

EXPLAINING EXTREME EVENTS OF 2014 FROM A CLIMATE PERSPECTIVE

Editors

Stephanie C. Herring, Martin P. Hoerling, James P. Kossin, Thomas C. Peterson, and Peter A. Stott

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ABSTRACT—Stephanie C. Herring, Martin P. Hoerling, James P. Kossin, Thomas C. Peterson, and Peter A. Stott

Understanding how long-term global change affects the intensity and likelihood of extreme weather events is a frontier science challenge. This fourth edition of explaining extreme events of the previous year (2014) from a climate perspective is the most extensive yet with 33 different research groups exploring the causes of 29 different events that occurred in 2014. A number of this year's studies indicate that human-caused climate change greatly increased the likelihood and intensity for extreme heat waves in 2014 over various regions. For other types of extreme events, such as droughts, heavy rains, and winter storms, a climate change influence was found in some instances and not in others. This year's report also included many different types of extreme events. The tropical cyclones that impacted Hawaii were made more likely due to human-caused climate change. Climate change also decreased the Antarctic sea ice extent in 2014 and increased the strength and likelihood of high sea surface temperatures in both the Atlantic and Pacific Oceans. For western U.S. wildfires, no link to the individual events in 2014 could be detected, but the overall probability of western U.S. wildfires has increased due to human impacts on the climate.

Challenges that attribution assessments face include the often limited observational record and inability of models to reproduce some extreme events well. In general, when attribution assessments fail to find anthropogenic signals this alone does not prove anthropogenic climate change did not influence the event. The failure to find a human fingerprint could be due to insufficient data or poor models and not the absence of anthropogenic effects.

This year researchers also considered other humancaused drivers of extreme events beyond the usual radiative drivers. For example, flooding in the Canadian prairies was found to be more likely because of human land-use changes that affect drainage mechanisms. Similarly, the Jakarta floods may have been compounded by land-use change via urban development and associated land subsidence. These types of mechanical factors reemphasize the various pathways beyond climate change by which human activity can increase regional risk of extreme events.

9. CAUSAL INFLUENCE OF ANTHROPOGENIC FORCINGS ON THE ARGENTINIAN HEAT WAVE OF DECEMBER 2013

A. HANNART, C. VERA, F. E. L. OTTO, AND B. CERNE

The Argentinian heat wave of December 2013 was likely caused in part by anthropogenic forcings. These forcings have increased the risk of such an event occurring by a factor of five.

Introduction. A heat wave occurred 13-31 December 2013 in the Northern and central area of Argentina as well as in Northern Patagonia (Fig. 9.1a). Since the beginning of the temperature record by the Argentine National Meteorological Service (SMN), the year 2013 had the hottest month of December ever recorded over the impacted area (+2.5°C anomaly with respect to 1961-90). In the Greater Buenos Aires area, which is the second largest urban area in South America, the event stands out as the single longest heat wave that ever occurred (18 days) over the observational period. This event had significant impacts in particular on the health and energy sectors.

The analysis of the atmospheric dynamics shows that the high temperatures were primarily associated with an intensification of the South Atlantic Convergence Zone (SACZ), which jointly caused simultaneous extreme rainfall events in Southeastern Brazil (Fig. 9.1b). While particularly apparent in the case of the December 2013 heat wave, the influence of the SACZ on heat waves over this subtropical region is a mechanism that has been previously well described (Cerne et al. 2007; Cerne and Vera 2011) and can be summarized as follows: (i) intensified SACZ promotes subsidence over Argentina, favoring clear sky conditions and increasing incoming solar radiation; (ii) which in turn generates high temperatures and lack of rainfall (Fig. 9.1a,b). Furthermore, this mechanism is reinforced in general, and in particular during December 2013, by the presence of an anticyclonic circulation anomaly over central Argentina, which in turn is modulated by Rossby wave trains extended along the South Pacific (Fig. 9.1c), likely induced in

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part by increased tropical convection over Northern Australia and the Maritime continent (not shown).

From a long-term climate perspective, previous studies (Rusticucci 2012) documented significant trends in some temperature features in the region: while minimum temperatures clearly exhibit positive trends—particularly in Central and Eastern Argentina during the second part of the twentieth century and first decade of the twenty-first century-maximum temperatures, on the contrary, show significant negative trends. On the other hand, there is only medium confidence in the increase of warm extremes of minimum temperature in this region, and trends in maximum temperature extremes are characterized by high spatial variability. Few studies analyzed the trends related to heat waves or warm spells, which appear to have increased in some areas and decreased in others (Seneviratne et al. 2012; Perkins et al. 2012; Rusticucci et al. 2015).

While the causes of the event associated with the short-term dynamics of the atmosphere are well understood, it is not clear at present whether or not the long-term climate response to anthropogenic forcing can also be held to have causally contributed to the occurrence of the event. The present work addresses the latter causal aspects.

Data and method. A heat wave is defined to occur when the index of surface temperature averaged over the relevant area (23°-45°S, 75°-55°W; see Fig. 9.1a) and over the month of December (Z hereafter) reaches or exceeds a threshold which will be discussed further. Observed values of Z over the instrumental period were calculated from the South American gridded dataset (SAG hereafter; Tencer et al. 2011). Since this dataset only covers the period 1960–2000, the periods 1901–60 and 2001–14 were obtained from HadCRUT4 (Morice et al. 2012) anomalies added to SAG climatology. The choice to use climatology from SAG is justified by the fact that it has better spatial

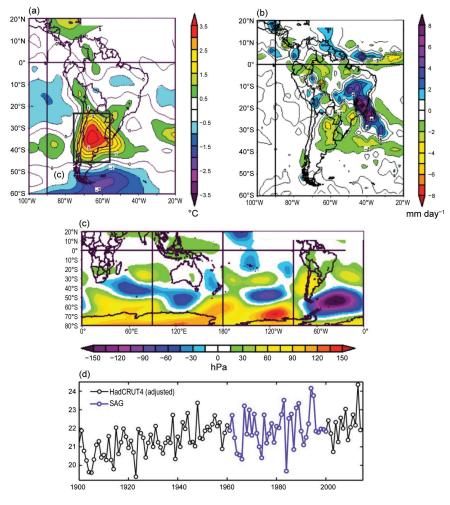


Fig. 9.1. Observations. (a,b,c) Atmospheric dynamics associated with the event: monthly anomalies with respect to 1981–2010 climatological average for Dec 2013 over South America (a,b) and Southern Hemisphere (c) obtained from NCEP-NCAR reanalysis dataset and plotted using the website esrl.noaa.gov/psd (NOAA/ESRL Physical Sciences Division, Boulder, CO, USA). (a) Surface temperature (°C). (b) Precipitation (mm day-1). (c) Sea level pressure (hPa). (d) Time series of the temperature index Z (space-time average over the domain 23°–45°S, 75°–55°W, month of Dec) over the instrumental period obtained from SAG and HadCRUT4.

coverage and underwent a more detailed quality control than HadCRUT4.

Following standard practices for causal attribution (Allen 2003; Hannart et al. 2015), we analyzed whether and to what extent anthropogenic climate change changed the odds of a heat wave in central Argentina in December. We thus compared P1, the probability of occurrence of a heat wave under the observed climate conditions of 2013 (referred to as factual conditions) to P0, the same probability in the world as it might have been without anthropogenic climate change (referred to as counterfactual). For this purpose, we followed the method of Schaller et al. (2014) and

we used the distributed computing framework weather@home to run the Met Office Hadley Centre atmosphere-only general circulation model HADAM3P (Massey et al. 2015), which has been shown to adequately represent some key features of atmospheric circulation over South America (Chou et al. 2012). We generated two ensembles: (i) observed climate conditions of 2013, forced with observed aerosols and greenhouse gas composition as well as SST and sea ice fraction values from 2013 obtained from the Operational Sea Surface Temperature and Sea Ice Analysis (OSTIA) dataset (Stark et al. 2007); (ii) counterfactual conditions, forced with preindustrial atmospheric gas composition, combined with the sea ice extent that corresponds to the year of maximum sea ice extent in each hemisphere of the OSTIA record, and with 11 different estimates of sea surface temperatures (SSTs) without human influence. These SSTs are obtained by subtracting

11 estimates of the human influence on SST from the 2013 OSTIA SST values. These 11 SST anomaly patterns are obtained by calculating the difference between nonindustrial and present-day simulations for available Coupled Model Intercomparison Project Phase 5 (CMIP5) models.

Two PDFs of the temperature index Z were estimated from these two ensembles, under Gaussian assumption. The so-called fraction of attributable risk (FAR = 1 - P0 / P1) was then derived; Hannart et al. (2015) have shown that in causal theory (Pearl 2000), the FAR may also be interpreted as the probability of necessary causation (PN) associated with

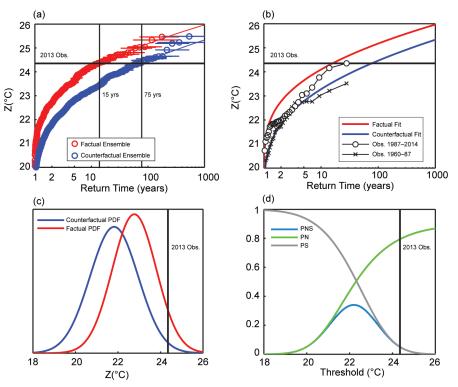
the causal link between the forcing and the event. A prominent feature of causal theory indeed consists of recognizing that causation corresponds to rather different situations and that three distinct facets of causality should be distinguished: (i) necessary causation, where the occurrence of the event requires that of the forcing but may also require other factors; (ii) sufficient causation, where the occurrence of the forcing drives that of the event but may not be required for the event to occur; (iii) necessary and sufficient causation, where (i) and (ii) both hold. The probability of necessary causation (PN) thus corresponds to only one of the three facets of causality, while the probability of sufficient causation PS = 1 - (1 - P1)/(1 - P0) is

its second facet, and the probability of necessary and sufficient causation PNS = P1 - P0 summarizes both. With these definitions, the choice of \widehat{o} 23 the threshold on Z that defines occurrence has critical implications on the amount and nature of causal evidence (Fig. 9.2d). In the context of extreme event attribution, many different factors are usually necessary to trigger the occurrence of a rare event and, conversely, no single factor will ever hold as a sufficient explanation thereof: maximizing PN at the expense of PS is thus arguably a relevant approach Here, this implies choosing the highest possible threshold for the index, that is 24.4°C (Fig. 9.2d).

time series over the period 1900-2014 is shown in Fig. 9.1d. The climatological mean calculated over the period 1961–90 is 21.9°C with anomalies

year 2013, which is therefore the record high year over the instrumental period.

The factual and counterfactual PDFs are shown in Figs. 9.2a-c. They differ significantly, and their difference can be described quite straightforwardly: it merely consists of a ~1°C gap in their mean, while their dispersion, shape, and tails are roughly unchanged. Thereby, the intensity level of a heat wave is higher in the factual world than it would be in the counterfactual one for any return period, and the level is increased by 1°C no matter the return period. Figs. 9.2a,b emphasize this finding: one can see that the return level curves match well with each other, up to a 1°C translation. On the other hand, the tail behavior



in the present context. Fig. 9.2. Simulations. (a,b,c) Comparison of the factual vs counterfactual PDFs of the temperature index Z. (a) Return level curves simulated by HADAM3P: factual values (red dots) and counterfactual values (blue dots), 95% confidence interval (thin horizontal bars), Gaussian fit (red and blue lines), observed Dec 2013 value (thick horizontal black line), return periods (thin vertical black lines). (b) Same as (a) with superimposed observations over the 1960-87 (black line with crosses) and 1987-2014 (black line with circles) periods. (c) Factual and counterfactual PDFs Results. The observed of Z (Gaussian fit), observed Dec 2013 value (thick vertical black line). (d) PN, PS, and PNS as a function of the threshold used for event definition. Adopting a less restrictive definition (i.e., smaller threshold) decreases the level of necessity but increases that of sufficiency—which of these matters most depends on the purpose of the causal question (Hannart et al., 2015). A possible "multipurpose" approach is to balance both quantities, but this leads here to a substantially lower threshold (u = 22.3°C; 1.5 and 3.5 year return periods) which no longer reflects the extreme nature of the event and yields a well-balanced but not very stringent ranging from -2.5°C in level of causal evidence. Maximizing PN at the expense of very low PS is arguably year 1923 to $+2.5^{\circ}$ C in more relevant here, yielding PN = 0.8 and PS = 0.07.

of the two PDFs appears to be accurately described by a Gaussian distribution. The quality of the Gaussian fit is apparent in Fig. 9.2a where the empirical and the theoretical Gaussian return level curves are almost indistinguishable.

Figure 9.2b shows that simulated PDFs represent reasonably well the distribution of observed values without requiring any bias correction. Indeed, the return level curve obtained from observations over the period 1987–2014 (respectively 1960–87) appears to match decently well with the simulated return level curve in the factual (respectively counterfactual) world.

Discussion and conclusion. The value of the index reached in 2013 corresponds to a return period of 15 years in the factual world and a return period of 75 years in the counterfactual one (Fig. 9.2a): exceeding the intensity of the December 2013 heat wave thus appears to be five times more likely in the world as it was in 2013, than it was in the same world with no anthropogenic forcings.

This ratio corresponds to FAR = 80%, and it is consequently tempting to claim that 80% of the risk of the December 2013 heat wave is attributable to anthropogenic forcings. But this statement may be considered as a somewhat misleading interpretation. Indeed, the FARs associated to other causes, whether natural or anthropogenic, could be high as well. In any case, the FARs associated to the many causes of an extreme event are never bound to sum up to one, therefore the FAR can not be interpreted as a "share" of causality as the above statement suggests. It could thus be more appropriate to rename the acronym FAR into "fraction of additional risk."

Finally, the probabilistic definitions of causality recalled above offer yet a different causal interpretation and formulation. In the present case, PN=0.8 and PS=0.07 yields the statement that anthropogenic forcings were likely a necessary cause of the December 2013 heat wave, and yet were very likely not a sufficient cause thereof—a statement which clearly reflects the fact that other causal factors were involved in this event, and which may be shortened into "The Argentinian heat wave of December 2013 was likely caused in part by anthropogenic forcings."

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Table 34.1. ANTHROPOGENIC INFLUENCE

ON EVENT STRENGTH † **INCREASE DECREASE NOT FOUND OR UNCERTAIN** Australia (Ch. 31) Australia, Adelaide & Melbourne (Ch. 29) Heat Europe (Ch.13) Australia, Brisbane (Ch.28) S. Korea (Ch. 19) Cold Upper Midwest (Ch.3) Eastern U.S. (Ch. 4) Winter Storms and N. America (Ch. 6) Snow N. Atlantic (Ch. 7) Jakarta**** (Ch. 26) Heavy Canada** (Ch. 5) United Kingdom*** (Ch. 10) **Precipitation** New Zealand (Ch. 27) Middle East and S.W. Asia E. Africa (Ch. 16) (Ch. 15) E. Africa* (Ch. 17) Drought N.E. Asia (Ch. 21) S. Levant (Ch. 14) Singapore (Ch. 25) Gonzalo (Ch. 11) **Tropical** Cyclones W. Pacific (Ch. 24) Wildfires California (Ch. 2) W. Tropical & N.E. Pacific (Ch. 20) Sea Surface **Temperature** N.W. Atlantic & N.E. Pacific (Ch. 13) Sea Level S. Australia (Ch. 32) Pressure Sea Ice Antarctica (Ch. 33)

Extent

[†] Papers that did not investigate strength are not listed.

^{††} Papers that did not investigate likelihood are not listed.

^{*} No influence on the likelihood of low rainfall, but human influences did result in higher temperatures and increased net incoming radiation at the surface over the region most affected by the drought.

^{**} An increase in spring rainfall as well as extensive artificial pond drainage increased the risk of more frequent severe floods from the enhanced rainfall.

^{***} Evidence for human influence was found for greater risk of UK extreme rainfall during winter 2013/14 with time scales of 10 days

^{****} The study of Jakarta rainfall event of 2014 found a statistically significant increase in the probability of such rains over the last 115 years, though the study did not establish a cause.

	ON EVENT LIKELIHOOD ††			Total Number
	INCREASE	DECREASE	NOT FOUND OR UNCERTAIN	of Papers
Heat	Argentina (Ch. 9) Australia (Ch. 30, Ch. 31) Australia, Adelaide (Ch. 29) Australia, Brisbane (Ch. 28) Europe (Ch. 13) S. Korea (Ch. 19) China (Ch. 22)		Melbourne, Australia (Ch. 29)	7
Cold		Upper Midwest (Ch.3)		1
Winter Storms and Snow	Nepal (Ch. 18)		Eastern U.S.(Ch. 4) N. America (Ch. 6) N. Atlantic (Ch. 7)	4
Heavy Precipitation	Canada** (Ch. 5) New Zealand (Ch. 27)		Jakarta**** (Ch. 26) United Kingdom*** (Ch. 10) S. France (Ch. 12)	5
Drought	E. Africa (Ch. 16) S. Levant (Ch. 14)		Middle East and S.W. Asia (Ch. 15) E. Africa* (Ch. 17) N.E. Asia (Ch. 21) S. E. Brazil (Ch. 8) Singapore (Ch. 25)	7
Tropical Cyclones	Hawaii (Ch. 23)		Gonzalo (Ch. 11) W. Pacific (Ch. 24)	3
Wildfires	California (Ch. 2)			I
Sea Surface Temperature	W. Tropical & N.E. Pacific (Ch. 20) N.W. Atlantic & N.E. Pacific (Ch. 13)			2
Sea Level Pressure	S. Australia (Ch. 32)			I
Sea Ice Extent			Antarctica (Ch. 33)	I
			TOTAL	32

 $[\]ensuremath{\dagger}$ Papers that did not investigate strength are not listed.

^{††} Papers that did not investigate likelihood are not listed.

No influence on the likelihood of low rainfall, but human influences did result in higher temperatures and increased net incoming radiation at the surface over the region most affected by the drought.

^{**} An increase in spring rainfall as well as extensive artificial pond drainage increased the risk of more frequent severe floods from the enhanced rainfall.

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^{****} The study of Jakarta rainfall event of 2014 found a statistically significant increase in the probability of such rains over the last 115 years, though the study did not establish a cause.