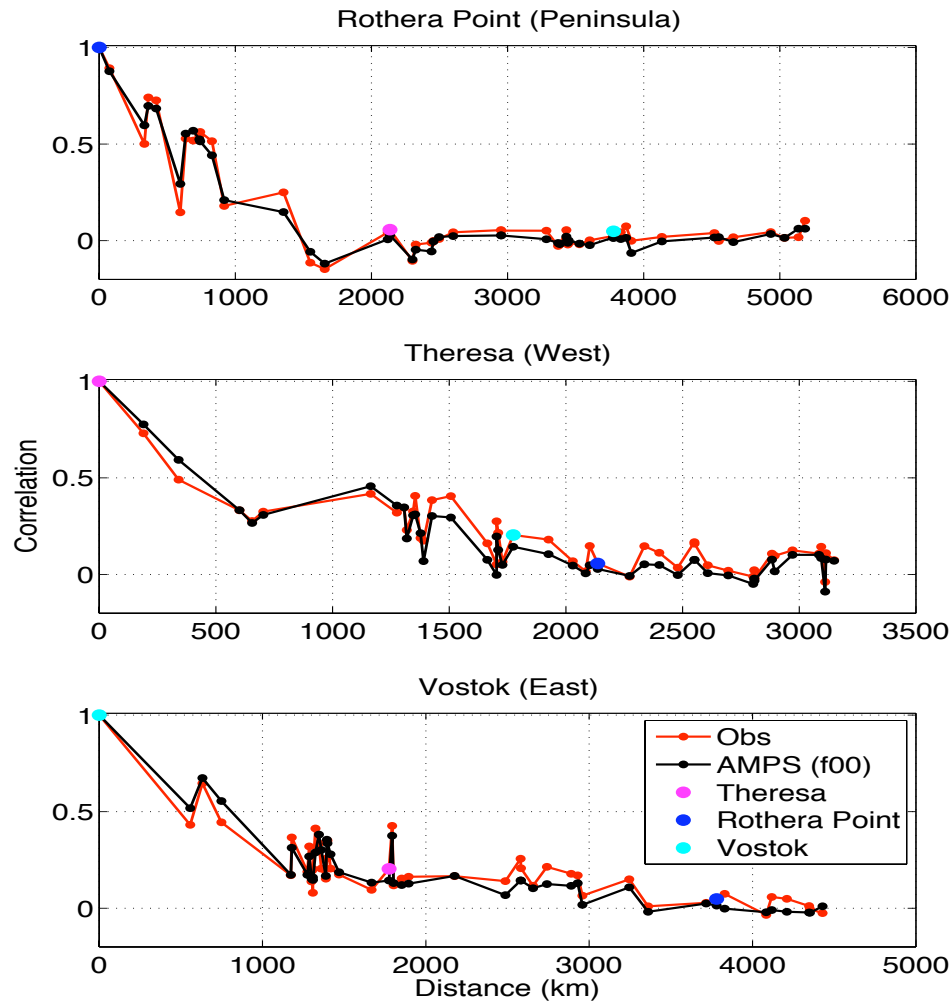
Alternative approach: relative entropy

$$D_{\text{KL}}(p_2 \| p_1)_{\text{cov}} = \frac{1}{2} \left\{ \text{tr} (\boldsymbol{\Sigma}_0^{-1} \boldsymbol{\Sigma}_1) - M - \ln \frac{|\boldsymbol{\Sigma}_1|}{|\boldsymbol{\Sigma}_0|} \right\}$$

Spatial correlation reduces the determinant (last term)

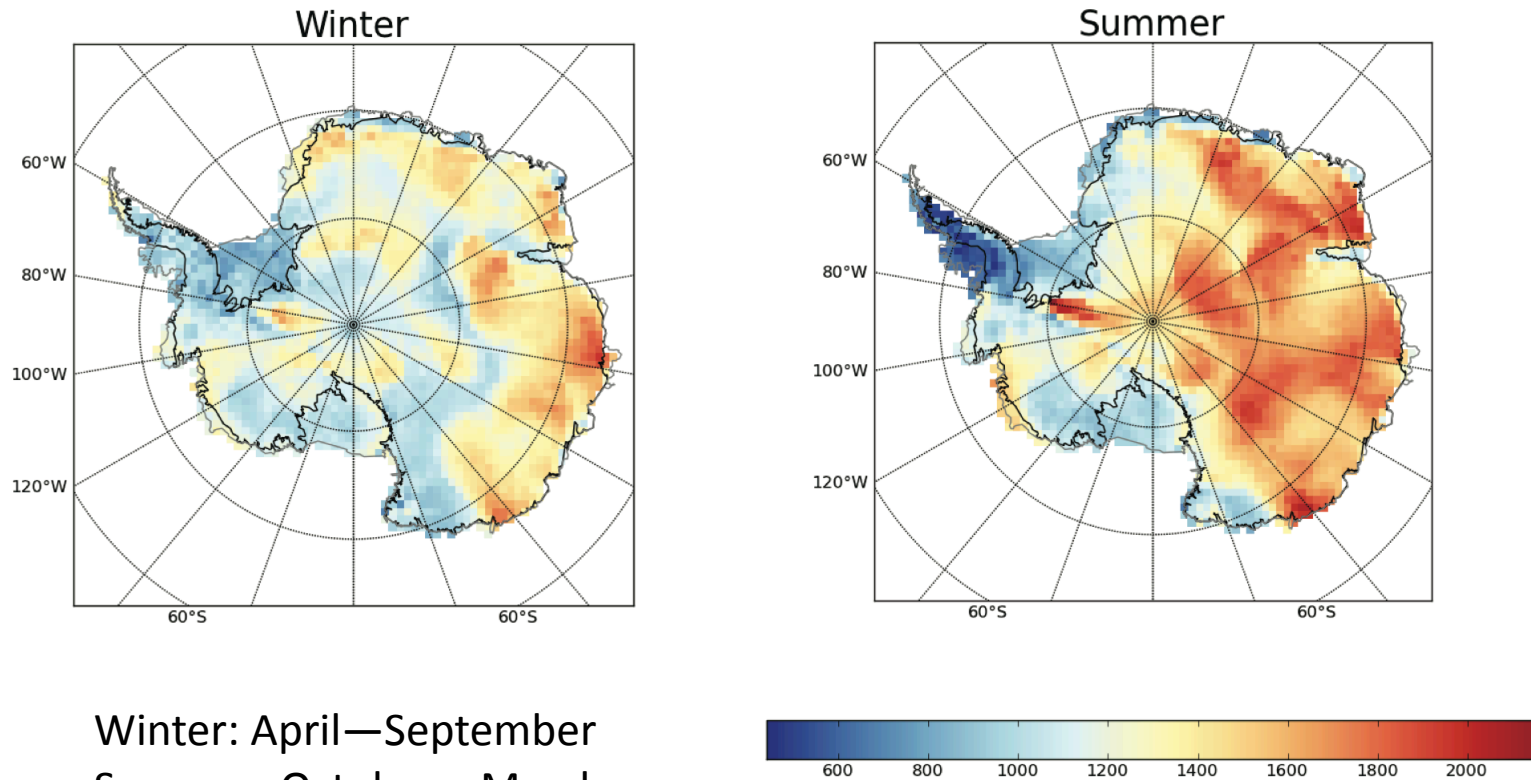
- Penalty on locations that increase correlations
- “covariance trace” favors those locations
- Interpretation benefits from knowing correlations

Correlation Length Scales



Compare obs and AMPS

Correlation Length Scales



Winter: April—September
Summer: October—March

Rich patterns. Main signals:

- Longer in summer
- Longer in East Antarctica
- Shortest on Peninsula

Summary

- Network design provides objective input for
 - New networks (clean slate)
 - Augmenting existing networks
 - Reconfiguring networks
- New extension allows many metrics at once
 - Good for spatial fields; soundings; multiple fields
 - More in next talk
- Current/future work
 - Soundings, and weather forecast improvement
- **Please give us feedback on metrics!**



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