

Towards an improved 20th Century reanalysis version “2c” (1851-2012)

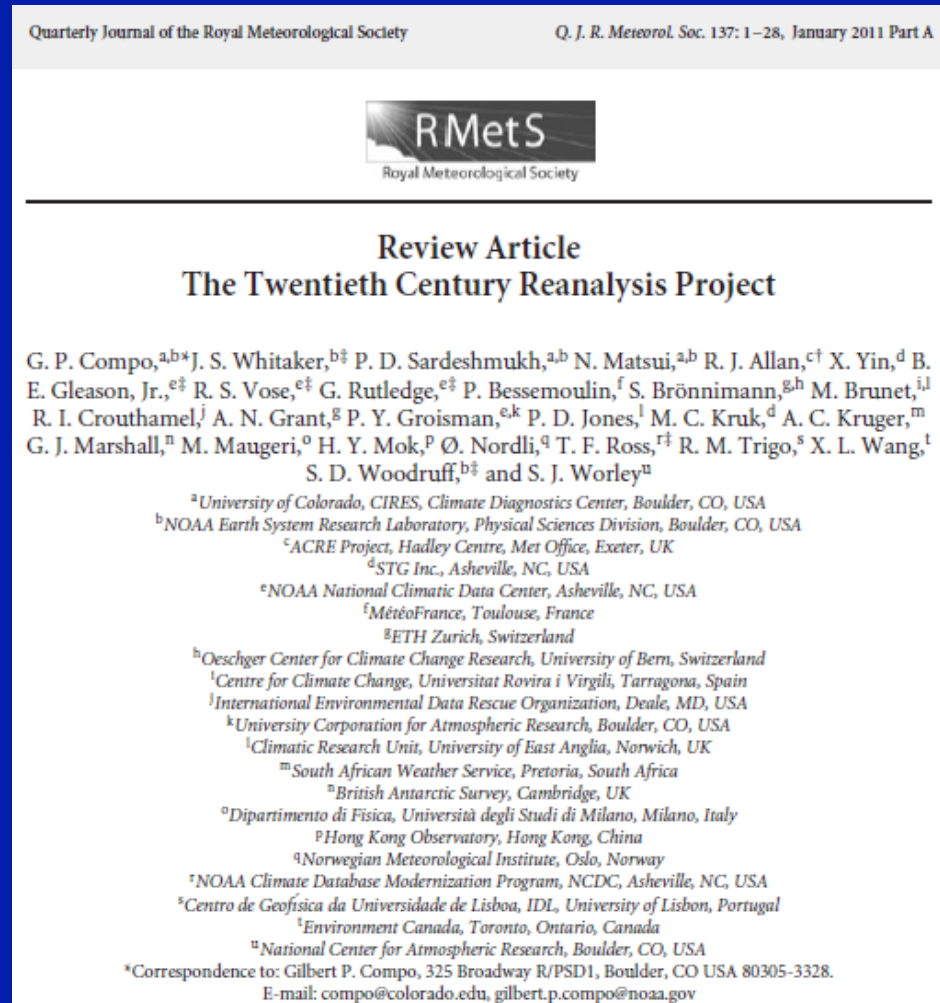
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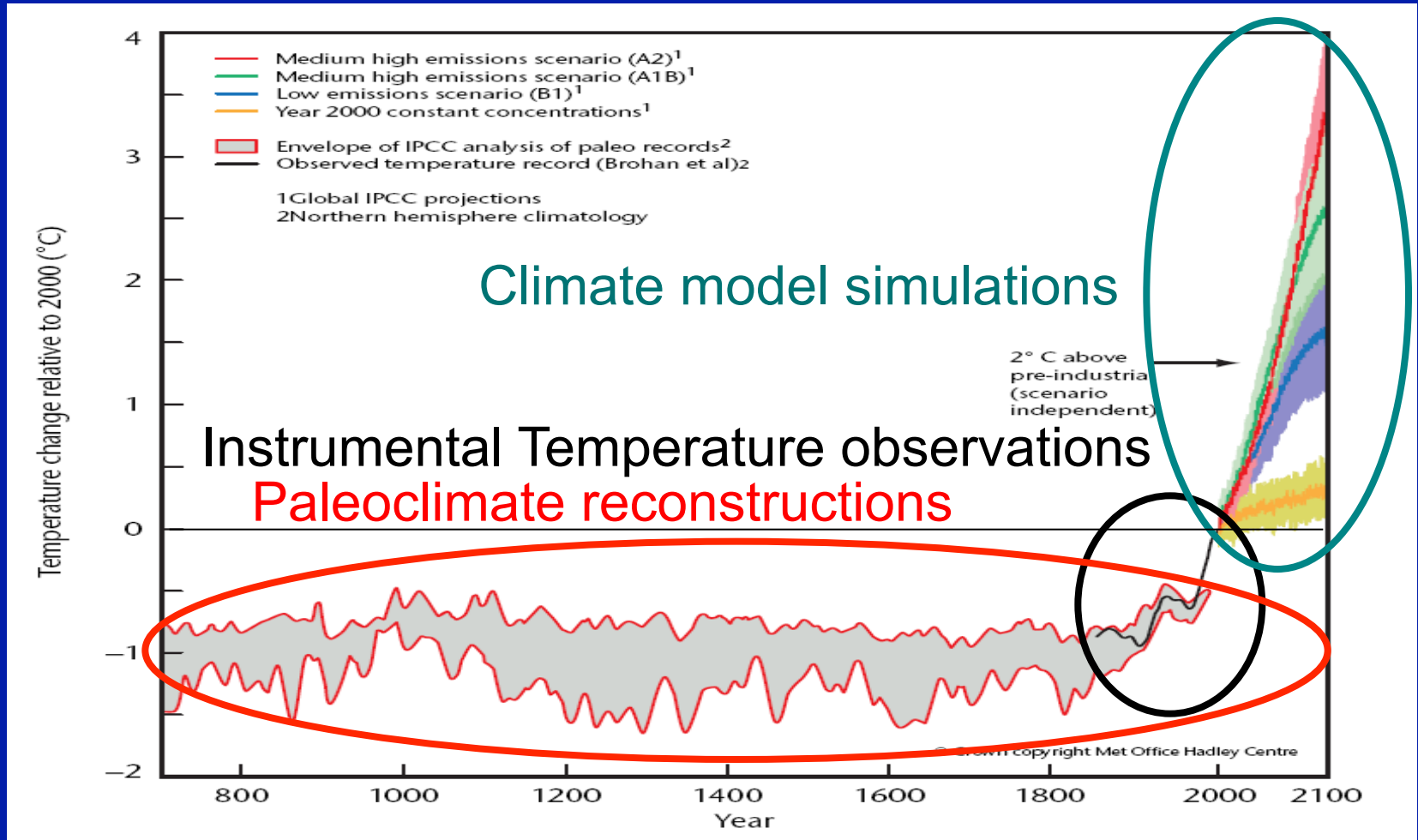


Compo et al. 2011 dx.doi.org/10.1002/qj.

776

Special thanks to Chesley McColl, NCEP/EMC, NCDC, Hadley Centre, ACRE partners

Illustration of Temperature Change relative to 2000



800

1200

1600

2000

Instrumental period is the short link between paleo reconstructions and climate model simulations

A Major Goal of the 20th Century Reanalysis project and the international Atmospheric Circulation Reconstructions over the Earth initiative:

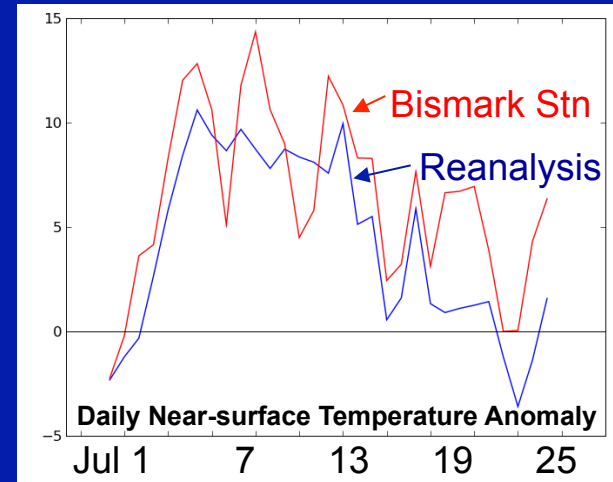
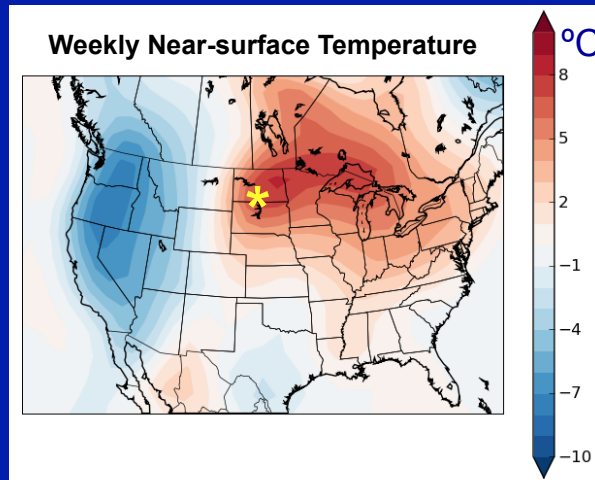
Use data assimilation to produce the longest possible *Instrument-based* estimates of global weather and climate for comparison to paleoclimate reconstructions and climate model simulations

The 20th Century Reanalysis Project version 2c (1851-2011)

Summary: An international project led by CIRES and NOAA to produce *4-dimensional* reanalysis datasets for climate applications extending back to the 19th century using an Ensemble Kalman Filter and *only surface pressure observations*.

Weekly-averaged anomaly during **July 1936** North American Heat Wave (> 2,000 dead during 10-day span)

Daily variations compare well with in-situ data.



The reanalyses provide:

- First-ever estimates of near-surface to tropopause 6-hourly fields extending back to the middle of the 19th century;
- Estimates of uncertainties in the basic reanalyses and derived quantities (e.g., storm tracks).

Examples of uses:

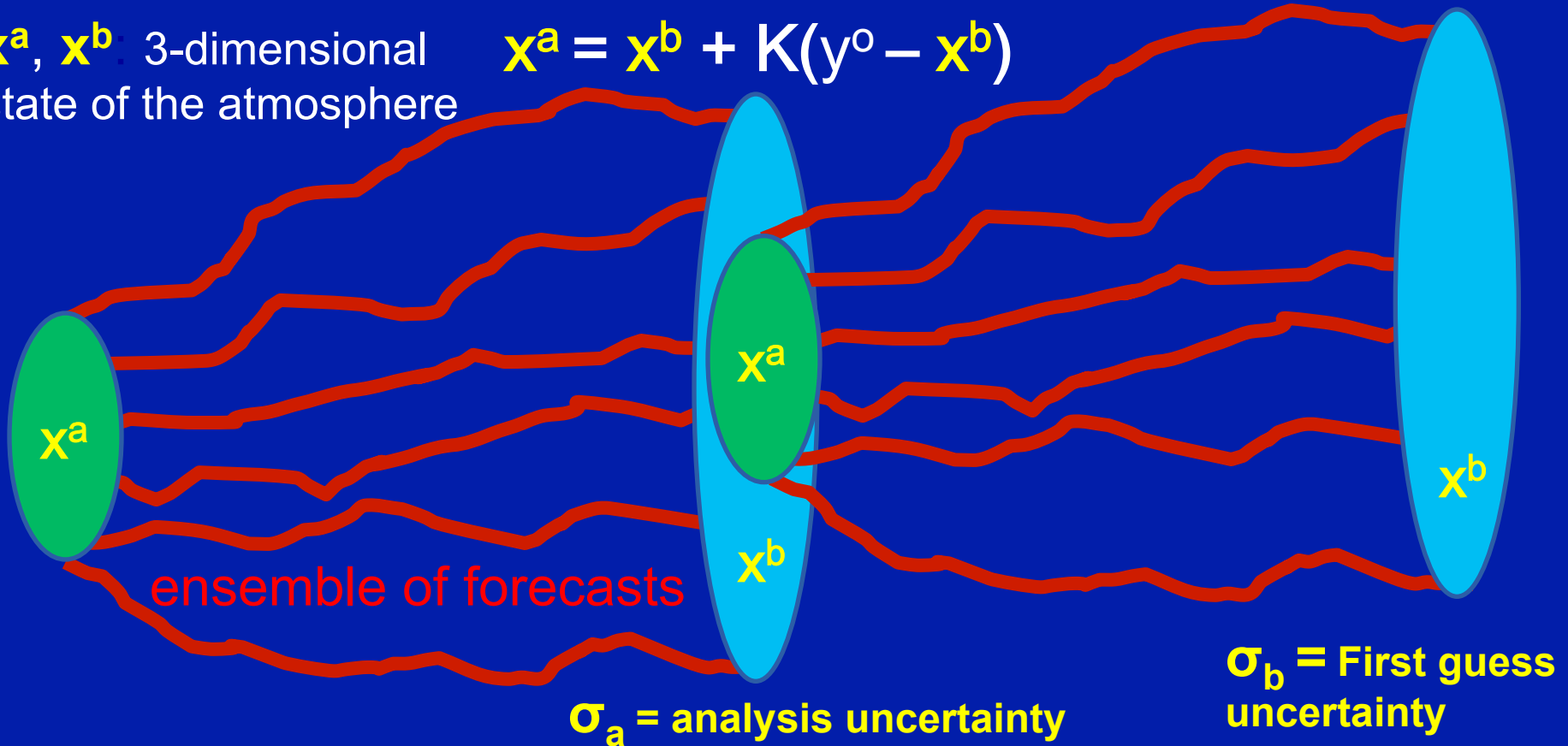
- Validating climate models.
- Determining storminess and storm track variations over the last 150 years.
- Understanding historical climate variations (e.g., 1930s Dust Bowl, 1920-1940s Arctic warming).
- Estimating risks of extreme events

Ensemble Data Assimilation (Whitaker and Hamill, 2002)

20CR analysis \mathbf{x}^a is a weighted average of the first guess \mathbf{x}^b and pressure observation y^o . Each observation is assimilated serially.

$\mathbf{x}^a, \mathbf{x}^b$: 3-dimensional state of the atmosphere

$$\mathbf{x}^a = \mathbf{x}^b + \mathbf{K}(y^o - \mathbf{x}^b)$$



ensemble of forecasts

σ_a = analysis uncertainty

σ_b = First guess uncertainty

the weight \mathbf{K} varies with the atmospheric flow and the observation network

analysis time (0Z)

analysis time (6Z)

analysis time (12Z)

Ensemble Filter Algorithm (Whitaker and Hamill, 2002)

Ensemble mean

$$\bar{\mathbf{x}}_j^a = \bar{\mathbf{x}}_j^b + \mathbf{K} \left(y^o - \bar{y}_j^b \right),$$

Ensemble deviations

$$\mathbf{x}_j'^a = \mathbf{x}_j'^b - \tilde{\mathbf{K}} \left(y_j'^b \right),$$

Sample
Kalman
Gain

$$\begin{aligned} \mathbf{K} &= \mathbf{P}^b \mathbf{H}^T (\mathbf{H} \mathbf{P}^b \mathbf{H}^T + R)^{-1} \\ &= \frac{1}{n-1} \sum_{j=1}^n \mathbf{x}_j'^b y_j'^b \left(\frac{1}{n-1} \sum_{j=1}^n y_j'^b y_j'^b + R \right)^{-1} \end{aligned}$$

Sample
Modified
Kalman Gain

$$\tilde{\mathbf{K}} = \left(1 + \sqrt{\frac{R}{\mathbf{H} \mathbf{P}^b \mathbf{H}^T + R}} \right)^{-1} \mathbf{K},$$

$\mathbf{x}_j = \mathbf{x} + \mathbf{x}'_j$ is pressure, air temperature, winds, humidity, etc. at all levels and gridpoints, every six hours.

y^o is only observations of hourly and synoptic surface pressure,

$y^b = \mathbf{H} \mathbf{x}^b$ is guess surface pressure

20th Century Reanalysis v2c implementation of Ensemble Filter Algorithm

(Whitaker et al. 2004, Compo et al. 2006, Compo et al. 2011)

Algorithm uses an ensemble of GCM runs to produce the weight **K** that varies with the atmospheric flow and the observation network every 6 hours

Using 56 member ensemble, new prescribed boundary conditions:

SODAsi.2c 18 member pentad SST and

COBE-SST2 monthly sea ice concentration

(corrects sea ice error in v2)

(Giese et al. 2015, Hirahara et al. 2014)

1851-2011:

T62 (~200km), 28 level NCEP GFS08ex atmosphere/land model

9 hour forecasts for 6 hour centered analysis window

- time-varying CO₂, solar and volcanic radiative forcing (Sato et al.)

<http://go.usa.gov/XTd> Compo et al. 2011, doi:10.1002/qj.776

20th Century Reanalysis v2c implementation of Ensemble Filter Algorithm (same as v2)

(Whitaker et al. 2004, Compo et al. 2006, Compo et al. 2011)

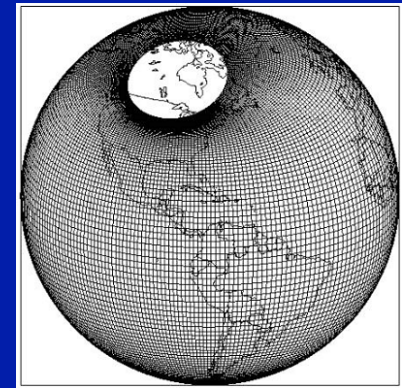
Algorithm uses an ensemble of GCM runs to produce the weight **K** that varies with the atmospheric flow and the observation network every 6 hours

Sampling and Model error parameterizations:

- Covariance localization (4000 km, 4 scale heights) and
- Latitude and time dependent multiplicative covariance inflation (alpha = 1.01 to 1.12) [*Anderson and Anderson, 1999; Houtekamer and Mitchell, 2001; Hamill et al. 2001; Whitaker et al., 2004*]

Every 5 years produced in parallel: 1851-1855, ..., 1881-1885, ..., 1996-2000, .., 2006-2011 after 14 month spin-up

SODA sparse input v2 (1846-2011)



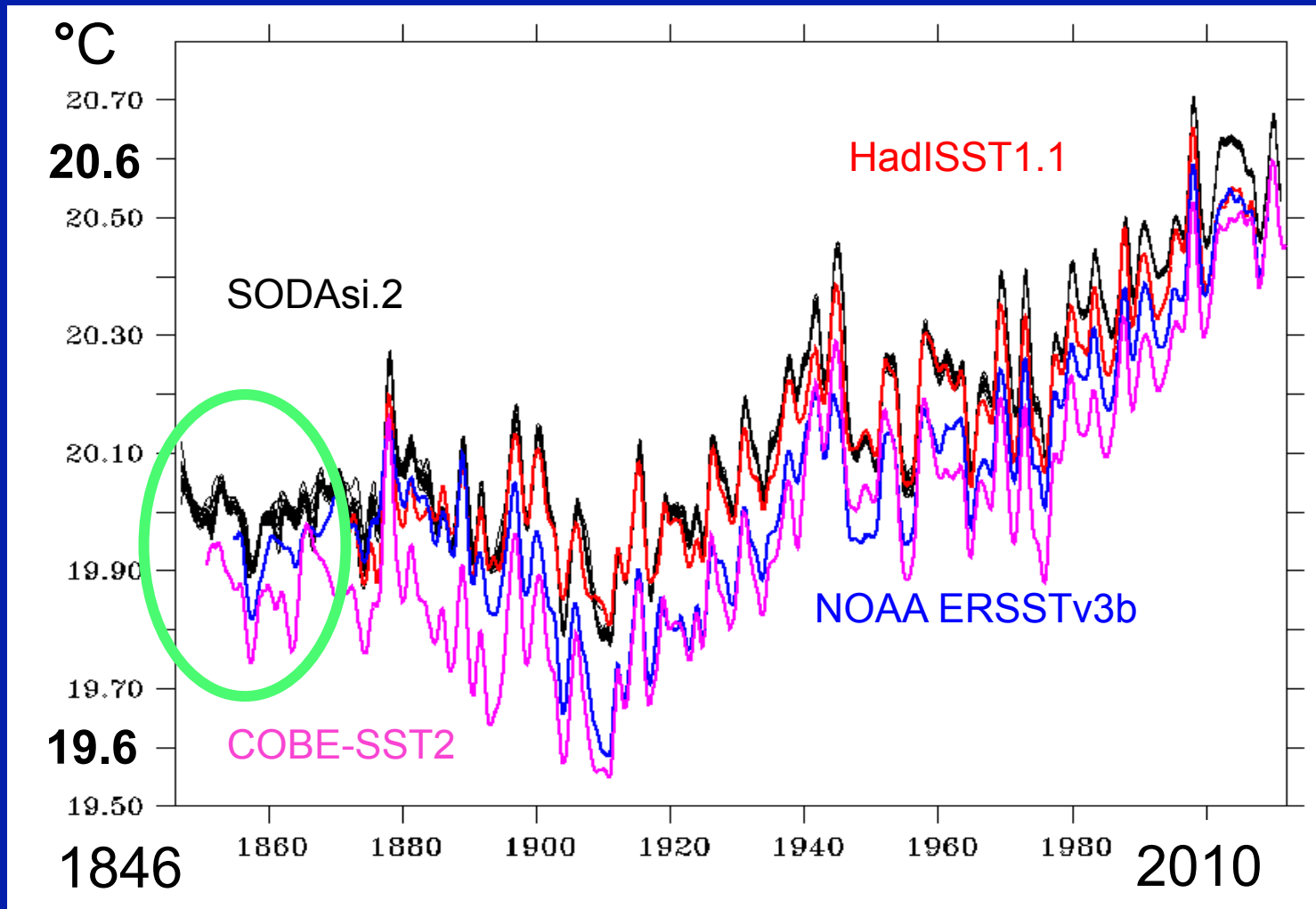
- 18 Ensemble Members
- Parallel Ocean Program v2.0.1
 - **0.4° longitude x 0.25° to 0.4° latitude with 40 levels**
- Winds
 - 20CRv2 ensemble member daily stress (1949 – 2011)
 - 20CRv2 system with ISPDv3.2.4 and HadISST1.1 (1871-1948)
 - with ISPDv3.2.4 and climatological SST (1846-1870)
- Heat and Salt fluxes
 - Bulk formulae using 20CRv2 daily variables
- SODAsi Observations
 - Only ICOADS 2.5 SST data with Hadley Bucket Correction

Gleckler et al. 2015

Simple Ocean Data Assimilation sparse input v2

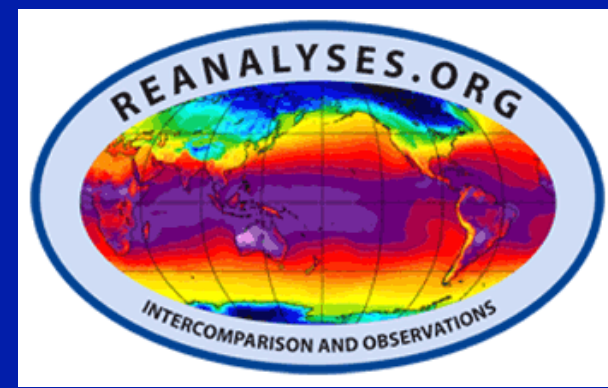
Global Ocean Annual Average (60N-60S)

Climo
SST
in 20CR
1846 to
1870
but
variation
in
SODAsi



SODAsi trends and decadal variability are consistent with statistical reconstructions. Generates interannual variations in late 1850s even when 20CR forcing had climatological SST.

International Surface Pressure Databank version 3 (ISPD)



Subdaily observations assembled in partnership with

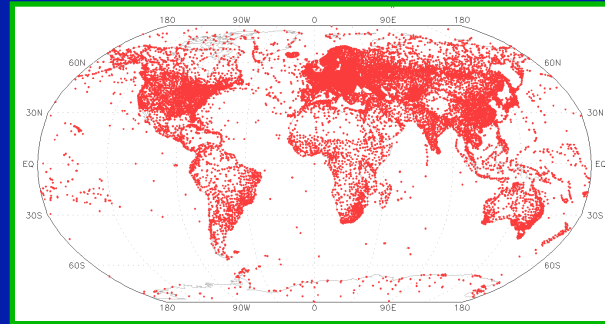
GCOS AOPC/OOPC Working Group on Surface Pressure

GCOS/WCRP Working Group on Observational Data Sets for Reanalysis

Atmospheric Circulation Reconstructions over the Earth (ACRE)

Land data Component: merged by NOAA NCDC, NOAA ESRL, and CU/CIRES

- 63 data sources
- 33,653 stations
- 1.7 billion obs
- 1768-2011



Marine data component: **ICOADSv2.5** merged by NOAA ESRL, NCDC, and NCAR;

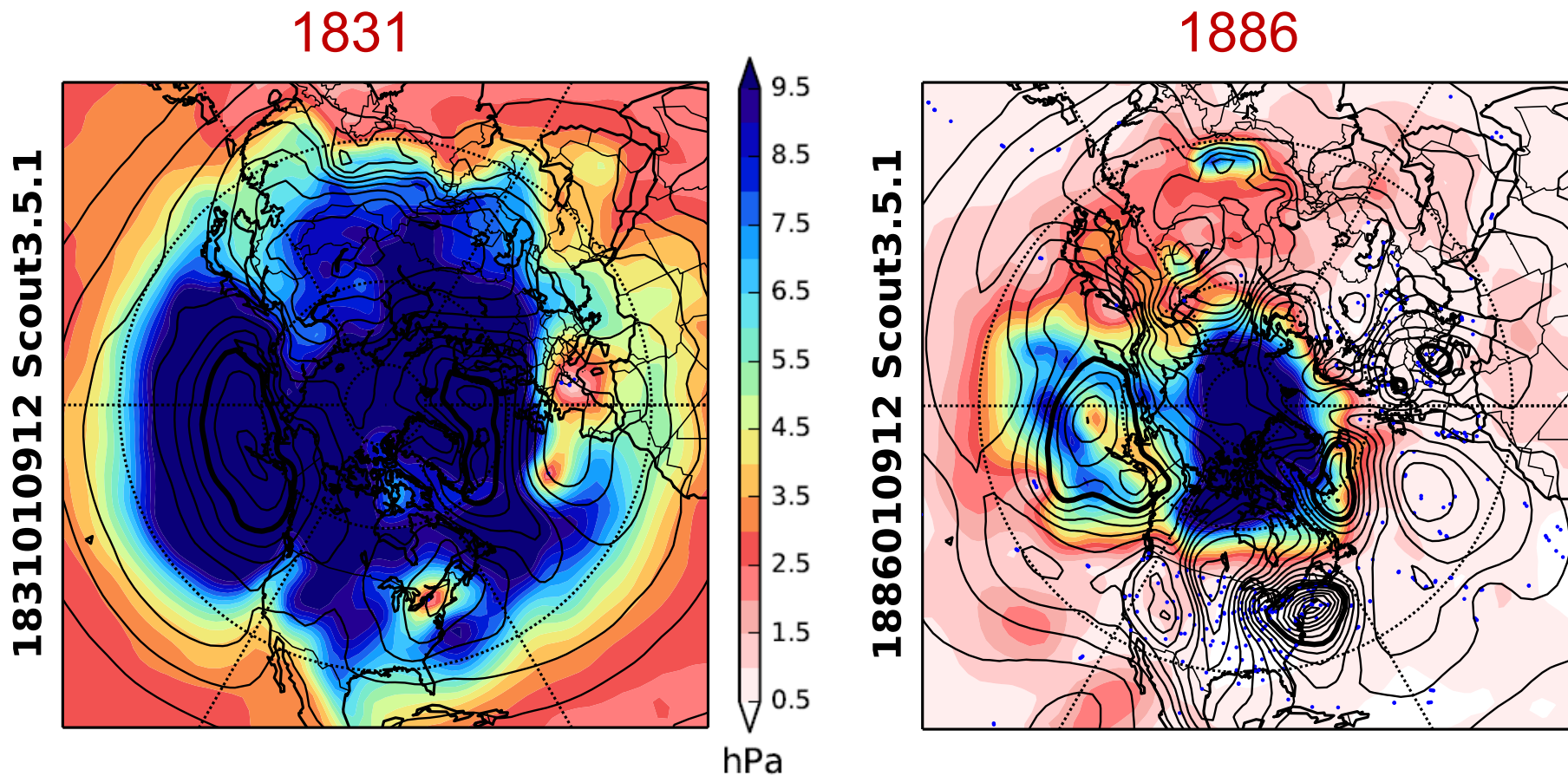
Oldweather.org, ACRE marine data

Tropical Cyclone Best Track data component: **IBTrACS** merged by NOAA NCDC

DATA ACCESS rda.ucar.edu/datasets/ds132.0 (T. Cram, NCAR DSS; C. McColl CIRES)

Reanalyses.org/observations/surface , Cram et al. 2015

20CRv2c Analyses of Sea Level Pressure for selected dates in 1831 and 1886



Contours-ensemble mean (ci: 4 hPa, 1000 hPa thickened)

Shading- blue: more uncertain, white: more certain

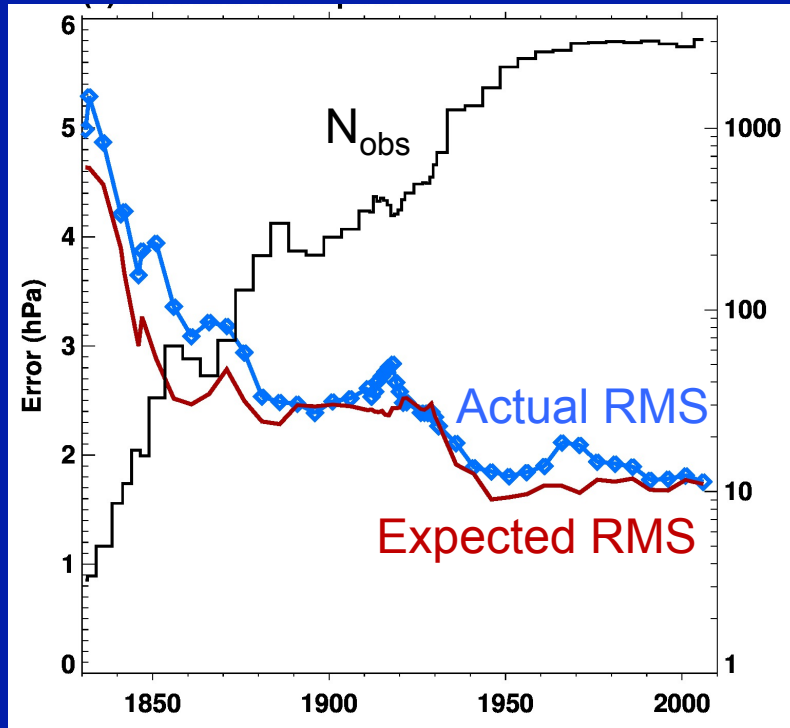
Analysis system responds to the observations and the flow, providing quantitative uncertainty for every variable at each analysis time.

v2c Surface Pressure uncertainty estimate poleward of 20(S,N)

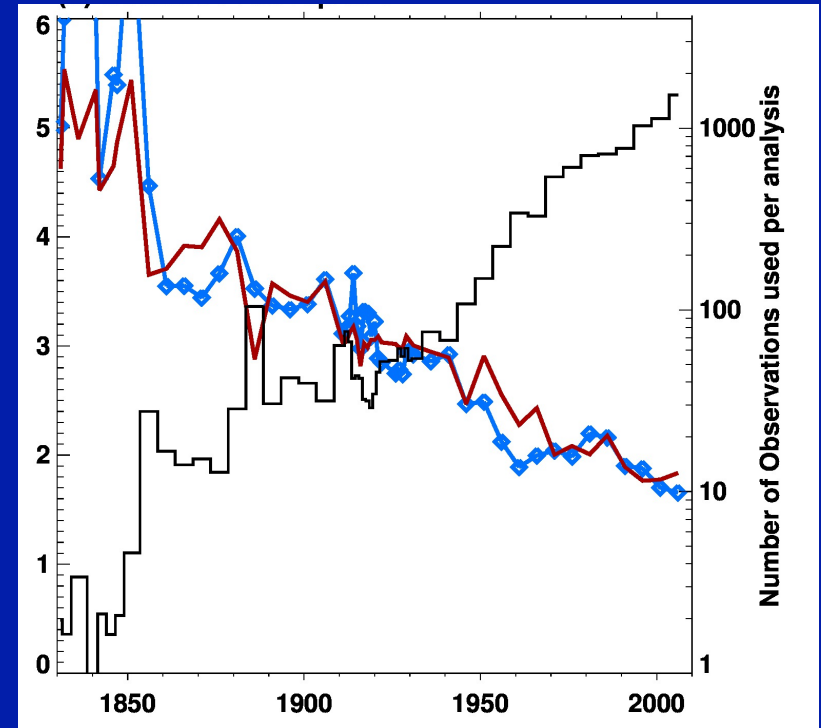
blue actual RMS difference obs minus first guess

red expected RMS difference obs minus first guess

Northern Hemisphere

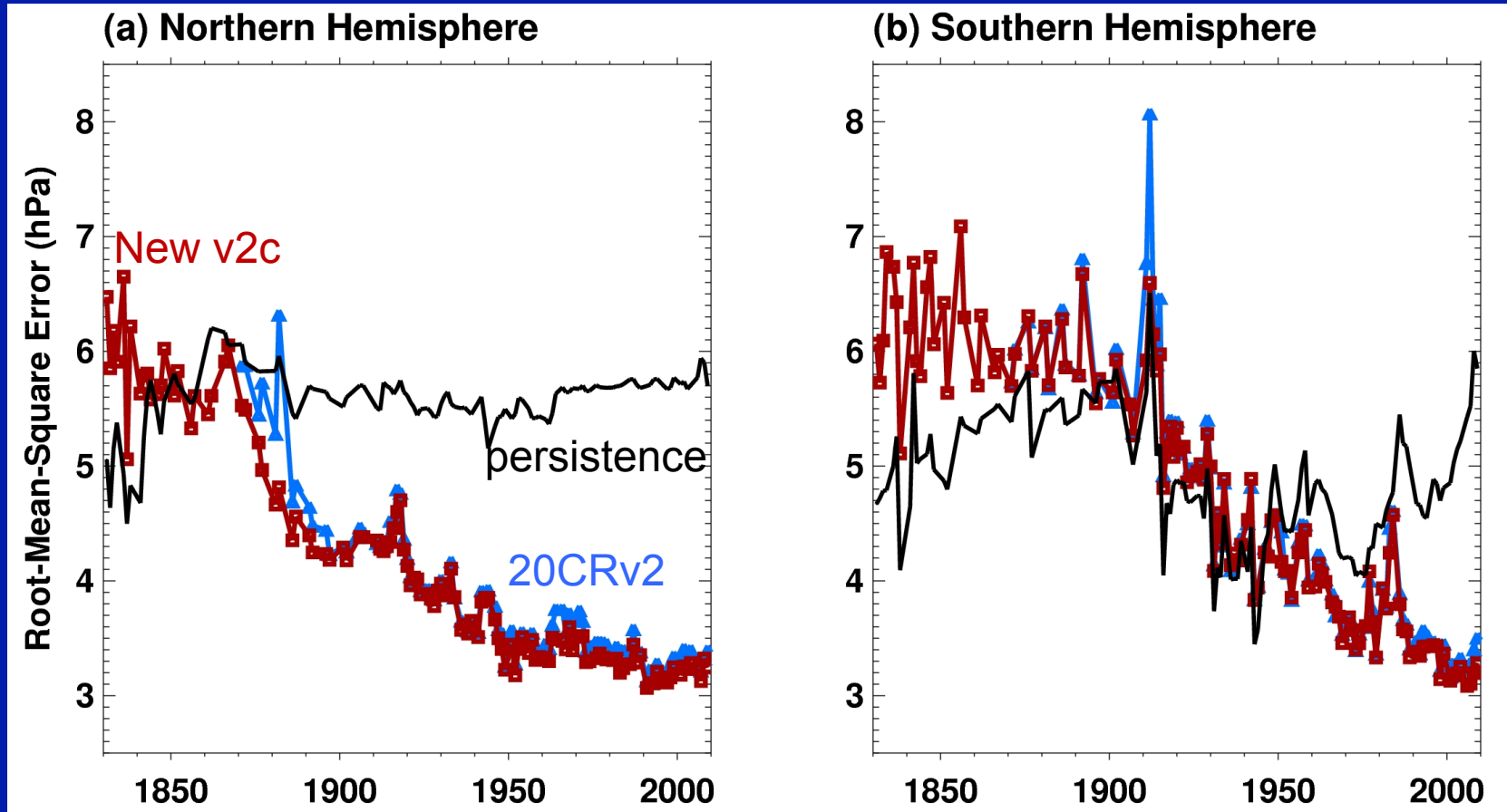


Southern Hemisphere



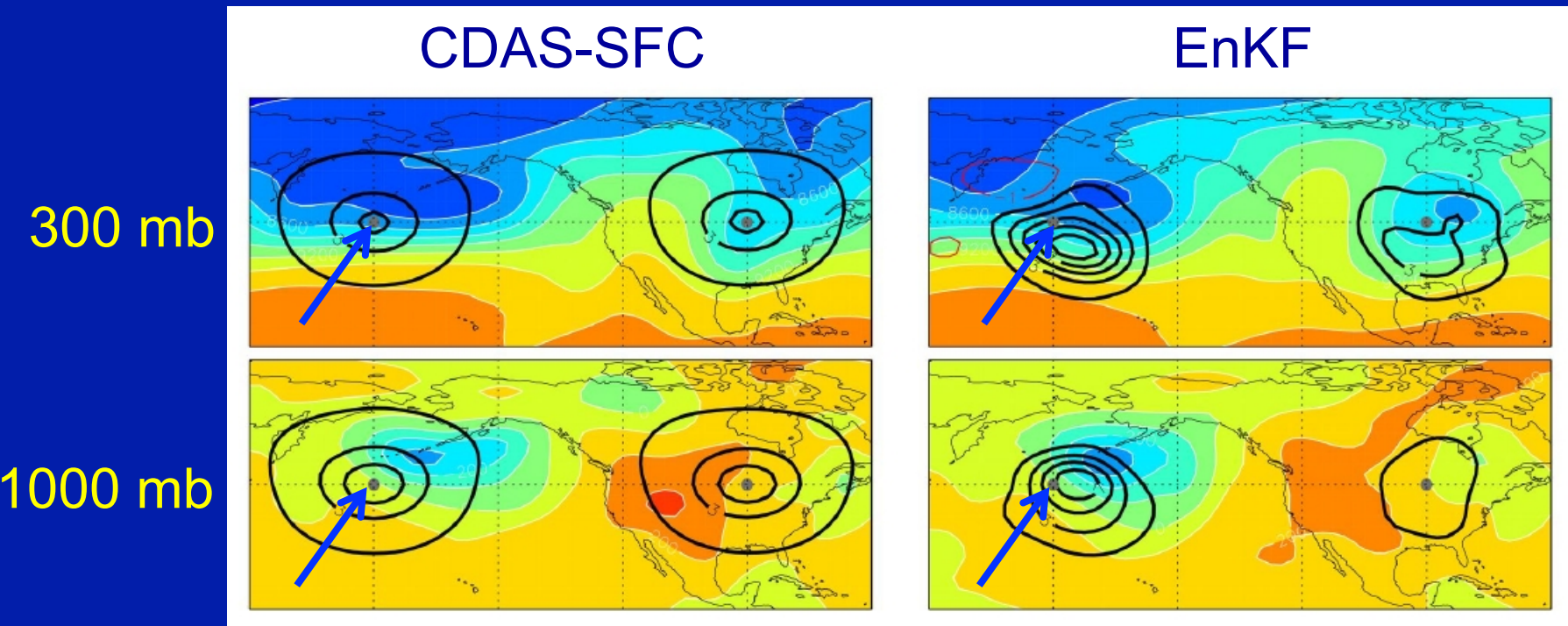
Uncertainty estimates are consistent with actual differences between first guess and pressure observations even as the network changes by three orders of magnitude over more than 150 years!
(This is not tuned).

Root Mean Square difference of Surface and Sea Level Pressure Observations and 24 hour Forecasts from 20CRv2 and v2c (Jan-Dec)



Northern Hemisphere 24 hr forecasts beat persistence even in 1850s. Southern Hemisphere has an analysis that produces forecasts comparable to persistence starting in 1900s. New v2c is an improvement.

Geopotential height first guess (colors) and
analysis minus first guess (lines)
for single pressure observations 1mb greater than first guess



Ensemble Kalman filter can extract spatially-varying structures
relative to the flow and the previous observational density.

Compo et al. 2006

Example uses of reanalysis

1. Effectively doubling the reanalysis record length 😊
2. Validating climate models for large-scale synoptic anomalies during extreme periods, such as droughts (30's, 50's).
3. Better understanding events such as the 1920-1940's Arctic warming.
4. Determining storminess and storm track variations over last 100-150 years.
5. Developing and improving forecasts of low-frequency (e.g., Pacific-North America pattern, North Atlantic Oscillation) atmospheric variations and their interannual to decadal variability.
6. Understanding changing atmospheric background state associated with interdecadal hurricane activity.
7. Discovering previously undocumented hurricanes.
8. Homogenizing upper-air and other independent observations.
9. Estimating historical probability distributions for wind energy.
10. Calibrating paleoclimate reconstructions.
11. Determining Weather effects on heroic journeys, e.g., death of Mallory and Irvine climbing Mt. Everest.
12. Tracking icebergs in the vicinity of the *Titanic*

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11. **Determining Weather effects on heroic journeys, e.g., death of Mallory and Irvine climbing Mt. Everest.**
12. **Tracking icebergs in the vicinity of the *Titanic*.**

Initial comparisons with ERA-20C

(Poli et al. 2014)

Assimilates

1) Pressure observations

ISPDv3.2.6: International effort to recover 100s of new stations, ACRE & ERA-CLIM data rescue, over 33 new organizations contributing

2) ICOADS R2.5 marine near-surface winds

4D-Var algorithm

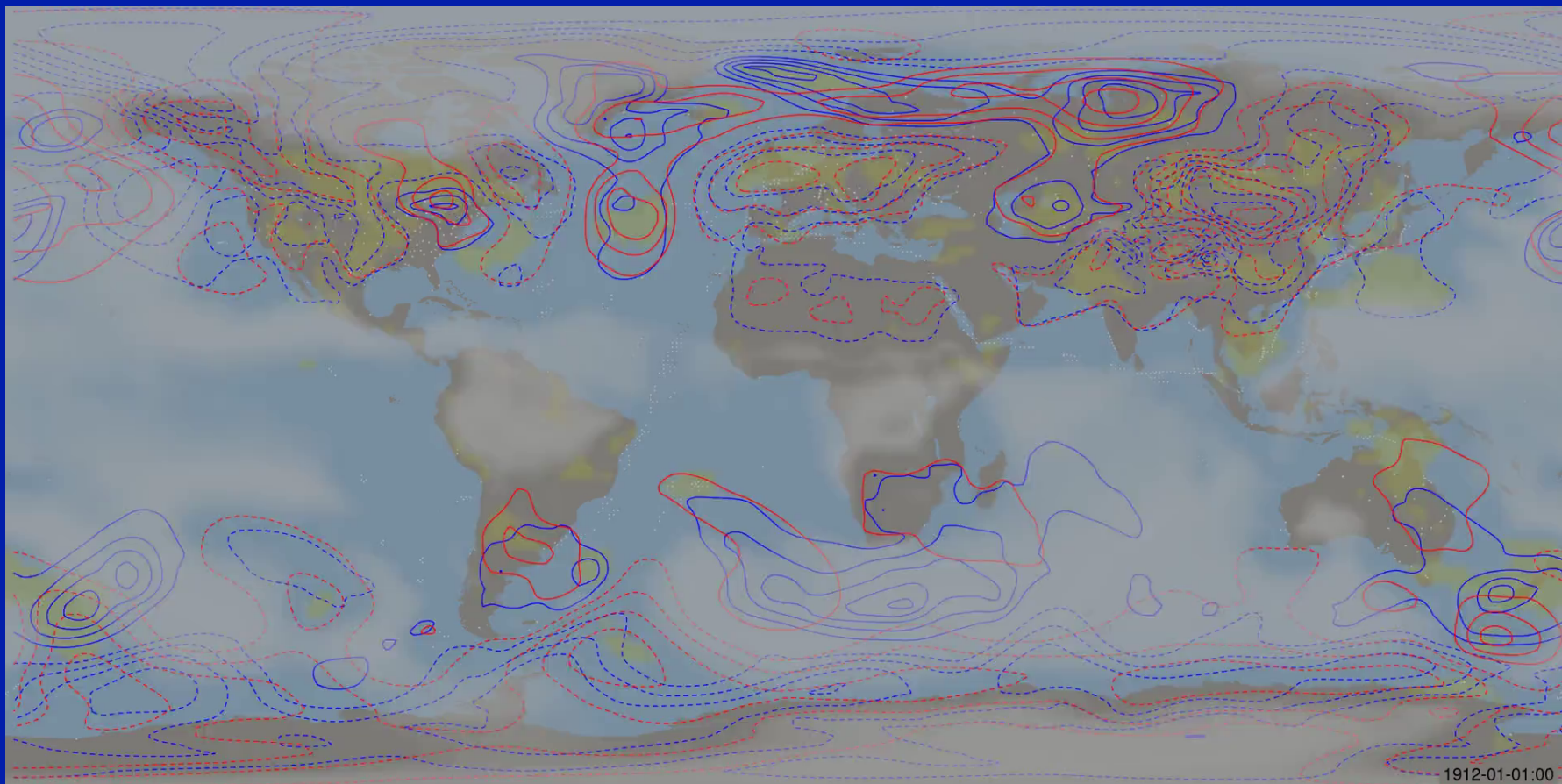
HadISST2.1.0.0 sea surface temperature and sea ice concentration

January 1912 Sea Level Pressure anomalies

20CRv2c ERA-20C

(gold=20CRv2c difference from ERA-20C > 3*spread)

Full Brohan movie at vimeo.com/109681668



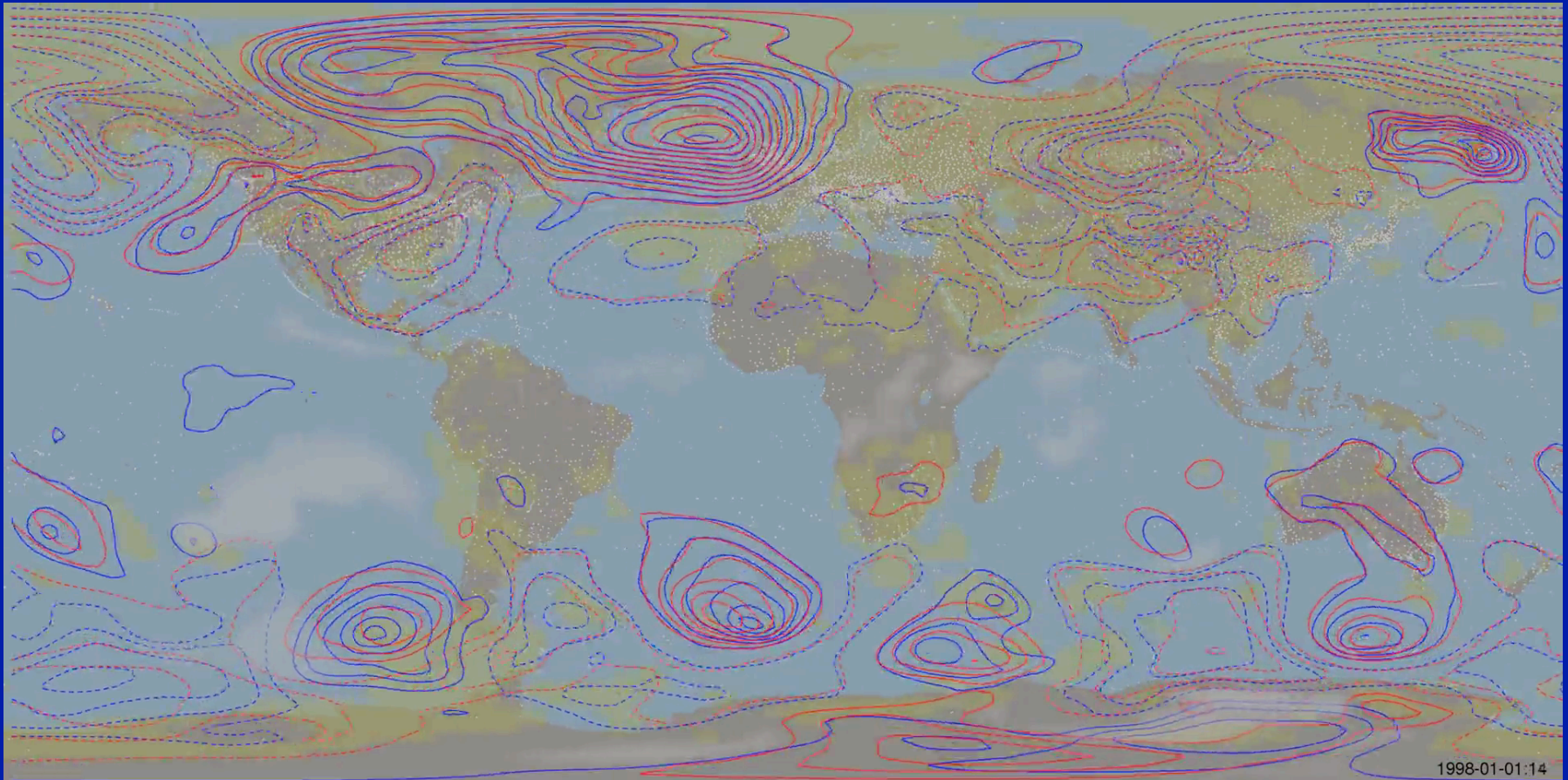
Grey dots=observations in 20CRv2c. Grey Fog=where 20CRv2c very uncertain.
20CRv2c differences with ERA-20C large over land.
Suspect latitudinally-constant covariance inflation is too small there.

January 1998 Sea Level Pressure anomalies

20CRv2c ERA-20C

(gold=20CRv2c difference from ERA-20C $> 3 \times$ spread)

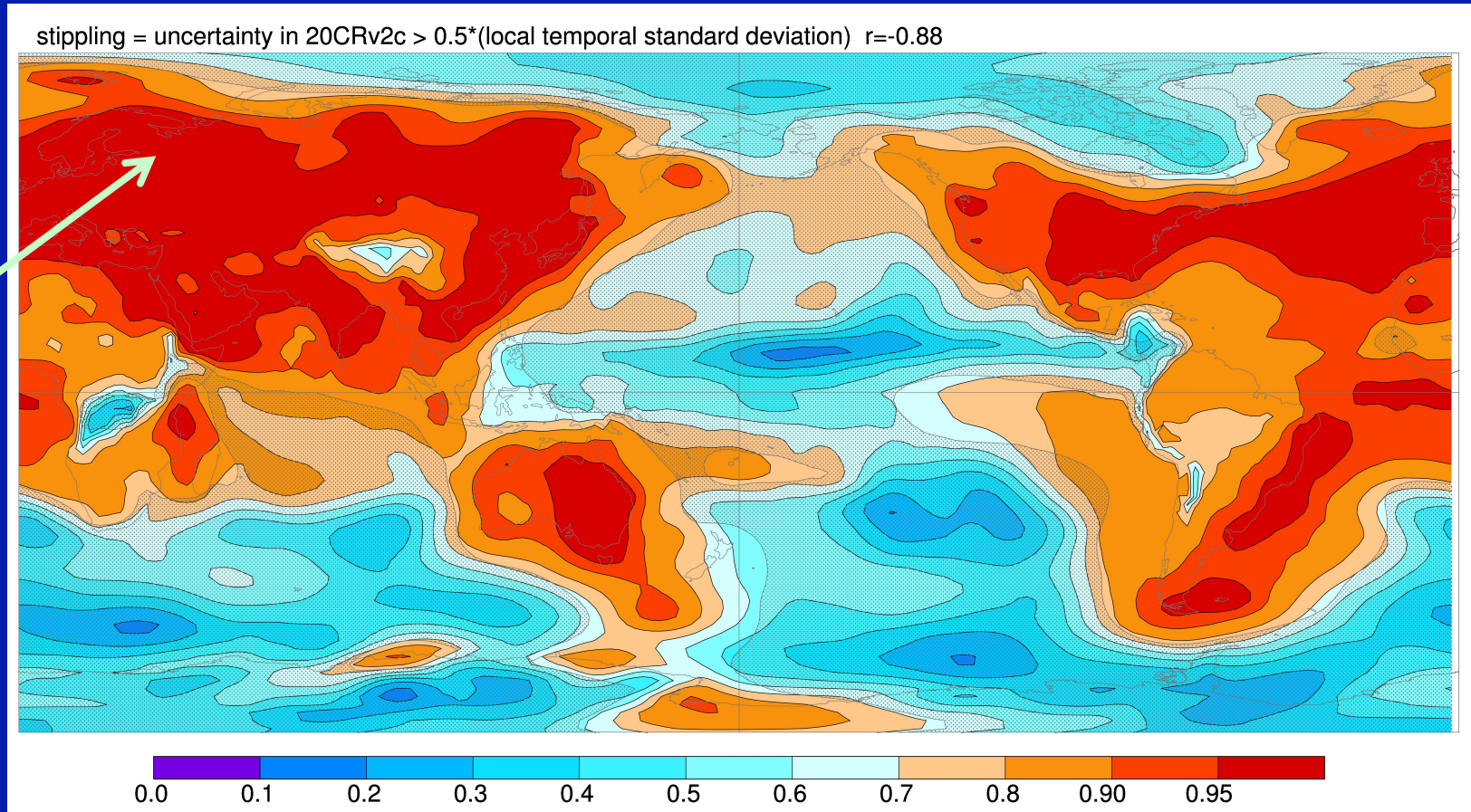
Full Brohan movie at vimeo.com/109684347



1998-01-01:14

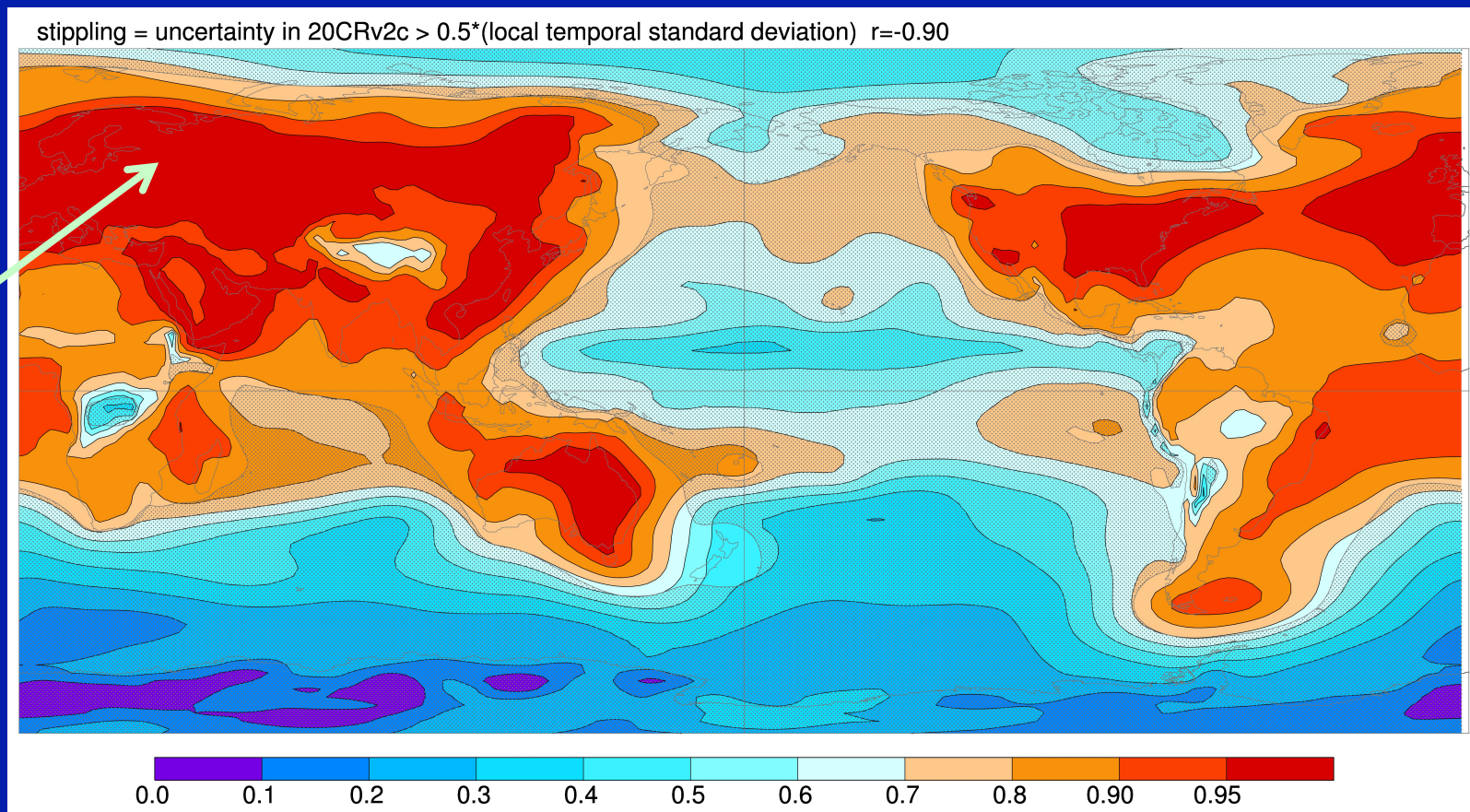
Grey dots=observations in 20CRv2c. Grey Fog=where 20CRv2c very uncertain.
20CRv2c differences with ERA-20C largest over land. Suspect latitudinally-constant covariance inflation is too small there.

Colors=Local Correlation between 20CRv2c and ERA20C
daily Sea Level Pressure during 1912
Grey Stippling= 20CRv2c is uncertain



Uncertainty field from 20CRv2c ensemble standard deviation is a good proxy for the dataset agreement (pattern correlation = -0.88).

Colors=Local Correlation between 20CRv2c and ERA20C
daily Sea Level Pressure averaged 1900-1919
Grey Stippling= 20CRv2c is uncertain



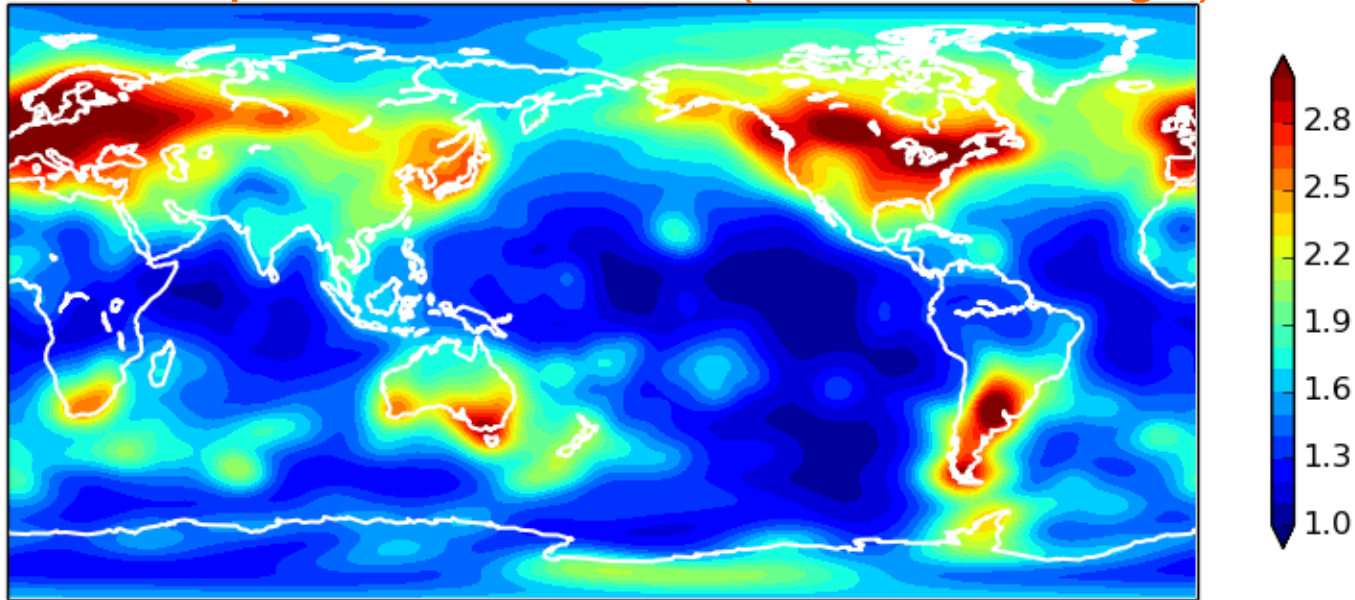
Uncertainty field from 20CRv2c ensemble standard deviation is a good proxy for the dataset agreement (pattern correlation = -0.9).

Developing 20CR version 3 Data Assimilation system

- Test Model: T126L64 GFS, circa 2013.
- 64 member EnKF
 - Varying covariance localization (based on estimated observation impact)
 - ‘Relaxation to prior spread’ inflation (larger where observations are dense).
 - New Quality Control that treats ob errors as non-Gaussian (based on Huber-norm VarQC algorithm developed at ECMWF).
 - No DFI - incremental analysis update (IAU) ending at analysis time instead.
- Sea-ice problem reduced.

New covariance inflation

Surface pressure inflation (annual average)



Relaxation to prior spread (RTPS) $\sigma^a \leftarrow (1 - \alpha)\sigma^a + \alpha\sigma^b$
which implies $\mathbf{x}'_i{}^a \leftarrow \mathbf{x}'_i{}^a \sqrt{\alpha \frac{\sigma^b - \sigma^a}{\sigma^a} + 1}$

Here we use $a = 0.9$

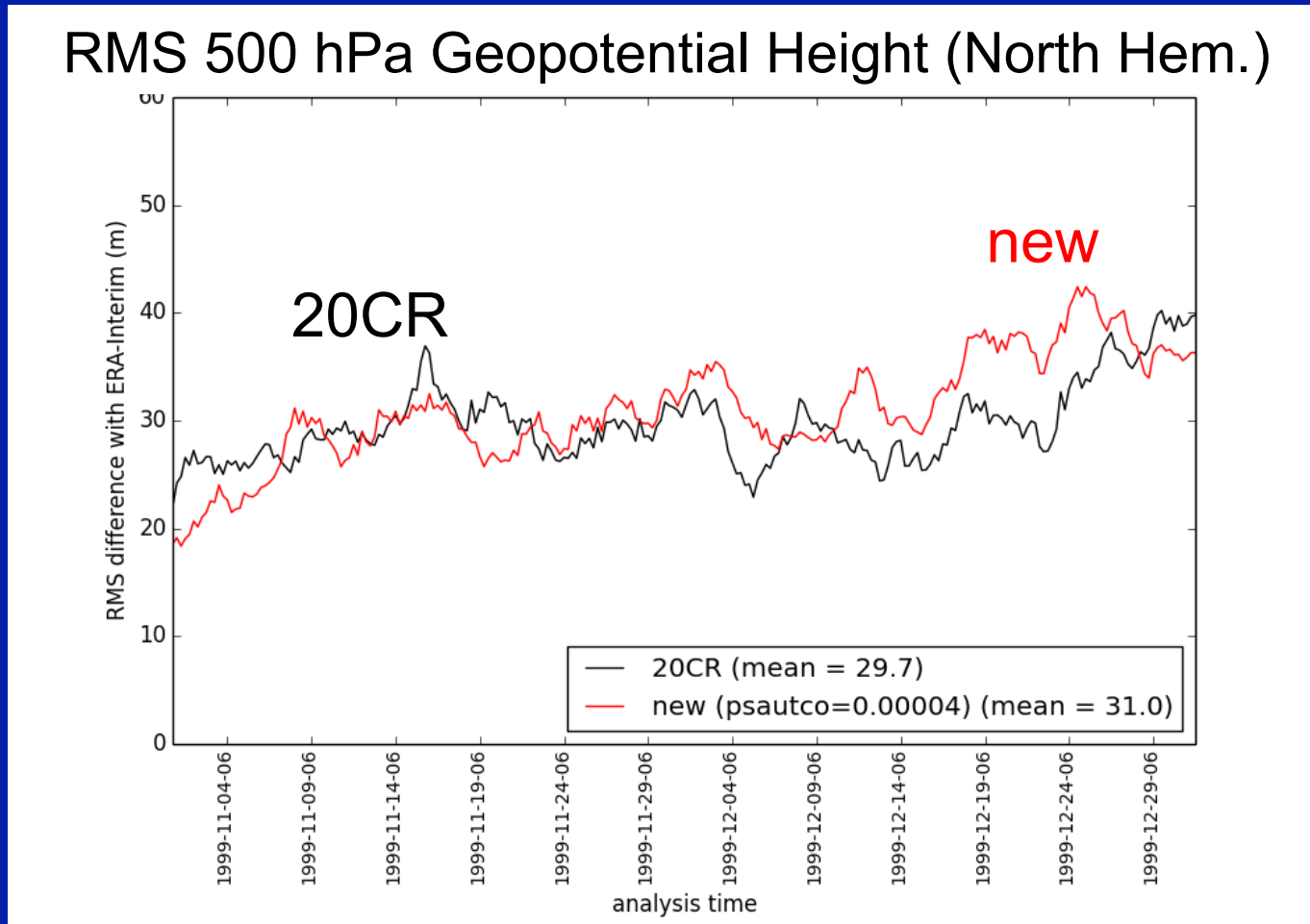
Previously, inflation was piecewise constant (NH,TR,SH)

Ref: journals.ametsoc.org/doi/pdf/10.1175/MWR-D-11-00276.1

Testing

- Start 1 September 1999 from 20CRv2 ensemble.
- Use same pressure obs and SSTs as 20CRv2, climatological sea ice.
- Run through 31 December 2000.
- Verify against ERA interim analyses.

First try using default GFS parameters



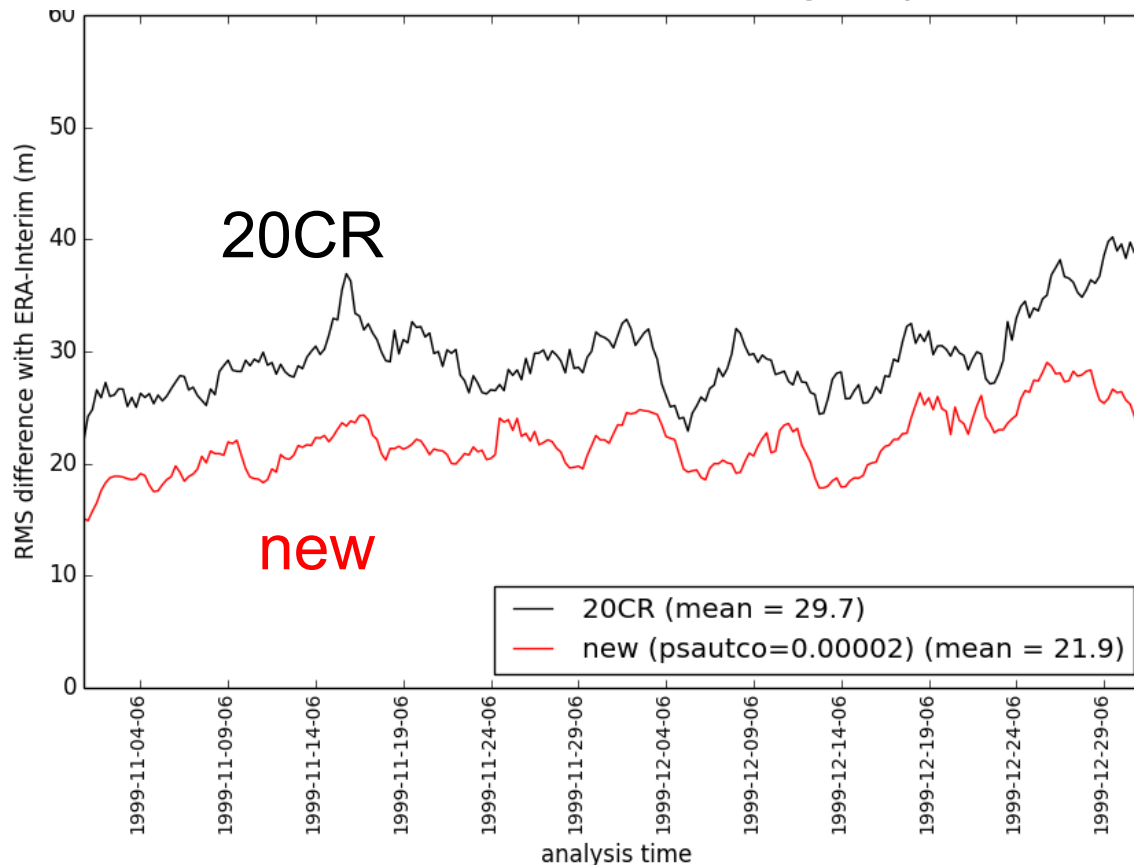
Worse than original 20CR. Large mid-upper tropospheric cold bias.

Cause: Not enough high cloud.

Solution: Reduce ice to snow auto-conversion rate in Zhao-Carr microphysics.

psautco halved

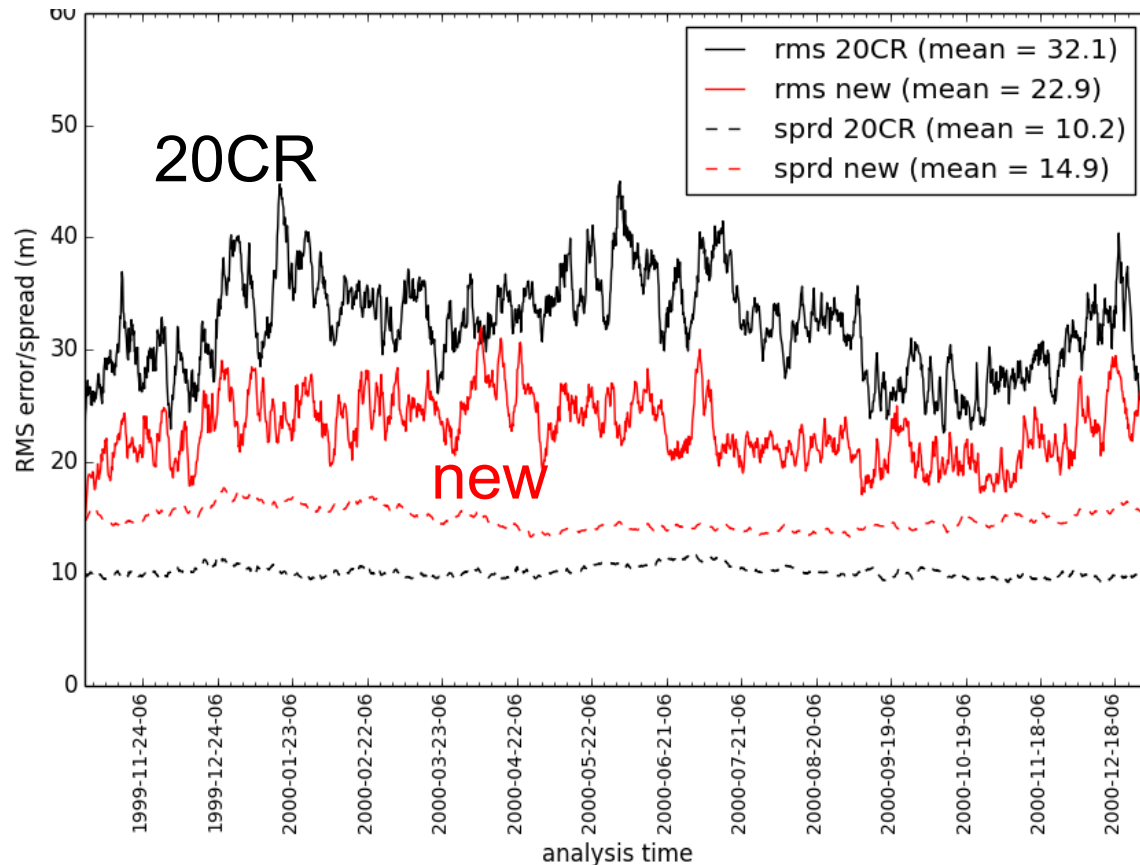
RMS 500 hPa Geopotential Height (North Hem.)



**Now much better than 20CR in Northern Hemisphere
(almost 25% reduction in error)!**

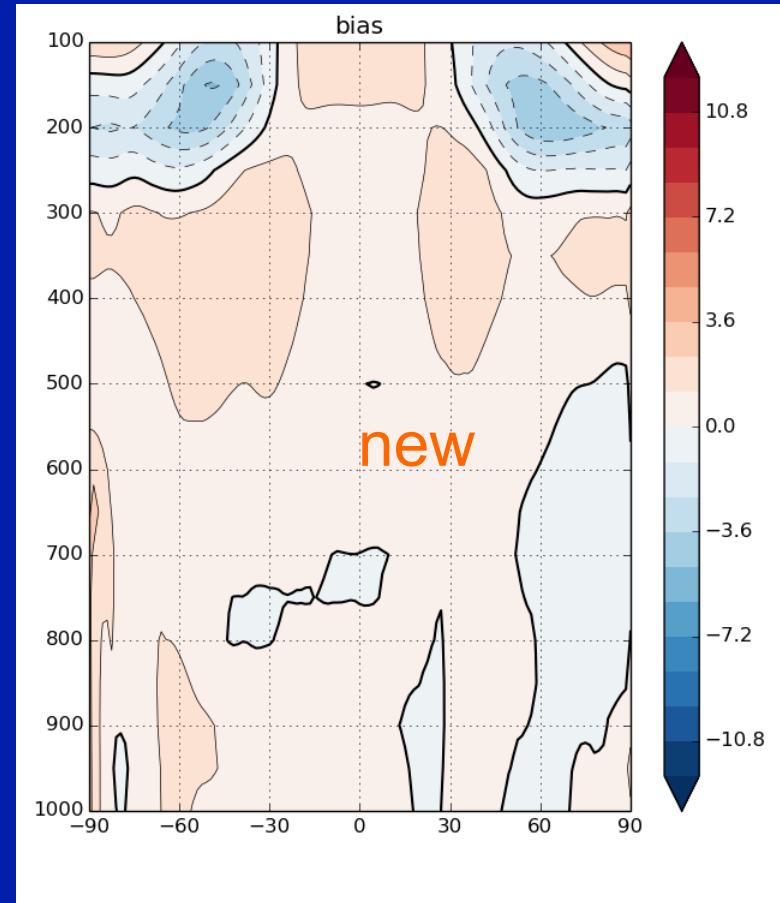
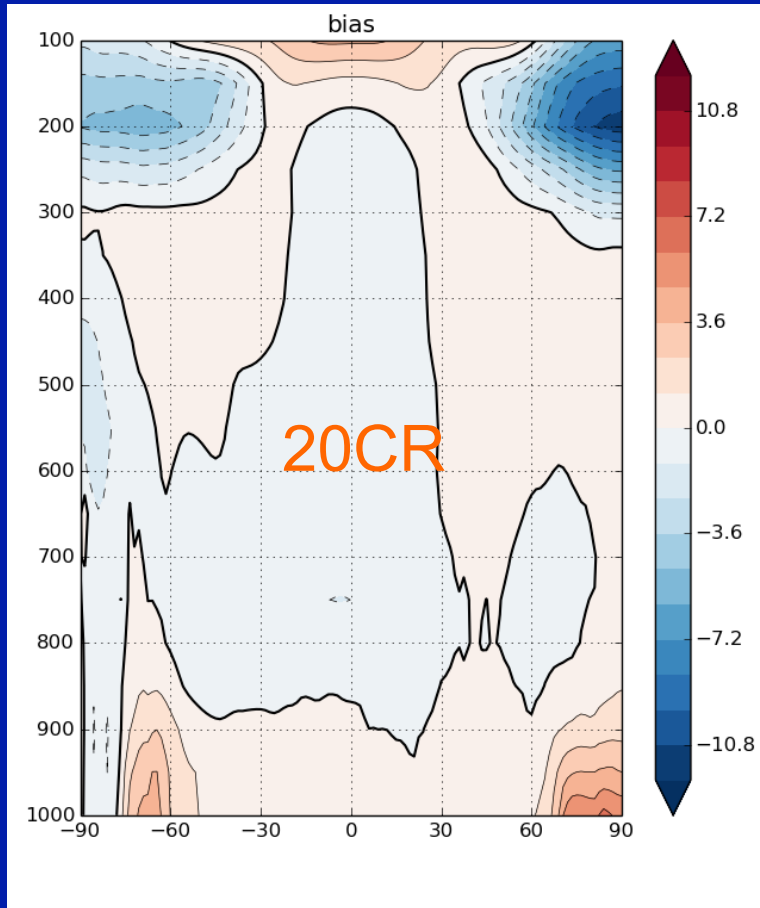
Results for full year

RMS 500 hPa Geopotential Height (North Hem.)



**28% reduction in RMS,
spread increased by 45% (now much closer to error)**

Zonal mean temperature bias



Upper trop warm bias in new system (too much high cloud?)
Near surface temp bias due to ice problem fixed.
Stratospheric cold bias reduced, but not eliminated.

Historical Reanalysis Status and Plans

20th Century Reanalysis Project http://www.esrl.noaa.gov/psd/data/20thC_Rean

- **Fall 2014:** 1871-2012 (includes time-varying CO₂, volcanic aerosols, GFS from NCEP). **Ensemble mean and spread and some individual member variables online now.**
 - http://www.esrl.noaa.gov/psd/data/gridded/data.20thC_ReanV2.html (NOAA ESRL)
 - <http://dss.ucar.edu/datasets/ds131.1> (NCAR)
 - http://portal.nerisc.gov/20C_Reanalysis **Every member** (US Dept of Energy, NERSC)
 - NERSC High Performance Storage System direct-from-tape distribution
 - Earth System Grid Federation ana4MIPS distribution and validation for IPCC AR5
 - British Atmospheric Data Center **Every member**

20CR v2c

- **Spring 2015:** 1851-2011

Very similar system to 20CRv2. 1851-2011

Corrects Sea ice issues using COBE-SST2 sea ice.

Uses ensemble of SODAsi.2 SST, more observations.

20CR v3

- **Spring 2017:** 1851-2016

Higher resolution, improved algorithm and observational quality control

Development of Suite of NOAA Climate Reanalyses

Joint NOAA NCEP, ESRL, NCDC, Univ. of Colorado CIRES

- Tiered assimilation approach
 - *0. Boundary forced* (equivalent to AMIP); 1850-present
 - *1. Historical using only surface pressure*; 1850-present
 - *2. Modern using surface and conventional data*; 1946-present
 - *3. Satellite using conventional and satellite data*; 1973-present
- Assimilation System
 - Hybrid Global Statistical Interpolation [3D-Var]/Ensemble Kalman Filter (GSI/EnKF)
 - Possibly T254L64 (~50 km resolution) for 0-2
 - Higher resolution for 3. Satellite

20CR: version "1815"

20th Century Reanalysis “1815” implementation of Ensemble Filter Algorithm

(based on Whitaker et al. 2004, Compo et al. 2006, Compo et al. 2011)

Algorithm uses an ensemble of GCM runs to produce the weight **K** that varies with the atmospheric flow and the observation network every 6 hours

Using 56 member ensemble,

prescribed 1861-1890 climatological boundary conditions:

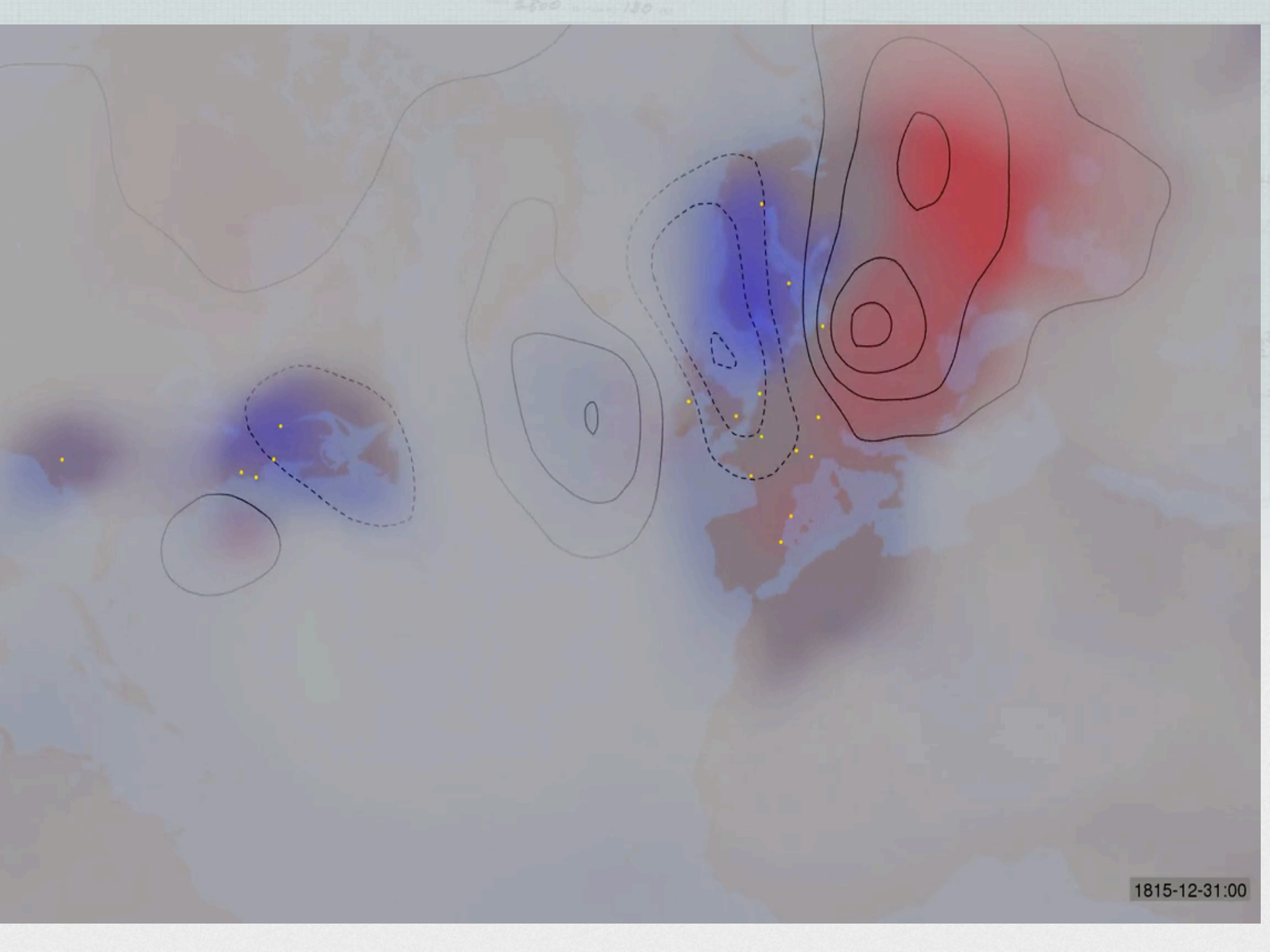
COBE-SST2 monthly SST and sea ice concentration (*Hirahara et al. 2014*)

1815-1850: T62 (~200km), 28 level NCEP GFS08ex atmosphere/land model
9 hour forecasts for 6 hour centered analysis window
- time-varying CO₂ , 11 year repeating solar cycle, and

Specified monthly volcanic aerosol optical depth:

-No Aerosols, Gao et al (2008), Crowley (2008)

Every 5 years produced in parallel: 1816-1820,..., 1846-1850, after 14 month spin-up

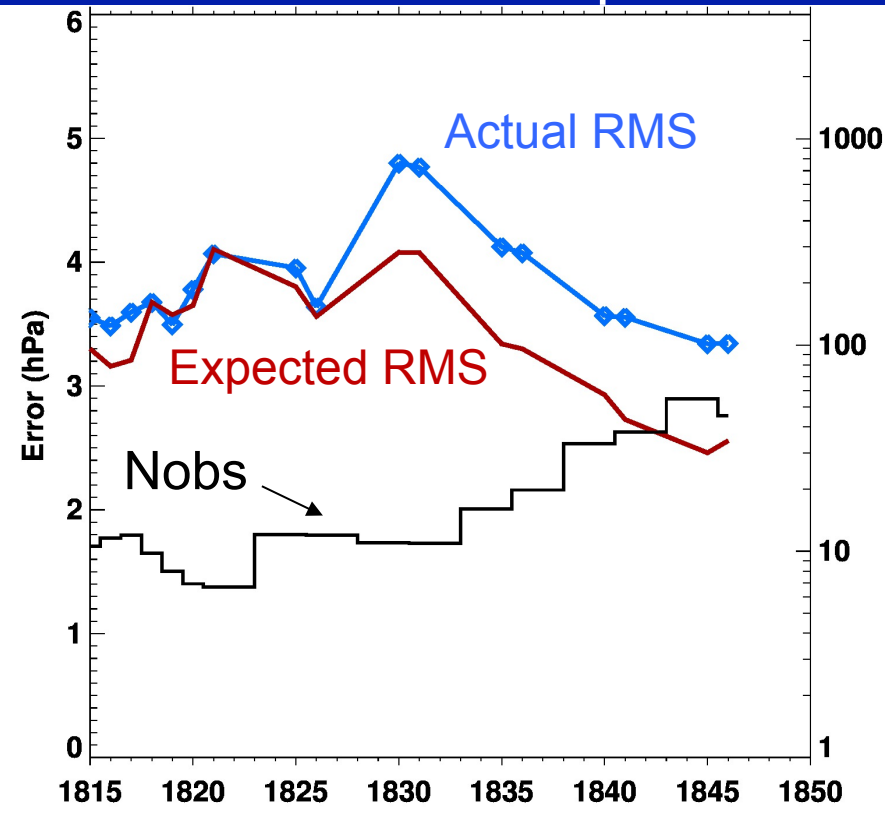


20CR:1815 Surface Pressure uncertainty estimate poleward of 20(S,N)

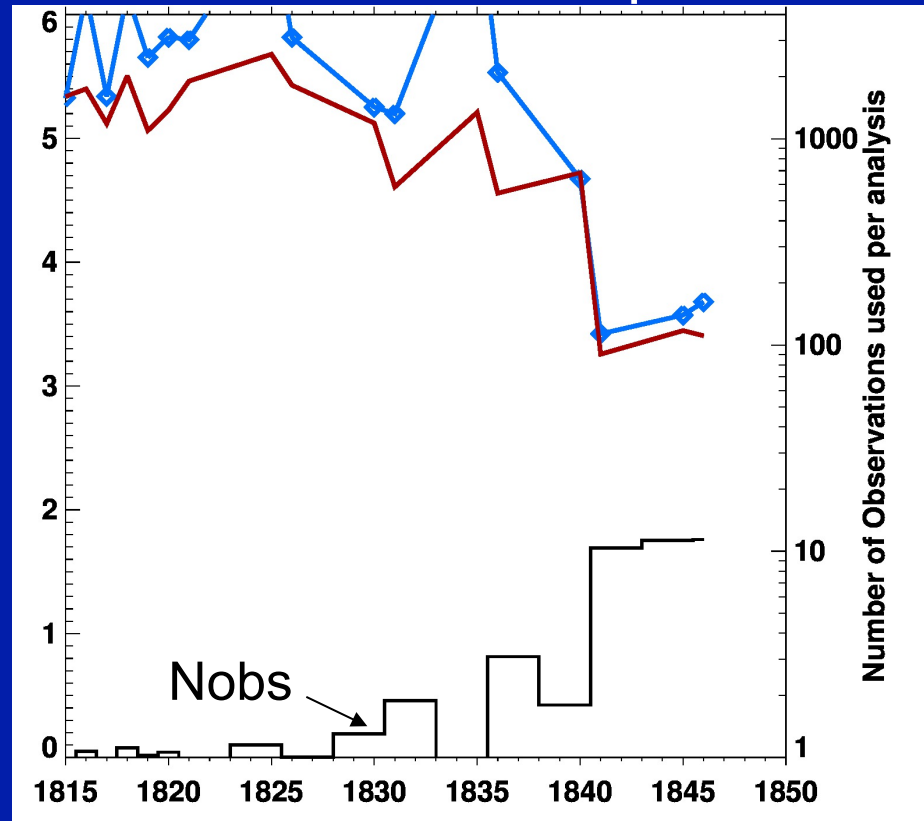
blue actual RMS difference

red expected RMS difference

Northern Hemisphere

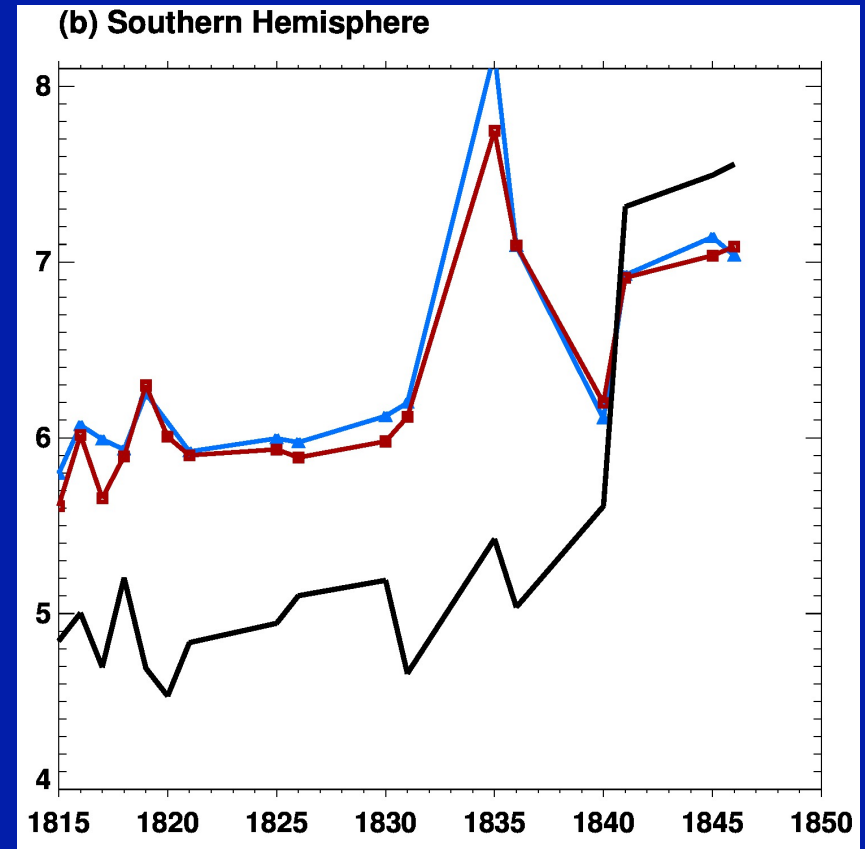
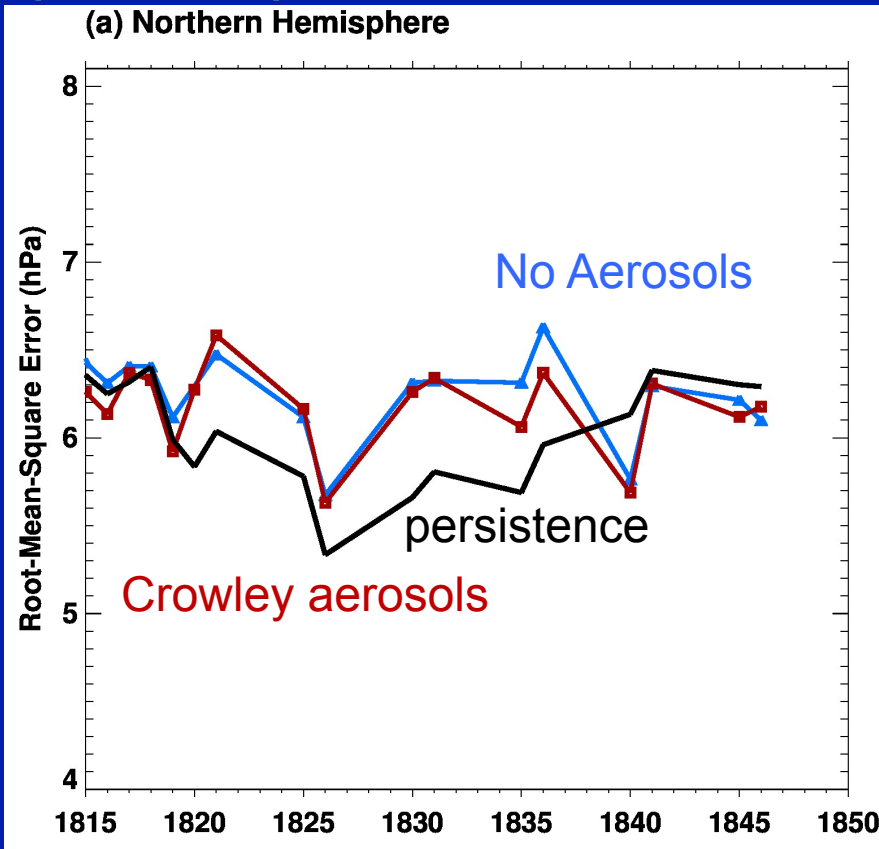


Southern Hemisphere

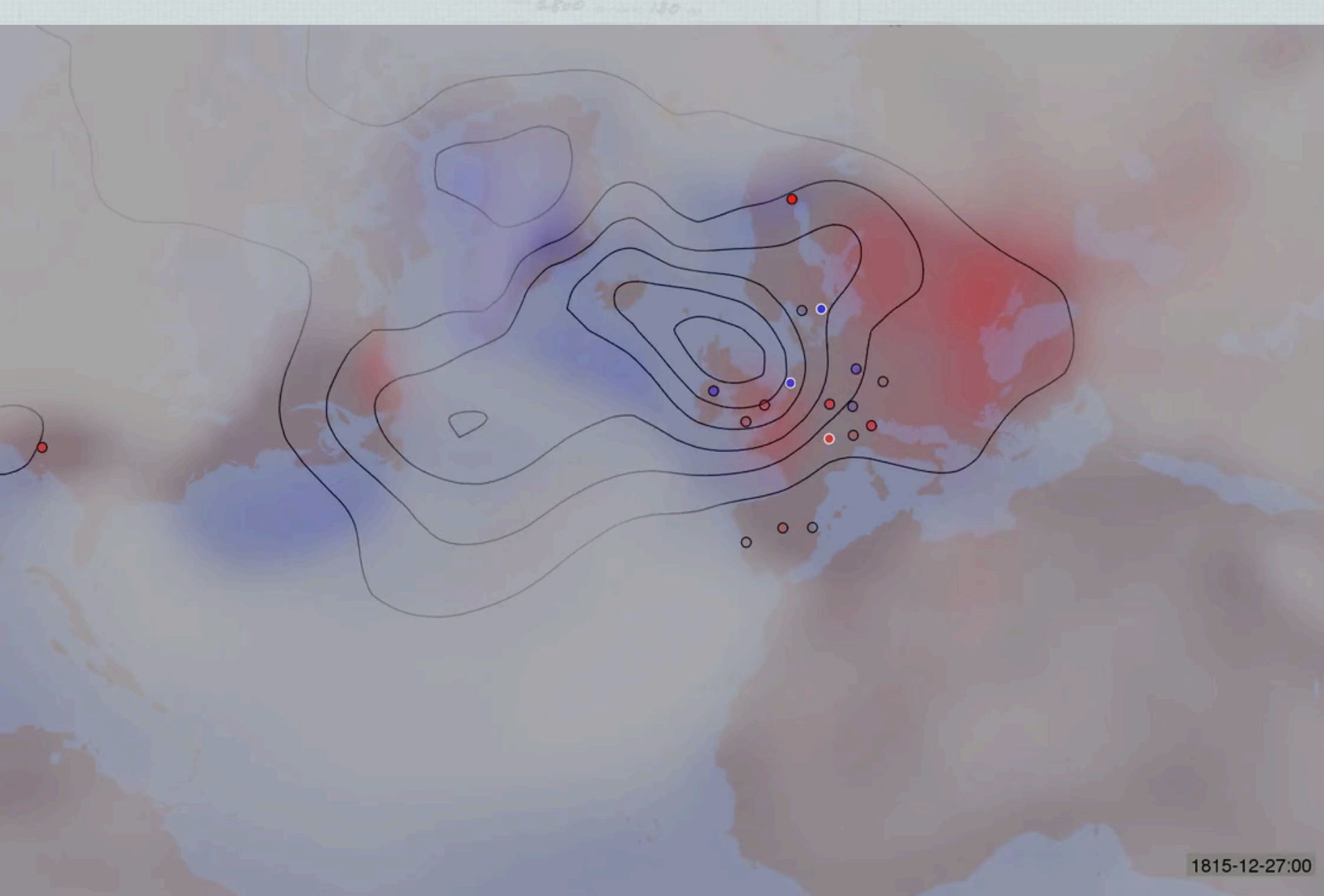


Uncertainty estimates are consistent with actual differences between first guess and pressure observations even in early 19th Century. Quantitative consistency degrades in NH after 1830s.

Root Mean Square difference of Surface and Sea Level Pressure Observations and 24 hour Forecasts from No Aerosols and Crowley Aerosols (Jan-Dec)



Northern Hemisphere 24 hr forecasts beat persistence even in 1815!
Southern Hemisphere has an analysis that produces forecasts comparable to persistence starting in 1840s with increased obs.



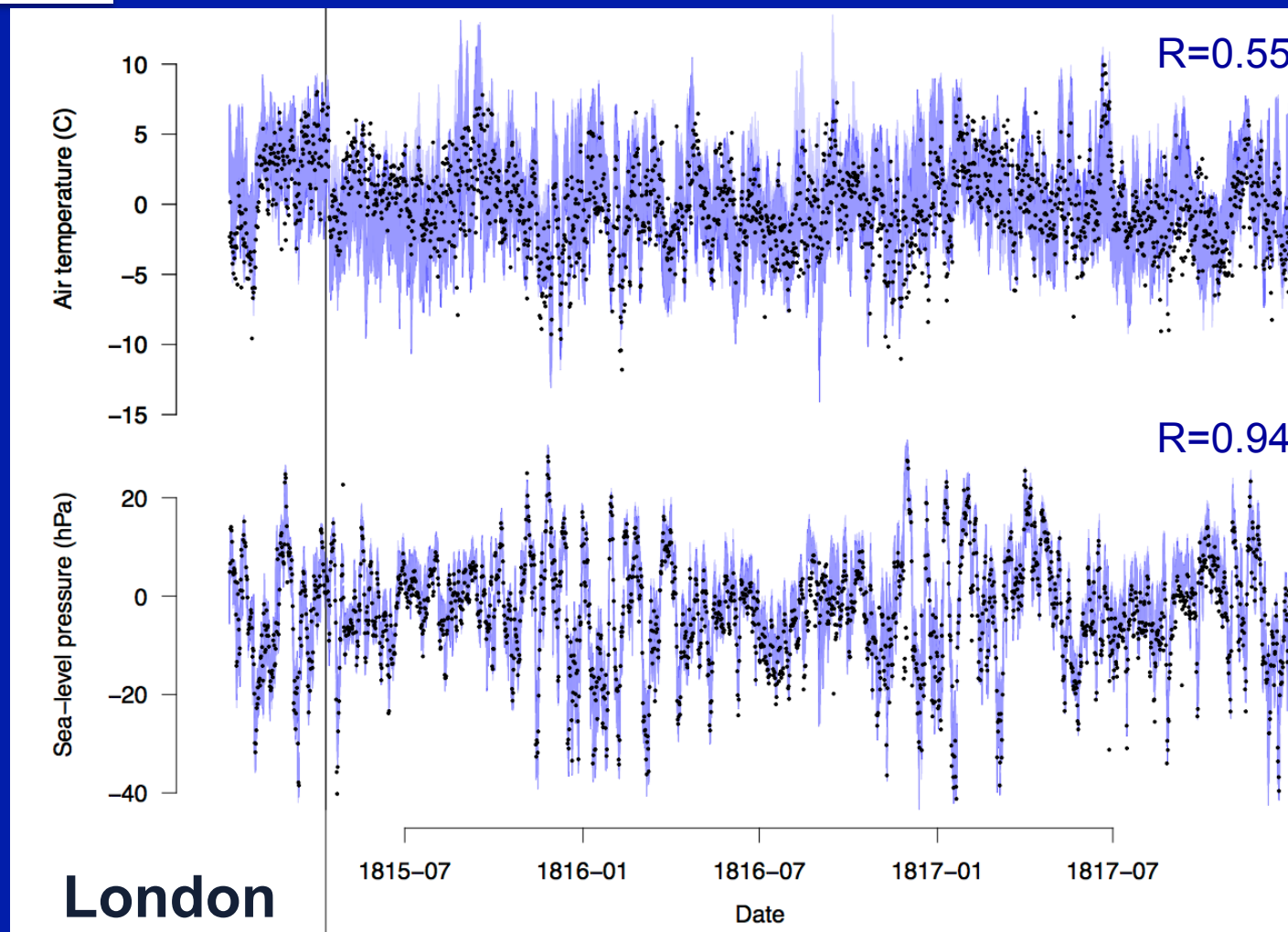
1815-12-27:00

Reconstructing the effects of Tambora 1815 and the Year Without a Summer of 1816

Comparison of anomalies from

Black dots: subdaily independent Air T and assimilated SLP from London

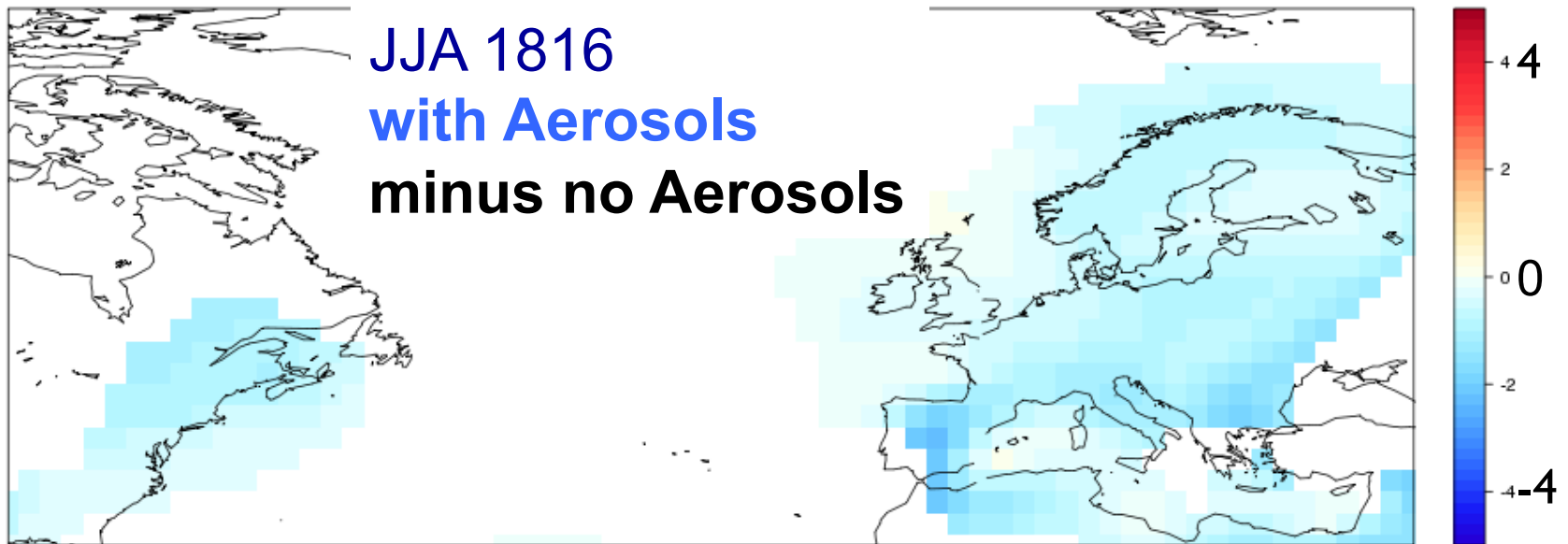
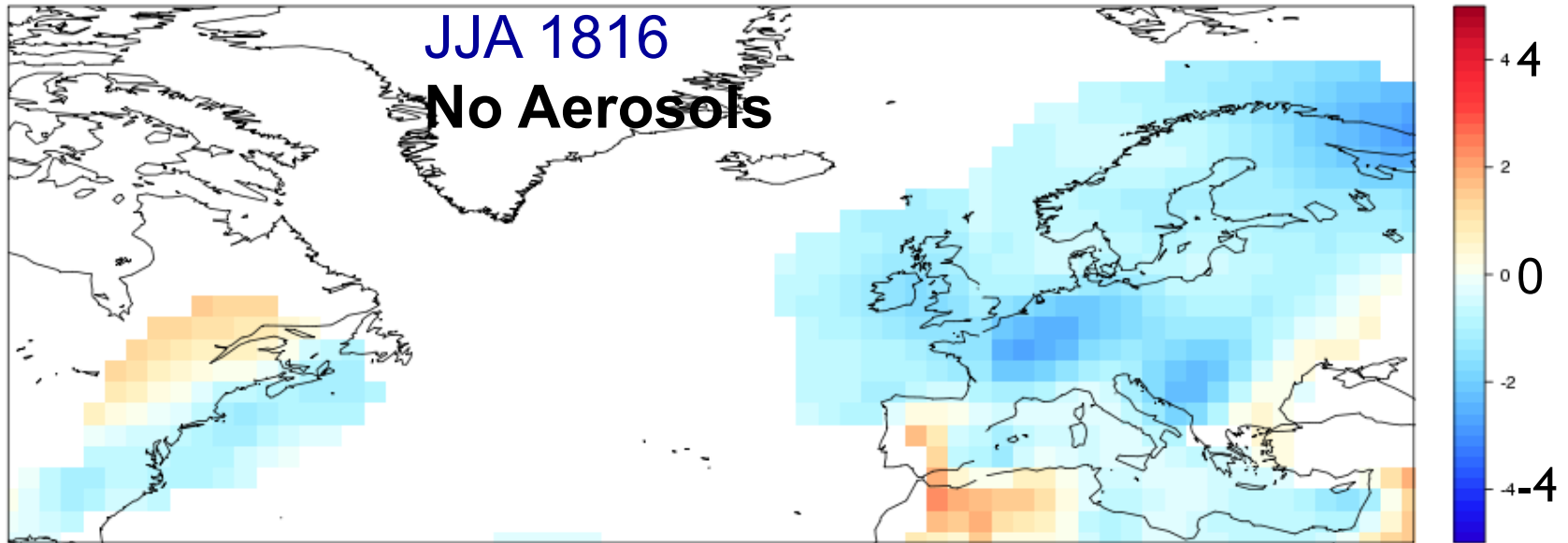
Purple swaths: 20CR-1815 ensemble range (1815 to 1817)



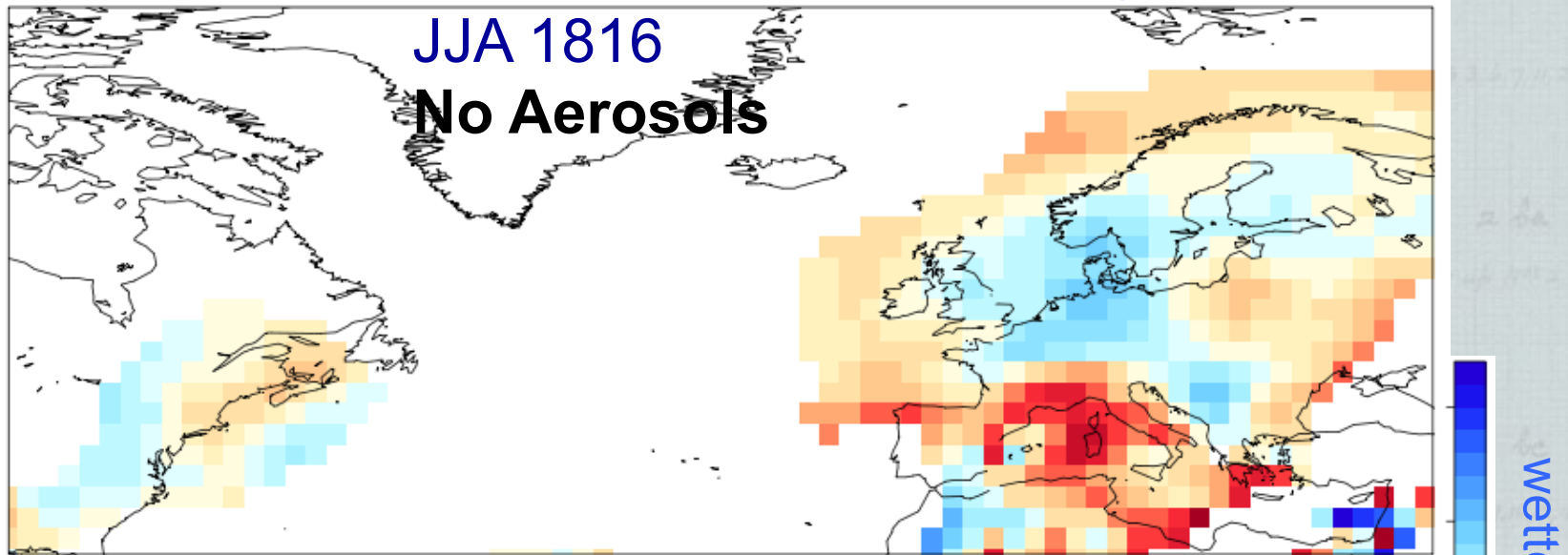
~4 other N. American and ~10 European Station, and ~10 Ship Obs assimilated each day.

2-4 times more possible.

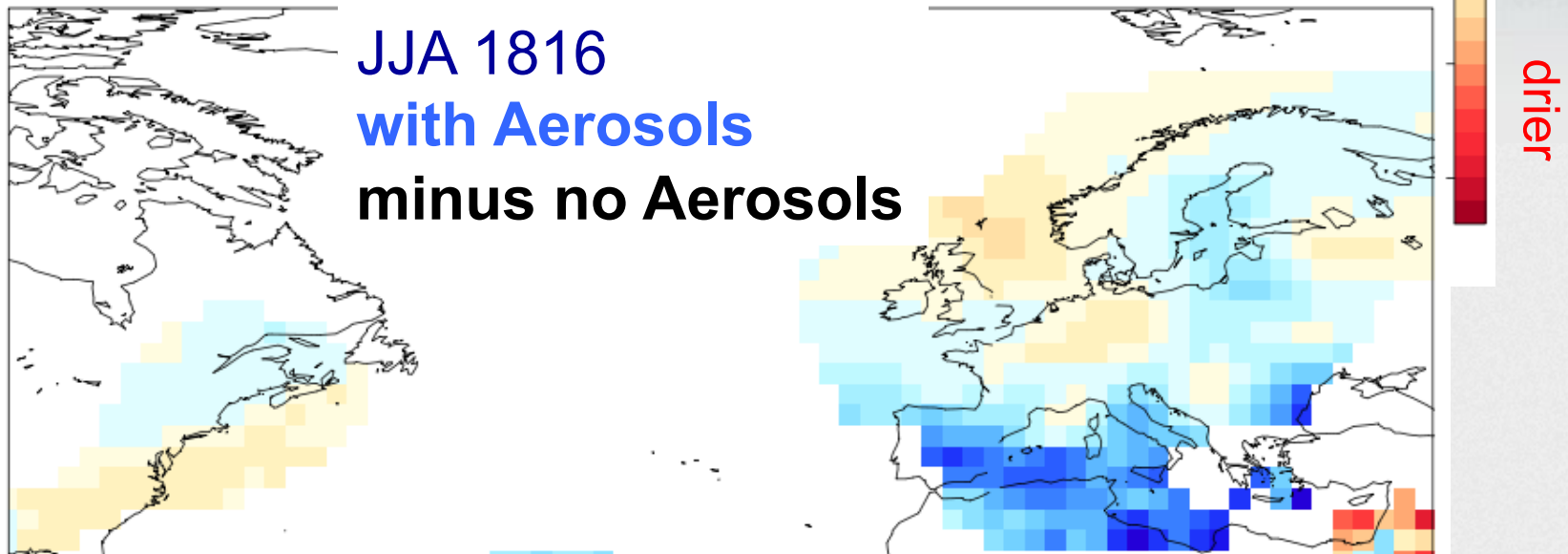
In regions such as Europe, 20CR-1815 compares well, showing skillful weather variability from the pressure observations. 1816 doesn't appear particularly anomalous in either dataset.



Additional cooling effect of adding Crowley aerosols is moderate but detectable.



20CR JJA 1816 precipitation anomaly ratio – **No Aerosols**



Additional effect of adding Crowley aerosols

Conclusions

1. Demonstrated that surface-based reanalyses throughout the troposphere are feasible using advanced data assimilation and surface pressure observations.
2. Almost tripling the reanalysis record length from ~60 year to more than 160 years, allowing current atmospheric circulation patterns to be placed in a broader historical context. 😊
3. Used in many studies of climate, weather and water variability and extremes
4. Unexpected uses: e.g., iceberg risk during the Titanic (1912), coastal storm surge risk, and Tse fly variability.
5. Coordinating with NOAA's National Centers for Environmental Prediction and National Centers for Environmental Information on a consistent suite of NOAA Climate Reanalyses using upper-air and satellite observations.
6. Data assimilation algorithm improvements can be offset by a bad model. Need to improve both.
7. Future: improved algorithm, higher resolution model, longer span (200 years could be possible! Quality would depend on region).

Thank you to organizations contributing observations to ISPD:

All Russia Research Institute of Hydrometeorological
Information WDC

Atmospheric Circulation

Reconstructions over the Earth (ACRE)

Australian Bureau of Meteorology

Australian Meteorological Association, Todd Project Team

British Antarctic Survey

Canadian Volunteer Data Rescue Project

Cook Islands Met Service

Danish Meteorological Institute

Deutscher Wetterdienst

EMULATE

Environment Canada

ERA-CLIM

ETH-Zurich

European Reanalysis and Obs for Monitoring

GCOS AOPC/OOPC WG on Surface Pressure

GCOS/WCRP WG on Obs Data Sets

Hong Kong Observatory

Icelandic Meteorological Office

IBTrACS

ICOADS

IEDRO

JAMSTEC

Japan Meteorological Agency

Jersey Met Dept.

Lamont-Doherty Earth Observatory

KNMI

MeteoFrance

MeteoFrance – Division of Climate

Meteorological and Hydrological Service, Croatia

National Center for Atmospheric Research

Nicolaus Copernicus University

Niue Met Service

NIWA

NOAA Climate Database Modernization Program

NOAA Earth System Research Laboratory

NOAA National Climatic Data Center

NOAA National Centers for Environmental Prediction

NOAA Northeast Regional Climate Center at Cornell U.

NOAA Midwest Regional Climate Center at UIUC

NOAA Pacific Marine Environmental Laboratory

Norwegian Meteorological Institute

Oldweather.org

Ohio State U. – Byrd Polar Research Center

Portuguese Meteorological Institute (IM)

Proudman Oceanographic Laboratory

SIGN - Signatures of environmental change in the
observations of the Geophysical Institutes

South African Weather Service

UK Met Office Hadley Centre

U. of Bern, Switzerland

U. of Colorado-CIRES/Climate Diagnostics Center

U. of East Anglia-Climatic Research Unit

U. of Giessen –Dept. of Geography

U. of Lisbon-Instituto Geofisico do Infante D. Luiz

U. of Lisbon-Instituto de Meteorologia

U. of Melbourne

U. of Milan-Dept. of Physics

U. of Porto-Instituto Geofisica

U. Rovira i Virgili-Center for Climate Change

U. of South Carolina

U. of Toronto-Dept of Physics

U. of Washington

World Meteorological Organization - MEDARE

ZAMG (Austrian Weather Service)

The new 20CRv2c (1851-2013)

20CRv2 system but with

SODAsi.2: 18 member ensemble of daily SSTs (1846-2011)

COBE-SST2: monthly sea-ice concentration (1850-2011)

ISPDv3.2.9: International effort to recover 100s of new stations,
new marine observations from Oldweather.org, ACRE data rescue,
over 33 new organizations contributing

Effect of some accounting for uncertainty in SST

Utility of new observations

After 14 month spin-up, 12 months produced for every 5th year.

1831, 1836,...,1851,1856,...,2007 complete

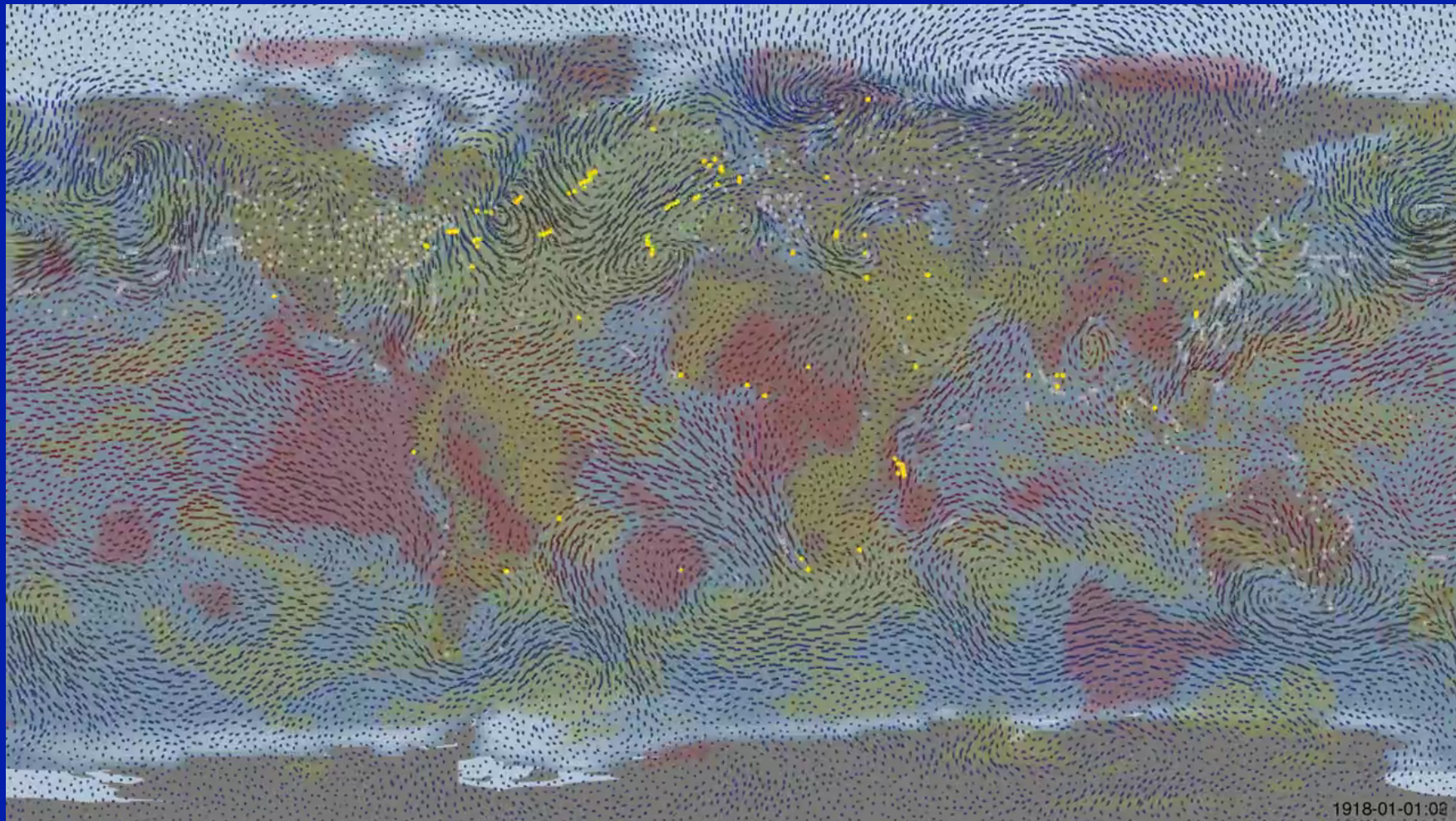
Compare to 1918 using same configuration but with

No Oldweather observations (spin up from July 1917)

January 1918

(20CRv2c worse than No Oldweather)

(20CRv2c improves upon No Oldweather)



1918-01-01:02

Grey dots=observations in 20CRv2c and in No OldWx

P. Brohan, UK Met Office

Gold dots=New observations from Oldweather.org used in v2c

Vectors=low-level wind, red=warm, blue=cold

5 Quality Control steps are part of the 20CR data assimilation system

1. **Plausibility check:** reduce to SLP, is observation reasonable (between 880 and 1060 hPa)?
2. **Background check:** check difference of observation and first guess. Flag $|y_o - x_b| > 3(\sigma_o^2 + \sigma_b^2)^{0.5}$
3. **Buddy check:** can return observations that fail the Background check.
4. **Bias correction of stations**
5. **Thinning:** F-test σ_a^2/σ_b^2 before each observation is assimilated, only use observations that significantly reduce spread.

Buddy Check makes an analysis using only one observation at a time

1. Make an analysis using only the current n th observation.

x_a = single observation analysis

2. Is the analysis error less than the first guess error when evaluated using the neighboring k observations within 1000 km

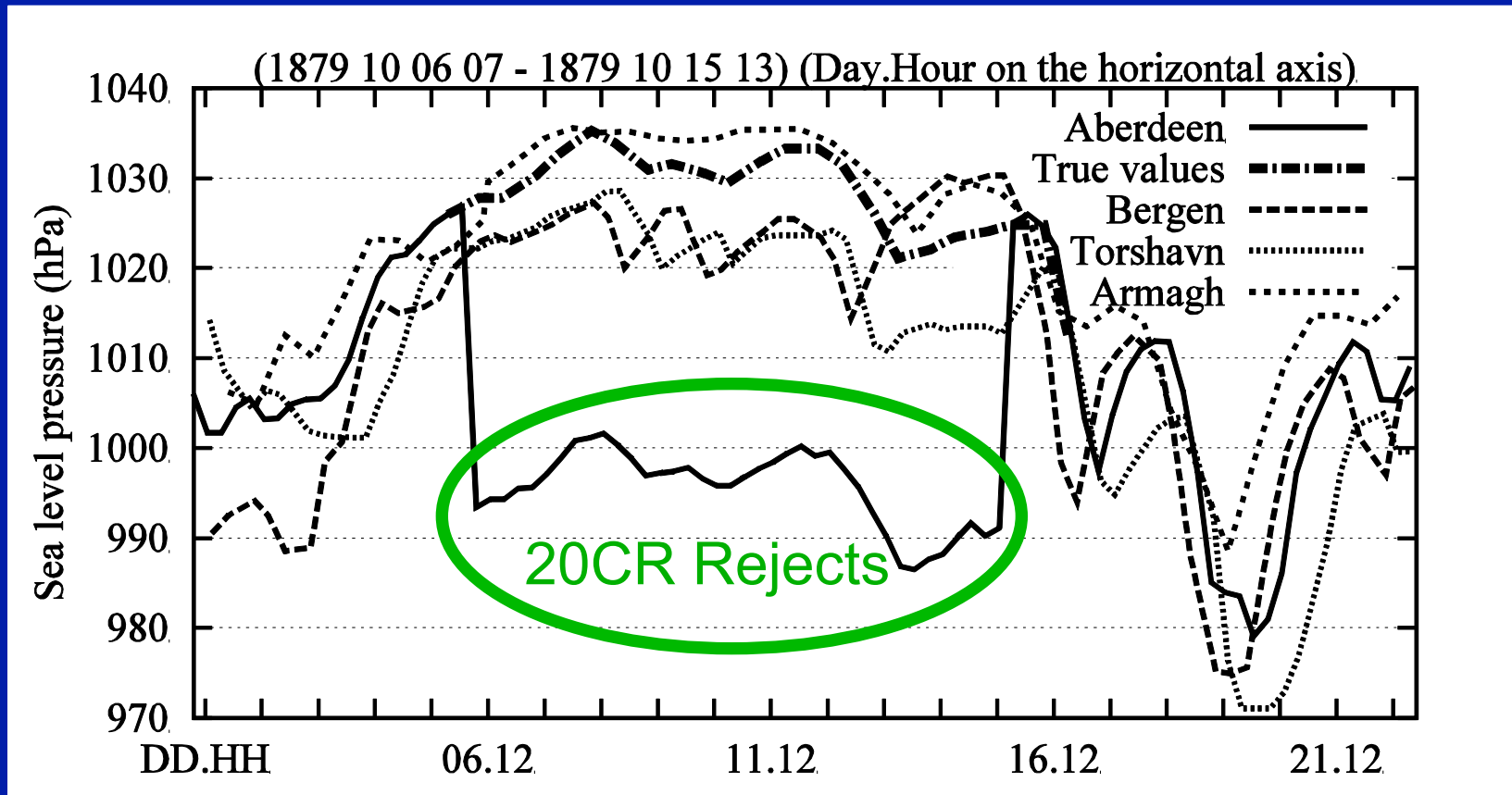
$$\sum_{k \neq n} |y_{ok} - x_a|^2 < \sum_{k \neq n} |y_{ok} - x_b|^2 ?$$

Yes: Retain the observation even if it failed the Background check and allow it to be a buddy on next iteration.

No: Reject the observation, even if it passed the Background check.

Example of correctly rejected data

SLP from multiple stations in the eastern North Atlantic
6-10 October 1878



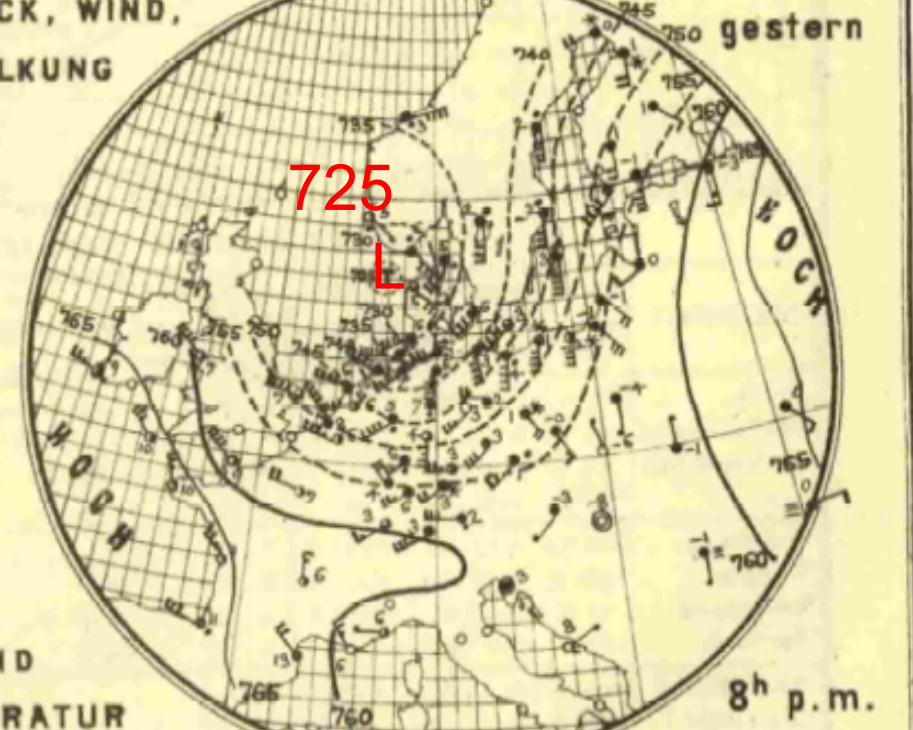
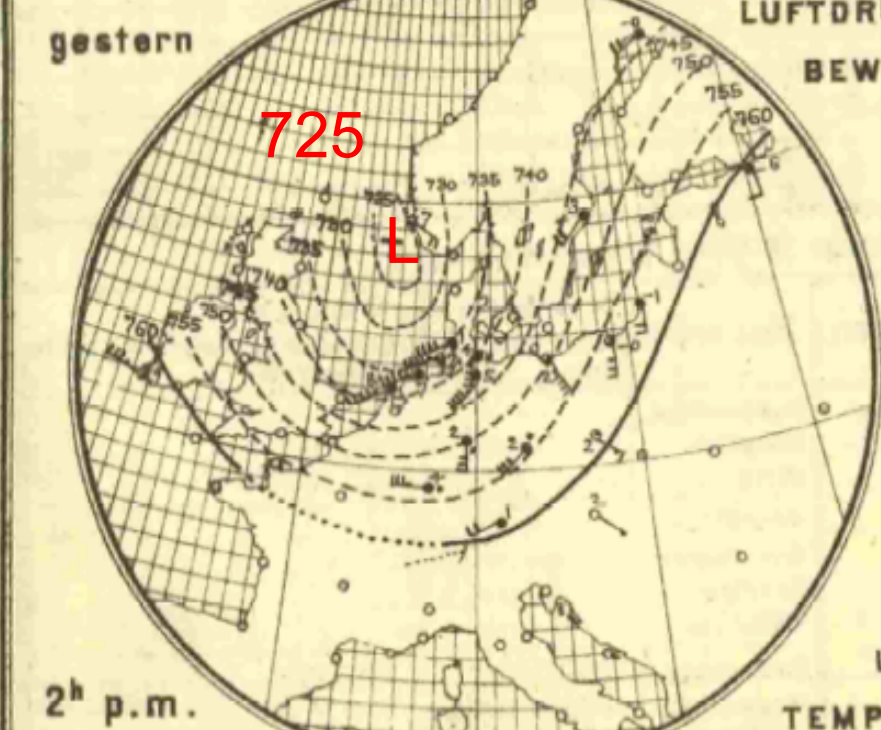
The extreme pressure gradient leads to large wind estimates in this decade if the data are used (Krueger et al. 2013).
20CR does not show such extremes.

Wang et al. 2013

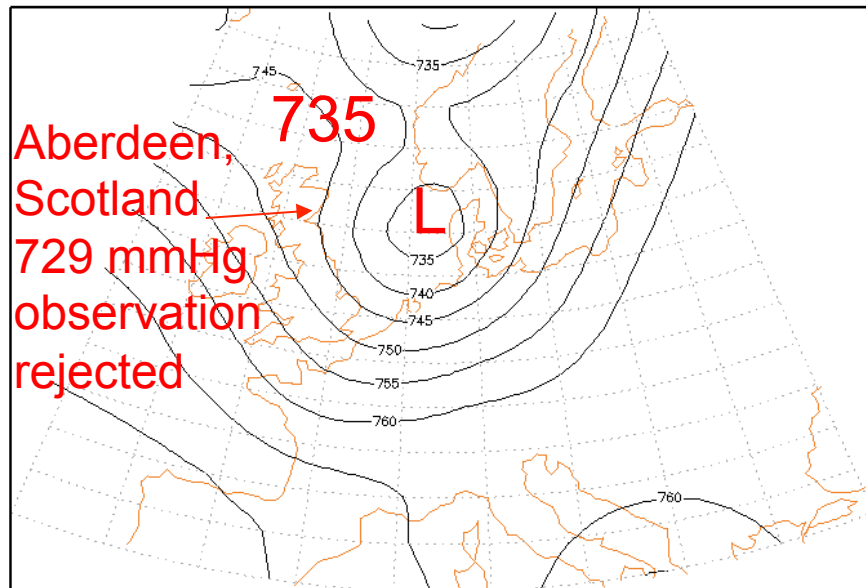
De Storm van 1894 (Zenit. 2010)

Henk de Bruin and Huug van den Dool

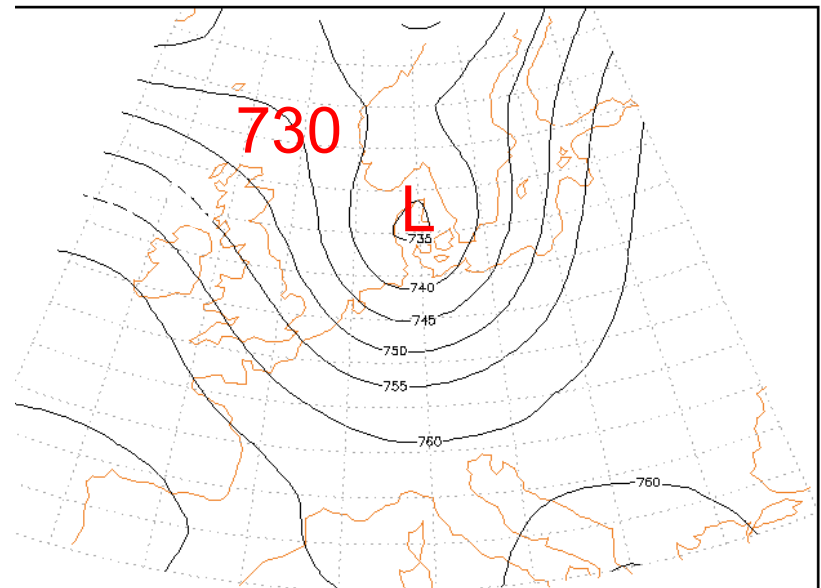
Frank Beyrich and Britta Bolzmann (DWD)
provided 1894 weather maps of the
Seewarte Hamburg



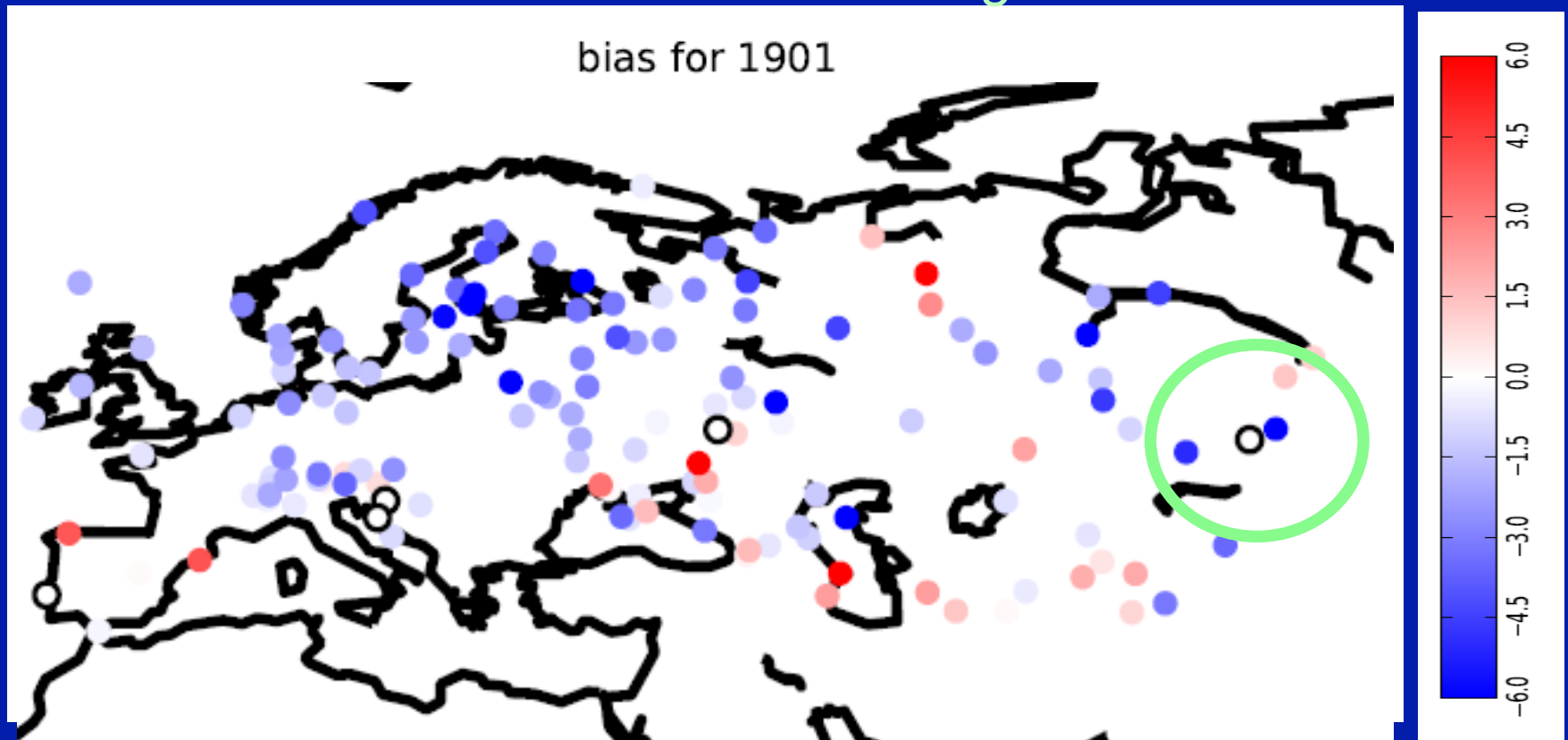
Reanalysis Storm 22 dec 1894 12Z



Reanalysis Storm 22 dec 1894 18Z

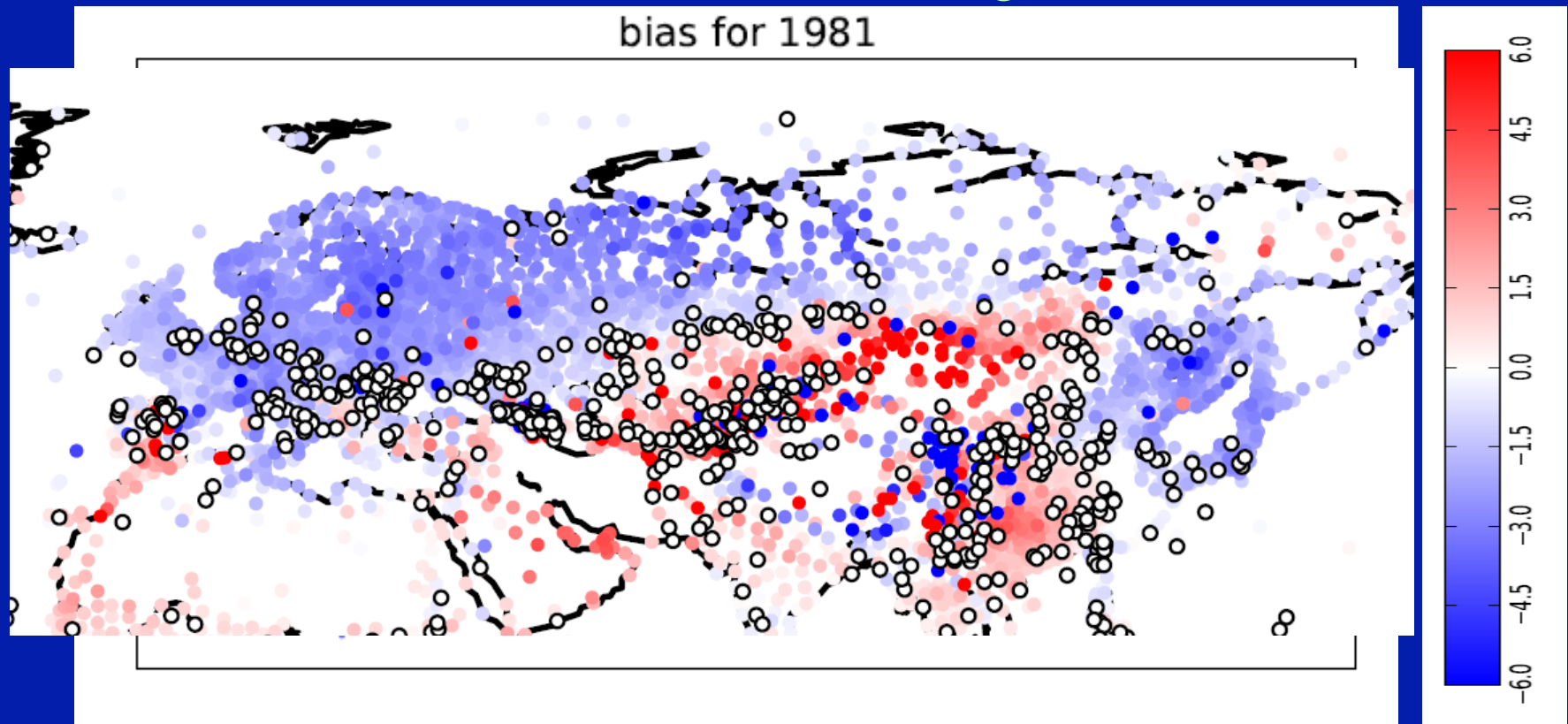


Station Bias correction algorithm corrects for Undocumented station moves, elevation errors
Remove statistically significant (paired t-test) differences between station observation and first guess.



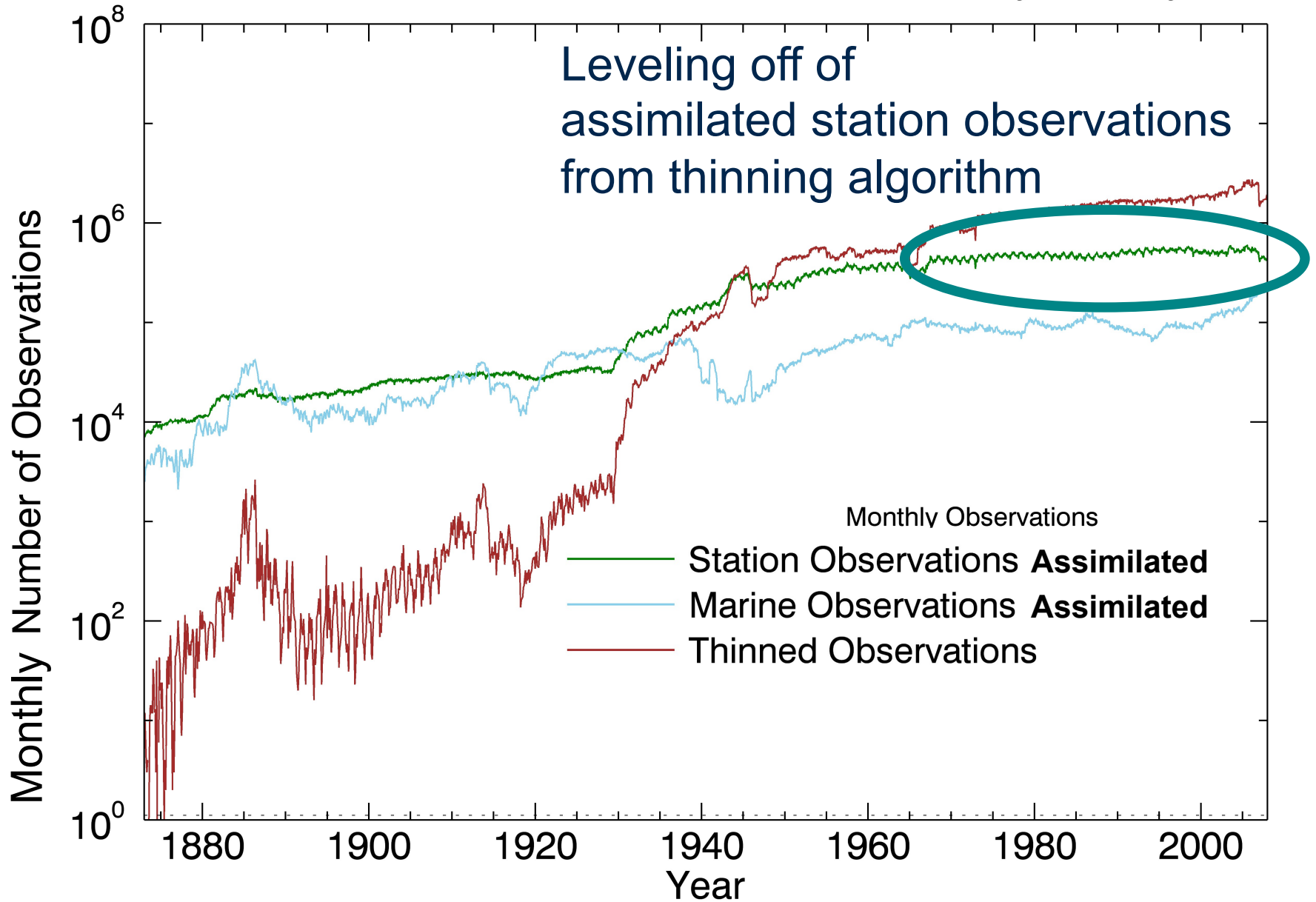
Large-scale coherence of the bias suggests large-scale model error may be attributed to stations. National boundaries suggest some issues may be national network wide.

Station Bias correction algorithm for such issues as undocumented station moves, elevation errors
Remove statistically significant (paired t-test) differences between station observation and first guess.

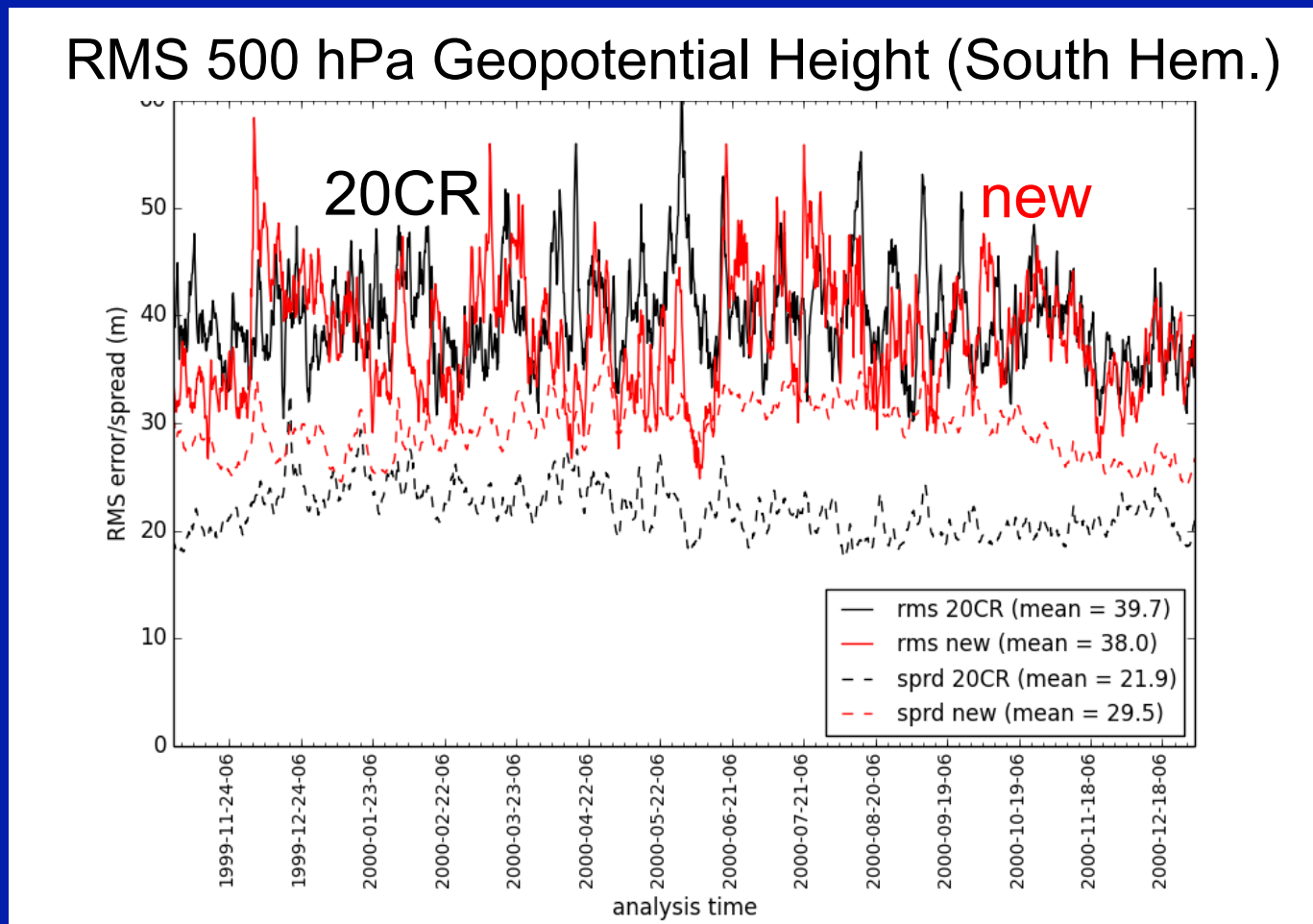


Large-scale coherence of the bias suggests large-scale model error may be attributed to stations. National boundaries suggest some issues may be national network wide.

Number of Useable Observations in 20th Century Reanalysis



Results of v3 test for full year



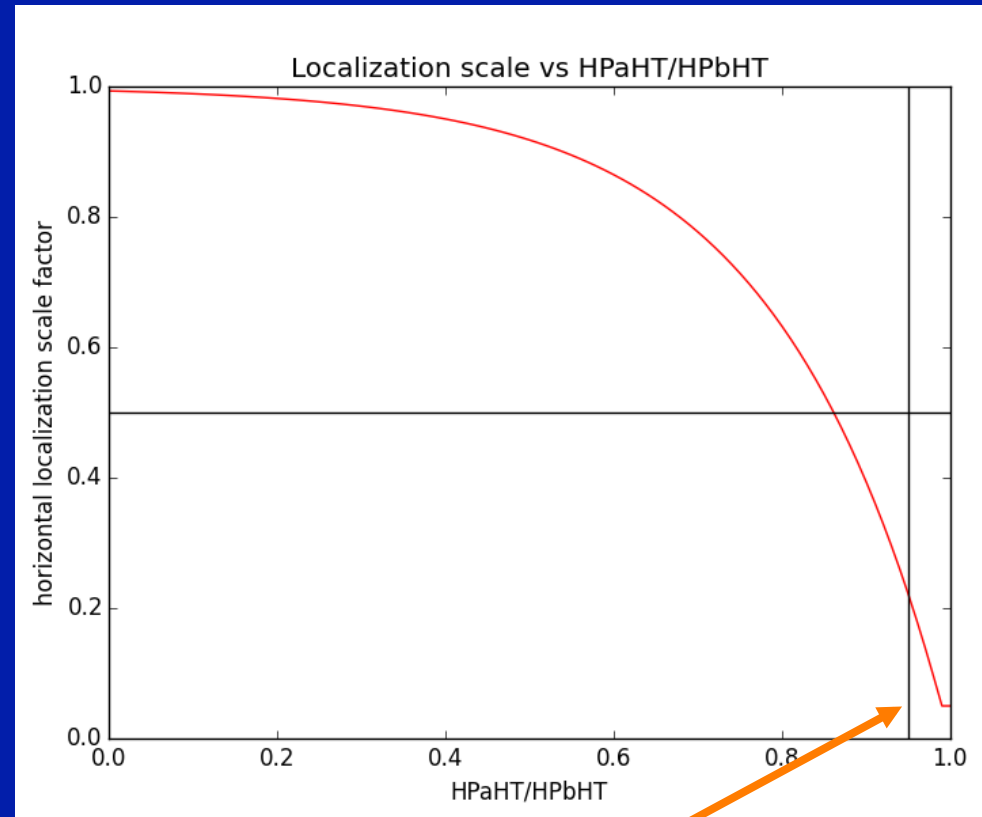
**Not much reduction in RMS for Southern Hemisphere
(but spread to RMS consistency improved)**

Adaptive Covariance Localization

Length to zero influence is
4000 Km times ratio of

Analysis spread $HPaH^T$ from
assimilating one observation
to
First Guess spread $HPbH^T$.

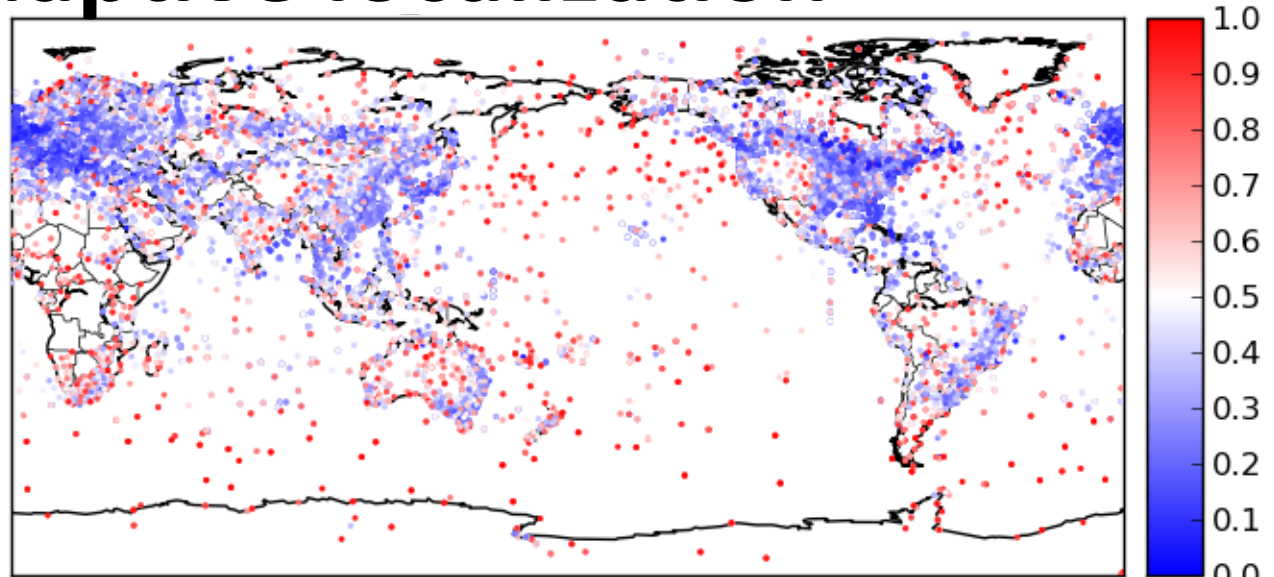
Analysis becomes
First Guess for next
observation.



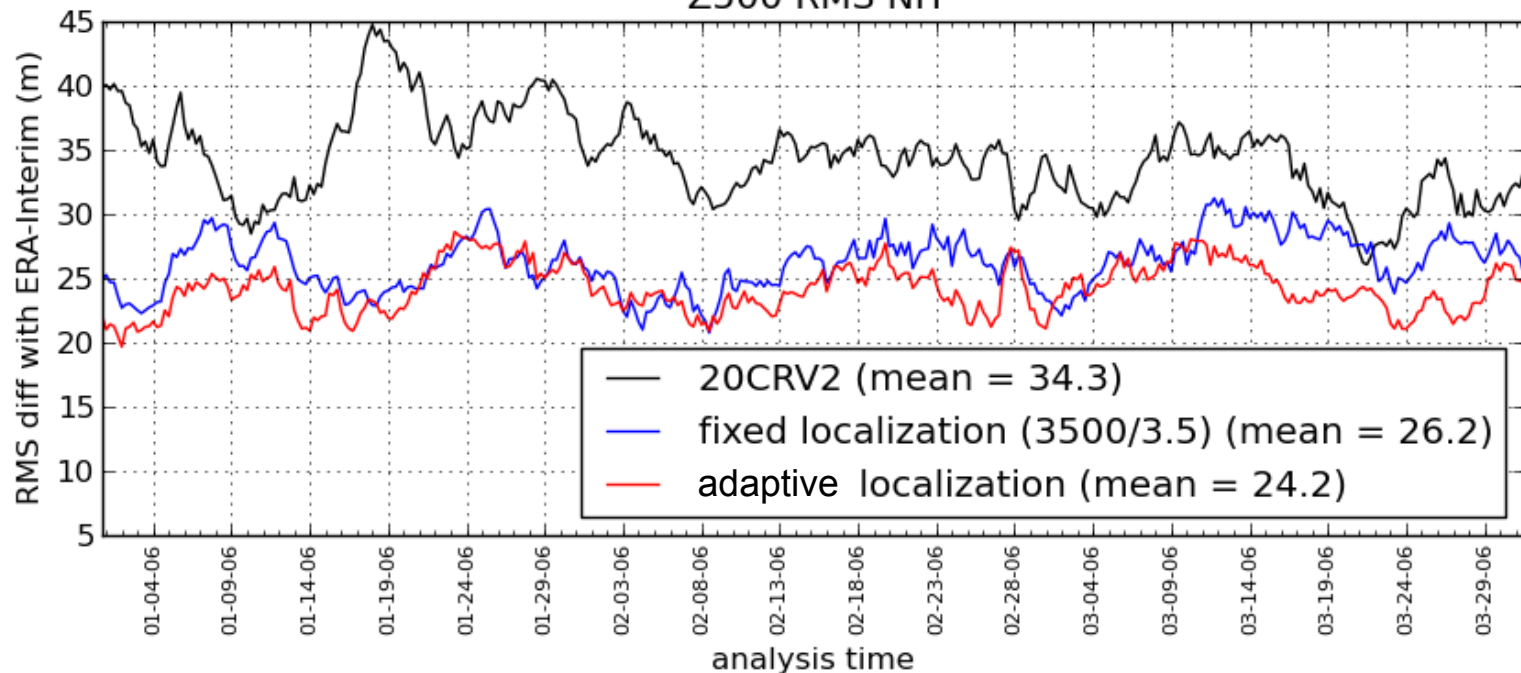
20CRv2 (and 2c) uses box car function.

Adaptive localization

Dots are ps ob locations.
Blue → short localization
Red → broad localization



Z500 RMS NH

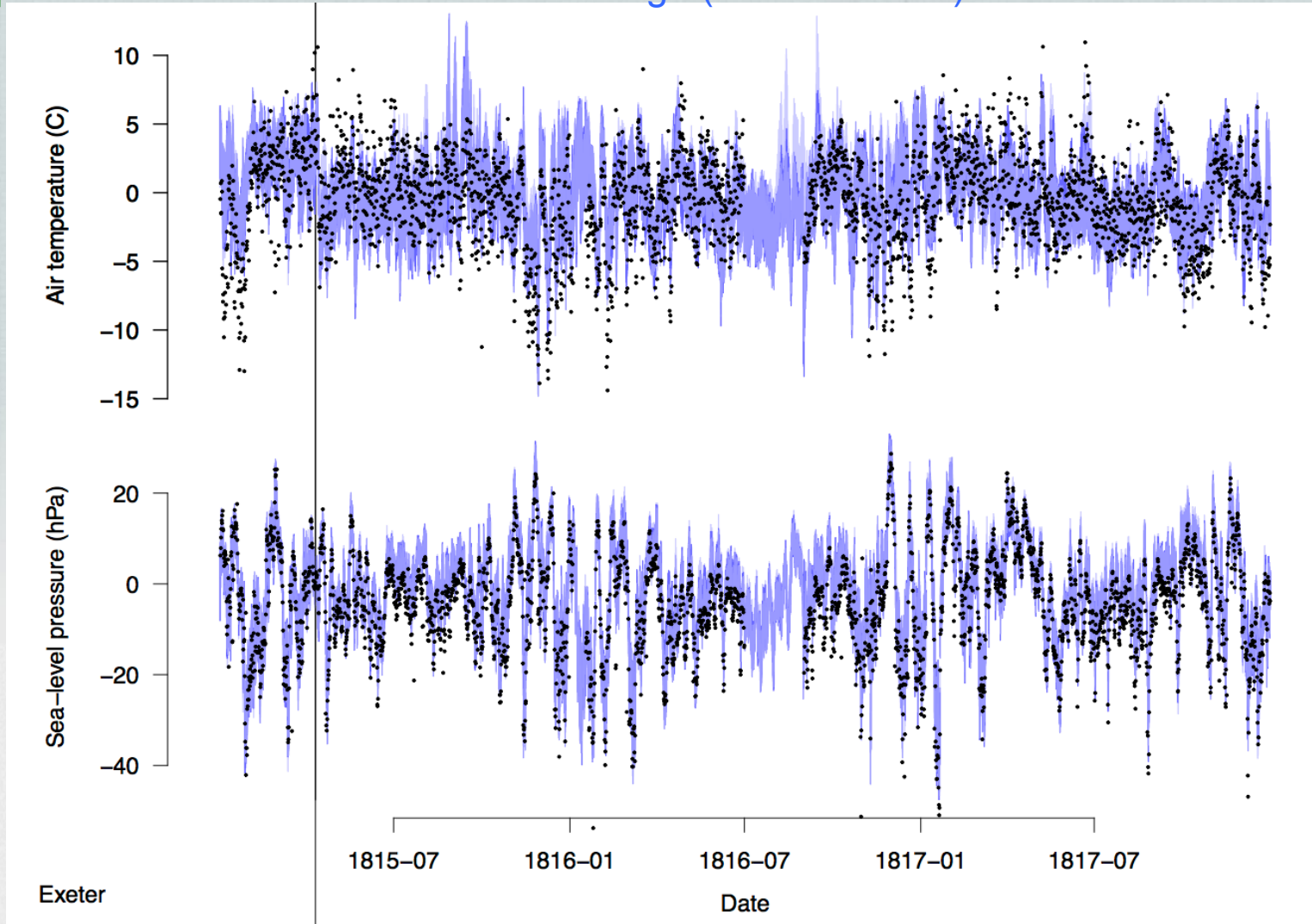


Adaptive
better about
85% of the
time

Comparison of anomalies of

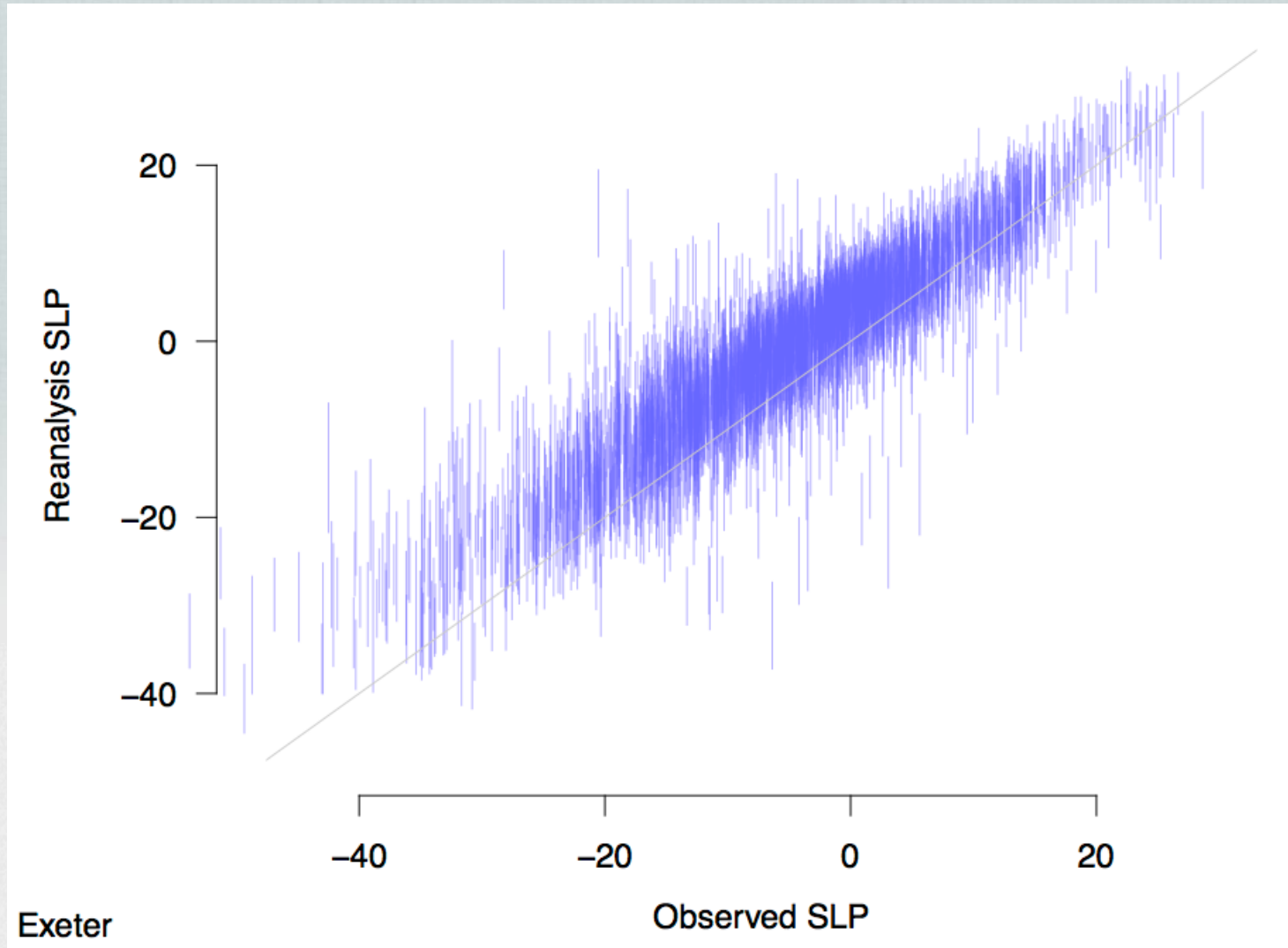
Black dots: subdaily independent Air T and independent SLP from Exeter

Purple swaths: 20CR-1815 ensemble range (1815 to 1817)



Both variables compare well, though pressure is more precise. Extreme temperature anomalies are muted

Comparison of anomalies from
Subdaily independent SLP observations from Exeter vs.
Reanalysis (20CR-1815) 2*ensemble spread (1815 to 1817)



Low pressure extremes are muted, but otherwise pressure successfully predicted at the independent station.