

Beyond the Hockey Stick: Climate Lessons from The Common Era

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Abstract: More than two decades ago, my co-authors Raymond Bradley, Malcolm Hughes and I published the now iconic “Hockey Stick” curve. It was a simple graph, derived from large-scale networks of diverse climate proxy (“multiproxy”) data such as tree-rings, ice cores, corals and lake sediments, that captured the unprecedented nature of the warming taking place today. And it became a focal point in the debate over human-caused climate change and what to do about it. Yet the apparent simplicity of the hockey stick curve betrays the dynamicism and complexity of the climate history of past centuries and how it can inform our understanding of human-caused climate change and its impacts. In this article, I discuss the lessons we can learn from studying paleoclimate records and climate model simulations of the “common era”, the period of the past two millennia during which the “signal” of human-caused warming has risen dramatically from the background of natural variability.

Significance Statement: *I review the significant developments, current challenges, and prospective future directions in the sub-field of paleoclimatology of the common era since the publication of the now iconic “hockey stick” curve by the author and collaborators more than two decades ago.*

Clearly there is a cautionary tale told by the hockey stick curve in the unprecedented warming that we are causing, but the lessons from the paleoclimate record of the common era go far beyond that. What might we infer, for example, about the role of dynamical mechanisms relevant to climate change impacts today from their past responses to natural drivers? Examples are the El Nino phenomenon, the Asian Summer Monsoon, and the North Atlantic ocean “conveyor belt” circulation. Are there potential “tipping point” elements within these climate sub-systems? How has sea level changed in

past centuries, and what does it tell us about future coastal risk? Are there natural long-term oscillations, evident in the paleoclimate record, that might compete with human-caused climate change today? Can we assess the “sensitivity” of the climate to ongoing human-caused increases in greenhouse gas concentrations from examining how climate has responded to natural factors in the past? And can better estimates of past trends inform assessments of how close we are to critical “dangerous” warming thresholds? In this article I seek to address such questions and offer thoughts about ways forward to more confident answers.

A Longer, Sturdier Hockey Stick

In the two decades since the original “hockey stick” work of Mann et al [1998(1) and 1999 (2), often referred to as “MBH98” and “MBH99” respectively] which extended back to 1000 CE, far more paleoclimate data have become available, more sophisticated methods have been developed and applied to these data, and longer reconstructions have been achieved using lower resolution but considerably longer paleoclimate proxy records. The net result is a veritable “hockey league” (3,4)--dozens of independent studies that come to similar conclusions, and a longer, sturdier “hockey stick” (Figure 1) that shows the recent warming to be anomalous in an even longer-term context--at least the past two millennia, and more tentatively, the past 20,000 years (5,6)--than originally concluded two decades ago by Mann et al (2). Studies using climate models driven by estimated natural (volcanic and solar) and anthropogenic forcing demonstrate that only the latter can explain this unprecedented warming trend (3).

From the standpoint of the public climate change discourse, these conclusions are significant. They underscore the profound, unprecedented impact human activity—fossil fuel burning in particular—is having on our planet. From a scientific standpoint, however, some of the more important insights--insights that indeed motivated the original work of MBH98--involve past patterns of variability and what they can tell us about the dynamics of the climate system and the response to external drivers or “forcings”. We explore these insights in the following section.

Dynamical Mechanisms and Responses

Many of the key impacts of climate change involve dynamical components of the climate system and their response to anthropogenic climate forcing. These include the El Nino/Southern Oscillation (ENSO) which influences weather patterns around the world, impacting western U.S. drought and Atlantic hurricane activity among other phenomena. These include the Arctic Oscillation (AO) or closely related North Atlantic Oscillation (NAO) which impact weather patterns in North America and Eurasia. They also include the Asian Summer Monsoon upon which more than a billion people depend for fresh water supply. Finally, there is the Atlantic Meridional Overturning Circulation (AMOC), often termed the “ocean conveyor belt” circulation. The AMOC delivers warm water to the high latitudes of the North Atlantic, warming neighboring regions, and circulating nutrients to North Atlantic surface waters, while suppressing sea level along parts of the U.S. east coast.

We can better understand these key dynamical components of the climate system and their potential role in climate change by studying how they have responded to past natural drivers, such as volcanic and solar radiative forcing. Moreover, we can assess whether dramatic changes in these systems are underway today by comparing recent trends against the record of the past one-to-two millennia.

ENSO

Drought in the desert southwest, which impacts large population centers in California, Nevada and Arizona, is modulated by ENSO. While the influence varies from one event to the next, El Nino years tend to be wetter than normal and La Nina years dryer than normal. ENSO also impacts Atlantic hurricane activity, with El Nino years less active and La Nina years more active than normal. Any changes in ENSO mean state and variability could have profound impacts on these and other phenomena impacting North and South America, Africa, Australia, Indonesia and other regions influenced by ENSO.

State of the art climate models such as those used in the most recent IPCC (AR5) assessment, however, provide limited guidance. They display a large range in the response of the tropical Pacific mean state to anthropogenic greenhouse forcing. Moreover the models, on the whole, are inconsistent with the observations in that they exhibit a trend toward an El Nino-like mean state, with a decreased contrast between the western Pacific warm pool and eastern Pacific “cold tongue”, while the observations, which show a neutral or even opposite, La Nina-like trend over the past half century. Seager et al (7) argue that this failure for models to produce the observed response to

heating is a consequence of a biased mean state in the models, associated with too strong upwelling in the cold of a “cold tongue” in the eastern equatorial Pacific. They warn that this may bias climate model projections in the many regions sensitive to tropical Pacific sea surface temperature (SST) forcing.

The dynamical mechanisms in question relate to earlier work by Clement and colleagues (8,9) who argued that the upwelling of cold waters and shallow thermocline in the eastern equatorial Atlantic oppose radiatively-forced warming in the eastern equatorial Pacific as the western equatorial Pacific continues to warm. The Bjerknes feedbacks amplify this forced response, creating stronger trade winds and a stronger zonal SST gradient, i.e. an overall “La Nina”-like relative pattern of SST change. Recent work yields a more nuanced picture, suggesting a transient tug-of-war between these dynamical responses and the thermodynamic response of a warming thermocline, with a net response that may be timescale dependent (10-14). The importance of the dynamical response to forcing nonetheless appear to remain relevant on the multidecadal-to-centennial timescales of interest.

Multiproxy reconstructions of surface temperature patterns spanning the past millennium based on tree-rings, corals, lake sediments, ice cores and other proxy sources, seem consistent with such a dynamical response, displaying a La Nina-like cooling in the eastern equatorial Pacific (15) and, a pattern of a dry desert southwest U.S. (16) and wet Pacific northwest U.S. (17) that is consistent with a La Nina-like state during the early

part (AD 1000-1400) of the past millennium. Consistent with the hypothesized dynamical response, that state coincides with a period of anomalous positive (high solar, low volcanic) natural radiative forcing. The La Nina-like pattern is not reproduced in global coupled climate model simulations of the past millennium (8), which could be due to uncertainties and biases in the proxy records available that far back in time and/or biases in the models. As noted earlier current generation climate models don't reproduce the observed historical trend of little or no warming in the eastern equatorial Pacific. The paleoclimate record of the past millennium, in that sense, seems to reinforce the notion that models aren't getting certain important dynamical response to forcing right, and may be underestimating key climate change impacts such as aridification in the western U.S. and heightened hurricane activity in the tropical Atlantic.

One important contributor to the medieval "La Nina" pattern is the relative absence of volcanic eruptions during the earlier centuries of the past millennium. A number of observational (18) and modeling (19,20) studies indicate a tendency for an El Nino-like response to volcanic radiative forcing, consistent with the thermostat response. One recent study based on a long coral record from the central equatorial Pacific (21) argues against such a response. But others (20) have noted that a coral proxy responding to local SST changes in the central equatorial Pacific might not detect a response at all, while remotely located coral or tree ring proxies, such as those used in large-scale climate reconstructions (15) might better detect an El Niño-like response. Additional high-resolution proxy records from key ENSO-sensitive regions spanning the past millennium might shed further light on this puzzle. At a time when one impacted region—the desert

southwest U.S.—is experiencing droughts that are unprecedented in at least 1200 years (22), this is an important puzzle to solve.

AO/NAO

Another intriguing dynamical response to forcing involves the AO/NAO, a pattern of variation in the winter storm track from year to year that is especially prominent over the North Atlantic sector, and impacts winter temperatures and precipitation over a large part of North America and Eurasia. Multiproxy reconstructions and model simulations suggest that the relative cold of certain regions like Europe during the so-called “Little Ice Age” (e.g. 15th-19th centuries) and, conversely, the relative warmth of those regions during the Medieval era (11th-14th centuries) are consistent with a negative and positive AO/NAO-like pattern, respectively, during those time intervals (15).

This response appears to be driven by the interaction between solar UV radiation and lower stratospheric/upper tropospheric atmospheric dynamics that leads to a negative AO/NAO pattern during periods of low solar irradiance (e.g. the “Little Ice Age”) and conversely, a positive AO/NAO pattern during Medieval times. Mann et al (15) show that a simulation of the past millennium with a model that includes interactive ozone photochemistry reproduces the pattern in the multiproxy reconstructions while a simulation with a model that lacks these processes does not. The absence of interactive ozone photochemistry in the vast majority of last millennium intercomparisons (e.g. CMIP5) is a severe limitation on the ability of those models to capture key regional climate responses (23), one that should be addressed in future such intercomparisons.

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171 *South Asian Summer Monsoon*

172 Precipitation tied to the South Asian Summer Monsoon (SASM) provides freshwater for
173 a large population in South Asia, and its potential future behavior under climate change is
174 a matter of considerable attention. While we can learn something about this important
175 component of the climate system from studying its response to past natural radiative
176 forcing, we must also be aware of the limitations and caveats involved.

177

178 The SASM is characterized by a largely meridional-vertical circulation with rising
179 motion over the Indian subcontinent driven by differential solar heating and orographic
180 lifting, and descending motion over the Indian ocean. It is sometimes equated with the
181 precipitation that occurs over land as a consequence of this circulation. However, the
182 monsoonal circulation and rainfall need not covary. In what has sometimes been referred
183 to as the “wind-precipitation paradox” (24,25), moistening of the mid-troposphere under
184 e.g. greenhouse warming leads to increased rainfall but stabilization of the mid-
185 troposphere in the sinking region of the SASM, inhibiting the Monsoonal overturning
186 circulation. A divergence is thus seen between SASM circulation and SASM-produced
187 precipitation during the 20th century and in future projections (24,25). Recent, high-
188 resolution model simulations demonstrate that the two quantities are in fact essentially
189 decoupled: it is possible to have a strong monsoon winds with no rain or monsoon rain
190 with weak meridional winds (26).

191

192 In an analysis of a simulation of the past millennium, Fan et al (27) found only moderate
193 covariation between the SASM circulation and SASAM rainfall during the pre-industrial
194 period and a marked decoupling during the modern era, where the SASM circulation
195 weakens but SASM precipitation remains roughly constant. Such findings present a
196 challenge for the interpretation of proxy reconstructions of the SASM over the past
197 millennium, and potentially explain the divergence seen among various purported SASM
198 proxies, some (e.g. tree-rings) being proxies for rainfall while others are proxies for wind
199 speed and direction (3,19,28).

201 There are additional challenges in drawing conclusions about the SASM from both
202 simulations and proxy data of the past millennium. There is an important influence of
203 ENSO on the SASM, with El Nino (La Nina) leading to a tendency for a weakened
204 (strengthened) monsoonal circulation. Fan et al (27) observe a significant weakening of
205 the SASM in response to past volcanic forcing, consistent with reduced surface solar
206 heating that induces a reduction in overturning circulation. However, the dynamical
207 response of ENSO to radiative forcing discussed earlier is not evident in most coupled
208 model simulations including the one they analyzed. This response should lead to further
209 weakening of the SASM owing to an El Nino-like response to volcanic forcing. A further
210 complication is the potential role of Indo-Pacific coupling that is known to impact the
211 SASM but is not well captured in models during the modern era (29) and appears to have
212 exhibited substantial variability over the past millennium (30).

Future proxy-based work should adopt a more systematic approach to defining past changes in the SASM, distinguishing in particular between proxies of circulation and proxies of precipitation. Future modeling work should employ higher-resolution, next generation climate model simulations where SASM dynamics are well resolved, and where key dynamical mechanisms such as the response of ENSO to radiative forcing, are more faithfully represented.

The AMOC

One of the most significant potential dynamical responses to anthropogenic climate forcing involves the behavior of the AMOC/ocean “conveyor belt”, a thermohaline overturning circulation that is driven by the sinking of cold, fresh water in high-latitude regions of the North Atlantic. While there has been much focus on the role of freshwater input and AMOC weakening during the last glacial termination and the so-called 8.2 ka event during the early Holocene, there has been less focus on the behavior of the AMOC during the common era. Rahmstorf et al (31) however estimated changes in AMOC over the past millennium using both an AMOC index derived from multiproxy-based North Atlantic SST reconstructions (15) and coral proxy $\delta^{15}N$ data that serve as tracers of Labrador slope water. They found an anomalous decrease in the strength of the AMOC over the past century in the context of the past millennium. More recent work by Caesar et al (32) using marine sediment silt data and foraminifera proxy data provide longer-term, additional evidence, suggesting that the AMOC slowdown over the past century is unprecedented since at least AD 400.

This finding is significant because AMOC collapse constitutes one of the potential “tipping point” responses to anthropogenic climate forcing. As climate models don’t predict a substantial weakening until the late 21st century, the fact that weakening may have already occurred during the 20th century suggests AMOC collapse could be proceeding faster than scheduled, perhaps due to earlier than expected freshwater input from Greenland melt. (31). While direct AMOC observations show conflicting trends (33), one recent study (34), based on a variety of complementary metrics, argues that AMOC collapse does appear underway.

Potential impacts of AMOC collapse include decreased marine productivity in the North Atlantic, accelerated sea level rise along parts of the U.S. east coast (arising from the geostrophic balance maintaining the northward current system) and the potential for greater tropical North Atlantic warming and increased Atlantic hurricane activity (both of which are discussed in the subsequent section). It is thus of vital importance to address the current discrepancies among models and observations. One current limitation is that interactive ice sheet meltwater coupling is not incorporated into multimodel climate change experiments (e.g. CMIP5 and CMIP6). Including such processes in future modeling experiments should allow for more confident conclusions. Meanwhile, a “future-proof” network based on existing and new AMOC observational approaches could better constrain historical trends (33).

Sea Level Rise and Tropical Cyclones

Climate change poses twin threats to coastal settlement in the form of sea level rise, and more intense tropical cyclones (TCs). Paleoclimate data and model simulations of the common era can inform our understanding of these threats and place them in a longer-term context.

Reconstructions of landfalling Atlantic tropical cyclones from coastal overwash deposits spanning the past two millennia suggest that the increase in basin-wide Atlantic TC activity over the past several decades is unusual, but not necessarily unprecedented, over that time frame (35). Comparison with a statistical model of Atlantic TC activity driven by proxy-reconstructed (15) indices of ENSO, tropical Atlantic SST and the NAO predict a period of high activity during the Medieval (associated with a warm tropical Atlantic and a prevalent La Nina-like state) that is matched by the overwash deposit evidence, but there is a discrepancy during the 15th century where the overwash deposits indicate a period of high activity that is not matched by the proxy climate index-driven statistical model.

Synthetic long-term hurricane datasets derived using downscaling approaches applied to coupled model millennium simulations of the past millennium have been used both to address statistical sampling issues (36) and examine long-term climate/TC relationship (36). Kozar et al (36) analyze downscaled synthetic TCs from an early forced millennial simulation of the NCAR CSM 1.4 coupled model, concluding that proxy composites of overwash deposits from a modest set of locations along the U.S. East, Gulf Coast and Caribbean are reasonably representative of basin-wide TC activity, though multidecadal

periods of divergence, such as observed during the 15th century, are consistent with expected sampling fluctuations. These conclusions are supported by additional work by Reed et al (37) examining downscaling results from the CMIP5 last millennium simulations. These simulations show only a weak long-term relationship between tropical Atlantic SST and measures of integrated TC intensity (e.g. the Power Dissipation Index), suggesting that the strong relationship between the two quantities during the modern era (38) might be specific to modern anthropogenic warming in recent decades and not generalizable. More confident assessments of long-term basin-wide trends should be possible as additional long-term paleohurricane records are recovered (39,40).

Sea level reconstructions based on coastal deposits indicate that the current rate of sea level rise is unprecedented over the past two millennia (41,42). The combination of rising sea level and more intense hurricanes that arise in downscaled historical simulations (43) has led to vastly reduced return periods for Superstorm Sandy-like storm surges for New York City. Garner et al (44) find that the 500 year return period for a 2.25 meter flood height during the pre-anthropogenic common era has decreased to roughly 24 years in the anthropogenic era. They find that a combination of projected future sea level rise and further intensification of hurricanes would likely yield permanent inundation for New York City under business-as-usual emissions, though a tendency for continued poleward shift in hurricane paths with climate change that has been noted in historical and paleoclimate observations (45,46) tends to mitigate risk for New York City at the expense of increased risk further north (e.g. Boston).

Internal Climate Oscillations

The analysis of paleoclimate data by Mann et al that led to the “Hockey Stick” reconstruction of Northern hemisphere average temperature (1,2) was an outgrowth of earlier work by Mann et al (47) analyzing networks of multiproxy data to assess evidence for natural long-term climate oscillations. Evidence of an apparent spatiotemporal mode of large-scale surface temperature variability emphasizing the North Atlantic basin, combined with evidence of a similar mode of climate variability in control simulations of early generation coupled ocean-atmosphere models (48) led to the notion of the “Atlantic Multidecadal Oscillation or “AMO”--a term originally coined by M. Mann (49) in an interview with Richard Kerr of *Science* magazine.

The notion that a natural, internal oscillation with a multidecadal (50-70 year period) timescale might be responsible for an array of climate trends including tropical Atlantic warming and increases in Atlantic tropical cyclone activity has since become widespread. A body of work over the past decade analyzing observations and climate model simulations, however, strongly calls into question whether such a mode of internal climate variability even exists (see ref. 49 for a review).

Mann et al (49) showed that control simulations of state-of-the-art (CMIP5) climate models don’t produce any consistent evidence for an internal “AMO”-like oscillation (or an interdecadal “Pacific Decadal Oscillation”). Indeed, they find they show no evidence of oscillatory variability other than the interannual ENSO phenomenon. An apparent ~50 year oscillatory signal in the instrumental surface temperature record is reproduced from

historical climate model simulations and seen to be an artifact of the competition between long-term greenhouse warming and the more recent decrease in sulphate aerosol cooling in the late 20th century, rather than a natural long-term climate oscillation.

Increasingly, the prevailing view is that decadal and longer timescale internal variability is indistinguishable from colored noise (50). One seeming contradiction with that interpretation, however, is the aforementioned evidence of interdecadal and multidecadal spectral peaks in the analysis of paleoclimate proxy data from past centuries (47).

Focusing on the putative ~40-60 year “AMO” signal, Mann et al (51) recently showed that multidecadal spectral peaks evident in analyses of climate model simulations of the past millennium (CMIP5 “Last Millennium” experiments) are a consequence of the coincidental multidecadal pacing by explosive volcanism in past centuries, explaining why these peaks are seen in the (forced) CMIP5 Last Millennium simulations but not control simulations of the very same models. Indeed, they show that these spectral peaks are evident in simple zero dimensional energy balance models forced by estimated past volcanic radiative forcing (Figure 2). Recent work by Waite et al (52) comes to similar conclusions based on a comparison of sclerosponge proxy data and CMIP5 Last Millennium simulations.

The best available evidence today thus argues against the existence of internal interdecadal and multidecadal climate oscillations, casting doubt on claims that an “AMO” oscillation is responsible for increases in tropical Atlantic SST or Atlantic hurricane activity and casting doubt on prospects for long-term predictability of internal

interdecadal and multidecadal climate variability. Further analysis of paleoclimatic data and direct comparisons with forced model simulations should allow for further validation of the recent conclusions by Mann et al (51) regarding the apparent volcanic origin of AMO-like variability in past centuries.

Climate Sensitivity

Paleoclimate plays an important role in informing assessments of the warming effect of an increase in greenhouse gas concentrations. That effect can be measured in terms of the transient climate response (TCR), defined as the warming at a given point in time when a doubling of the concentration of carbon dioxide in the atmosphere (equivalent to a roughly 3.7 W/m^2 radiative forcing) has been achieved. Most paleoclimate studies however focus on the so-called equilibrium climate sensitivity or “ECS”, the estimated warming that results, in equilibrium, from that same doubling. This is the “Charney” definition of climate sensitivity and accounts for “fast” feedbacks related to clouds, water vapor, ice, etc. (an alternative related quantity, known as “Earth System Sensitivity” (“ESS”) takes into slow feedbacks related to ice sheet dynamics, long-term shifts in vegetation zones, etc.).

Though defined in the context of greenhouse warming, ECS is often regarded as universal enough of a quantity that it can be measured from the response of the climate to other radiative forcings, including natural solar and volcanic radiative forcing (though caveats herein are noted below). The historical record provides relatively poor constraint on estimates of ECS owing to the shortness of the record and the fact that there are

multiple competing and uncertain (particularly anthropogenic aerosol) radiative forcings over the duration of the record. ECS values lower than 1.5C or higher than 8.5C cannot be ruled out with a reasonable (68%) degree of confidence from this line of evidence alone (Figure 3).

The “likely” range of ECS, based on a variety of lines of independent (or largely independent) evidence, is generally considered to be ~1.5C-5C (Figure 3), with a central estimate of 3C. Such evidence (53,54) includes the match of model simulations to current climatological averages, the average value of ECS as diagnosed directly from climate models, the response of the climate to volcanic eruptions, and last, models and observations from the Last Glacial Maximum (“LGM”), geological evidence over millions of years, expert judgment, and, last but not least, comparisons of paleoclimate observations and model simulations over the past millennium.

It is notable that this last line of evidence leads to the lowest estimate of ECS of all, just above 2C (Figure 3), whereas the average among all lines of evidence is closer to 3C. In most last millennium studies, ECS estimate are typically obtained by varying the ECS value in an energy balance model simulation (where it is a simple adjustable parameter) and determining the value at which the best fit is achieved between simulated and proxy-reconstructed global or hemispheric mean temperatures. The dominant source of radiative forcing during the pre-industrial common era is the cooling response to volcanic aerosol loading of the stratosphere during the years following an explosive (typically tropical) volcanic eruption. In fact, one prominent study (55) only used data back to 1270 CE to

avoid the largest estimated volcanic forcing event of the past millennium (the 1258 CE eruption) where a dramatic data/model mismatch is observed. While the eruption is estimated to have given rise to a radiative forcing of about -12 W/m^2 (roughly four times larger than the 1991 Pinatubo eruption) little or no response is seen in tree-ring temperature proxies, which make up a significant component of most proxy reconstructions of hemispheric and global mean temperature. Though this problem is highlighted by the 1258 CE discrepancy, it is hardly likely to be limited to that year.

Mann et al (56) use a combination of model-simulated temperature over the past millennium and a simple simulation of tree ring responses to argue that the reliance on treeline-proximal tree-ring sites in temperature reconstructions leads to both an underestimation bias in recording very large eruptions (which dominate the forced climate response prior to the industrial era) and chronological errors that accumulate back in time. This arises from the fact that very cold summers may lie below the minimal temperature threshold for growth—a problem that leads to a loss of sensitivity to cooling (correlated over large spatial regions) and the potential accumulation of chronological errors back in time (if there is no growth during a given summer season, then there is no ring recorded). This effect leads to an attenuation and smearing of the apparent response to very large eruptions that increases back in time that is reproduced by the simulated tree growth response (56) and a substantial potential bias in estimating ECS values, yielding an estimate $\text{ECS} \sim 2.0^\circ\text{C}$ when the true value is 3.0°C (57).

While tree-ring researchers have strongly objected to these conclusions (58), there are additional lines of evidence that support them including (a) resampling experiments that show that shifts consistent with the estimated chronological errors in age models yield surrogates with large simultaneous responses to the 1258 CE and 1815 CE eruptions consistent with the model-simulated response (56) and (b) a realignment of specific tree-ring series consistent with the estimated chronological error range yields a much larger and sharper hemispheric mean cooling response to the 1258 CE eruptions (59).

Regardless of the source of the discrepancy, it is clear that the mismatch between the reconstructed and modeled volcanic cooling leads to an ECS underestimation bias.

Analyzing the CMIP5 “Last Millennium” forced simulations, Schurer et al (60) show that simply removing the few largest volcanic forcing events from model/data comparisons results in a substantially larger inferred forced response from proxy temperature reconstructions that is consistent with model simulations and the average CMIP5 model ECS value of roughly 3.2C. A further complication is the fact that volcanic events measure a short-term transient response to forcing that is arguably more of a measure of TCR than ECS, and there are substantial uncertainties involved in translating TCR to an equivalent ECS estimate (61).

There is a larger potential problem here, however, that goes beyond the issue of how well proxy data record past climate change. Forced climate responses during the common era are dominated by forcings (i.e. explosive volcanic eruptions) that lead to substantial *cooling* relative to current temperatures. This is relevant because ECS is not a universal

quantity. It involves feedback processes that are in general not the same for cold and warm global climates. Cold global climates, for example, are more likely to be impacted by cryosphere responses such as ice cover-related albedo changes, while warm global climates are more likely to be impacted by carbon cycle feedbacks related to permafrost melt and methane release or warm-climate threshold cloud responses (62). We are still in the process of better understanding the potential feedback processes that may arise in hothouse climates (63,64).

This asymmetry could mean that ECS values obtained from past “cold climate” responses are not especially instructive when it comes future potential greenhouse warming. A worst case scenario could take us to CO₂ levels not seen in tens of millions of years. Herein we face a “catch 22”. Our most reliable paleoclimate constraints include the more recent past, e.g. the common era and the last glacial maximum (LGM) where both paleodata and relevant forcings are best known. Yet both yield “cold climate” estimates of ECS. To find analogs for present greenhouse gas levels, we must go back to at least the early Pliocene 5 million years ago. And to find “warm climate” analogs for greenhouse gas levels of ~1200 ppm CO₂ equivalent, which we could reach by the end of the century under a worst-case scenario (i.e., no substantial reductions in carbon emission), we must go back to the early Eocene, ~50 million years ago.

While forcings and response are highly uncertain that far back in time, model/data comparison studies of the early Paleogene (65-35 Myr BP) suggest a state-dependent ECS that increases with warming, both due to an increase in fast (i.e. Charney) climate

465 feedbacks associated with cloud property adjustments and a non-logarithmic increase in
466 CO₂ opacity (63). Shaffer et al (64) used estimates of CO₂ and global temperature
467 change to estimate ECS both before and during the Paleocene/Eocene Thermal Maximum
468 (PETM) natural carbon release event ~56 Myr ago. They estimate that ECS increased
469 from a range of 3.3–5.6C to 3.7–6.5C. Comparing these estimates with ECS estimates
470 from the LGM and modern era suggests a significant increase in ECS with warming for
471 greenhouse climates relative to colder climates.

472
473 Sherwood et al (65) employed a Bayesian statistical approach to combine various lines of
474 paleoclimate evidence in an attempt to reduce the current uncertainty range in ECS. They
475 produce a revised “likely” (66% probability), “robust” range of 2.3-4.5C, reduced relative
476 to the canonical 1.5-5C range cited earlier, and a “very likely” (95% probability) range of
477 2.0-5.7C. The strongest constraints at the upper end of the range, in their analysis, come
478 from paleoclimate evidence from cold climates. That finding seems to be contradicted by
479 evidence cited above for substantially higher sensitivities from hothouse climates of the
480 distant past. Neither the cooling during the largest volcanic eruptions of the common era
481 nor the cooling during the LGM can provide any constraint on feedback processes that
482 are specific to hothouse climates. Even the most sophisticated statistical analysis cannot
483 account for physical responses that lie outside the range of the data analyzed.

484 485 **Dangerous Warming**

486 One uncertainty in evaluating the carbon budget left for avoiding critical warming
487 thresholds such as the 2C and (aspirational) 1.5C warming limits adopted by the Paris

climate accord involves the definition of the pre-industrial baseline with respect to which warming is measured. Many studies have, for simplicity, adopted a late 19th century (e.g. 1850-1900) baseline, since a reliable global surface temperature record is only available back to the mid 19th century (66). However, fossil fuel burning and the rise in global CO₂ concentrations began in the 18th century, and models predict that some anthropogenic greenhouse warming had occurred prior to the mid 19th century (67). Given that the warming recorded by the instrumental record is already ~1.2C, even a tenth of a degree C has a large impact on how close we are to the 1.5C (or 2C) thresholds and the carbon budgets left for avoiding those thresholds.

Given the uncertainties that exist in proxy reconstructions of global mean temperature during the common era (refer back to Figure 1), these reconstructions provide relatively little constraint on pre-instrumental warming. Climate model simulations, on the other hand, can provide more precise estimates of that warming. Schurer et al (67) use the CMIP5 Last Millennium simulations to estimate how much anthropogenic warming had occurred prior to the late 19th century period typically used to define the pre-industrial baseline, finding evidence for anywhere from 0.1-0.2C additional warming, depending on the precise pre-industrial time period used, since there are centennial-scale pre-industrial temperature fluctuations driven by natural (primarily solar and volcanic) radiative forcing. Taking into account this additional pre-instrumental warming, Schurer et al estimate as much as a 40% reduction in the carbon budget available for avoiding 2C (and even greater reduction in the carbon budget for 1.5C). Along with other considerations, including how surface air temperature and sea surface temperatures are blended in

calculating global mean temperature in models and observations, and how instrumental and global temperature series are merged (68), such technical considerations demand greater precision in how warming targets and carbon budgets are defined by policymakers and other stakeholders.

Looking Forward

The study of the common era can inform many of the key scientific questions that remain regarding climate dynamics and climate change. The large-scale warming trend of the past century is seen to be unprecedented in millennia (and likely even a longer timeframe), confirming the unprecedented nature of human-caused climate change. Past relationships between natural solar and volcanic forcing hint at potential dynamical responses to human caused warming (e.g. regional responses related to modes of variability such as El Nino, the Asian summer monsoon and the Atlantic “conveyor belt” ocean circulation, that remain uncertain). As these responses are likely to influence many key regional climate change impacts, it is critical to better understand them.

The pre-industrial common era can also afford us an expanded view of natural climate variability. Analyses of the past millennium, for example, are seen to cast doubt on the existence of AMO-like internal multidecadal oscillations that have been invoked to argue against an impact of anthropogenic climate change on key climate impacts such as the observed increase in Atlantic hurricane activity. These analyses furthermore suggest limited potential for long-range climate predictability through initialized model prediction beyond the seasonal predictability afforded by ENSO.

534

535 Finally, studies of the common era inform important climate policy assessments. These
536 include the evaluation of the increased coastal risk from sea level rise and tropical storm
537 intensification, the estimation of climate sensitivity to greenhouse gas increases, and
538 estimates of the carbon budget remaining for keeping warming below critical 1.5C and
539 2.0C planetary danger limits.

540

541 What is the path forward to more confident insights? It is a truism that better models,
542 more and higher-quality paleoclimate proxy data, and more careful comparisons of the
543 two can yield more confident inferences. But it's helpful to be more precise than that.
544 Clearly, as we have seen, volcanic forcing places a critical role in forced climate change
545 during the common era, and yet the estimates of the forcing remain widely variable (69).
546 Efforts to reduce that uncertainty would clearly pay dividends, but so would efforts to
547 expand and diversify the available networks of high-resolution proxy given the potential
548 biases and limitations in dating, interpretation, and climate signal sensitivity that are
549 specific to individual proxy types such as tree-rings, corals and ice cores.

550

551 Equally important, however, are current limitations in climate models used in "Last
552 Millennium" experiments (e.g. CMIP5 and now CMIP6). As we have seen, ENSO-
553 related dynamics play a particularly important role in regional climate responses. Yet
554 there is reason to believe that most current generation climate models may not exhibit the
555 correct response of those dynamics to forcing. Furthermore, as noted earlier, potentially
556 important AO/NAO response to solar forcing require interactive ozone photochemistry,

which has not been incorporated into most model simulations including the CMIP5 Last Millennium simulations (16).

Of course, limitations in current generation climate models lead not only to uncertainties in forced responses but internally-generated variability as well. The possibility cannot be ruled out, for example, that the absence of AMO-like climate oscillations in current generation coupled models (35,37), rather than indicating the absence of such oscillations in the real world, reflects a limitation in the representation of oceanic boundary currents, gyre and overturning circulations and/or surface ocean-atmosphere coupling.

One particularly promising path forward combines the multiple sources of information we have--paleoclimate proxy data, models and forcing estimates--in the form of data assimilation experiments. These experiments can be used to provide better constraint on key parameters of the climate system (70). More often they are used to merge proxy data and model simulations in the process of reconstructing past climate fields (71-75).

Earlier proxy data assimilation efforts sought to reduce computational demand by employing “particle filter” approaches to assimilation with climate models of intermediate complexity (71, 72). More recent efforts such as the Last Millennium reanalysis project (73-75) have made use of long-term forced coupled model integrations (CMIP5 Last Millennium experiments). But, as always, caveats and limitations must be taken into account. The final product is no better than the models, data and forcings that go into it, and the previously discussed limitations in each must be kept in mind.

580

581 In data assimilation, the climate physics of the model is in essence used to fill in missing
582 information, a form of “smart interpolation” that yields complete climate fields. But the
583 accuracy of these reconstructed fields is limited by the ability of the models to reproduce
584 real-world ocean and atmospheric circulation patterns. When the patterns present in the
585 proxy data don’t fit with the patterns produced by the models, the assimilation product
586 represents an imperfect compromise between the conflicting sources of information (70),
587 smearing out and distorting any real-world climate features. A specific source of concern
588 is the fact, discussed earlier, that current climate models do not appear to reproduce SST
589 gradients in the tropical Pacific during the historical era and don’t match longer-term
590 trends in those gradients inferred from paleoclimate data. Data assimilation is unlikely to
591 resolve such fundamental discrepancies and may simply obscure what is actually
592 happening.

593

594 There are of course the limitations in the underlying proxy data themselves and how they
595 are assimilated into the models. Past studies have generally (70-74) assumed simple
596 linear relationships between proxy data and the model variables (temperature,
597 precipitation, upwelling, etc.) they are purported to represent. However, as discussed
598 earlier, there are open questions about tree-rings and their ability to record the largest
599 volcanic cooling events, complications in tree-rings, corals and ice cores due to mixed
600 precipitation and temperature signals, the problem of threshold response limits of proxy
601 responses. Such non-linearities *could* in principle be accounted for through the use of
602 non-linear regression methods such as neural nets, or the use of forward models

connecting the proxy data to the target model variables that account for non-linearities and threshold response limits. One particularly attractive prospect is the availability of interactive tools that allow users to diagnose the impact of adding a particular proxy record or set of proxy records to the data assimilation product, building on earlier efforts to investigate proxy sampling strategies (76,77).

As important as paleoclimate is to addressing fundamental questions today regarding climate science, climate impacts and climate policy, we must make sure not to *overpromise* what it can provide, particularly when there is the potential for findings to be used in crafting climate policy. Consider for example the previous discussion of equilibrium climate sensitivity (“ECS”). While paleoclimate data might provide solid constraints on the “low end” of the ECS spectrum, there is reason to be skeptical about efforts to narrow the high end of the spectrum based primarily on “cold climate” constraints, especially when work focused specifically on past hothouse climates suggests substantially higher climate sensitivities. What might be more relevant to projected future climate change is the warm climate, transient response (TCR) to increasing greenhouse gas concentrations.

We should furthermore not brush real discrepancies between models and proxy observations “under the rug”. We shouldn’t (78), for example, dismiss systematic underestimates of reconstructed responses to forcing as simply an artefact of presumed errors in the models or the forcing estimates. There is good reason to suspect potential systematic biases and limitations in the underlying proxy data themselves.

It is important to recognize the potential limitations of paleoclimate studies in addressing some outstanding questions. There is no shame in the paleoclimate research community acknowledging that paleoclimate studies cannot address all outstanding questions regarding climate dynamics, climate variability and climate change. Paleoclimate evidence should instead be viewed as providing one very valuable source of information that, combined with other sources, can provide a fuller understanding and appreciation of the climate system. Constructive feedback from other sectors of the climate research community should be taken in good faith and recognized as critical to continued progress in the field. Two decades after my co-authors and I published the “Hockey Stick” curve, I look forward to both observing and participating in that further progress.

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FIGURE CAPTIONS

Fig. 1. Comparison of Temperature Reconstructions Spanning the Common Era including the original Mann et al (1999) “hockey stick” reconstruction (1,2) and its 95% uncertainty range and several different versions of the PAGES 2k reconstruction (4) and uncertainty range, as well as the lower-resolution reconstruction of Marcott et al (5) and its uncertainties. The smoothed HadcRUT4 instrumental global temperature series is shown for comparison.

Fig. 2. EBM Simulations and associated spectra (from ref. 51). Shown are (left) global mean surface temperature anomaly series from 1000-1835 CE and (right) corresponding MTM power spectra, using both solar and volcanic forcing (blue), solar only (green) and volcanic only (red). Forcings used correspond to CEA volcanic series and SBF solar series (top), GRA volcanic series and SBF solar series (middle), CEA volcanic series and VSK solar series (bottom).

Fig. 3. Estimates of Equilibrium Climate Sensitivity (ECS) from various lines of evidence [from Mann, *Scientific American*, 2014; adapted from ref. 53].