

1. Introduction

Our previous work has established that the dominant modes of Pacific sea surface temperature (SST) influence the summer climate of North America via remote forcing of the large-scale circulation, or teleconnections. These teleconnections excite in time and are most apparent during the early part of the summer, affecting the onset of the North American monsoon and the end of the late spring wet period in the central US. Our companion paper presented at the meeting summarizes and extends the physical mechanisms of the physical mechanisms of the central sector of regional climate model (RCM) framework. It has also been established via RCMs that the land surface influence of soil moisture and vegetation may significantly impact summer climate. However, current RCM studies have focused on only one or a few sites, typically with extreme climate conditions like 1988 and 1993. The hypothesis posed here is that the land surface influences become more important during the latter part of the summer, when the influence of remote Pacific SST forcing diminishes. Our goal is to eventually test this hypothesis using a RCM as a complement to our previous work, with idealized land surface forcing corresponding to the statistically significant spatiotemporal patterns of soil moisture and vegetation that have been associated with the largest atmospheric variability and most act synergistically with the summer teleconnections to create extreme conditions. The results presented herein show our initial statistical analysis of long-term observational soil moisture and vegetation data. The most significant findings from the latter two products will be used to drive future idealized RCM simulations with the Regional Atmospheric Modeling System (RAMS).

2. Description of Soil Moisture and Vegetation Datasets

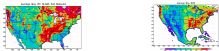


Figure 1: Climatological time-averaged soil moisture (left) and NDVI (right) derived from VIC model output.

Soil moisture data is from a long-term integration of the Variable Infiltration Capacity (VIC) hydrologic model over the North American Land Data Assimilation System (NLDAS) domain (one-degree grid resolution).

Vegetation greenness is defined using the Normalized Difference Vegetation Index (NDVI) available from Global-NDVI data from 1981-present at 1/4° resolution (Justice et al., 2002). NDVI can be used to derive leaf area index (LAI) per GCM vegetation type with a transfer algorithm.

3. Statistical Analysis Methods

Principal Component Analysis: Allow for the detection and reconstruction of quasi-stationary spatio-temporal climate signals that exhibit evidence of spatially correlated behavior. Principal Component (PC) analysis of the principal eigenmodes (1) statistical confidence intervals for the LTV spectrum; and (2) statistical confidence intervals corresponding to the significant time-varying forcing (e.g., Rajagopalan et al., 1994). The reference point for soil moisture and vegetation anomalies is defined as being in the central US for the present analysis, unless otherwise specified.

Wavelet Analysis: Decompose a time series into its multi-frequency components, providing information on periodic signals and how they vary in time (Torrence and Compo 1998). Used here to confirm interdecadal oscillation results by another method.

4. Dominant Spatiotemporal Modes of Global SST

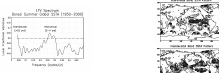


Figure 2: Principal eigenmode LTV spectrum of Global SST for 1950-2000. Dashed line indicates statistical significance of the 95% level. From Castro et al. (2006).



Figure 3: In-phase normalized SSTA associated with and interdecadal trends associated with the leading principal component. Values range from -1 to 1 (°C) (shaded red (right), Contour interval 0.25 units. From Castro et al. (2006).

Two statistically significant spatiotemporal modes of global SST are related to Indo and Indo-Sea decadal variability in the Pacific, of the cycle of 3-4 years and about 22 years, respectively. The patterns shown here are for boreal summer, as this is the season that is influenced in our present work. These SST patterns are related to distinct atmospheric teleconnections in both warm and cold seasons, as shown for example on our companion paper and Castro et al. (2006). We therefore expect a clear link between the significant patterns of central, soil moisture, and vegetation greenness would reflect the combination of SST forcing from both of these modes.

5. Standardized Precipitation Index (SPI)

The standardized precipitation index (SPI; McKeen et al., 1998) normalizes a given precipitation total at each point to a gamma distribution, allowing for comparison of precipitation anomalies over varying climate regimes on a continental scale. In the present analysis, the conventional SPI is used for the CPC US-6-month long term gauge-derived dataset (Pielke et al., 1994) for the period 1950-2000. The LTV spectrum for SPI considering all seasons (Fig. 5) shows significant interannual variability of 1° frequency of about seven years which is the band of vegetation greenness due to significant variability in soil moisture and vegetation greenness occur at approximately the same time scale. This modes sense shows that the most coherent patterns of atmospheric forcing, for both cold and warm regions, occur when the interannual and interdecadal variability in the Pacific sea in phase. The spatial pattern of variability reflects the well-established teleconnections between North American winter precipitation and Indo-Pacific winter (Fig. 6) and the corresponding wet/dry events (Fig. 7) show that the inter-annual variability in central US SPI is more dominant during the early part of the record and occur at a slightly lower frequency.

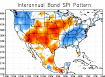


Figure 4: Interannual Band SPI Pattern.

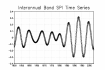


Figure 5: Interannual Band SPI Time Series.

Figure 6: Spatial pattern (left) and time series (bottom) of in-phase SPI in the interannual band (right) referenced to the central US.

Figure 7: Spatial pattern (left) and time series (bottom) of in-phase SPI in the interannual band (right) referenced to the central US.

5. SPI Cold vs. Warm Season

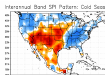


Figure 8: Interannual Band SPI Pattern: Cold Season.

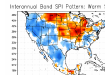


Figure 9: Interannual Band SPI Pattern: Warm Season.

Figure 10: Spatial pattern (left) and time series (bottom) of in-phase SPI in the interannual band (right) referenced to the central US.

Figure 11: Spatial pattern (left) and time series (bottom) of in-phase SPI in the interannual band (right) referenced to the central US.

A similar analysis of SPI in the interannual band considering total precipitation of the cold season (Fig. 8) September to May) and warm season (Fig. 9) June to August) repeatedly reveals several additional important features of interannual precipitation variability. The cold season SPI pattern repeats that for all seasons (Fig. 4) indicating that winter precipitation is likely more important than summer precipitation in determining drought conditions on interannual timescales. Central US precipitation anomalies associated with interannual precipitation variability in the same way throughout the year. Therefore, the climate in this region are very sensitive to the large-scale atmospheric forcing provided by Pacific SST. The most dramatic seasonal shift in the sign of atmospheric teleconnections associated with interannual variability occurs in the core monsoon region. This reflects the fact that interannual variability in winter summer precipitation is inversely related in this region. Therefore, a wet (dry) winter tends to be followed by a dry (wet) monsoon.

6. VIC NLDAS Soil Moisture

The statistically significant modes of interannual variability of VIC NLDAS soil moisture at a slightly lower frequency than that of the SPI (Fig. 10). The most coherent in interannual soil moisture variability occur in the central US (Fig. 11) and this is not surprising because the sign of interannual variation in precipitation there is consistent for the entire year (Fig. 8 and 9). The interannual band time series for soil moisture in the central US (Fig. 11) matches the SPI time series well and this is particularly true during the very wet. In particular, long-term drought occur on an approximately multi-decadal frequency. In the observational record, these include the Dust Bowl of the mid-1930s, mid-1950s, mid-1970s, 1980s, and the most recent drought of 1999-2002. We hypothesize that the most acute drought conditions in these periods are near the tail end of a cycle of relatively persistent Pacific SSTs, when the soil moisture deficit starts to act synergistically with atmospheric forcing (Fig. 1950-54, 1974-77, 2000-02).

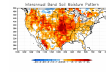


Figure 12: Interannual Band Soil Moisture Pattern.

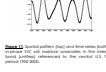


Figure 13: Interannual Band Soil Moisture Time Series.

Figure 14: Spatial pattern (left) and time series (bottom) of in-phase VIC soil moisture with the interannual band (right) referenced to the central US for the period 1950-2000.

Figure 15: Spatial pattern (left) and time series (bottom) of in-phase VIC soil moisture with the interannual band (right) referenced to the central US for the period 1950-2000.

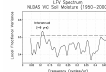


Figure 16: Principal eigenmode LTV spectrum of soil moisture for VIC NLDAS soil moisture (1950-2000). Dashed line indicates statistical significance of the 95% level.

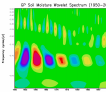


Figure 17: Wavelet spectrum of VIC soil moisture (left) and time series (bottom) of the interannual band (right) referenced to the central US for the period 1950-2000.

7. Satellite-Derived NDVI

Significant interannual variability in AVHRR Global-NDVI occurs on a timescale of about seven years (Fig. 13), unlike soil moisture, long-term variability in vegetation greenness is a maximum in the southwest US, and does not correspond well with precipitation variability (Fig. 14). The likely reason for this is because vegetation growth is influenced by factors other than precipitation in this area. The surface temperature and availability of sunlight and not moisture availability may be the most dominant factors. Variability in vegetation greenness in the core monsoon region indicates a dependence on unmodeled factors. Variability in vegetation greenness in the core monsoon region indicates a dependence on unmodeled factors. Variability in vegetation greenness in the core monsoon region indicates a dependence on unmodeled factors. We note there is a problem with NDVI data for 1994, which affect the NDVI time series and wavelet analysis (Fig. 14 and 15). Also, the wavelet analysis is biased for the southwest US instead of the central US, in this case.

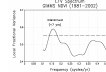


Figure 18: Principal eigenmode LTV spectrum of AVHRR Global-NDVI (1981-2000). Dashed line indicates statistical significance of the 95% level.

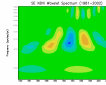


Figure 19: Wavelet spectrum of AVHRR Global-NDVI (left) and time series (bottom) of the interannual band (right) referenced to the central US for the period 1980-2000.