# Blended global temperatures versus global surface air temperature: a view from reanalysis

Taylor, M.<sup>1\*</sup>, Osborn, T.<sup>1</sup>

<sup>1</sup>Climatic Research Unit, University of East Anglia

\* <u>michael.a.taylor@uea.ac.uk</u>





## A GloSAT story: Part I (land obs spatial coverage evolution)

60°N Martin Stendel 30°N Help! Data coverage of observations available for use by the 0 @ECMWF data assimilation system today at 00 UTC. I am looking for either similar plots for the past or 30°S timeseries (e.g. radiosonde data) over the course of the years. 60°S I know such figures exist, but can't find them. Thx! ECMWF data coverage (all observations) - SYNOP-SHIP-METAR 29/09/2020 00 180°W 120°W 60°W 0° 120°E 180°E ٩°E Total number of obs = 121372 Automatic Land SYNOP (15580) Manual Land SYNOP (9035) METAR (1999) VALUE AND LODGE 100 10 1000 1 SHIP (715) Abbroulated SHEP (1975) GloSATp02 temporal coverage (1658-2020) by latitudinal band BUFR LAND SYNOP (33249) CRUTEM4.6.0.0-2019-12 percentage coverage (1850-2019) 8000 7000 ₿ 6000 £ 5000 t 4000 gu 000E prt global (90°S,60°S) n(stations)=72 Ö 2000 (59°S,30°S) n(stations)=618 (29°S,00°S) n(stations)=1210 (01°N,30°N) n(stations)=1736 1000 (31°N,60°N) n(stations)=5846 2020 NH coverage ~ 40% 10:32 AM · Sep 29, 2020 · Twitter Web App (61°N,90°N) n(stations)=1200 1700 1750 1800 1850 1900 1950 2000 2020 SH coverage ~ 20% Year







## A GloSAT story: Part II (land obs get us part of the way there)







# **Observations and coverage:** The actual picture is more like this











NB: cosine-weighted average of grid cells which are bigger in the tropics

2000

2020



## **WMO Data Conference 16-19 Nov 2020:** GloSAT and data rescue efforts will help a lot





Models (and satellite EO) rely on observations (and vice-versa)

Big issue in some of the poorest countries in Africa and large sea areas in the southern hemisphere like the Indian Ocean and in the Pacific where there are large areas with few or no observations

Big issue in regions affected by extreme weather

Big issue in regions where climate tipping points need careful monitoring





### **Blending of temperatures**

Kevin, Ed & colleagues have shown how to compare obs with climate models by calculating global mean surface temperature (GMST) anomalies (from 1961-1990) by blending air temp over land (TAS) with SST over oceans (TOS) for various sea-ice concentration (SIC) and land/sea mask (SFTOF) configurations



### **@AGU**PUBLICATIONS



#### **Geophysical Research Letters**

RESEARCH LETTER 10.1002/2015GL064888

#### Robust comparison of climate models with observations using blended land air and ocean sea surface temperatures

#### Key Points:

 There is a systematic bias in model-observation comparisons from blending air and sea temperatures A further bias arises from using anomalies in regions where the sea ice boundary has changed Correcting these accounts for a quarter to half of the discrepancy between models and observations

#### Supporting Information:

Supporting Information S1

#### Correspondence to: K. Cowtan.

kevin.cowtan@york.ac.uk

#### Citation: Cowtan, K., Z. Hausfather, E. Hawkins

P. Jacobs, M. E. Mann, S. K. Miller, B. A. Steinman, M. B. Stolpe, and R. G. Way (2015), Robust comparison of climate models with observations using blended land air and ocean sea surface temperatures, *Geophys. Res. Lett.*, 42, 6526–6534, doi:10.1002/2015GL064888.

#### Kevin Cowtan<sup>1</sup>, Zeke Hausfather<sup>2</sup>, Ed Hawkins<sup>3</sup>, Peter Jacobs<sup>4</sup>, Michael E. Mann<sup>5</sup>, Sonya K. Miller<sup>5</sup>, Byron A. Steinman<sup>6</sup>, Martin B. Stolpe<sup>7</sup>, and Robert G. Way<sup>8</sup>

<sup>1</sup>Department of Chemistry, University of York, York, UK, <sup>2</sup>Energy and Resources Group, University of California, Berkeley, California, USA, <sup>3</sup>National Centre for Atmospheric Science, Department of Meteorology, University of Reading, Reading, UK, <sup>4</sup>Department of Environmental Science and Policy, George Mason University, Fairfax, Virginia, USA, <sup>5</sup>Department of Meteorology and Earth and Environmental Systems Institute, Pennsylvania State University, University of Minnesota, Duluth, USA, <sup>6</sup>Department of Earth and Environmental Sciences, Large Lakes Observatory, University of Minnesota, Duluth, Duluth, Minnesota, USA, <sup>7</sup>Institute for Atmospheric and Climate Science, ETH Zurich, Zurich, Switzerland, <sup>8</sup>Department of Geography, University of Ottawa, Ottawa, Ontario, Canada

Abstract The level of agreement between climate model simulations and observed surface temperature change is a topic of scientific and policy concern. While the Earth system continues to accumulate energy due to anthropogenic and other radiative forcings, estimates of recent surface temperature evolution fall at the lower end of climate model projections. Global mean temperatures from climate model simulations are typically calculated using surface air temperatures, while the corresponding observations are based on a blend of air and sea surface temperatures. This work quantifies a systematic bias in model-observation comparisons arising from differential warming rates between sea surface temperatures and surface air temperatures over oceans. A further bias arises from the treatment of temperatures in regions where the sea ice boundary has changed. Applying the methodology of the HadCRUT4 record to climate model temperature fields accounts for 38% of the discrepancy in trend between models and observations over the period 1975–2014.

#### HadCRUT calculation:

No sea-ice: TOS(SIC>0.05)=NaN Trim to land: TAS(SFTOF>0.99)=NaN Downscaled to 5x5 grids NB: may also mask by incomplete geographical coverage of observations ( some datasets interpolate to fill in the gaps )

Air temp over sea-ice is usually estimated from obs over land





## **Blending using reanalysis**

Q. How well do blended (and masked) global mean surface temperatures (GMST) from reanalysis agree with osbervations of global surface air temp (GSAT) ?Q. What is the effect of masking by observational coverage ?Q. What is the effect of variable sea-ice coverage ?

<u>Reanalysis datasets</u>: **ERA5** (incl. BE) 1950-2020 & **JRA-55** 1958-2019 re-gridded at 5x5 and 1x1 using CDO and NCO <u>Temperature datasets</u>: HadCRUT.4.6.0.0, CRUTEM.4.6.0.0 and HadSST.3.1.1.0 from 1850-2020

Scenarios:

1) Reanalysis with **land/sea mask** and variable **sea-ice conc** 

2) HadCRUT with **separate masks** from CRUTEM and HadSST

Stats: bootstrap calculation of CI on the GSAT / GMST ratio





### **Reanalysis inputs:** Land/sea masks and sea-ice boundary changes







### **Analysis:** ERA5 (blended-air) scenario effects < 0.01K



xxf - unmasked, blended abs temp, fixed ice

mxx – HadCRUT-masked, blended abs temp, variable ice mxf - HadCRUT-masked, blended abs temp, fixed ice

xax - unmasked, blended anomalies, variable ice
 xaf – unmasked, blended anomalies, fixed ice
 max - HadCRUT-masked, blended anomalies, variable ice
 maf - HadCRUT-masked, blended anomalies, fixed ice

Similar overall trends Sensitivity to variable ice Crossover ~ middle of 1961-1990 baseline





# **Analysis:** JRA-55 (blended-air) scenario effects < 0.01K



Similar overall trends (<u>but different to ERA5</u>) Sensitivity to variable ice (arguably less than ERA5) Crossover ~ middle of baseline





# **Analysis:** The HadCRUT calculation (separate land & sea masks)



**GSAT** has a large discrepancy from the **GMST** obtained when the reanalysis is blended using a separate land mask from CRUTEM and sea mask from HadSST and rises to ~0.1K by 2020







# **Analysis:** Sea-ice resolves the HadCRUT calculation discrepancy

#### Discrepancy is barely visible



HadCRUT calculation: **no sea-ice** 

maf - HadCRUT-masked, blended anomalies, fixed ice

GSAT discrepancy from GMST obtained when the reanalysis is blended the HadCRUT mask and known sea-ice seems to resolve this problem





## **Analysis:** General GSAT/GMST blended cases – sliding trend ratios



Although the anomaly discrepancy is ~ 0.1K, the impact on the GSAT / GMST is large e.g. a ratio of 1.7 translates to 70% increase over the blended anomaly value Q. Does repeating this analysis using LOESS fits (30-yr span) change things ?





### **Analysis:** General GSAT/GMST blended cases – bootstrap CI on ratio



We then bootstrap to find the likelihood of the 5-95% bounds on the confidence interval (CI) for the ratio We do this for all cases (including the HadCRUT calculation)





### **Analysis:** Blended ratio bounds~[0.99,1.01] (HadCRUT case ...)



ERA5 (incl. BE): 5°x5° grid: bootstrap estimates based on sliding 30 year trend ratios

University of East Anglia



This study reiterates the value of the **blending approach for robust comparisons**:

A. Blended (and masked) temperatures from reanalysis <u>agree well with 'true' GMST</u> <u>observations</u> (some minor difference between ERA5 & JRA-55)
A. Using separate land/sea coverage masks <u>and no sea-ice</u> has the strongest impact ~ 0.1K (the HadCRUT calculation)
A. Variable sea-ice and HadCRUT mask effect ~ 0.01K

**Robust statistics** can be obtained from sliding trends and **bootstrapping** to determine the 5-95% likelihood of **CI on the GSAT / GMST ratio:** [0.99,1.01] <u>depending on reanalysis</u>





## A GloSAT story: Part III (nice collaborations and rescued obs)

#### Daily Record, 1 July 1776-10 January 1777

### Massachusetts Historical Society: Coolidge Collection of Thomas Jefferson Manuscripts, 1705-1827, Other Volumes, 1766-1824, Memorandum Book, 1776-1778.

5 pp., bound at the end of a volume with Timothy Telescope [Robert Aitken], The Philadelphia Newest Almanack, for the Year of Our Lord 1776 (Philadelphia, 1776), 70 pp. of Jefferson's cash accounts, and 2 pp. of miscellaneous memoranda.

Printed in Memorandum Books, 1:432-5; see also 1:xxxix.

At head of first page: "Observations on the weather / Philadelphia 1776." Headings of first column on first page: "July. | hour. | thermom."; and second column: "day | h - m | °."

At 17 July 1776, Jefferson put an asterisk in front of his evening observation and put a note at the foot of the page: "this mark \* denotes there had been rain previous." He employed the asterisk in other entries in this record up to 26 August 1776.



#### **Rescuing land air temp observations**

#### Martin Stendel @MartinStendel

In the Danish National Archives, @Rigsarkivet, 3 centuries of Danish ship logbooks are kept, offering an invaluable wealth of information about temperature, pressure, wind and weather. @isterbandit, @ed\_hawkins and I are building a consortium to digitize ~3 million pages. @dmidk



#### **Rescuing marine air temp observations**





### Many thanks

#### **Kevin Cowtan**

For providing python blending code and guidance: <u>https://www-users.york.ac.uk/~kdc3/papers/robust2015/methods.html</u> **NCAR/UCAR & JMA** For providing JRA-55 reanalysis data 1958-2019 and NCO: <u>https://rda.ucar.edu/datasets/ds628.1</u> **C3S CDS / ECMWF** For providing ERA5 (incl. BE) reanalysis data 1950-2020:

https://cds.climate.copernicus.eu/#!/search?text=ERA5

#### UKMO / Hadley Centre

For providing HadCRUT.4.6.0.0 and HadSST.3.1.0.0 anomaly data 1850-2020:

https://www.metoffice.gov.uk/hadobs

#### CRU / UEA

For guidance and inspirations and huge effort providing CRUTEM4.6.0.0 anomaly data 1850-2020: <u>https://crudata.uea.ac.uk/cru/data/temperature/</u>

#### GloSAT / NOC Team & JASMIN / CEDA

For super project support + GWS access <a href="https://www.glosat.org/">https://www.glosat.org/</a>



