Functional response of U.S. grasslands to the early 21st-century drought

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Abstract. Grasslands across the United States play a key role in regional livelihood and national food security. Yet, it is still unclear how this important resource will respond to the prolonged warm droughts and more intense rainfall events predicted with climate change. The early 21st-century drought in the southwestern United States resulted in hydroclimatic conditions that are similar to those expected with future climate change. We investigated the impact of the early 21st-century drought on aboveground net primary production (ANPP) of six desert and plains grasslands dominated by C4 (warm season) grasses in terms of significant deviations between observed and expected ANPP. In desert grasslands, drought-induced grass mortality led to shifts in the functional response to annual total precipitation (PT), and in some cases, new species assemblages occurred that included invasive species. In contrast, the ANPP in plains grasslands exhibited a strong linear function of the current-year PT and the previous-year ANPP, despite prolonged warm drought. We used these results to disentangle the impacts of interannual total precipitation, intra-annual precipitation patterns, and grassland abundance on ANPP, and thus generalize the functional response of C4 grasslands to predicted climate change. This will allow managers to plan for predictable shifts in resources associated with climate change related to fire risk, loss of forage, and ecosystem services.

Key words: climate change; desert; extreme events; grassland production; invasive species; plains; precipitation variability; resilience; warm drought.

INTRODUCTION

General circulation models predict that climate change will result in an unprecedented concurrence of regional drying and warming (e.g., Seager et al. 2007). Further, droughts induced by climate change are predicted to be spatially expansive and temporally synchronous (Weiss et al. 2009). These droughts are also predicted to include novel precipitation patterns characterized by more intense rainfall events and longer dry interstorm intervals (Easterling et al. 2000, Solomon et al. 2007), potentially inducing dramatic region-wide changes in ecosystem productivity. In the western United States, recent episodes of drought-induced perennial plant mortality in forests (Allen et al. 2010), woodlands (Breshears et al. 2005), shrublands (McAuliffe and Hamerlynck 2010), and grasslands (Scott et al. 2010) have been attributed to warmer prevailing temperatures that exacerbated drought conditions. Such mortality and potential subsequent development of novel species assemblages (Williams and Jackson 2007, Willis and Bhagwat 2009) could have significant implications for the relationship between annual precipitation and ecosystem productivity.

North American grassland ecosystems are expected to be sensitive to global climate change because they have the capacity for large and rapid responses of production to annual precipitation (Sala et al. 2000, Knapp and Smith 2001, Suttle et al. 2007, Reichmann et al. 2013). However, there is uncertainty how grassland production will respond to new hydroclimatic conditions because there is still no consensus on the underlying mechanisms driving these responses even though individual grassland sites have been studied for long-term responses, and these sites have been compared for relative sensitivities (Epstein et al. 2002, Morgan et al. 2011). As a result, there are few generalizations about the regional behav-
ior of grasslands with prolonged warm drought and more extreme precipitation events (Peters et al. 2004).

Studies of grassland productivity in arid and semi-arid ecosystems are generally based on the response of aboveground net primary production (ANPP or total new biomass produced aboveground during a specific interval) to variability in total annual precipitation (P<sub>T</sub>). There is a well-accepted relation in North America between mean annual precipitation (MAP, the long-term average of P<sub>T</sub>) and ANPP across sites from arid to mesic ecosystems, termed the spatial model (Sala et al. 1988b, Knapp and Smith 2001, Huxman et al. 2004), and there is some understanding about how this spatial model will respond to climate change (Sala et al. 2012, Ponce-Campos et al. 2013). However, the cross-site spatial model differs substantially from the within-site temporal model (e.g., Huxman et al. 2004, Sala et al. 2012), and there is no consensus on the biogeochemical, physiological, or community-level mechanisms that may explain this disparity (Lauenroth and Sala 1992, Muldavin et al. 2008). Understanding this distinction is important because it is the temporal model that will be used to manage grassland resources and to plan for shifts in functional processes associated with climate change.

Although there is general agreement that the temporal model is contingent upon the life history of organisms, geography, and limitations of nutrients, light, and water, there is little consensus on the underlying mechanisms driving the regional patterns because most temporal models have been developed for individual sites. Knapp and Smith (2001) reported a temporal model for grasslands where the pulses in ANPP in years of above-average precipitation were larger than declines in ANPP during years with below-average precipitation. They attributed this asymmetry to buffering mechanisms that reduced the impact of drought. Using largely the same data set, Huxman et al. (2004) reported that temporal models for single sites followed a pattern where rainfall use efficiency (RUE is ANPP/P<sub>T</sub>) increased from high- to low-production biomes because resources other than soil water became limiting in the high-production biomes. With a data set limited to those low-production–high-RUE sites (subhumid grass and shrub-dominated rangelands), Sala et al. (2012) reported that the temporal model was best characterized by the current-year precipitation and the previous-year ANPP because of time lags in response associated with meristem density at the plant scale and grass abundance at the plot scale. Their results suggest that in such rangeland systems, the response of ANPP to long-term drought would increase with time as the lag-effect continued, ultimately leading to functional and structural changes as hypothesized by Smith et al. (2009). Ponce-Campos et al. (2013) suggested that with prolonged warm drought, grassland biomes would reach a maximum RUE, approaching a threshold which, when crossed, would result in biome reorganization. The concurrence of larger storms and longer dry intervals has been reported to decrease ANPP in intermountain desert grasslands (Zhang et al. 2013) and may increase ANPP in Great Plains grasslands (Heisler-White et al. 2009, Cherwin and Knapp 2012).

This ecological complexity has made regional generalizations of temporal models of the ANPP-to-P<sub>T</sub> relation difficult. Our objective was to develop a temporal model of the relation between ANPP and P<sub>T</sub> for arid and mesic regions of the C<sub>4</sub> grasslands of the southwestern United States. Deviations between observed and expected ANPP were used to determine changes in function related to the long-term ANPP-to-P<sub>T</sub> relation.

We focused on grasslands of the United States for several reasons. First, a basic goal of ours was to provide a climate change context to the temporal model. Recent hydroclimatic conditions in the United States provided an opportunity to study the functional response of biomes dominated by C<sub>4</sub> grasses to warm drought and extreme precipitation patterns. In the southwest United States, the early 21st-century drought has resulted in regional annual precipitation 25% below 20th-century means and an average annual temperature 0.8°C warmer than the 20th-century mean; these conditions are similar to those expected with climate change (MacDonald 2010). Second, the continental United States is characterized by two distinct regions supporting extensive grasslands: the Great Plains and the intermountain desert (hereafter referred to as plains and desert grasslands). This allowed us to investigate the impact of life history on the temporal model related to climate and grazing pressures. Third, our research was intentionally designed to utilize in situ knowledge to understand the mechanics behind site-specific behavior. Within this region, there are six USDA experimental sites with continuous research and data collection over the past 20–100 years. We combined these case studies to understand the mechanisms underlying deviations from predicted ANPP-to-P<sub>T</sub> relations and make regional generalizations that could apply to similar grasslands around the world.

**Materials, Methods, and Models**

**Study sites**

We focused on six USDA long-term experimental sites during the period from 2000 to 2011 (Fig. 1). These study sites encompass a range of precipitation and temperature regimes and are dominated by C<sub>4</sub> grasses (Table 1). Locations include three arid intermountain desert grasslands (mean annual precipitation [MAP] <350 mm/yr), Desert Experimental Range (DER) in Utah, Jornada Experimental Range (JRN) in New Mexico, and Walnut Gulch Experimental Watershed (WGE) in Arizona; and three mesic Great Plains grasslands (MAP ≥ 350 mm/yr), Central Plains Experimental Range (CPL) in Colorado and Southern Plains Experimental Range (SPL) and Little Washita River...
Experimental Watershed (LWA) in Oklahoma. Site-level experts were available at each experimental site to interpret and validate results based on long-term records.

These grassland ecosystems in the continental United States need to be set in context of the historical evolution across North America that has resulted in regional structure and function. North American grasslands occur in transitional climate regions where prolonged dry periods favor perennial grass persistence; thus most plants are tolerant of short-term drought, but the frequency of species with tolerance to severe, long-

**FIG. 1.** Locations of the six grassland sites across the southwest United States, overlain on an image of regional Palmer drought severity index (PDSI) over the NOAA/NCDC polygons for year 2003 during the early 21st-century drought (from http://www7.ncdc.noaa.gov/CDO/cdo). CPL is Central Plains Experimental Range, DER is Desert Experimental Range, SPL is Southern Plains Experimental Range, LWA is Little Washita River Experimental Watershed, JRN is Jornada Experimental Range, and WGE is Walnut Gulch Experimental Watershed.

**TABLE 1.** Characteristics of the six grassland sites, including the dominant grasses at the beginning of the decade (year 2000), mean annual sum of precipitation (MAP), and mean annual maximum temperature (MAT) with standard deviations in parentheses.

<table>
<thead>
<tr>
<th>Site name and location</th>
<th>Dominant grasses in year 2000</th>
<th>MAP (mm)</th>
<th>MAT(°C)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intermountain desert grasslands</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DER, Desert Experimental Range, Utah</td>
<td>(C_4): Sporobolus cryptandrus</td>
<td>179 (58)</td>
<td>20 (0.8)</td>
</tr>
<tr>
<td></td>
<td>(C_3): Oryzopsis hymenoides</td>
<td></td>
<td></td>
</tr>
<tr>
<td>JRN, Jornada Experimental Range, New Mexico</td>
<td>(C_4): Bouteloua eriopoda;</td>
<td>241 (73)</td>
<td>25 (0.8)</td>
</tr>
<tr>
<td></td>
<td>Sporobolus flexuosa</td>
<td></td>
<td></td>
</tr>
<tr>
<td>WGE, Walnut Gulch Experimental Watershed,</td>
<td>(C_4): Bouteloua eriopoda;</td>
<td>305 (91)</td>
<td>25 (0.9)</td>
</tr>
<tr>
<td>Arizona</td>
<td>Bouteloua curtipendula</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Great Plains grasslands</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>CPL, Central Plains Experimental Range,</td>
<td>(C_4): Bouteloua gracilis</td>
<td>381 (91)</td>
<td>16 (1.5)</td>
</tr>
<tr>
<td>Colorado</td>
<td>(C_4): Hesperostipa comata</td>
<td></td>
<td></td>
</tr>
<tr>
<td>SPL, Southern Plains Experimental Range,</td>
<td>(C_4): Bouteloua curtipendula</td>
<td>587 (165)</td>
<td>22 (0.9)</td>
</tr>
<tr>
<td>Oklahoma</td>
<td>Andropogon hallii</td>
<td></td>
<td></td>
</tr>
<tr>
<td>LWA, Little Washita River Experimental</td>
<td>(C_4): Schizachyrium scoparium</td>
<td>794 (197)</td>
<td>24 (1.0)</td>
</tr>
<tr>
<td>Watershed, Oklahoma</td>
<td>Andropogon gerardii</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Note:** Averages represent the 32-year period 1980–2011 for all sites except DER (18-year period 1994–2011). All sites are in the United States.
term drought increases as MAP decreases (McPherson 1997, McCarron and Knapp 2001, Peltzer and Köchy 2001, Briggs et al. 2005). Plains grasslands are dominated by sod-forming bunchgrasses capable of tolerating prehistoric and historic grazing pressure, especially by large ungulates (Collins et al. 1998, Lyman and Wolverton 2002). This region encompasses mesic tallgrass prairie, mixed-grass prairie, and semiarid shortgrass steppe and follows a declining east–west precipitation gradient with similar seasonal rainfall distribution peaking in the summer when temperatures are warmest (Borchert 1950). In contrast, desert grasslands are characterized by widely spaced, drought-tolerant bunchgrasses, and the region did not historically support high densities of large ungulate populations (McClaran 1997, Lyman and Wolverton 2002). These arid grasslands experience lower, more variable interannual rainfall, with distinct summer and winter peaks that reflect the relative strengths of North American monsoon and frontal systems influenced by El Niño Southern Oscillation (ENSO) and Northern Annular Mode dynamics (Sheppard et al. 2002, McAfee and Russell 2008).

Meteorological data

Long-term in situ meteorological measurements were used to derive climatic variables to characterize the hydroclimatic conditions. The \( P_T \) was computed as a sum of daily precipitation based on the water year (1 October–30 September). The Palmer drought severity index (PDSI) was computed from measurements of monthly precipitation and temperature from 1960 to 2011, where negative and positive values indicated dry and wet conditions, respectively (Dai 2011). For the DER site in Utah, the record of continuous in situ meteorological measurements was only available from 1994 to present, and consequently, PDSI was computed using extended records at nearby NOAA stations.

National networks with instrumentation near the study sites provided additional information. The NOAA National Climate Data Center (NCDC) provided regional annual PDSI values from 1900 to present based on measurements that included monthly precipitation and average air temperature and estimates of local soil available water content. These data were used to compare the 1930s dust bowl drought, the 1950s drought that affected much of the southwest and southern Great Plains of the United States, and the contemporary early 21st-century drought.

We defined a hydroclimatic index based on PDSI to be used as a predictor of ecosystem functional response to drought based on findings that the number of consecutive months of drought extending over multiyear periods was a good predictor of plant mortality (McAuliffe and Hamerlynck 2010). Since monthly PDSI was computed based on conditions during the prior 12 months, we considered the PDSI in September (PDSI\(_{Sep}\)) to be an indicator of the consecutive months of drought weighted more heavily on the growing season of that water year. The distribution of PDSI\(_{Sep}\) during the late 20th century (1980–1999) was used to determine the values associated with an extreme wet year (in top fifth percentile), a wet year (in top 20th percentile), a warm drought year (in bottom 20th percentile), and an extreme drought year (in bottom fifth percentile). That is, the bottom fifth percentile translates to the driest year in the 20-year record (1980–1999), and similarly, the bottom 20th percentile translates roughly to the fourth driest year.

To characterize the seasonal precipitation pattern, we used an R95 index that was proposed by Frich et al. (2002) and has been adopted as a standard output in the Intergovernmental Panel on Climate Change (IPCC) AR2 (second assessment report; Solomon et al. 2007). R95 is the annual precipitation amount due to daily precipitation exceeding the 95th percentile of the full temporal record (1970–2011). To make this value comparable across sites along a precipitation gradient, we normalized R95 by \( P_T \) to derive R95%. The index R95% is the fraction of \( P_T \) due to the events above the 95th percentile and provides a standardized annual index of the frequency of large storms.

Aboveground net primary production (ANPP)

Many studies of cross-site ecosystem production are based on a compilation of in situ measurements of ANPP from long-term experimental sites across the United States and around the world. This approach can lead to some uncertainty because procedures are not consistent across sites, and in some cases, not consistent over time at a given site (Sala et al. 1988, Biondini et al. 1991). Further, the general approach for estimating ANPP for grasslands is to schedule the destructive sample for dates associated with peak biomass.

Instead, we used satellite observations of the enhanced vegetation index (EVI) from NASA’s moderate-resolution imaging spectroradiometer (MODIS) as a surrogate for annual ANPP. At each location, a site of \( \sim 3 \times 3 \) km was chosen where vegetation was dominated by C\(_4\) grasses and in situ meteorological records representative of the location could be obtained. MODIS EVI data were extracted for this location, where

\[
EV = \frac{\rho_{\text{NIR}} - \rho_{\text{red}}}{\rho_{\text{NIR}} + C_1 \rho_{\text{red}} - C_2 \rho_{\text{blue}} + L}
\]

and \( \rho_{\text{NIR}} \), \( \rho_{\text{red}} \), and \( \rho_{\text{blue}} \) are atmospherically corrected surface near-infrared, red, and blue reflectance bands, respectively; \( G \) is a gain factor; \( C_1 \) and \( C_2 \) are the coefficients of the aerosol resistance term; and \( L \) functions as the soil-adjustment factor (Huete et al. 2002). Values of EVI were averaged over an area of 2.25 \( \times \) 2.25 km (\( 9 \times 9 \) MODIS pixels) extracted from each MODIS scene (23 per year \( \times \) 10 years = 230 scenes) representing the dominant vegetation type at the site. Using the software TimeSat (Jönsen and Eklundh...
2004), we applied pixel-based quality control to remove noise from the time series and smoothed the time series to integrate the EVI over the active growing season (A in Fig. 2a; hereafter referred to as the integrated EVI or iEVI).

The MODIS iEVI has been used to quantify the dynamics of ANPP across biomes ranging from arid grassland to forest (Ponce-Campos et al. 2013, Zhang et al. 2013). For this study, plot-scale measurements of ANPP at JRN and CPL sites during the 2000–2011 study period (Morgan et al. 2011, Peters et al. 2012) were compared with iEVI measurements. Though there are scale differences between measurements of iEVI and ANPP (i.e., ANPP measurements are commonly made at scales from 1 m² to 0.01 km² vs. iEVI at ~2 × 2 km), ground-based measures of ANPP were significantly related to iEVI for these select grasslands over the time period 2000–2011 in a log-log relation ($r^2 = 0.71$ and 0.91, $P < 0.01$; Fig. 2b). The log-log relation accounted for the uneven distribution of ANPP estimates over time.

A simple test was conducted to compare in situ ANPP estimates made near the peak of the growing season to iEVI values integrated over the entire growing season. At CPL, it is the protocol to make ANPP destructive samples in the first week of August. We integrated the EVI from the beginning of the growing season to the first week in August (rather than the full water year) to be comparable to the time of the CPL estimate. The log-log relation improved from $r^2 = 0.71$ (Fig. 2b) to $r^2 = 0.82$ with this synchronization of time periods. This result implies that the iEVI may provide more temporal stability, and less uncertainty, than in situ estimates of annual ANPP associated with destructive samples scheduled to coincide with the elusive date of peak greenness. Thus, we assumed that the iEVI was a reasonable surrogate for ANPP interannual variability at all six sites in our analyses, and the terms iEVI and ANPP are used interchangeably in further discussion.

Standardized values of $P_T$ and iEVI were computed for intersite comparison ($P_{TS}$ and iEVI$_S$), as the deviation of the $i$-year value from the 11-year average in units of standard deviation ($\sigma$), i.e.,

$$P_{TS} = (P_{Ti} - \overline{P_T})/\sigma_{P_T}$$

and

$$i\text{EVI}_S = (i\text{EVI}_i - \overline{i\text{EVI}})/\sigma_{\text{EVI}}.$$  \hspace{1cm} (2)

Differences greater than 1σ between iEVI$_S$ and $P_{TS}$ for any given year indicated an anomaly in the functional response of the grassland production to precipitation and a potential indication of impact of the warm drought conditions in the early 21st century.

**Temporal model**

Most of the previous studies have focused on the spatial model of ANPP and $P_T$ for MAP ranging from ~100 to ~3000 mm/yr. This is summarized here and interpreted for application to a temporal model for grasslands. There is general agreement that the spatial model of ANPP and $P_T$ is exponential, with a form $\text{ANPP} = b_0(1 - e^{b_1P_T})$ (Huxman et al. 2004, Ponce-Campos et al. 2013). Zhang et al. (2013) proposed a multiple nonlinear regression model to account for the significant effects of extreme precipitation patterns, where $i\text{EVI} = (b_0 + b_1\text{R95}\%)(1 - e^{b_2P_T})$. They found that R95% and $P_T$ together explained 88% of the variance in observed ANPP across biomes from desert grassland to temperate forest. The temporal model for grassland sites with MAP <1000 mm has been approximated with a linear fit, where $\text{ANPP} = b_0 + b_1P_T$ (e.g., Briggs and Knapp 1995). For arid to
subhumid grass- or shrub-dominant systems (i.e., rangelands), Sala et al. (2012) suggested that ANPP was best characterized by the current-year precipitation ($P_{T(t)}$, where $t$ is the current year) and the previous-year ANPP ($ANPP_{(t-1)}$), where $ANPP = b_0 + b_1P_{T(t)} + b_2ANPP_{(t-1)}$. Though no correlation coefficient was provided for this unified model, the logic that previous-year ANPP controls a significant fraction of current-year production was discussed in Introduction.

An equivalent set of relations specific to this study of six grassland sites was derived by (1) substituting the linear temporal ANPP-to-$P_T$ relation in place of the exponential spatial relation and (2) using $iEVIS_5$ as a proxy for ANPP. As a result

$$iEVIS_5 = b_0 + b_1P_{TS(t)}$$

from Briggs and Knapp (1995);

$$iEVIS_5 = b_0 + b_1P_{TS(t)} + b_2iEVIS_{(t-1)}$$

from Sala et al. (2012); and

$$iEVIS_5 = b_0 + b_1P_{TS(t)} + b_2R95\%$$

from Zhang et al. (2013).

Combining these relations, it is reasonable to propose that

$$iEVIS_5 = b_0 + b_1P_{TS(t)} + b_2iEVIS_{(t-1)}b_3R95\%.$$  \hspace{1cm} (6)

We fit models for all possible combinations of predictors (Eqs. 3–6) to data for the three combined desert grassland sites and to data for the three combined plains grassland sites. This provided the expected response of $iEVIS_5$ to model parameters for detection of significant deviations between observed and expected ANPP. We assessed models based on Akaike’s information criterion adjusted for small sample sizes (AICc) to evaluate the benefit of increasing model complexity from Eq. 3 to Eq. 6 (Burnham and Anderson 2002). We selected models of greater complexity than Eq. 3 when the inclusion of an additional model parameter reduced AIC by more than 2.0 (Burnham and Anderson 2002). We also report regression correlation coefficients ($r^2$) as an absolute measure of model fit. We note that intercepts for all models were close to zero such that including intercepts did not produce more parsimonious models ($\Delta$AIC <2 for models with vs. without intercepts in all cases); hence, intercepts were not included in the models.

RESULTS

Characteristics of the early 21st-century drought

During the early 21st century, these six grasslands experienced prolonged drought with warm season temperatures higher than any period in the 20th century, representing new hydroclimatic conditions (Fig. 3). Low precipitation during the early 21st-century drought was comparable to that during the 1930s dust bowl drought. However, temperature deviations from the historic mean during the early 21st-century drought were nearly double those reported for the 1930s drought, rising from 0.2° to 0.8°C during the cold season and from 0.6° to 1.1°C during the warm season. The decrease in precipitation in the 1950s drought was greater than the decrease in the 1930s and early 21st-century droughts, but the temperature during the warm season was only 0.6°C above the historic mean.

During the early 21st-century drought, the three desert grasslands (DER, JRN, and WGE) and two of the three plains grasslands (CPL and LWA) recorded warm drought periods ranging from five to eight years, and all sites recorded two or more consecutive warm drought years (Fig. 4). All sites recorded at least one extreme warm drought year, and CPL recorded six consecutive extreme warm drought years, where extreme warm drought years were within the bottom fifth percentile and thus drier than the driest year between 1980–1999. For all sites, there were only three years (i.e., 2006–2008 at JRN) when PDSI$_{Sep}$ fell within the top 20th percentile and one year when PDSI$_{Sep}$ fell within the top fifth percentile of the 1980–1999 distribution (i.e., 2011 at DER). Warm season temperatures during the most extreme drought years exceeded the 12-year (2000–2011) average by values ranging from 0.4° to 1.7°C for the desert grasslands and 1.3° to 2.5°C for the plains grasslands.

The R95% was used as an index of the number of large storms in each year standardized to the long-term average for each site, respectively. We used the Hartigan’s dip statistic (HDS) to detect the presence of bimodality in the data, and this yielded the mean and standard deviation of each distribution (Fig. 5). There was a shift in the peaks of this mixed-density function toward larger storms (higher R95%) in the 2001–2011 time period, determined with one-way analysis of variance. The mean of the leftmost distribution shifted from 8.6 to 13.1 for the desert grasslands and 13.1°C above the historic mean.

A temporal model for C$_4$ grasslands during altered hydroclimatic conditions

For desert grasslands, the model based on $P_{TS(t)}$ alone (AIC = 67.1) explained 60% of the variance in $iEVIS_5$. More complex models that included $iEVIS_{(t-1)}$ and R95% (AIC = 73.1) or R95% alone (AIC = 74.8) were considerably less parsimonious (AAIC = 7.3 and 9.0; Table 2). A model that included $iEVIS_{(t-1)}$ plus $P_{TS(t)}$ reduced AIC by 1.3 relative to the $P_{TS(t)}$ model. Because the small increase in AIC (AAIC <=2) came at the cost of an additional model parameter, we rejected this model and retained the simplest model that predicted $iEVIS_5$ only on the basis of $P_{TS(t)}$ (Table 2), where

$$iEVIS_5(\text{Desert}) = 0.7772\left(P_{TS(t)}\right).$$  \hspace{1cm} (7)

For plains grasslands, the model based on $P_{TS(t)}$ alone (AIC = 50.0) explained 75% of the variance in $iEVIS_5$. A
model including both \( P_{TS(t)} \) and \( iEVIS(t/C0) \) was more parsimonious (\( DAIC = -3.7 \)) and explained 79% of the variance in \( iEVIS \). The selected model was

\[
iEVIS(\text{Plains}) = 0.8811 \left( P_{TS(t)} \right) - 0.1522 \left( iEVIS(t/C0) \right).
\]  

(8)

Models including R95% were considerably less parsimonious (\( DAIC > 8.6 \)). The intra-annual precipitation pattern (larger storms and longer interstorm dry periods) had only a secondary impact on ANPP and that impact was greater for desert than for plains grasslands (Table 2; and Zhang et al. 2013). To achieve model parsimony, that is, the desire to explain phenomena using fewer parameters, models including R95% were not considered for these grasslands despite potential improvements in explaining \( iEVIS \) variance.

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The residuals of the selected models (Eqs. 7 and 8) were symmetric across the precipitation gradient (Fig. 6); however, the most striking pattern in the residuals is related to the outliers. After three to four near-consecutive years of warm drought, residuals for desert grasslands (DER, JRN, and WGE) were greater than \( 1 \sigma \) or less than \(-1 \sigma \). Whereas, the \( iEVIS \) for plains grasslands (CPL, SPL, and LWA) corresponded well with modeled values for most years (within \( \pm 1 \sigma \)) and did not exhibit the dramatic anomalies observed for desert grasslands. Thus, the root mean squared error (RMSE) of the difference between measured and modeled \( iEVIS \) was lower for plains grasslands than for desert grasslands (0.46 and 0.6, respectively).

Of the three outliers in the \( iEVIS/P_{TS} \) relationship for desert grasslands, two were associated with plant mortality during prolonged drought, and one involved a lag in the response of ANPP to \( P_T \) associated with changes in species dominance (Fig. 7; Table 3). After three near-consecutive years of warm drought (2002–2003 and 2005), the native grasses at WGE experienced near 100% mortality (Scott et al. 2010). In 2006, WGE was dominated by opportunistic forbs with ANPP greater than expected if produced by native grasses (\( iEVIS > 1 \sigma \)). This was concurrent with the spread of the exotic South African grass \textit{Eragrostis lehmanniana} (Lehmann lovegrass), which then attained dominance by 2007. This near 100% mortality of native grasses and replacement with forbs and invasive perennial grass was associated with extreme warm drought (Fig. 3).
Similarly, the unexpectedly high iEVIS relative to P$_{TS}$ at DER in 2010 was associated with an increase in the introduced annual weed *Salsola* sp. (Russian thistle) documented during ANPP measurements (Fig. 7a). After two consecutive years of extreme warm drought at JRN (2002–2003), ANPP decreased in 2004 by 70% of the predrought mean despite greater than average precipitation (Peters et al. 2012), followed by an increase in the production of annuals (2005 and 2006) and an increase in the production of the dominant native grasses by 2008. A similar lag in the response of ANPP to $P_T$ was observed at DER in the postdrought years 2005–2006 after five consecutive years of warm drought (Fig. 7a and 7b). No lags nor new species assemblages were observed in plains grasslands over the study period (Table 3). Though CPL and LWA sites experienced similarly prolonged warm drought, the ANPP remained primarily responsive to $P_T$ (Fig. 7d and 7f) and secondarily to previous year ANPP (Eq. 8).
DISCUSSION

Results illustrated the regional variation and complex functional response of grassland production to extreme conditions that are analogous to predicted climate change and the importance of multisite experiments for making generalizations. After only a few years of warm drought, desert grasslands entered a long recovery period that shifted the expected ANPP response to \( P_T \) for several years, and in some cases, resulted in changed community composition. Though grasses in these arid regions are well adapted to persist through typical seasonal dry periods and episodic, chronic drought (McClaran 1997), this series of warm drought years disrupted the typical productivity response to \( P_T \). Plains grasslands exposed to prolonged warm drought responded strongly to both wet and drought conditions (Fig. 6b), a dynamic that confirms the linear temporal model. This led to linear models (Eqs. 7 and 8) across all sites and years that improved the prediction of ANPP, with anomalies attributed to changes in dominant species and lags in the response of ANPP to \( P_T \) in desert grasslands (Fig. 7; Table 3).

This was a first attempt to model the response of ANPP to the interdependent factors of current-year precipitation, intra-annual rainfall distribution, and previous-year ANPP in a natural setting. We confirmed that ANPP responded primarily to current-year precipitation in both desert and plains grasslands, and to previous-year ANPP only in the mesic plains grassland. When all factors covaried, the impact of larger storms on grassland ANPP was minimized, in contrast to studies that reported a significant impact of larger storms and longer dry periods on grassland ANPP when precipitation amount was held constant in a manipulated experiment (e.g., Heisler-White et al. 2009) or when effects of precipitation patterns were isolated from effects of \( P_T \) (e.g., Zhang et al. 2013). Interestingly, the influence of previous-year ANPP in mesic plains grasslands did not lead to the type of anomalous ANPP-to-\( P_T \) relationship that might be expected with chronic drought (Smith et al. 2009). Instead, the anomalous behavior associated with lags, mortality, and new species assemblages was limited to prolonged droughts in the arid desert grasslands, where previous-

TABLE 2. Comparison of models predicting current-year ANPP (integrated standardized enhanced vegetation index, iEVIS) as a function of standardized current-year \( t \) precipitation (\( P_{TSO}(t) \)), standardized previous-year ANPP (iEVIS\(_{S(t-1)} \)), and/or an index of large precipitation events (R95%).

<table>
<thead>
<tr>
<th>Model</th>
<th>AIC</th>
<th>AAIC</th>
<th>( r^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Desert grasslands</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( P_{TSO} )</td>
<td>64.5</td>
<td>1.1</td>
<td>0.60</td>
</tr>
<tr>
<td>( P_{TSO} ), iEVIS(_{S(t-1)} )</td>
<td>63.4</td>
<td>0.0</td>
<td>0.56</td>
</tr>
<tr>
<td>( P_{TSO} ), R95%</td>
<td>73.5</td>
<td>10.1</td>
<td>0.61</td>
</tr>
<tr>
<td>( P_{TSO} ), iEVIS(_{S(t-1)} ), R95%</td>
<td>72</td>
<td>8.6</td>
<td>0.57</td>
</tr>
<tr>
<td>Plains grasslands</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( P_{TSO} )</td>
<td>50.0</td>
<td>3.7</td>
<td>0.75</td>
</tr>
<tr>
<td>( P_{TSO} ), iEVIS(_{S(t-1)} )</td>
<td>46.3</td>
<td>0.0</td>
<td>0.79</td>
</tr>
<tr>
<td>( P_{TSO} ), R95%</td>
<td>58.9</td>
<td>12.6</td>
<td>0.75</td>
</tr>
<tr>
<td>( P_{TSO} ), iEVIS(_{S(t-1)} ), R95%</td>
<td>55.6</td>
<td>9.3</td>
<td>0.83</td>
</tr>
</tbody>
</table>

**Notes:** Selected models are shown in boldface. AIC is Akaike’s information criterion. Where \( AAIC \) values for two competing models differed by less than 2, we selected the model with the fewest parameters. \( R95 \) is the annual precipitation amount due to daily precipitation exceeding the 95th percentile of the full temporal record (1970–2011). To make this value comparable across sites along a precipitation gradient, we normalized R95 by \( P_T \) to derive R95\%. The index R95\% is the fraction of \( P_T \) due to the events above the 95th percentile and provides a standardized annual index of the frequency of large storms.
Fig. 6. Relation between measured and modeled standardized integrated enhanced vegetation index (iEVIₘ) based on Eqs. 7 and 8 for desert (DER, JRN, and WGE; mean annual precipitation [MAP] <350 mm/yr) and plains (CPL, SPL, and LWA; MAP ≥350 mm/yr) grasslands, respectively. Sites and years for which the difference between measured and modeled iEVIₘ exceeded ±1σ (±1 SD) are labeled. RMSE is the root mean squared error of the difference between measured and modeled iEVIₘ and standardized total precipitation, Pₜₛ. Location site abbreviations are defined in Fig. 1.

Fig. 7. Trends in standardized iEVI and Pₜ (iEVIₘ and Pₜₛ, respectively) for each grassland site. Black lines represent iEVIₘ < Pₜₛ (ANPP is lower than expected for the given precipitation, see anomalous negative residuals in Fig. 6), and gray lines represent iEVIₘ > Pₜₛ (ANPP is higher than expected for the given precipitation, see anomalous positive residuals in Fig. 6). The thicker lines indicate a larger deviation (in units of standard deviation). The years with PDSIₛₑₚ below the 20th percentile of the 1980–1999 distribution (from Fig. 4) are represented by dark gray vertical bars. Arrows point to years of documented lag in the response of ANPP to Pₜ (Lag) and new species assemblages (NSA; from Table 3).
Table 3. Case studies at long-term experimental sites explaining the mechanisms driving the site-by-site, year-by-year relation between standardized iEVI and standardized total precipitation $P_T$ (iEVI$_{TS}$ and $P_{TS}$), where the number of consecutive months with PDSI < 0 and number of consecutive years with PDSI in September (PDSI$_{Sep}$) < 20th percentile express the prolonged warm drought.

<table>
<thead>
<tr>
<th>Site</th>
<th>Ecosystem</th>
<th>PDSI &lt;0 (consecutive months)</th>
<th>PDSI$_{Sep}$ &lt;20th percentile (consecutive years)</th>
<th>Observed functional response</th>
<th>In situ knowledge</th>
</tr>
</thead>
<tbody>
<tr>
<td>DER</td>
<td>desert</td>
<td>62</td>
<td>5</td>
<td>Lag. 2005</td>
<td>S. G. Kitchen, field notes</td>
</tr>
<tr>
<td>DER</td>
<td>desert</td>
<td>97 of 124</td>
<td>8 of 10</td>
<td>NSA, 2010</td>
<td>S. G. Kitchen, field notes</td>
</tr>
<tr>
<td>WGE</td>
<td>desert</td>
<td>78 of 97</td>
<td>3 of 4</td>
<td>NSA, 2006–present</td>
<td>Scott et al. (2010)</td>
</tr>
<tr>
<td>CPL</td>
<td>plains</td>
<td>106 of 108</td>
<td>7</td>
<td>responsive to $P_{TS}$</td>
<td>Evans et al. (2011)</td>
</tr>
<tr>
<td>SPL</td>
<td>plains</td>
<td>no prolonged drought</td>
<td>0</td>
<td>responsive to $P_{TS}$</td>
<td>S. A. Gunter, personal communication</td>
</tr>
<tr>
<td>LWA</td>
<td>plains</td>
<td>99 of 101</td>
<td>5</td>
<td>responsive to $P_{TS}$</td>
<td>P. J. Starks, personal communication</td>
</tr>
</tbody>
</table>

Notes: The functional responses include the lag in the response of ANPP to $P_T$ (Lag) and new species assemblages (NSA). The in situ knowledge used to understand the mechanics of site behavior include field notes, recent publications, and personal communications. Sites listed as responsive to $P_{TS}$ are those for which no outliers exceeded ± 1σ (± 1 SD; see Fig. 6). See Table 1 for site names.

year ANPP was not included in the ANPP model. This may be explained by exceptions to the response trajectory associated with ecosystems that are near thresholds of change (Peters et al. 2004, Smith et al. 2009). Continuing drought in plains grasslands could eventually lead to altered ecosystem functioning similar to that reported here for desert grasslands. This is supported by findings that the persistence of water deficit largely determines the sensitivity of biomes to drought and that there are different adaptation mechanisms in arid and mesic biomes (Vicente-Serrano et al. 2013). Vicente-Serrano et al. (2013) emphasized the role of temperature in the response of vegetation to climate variability by correlating the standardized precipitation evapotranspiration index (SPEI) with vegetation activity and growth.

Together, the findings of Ponce-Campos et al. (2013) and Zhang et al. (2013) and our findings (Fig. 7) suggest that there was shift in the functional response of desert grasslands to $P_T$. In contrast, the plains grasslands are following a predictable trajectory of ecosystem response to chronic drought that is expected to result in altered ecosystem functioning at the scale of months to decades (Smith et al. 2009). This difference in response between plains and desert grassland may be related to the adaptive responses of their dominant grass species to water availability. Plains grasslands historically and currently support large grazing populations (Briggs et al. 2005). The dominant grass assemblages of the plains grasslands maintain aboveground–belowground ratios which can impart greater capacity to physiologically adjust to climate variability, as well as facilitate community shifts, supporting greater regenerative capacity following disturbance (Knapp and Smith 2001, Zhou et al. 2009). Convergent selection pressure exerted by the combination of grazing and aridity in the plains grasslands may also be important in selecting for traits that impart substantial resistance to both disturbances (Milchunas et al. 1988, Quiroga et al. 2010). Desert grasslands, in contrast, did not coevolve with consistent grazing pressure and evolved the ability to rapidly respond to and effectively use highly variable and extremely pulsed precipitation (Bock et al. 1984, McClaran 1997). Once grass biomass is lost, grasslands can take years to recover (Peters et al. 2012), as evidenced by the lags observed for desert grasslands (Fig. 7). Such lags in grass productivity are a contributing factor to historical expansion of woody species into desert grassland systems (McClaran et al. 2003, McClaran et al. 2010), as well as contributing to altered community structure at DER and WGE that affected ANPP precipitation responses (Fig. 6). Collectively, these observations suggest that North American grasslands will undergo predictable, but regionally distinct, responses to the prolonged warm, dry conditions which are characteristic of climate change.

Results reported here for the early 21st century may be temporary, and the long-term impact on ecosystem functional integrity across the North American grassland biome is yet unknown. Interannual trends in plant phenology should provide a better understanding of the reported differences between arid and mesic sites, and the ecohydrological feedbacks leading to grassland mortality (Pennington and Collins 2007, Notaro et al. 2010). Also, the site-level mechanisms associated with disease and pests have not been a factor at these sites, but will likely further influence ecosystem function (Rosenzweig et al. 2001). Rising atmospheric CO$_2$ may partially counter the effects of prolonged warm drought by increasing plant biomass (Morgan et al. 2011, Donohue et al. 2013). These variant outcomes could be attributed to the contrast between short-term experiments and studies lasting longer than a decade related to biochemical changes (Buldocchi 2011). With a longer postdrought time period, it may be possible to investigate the potential for, and dynamics of, grassland recovery and rehabilitation (e.g., Peters et al. 2012).
Providing a prediction of the functional response of grasslands to altered hydroclimatic conditions and an operational means to monitor it has important implications. First, our work illustrates the value of experimental sites with long-term, species-scale observations of grassland dynamics combined with continuing satellite-based observations such as EVI in future studies and modeling efforts because in our case the species-specific information aided interpretation of the ANPP–P$_T$ anomalies. Second, our results show that regional management of grassland resources related to fire risk, loss of forage, and ecosystem services can be designed to account for predictable shifts in functional processes associated with climate change. These compelling results in a natural setting at the regional scale should play a role in future grassland research, management, and policy.

**Acknowledgments**

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**LITERATURE CITED**


