

## Authors response to comments of L.A. Smith

*Visser et al (2017) provide an interesting and insightful discussion of signal detection in global mean temperature (GMT), focusing on the 1.5 degree target of the Paris Agreement of 2015. This paper could be made more informative by further consideration of three topics: (1) clarifying what is meant by “signal” and by “noise”, and more specifically how (whether) natural variability can be “corrected for” in an evolving nonlinear system, (2) implications of using CMIP5 models, given that those models display a wide range of values for today’s GMT, and (c) a cleaner definition of how one would detect failure to stay “well below” a temperature target, or to exceed it. These points are expanded upon below.*

### **Specific Comments**

*“Natural variability” is said to be a dominant source of uncertainty which has been “corrected for” (24). Although discussions of a climate signal coming “out of the noise” are common, the notions underlying the distinction between signal and noise in the climate context is unclear; it is not the traditional distinction of observational noise superimposed on a imprecisely measured but well-defined signal. Superposition can only be assumed in nonlinear systems given purely observational noise that has no impact on the system: natural variability, internal variability and the like alter the dynamics, and thus the “signal” itself, if such a separation exists (Smith (2001,2002)). A more appropriate conceptualization in nonlinear systems is found in consideration of an ensemble of systems each subject to a common driving and independent realizations of the relevant noise. In this case, the ensemble median would provide a well-defined signal while the distribution about it would capture the effects of noise processes. This view is of limited utility in climate science, where there is only one realization (the Earth): particular realizations need not reflect the (unobservable, non-empirical) “signal”; indeed they can diverge arbitrarily far from it. So in no sense can one expect “the” signal to emerge from the noise, given observations of a single realization. While vague appeals to something somewhat reminiscent of an adiabatic change in thermodynamics may be voiced, clear clarification of the meaning of signal and noise in the climate context would be of value.*

*In short: it would be useful to clarify how “natural variability” and “internal variability” might be isolated in the case of a complicated, nonlinear, evolving planetary system. How are we to make sense of the traditional notions of “signal” and “noise” given that the “noise” is not mere observational noise but actually a component of the system dynamics, and given that in nonlinear systems we cannot appeal to a principle of superposition of solutions (Smith, 2002).*

The modeling of climate data by stochastic climate models have been described in Mudelsee (2014, sections 2.5.1 and 2.6). He describes the suitability of climate modeling with AR(1) processes (and the more general ARIMA models as well) to describe the persistence in data.

The reviewer is right that correlated noise is not the same as climate variability arising from nonlinear systems. However, statistical modeling has proven fruitful in a wide field of ecological modeling. To stick to the modeling of global mean temperatures, we refer to our review of (statistical) trend analyses in the peer-reviewed literature in the Supplementary Material section of our manuscript (table S.1). Furthermore, Visser et al. (2015) show in their table 1 that researchers in the field of sea level rise apply 30 trend methods for quantifying "the signal" in sea level data, all with different mathematical formulations.

Note: we do not use trend models for **prediction**. Next to that, projections up to the year 2100 will be removed.

*It is also worth noting that the statistics community and the physical science community often hold very different notions of what a trend is: for the first, it is a statistically consistent combination of two well-defined models (the trend model and the noise model), while for the second it is merely a systematic, often obvious drift. Statisticians require, and quantify, consistency between these two components, and reject identification of a trend if that consistency is lacking. Physical scientists often require the observations to look trendy, and the ability to reject simple statistical models given the data, when those models are known by construction not to admit a trend. The second bar is much lower.*

*The claim that modelling groups “have not been very successful in tuning to the observed trend” (299) suggests some knowledge as to how large the spread would be in the absence of each group knowing the observed trend (aiming for the same target). It has been argued elsewhere that knowledge of such spread would be very useful to have if, perhaps, impractical to obtain.*

*Visser et al (2017) state that “mean progression derived from GCM-based GMTs appear to lie within the range of the trend-dataset combinations” (311). It would be interesting to see the variations among individual CMIP5 simulations (not the mean over them, but their distribution). The IPCC AR5 reports that variations in the global mean temperature of today’s CMIP5 GCMs have a range exceeding 2.5 degrees (see right side axis labels of Figure 9-08 of Flato et al (2013)); what are the implications of our best models showing a range of GMT almost twice the 1.5 degree target? Physical and biological processes are driven by actual temperature, not anomalies. Given the current (limited) level of realism in these models, and the fact there is a great deal more in them than their basis in physical understanding, the authors might wish to reconsider calling today’s GCMs “fully physics-based” (86).*

The upper panel of figure 4 shows in part what the reviewer asks for. We will discuss the implication of the wide range of incremental values at the end of the discussion section. Here, we will argue that GCM simulations are less suited for tracking the signal in GMTs due to their wide range. Another argument will be that GCM simulations in CMIP5 are up to date up to the year 2005. Estimates for the period 2006-2016 are less reliable.

We propose to add the important comment of the reviewer that GCMs give a wide range of estimates for the global temperature over the period 1961-1990. Not as anomalies but in **absolute temperatures**. Indeed, figure 9.8 of the AR5 WGI report (2013, page 768) shows a range from 12.6 °C to 15.3 °C, based on 36 models. This range is almost the double of the 1.5 °C limit.. Also see figure 1 upper panel in Hawkins and Sutton 2016 BAMS 963-980.

Finally, we propose to remove the expression that GCMs are 'fully physics based'. That is, indeed, not true.

*Lastly: what precisely does it mean to hold GMT “well below” (14) some temperature threshold? How would we know if we had missed this target? Can this be phased with sufficient precision to allow, say, an insurance contract or legal wager to hinge on its occurrence? Issues include the duration for which the threshold is exceeded (An instant? A month? A year? A decade?) and how to deal with the imprecision in measuring*

101 *the global mean temperature, even today. In practice, simply setting the target*  
102 *as an absolute value of GMT, inspired by the agreed 1.5 change, would prove more*  
103 *straightforward both scientifically and legally, even if not politically or diplomatically.*  
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105 Good point. However, we propose to remove Section 5.2 where we extend the historical  
106 analysis to the year 2100. Therefore, this important comment is not directly applicable to our  
107 revised text.  
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