A large discontinuity in the mid-twentieth century in observed global-mean surface temperature

David W. J. Thompson¹, John J. Kennedy², John M. Wallace³ & Phil D. Jones⁴

Data sets used to monitor the Earth’s climate indicate that the surface of the Earth warmed from ~1910 to 1940, cooled slightly from ~1940 to 1970, and then warmed markedly from ~1970 onward¹. The weak cooling apparent in the middle part of the century has been interpreted in the context of a variety of physical factors, such as atmosphere-ocean interactions and anthropogenic emissions of sulphate aerosols². Here we call attention to a previously overlooked discontinuity in the record at 1945, which is a prominent feature of the cooling trend in the mid-twentieth century. The discontinuity is evident in published versions of the global-mean temperature time series³, but stands out more clearly after the data are filtered for the effects of internal climate variability. We argue that the abrupt temperature drop of ~0.5 °C in 1945 is the apparent result of uncorrected instrumental biases in the sea surface temperature record. Corrections for the discontinuity are expected to alter the character of mid-twentieth century temperature variability but not estimates of the century-long trend in global-mean temperatures.

The time series of global-mean surface temperatures (T_G) reflects the influence of both anthropogenically induced and internally driven climate variability. A common method used to identify the signature of anthropogenic emissions in T_G is to compare the observed temperature variability with the output of climate change simulations. An alternative approach used to isolate the anthropogenic signal in T_G is to filter out the variance associated with internally driven climate phenomena and then interpret the residual time series. Here we exploit the latter technique to filter out the variance in T_G associated with two prominent internally driven climate phenomena: (1) the El Niño/Southern Oscillation (ENSO) and (2) temperature advection over the high latitudes of the Northern Hemisphere during winter.

The data and details of the filtering method are described in Methods. Briefly, the signal of ENSO in T_G is modelled as the damped thermodynamic response of the tropics to anomalous sea-air fluxes of sensible and latent heat in the dynamically active cold-tongue region of the eastern tropical Pacific³. The resulting ENSO index time series (T\_ENSO, Fig. 1) is analogous to a lagged and low-pass-filtered time series of sea surface temperatures (SSTs) averaged over the eastern tropical Pacific³ and provides a substantially better rendition of the ENSO signal in globally averaged temperatures than the more commonly used cold-tongue index (Table 1).

The signal of temperature advection over the high latitudes of the Northern Hemisphere is manifested in the so-called “cold oceans-warm land” (COWL) pattern⁴. During months with abnormally strong westerly winds at the surface, there is enhanced advection of relatively warm marine air masses over the colder continents and cold continental air masses over the warmer oceans. The continents have a lower heat capacity than the oceans, and the warming of the land exceeds the cooling of the ocean, and the global-mean temperature for that month is anomalously high. Months with abnormally weak surface westerlies are marked by global-mean temperature anomalies in the opposite sense. The calculation of the COWL pattern and its associated index time series is described in Methods. The COWL index time series (T\_COWL) accounts for a substantial amount of the month-to-month weather-related “noise” in T_G but also has weak secular variability due in part to trends in the atmospheric circulation (Fig. 1).

The influences of ENSO and the COWL pattern on surface temperatures were removed by subtracting the linearly fitted T\_ENSO and T\_COWL index time series from T_G (Methods). The resulting residual global-mean temperature time series (T\_G\_residual) is shown at the bottom of Figs 1 and 2. Filtering out ENSO and the COWL pattern reduces substantially the amount of interannual and month-to-month variance in T_G without reducing its temporal resolution. Consequently, the residual time series provides a cleaner rendition of the interdecadal variability in the time series of twentieth-century global-mean temperatures while retaining and increasing the prominence of numerous discrete drops in it. Most of the more prominent drops that are apparent in T\_G\_residual coincide with large tropical

---

¹ Department of Atmospheric Science, Colorado State University, Fort Collins, Colorado 80523, USA. ²Met Office Hadley Centre, Exeter EX1 3PB, UK. ³Department of Atmospheric Sciences, University of Washington, Seattle, Washington 98195, USA. ⁴Climatic Research Unit, School of Environmental Sciences, University of East Anglia, Norwich NR4 7TJ, UK.
volcanic eruptions (Fig. 2; solid vertical lines). However, the most pronounced drop occurs in late 1945 and is not associated with any known climate phenomenon. The signal of the volcanic eruptions in the residual data is discussed elsewhere (D.W.J.T., et al., manuscript in preparation); here we focus on the discontinuity in late 1945.

The discontinuity in global-mean surface temperatures in late 1945 is evident in the unfiltered global-mean time series, but its prominence and unique character are highlighted by the removal of the ENSO and COWL-related variability (Fig. 2). The amplitude of the step is substantial: temperatures dropped by ~4.3°C during the ~6-month period following August 1945, thus the amplitude of the drop is roughly 40% as large as the ~0.75°C rise in T°C from 1900 to 2006, and is larger than the drop in 1991 following the June 1991 eruption of Mt. Pinatubo (Fig. 2 bottom time series). Clearly, the step has a profound impact on the historical record of twentieth-century surface temperatures.

The step in late 1945 does not appear to be related to any known physical phenomenon. No substantial volcanic eruptions were reported at the time, and the nuclear explosions over Hiroshima and Nagasaki are estimated to have had little effect on global-mean temperatures (~1°C rise). Instead, the step appears to be a result of the change in oceanic and atmospheric conditions associated with the end of World War II and the subsequent onset of the Cold War.

References therein). Different measurement techniques are prone to different physical biases. For example, bucket measurements are affected by the exchange of sensible, latent and radiant heat with the surrounding environment, and are often biased too cool relative to the actual temperature. Engine room intake measurements are influenced by the proximity of the ship's engine and the depth of the intake water, and are most often biased warm15. The effect of changes in instrumentation on SST measurements can be corrected only with knowledge of both the physical biases and a time history of the way in which measurements were taken.

The results shown here are based on the current version of the UK Met Office Hadley Centre SST data set (HadSST3 ref. 9). Like all historical SST products, the HadSST3 data are derived from the database of raw (i.e., uncorrected) marine observations provided by the International Comprehensive Ocean-Atmosphere Data Set (ICOADS ref. 10) or one of its predecessors. The SST data in the ICOADS archive are affected by numerous changes in instrumentation. The HadSST3 data have been corrected for the widespread use of uninsulated buckets in the nineteenth and early twentieth century, and for the large change in instrumentation which occurred when US Merchant Marine ships switched from uninsulated bucket measurements to engine room intake measurements between 1939 and the end of 1941 (refs 6, 9). The corrections were made by adjusting the data before 1941 so that they are compatible with the data obtained from the mix of measurements during the 1961–90 period, and they

![Figure 3](image-url)
result in the large differences between the ICOADS and HadSST2 data before 1941 (Fig. 4). However, these are the only substantial corrections that have been applied to the HadSST2 data, in part because important information about the mix of measurements for the post-Second World War era has only recently been digitized\(^\text{11}\). From the country-of-origin information in Fig. 4, it is clear that the SST archive—and hence the mix of measurement methods—continued to evolve considerably during the decades following 1941.

The most notable change in the SST archive following December 1941 occurred in August 1945. Between January 1942 and August 1945, ~80% of the observations are from ships of US origin and ~5% are from ships of UK origin; between late 1945 and 1949 only ~30% of the observations are of US origin and about 50% are of UK origin. The change in country of origin in August 1945 is important for two reasons: first, in August 1945 US ships relied mainly on engine room intake measurements whereas UK ships used primarily uninsulated bucket measurements\(^\text{12}\), and second, engine room intake measurements are generally biased warm relative to uninsulated bucket measurements\(^\text{12}\).

Hence, the sudden drop in SSTs in late 1945 is consistent with the rapid but uncorrected change from engine room intake measurements (US ships) to uninsulated bucket measurements (UK ships) at the end of the Second World War. As the drop derives from the composition of the ICOADS data set, it is present in all records of twentieth-century climate variability that include SST data.

The Met Office Hadley Centre is currently assessing the adjustments required to compensate for the step in 1945 and subsequent changes in the SST observing network. The adjustments immediately after 1945 are expected to be as large as those made to the pre-war data (~0.3°C; Fig. 4), and smaller adjustments are likely to be required in SSTs through at least the mid-1960s, by which time the observing fleet was relatively diverse and less susceptible to changes in the data supply from a single country of origin\(^\text{9}\). The new adjustments are likely to have a substantial impact on the historical record of global-mean surface temperatures through the middle part of the twentieth century. The adjustments are unlikely to significantly affect estimates of century-long trends in global-mean temperatures, as the data before ~1940 and after the mid-1960s are not expected to require further corrections for changes from uninsulated bucket to engine room intake measurements. However, compensation for a different potential source of bias in SST data in the past decade—the transition from ship- to buoy-derived SSTs—might increase the century-long trends by raising recent SSTs as much as ~0.1°C as buoy-derived SSTs are biased cold relative to ship measurements\(^\text{10}\).

METHODS SUMMARY

The data used in this study are the land-surface (CRUTEM3), sea-surface (HadSST2) and combined land- and sea-surface temperature (HadCRUT3) data obtained from the Climatic Research Unit (http://www.cru.uea.ac.uk) and described in refs 9 and 13. The methodologies used to form the ENSO and COWL index time series, and to filter out the ENSO and COWL index time series from the global-mean time series, are described in the Methods section.

Full Methods and any associated references are available in the online version of the paper at www.nature.com/nature.

Received 28 January; accepted 4 April 2008.


---

**Figure 4** | The HadSST2, ICOADS and country-of-origin time series. a, Top, the ENSO residual global-mean SST time series reproduced from Fig. 3. Middle, as in the top time series but for data from ICOADS. Bottom, the percentage of observations which can be positively identified as coming from US (blue) and UK (red) ships. The vertical line denotes December 1941. Data to the left of the vertical line were corrected in the Met Office Hadley Centre data\(^\text{9}\); data to the right of the vertical line have not been adjusted in the HadSST2 data. b, As a but focused on the period 1940–60. Left vertical axis shows temperature anomalies (°C); right vertical axis shows percentage of observations. Horizontal lines in the temperature time series denote the mean for 1961–90.

Acknowledgements: We thank D. Parker and S. Solomon for comments on the manuscript, and R. Reynolds for reviews. D.W.J.T. and J.M.W. were supported by the NSF Climate Dynamics Program under budget numbers ATM-0132190 and ATM-0263082 (D.W.J.T.) and ATM-0318675 (J.M.W.). J.M.W. was supported by the Joint Defra and MoD Programmes GA0101 (Defra) and CD/D/38/0317 Annex C5 (MoD). P.D.J. was supported by the US Department of Energy (DE-FG02-98ER62601).

Author Information: Reprints and permissions information is available at www.nature.com/reprints. Correspondence and requests for materials should be addressed to D.W.J.T. (dave@atmos.colostate.edu).
METHODS

All results and analyses are based on monthly-mean data. Following ref. 3, the tropical-mean surface temperature response to ENSO is modelled as

\[
C \frac{dT_{\text{ENSO}}(t)}{dt} = R(t) - \beta T_{\text{ENSO}}(t)
\]

(1)

where \(T_{\text{ENSO}}(t)\) denotes the simulated response of monthly-mean tropical-mean surface temperature anomalies to ENSO variability; \(R(t)\) is the anomalous flux of sensible and latent heat in the eastern tropical Pacific, parameterized as the average SST anomaly in the dynamically active cold-tongue region (5°N–5°S, 160°W–90°W) multiplied by (1) the fractional area of the tropics covered by the cold-tongue region (~5%), and (2) a coupling coefficient of 25 W m\(^{-2}\) K\(^{-1}\); \(\beta\) is a linear damping coefficient found by linearizing the Stefan-Boltzmann law, \(\beta = 4\sigma T_{\text{B}}^4\), where \(\sigma\) is the Stefan-Boltzmann constant and the mean temperature of the tropical atmosphere. \(T_{\text{B}}\) is assumed to be ~255 K. \(C\) is the heat capacity of the tropics per unit area and is determined empirically such that the correlation between \(T_{\text{ENSO}}(t)\) and tropical-mean surface temperature anomalies is maximized (the resulting heat capacity corresponds to the entire atmosphere and the top ~8 m of the tropical ocean; \(C = 5.2 \times 10^{10} \text{ J m}^{-2} \text{ K}^{-1}\)). The model was initialized with anomalies in the cold-tongue region starting at 1870 and the output \(T_{\text{ENSO}}\) was used from 1875 to 1995.

Following ref. 4, the COWL pattern loadings are found as

\[
COWL(x) = \frac{\langle T(x,t) - T_{\text{NH}}(t) \rangle T_{\text{NH}}(t)}{ \langle T(x,t) - T_{\text{NH}}(t) \rangle^2}
\]

(2)

where \(T(x,t)\) are Northern Hemisphere (30°N–90°N) temperature anomalies given as a function of space, \(x\), and time, \(t\); \(T_{\text{NH}}(t)\) denotes the time series of Northern Hemisphere (30°N–90°N) mean surface temperature anomalies, and the overbar denotes the time mean. The time series of the COWL pattern \(COWL(x)\) is found by projecting \(\langle T(x,t) - T_{\text{NH}}(t) \rangle\) onto \(COWL(x)\) at each time step.

The \(T_{\text{ENSO}}\) and \(COWL\) index time series were fitted to global-mean temperatures as

\[
x_{\text{ENSO}}(t) = \frac{x(t) - T_{\text{ENSO}}(t)}{x(t)}
\]

where \(x(t)\) corresponds to the monthly-mean \(T_{\text{ENSO}}\) or \(COWL\) index time series; the fractional term on the right-hand side denotes the regression of \(T(x,t)\) onto \(x(t)\); and \(x_{\text{ENSO}}\) corresponds to the component of \(T(x,t)\) that is linearly congruent with variations in \(T_{\text{ENSO}}\) or \(COWL\). The regression coefficients are based on detrended values of the data so that shared trends in the time series do not contribute to the linear fits (the data are detrended for the 1950–2006 period), and are calculated for the period 1950–2006 when the data coverage is best (the results are not sensitive to reasonable changes in the period of analyses and are virtually unchanged if the data are not detrended). The regressions were done separately for \(T_{\text{ENSO}}\) and \(COWL\) as the time series are effectively uncorrelated; virtually identical results were obtained for multiple regression. In Fig. 1, the fit to the \(T_{\text{ENSO}}\) and \(COWL\) index time series was computed for global-mean combined ocean and land temperature data in Figs 3 and 4, the fits were computed for global-mean land and ocean temperature data separately.