I am a Professor of Climatology at the University of Delaware and I served as the Delaware State Climatologist from 2005 to 2011. I also am an adjunct faculty member in the Department of Agricultural Economics & Statistics and the Physical Ocean Science and Engineering Program. I received a B.A. in Mathematics and Geography, a M.S. in Geography, and a Ph.D. in Climatology, all from the University of Delaware. I served on the faculty of the University of Oklahoma and Louisiana State University before returning to the University of Delaware in 1999. I was part of the US delegation that negotiated a protocol for the first climate data exchange program with the Soviet Union in 1990. I am recognized as a Certified Consulting Meteorologist by the American Meteorological Society and was the recipient of the 2002 Boeing Autometric Award in Image Analysis and Interpretation by the American Society of Photogrammetry and Remote Sensing.

I would like to thank the Chair and the Committee for the privilege to offer my views and my thirty years of experience on climate change from the perspective of a climatologist. My expertise lies in statistical methods in climatology, particularly as it relates to the hydrologic cycle – precipitation and soil moisture. For my dissertation, I developed the first digital and gridded global precipitation and air temperature dataset that specifically incorporates biases arising from the precipitation gage measurement process. This database is still used today in climatology as a standard against which climate model-derived fields and regional assessments are compared. I also have published several important articles that discuss the impact of precipitation variability on soil moisture in regional and global studies. In the following discussion, I will address the potential impact of climate change on agriculture and relay some of my pressing concerns that are related to the treatment of climate scientists who do not agree with the anthropogenic global warming disaster scenarios.
Global Warming and Agricultural Impacts

One of the important questions raised by the response of increasing atmospheric carbon dioxide concentrations is the possible impacts on agriculture, aquaculture, and commercial/recreational fishing. Considering that CO$_2$ is food for plants and animals, this is seen as a positive and any potential negative effects are minimal. But if global surface air temperatures do rise for any reason, this will undoubtedly increase the length of the growing season which, in turn, will enhance the amount and diversity of crops that can be grown. Moreover, it will allow for more areas of the planet to be farmed, primarily in the Northern Hemisphere, thereby increasing crop productivity. Billions of acres of land in northern Canada and Russia could become cultivable. The limiting factor, however, is the moisture availability to plants as agriculture in much of the world is restricted by water availability both from precipitation and surface/groundwater reserves.

A discussion of the possible results of soil moisture availability in a warmer world depends on a complicated interaction of two factors – changes in the precipitation climatology and increases in evapotranspiration (the combined effect of soil evaporation and plant transpiration). The impacts of these two factors are opposite in sign; precipitation, when it occurs, is likely to increase but the potential for evapotranspiration also is likely to increase, both due to the increase in the saturation vapor pressure as a function of increasing air temperature. The question then is which dominates – does the increase in precipitation compensate for the increase in the evapotranspiration demand or does the increase in air temperature reduce soil moisture reserves such that droughts will become more likely? Complicating this discussion is the fact that atmospheric circulation changes may affect the patterns of precipitation so that some areas may become more drought-prone while others may become less so. **Pinpointing the exact geographical areas for which drought/increased rainfall are likely to occur lie far beyond our technology for the foreseeable future.**

To answer the questions, climatologists employ two methods. In one, historical patterns and trends over the last century are extrapolated to provide a forecast of what might happen in the future. From the demise of the Little Ice Age – a relatively cold period between about 1300 and 1850 A.D. (Soon et al. 2003) that is concomitant with decreased solar output – to the late 1990s,
air temperatures increased about 0.6°C (~1.1°F). We can use this rising trend in air temperature to make prognostications as to what we might expect from a warming world in the future. The second method involves climate models – mathematical/statistical representations of the climate system. These models are used to simulate future climate scenarios from which patterns of climate change are inferred. We will examine the results using both of these methods.

**Historical Patterns and Trends in Drought**

Several analyses have focused on patterns and trends associated with drought. Hao *et al.* (2014) used satellite analysis to examine global patterns of drought from June 1982 through December 2012 (Figure 1). Only a slight decrease in abnormally dry and moderate drought conditions has occurred, though it is not statistically significant. Note particularly the increase in global drought

![Figure 1](image)

**Figure 1:** Fraction of the global land in D0 (abnormally dry), D1 (moderate), D2 (severe), D3 (extreme), and D4 (exceptional) drought condition (adapted from Figure 5 of Hao *et al.* (2014)).
in 1998 resulting from the rather strong naturally-occurring El Niño of that year. Patterns in precipitation for the Twentieth Century show no observable trend over the entire period of record for either the globe or for either hemisphere (New et al. 2001 – Figure 2). Regionally, the only statistically significant pattern occurs for the upper latitudes of the Northern Hemisphere (where snowfall is better measured in the latter portion of the record due to better snow-gage instruments) and for the lower latitudes of the Northern Hemisphere (dominated by the Sahel region in Africa, where overgrazing has substantially changed the landscape and, consequently, the precipitation climate of the region). Sheffield et al. (2012) concur with the results of Hao et al. (2014): “more realistic calculations…suggest there has been little change in drought over the past 60 years.”

Much more research, however, has been conducted in the United States where observations are more dense and reliable. Generally, precipitation tended to increase over much of the United States between 1895 to 2012, although with much lower certainty in the record prior to 1950 (Vose et al. 2014). Groisman and Knight (2008), however, argued that “prolonged dry episodes” of precipitation have increased over the southwestern United States. McCabe et al. (2010) addressed this issue by examining a more complete dataset and concluded that there is “little
evidence of long-term positive trends in dry event length in the southwestern United States.” We concluded that El Niño and La Niña events and the Pacific Decadal Oscillation are largely responsible for the variability in trends in dry event length in the southwestern United States. Station network limitations and the treatment of missing data adversely affected the results of Groisman and Knight (2008).

Again, however, the main concern focuses on the change in precipitation relative to the change in evapotranspirative demand. Senate testimony by John Christy of the University of Alabama in Huntsville (Christy 2012) has shown that the daily all-time record high temperatures from 970 weather stations with at least eighty years of record peaked in the 1930s and the numbers since 1955 have not increased (Figure 3). This trend also is consistent for a subset of stations in the central United States and along the US West Coast (Christy 2012).

**Figure 3:** Frequency of all-time high maximum daily air temperatures, 1895 to 2011, at 970 USHCN stations with at least eighty years of observations (from Figure 1.2 of Christy 2012).
However, our primary concern in agriculture is the statistics of drought – changes in its intensity, frequency, and duration. Woodhouse and Overpeck (1998), comparing drought variability in the Central United States over the last two millennia concluded, “The droughts of the 20th century have been characterized by moderate severity and comparatively short duration, relative to the full range of past drought variability.” A plot of the Palmer Drought Severity Index, averaged for the contiguous United States, shows considerable variability from 1900 to 2012 with the droughts of the 1930s standing out, but without any long-term trend. This pattern has also been noted by the US Climate Change Science Program (2008) – “When averaged across the entire United States, there is no clear tendency for a trend…long-term trends (1925-2003)…show that droughts have, for the most part, become shorter, less frequent, and cover a smaller portion of the United States over the last century.”

![Palmer Drought Severity Index](image)

**Figure 4:** Percentage of the United States in moderate or extreme drought from January 1900 to June 2012 (from the National Climate Data Center, Asheville NC).

Regionally, there have been numerous studies and their results have been similar. For example, Bekker *et al.* (2014), using a 576-year reconstruction of flood conditions, concluded that droughts of greater magnitude, duration, and intensity have occurred previously in Utah. Knapp *et al.* (2002) found that the period since 1950 was “anomalous in the context of this [500-year]
record for having no notable multiyear drought events.” For the Idaho, Montana, and Wyoming region, Wise (2010) argued that “the instrumental record (i.e., since the late 1800s) does not contain a drought of the extent seen in the mid-1600s.” Gray et al. (2004) found that dry conditions in the Sixteenth Century (i.e., during the Little Ice Age) were greater in magnitude and duration than anything seen in the Twentieth Century for the same region.

Andreadis and Lettenmaier (2006) concluded that Midwestern droughts “have, for the most part, become shorter, less frequent, less severe, and cover a smaller portion of the country over the last century.” Even the Special Report of the IPCC (IPCC 2012) concluded “…In some regions droughts have become less frequent, less intense, or shorter, for example, in Central North America.” Indeed, NOAA scientists (NOAA 2013) concluded that the 2012 Central Great Plains drought “resulted mostly from natural variations in weather.”

In the Eastern United States, Pederson et al. (2012) concluded that recent droughts are not unprecedented over the last 346 years, with more frequent droughts occurring between 1696 and 1820 A.D. during the colder conditions of the Little Ice Age. Indeed, Quiring (2004) concurred that “…the recent growing-season moisture anomalies that occurred during 2002 and 2003 can only be considered rare events if they are evaluated with respect to the relatively short instrumental record (1895-2003)” and that condition during the 16th Century were longer and more severe.

My overall conclusion is that droughts in the United States are more frequent and more intense during colder periods. Thus, the historical record does not warrant a claim that global warming is likely to negatively impact agricultural activities.

Model-derived Trends and Patterns of Drought
Global climate models (or General Circulation Models – GCMs) are only as good as their ability to simulate precipitation. They are descriptions of the full three-dimensional structure of the Earth's climate and often are used in a variety of applications, including the investigation of the possible role of various climate forcing mechanisms and the simulation of past and future climates. There are, however, several important issues to remember with GCMs. First, they are
limited by our incomplete understanding of the climate system and how the various atmospheric, land surface, oceanic, and ice components interact with one another. They are further limited by our ability to transform this incomplete understanding into mathematical representations. We may have a general feel for the complex interrelationships between the atmosphere and the oceans, for example, but expressing this understanding in a set of mathematical equations is much more difficult. Second, GCMs are limited by their own spatial and temporal resolutions. Computational complexity and finite restrictions on computing power reduce GCM simulations to coarse generalities. As a result, many small-scale features, which may have significant impact on the local, regional, or even global climate, are not represented. Thus, we must recognize that GCMs, at best, can only present a gross thumbnail sketch. Regional assessments over areas encompassing many GCM grid cells are the finest scale resolution that can be expected. It is inappropriate, and grossly misleading, to select results from a single grid cell and apply it locally. It cannot be over emphasized that GCM representations of the climate can be evaluated at a spatial resolution no finer than large regional areas, seldom smaller than a region defined by a square several hundred miles (at least several GCM grid cells) on a side. Even the use of "nested grid models" (models which take GCM output and resolve it to finer scale resolutions) does not overcome this limitation, since results from the GCM simulation drives such models and no mechanism is available to feedback the results of such finer-scale models to the GCM.

Another limitation in GCMs is that given the restrictions in our understanding of the climate system and its computational complexity, some known phenomena are simply not reproduced in climate models. Hurricanes and most other forms of severe weather (e.g., nor'easters, thunderstorms, and tornadoes) simply cannot be represented in a GCM owing to the coarse spatial resolution. Other more complex phenomena resulting from interactions among the elements that drive the climate system may be limited or even not simulated at all. Such indicators should be flags that something fundamental is lacking in the GCM. These phenomena should be produced in the model as a result of our specification of climate interactions and driving mechanisms; their absence indicates a fundamental flaw in our understanding of the climate system, our mathematical representation of the process, the spatial and temporal limitations imposed by finite computational power, or a combination of the above.
An assessment of the efficacy of any climate model, therefore, must focus on the ability of the model to simulate present climate conditions. If a model cannot simulate what we know to be true, then it is unlikely that model prognostications of climate change are believable. However, a word of caution is warranted. It is common practice to “tune” climate models so that they better resemble present conditions. This is widely acceptable, because many parameters in GCMs cannot be specified directly and their values must be determined through empirical trial-and-error. However, this raises the concern that a GCM may adequately simulate the present climate, not because the model correctly represents the processes that drive the climate; but rather, because it has been tuned to do so. Thus, the model may appear to provide a good simulation of the climate, when in fact the model may poorly simulate climate change mechanisms. **In other words, a GCM may provide an adequate simulation of the present-day climate conditions, but it does so for the wrong reasons.** Model efficacy in simulating present-day conditions, therefore, is not a guarantee that model-derived climate change scenarios will be reasonable. To address this question, modelers often employ simulations of past climates, such as the Holocene or the Pleistocene, to see if the model provides the kind of climate that we can infer existed during such epochs. Of course, our knowledge of pre-historical climate conditions is tenuous and extremely crude, which limits the utility of such evaluations.

A final limitation in climate modeling is that in the climate system, everything is interconnected. In short, anything you do wrong in a climate model will adversely affect the simulation of every other variable. The most problematic variable is precipitation. Precipitation requires moisture in the atmosphere and a mechanism to cause it to condense (causing the air to rise over mountains, by surface heating, as a result of weather fronts, or by cyclonic rotation). Any errors in representing the atmospheric moisture content or precipitation-causing mechanisms will result in errors in the simulation of precipitation. Thus, GCM simulations of precipitation will be affected by limitations in the representation and simulation of topography, since mountains force air to rise and condense to produce orographic (mountain-induced) precipitation (e.g., the coastal mountain ranges of Washington and Oregon). Incorrect simulations of air temperature also will adversely affect the simulation of precipitation since the ability of the atmosphere to store moisture is directly related to its temperature. If winds, air pressure, and atmospheric circulation are inadequately represented, then precipitation will be adversely affected since the atmospheric
flow of moisture that may condense into precipitation will be incorrect. Plant transpiration and soil evaporation also provide moisture for precipitation; therefore, errors in the simulation of soil moisture conditions will adversely affect the simulation of precipitation. Simulation of clouds solar energy reaching the ground will affect estimates of surface heating which adversely affects the simulation of precipitation. Even problems in simulating oceanic circulation or sea ice concentrations will affect weather patterns, which affect precipitation simulations.

Equally important is the fact that inaccuracies in simulating precipitation, in turn, will adversely affect the simulation of virtually every other climate variable. Condensation releases heat to the atmosphere and forms clouds, which reflect energy from the sun and trap heat from the Earth’s surface – both of which affect the simulation of air temperature. As a result, this can affect the simulation of winds, air pressure, and atmospheric circulation. Since winds drive the circulation of the upper layers of the ocean, the simulation of ocean circulation also is affected. Air temperature conditions also contribute to the model simulation of sea ice formation, which would be adversely affected. Precipitation is the only source of soil moisture; hence, inadequate simulations of precipitation will adversely affect soil moisture conditions and land surface hydrology. Vegetation also responds to precipitation availability so that the entire representation of the biosphere can be adversely affected. Clearly, the interrelationships among the various components that comprise the climate system make climate modeling difficult. Keep in mind, however, that it is not just the long-term average and seasonal variations that are of interest. Demonstrating that precipitation is highest over the tropical rainforests and lowest in the subtropical deserts is not enough. Climate change is likely to manifest itself in small regional fluctuations. Moreover, we also are interested in intra-annual (year-to-year) variability. Much of the character of the earth's climate is in how it varies over time. A GCM that simulates essentially the same conditions year after year clearly is missing an important component of the earth's climate. Thus, the evaluation of climate change prognostications using GCMs must be made in light of the model’s ability to represent the holistic nature of the climate and its variability. In sum, the simulation of precipitation, and subsequently soil moisture, is adversely affected by inaccuracies in the simulation of virtually every other climate variable while, in turn, inaccuracies in simulating precipitation adversely affect virtually every other variable in the model.
It should be noted that GCMs are not weather prediction models. Their utility is not in predicting, for example, whether it will rain in southern England on the morning of July 14, 2087. Rather, we are interested in determining whether the probability of precipitation will be substantially different from what it is today – in both the frequency and intensity of precipitation events. In general, we want to know whether the summer of 2087 is likely to be wetter or drier than present conditions, and by how much. As such, GCMs are only used appropriately to address the likelihood of changes over large spatial and temporal scales -- assessing changes for specific dates or locations are beyond the scope of GCM utility.

But this is my biggest concern. If a climate model simulates an increase in precipitation for the near or distant future, I want to know why. In particular, I want to verify that it is because a specific precipitation-producing mechanism has changed. Are there more tropical storms or nor’easters simulated? More frontal precipitation? Is there more convective activity from surface heating that leads to more rising air? Or has the atmospheric circulation changed such that orographic precipitation is enhanced?

Unfortunately, this is where over-reliance on GCMs forecasts can betray us. In these models, precipitation is produced almost exclusively from a single mechanism – surface convection – and is often termed “popcorn precipitation” since it occurs over large regions and relatively frequently (see Zolina 2014)). When models are averaged over seasons, the classic pattern of global precipitation emerges with a moist equatorial region, decreased precipitation in the Subtropics, and increased precipitation in mid-latitudes that tapers off with colder temperatures toward the poles. While this may appear correct in the aggregate, it has achieved its apparent success without properly simulating the mechanisms that create precipitation in the real world. How possibly, therefore, can the models make accurate prognostications of precipitation when they do not simulate correctly the mechanisms that drive precipitation? And if precipitation is not modeled properly, how then can soil moisture estimates be used to prepare farmers for an uncertain future?
Stephens *et al.* (2010) identifies this problem with three state-of-the-art climate models and numerical weather prediction models. Using high resolution CloudSat observations over the oceans (where precipitation is more uniform spatially), they demonstrate that the differences between the models and the observations are much greater than observational and averaging errors. They conclude “the general tendency is for models to produce precipitation that is far too frequent, especially in midlatitudes” (*i.e.*, the United States). Note that tropical precipitation is largely convective (although some stratiform precipitation does occur – Janowiak *et al.* 1995) but that in midlatitudes, precipitation arises from a variety of mechanisms. Instead of simulating frontal passages and organized weather systems, the models exhibit “popcorn precipitation” where it rains far too often. As a consequence of having it rain too frequently, the intensity of modeled precipitation is that when it occurs, its intensity is much lower than observed. Thus, the total precipitation is reasonable but its distribution (frequency and intensity) is grossly in error. Even models that have spatial resolutions as fine as 7-14 km (4.4-8.8 mi) exhibit these problems. When averaged to seasonal averages for the globe, the models do remarkably well. However, they achieve this level of success for the wrong reasons. Regionally, the GCMs “tend to produce too much precipitation over the tropical oceans and too little in midlatitudes”. Moreover, this is where soil moisture is greatly affected – models that rain too frequently with lighter amounts will necessarily overestimate soil moisture conditions because soil moisture responds not just to the amount of precipitation but is very dependent on its timing.

As Dr. John Christy demonstrated in his Senate Testimony (Christy 2012), the March-to-July precipitation, as simulated by most climate models, exhibits considerable variability between the models but does not exhibit a long-term trend. For the Southeastern United States (Figure 5, top), the models vary from an average of less than 15 inches to more than 25 inches and most models tend to underestimate the observed precipitation from 1890 to 2012. Similarly, the models also vary from below 15 inches to more than 24 inches for the Midwestern United States (Figure 5, bottom) although the models tend to be wetter than observations.

If models indicate that precipitation is not forecast to change over this century, how do models suggest an adverse impact on agriculture will occur? Models suggest that air temperatures will increase substantially over the next century, rising by as much as 6°C (10.8°F). This indicates
that the evapotranspirative demand will increase substantially and result in lower soil moisture conditions and hence more droughts. However, these models have significantly overestimated the warming of the last fifteen years (Figure 6) such that they command little confidence.

The consistent and substantial over-predictions of the general-circulation climate models are reflected in those of the Intergovernmental Panel on Climate Change, which, in 1990, predicted that near-term warming would occur at a rate exactly double what has actually occurred. Furthermore, none of the models predicted that for 17 years 9 months, or more than half the entire satellite temperature record, there would be no global warming at all (Figure 7).
Figure 6: Global surface air temperature (in °C) as simulated by 102 climate models runs (colored, dashed lines). The solid red line is the model average, the circles are the balloon observations, and the squares are the satellite measurements (from Pielke Sr. 2014).

As Dirmeyer (2014) argues, “The problem is that coupled land-atmosphere models used for weather and climate forecasting and research have never been thoroughly validated in terms of their simulation of the coupled processes that provide predictability.” Even if the land surface model was perfect, it will provide bad simulations if forced by “an atmospheric model with serious systematic biases or inadequately represented physical processes” (see also Steinhaeuser and Tsonis 2014). Given the limitations of the models not only in predicting global air temperatures but also in estimating precipitation and soil moisture conditions, it seems that a more reasonable approach is not to rely on the model prognostications; but rather, to focus on policies that allow for adaptation to the observed variability in precipitation and soil moisture. **Droughts that have happened in the past are likely to occur again, and with likely similar frequencies and intensities; thus, preparation for their return is a better strategy than trying to mitigate them through draconian CO₂ emission control policies.**
Figure 7: Monthly global mean surface air temperature anomalies monitored by satellite (RSS), August 1996 to April 2014. For 213 months, or more than half the entire satellite record, the least-squares linear-regression (thick bright blue line) has been zero.

The Scientific Method versus Post-Normal Science

The scientific method has long been the ‘gold standard’ among scientists. It is the empirical evidence that separates science from mythology and is the key to finding scientific truth (Legates et al. 2013). Indeed, it is the evaluation of theories with observations that have trumped appeals to authority or consensus or the longevity of a theory (Legates et al. 2014). As Legates et al. (2013) argued, “results from climate models are often erroneously posited as observations themselves or even data and even when they diverge considerably from the real observations, they are used to drive theory construction…results from climate models should be used with extreme care and not be taught as scientific fact.”

As a response to policy-making when a ‘solution’ is demanded immediately and the facts are obscured by error, widely divergent views exist, models are inherently uncertain, Post-Normal Science emerges where ‘science by consensus’ reigns. It has been strongly argued that even in its early days, the Intergovernmental Panel on Climate Change abandoned the scientific method in favor of this new paradigm (Saloranta 2001, Legates et al. 2013). This inherently morphs the role of the scientist from an impartial observer and seeker of the truth to one who dons the hat of
an advocate. This is where the so-called ‘consensus arguments’ arise where an appeal to some very large percentage of scientists appears to give credibility to a particular viewpoint. Most of these consensuses are contrived (see Legates et al. 2014) and serve to push an agenda that diverges widely from truth-seeking. The scientific method has been abandoned by many in the climate change discussion with an appeal to the masses through an imaginary consensus of scientists. This has greatly undermined both the quest for truth in this debate and the respect the general public has for scientists who advocate for anthropogenic global warming disaster scenarios.

Cited Literature


