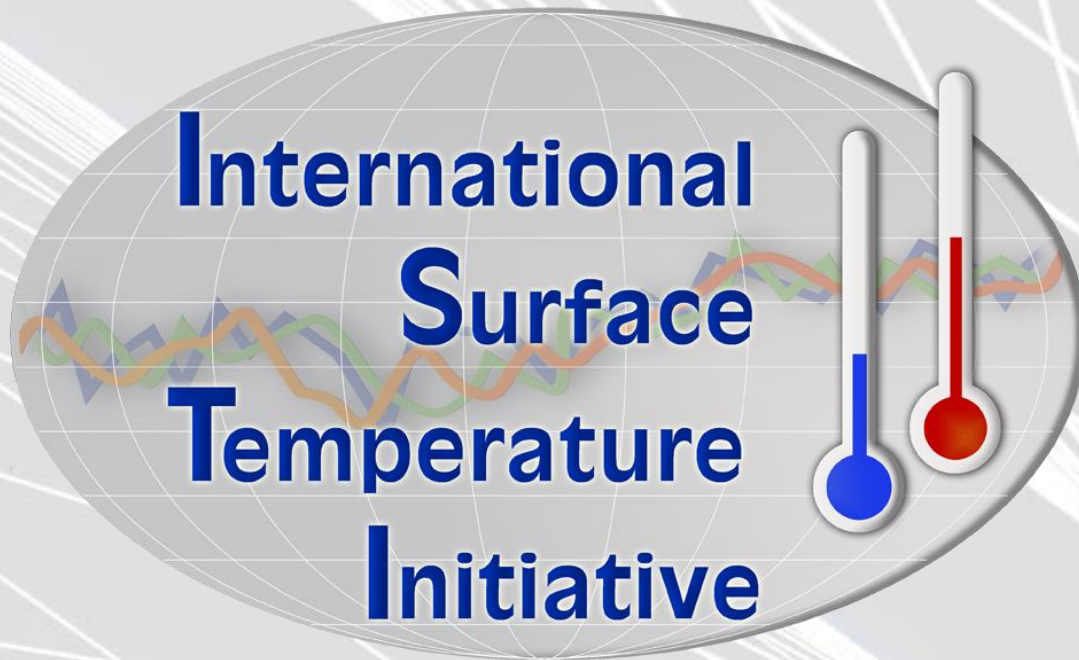


# The ISTI: Land surface air temperature datasets for the 21st Century



WRCP Extremes, Feb 2015

**Kate Willett**, with thanks to many Initiative participants

# The real world observing system is not perfect ...



US Climate Reference Network website

# Its more like these ...



Huge range of instrument types, siting exposures etc. regionally, nationally and globally with many changes over time.

More examples on [www.surfacestations.org](http://www.surfacestations.org)

# Effects of Changes that are not of Climate Origin



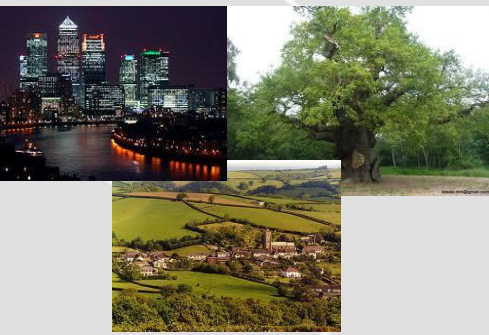
**STATION MOVE:** EXPOSURE AND MICROCLIMATE = abrupt change in mean and diurnal extremes - may affect seasonal cycle extremes

**SHELTER CHANGE:** EXPOSURE = abrupt change in diurnal extremes - may affect seasonal cycle extremes

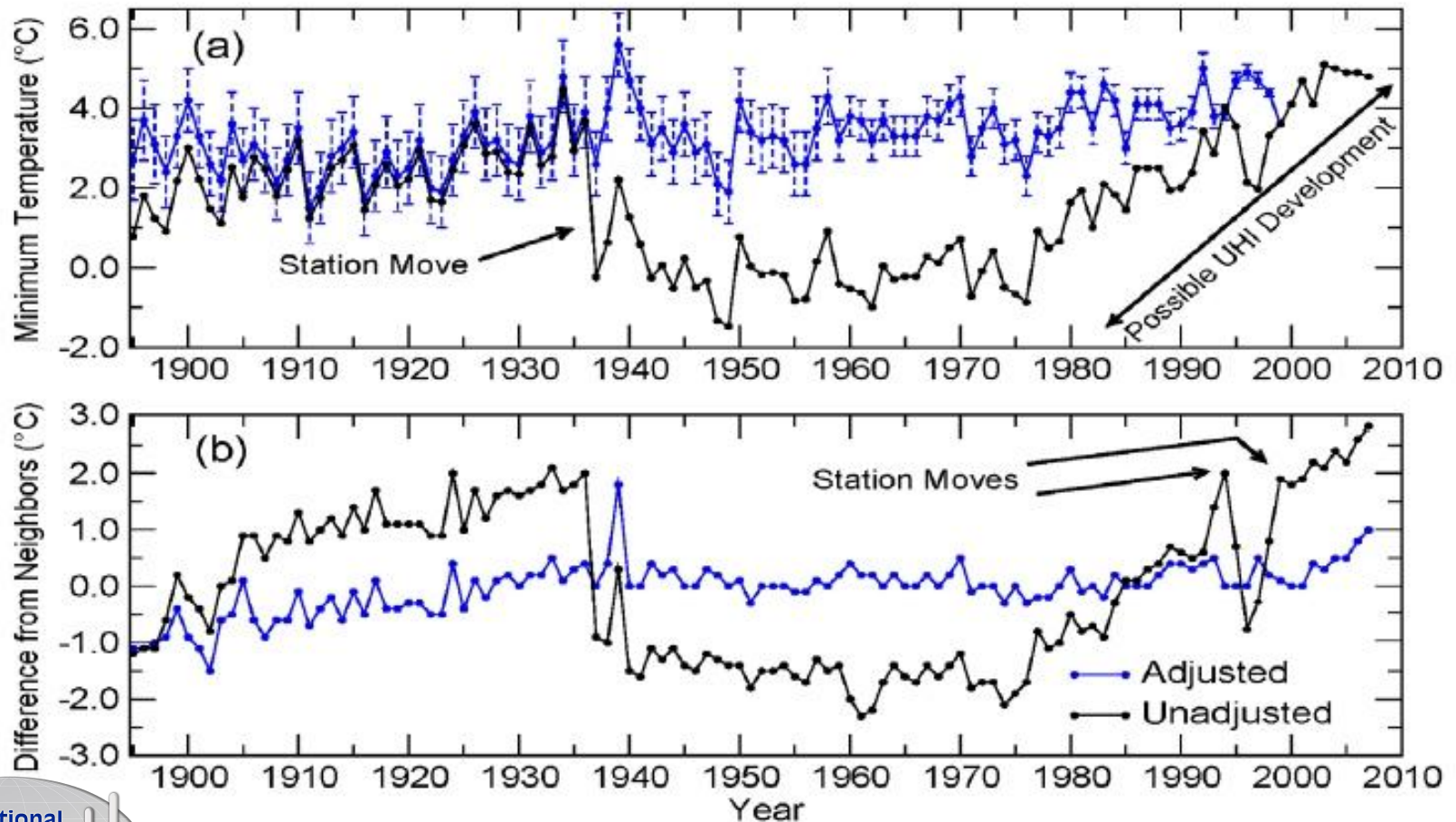
**OBSERVING PRACTICE CHANGE:** SAMPLING = abrupt change possible in mean and extremes

**INSTRUMENT CHANGE:** CALIBRATION = abrupt change in mean and possibly extremes

**LANDUSE CHANGE:** EXPOSURE AND MICROCLIMATE = gradual change in mean and diurnal extremes - may affect seasonal cycle extremes



# Inhomogeneities: annual mean minimum temperature at Reno, Nevada, USA

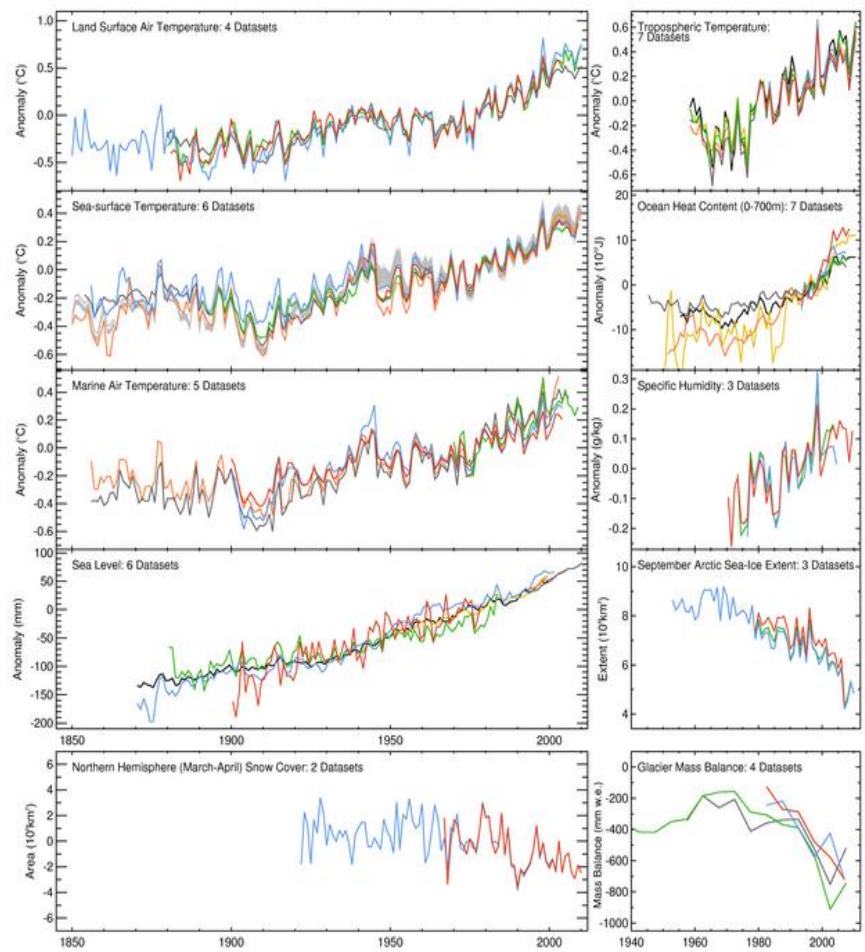


(Matt Menne and Claude Williams, NOAA National Climatic Data Center)

# Underlying these are four fundamental issues ...

- A lack of **traceability** to known standards and original hard copy data sources for most historical records
- A lack of adequate **documentation** of the (ubiquitous) changes (station location, shelter, observing time etc.) sufficient to characterize their changing measurement characteristics
- A lack of '**one-stop-shop**' for all land meteorological data (like ICOADS for land) – both raw and CDR/value-added-products
- A lack of set **benchmarks** with which to comprehensively test Climate Data Record development

# No doubt that it is warming – the rate and temporal / spatial details are the issue



Is our climate changing?

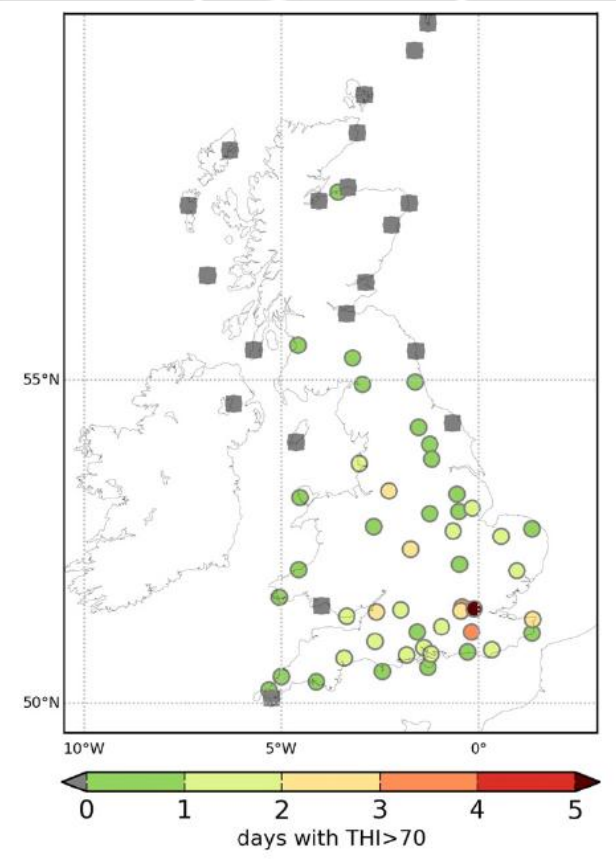
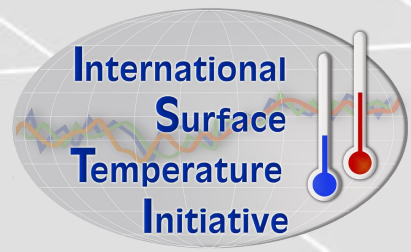


Figure 2. The average number of days (over 1973–2012) with THI > 70 for each of the 68 stations. Stations with no days with THI > 70 are shown as grey squares.

What about the cows?

# ISTI: Creating a framework to enable advances

1. Basic environmental data provision
2. Benchmark assessment of uncertainty relating to methodological choices
3. User advice

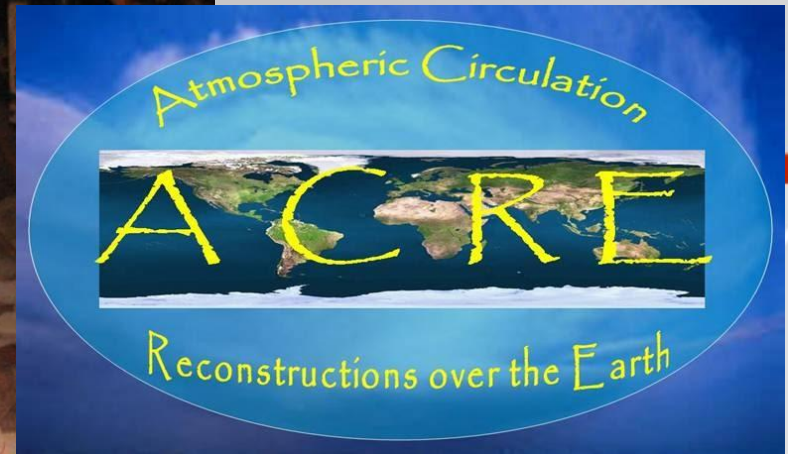




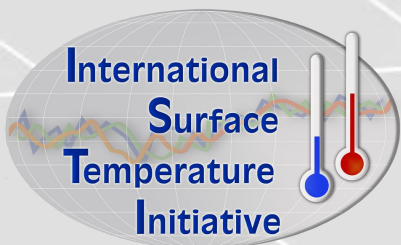
# Step 1: Data rescue and provision



Jay Lawrimore, Jared Rennie and Peter Thorne (2013) Responding to the Need for Better Global Temperature Data, EOS, 94 (6), 61–62  
DOI: 10.1002/2013EO060002



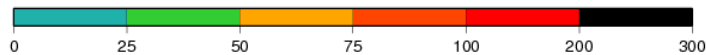
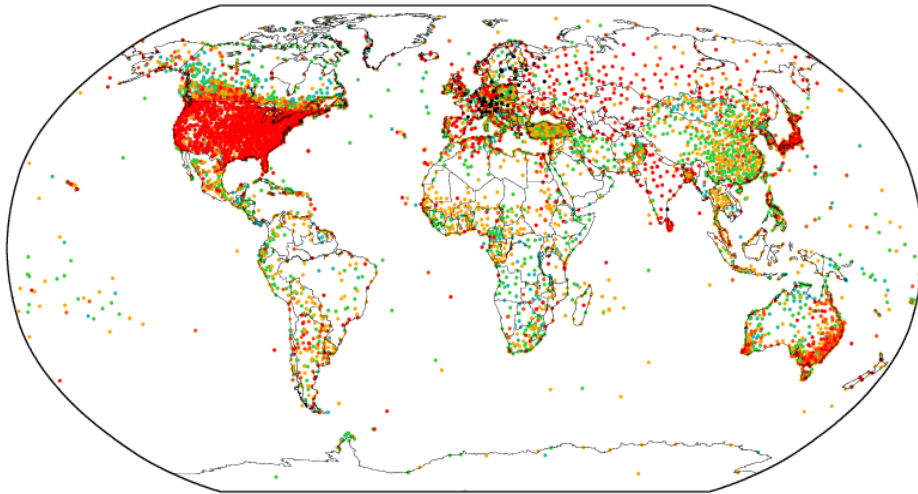
[www.met-acre.org](http://www.met-acre.org)



# ISTI Stage 3 vs GHCNv3 data

## GHCNM v3

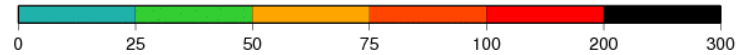
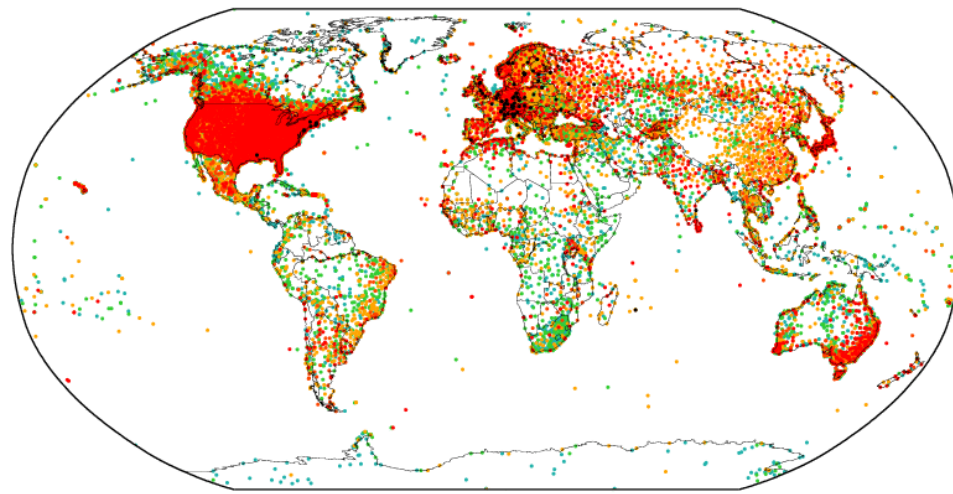
Number of Stations: 7280



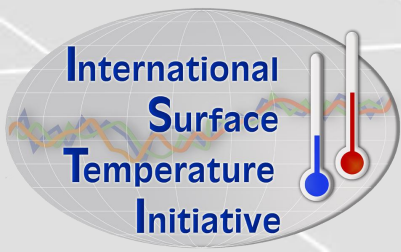
Number of Years

## Stage Three (Recommended Merge)

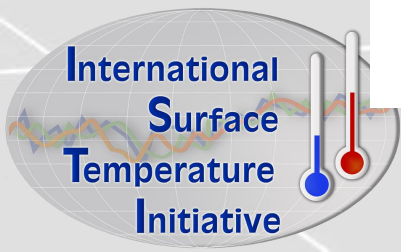
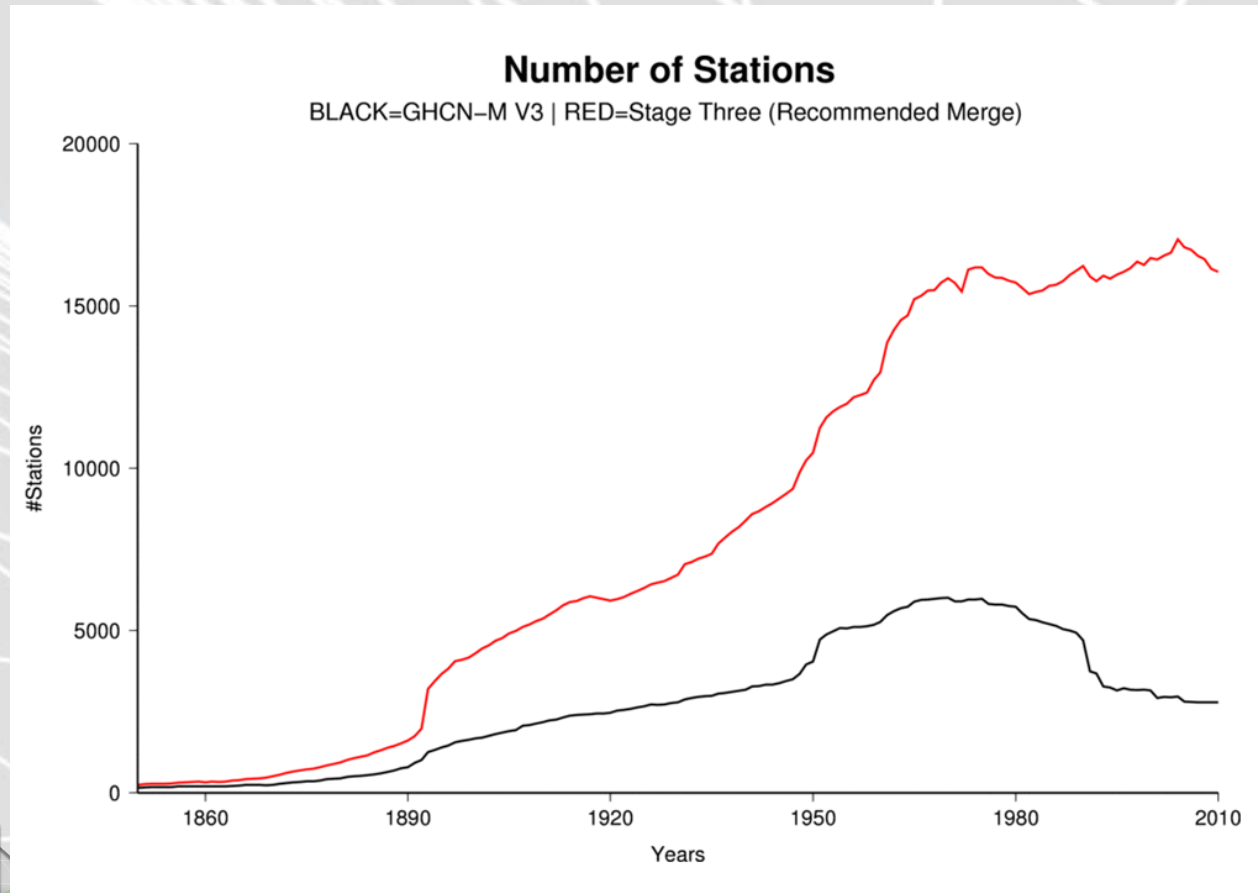
Number of Stations: 31999



Number of Years

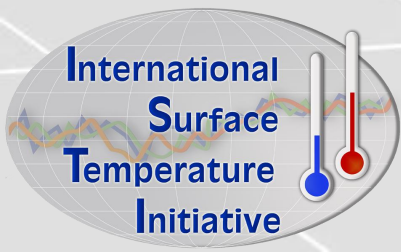


# More than a little better?



# Step 2: Benchmarking and Assessment

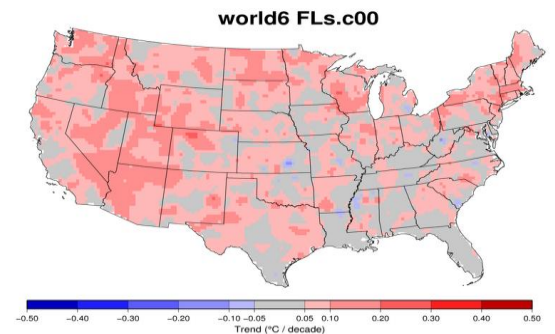
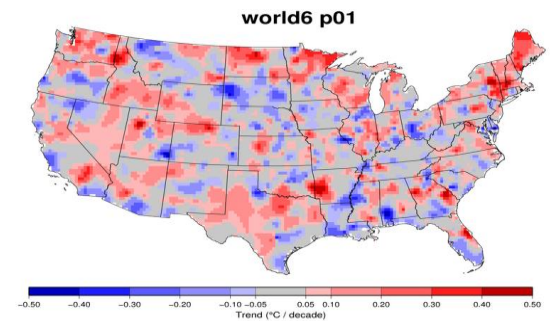
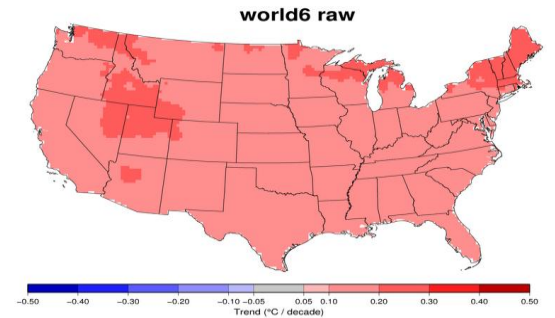
- With real world data we do not have the luxury of knowing the truth – we CANNOT measure performance of a specific method or closeness to real world truth of any one data-product.
- We CAN focus on performance of underlying algorithms (AKA software testing)
- Consistent synthetic test cases, simulating real world noise, variability and spatial correlations potentially enable us to do this



# Benchmarking Cycle

## Create c.10 analog-error-worlds

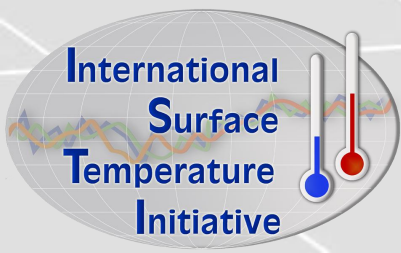
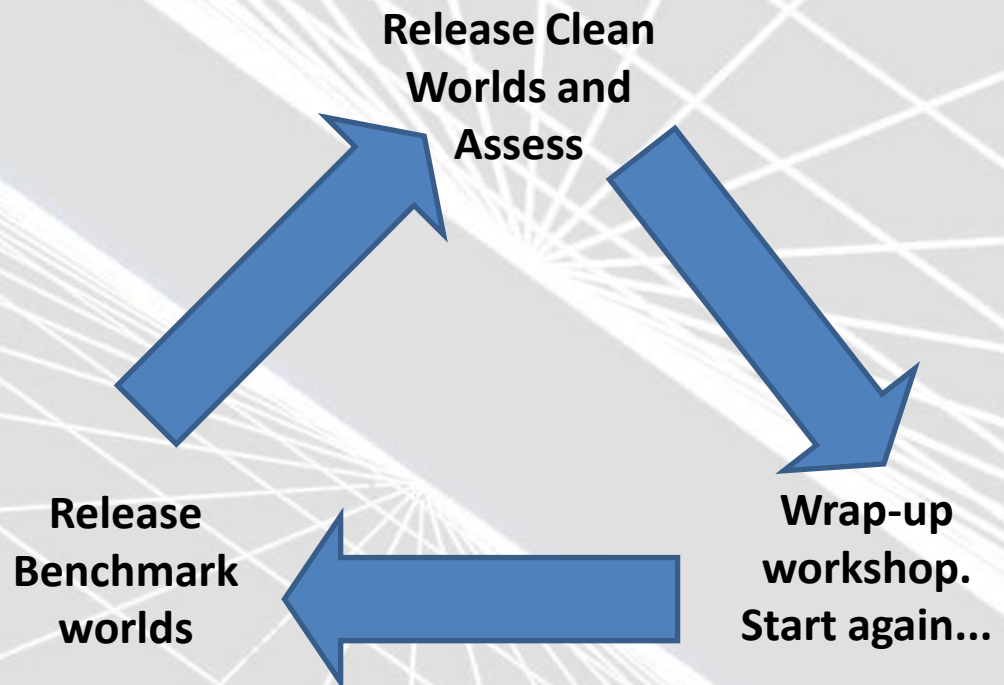
- Simulate 'clean' spatio-temporal characteristics of actual stations underpinned by low frequency variability from a climate model to maintain plausible spatial correlation
- Add abrupt and gradual changepoints to approximate our best guess real world error structures
- Run homogenisation algorithms on the test data and assess ability to recover original 'clean' data
- Useful for further improvement of algorithms

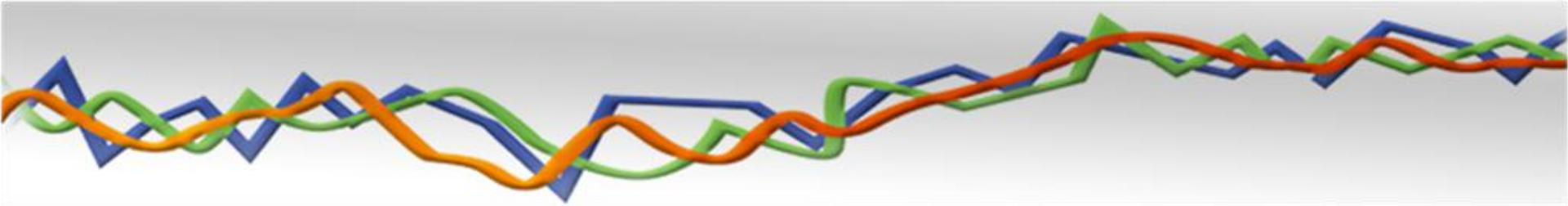


Example use of benchmark data  
for USHCN

# Benchmarking cycle

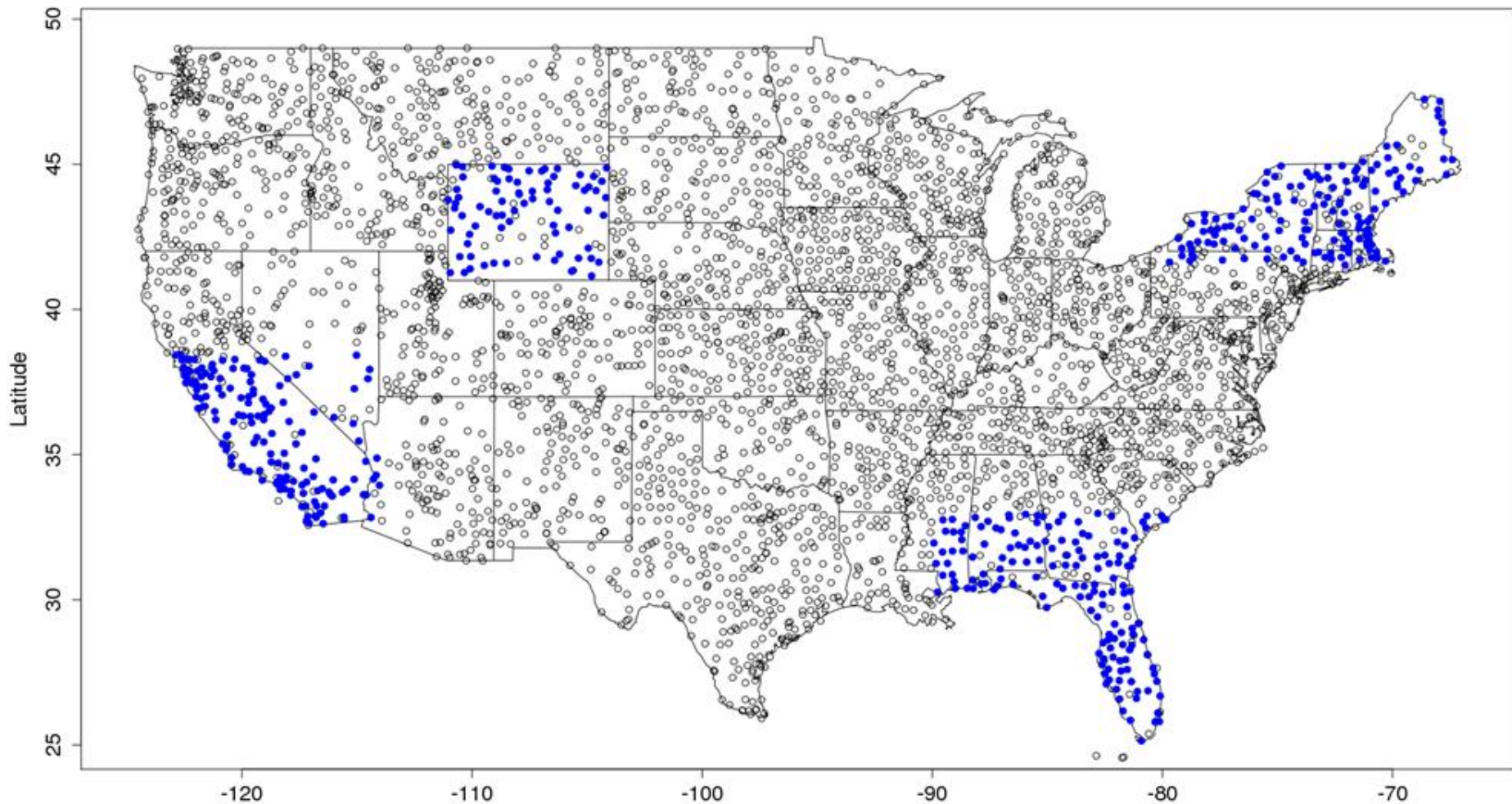
**Willett, et al., 2014:** A framework for benchmarking of homogenisation algorithm performance on the global scale, *Geoscientific Instrumentation, Methods and Data Systems*, 3, 187-200, doi:10.5194/gi-3-187-2014.





## Daily Benchmarks for the USA using a GAM

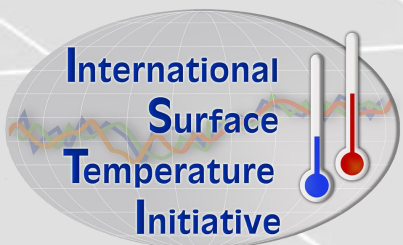
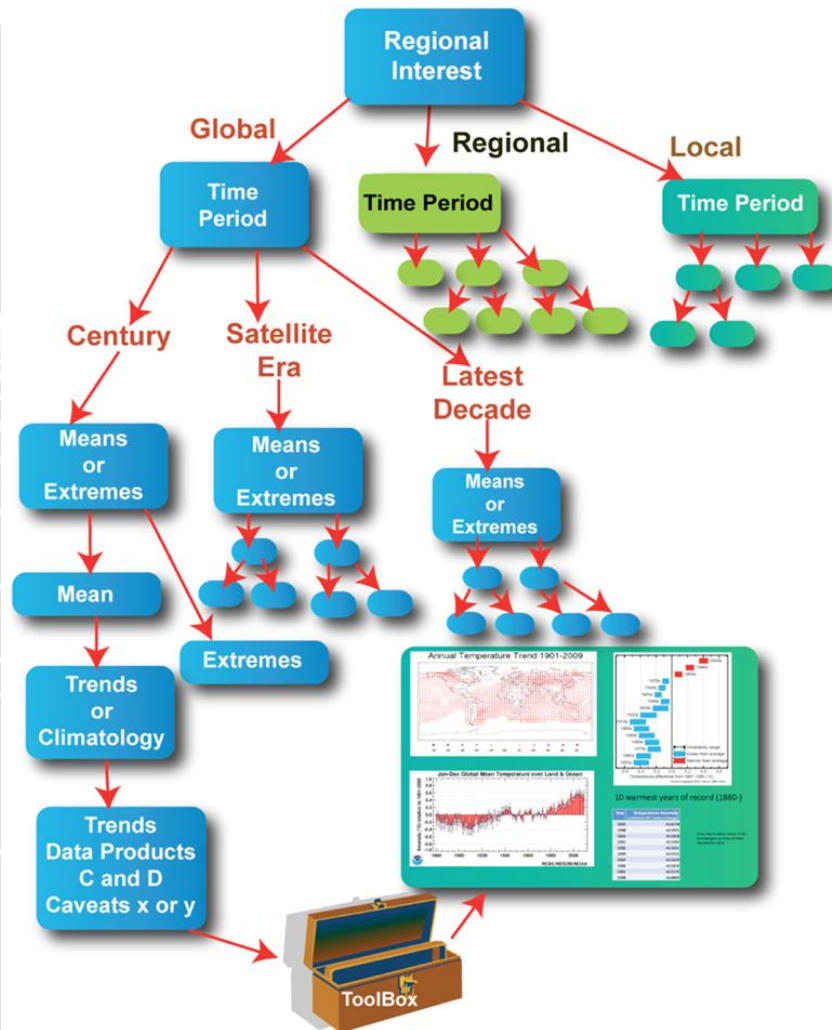
Stations with temperature records from 1970 to 2011 in the contiguous USA:  
Blue = Focus stations with no more than a quarter of the record missing





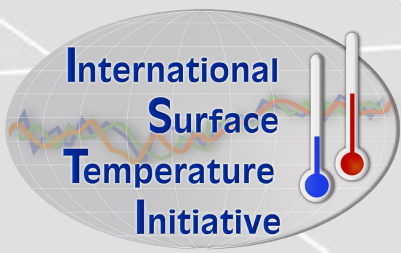
# Step 3: Serving products and aiding users

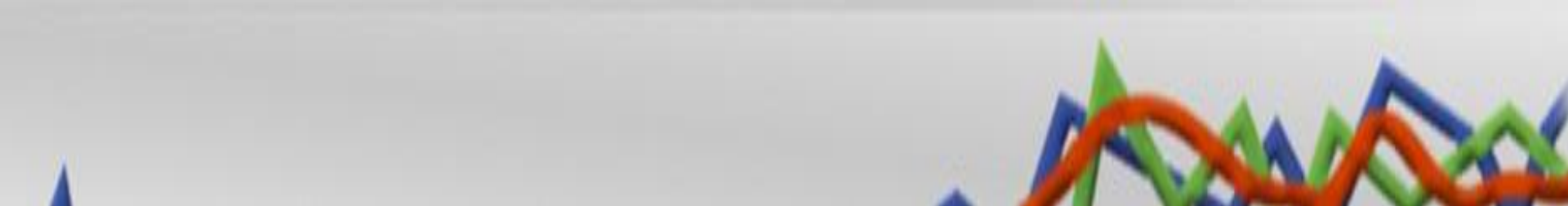
Hypothetical Decision Tree



# What happens if you build a state of the art playground and nobody turns up?

- The Initiative will have provided a framework which should be conducive to scientists coming and having a 'play'
- The Initiative cannot compel scientists to 'come and play'
- Nor does it have dedicated funding support to offer ...





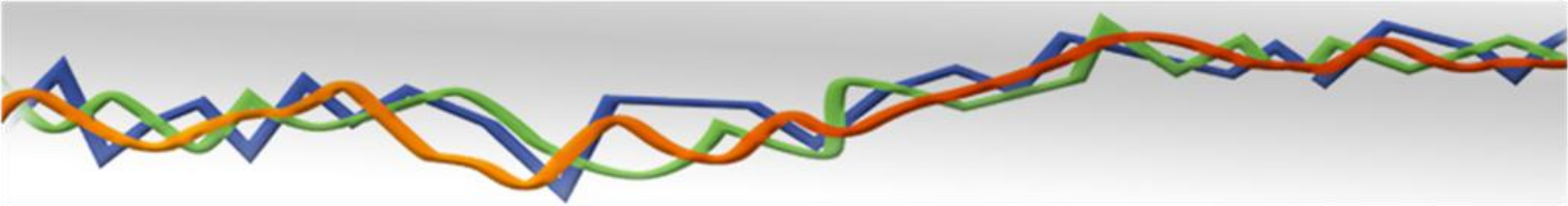
# Q & A

[www.surface temperatures.org](http://www.surface temperatures.org)

Bull. Amer. Met. Soc. doi: 10.1175/2011BAMS3124.1

[General.enquiries@surface temperatures.org](mailto:General.enquiries@surface temperatures.org)

[Data.submission@surface temperatures.org](mailto:Data.submission@surface temperatures.org)



# **Parallel Observations Science Team (POST)**

([http://www.surface temperatures.org/databank/parallel\\_measurements](http://www.surface temperatures.org/databank/parallel_measurements))

Long instrumental climate records are usually affected by non-climatic changes, such as relocations and changes in instrumentation, instrument height or procedures. They can hamper the assessment of trends and variability in climate records by distorting the climate signal, especially when it comes to trends in extreme weather.

The most direct way to study the influence of these non-climatic changes and to understand the reasons for these biases is the analysis of parallel measurements representing the old and new situation (for example instruments or locations).

## A GLOBAL PARALLEL CLIMATE DATASET

Current studies of non-climatic changes using parallel data are limited to local and regional case studies. However, the effect of specific transitions depends on the local climate and the most interesting questions are about the systematic large-scale biases

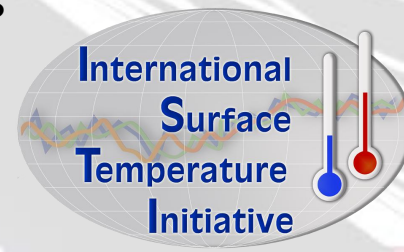
produced by transitions that occurred in many regions. Important potentially biasing transitions are the adoption of Stevenson screens or the introduction of automatic weather stations. Thus a large global parallel dataset is highly desirable as it enables the study of systematic biases in the global record.



A standard Stevenson screen and a historical French screen measuring in parallel at an observatory in Basel, Switzerland.

# Multiplicity of data products

- .Quantifying structural uncertainty is key
- .Raw data is far from traceable to international measurement standards.
- .Data artifacts are numerous and have myriad causes
- .Metadata describing station histories is patchy at best and often non-existent
- .Data is discrete in both space and time
- .No “how to” ... rather very many cases of “it may work ...”
- .Multiple subjective decisions required even in automated procedures (thresholds, periods)
- .Different approaches may have different strengths and weaknesses
- .No single dataset can answer all user needs



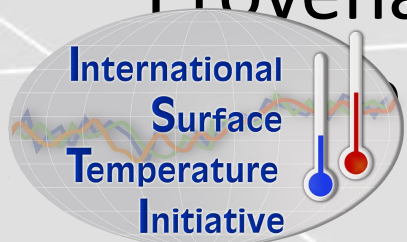






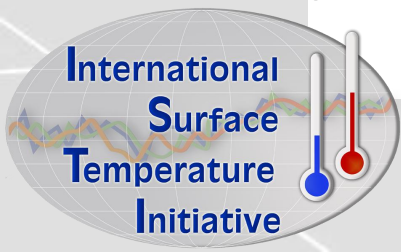
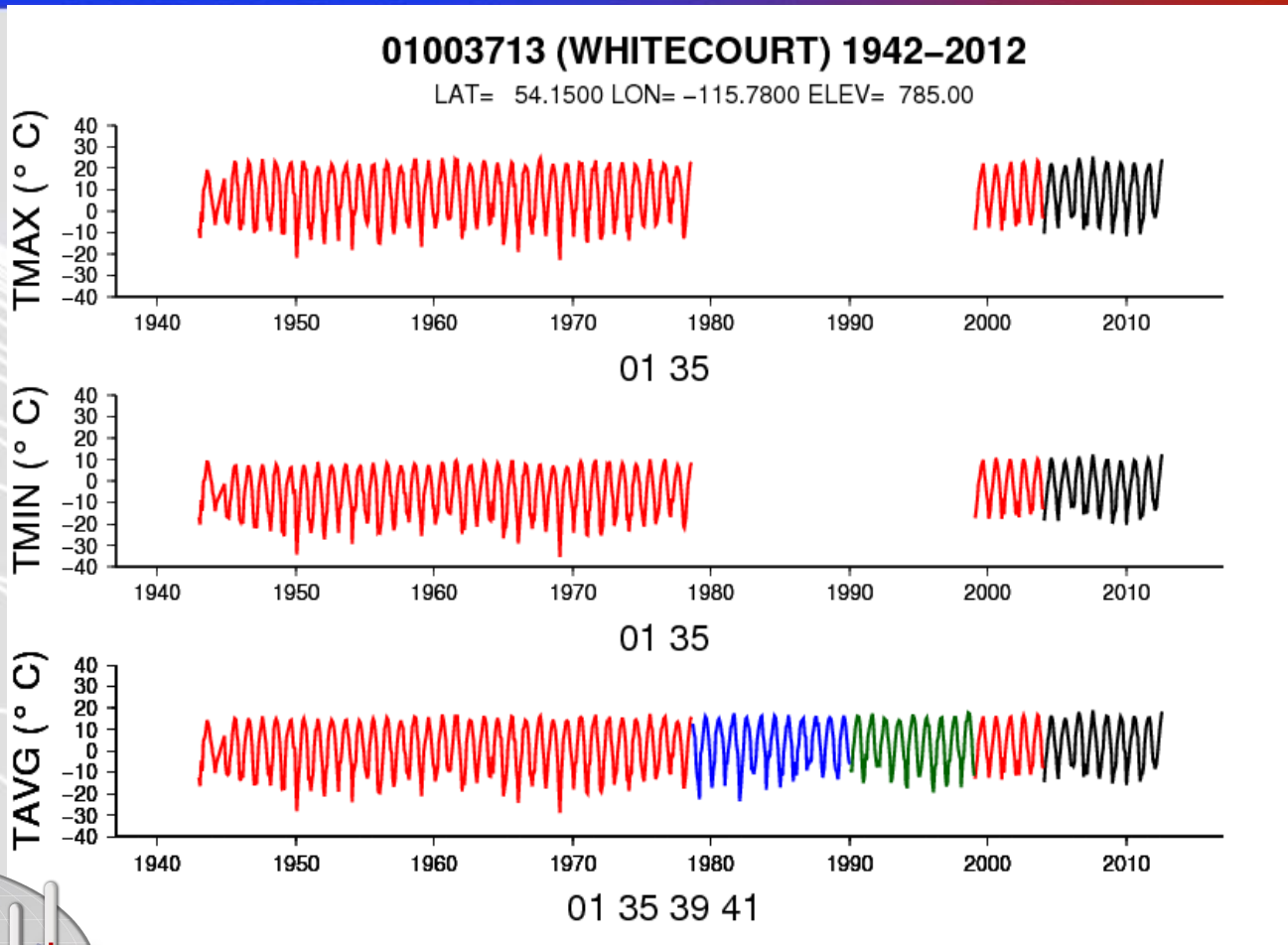
# Stage 3 (under beta)

- Same format as stage 2
- Optimised station merging of non-unique records
- One unique version for each station – recommended version for most users
- Forms basis for creation of benchmarking analog stations (see later)
- Provenance tracking ensures an unbroken chain

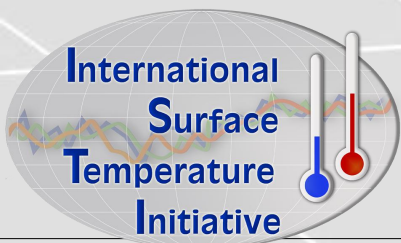


or stages

# Station series example

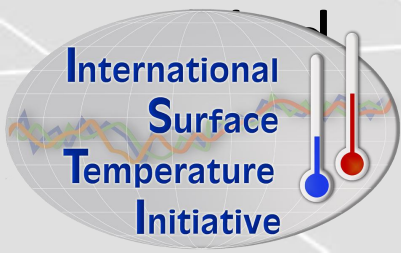


# Benchmarking cycle

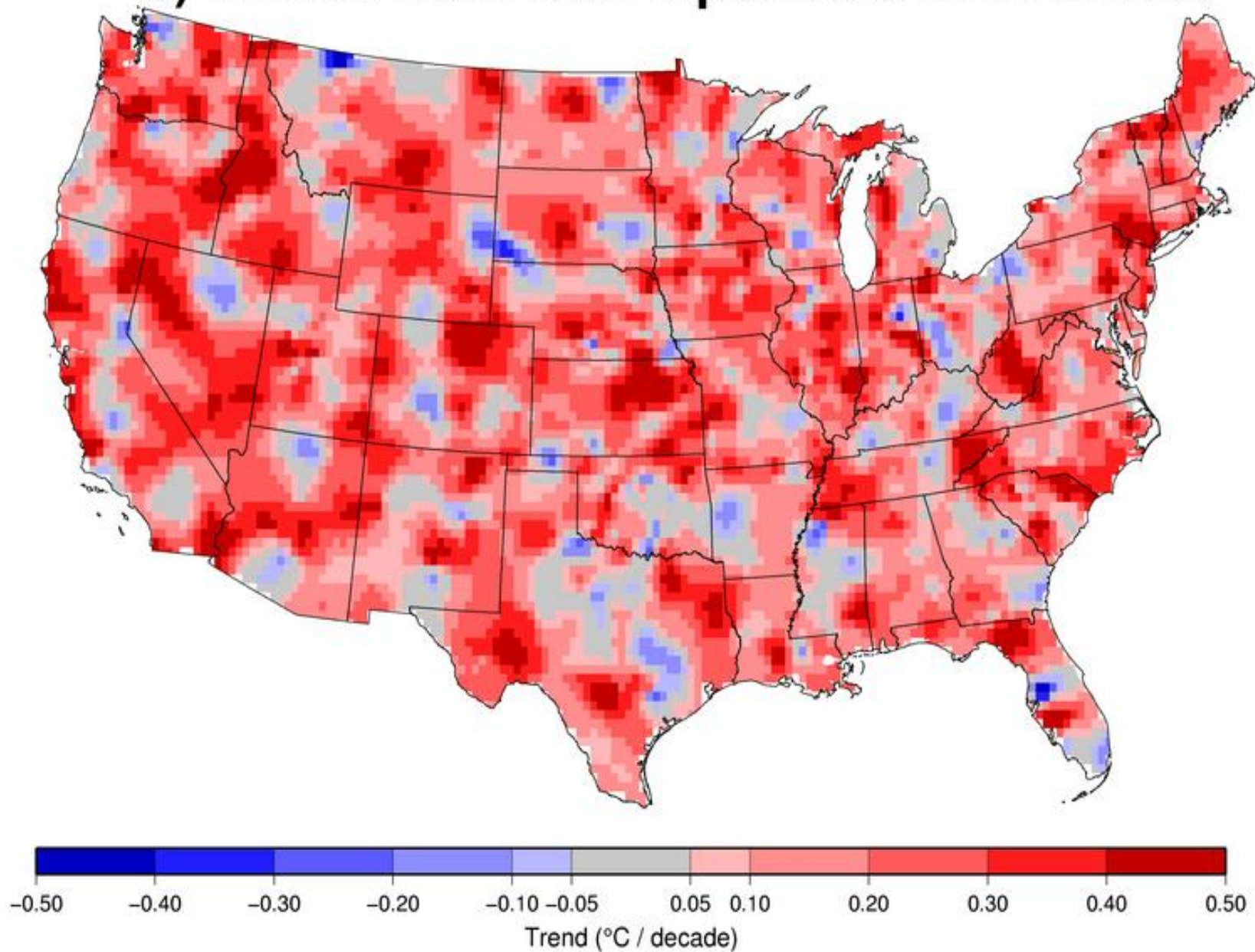


# Benchmarking example

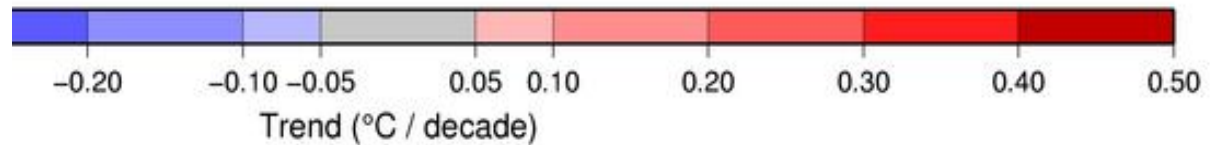
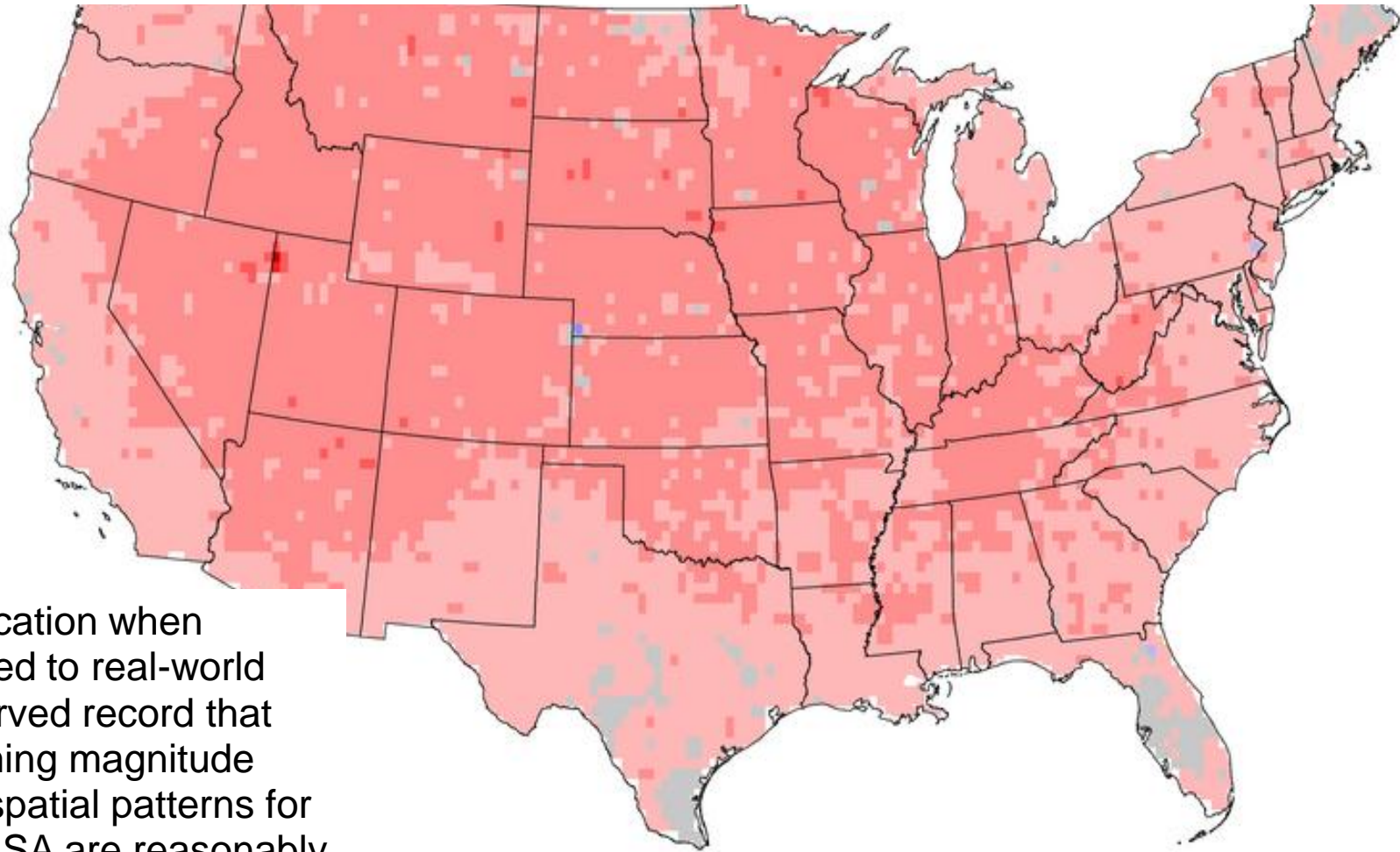
- For USHCN (lower 48 states)
- 100 member perturbed ensemble of the NCDC pairwise algorithm was run on 8 analogs (Williams et al., 2012, JGR-A, 117, D05116, doi:10.1029/2011JD016761)
- Consideration solely of timeseries and trends
- Analyses that follow are for the hardest analog with frequent predominantly small breaks



# a) Trends from Raw Input Data With Errors

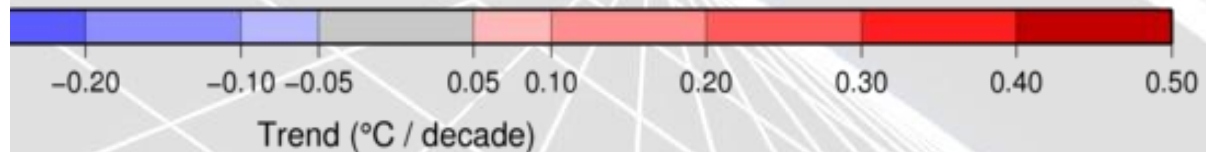
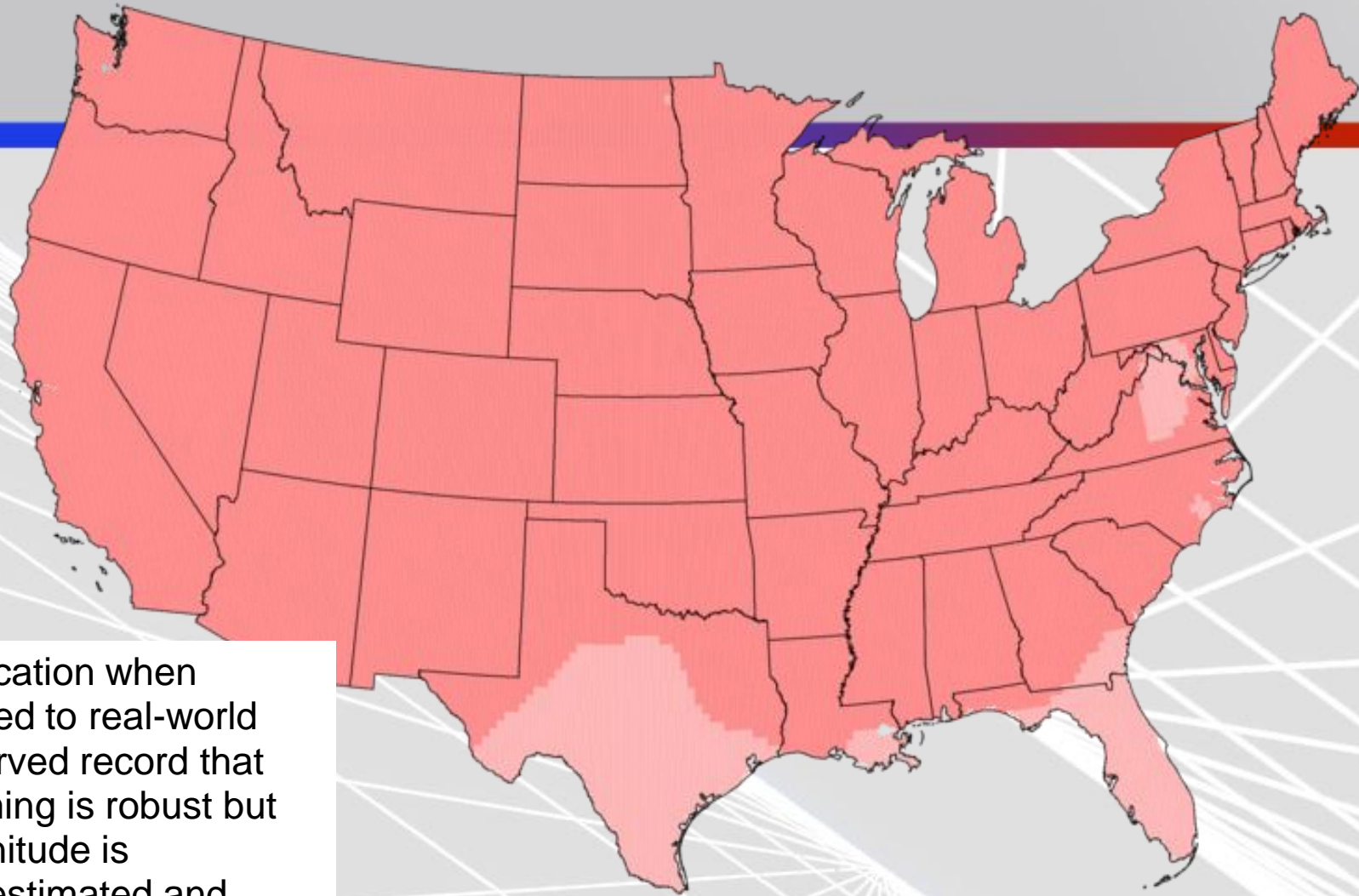


## c) Homogenized Data (NOAA/NCDC Pairwise Algorithm)



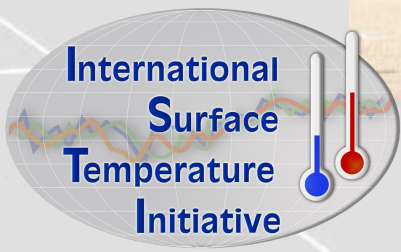
Implication when applied to real-world observed record that warming magnitude and spatial patterns for the USA are reasonably well captured with some regional discrepancies

## d) Homogenized Data (Berkeley Earth Surface Temperature Method)



Implication when applied to real-world observed record that warming is robust but magnitude is overestimated and spatial patterns not well captured for the USA

# By analogy ...





# The field is wide open ...

- We have thus far sampled only a small area of solution space which has many d.o.f
- We need to far more fully explore the plausible solution space
- Many possibilities exist

Bottom line:  
Please please please  
come and play

