Long-term temperature and precipitation trends at the Coweeta Hydrologic Laboratory, Otto, North Carolina, USA
Stephanie H. Laseter, Chelcy R. Ford, James M. Vose and Lloyd W. Swift Jr

ABSTRACT

Coweeta Hydrologic Laboratory, located in western North Carolina, USA, is a 2,185 ha basin wherein forest climate monitoring and watershed experimentation began in the early 1930s. An extensive climate and hydrologic network has facilitated research for over 75 years. Our objectives in this paper were to describe the monitoring network, present long-term air temperature and precipitation data, and analyze the temporal variation in the long-term temperature and precipitation record. We found that over the period of record: (1) air temperatures have been increasing significantly since the late 1970s, (2) drought severity and frequency have increased with time, and (3) the precipitation distribution has become more extreme over time. We discuss the implications of these trends within the context of regional and global climate change and forest health.

Key words | climate, long-term, precipitation, quantile regression, temperature, time series

INTRODUCTION

The Coweeta Hydrologic Laboratory is among the oldest, continuously operating, environmental study sites in North America and is located in the eastern deciduous forest of the southern Appalachian Mountains in North Carolina, USA (Figure 1). The laboratory was established in 1934 to determine the fundamental effects of forest management on soil and water resources and to serve as a testing ground for theories in forest hydrology (Swank & Crossley 1988). To facilitate this, a network of climate and precipitation stations was established across the site (Figure 1, Tables 1 and 2). The research program has since expanded its focus to encompass watershed ecosystem science. The original climate and precipitation network continues to facilitate these studies and serves as the foundation of the long-term data record.

Forested ecosystems are inherently complex and require a long-term perspective to evaluate responses to natural disturbances and management. Long-term climate data can be an especially important part of this perspective, particularly when evaluating watershed responses to pulse and press climatic events. Without long-term datasets that encompass a wide range of conditions, quantifying hydrologic and ecologic thresholds can be challenging, and identifying cyclical trends or changes in key climate variables can be impossible (Moran et al. 2008). A comprehensive description and analysis of the first 50 years of the climate data recorded at Coweeta was published in 1988 (Swift et al. 1988). In that study, few climate trends were evident. For example, no significant trends in maximum or minimum annual temperatures, or changes in the distribution of precipitation were detected. Since then, in the southeastern USA, the last 25 years have been characterized by marked changes in key climate variables, including increases in precipitation (Karl & Knight 1998; Groisman et al. 2004) leading to greater streamflow (Groisman et al. 2004), increased minimum temperatures (Portmann et al. 2009), especially minimum temperatures in the summer months (Groisman et al. 2004), and increased cloud cover (Dai et al. 2006). In addition, the variability of precipitation has also changed in the southeastern USA, with increases in extreme precipitation events (Groisman et al. 2004) including high intensity rainfall as well as extreme droughts (Karl & Knight 1998). The ability to separate actual changes or significant trends in climatic variables from
natural variability requires long-term records. Hence, these long-term records are critical for detecting historical changes in climate and they can serve as benchmarks for detecting future change. Our objectives were to provide an update of the climate and precipitation network in the Coweeta Basin, summarize the long-term temperature and precipitation data in the Coweeta Basin, and analyze the temporal variation in these two data series.
SITE DESCRIPTION

The Coweeta Hydrologic Laboratory is located within the Nantahala Mountain Range of western North Carolina, USA, latitude 35°03′ N, longitude 85°25′ W (Figure 1). The Coweeta Hydrologic Lab is 2,185 ha in area and comprised of two adjacent, east-facing basins. The larger of the two basins (1,626 ha), Coweeta Basin, has been the primary...
focus of watershed experimentation. Within the basin, elevations range from 675 to 1,592 m. Climate is classified as maritime, humid temperate (Trewartha 1954; Critchfield 1966).

Historic vegetation patterns in the basin have been influenced by human activity, primarily through small homestead agriculture, both clear-cut and selective logging, the introduction of chestnut blight (Cryphonectria parasitica) (Elliott & Hewitt 1997) and hemlock woolly adelgid (Adelges tsugae) (Nuckolls et al. 2009), and fire management (Hertzler 1936; Douglass & Hoover 1988). The resulting unmanaged forests are relatively mature (~85 years old) oak-hickory (at lower elevations) and northern hardwood forests (at higher elevations) with an increasing component of fire-intolerant species (Elliott & Swank 2008). Bedrock is comprised of granite-gneiss and mica-schist. Soils are immature Inceptisols and older Ultisols and are relatively high in organic matter and moderately acid with both low cation exchange capacity and base saturation.

**METHODS**

**Climate network**

Daily temperature and precipitation data have been recorded at the main climate station (CS01) continuously since 1934 (Table 1). In addition to CS01, there are four climate stations across the basin (Table 1). In the 1980s, measurements at each of these stations were expanded in scope and temporal resolution. Each of these stations now continually measures and records (CR10X, Campbell Scientific, Inc., Logan, UT, USA). The following variables every 5 min: temperature and relative humidity in both an open field and forested setting (HMP45c, Campbell Scientific, Logan, UT, USA), photosynthetically active radiation (LI-190-SB, Campbell Scientific), soil and litter temperature under forest cover (107-L, Campbell Scientific) and wind speed and wind direction at canopy height (014A and 024A, MetOne Instruments, Grants Pass, OR, USA). Barometric air pressure (Vaisala CS106, Campbell Scientific), solar radiation (model 8-48, Eppley Lab, Inc., Newport, RI, USA), atmospheric CO₂ concentration (Licor LI-820, Licor, Lincoln, NE, USA) and pan evaporation are measured only at CS01. Data retrieval from the climate network is via wireless remote access. All data recorded to the CR10X datalogger are transmitted via radio frequency (Free-wave Technologies, Inc., FGR-115RC, Boulder, CO) from each of the four climate stations to a computer server in the data processing office.

Temperature is recorded at CS01 at 0800 EST daily using a National Weather Service (NWS) maximum, minimum and standard thermometer. Daily minimum and maximum temperatures are recorded and then averaged to determine the average minimum or maximum temperature for the month. In addition, air temperature is digitally recorded on a 5 min increment (CR10X, Campbell Scientific). These values are averaged and hourly maximum, minimum and average temperatures are stored. Weekly absolute maximum and minimum temperatures are recorded at all other climate stations with NWS maximum and minimum thermometers.

Total daily precipitation is collected by an 8 in. Standard Rain Gage (NWS). Rainfall volume and intensity are recorded by Recording Rain Gage (Belfort Universal Recording Rain Gage, Belfort Instrument Co., Baltimore, MD, USA). A network of nine Recording Rain Gages and 12 Standard Rain Gages are located throughout the basin (Table 2). (The use of trade or firm names in this publication is for reader information and does not imply endorsement by the US Department of Agriculture of any product or service.)

**Statistical models**

To describe the climate data, we present means, extremes, deviations from long-term means, and simple Pearson’s correlation coefficients (R) among climate variables and time. To test the hypotheses that mean, maximum and minimum annual air temperature (T, °C) has been increasing in the recent part of the record by fitting a time series intervention models to T data, method is described in detail in Ford et al. (2005). Candidate models were a simple level, or a mean level plus a linear increase starting at time t. Each potential starting time in the 1975–1988 range, which was the visual range of the temperature increase, was evaluated sequentially (PROC ARIMA, SAS v9.1, SAS Institute, Inc.). We computed Akaike’s
information criterion (AIC) for each model, which is a statistic used to evaluate the goodness of fit and parsimony of a candidate model, with smaller AIC values indicating a better fitting and more parsimonious model than larger values (Johnson & Omland 2004). We used the differences in the AIC values among candidate models with all starting times (Δ = AIC_i – AIC_min) to compute a relative weight (w_i) for each model relative to all models fit:

\[ w_i = \frac{e^{-0.5\Delta_i}}{\sum_{i=1}^{R} e^{-0.5\Delta_i}} \]  

(1) 

with the sum of all w_i equal to 1. The final model selected was the model with the highest w_i (Burnham & Anderson 2002; Johnson & Omland 2004).

We explored whether the high and low ends of the precipitation distribution were changing over time with quantile regression (Cade & Noon 2003). We analyzed linear trends in all quantiles of precipitation (P, mm) to quantify changes to the distribution of annual and monthly precipitation. We used data from the high- and low-elevation standard rain gages (SRG 19 and 31, Table 2) for the entire period of record. Our model predicted the precipitation amount as a function of year, with elevation as a covariate. All models were fit using PROC QUANTREG in SAS (v9.1, SAS Institute, Inc.). If the bootstrapped 95% confidence interval around the estimated coefficient for the quantile overlapped zero, we interpreted this as no significant time trend. To check whether annual precipitation totals from the two gages were consistent over time, we fit a linear model predicting precipitation from SRG 19 as a function of precipitation from SRG 31, year, and the interaction of year and SRG 31 (PROC GLM).

RESULTS

Temperature

Daily and seasonal temperatures do not fluctuate widely (Figure 2). Temperatures are most variable in the winter months. The warmest year on record occurred in 1999 (average maximum temperature of 21.5 °C). The coldest year on record was 1940 with a minimum of 4.1 °C.

Average annual, maximum and minimum air temperatures at the site have increased significantly relative to the long-term mean (Figure 3). The warming trend is apparent
in all temperature series (for growing season temperature analysis see Ford et al. (2011)), and appeared to begin in the late 1970s to late 1980s. The most parsimonious statistical models indicate a significant increase in average minimum temperatures beginning in 1976 (\(w_i = 0.16\)), with a rate of increase away from the 5.4 °C long-term mean of 0.5 °C per decade. This same rate of increase occurred in the average annual data series in 1981 (\(w_i = 0.13\)), and in the annual maximum temperature series in 1988 (\(w_i = 0.17\)). Average dormant and growing season temperatures also show this trend (data not shown).

### Precipitation

Annual precipitation at Coweeta is among the highest in the eastern USA, averaging 1,794 mm (CS01 station, Figure 4 (a)). The basin receives frequent, small, low-intensity storms in all seasons, with the wetter months in late winter and early spring. Fall months are drier (Figure 4 (b)), but generally have larger, more intense tropical storm activity. Elevation has a strong influence on precipitation amount (Figure 4(b)). For example, precipitation at 1,398 m in elevation is 32% (±6% SD) higher than that at 685 m. Although \(P\) increases with elevation at the site, \(P\) amounts among the rain gages are highly correlated (0.96 < \(R^2\) < 0.99), and this relationship is consistent over time (\(P\) vs. time \(R = 0.02\); no \(P\) by year interaction \(Pr = 0.73, F_{1,73} = 0.12\)).

Annual precipitation totals are also becoming more variable over time, with wetter wet years and drier dry years (Figure 5(a) and (b)). Coweeta experienced the wettest year on record in 2009 with a total of 2,375 mm. Only 2 years prior, in 2007, the driest year on record occurred with 1,212 mm. Low quantiles, ~10–20%, had a significant negative slope over time. Higher quantiles, 65–75%, had a significant positive slope over time. This indicates that the low and high ends of the annual precipitation distribution in the basin changed during the period of record. During the wettest years, not all months were wetter; and similarly, during the driest years, not all months were drier. Our results show that the summer months became drier over time, while the fall months became more wet (Figure 5(c)–(f)). In general, most quantiles describing July precipitation declined over time. In September, only the most extreme part (>85%) of the distribution increased over time due to an increase in high intensity, shorter duration storm events, such as tropical storms, as opposed to an increase in the number of storms per month. For example, the number of storms occurring in September did not increase over time (\(R = -0.01\), Figure 6(a)), but the percentage of September storms that fell above the 75th percentile (16.5 mm) appears to increase substantially in the latter part of the record (Figure 6(b)). Other fall months became wetter over time, mainly due to increases in the lower percentile storms (data not shown).

In addition to more intense precipitation, recent climate patterns trend toward more frequent periods of prolonged
drought (Figure 7) as inferred by comparing annual totals against the long-term mean. In addition, drought severity (accumulated deficit in precipitation over time) is increasing with time ($R = -0.35$, $t_{0.05, 37} = -2.29$, $Pr = 0.01$). Beginning with a severe drought in 1985, a 1,600 mm deficit in rainfall accumulated through 2008.

Figure 5 | Annual (a), July (c) and September (e) total precipitation from CS01 (see Table 1). Lines show the modeled qth quantile as a function of time. Slopes of all lines shown are significantly different than zero. Panels on right show the modeled estimates (symbols) of the quantile slope conditional on year, for annual (b), July (d) and September (f) precipitation totals. Bootstrapped upper and lower 95% confidence intervals also shown for parameter estimates (grey area).
DISCUSSION

Climate change

The increased mean annual air temperature (i.e. 0.5 °C per decade) observed since the early 1980s at Coweeta is consistent with other global and regional observations. In the observed climate records, globally, the 20 warmest years have all occurred since 1981 (Peterson & Baringer 2009). Across the USA, a significant warming trend in air temperature also began in the late 1970s to early 1980s (Groisman et al. 2007). This warming trend is predicted to continue; ensemble atmosphere-ocean general circulation models (AOGCMs) predict that by the early 21st century (2030), southeastern US air temperatures will increase at a rate of 0.5 °C per decade (IPCC 2007).

Causes for the increases in air temperature include both natural (i.e. surface solar radiation) and anthropogenic (e.g. ozone and CO₂ concentrations) sources, and potential interactions of sources (IPCC 2007). For example, as aerosols increase, surface solar radiation is reduced (i.e. global dimming), which decreases surface temperatures (Wild 2009). One explanation of the lack of significant warming in the 1950–1980 period, followed by the rapid warming in the last 25 years, is that the lower surface solar radiation in the former period masked the warming trend, while the higher surface solar radiation in the latter period apparently accelerated warming (Wild 2009). Although cause and effect is difficult to establish, the timing of increased surface solar radiation is coincident with the passage and implementation of the 1977 Clean Air Act (CAA) and the 1990 CAA Amendments in the USA, which have been quite effective at reducing anthropogenic aerosols (Streets et al. 2006). Similar to our hypothesis here, over Europe a 60% reduction in aerosols has been linked to the 1 °C increase in surface temperature since the 1980s due to impacts on shortwave and long-wave forcing (Philipona et al. 2009).

Precipitation is increasing in the southeast, similar to what we observed in the Coweeta Basin (Karl & Knight 1998; Groisman et al. 2004). These increases in precipitation have translated to increases in streamflow according to long-term US Geological Survey streamflow data (Karl & Knight 2009).
Recent trends in eastern US precipitation, and specifically those in the Coweeta Basin, have been linked to regular patterns in the North Atlantic Oscillation (Riedel 2006a, b). A trend in drier summers since the 1980s has occurred for the southeast (Groisman et al. 2004; Angert et al. 2005). Simultaneously, a trend in wetter fall months has also occurred (Groisman et al. 2004). Our findings in the Coweeta Basin are consistent with both of these larger-scale regional patterns. Whether the trend of increasing precipitation will continue for the region in a warmer, higher-CO2 scenario is uncertain. Most AOGCMs do not agree on the predicted change in direction of future precipitation for the southern Appalachians and southeast USA, e.g. wetter vs. drier (IPCC 2007).

Many regions of the USA have experienced an increased frequency of precipitation extremes, droughts and floods over the last 50 years (Easterling et al. 2000; Groisman et al. 2004; Huntington 2006; IPCC 2007). As the climate warms in most AOGCMs, the frequency of extreme precipitation events increases across the globe, resulting in an intensification of the hydrologic cycle (Huntington 2006). For example, the upper 99th percentile of the precipitation distribution is predicted to increase by 25% with a doubling of CO2 concentration (Allen & Ingram 2002). The lower end of the precipitation distribution is also predicted to change. Forecasts of the drought extent over the next 75 years show that the proportion of land mass experiencing drought will double from 15 to 50% (Burke et al. 2006). For example, with a doubling in the peak CO2 concentration, dry season precipitation is expected to decline irreversibly on average by 15% on most land masses (Solomon et al. 2009). The timing and spatial distribution of extreme precipitation events are among the most uncertain aspects of future climate scenarios, however (Karl et al. 1995; Allen & Ingram 2002). Our results show that the extremes of the annual precipitation distribution are increasing in magnitude, with recent increases in the frequency of drought, and wetter wet years and drier dry years. We have observed the most extreme precipitation changes in the fall months, with increases in intense rainfall in September in particular. This is partly associated with precipitation generated from tropical storm events. However, a wetter fall season is also being observed due to an increase in the low percentile rain events in November (Ford et al. 2011).

**Effects on forest function and health**

Observed changes in temperature and precipitation distributions that have occurred both locally and regionally have significantly affected forest function and health. The eastern USA has experienced an earlier onset in spring due to rising temperatures (Czikowsky & Fitzjarrald 2004), which has increased spring forest evapotranspiration (Czikowsky & Fitzjarrald 2004) and growth (Nemani et al. 2003; McMahon et al. 2010). However, the increase in spring growth has been largely offset by drier summers in the southeast (Angert et al. 2005), and areas with observed sustained increases in forest growth over time have been those in the northeast (McMahon et al. 2010) and at the highest elevations (Salzer et al. 2009), where temperatures are more limiting than water, and the tropics (Nemani et al. 2003), where radiation is the primary limiting factor.

Whether forest productivity is experiencing recent increases concomitant with temperature increase in the Coweeta Basin has not yet been reported. The effects of extreme events, most notably drought, have had a more dramatic effect on forest health and forest species composition than the trends in temperature. Native insect outbreaks, e.g. southern pine beetle (Dendroctonus frontalis), are triggered by drought. The successive droughts in the 1980s and late 1990s caused widespread southern pine beetle infestations in Coweeta watersheds, and throughout the southern Appalachians. As a result of these outbreaks, a decrease in pitch pine (Pinus rigida) stands and increased canopy gap area due to dead or dying snags (Clinton et al. 1993; Vose & Swank 1994; Kloeppel et al. 2003) occurred. The growth of eastern white pine (Pinus strobus) has also been significantly reduced by drought (Vose & Swank 1994; McNulty & Swank 1995).

Deciduous hardwood ecosystems have also been impacted by droughts. For example, accelerated mortality of oaks in the red oak group (especially Quercus coccinea) occurred during the successive droughts in the 1980s and late 1990s, and interestingly, larger trees were more vulnerable than smaller trees (Clinton et al. 1993). The earliest reports of drought in the area in the 1920s also noted that oak mortality was higher than other deciduous tree species (Hursh & Haasis 1931). Growth rate data for oaks
in the 1980–1990 period showed comparable growth rates during wet and dry periods, suggesting either deep rooting and access to stored water in the deep soils at Coweeta (Kloeppel et al. 2003), or highly conservative gas exchange in both wet and dry periods (Bush et al. 2008; Ford et al. 2010).

SUMMARY AND CONCLUSIONS

Analysis of 75 years of climate data at the Coweeta Hydrologic Laboratory has revealed a significant increase in air temperatures since the late 1970s, an increase in drought severity and frequency, and a more extreme precipitation distribution. Cause and effect are difficult to establish, but these patterns are consistent with observations throughout the southeastern USA that suggest linkages between reduced aerosols and temperature patterns, and linkages between the North Atlantic Oscillation and increased precipitation variability. Similar patterns are predicted with AOGCMs under climate change and this recent variability in the observed record may be providing a glimpse of future climatic conditions in the southern Appalachian. Based on observed ecosystem responses to climate variability over the past 20 years, we anticipate significant impacts on ecosystem structure and function.

Climate change during the 21st century is predicted to include novel climates – combinations of seasonal temperature and precipitation that have no historical or modern counterpart (Williams et al. 2007). In the USA, the southeastern region is predicted to be the most susceptible to novel climates (Williams & Jackson 2007; Williams et al. 2007). Detecting ecosystem change, including ecological ‘surprises’, will require long-term data from monitoring networks and studies (Lindenmayer et al. 2010), such as those presented here from the Coweeta Hydrologic Laboratory.

ACKNOWLEDGEMENTS

This study was supported by the United States Department of Agriculture Forest Service, Southern Research Station, and by NSF grants DEB0218001 and DEB0823293 to the Coweeta LTER program at the University of Georgia. Any opinions, findings, conclusions, or recommendations expressed in the material are those of the authors and do not necessarily reflect the views of the National Science Foundation or the University of Georgia. We acknowledge the support of many individuals, past and present, including Wayne Swank, Charlie L. Shope, Charles Swafford, Neville Buchanan, ‘Zero’ Shope, Bryant Cunningham, Bruce McCoy, Chuck Marshall, Mark Crawford, and Julia Moore as well as the long-term climate and hydrologic data network at the Coweeta Hydrologic Lab.

REFERENCES


First received 18 April 2011; accepted in revised form 18 August 2011. Available online 22 February 2012