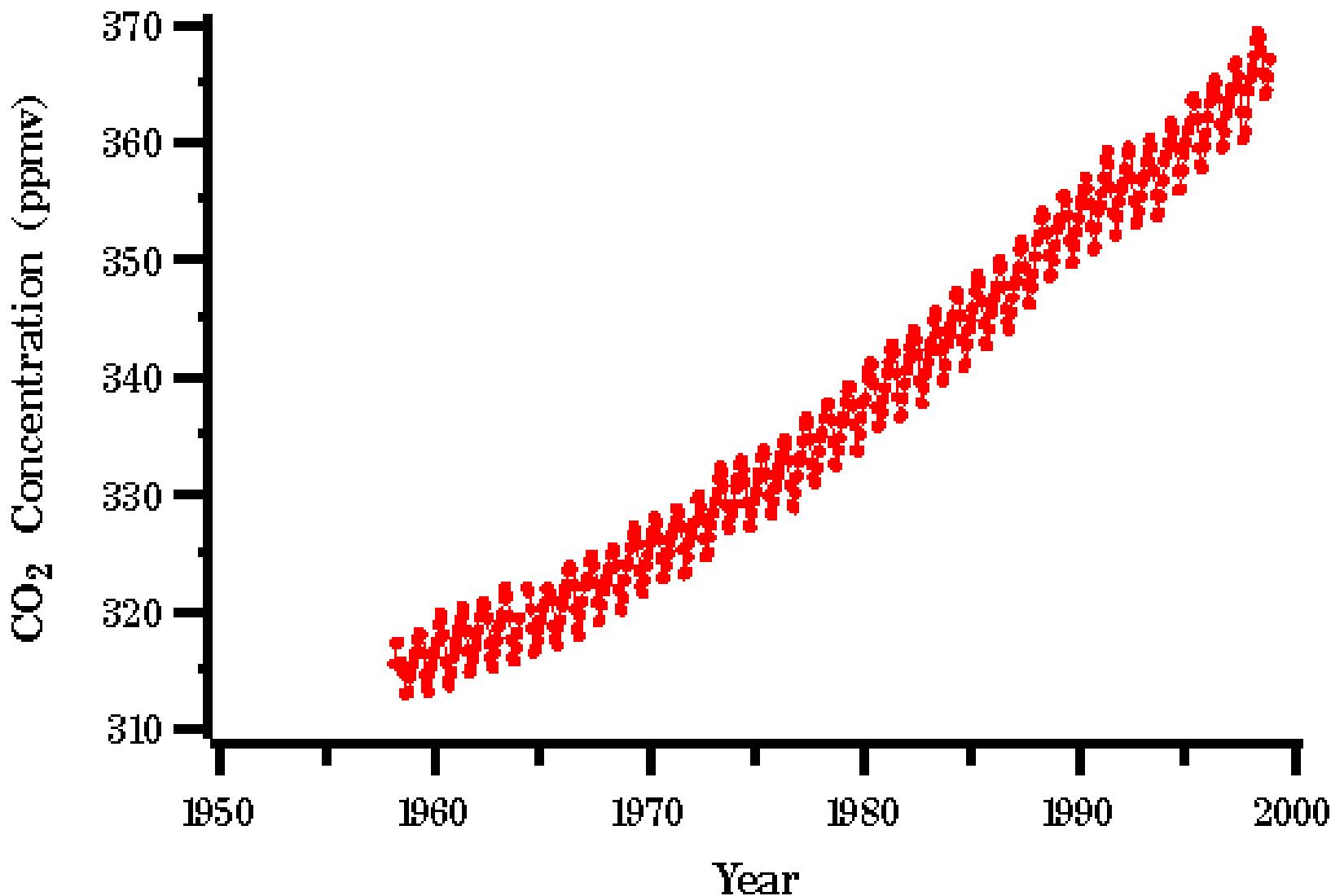


Spatial and Temporal Considerations for a Reference Network

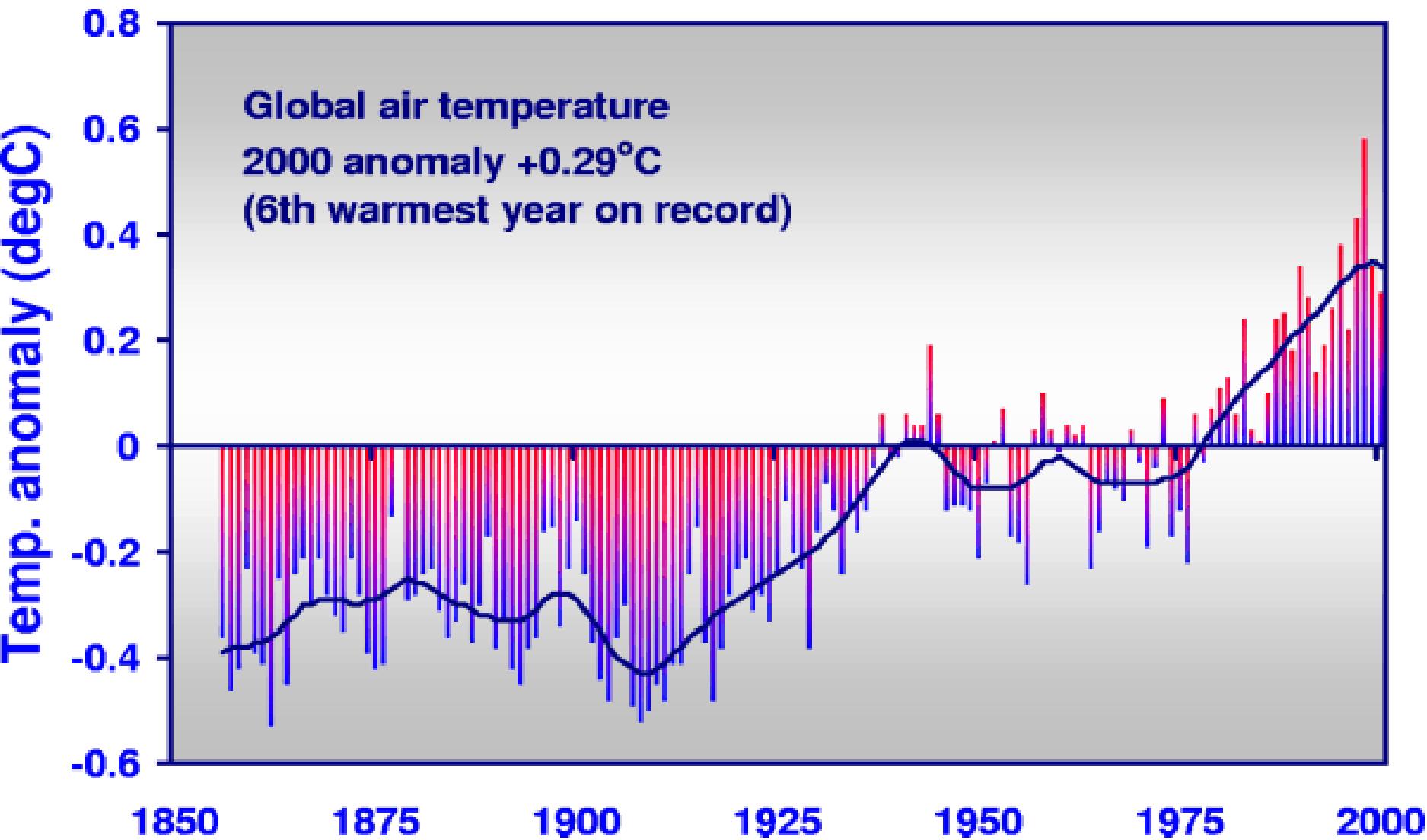
Betsy Weatherhead

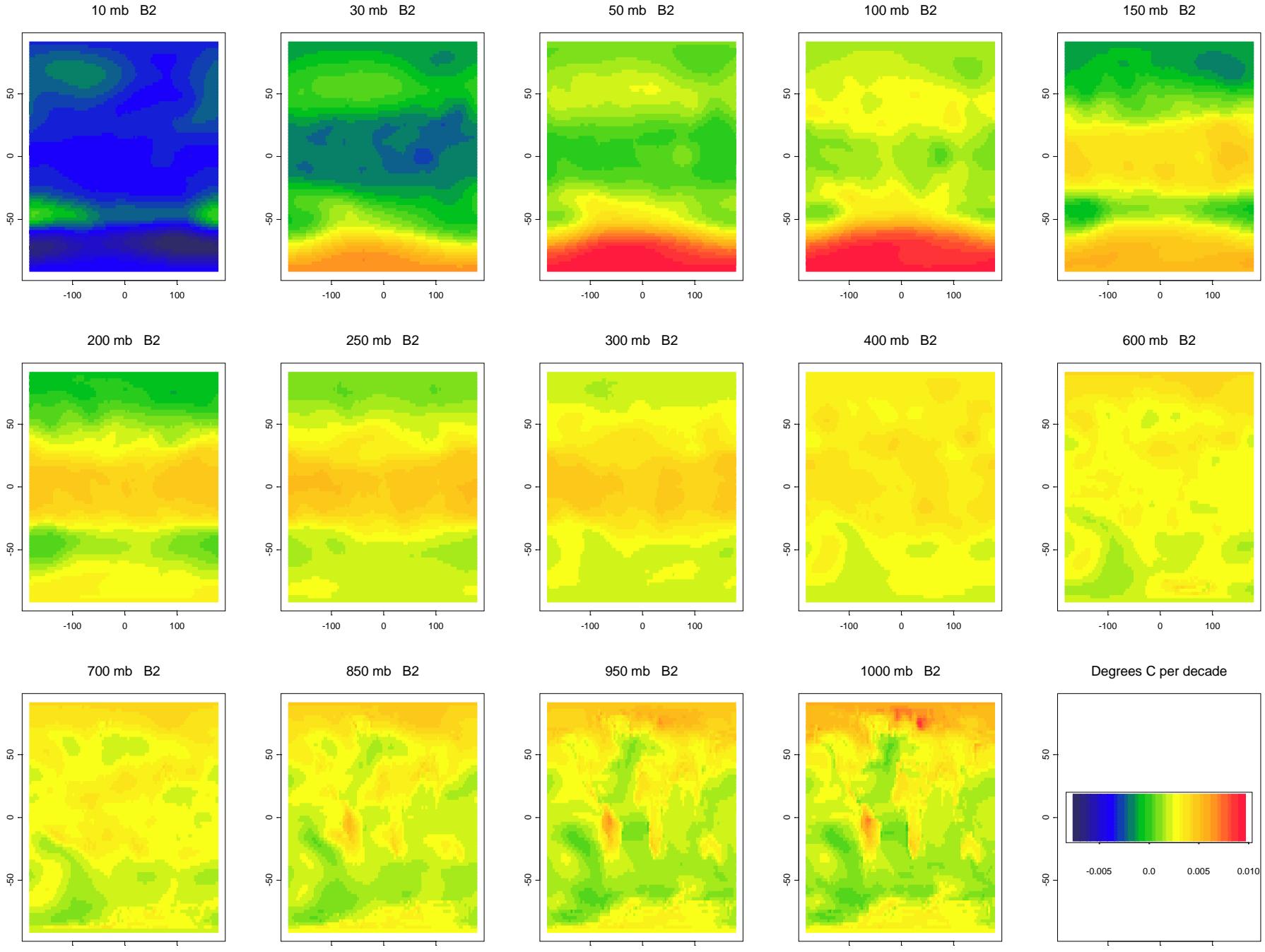
Mauna Loa, Hawaii



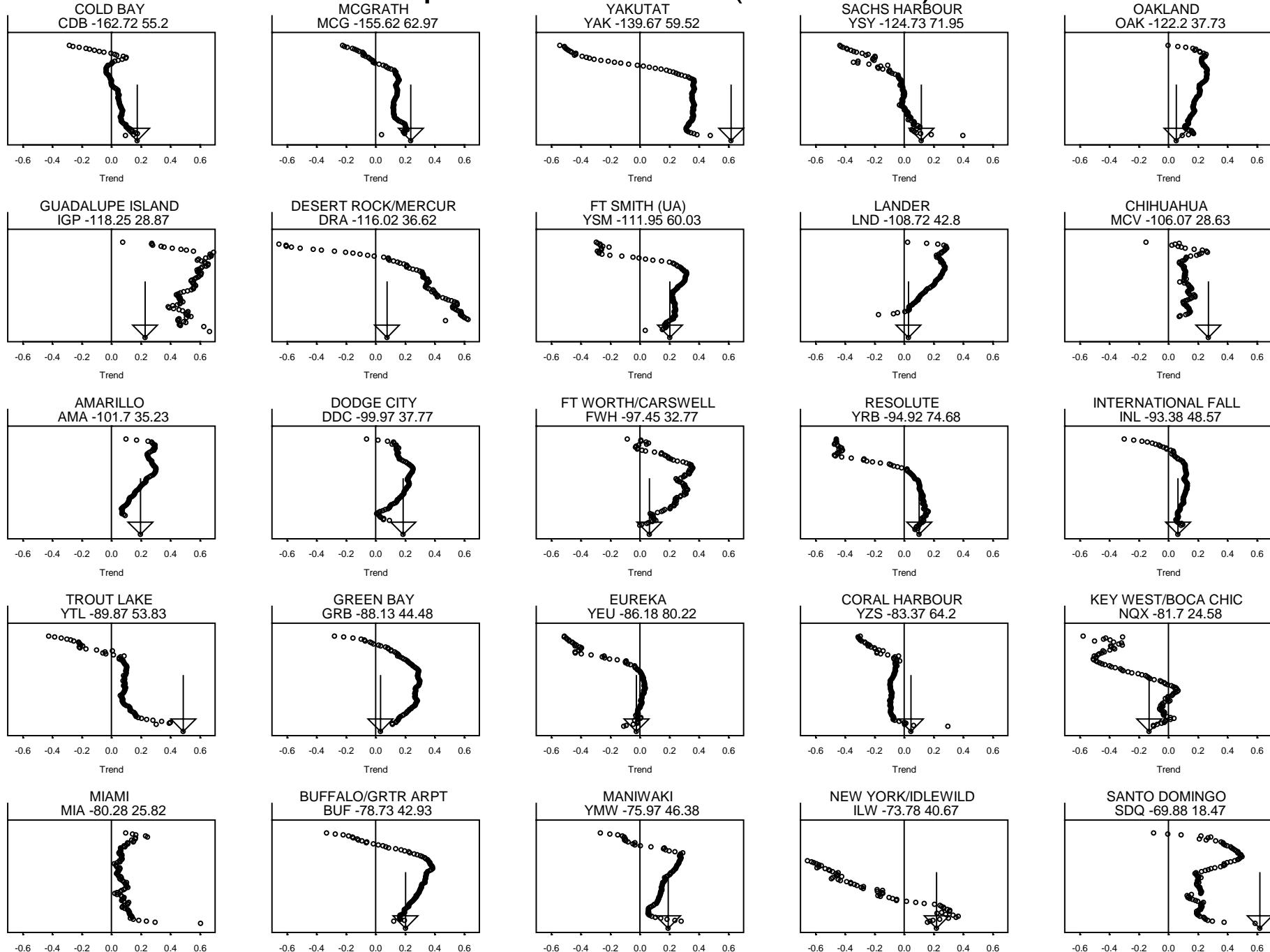
Source: Dave Keeling and Tim Whorf (Scripps Institution of Oceanography)

Nature's response is not always so linear.

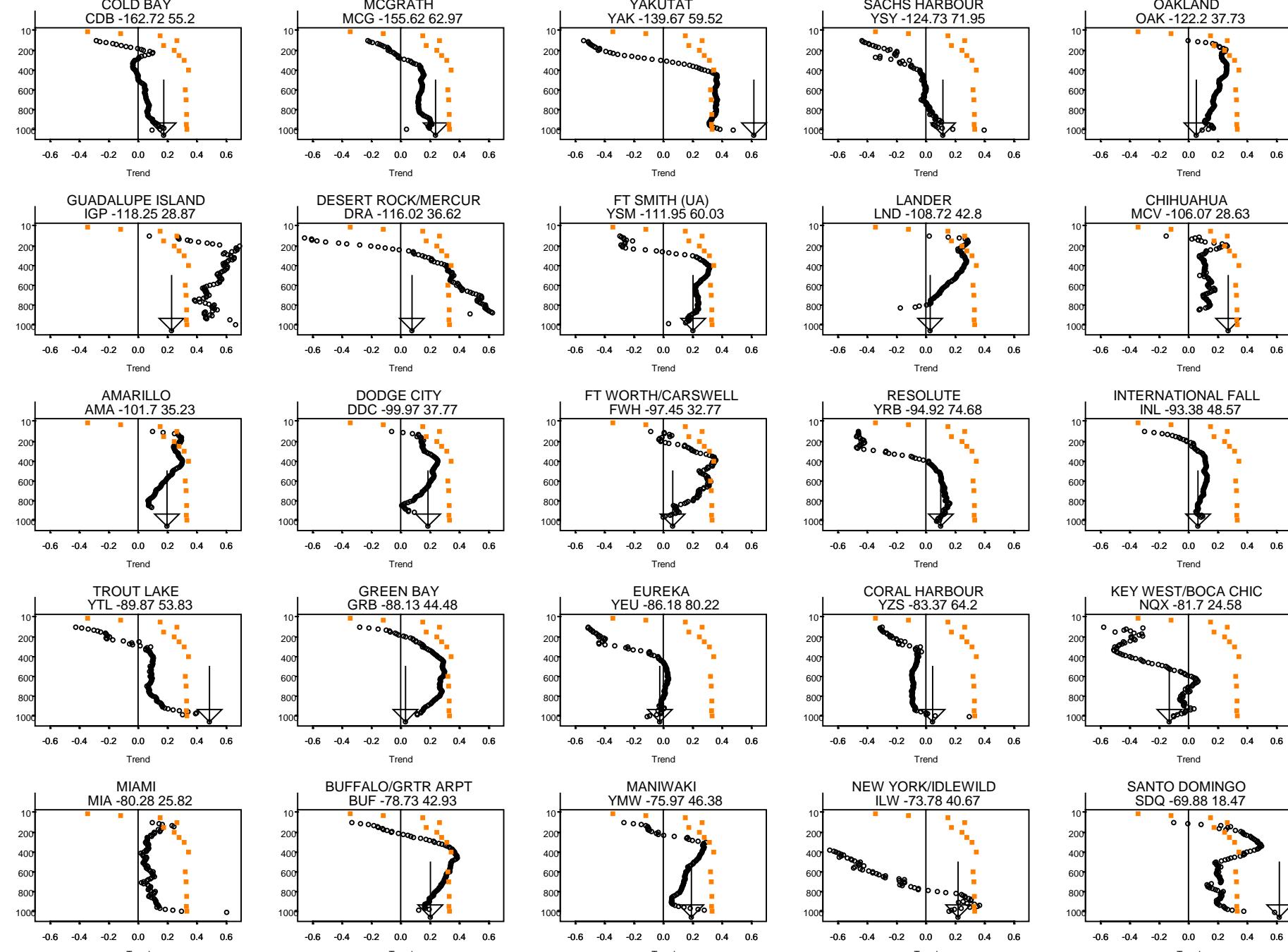




Sonde Temperature Trends (C/decade) at 0 Z



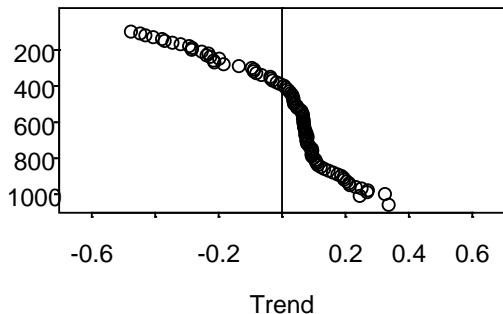
Sonde Temperature Trends (C/decade) at 0 Z



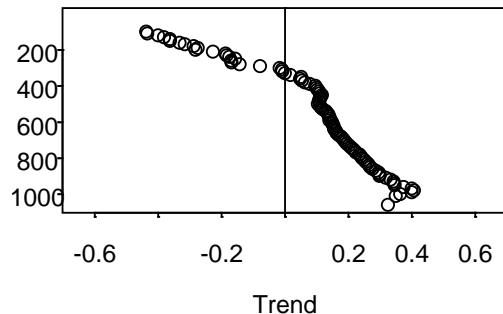
Are there patterns?

Sonde Temperature Trends (C/decade) at 12 Z

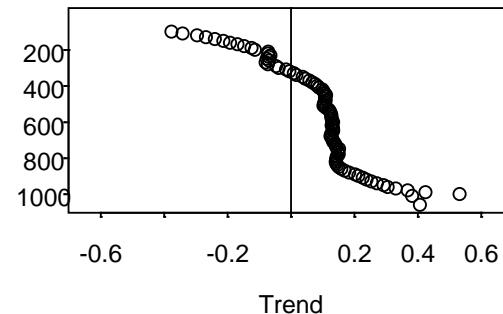
NOME FED BLDG
OME -165.4 64.5



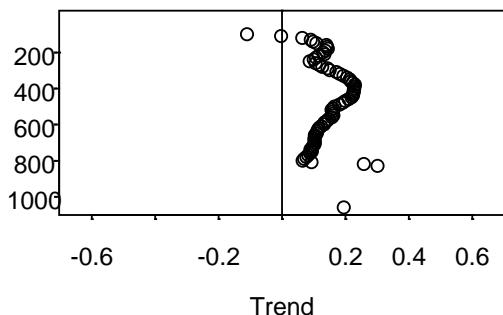
POINT BARROW
BRW -156.78 71.3



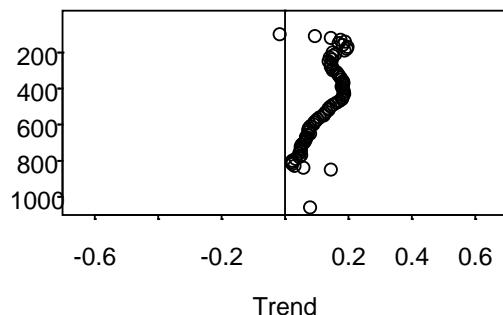
NAKNEK
AKN -156.65 58.68



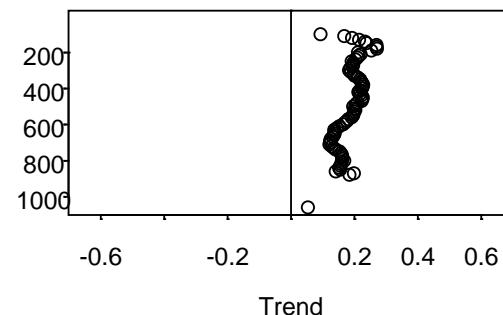
LANDER
LND -108.72 42.8



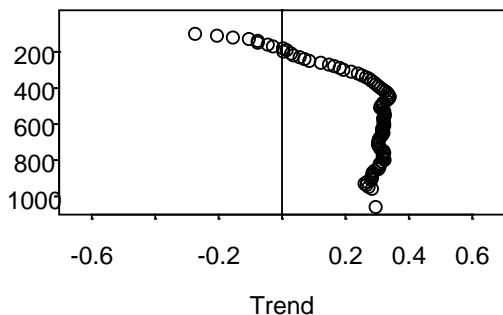
GRAND JUNCTION
GJT -108.53 39.12



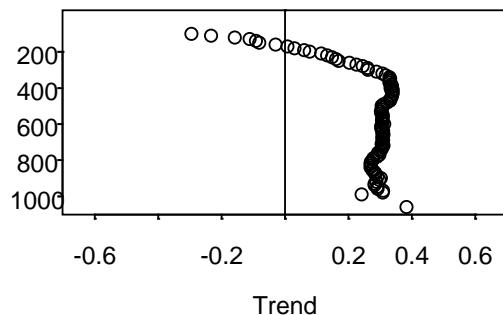
EL PASO
ELP -106.4 31.82



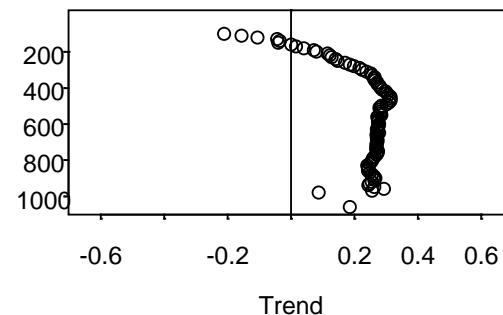
DAYTON/WRIGHT PATT
DAY -84.12 39.87



HUNTINGTON
HTS -82.55 38.37

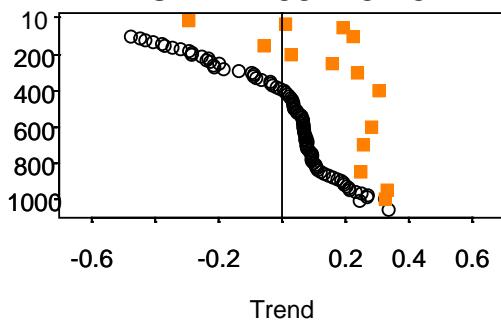


PITTSBURGH/PITTSBG
PIT -80.22 40.5

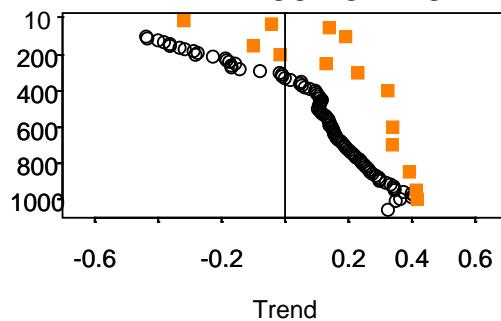


Sonde Temperature Trends (C/decade) at 12 Z

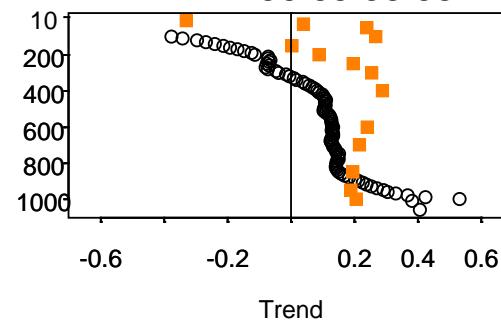
NOME FED BLDG
OME -165.4 64.5



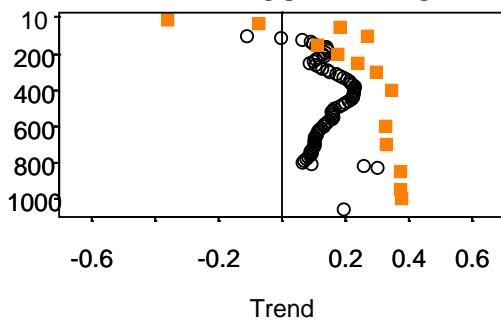
POINT BARROW
BRW -156.78 71.3



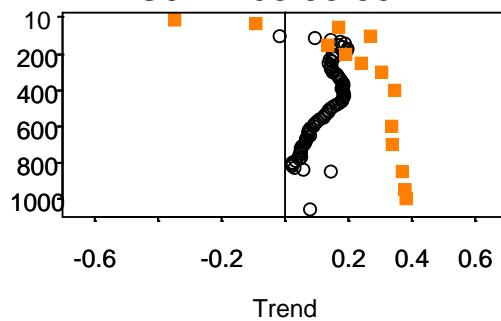
NAKNEK
AKN -156.65 58.68



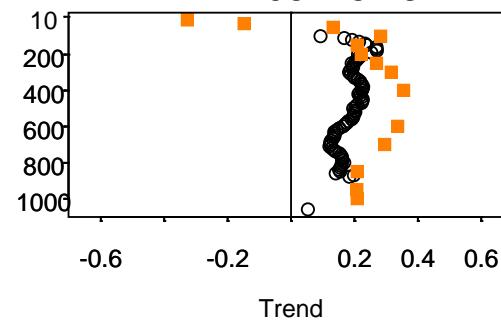
LANDER
LND -108.72 42.8



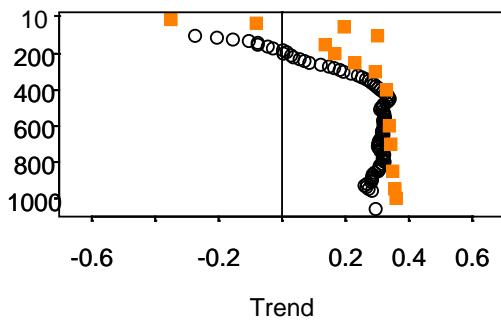
GRAND JUNCTION
GJT -108.53 39.12



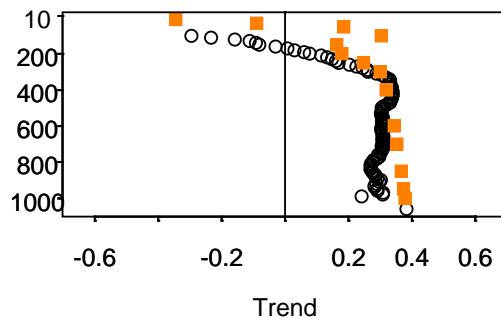
EL PASO
ELP -106.4 31.82



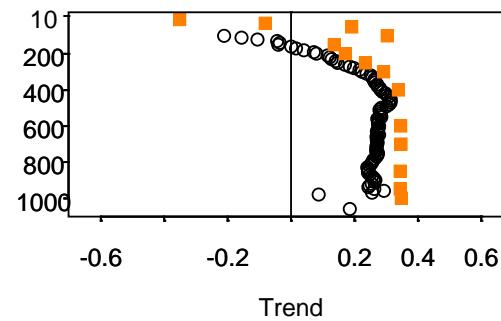
DAYTON/WRIGHT PATT
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HUNTINGTON
HTS -82.55 38.37



PITTSBURGH/PITTSBG
PIT -80.22 40.5



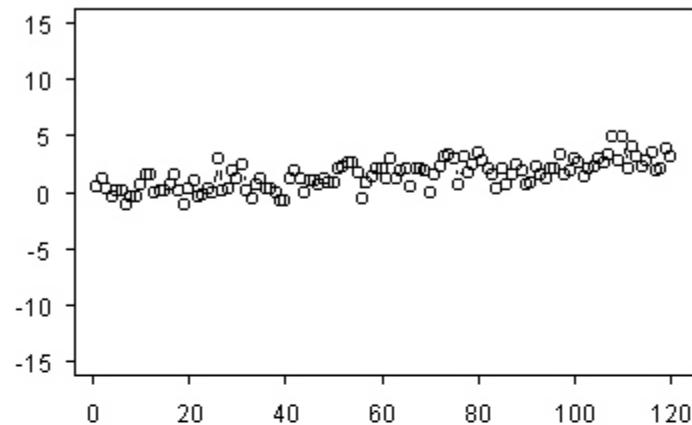
What is the primary goal of a Reference Network?

- Detection of Representative Trends?
 - Spatially: site in representative areas
 - Temporally: monitor consistently to establish trends
- Understand errors and transfer standards?
 - Spatially: site with existing instrumentation and expertise
 - Temporally: make measurements when they can add to knowledge.

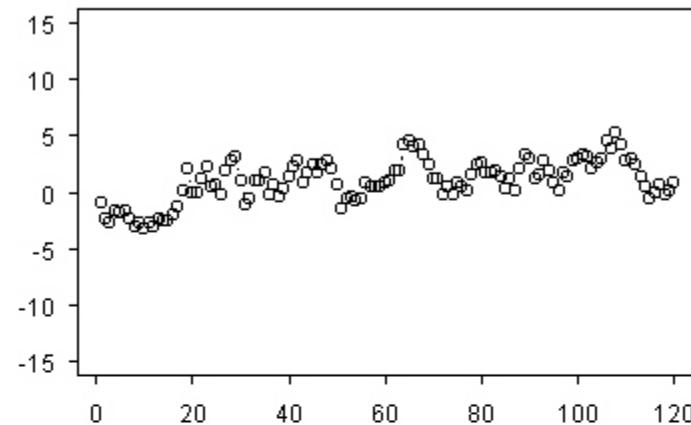
Trend Detection

- “Finding a change which is large relative to natural variability.”
- For environmental data both the magnitude of variability and the memory (autocorrelation) hinder our ability to detect trends.

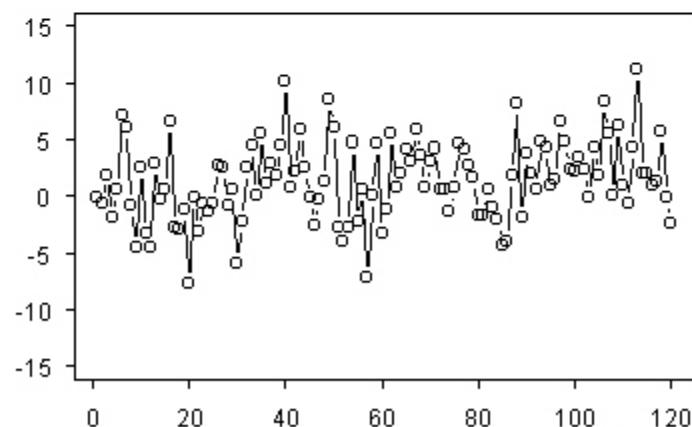
Low Noise, Low Autocorrelation



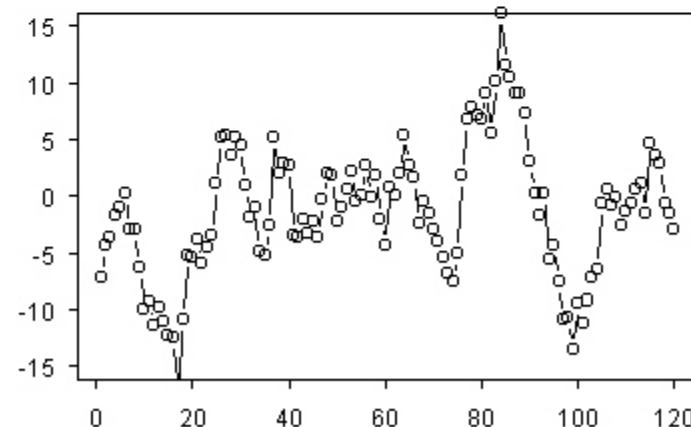
Low Noise, High Autocorrelation



High Noise, Low Autocorrelation



High Noise, High Autocorrelation

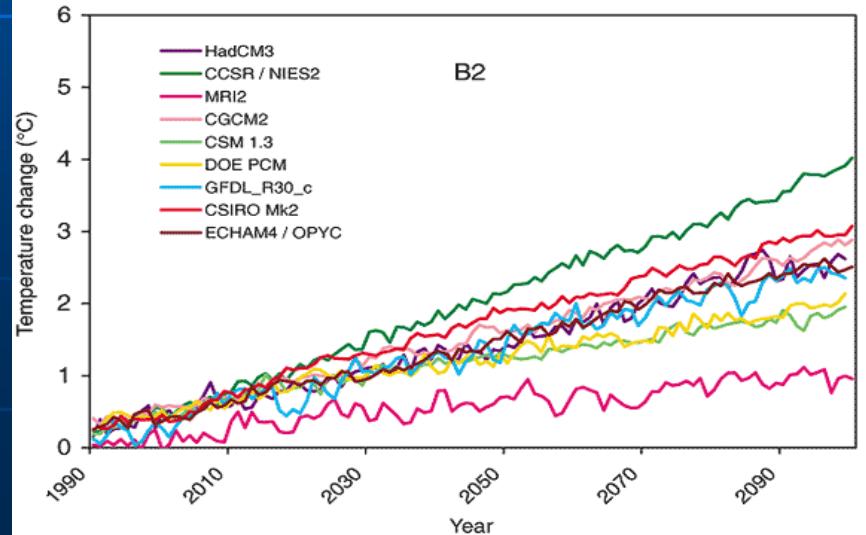
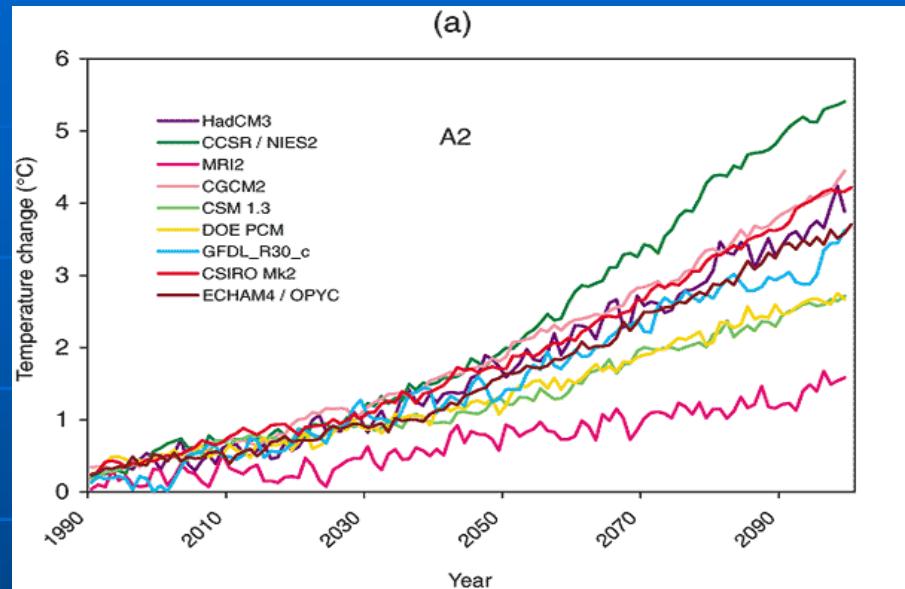


The key

- All four parameters which affect our ability to detect trends vary by location:
 - Magnitude of variability
 - Autocorrelation
 - Size of the trend
 - Stability of the measurements

One example: temperature trends

- Temperature trends are predicted by a number of different models.
- How long will we need to monitor to detect trends?



Temperature Trend Analysis

- As for most environmental data, trends are usually derived using a statistical model such as:
- Temperature = trend + seasonal + Noise
 - Where the trend may be linear or not.
 - Where the noise involves both the magnitude of variability and autocorrelation.

Estimating the Number of Years for Trend Detection

- If we understand the size of trend we are looking for;
- If we know the typical magnitude of variability;
- If we know the amount of temporal memory in the system;
- We can estimate how long we need to monitor.

Number of Years needed to detect a trend

- Approximately:

$$n = \{ (2 * \sigma_n / |\omega_0|) \sqrt{(1 + \phi) / (1 - \phi)} \}^{2/3}$$

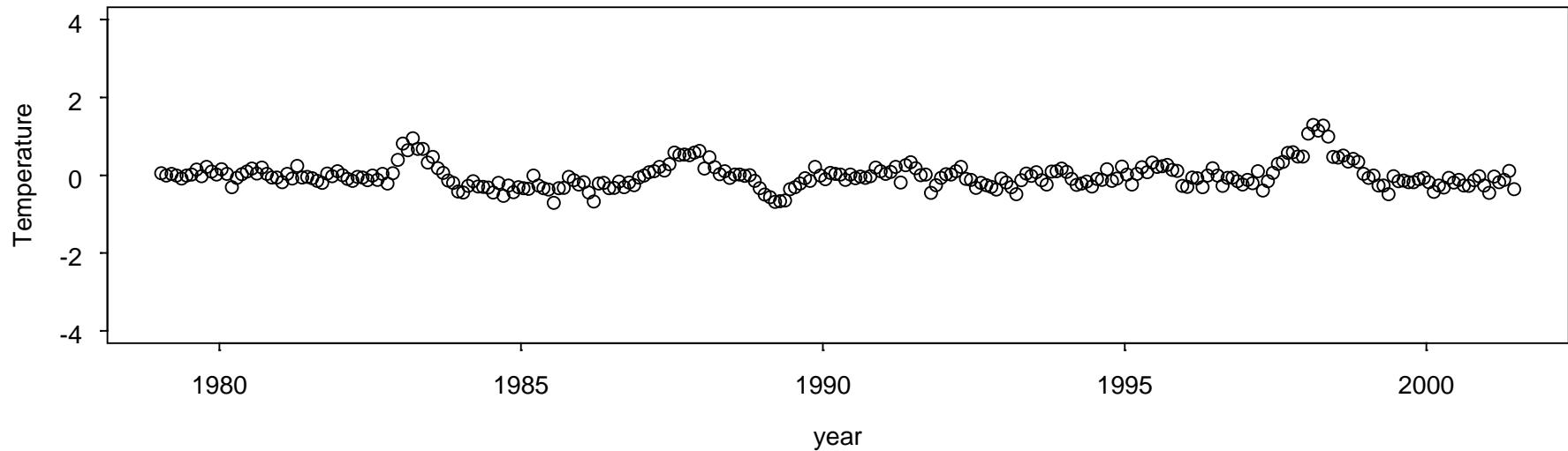
- Assuming that detection is declared at the 95% confidence level
- This estimate allows for 50% likelihood of detection

Years to Detect .2 Degrees per Decade Trend in Temperature

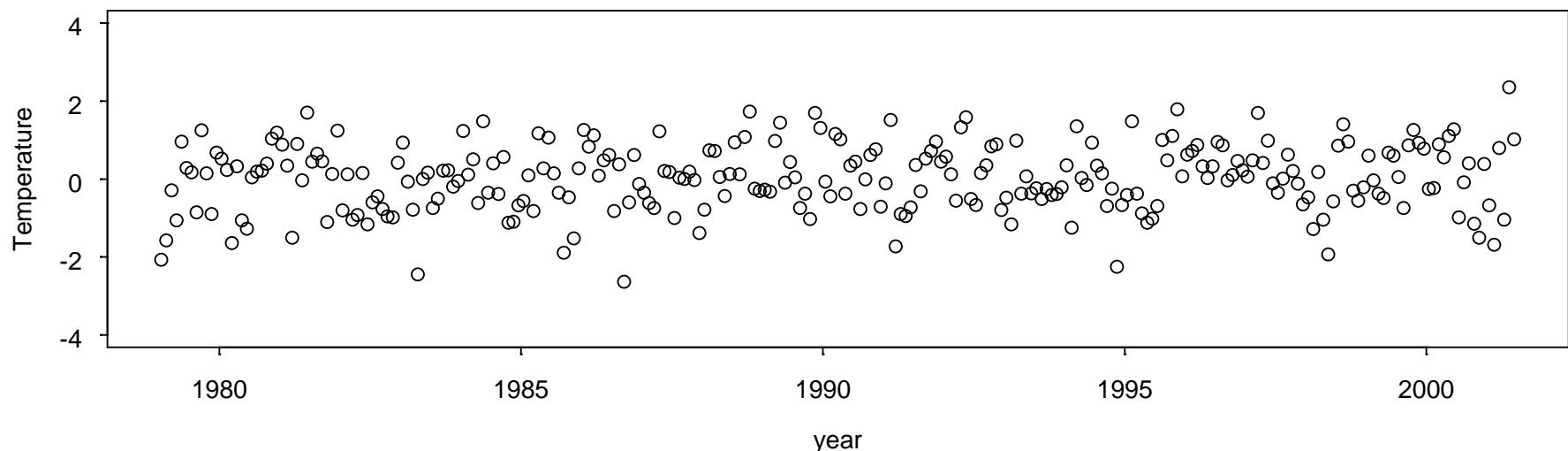
	0	.3	.6	.9
.5	14	17	21	35
1	22	26	34	56
3	45	55	71	100+
5	63	77	100+	100+

MSU Channel 2

Equator



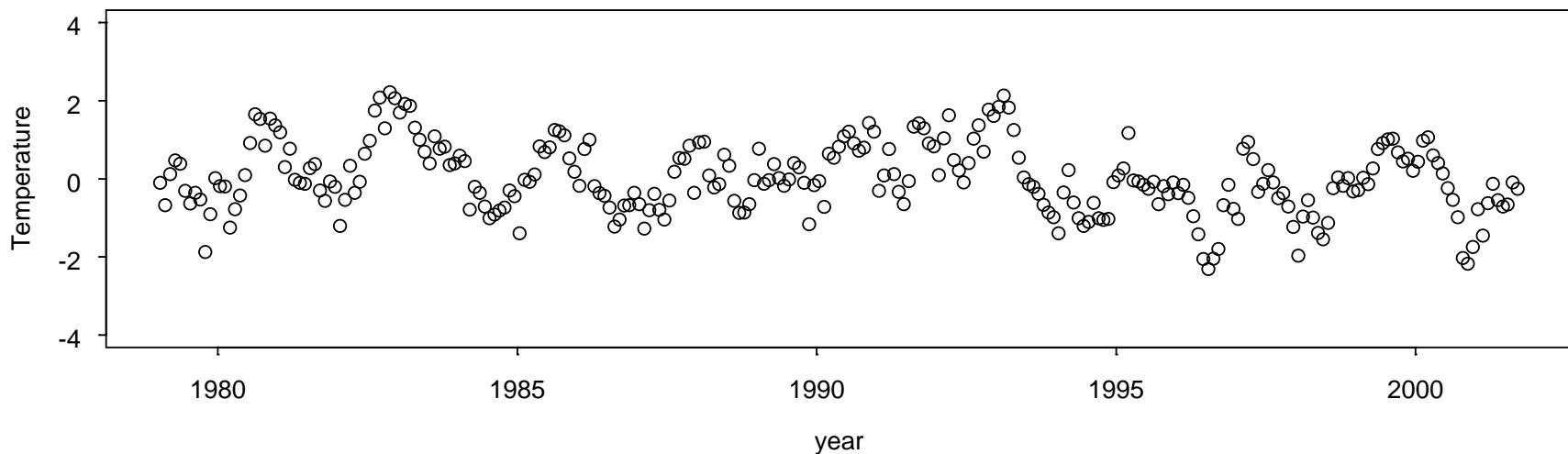
San Francisco



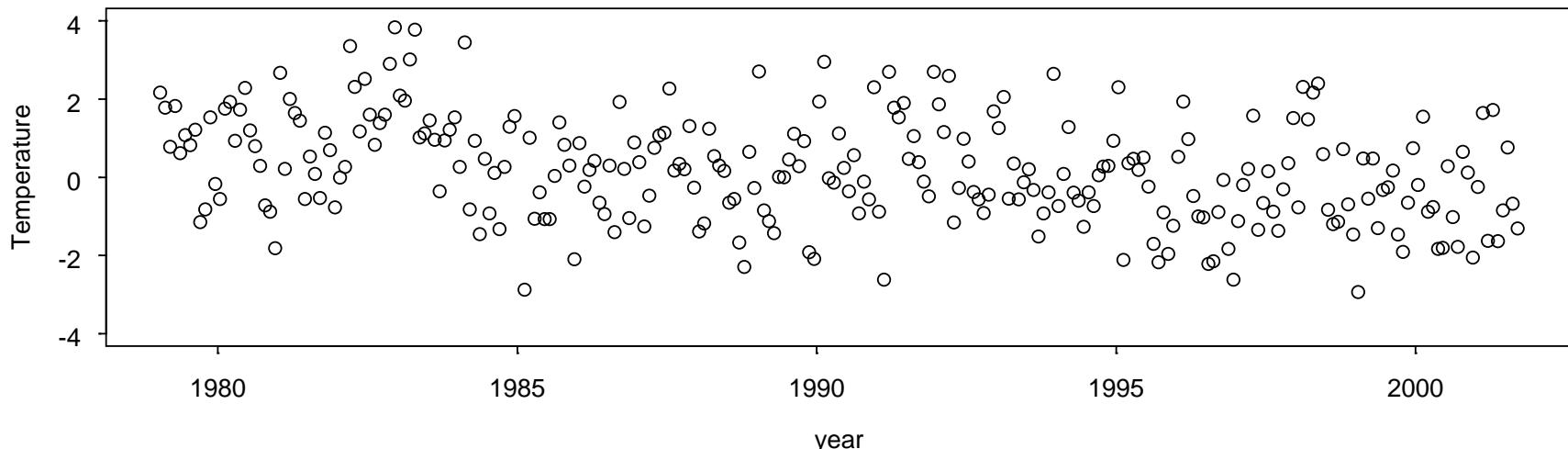
Weatherhead Tue Dec 11 20:36:28 2001

MSU Channel 4

Equator



San Francisco



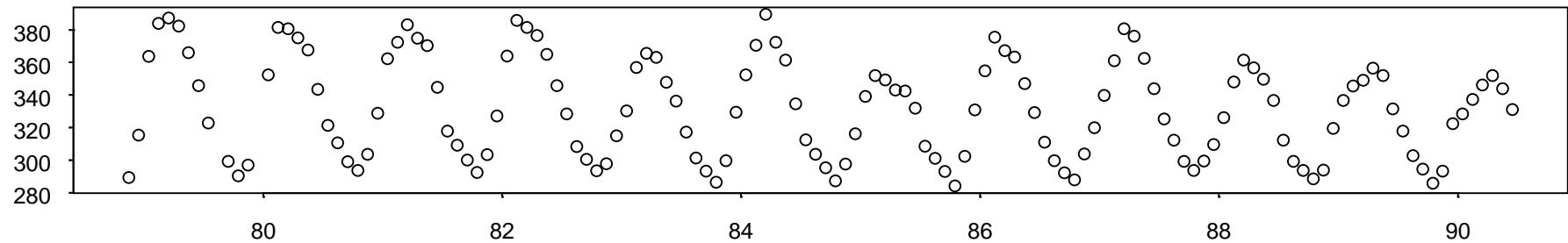
Weatherhead Tue Dec 11 20:36:21 2001

Visual Example

- How many years does it take to detect a trend in ozone?
- Use our understanding of variability;
- Use our understanding of the predicted trends
- Estimate visually how long it will take to detect a trend.

SBUV OZONE TOTAL COLUMN OZONE - 40N

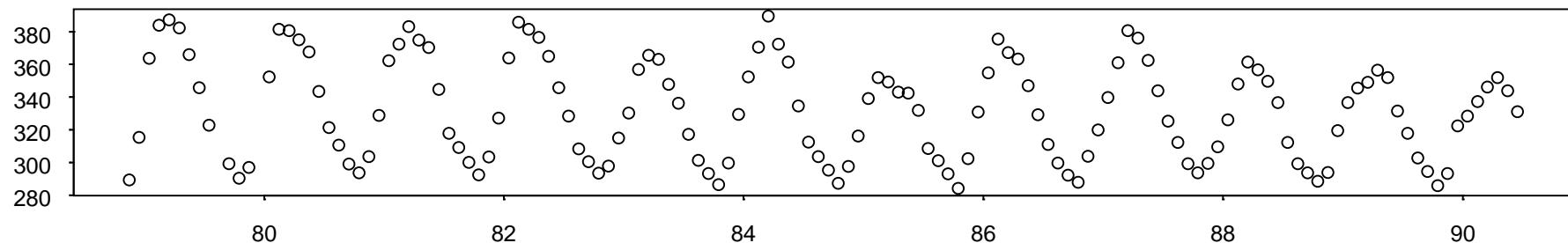
Original Monthly Averaged Data



Weatherhead Fri Nov 2 11:38:10 2001

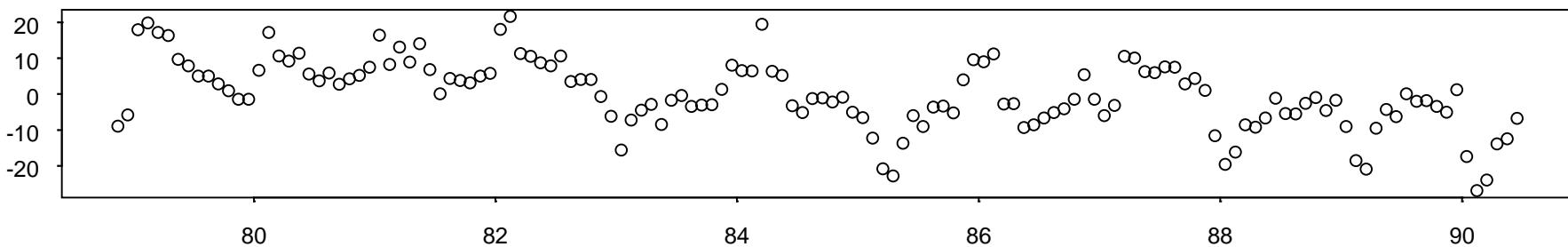
SBUV OZONE TOTAL COLUMN OZONE - 40N

Original Monthly Averaged Data



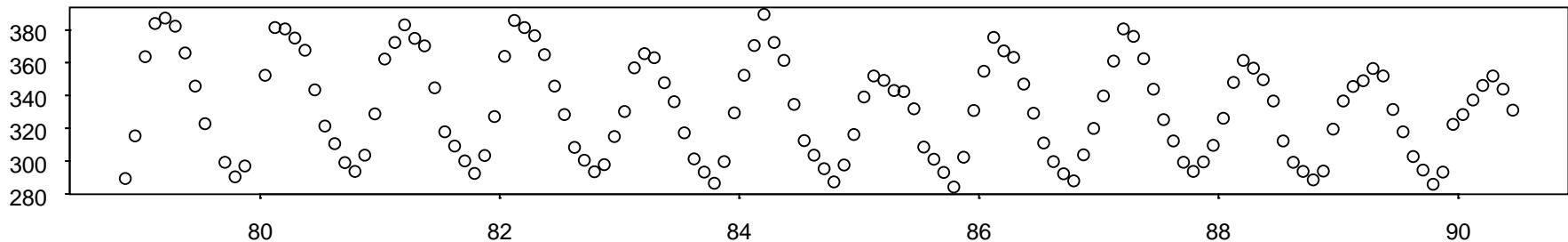
Weatherhead Fri Nov 2 11:48:50 2001

Monthly Means Removed, Lowess Line Fit Superimposed



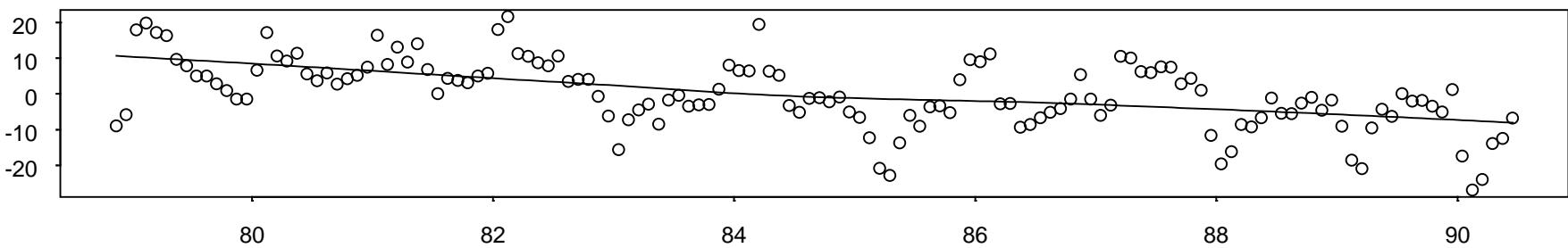
SBUV OZONE TOTAL COLUMN OZONE - 40N

Original Monthly Averaged Data



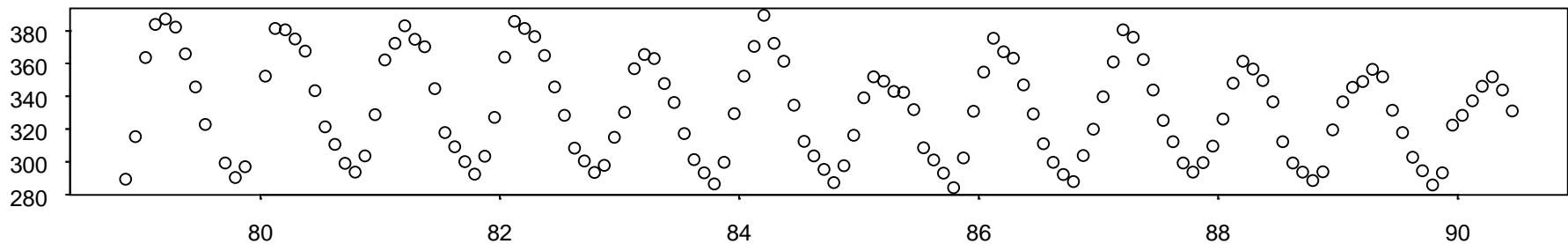
Weatherhead Fri Nov 2 11:48:50 2001

Monthly Means Removed, Lowess Line Fit Superimposed



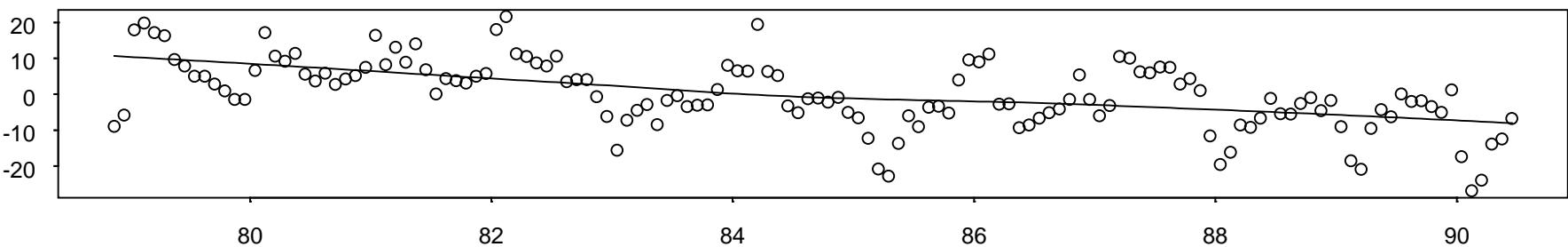
SBUV OZONE TOTAL COLUMN OZONE - 40N

Original Monthly Averaged Data

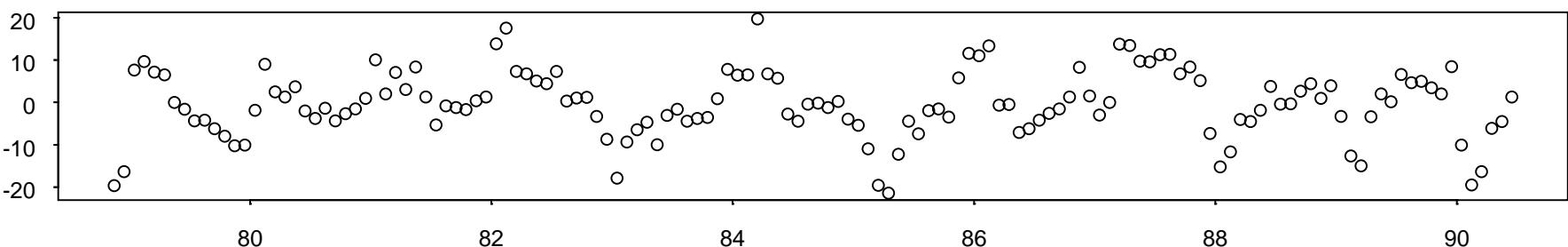


Weatherhead Fri Nov 2 11:48:50 2001

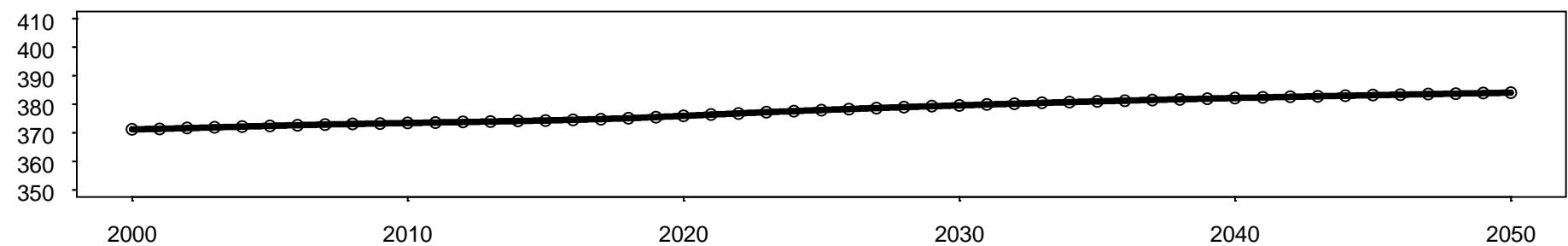
Monthly Means Removed, Lowess Line Fit Superimposed



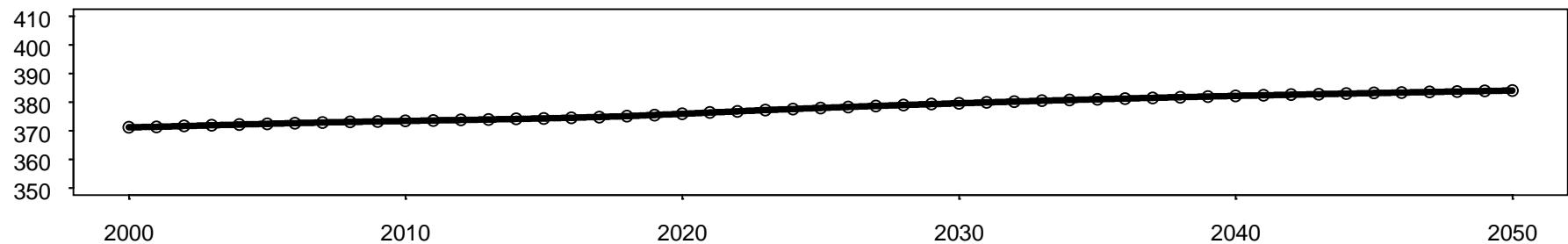
Residuals From Lowess Line Fit



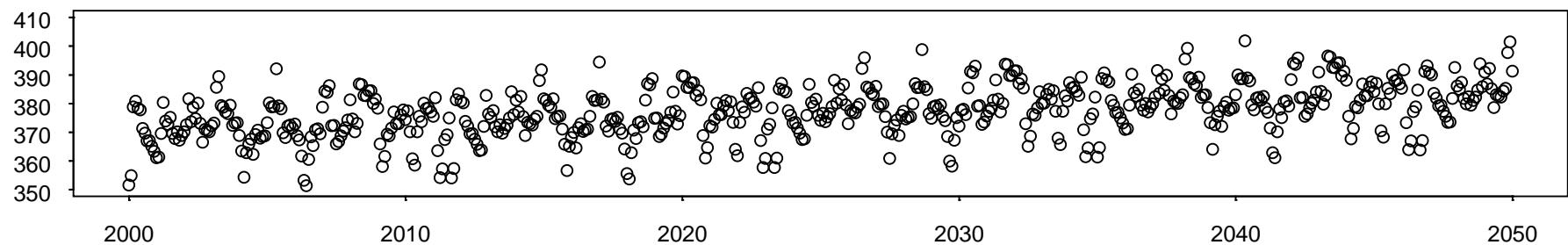
GSFC Predictions - without climate change



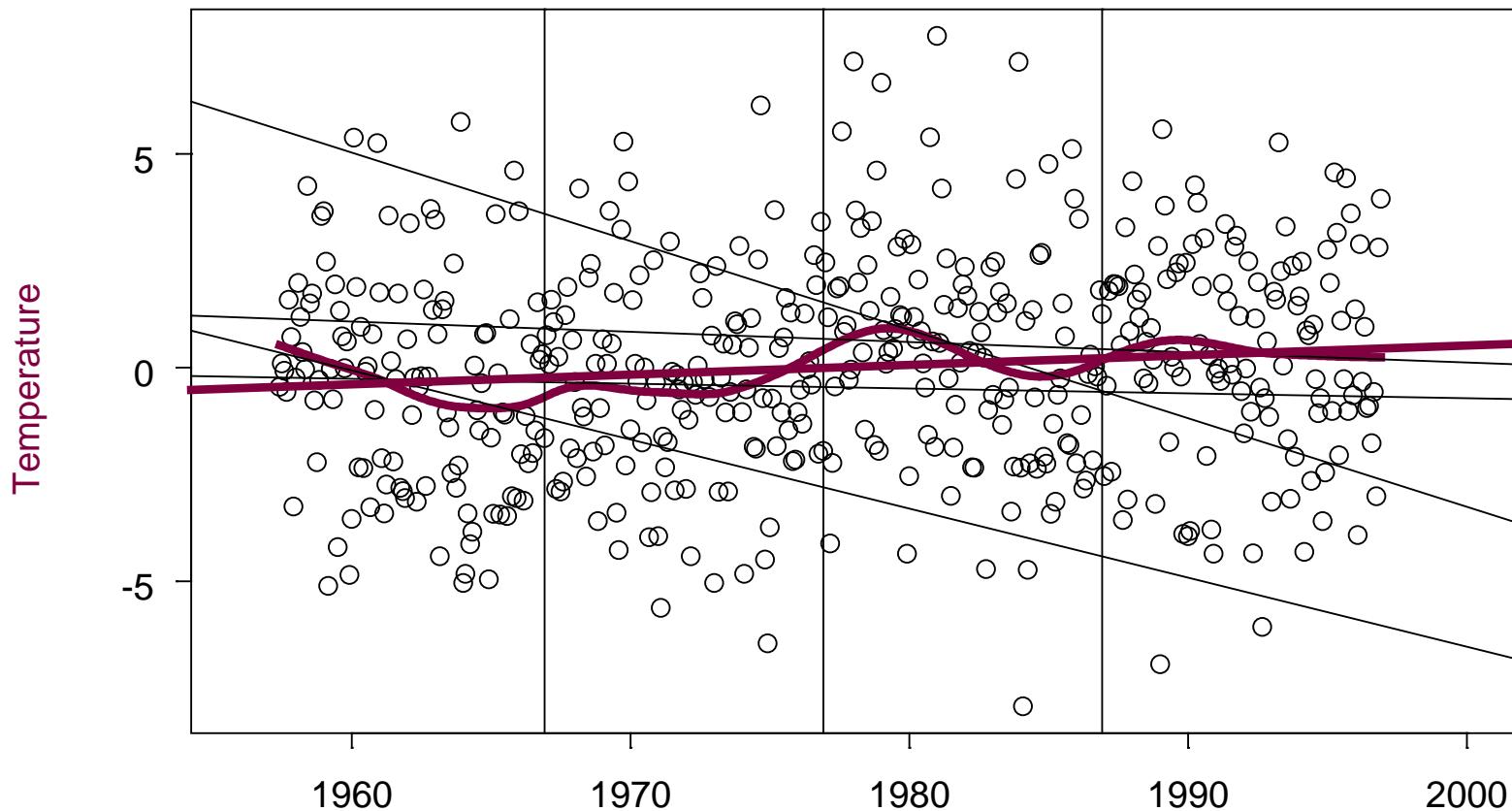
GSFC Predictions - without climate change



GSFC Predictions with SBUV Lowess Residuals



BARROW - 800 mb - 0Z - temperature anomalies



Metric: Number of years

- Our ability to detect trends is limited by natural variability
- We can estimate how long it will take to detect trends
- Some parameters, some places, some monitoring approaches may take considerably less time than others.

■ What can we control?

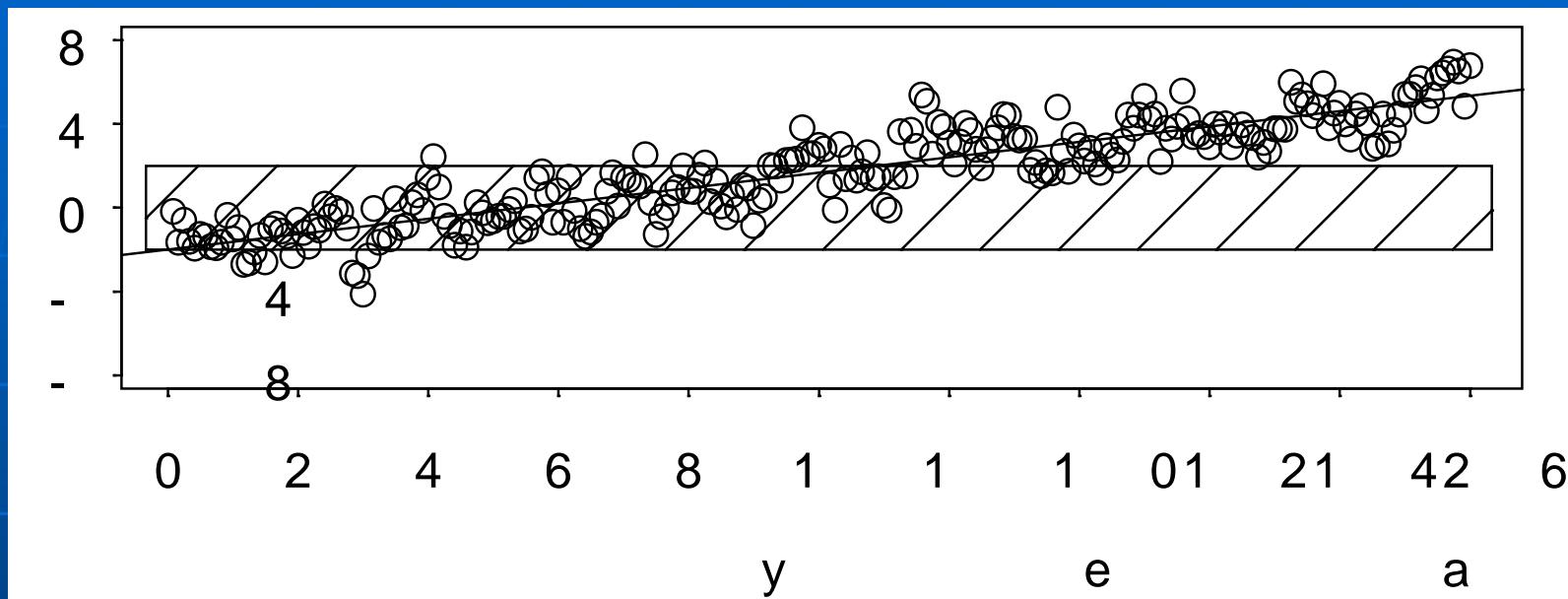
We can control only four aspects of monitoring to detect trends

- **What we monitor**
- **What accuracy**
- **Where we monitor**
- **What frequency**

What accuracy?

- Relative accuracy is all that's needed for trend detection.
- Relative accuracy is extremely hard to maintain for decades without absolute accuracy.
- Improved accuracy may save decades in monitor or may be irrelevant.

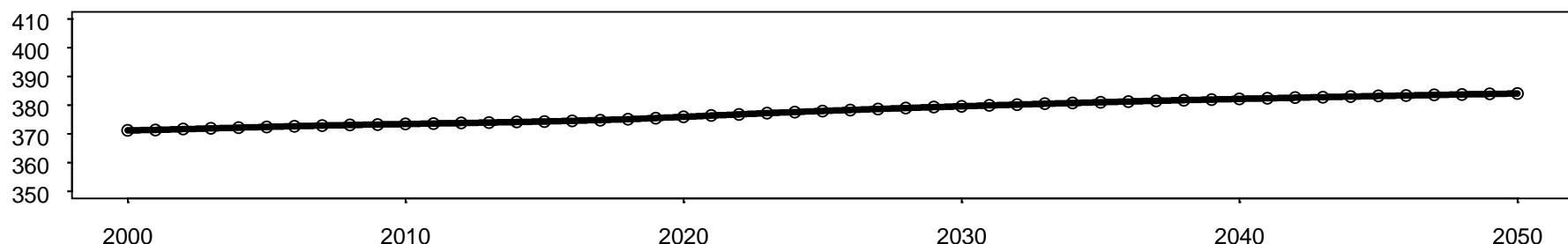
Case Example



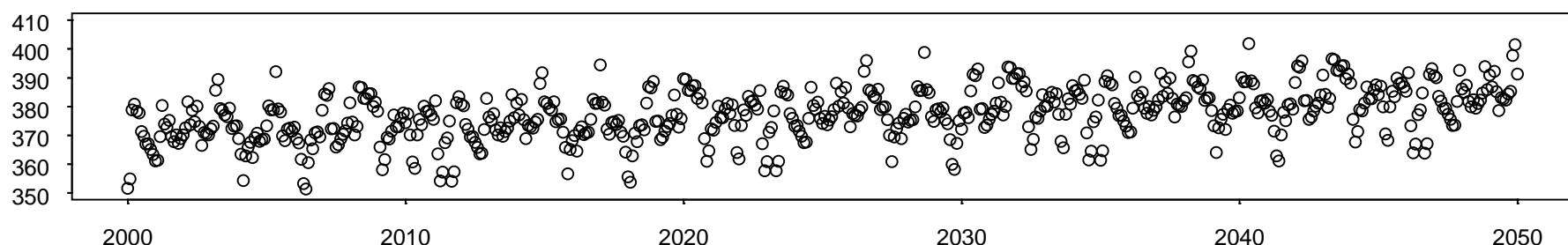
- Uncertainty: $\pm 2\%$; Trend: 4% per decade
- Result:
 - First ten years of data are still unsubstantial
- Improving Accuracy to $\pm 1\%$ saves five years of monitoring

GSFC 2d Predictions with SBUV Residuals of Total Col. Ozone (d.u.) 40N

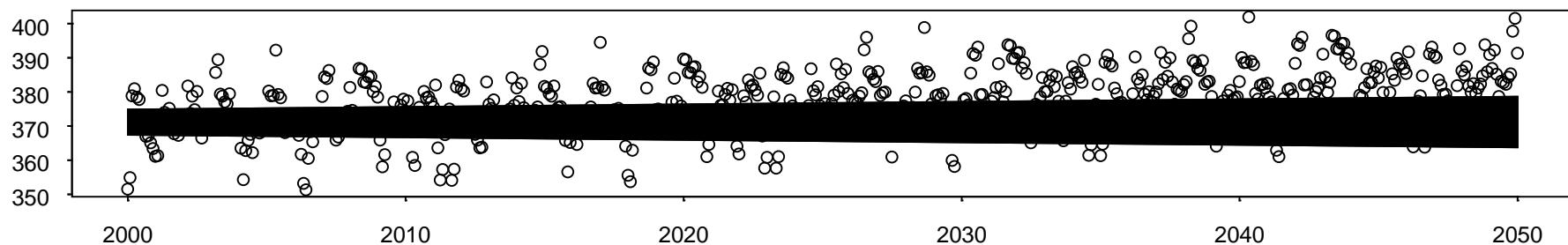
GSFC Predictions - without climate change



GSFC Predictions with SBUV Lowess Residuals



with +/-1% error plus +/-1% drift



Incorporating Long Term Stability estimates in our estimate of our ability to detect trends

- In some cases, our measurement uncertainty is considerably larger than the signal we want to detect.
- Estimating appropriate measurement uncertainty over decades of monitoring is extremely difficult.
- Measurement stability and statistical variability are likely to be independent, thus:

$$\sigma^2_{\text{total}} = \sigma^2_{\text{statistical}} + \sigma^2_{\text{stability}}$$

For temperature: 0.1 may add ten years to monitoring

For humidity: uncertainty is larger.

Identifying a metric gives guidance to decision making and resource allocation.

We can control only four aspects of monitoring to detect trends

- **What we monitor**
- **What accuracy**
- **Where we monitor**
- **What frequency**

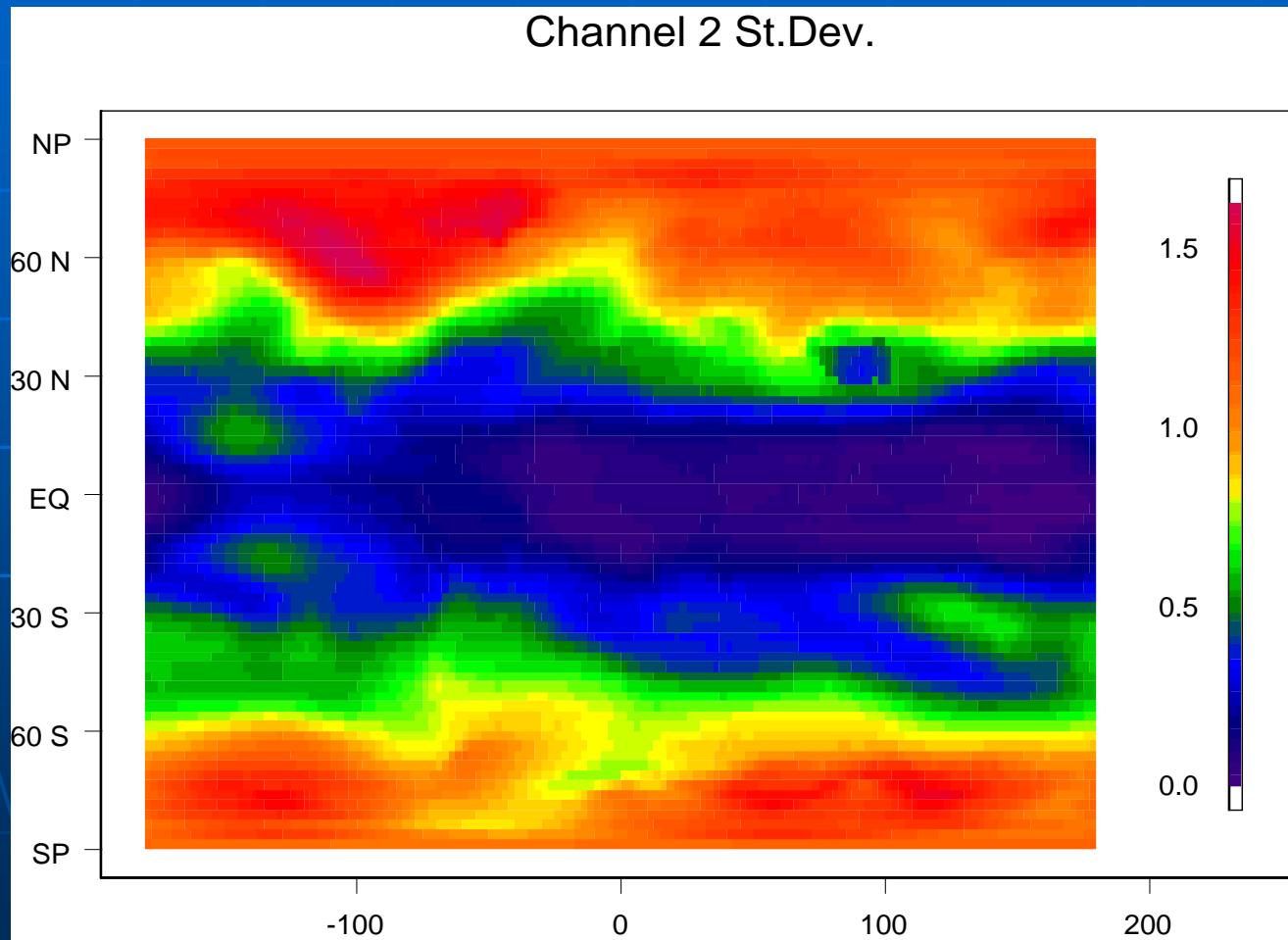
Where do we monitor?

- Some places are inherently better for detecting trends than others.
- Monitoring by satellite involves averaging over height, longitude and latitude.
 - Measurement smoothing can damage our ability to detect trends

Where do we monitor: single locations

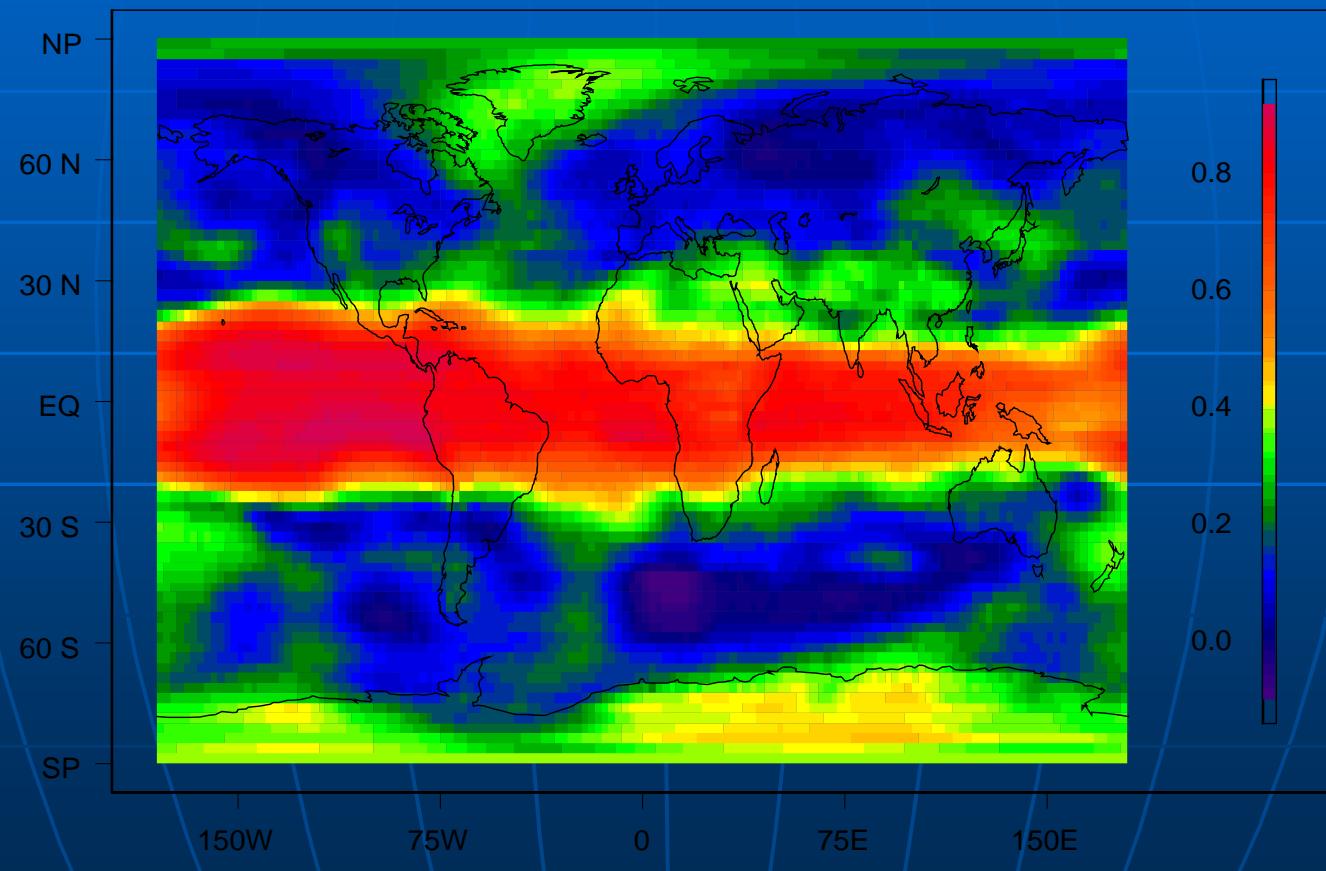
- Some places are inherently better for detecting trends than others.
- Natural variability, memory and magnitude of trend vary by location
- The difference in number of years can vary by more than a factor of two.

Magnitude of temperature swings

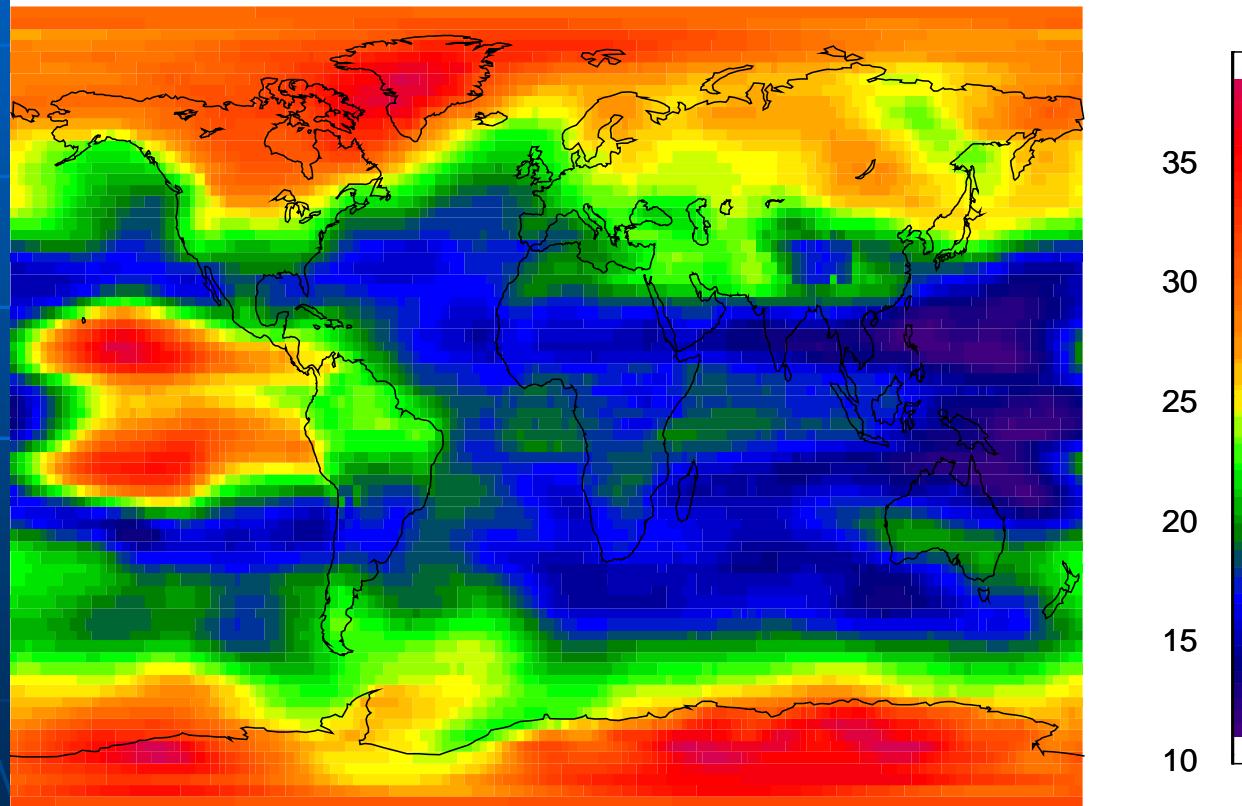


Autocorrelation of monthly data

Channel 2 Autocorrelation

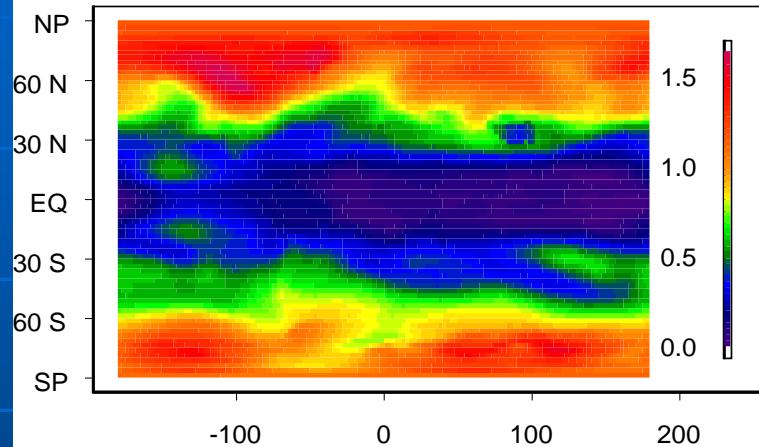


Channel 2: Years to Detect 0.2 deg/decade

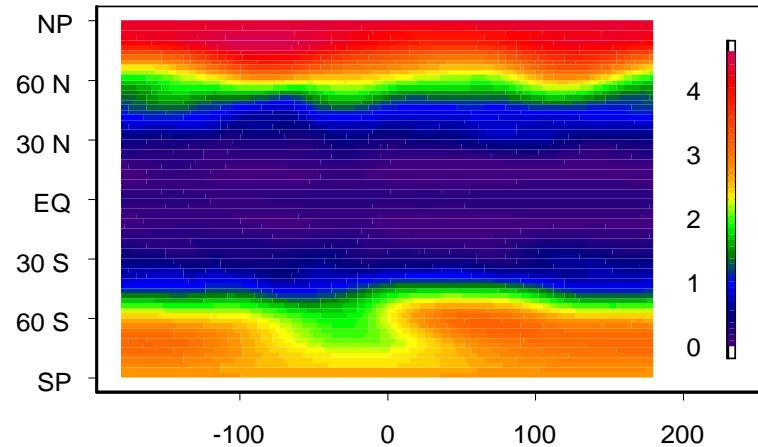


MSU channels 2 & 4 characteristics

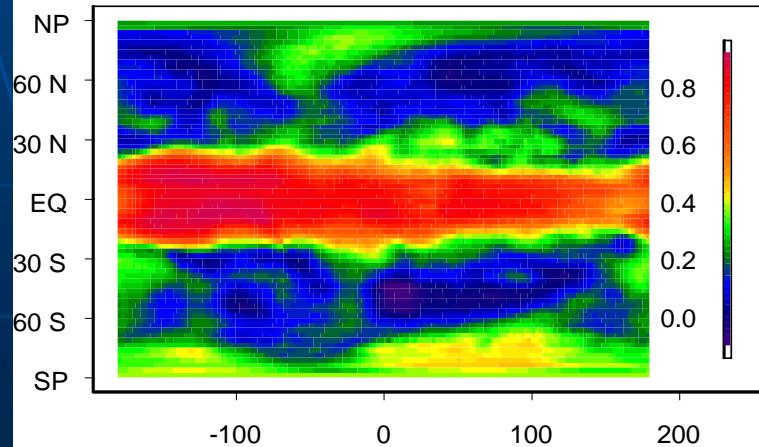
Channel 2 St.Dev.



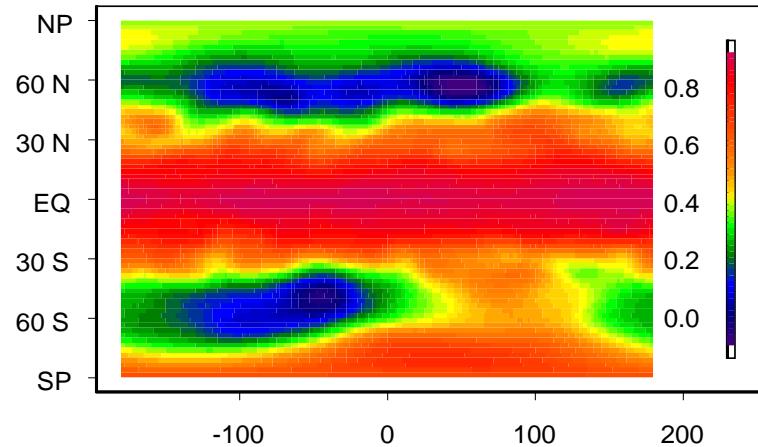
Channel 4 St.Dev.



Channel 2 Autocor.



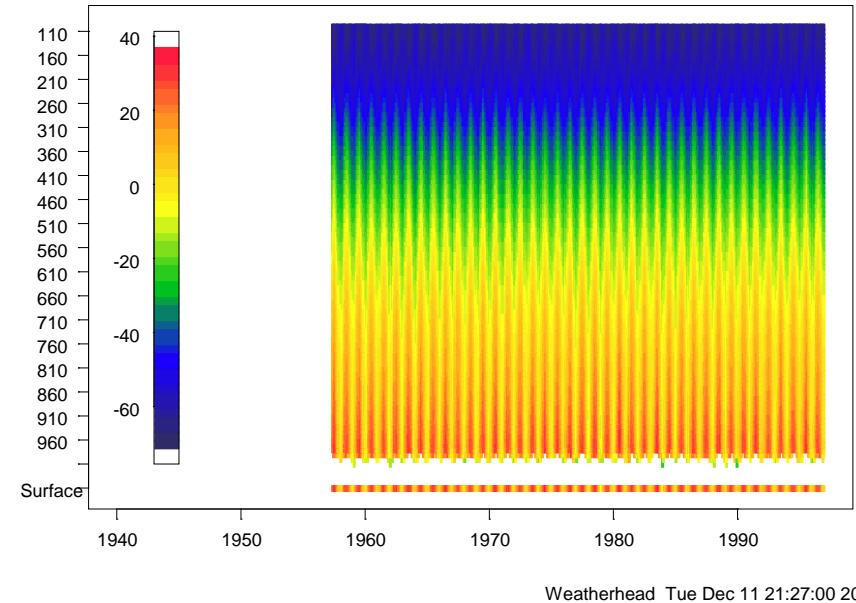
Channel 4 Autocor.



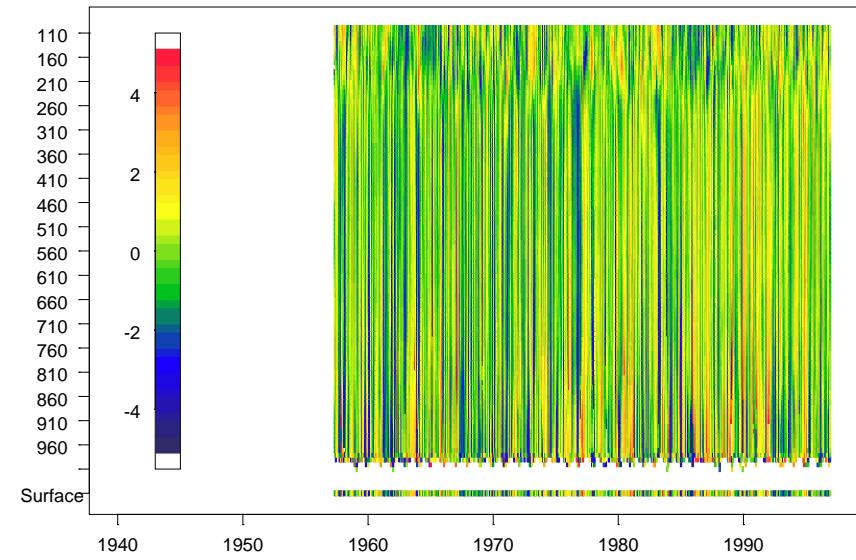
Looking in the vertical

- Near constant trends are predicted throughout the troposphere
- Is there an optimal place to detect trends?

Topeka Sonde Temperature Data at 0 Z

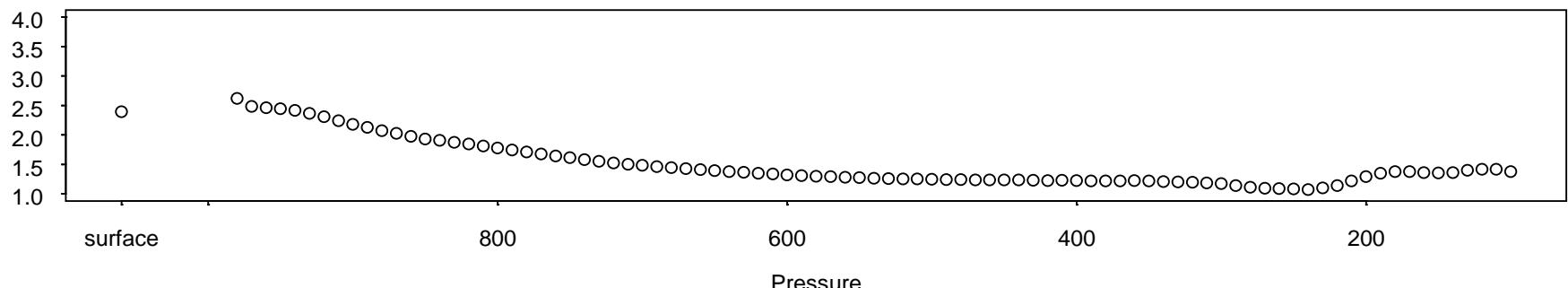


Topeka Sonde Temperature Data at 0 Z

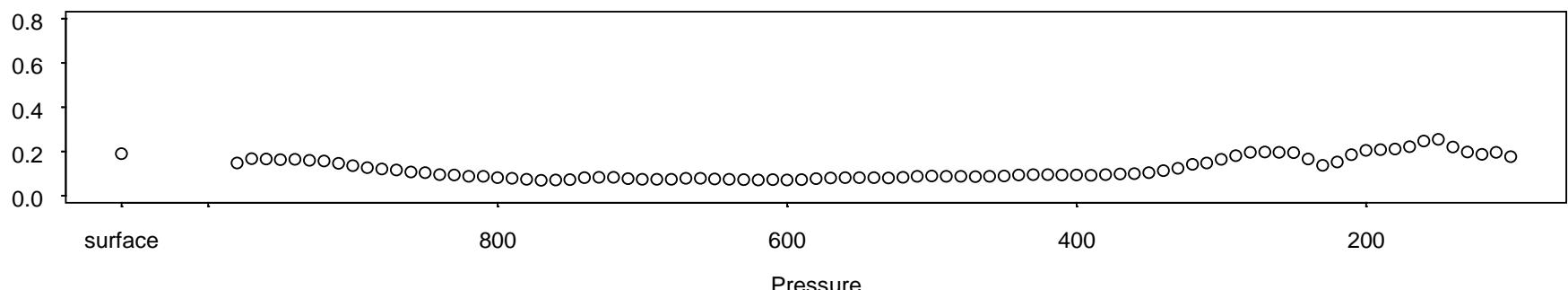


Topeka temperature detectability

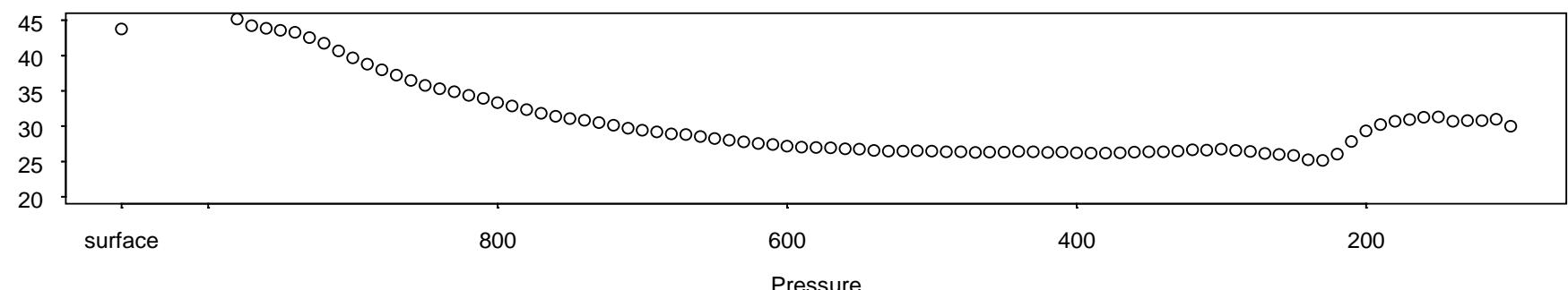
Noise



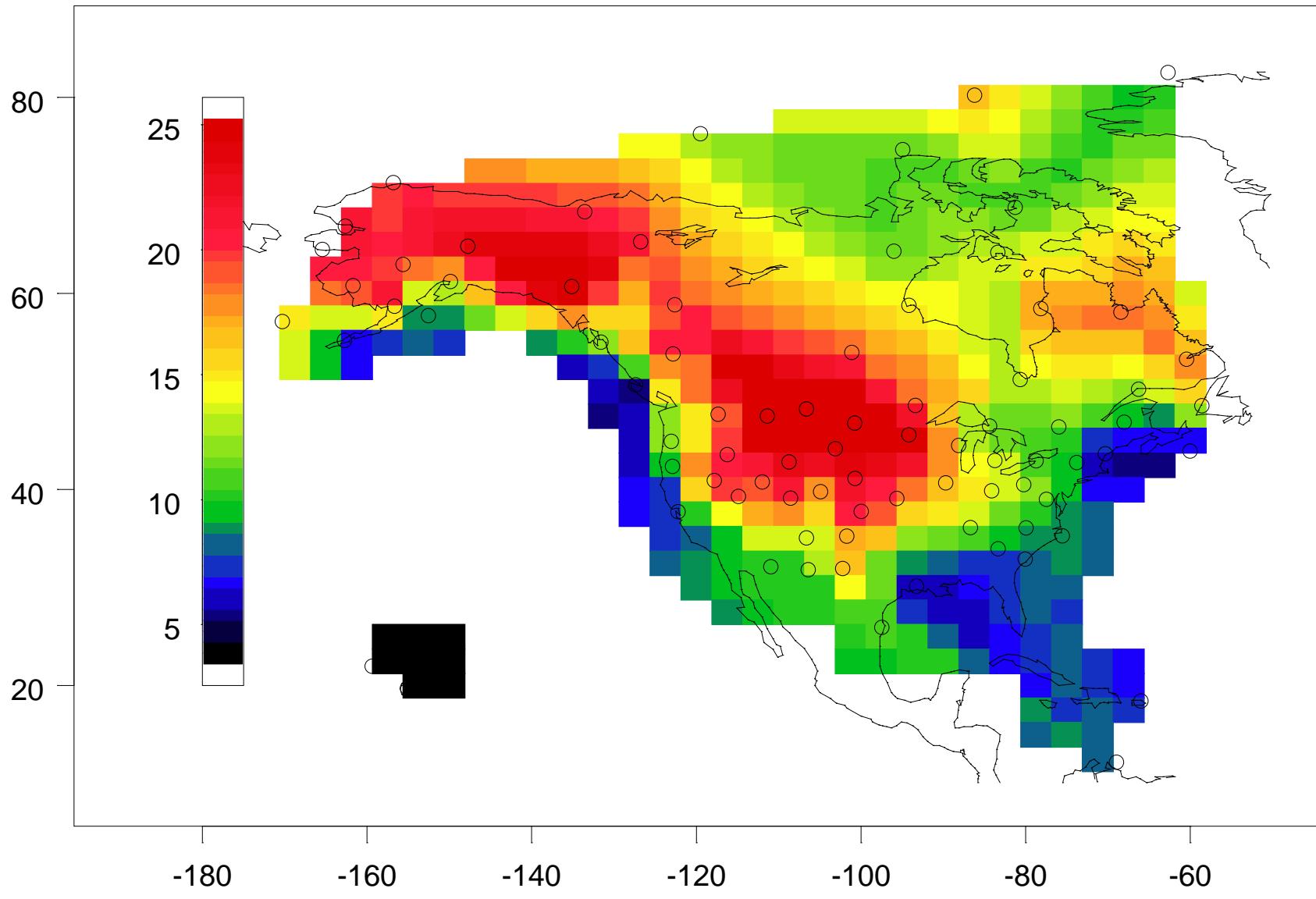
Autocorrelation



Years to Detect a .2 deg/decade trend



Years Saved by Monitoring Free Troposphere OZ

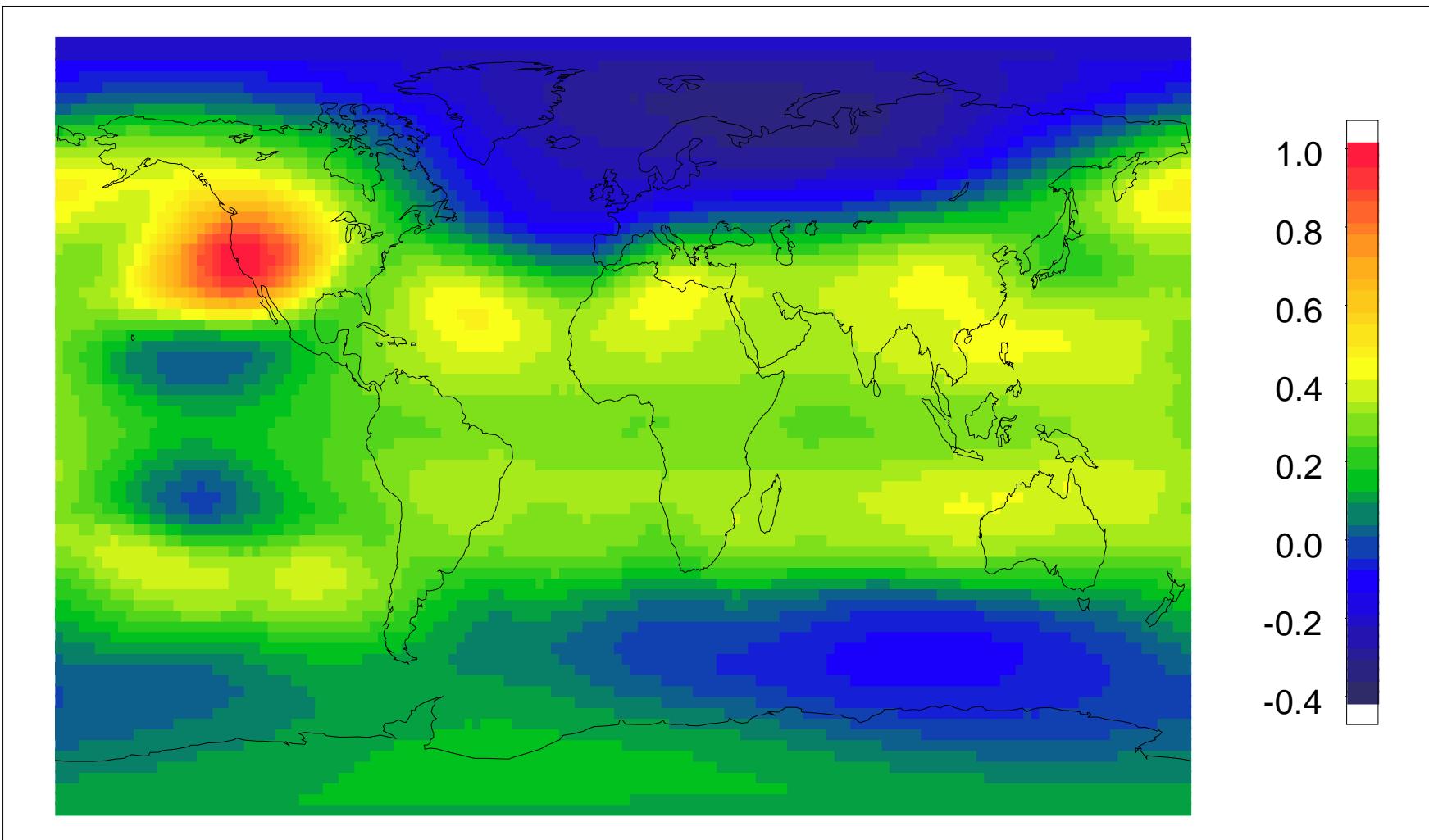


Weatherhead Tue May 21 12:14:04 2002 True Range of Data: (3 to 36)

How many single stations do we need?

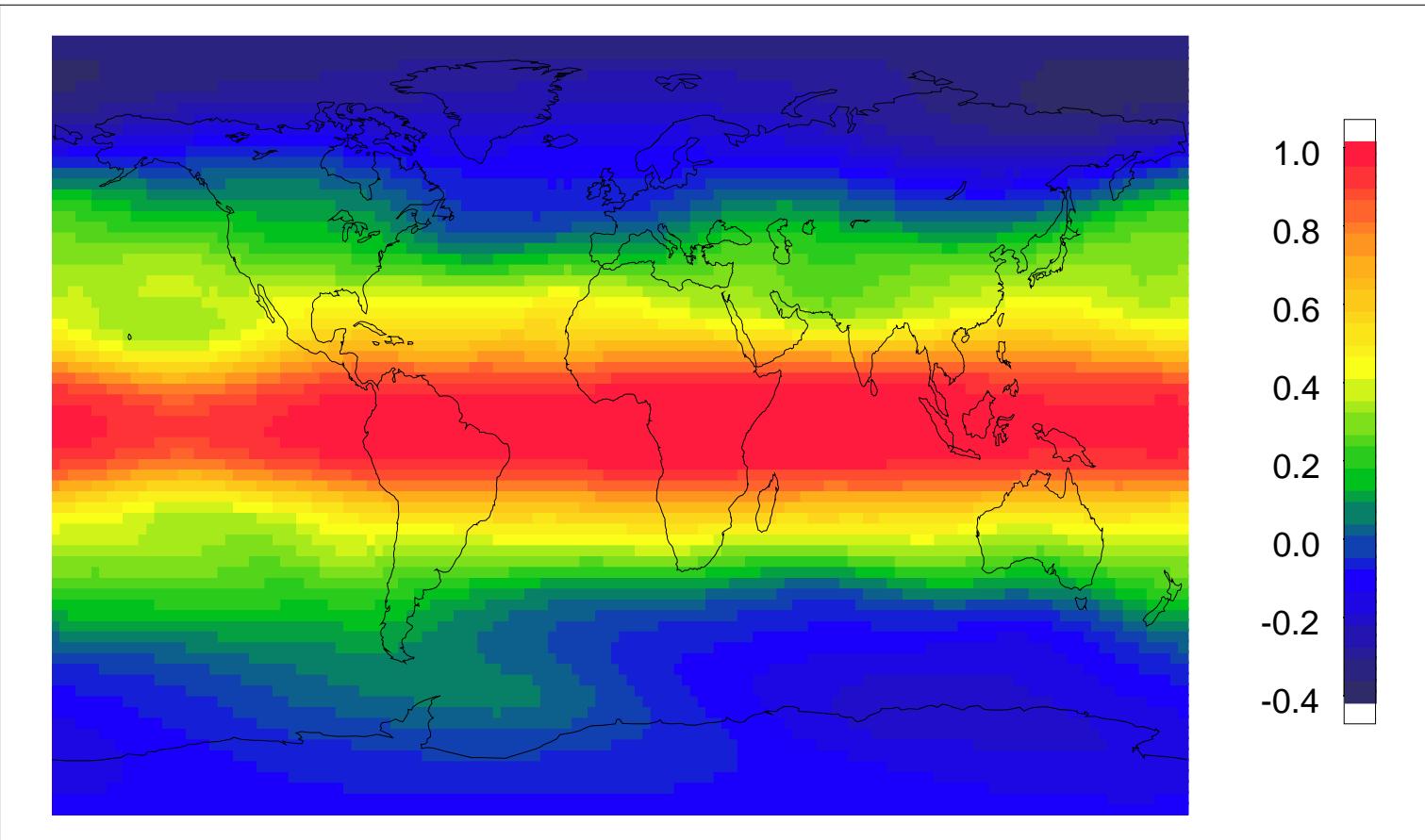
- Spatial coherence means that averaging many different locations does not always reduce error bars significantly.
- Spatial coherence can be estimated from past data.

MSU Channel 4 Correlation with S.F.



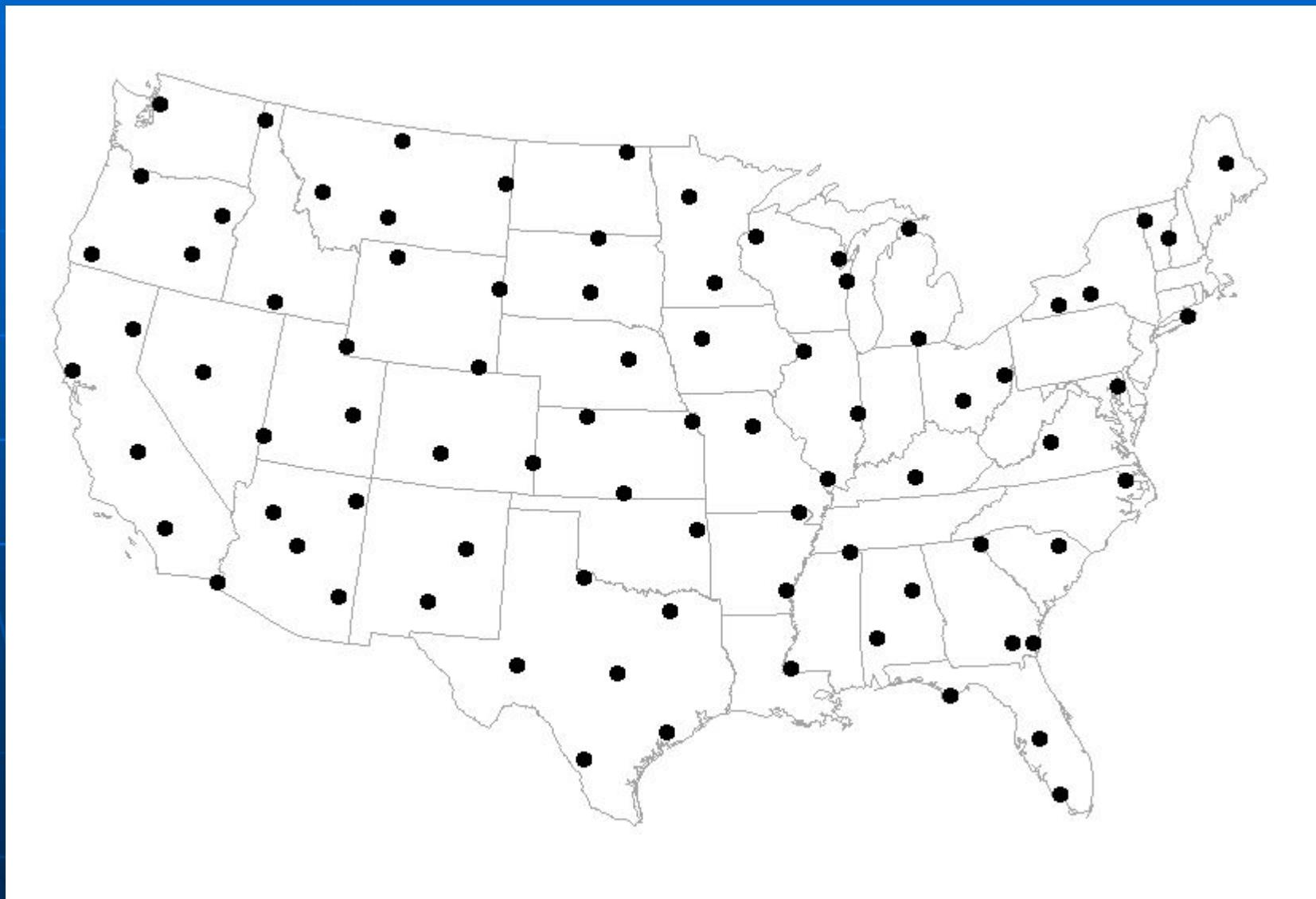
MSU Channel 4

Correlation with lat=0 and long=0



How does spatial
redundancy affect our
ability to detect trends?

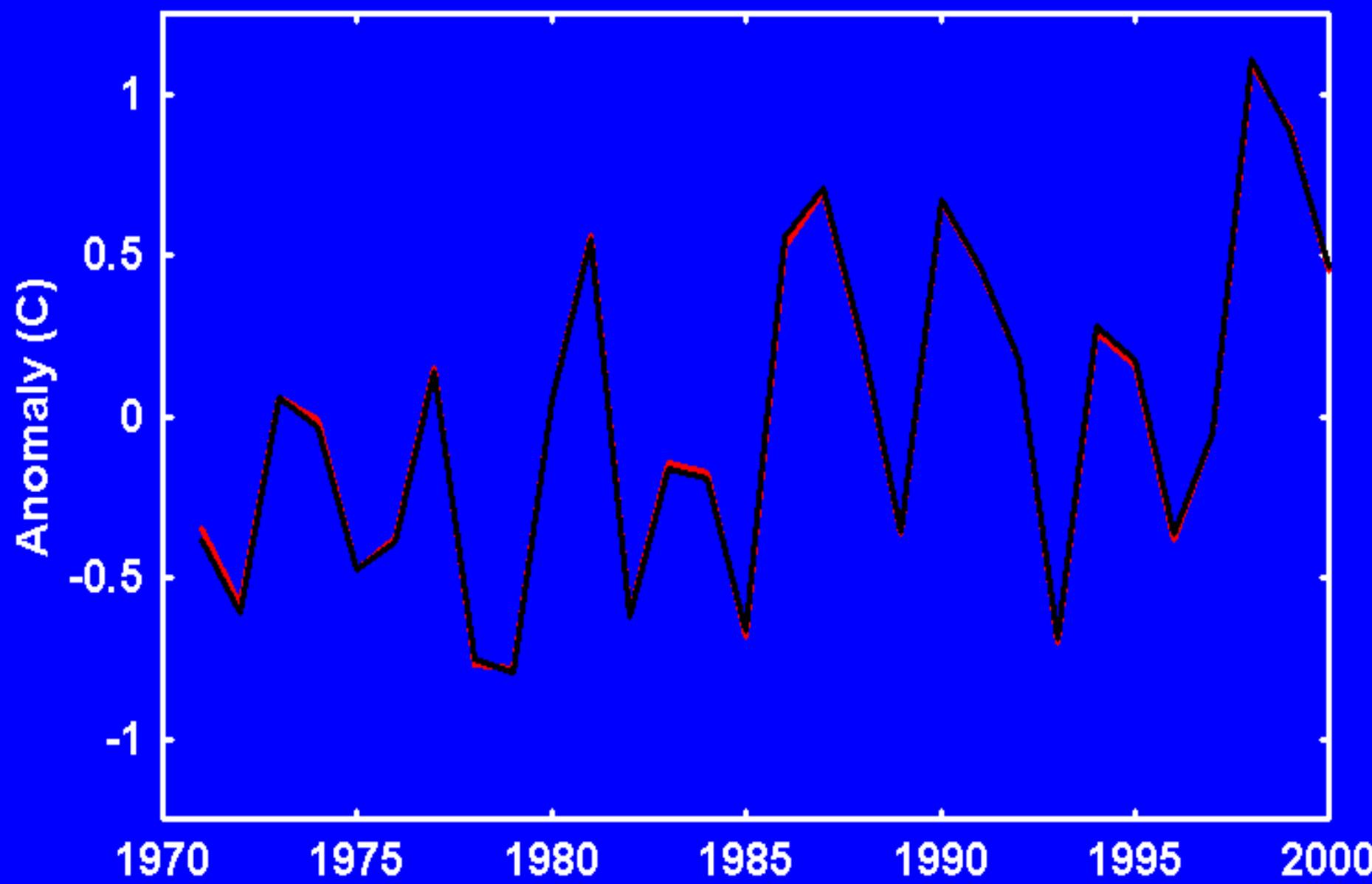
82 Station Subset of HCN Network



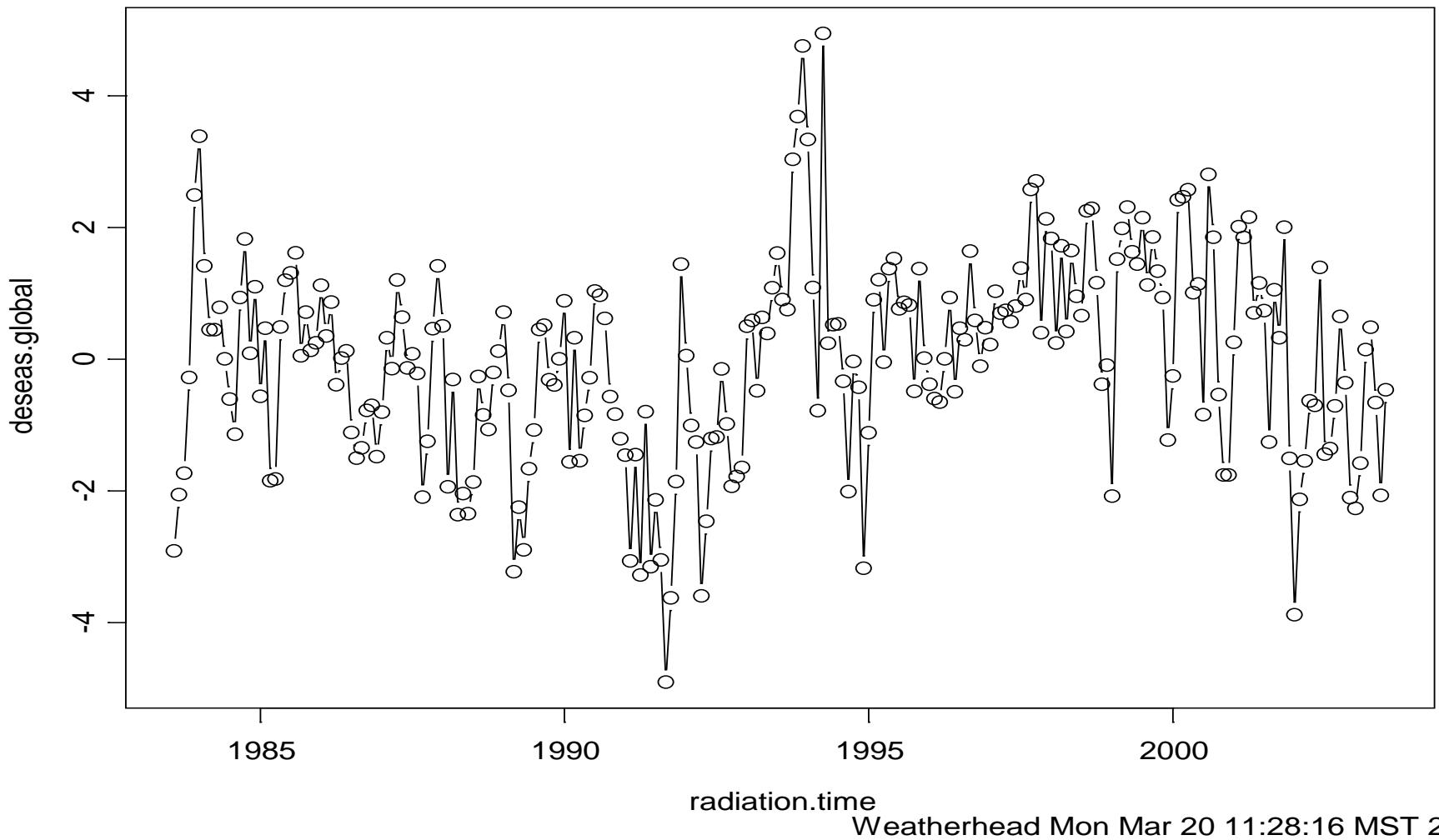
225 Station Subset of HCN Network



U.S. Annual Temperature Series

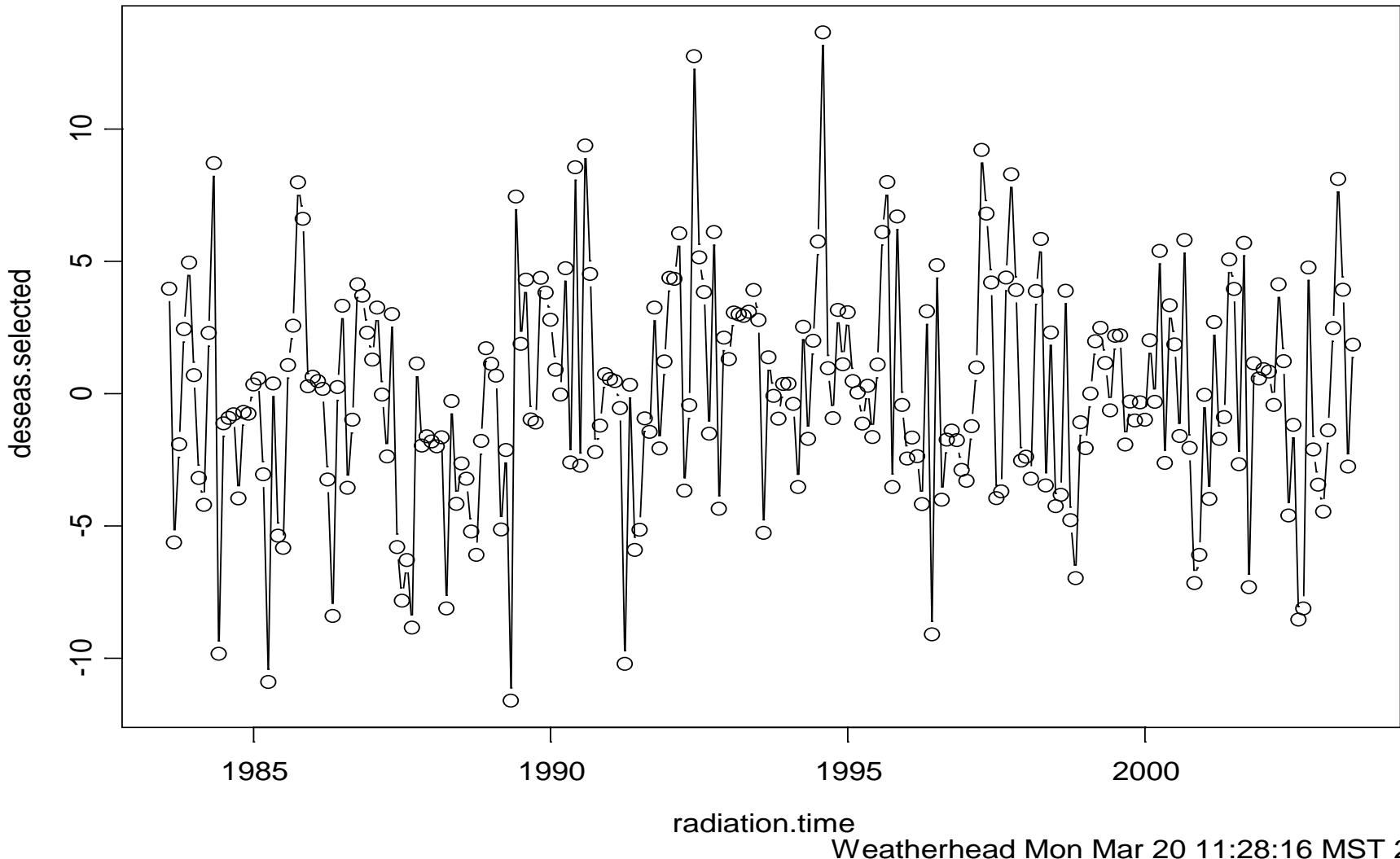


Deseasonalized Global (W m⁻²)



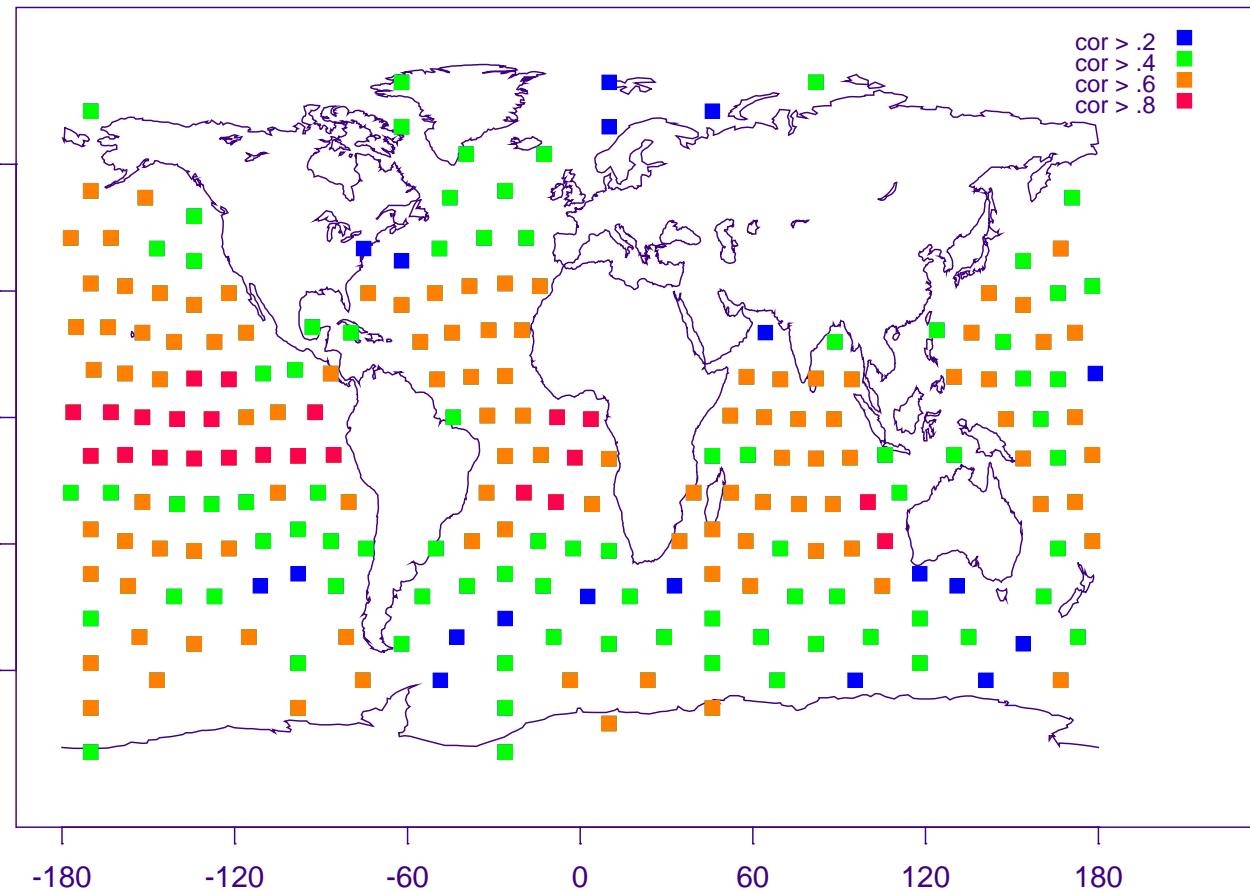
From Laura Hinkelman

Deseasonalized Selected (W m⁻²)



From Laura Hinkelman.

Redundancy in Surface Data for Proposed Locations.



Choosing locations for monitoring

- Some locations are better than others for detecting trends.
- Sub-regional differences are likely small.
 - Define spatial scales
- What is the primary goal:
 - Estimating global change
 - Annual/seasonal/diurnal trends?
 - Understanding specific regional change
 - QBO, AO, NAO, ENSO effects?
 - Transferring standards/undersanding errors.
- Once the goal is clear, we can quantify the best approach.

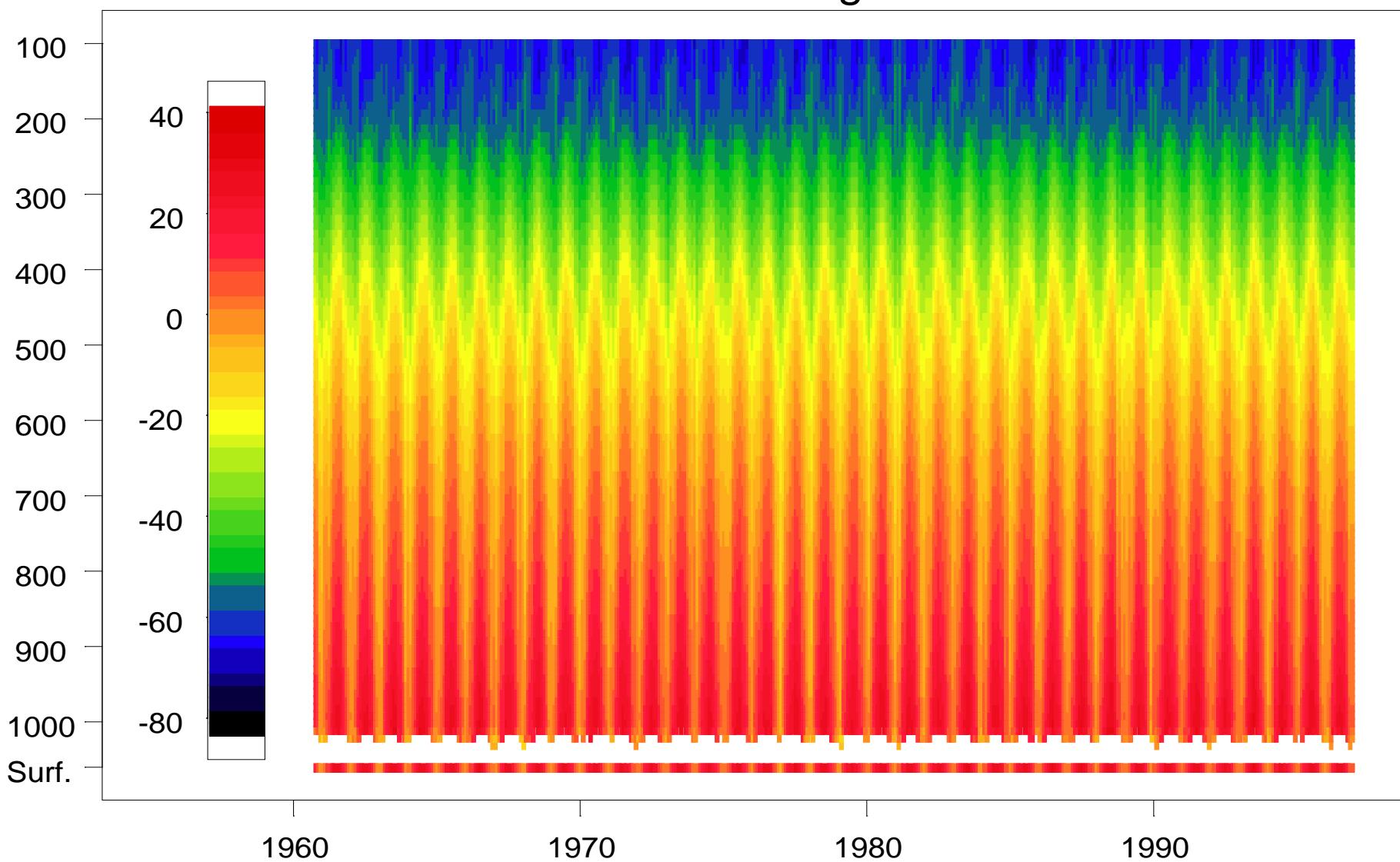
We can control only four aspects of monitoring to detect trends

- **What we monitor**
- **What accuracy**
- **Where we monitor**
- **What frequency**

What frequency?

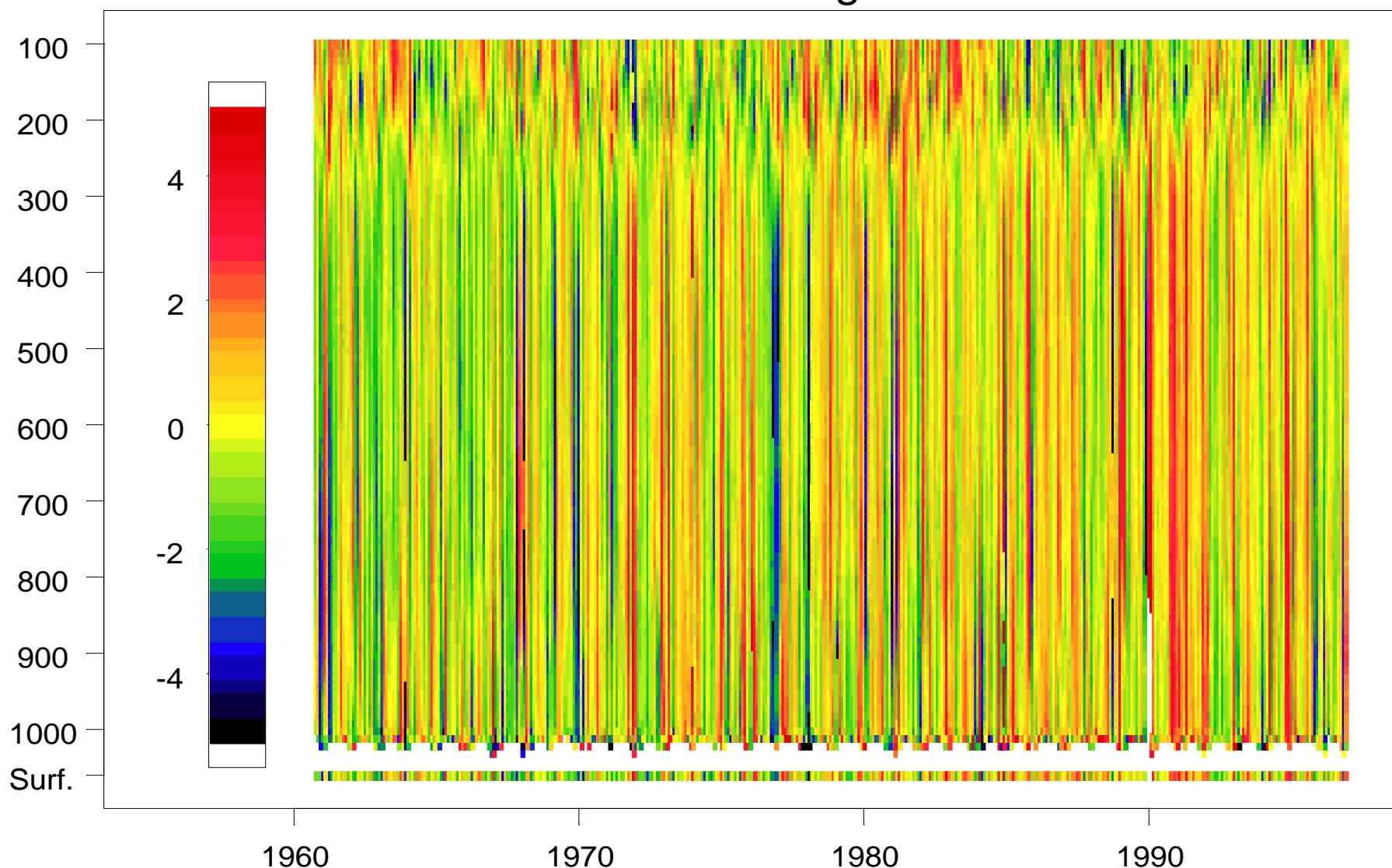
- Inherent memory in environmental data results in redundancy of measurements.
- Daily data may be more than needed.
- Less than daily measurements may obscure diurnal changes

STERLING(WASH DULL 0 Z temp
Lat. = 38.98 Long. = -77.47



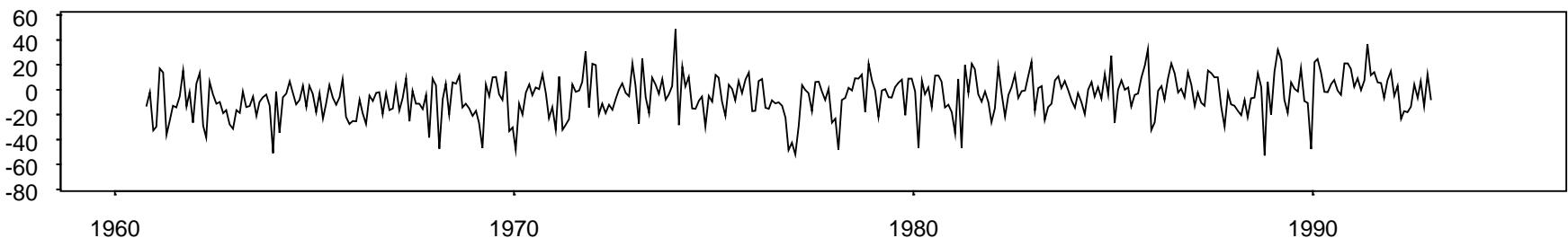
Weatherhead Wed Mar 13 13:30:52 200

STERLING(WASH DULL 0 Z
Lat. = 38.98 Long. = -77.47

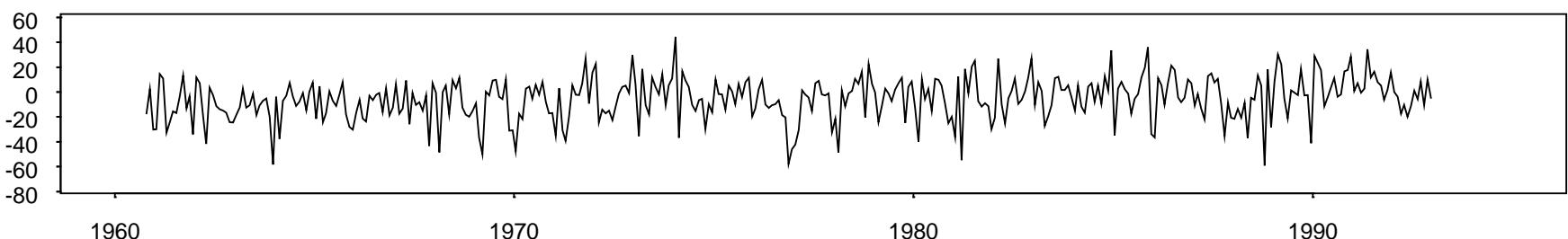


Monthly Deseasonalized Averages, 500 mb Temperature, Dulles

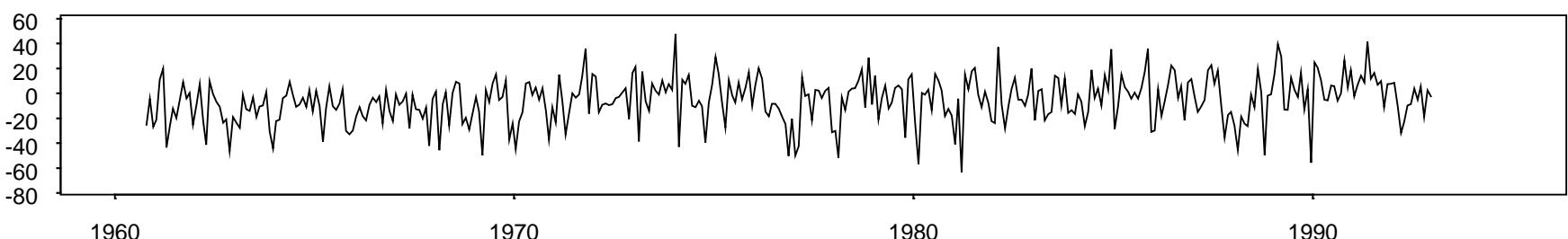
Measure Every 1 Days



Measure Every 2 Days

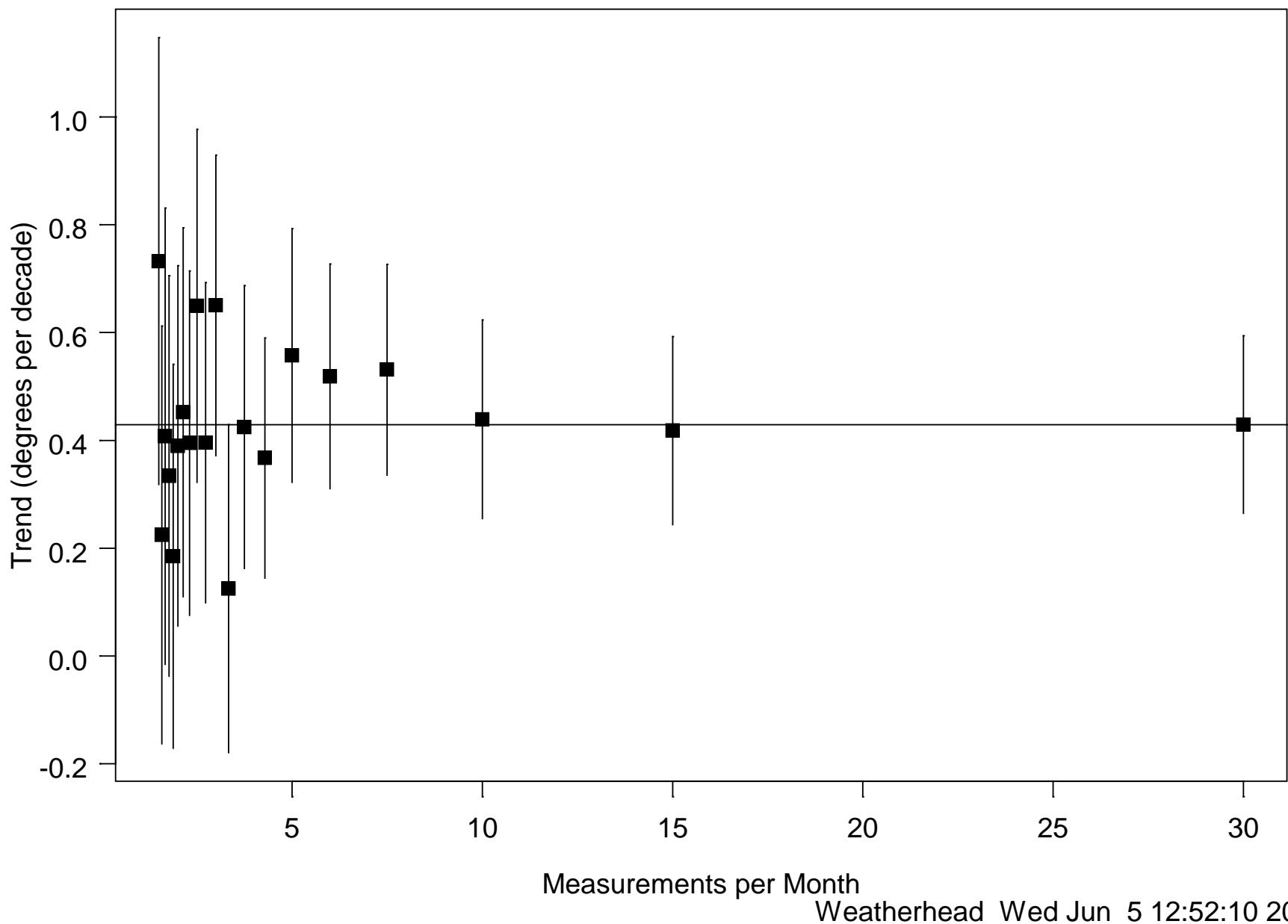


Measure Every 3 Days



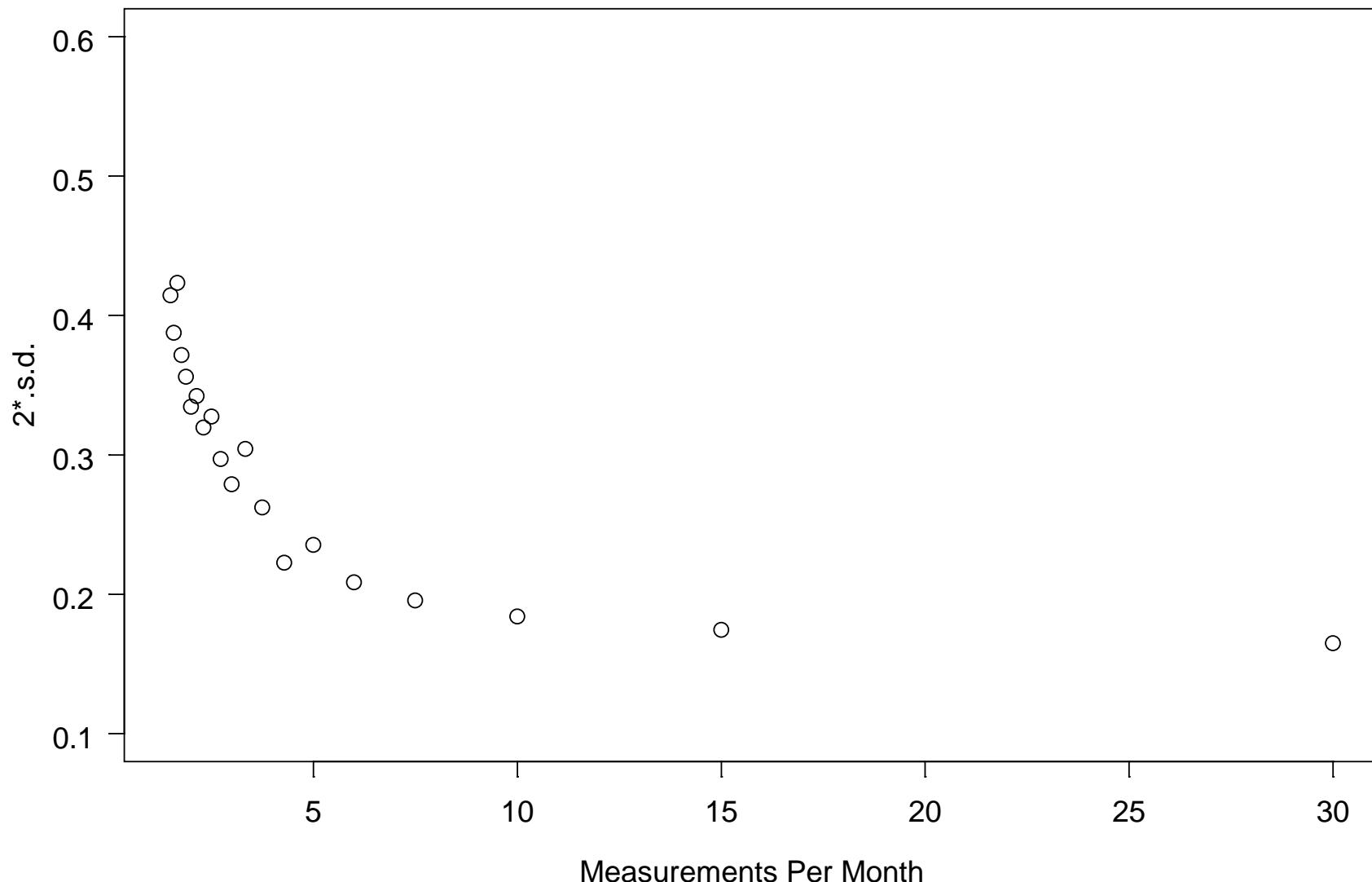
How do the trends change
when we take data less
frequently than every day?

500 mb Temperature Trend, Dulles



How do the error bars on
our trends change when we
take data less frequently
than every day?

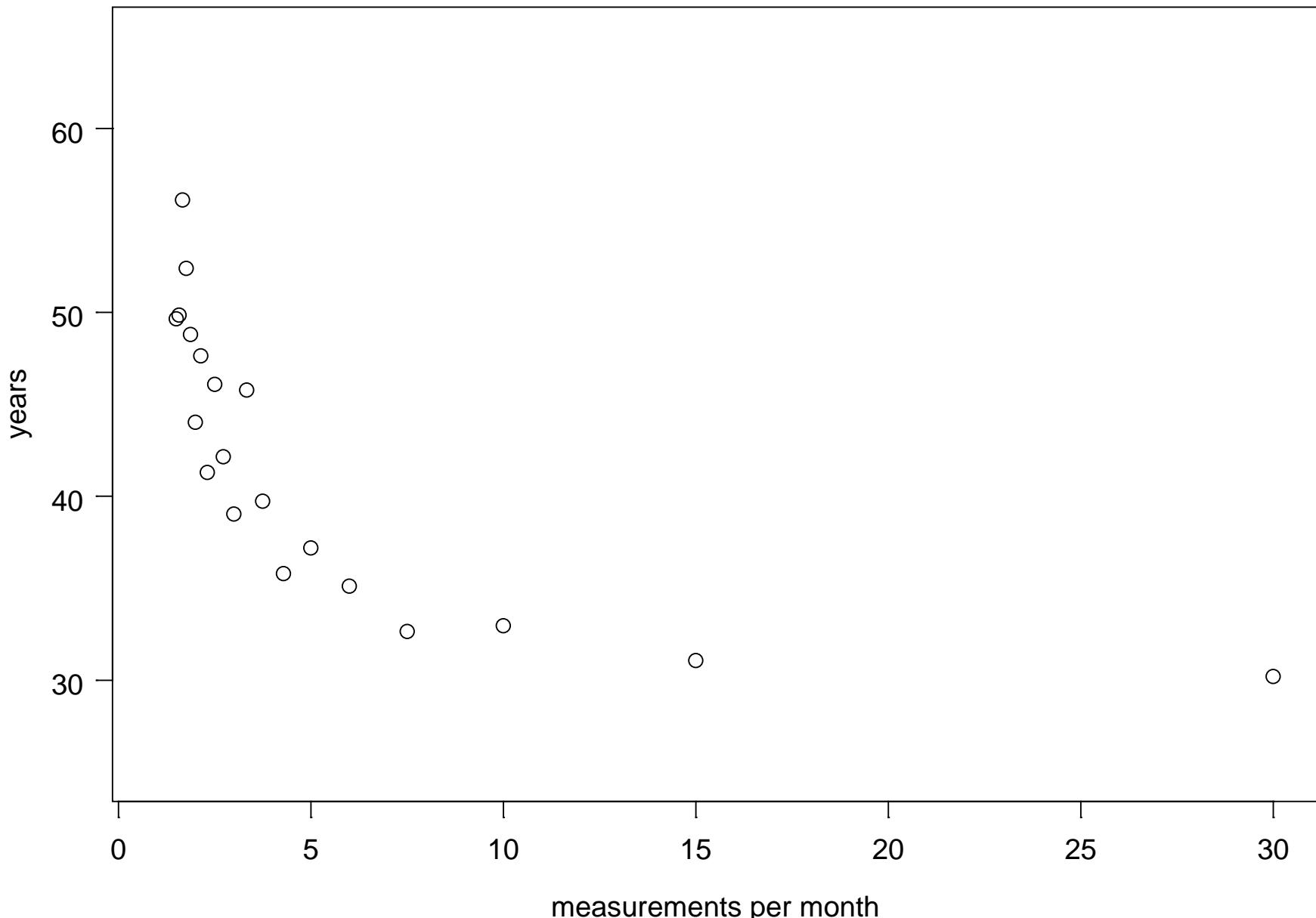
Estimated Error in 500 mb Temperature Trend (2 sigma) Dulles



How long will it take to detect trends?

Years to Detect 0.2 degrees per Decade

Dulles 0Z 500 mb



Weatherhead Wed Jun 5 12:45:30 2002

How does frequency of measurement affect how long we will have to monitor to detect trends?

In general: Monitoring less frequency:

- Increases magnitude of variability (bad for trends)
- Decreases autocorrelation (good for trends)
- Reduces representativeness (do we really know what happened?)

Decreasing the data frequency

- We can estimate the cost (in number of years of monitor) to decrease the data frequency.
- Decreasing the data frequency can reduce our ability to:
 - Detect extreme events
 - Detect diurnal (or perhaps seasonal) signals

We can control only four aspects of monitoring to detect trends

- **What we monitor**
- **What accuracy**
- **Where we monitor**
- **What frequency**

Integration

- We make choices about all four of the parameters we control.
- These choices have direct impact on how long we will likely need to monitor in order to detect trends.
- Optimal choices exist.
 - e.g. More sites or higher accuracy?
- All choices will affect our ability to detect trends and the scientific questions we may ask of the emerging data.

Improved Accuracy or More Sites?

- **Improved Accuracy**

- Clearer understanding of what we've measured
- Costs often increase exponentially
- Time for trend detection decreases

- **Additional Sites**

- Costs increase in a known manner
- Time for trend detection decreases - usually slightly
- Representativeness improves and expands
- "Insurance" for site failures

Conclusion

1. We can control only four aspects to detect trends:
 - What we monitor; Where we monitor;
 - What frequency; What accuracy
2. We can optimize systems when we have a clear goal.
 - Establish accuracy and transfer ability
 - Establish global trends
 - Establish regional trends
 - Monitor key climate features
3. Optimization provides the following benefits:
 - Answering scientific questions earlier
 - Confirming, improving models
 - Allowing for earliest policy decisions
 - Maintaining prudent use of available funds