1	El Niño phases embedded in Asian and North American drought
2	reconstructions
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ABSTRACT

2	The amplitude of El Niño-Southern Oscillation (ENSO) varies substantially at each
3	phase of its evolution, affecting the timing and patterns of atmospheric teleconnections
4	around the globe. Instrumental records are too short to capture the full behavior of ENSO
5	variability. Here we use the well-validated Monsoon Asia Drought Atlas (MADA) and
6	North America Drought Atlas (NADA) for the past 700 years, and show that tree-ring
7	records from different regions represent tropical sea surface temperature (SST) conditions at
8	various phases of ENSO. Three modes of tree-ring based summer drought variability are
9	found to be correlated with ENSO: summer droughts over the Maritime Continent and
10	Southwest North America (NA), and a dipole mode between Central and South Asia. A
11	lagged correlation analysis is performed to determine the time when precipitation and
12	temperature anomaly imprints on summer droughts as recorded in tree-rings. Drought
13	anomalies in the Maritime Continent and Southwest NA represent ENSO at the developing
14	and peak phases respectively, while those over Central/South Asia are associated with
15	tropical wide SST anomalies (including the Indian Ocean) at the decay phase of ENSO.
16	Thus proxy records from different regions can provide valuable information on long-term
17	behavior of ENSO at different phases.
10	

19 Keywords: ENSO; tree-rings; drought; Monsoon Asia; North America

1 1. Introduction

2 The El Niño-Southern Oscillation (ENSO) phenomenon dominates tropical ocean-3 atmosphere variability on interannual timescales, with profound impacts on global weather 4 and climate (McPhaden et al. 2006; Deser et al. 2010). One fundamental property of El Niño, 5 the ENSO warm phase, is that its evolution tends to be phase-locked to the annual cycle, 6 with peak warming around the end of the calendar year (Rasmusson and Carpenter, 1982; 7 Trenberth, 1997). Because of evolving patterns of tropical sea surface temperature (SST) 8 anomalies and atmospheric teleconnections, ENSO effects display strong seasonality in 9 various parts of the globe. Droughts tend to occur in India (Indonesia) in summer (fall) of 10 the El Niño developing year (Shukla and Paolino, 1983; Ropelewski and Halpert, 1987; 11 Haylock and McBride, 2001). Precipitation and temperature anomalies tend to peak over 12 North America during winter at the peak phase of ENSO (Trenberth et al., 1998; Alexander 13 et al., 2002; Johnson and Feldstein, 2010). Warm temperature over India and reduced 14 precipitation over the northwest Pacific east of the Philippines linger through the El Niño 15 decay summer (Xie et al. 2009; Du et al. 2011). 16 Each El Niño event evolves differently from others, with the amplitude varying 17 substantially throughout the life cycle at the developing, peak, and decay phases (Fig. 1). 18 The large variations in El Niño amplitude, plus the interactions with other climate 19 phenomena such as the Indian Ocean Dipole (IOD), have caused marked changes in ENSO 20 teleconnection patterns around the globe (Krishna Kumar et al. 1999; Annamalai et al., 2005; 21 Xie et al., 2010; Chowdary et al., 2012; Li et al., 2013). Since the 1970s, for example, the 22 ENSO effects weaken on India summer rainfall (Krishna Kumar et al. 1999) but strengthen 23 on Indian Ocean SST and summer rainfall over the northwest Pacific and East Asia (Xie et

2	variations at each phase of its evolution, a crucial step toward understanding changes in the
3	teleconnection patterns.
4	Instrumental records indicate that ENSO exhibits strong variability on the classical 2-7-
5	year band in the past 150 years, and that its amplitude and frequency are modulated at
6	decadel to interdecadal timescales (Fang et al., 2008; Deser et al., 2010; Li et al., 2011). In
7	light of such decadel to interdecadal modulation, existing instrumental records are too short
8	to characterize the full behavior of ENSO variability (Guilyardi et al., 2009; Wittenberg,
9	2009; Yeh et al., 2011; Stevenson et al., 2012). Long integrations of coupled general
10	circulation models (GCMs) have been used to evaluate slow modulations of ENSO
11	(Wittenberg, 2009; Yeh et al., 2011). However, current GCMs have limited ability to
12	reproduce historical ENSO variability, and that they yield a wide range of projections for
13	future ENSO variability (Guilyardi et al., 2009; Yu and Kim, 2010; Ham and Kug, 2012). A
14	better understanding of long-term ENSO variability is necessary for improving climate
15	models and their projection of ENSO behavior linked to global warming.
16	Proxy records are the primary source to study climate variability beyond the
17	instrumental period. Many proxies, in particular seasonally to annually resolved tree-rings
18	and corals, have been used to reconstruct ENSO variability during the past millennium (e.g.,
19	Stahle et al., 1998; Cobb et al., 2003; D'Arrigo et al., 2005; Braganza et al., 2009; Wilson et
20	al., 2010; Emile-Geay et al., 2013; Li et al., 2013). These reconstructions have provided
21	valuable information on long-term ENSO behaviors, such as its frequency change,
22	amplitude modulation, and possible effects of solar and volcanic forcing on ENSO

al. 2010; Chowdary et al. 2012). Therefore, it is important to determine El Niño amplitude

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23 occurrence. Based on various proxy records and statistical methods, however, these

1	reconstructions exhibit large inconsistency in ENSO variability in both low- and high-
2	frequency components (D'Arrigo et al., 2005; McGregor et al., 2010; Wilson et al., 2010).
3	Part of the ENSO reconstruction uncertainty was introduced by statistical treatment of
4	proxy data (i.e., removal of biological trend in proxies) and/or different method used for
5	each reconstruction (von Storch et al., 2004; Mann et al., 2007; Wilson et al., 2010, Emile-
6	Geay et al., 2013). This type of uncertainty could be minimized by rigorously evaluating and
7	improving the statistical methods (Mann et al., 2007; Smerdon et al., 2010, 2011).
8	Meanwhile, the reconstruction uncertainty was equally likely caused by shifts in ENSO
9	teleconnection patterns, as most of previous studies did not consider seasonal evolution and
10	spatial structure of ENSO and their influence on the spatial/temporal stability of
11	teleconnections (Cole and Cook, 1998; Trenberth and Stepaniak, 2001; Kim et al., 2009).
12	Moreover, proxy records in different regions have distinct phenological cycle, often most
13	sensitive to climate anomalies of a particular season. Without scrutinizing the ENSO signals
14	encoded in proxy records, a simple combination of them for reconstruction will lead to large
15	inconsistencies, a problem that could be worsened by the improper use of statistical
16	techniques. In light of this, we argue that a careful examination of ENSO phases in global
17	proxy records is essential for improving ENSO reconstructions. This assessment is
18	becoming increasingly important, as the need rises to combine and synthesize accurately
19	dated, high-resolution proxy records around the globe.
20	This study aims to systematically examine the spatiotemporal structures of ENSO effects
21	on summer droughts over Monsoon Asia (MA) and North America (NA), by employing the
22	tree-ring derived Monsoon Asia Drought Atlas (MADA) and North America Drought Atlas
23	(NADA) (Cook et al., 2004, 2008, 2010). This work will pinpoint ENSO phases encoded in

tree-rings, and will shed light on the general pattern of tree growth response to ENSO over MA and NA. We hope this work will inspire more exploration of tree-rings and other proxies around the globe to distinguish the encoded ENSO signals. These efforts will eventually lead to more skillful reconstructions to improve our understanding of the complex ENSO system.

6 The current paper is organized as follows. Section 2 describes instrumental and proxy 7 data as well as the analytical techniques employed in this study. Section 3 describes the two 8 dominant drought patterns over MA and NA for the past seven centuries. In section 4 we 9 demonstrate that the dominant drought patterns over MA and NA are indeed related to 10 different ENSO phases. Section 5 illustrates the seasonal relationships of summer droughts 11 to current and antecedent precipitation/temperature anomalies, and the physical processes 12 for the ENSO-drought connections in each region. Based on the identified phase relationship 13 with ENSO, section 6 discusses the covariability of summer droughts between MA and NA 14 for the past seven centuries. Section 7 is a summary.

15

16 **2. Data and methods**

17 2.1. The Monsoon Asia and North America Drought Atlas

18 We employed the well-validated MADA and NADA to represent climate signals

19 encoded in moisture sensitive tree-rings across the two continents (Cook et al., 2004, 2008,

20 2010). The MADA and NADA are gridded datasets of tree-ring derived summer (June-

- 21 August) Palmer Drought Severity Index (PDSI; Palmer, 1965) reconstructions. The PDSI is
- 22 an index of meteorological drought that incorporates both precipitation and temperature into
- 23 a two-layer soil moisture reservoir model, such that moisture supply (rainfall or snowfall

1	water equivalent) and demand (water loss through evapotranspiration as a function of
2	temperature) could be accounted for in the model (Palmer, 1965; Heim, 2002). As a means
3	to measure soil moisture content, the PDSI was designed to include a built-in persistence
4	term, which means the PDSI for a given month integrates current and antecedent moisture
5	conditions for a few months to seasons. Moreover, the PDSI is scaled to remove differences
6	between regional climatology, thereby allowing direct comparison of spatiotemporal
7	moisture variations over a large region. By design, positive (negative) values of the PDSI
8	indicate wetter (drier) than normal conditions, generally within a range of -6 to 6. We note a
9	few caveats and limitations on the use of the PDSI, such as its inappropriateness for winter-
10	season and short-term droughts, its slow response to developing and diminishing droughts,
11	and its fixed weighting factors for regions with diverse climatology (Alley 1984; Guttman et
12	al., 1992; Heim, 2002; Wells et al., 2004). Nonetheless, the PDSI remains one of the most
13	widely used drought indices in the United States and for much of the globe (Ntale and Gan,
14	2003; Dai et al., 2004; Li et al., 2009a; Dai, 2011).
15	The MADA consists of summer PDSI reconstructions on 534-point grid derived from
16	327 tree-ring chronologies, available on a $2.5^{\circ}x2.5^{\circ}$ regular grid over MA (Fig. 2). The final
17	MADA product is the ensemble mean of 24-member reconstructions for each grid point
18	using a modified "point-by-point regression (PPR)" method (Cook et al., 1999, 2010). Each
19	PPR model was calibrated over the period 1951-1989, and was verified against instrumental
20	data over 1920-1950 for testing the accuracy of tree-ring estimates. Calibration and
21	verification statistics indicate significant reconstruction skill over most of the domain as far
22	back as 1300. Given that the reconstructions ended in 1989, we appended them with

1	instrumental data from 1990 to 2005, and adjusted the means and variances of the
2	reconstructions accordingly. As a result, the MADA spans 1300-2005 (Fig. 3).
3	The NADA is a set of tree-ring derived summer PDSI reconstructions over most of NA,
4	based on a 286-point 2.5°x2.5° regular grid (Fig. 2). In its current version (NADAv2a;
5	available online at: http://www.ncdc.noaa.gov/paleo/pdsi.html), the NADA was developed
6	using 1845 annually resolved tree-ring records across Canada, the United States, and
7	northern Mexico (Cook et al., 2008). The PPR method was used in a nested manner in order
8	to utilize the full length of the available tree-ring records to extend the reconstructions as far
9	back as possible (Cook et al., 2004). Each PPR model was developed over the fixed
10	calibration period 1928-1978, and was verified against instrumental PDSI data over 1900-
11	1927. The model test results indicate that the NADA is highly reliable back to 1300, and is
12	still quite useful back to 800. In addition, instrumental PDSI data were appended to the end
13	of the reconstructions to bring them up to 2006. As a result, the NADAv2a covers AD 0-
14	2006, with reconstructions available at over 70% of all grid points during most of the last
15	millennium (Fig. 3). In this study, we focus on the reconstructions over 1300-2005, a period
16	most reliable and consistent with the time span of MADA.
17	

18 2.2. Instrumental climate data

Monthly precipitation and temperature data were obtained from the Climatic Research
Unit (CRU) TS 3.1 global climate dataset, available for 1901-2009 on a half-degree grid
(Mitchell and Jones, 2005). These climate data were used to investigate how summer
droughts over MA and NA are related to current and antecedent precipitation and
temperature anomalies.

1	The primary SST data we used are the National Climatic Data Center (NCDC) extended
2	reconstructed SST dataset version 3b (ERSST.v3b), available on a monthly 2-degree grid
3	(Smith et al., 2008). This dataset starts in 1854, but its variability is heavily damped before
4	1870 due to sparse data availability. We therefore only used the data after 1870. We
5	repeated our analyses by substituting the ERSST with the Hadley Center and Kaplan SST
6	products (Kaplan et al., 1998; Rayner et al., 2006), and found them very consistent in
7	exhibiting the relationships between global SSTs and drought anomalies over MA and NA.
8	Here we report the results obtained from the ERSST dataset.
9	
10	2.3. Analytical techniques
11	We adopted an objective approach to investigate the influence of ENSO on summer
12	droughts over MA and NA. That is, we first take advantage of MADA and NADA to define
13	the spatially coherent drought patterns in each continent for the past seven centuries, and
14	then analyze if temporal changes of each drought pattern are related to ENSO variability.
15	This method allows for drought patterns defined in terms of their own long-term
16	spatiotemporal behaviors, avoiding a somewhat arbitrary selection of a region of interest
17	based on its association with ENSO during the instrumental period. This consideration
18	matters in that the ENSO teleconnection patterns might vary over time, and the instrumental
19	data only become available for most of MA untill the 1950s.
20	Empirical Orthogonal Function (EOF) analysis was used to objectively define the
21	dominant modes of summer drought variability over MA and NA. We performed EOF
22	analysis on MADA and NADA separately, and focused on the first two leading modes in
23	each domain and their temporal variability (i.e., the associated principal components (PCs)

1 of each mode) over 1300-2005.

2	Because El Niño events are phase-locked to the annual cycle (Fig. 1), summer droughts
3	may result from teleconnections induced by either the preceding or the following El Niño
4	event. In order to pinpoint the influence of ENSO on summer droughts, we investigated their
5	relationships for the full El Niño cycle that spans two calendar years by calculating the
6	seasonally lagged correlations of the MADA/NADA PCs with global SSTs from the
7	precedent to the following winter for their common period 1870-2005. The lagged
8	correlations were presented in a progressive manner, so that the El Niño phases embedded in
9	MADA and NADA could be easily identified.
10	Summer soil moisture content incorporates climate signals of both current and
11	antecedent seasons (Palmer, 1965; Heim, 2002; St George et al., 2010). In order to unravel
12	physical processes by which ENSO imprints on summer droughts, we calculated the
13	seasonally lagged correlations of the MADA/NADA PCs with the CRU precipitation and
14	temperature data over 1951-2005. These correlations were calculated from the precedent to
15	the following winter, and were presented in a progressive manner. This facilitates
16	identifying the seasonality and the relative role of precipitation versus temperature on
17	summer droughts over MA and NA.
18	
19	3. Dominant drought patterns over MA and NA

20 3.1. Dominant MADA drought patterns

The MADA EOF1 accounts for the largest percentage (15.2%) of variance in the field, and is characterized by a distinct dipole mode, with one center in the mid-latitude Central Asia and the other in India with extensions to the southern Tibetan Plateau and northern

1	Southeast Asia (Fig. 4A). The EOF loading over Central Asia is much higher than that over
2	South Asia, suggesting that this dipole mode is dominated by moisture variability over
3	Central Asia. The opposite sign of loadings indicates that there exists a seesaw pattern
4	regarding summer moisture change in the two regions, with concomitant moisture increase
5	in Central Asia and a decrease in South Asia, or vice versa. The MADA EOF2 accounts for
6	9.6% of the total variance, and represents a distinct moisture pattern with strong positive
7	loading over the Maritime Continent (Fig. 4B). The first EOF pattern identified here from
8	standard EOF analysis is nearly identical to the first distinct EOF (DEOF1) mode for the
9	same period (Cook et al., 2010). The second EOF pattern agrees well with the DEOF5 mode
10	in Cook et al. (2010), whereby the explained variance of DEOF5 (6.39%) is as high as that
11	of DEOF2 (6.57%). These agreements suggest that the identified drought patterns for
12	MADA are insensitive to the orthogonality constraint of EOF analysis, and are robust
13	features of summer drought variability over MA, at least for the past seven centuries.
14	The MADA PC1 represents summer moisture change over Central and South Asia
15	during 1300-2005, with positive scores indicating concomitant pluvials in Central Asian and
16	droughts in South Asia, and vice versa for negative scores (Fig. 5A). The PC scores indicate
17	two pronounced pluvial (drought) conditions over Central (South) Asia around the 1340s-
18	60s and 1420s, both recorded by the Aral Sea sediments in Central Asia (Boomer et al.,
19	2009) and speleothem records in east-central India and west-central China (Zhang et al.,
20	2008; Sinha et al., 2011). These two megadroughts in South Asia indicate severe and
21	sustained monsoon failure during these periods, a vital factor for the demise of the Khmer
22	civilization at Angkor in Cambodia (Buckley et al., 2010). Other megadroughts in South
23	Asia indicated by the MADA PC1, such as those around the 1560s-90s, 1680s-90s, 1760s

and 1790s, are widely recorded in ice cores on the southern Tibetan Plateau (Thompson et
al., 2000), speleothem records in east-central India and west-central China (Zhang et al.,
2008; Sinha et al., 2011), and an independent tree-ring record in southern India (Borgaonkar
et al., 2010). The agreement with multiple proxies validates the reliability of MADA PC1 in
representing regional moisture change for the past seven centuries.

6 The MADA PC2 represents summer moisture change over the Maritime Continent, with 7 positive (negative) scores indicating pluvial (drought) conditions, respectively (Fig. 5B). 8 Many of the megadroughts observed in South Asia, such as those around the 1340s-60s, 9 1420s and 1760s, are also found to occur over Southeast Asia. However, some other 10 megadroughts in South Asia, such as those around the 1560s-90s and 1790s, have no 11 parallels in Southeast Asia. These results suggest that the spatially coherent megadroughts 12 over South and Southeast Asia waxed and waned over time, underlining marked changes in 13 the large-scale circulations of the Asian monsoon system. Future work towards 14 understanding the shifts of these modes will provide deeper insights into the Asian monsoon 15 dynamics. Moreover, a persistent drying trend occurred over Southeast Asia since the 1960s, 16 with its duration and magnitude unprecedented for the past seven centuries (Fig. 5B). This 17 change is likely due to the decrease in both precipitation and SST warming associated with 18 the weakening of the Indo-Pacific Walker circulation (Tokinaga et al., 2012). A similar 19 drying trend is observed in East Asia (Jiang and Wang, 2005; Li et al., 2009b). These 20 persistent drying trends over East and Southeast Asia may suggest a significant weakening 21 of the Asian monsoon system in recent decades, which is thought to be in large part a 22 consequence of global warming (Zhou et al., 2008; Li et al., 2010).

23

1 3.2. Dominant NADA drought patterns

2 The NADA EOF1 accounts for 24.9% of the total variance, and shows a distinct pattern 3 centered over Southwest NA (Fig. 4C). The NADA EOF2 explains 13.7% of the total 4 variance, and is characterized by a dipole mode with opposite loadings over the Pacific 5 Northwest (PNW) and the Texas-Mexico (TexMex) region (Fig. 4D). These two leading 6 EOF patterns for 1300-2005 are nearly identical to the first two EOFs for 1400-2005 using 7 the same dataset (Cook et al., 2011), and are similar to the first and the third mode in a 8 rotated EOF analysis of NADAv1a spanning 1000-2003 (Herweijer et al., 2007). Although 9 there are biological limitations on using total ring-width as a proxy for summer moisture, 10 NADA is capable of representing observed summer droughts over most of western NA 11 (Cook et al., 2004; St George et al., 2010). Therefore, the two identified drought patterns for 12 NADA are likely robust features of summer moisture change over NA, rather than artifacts 13 of tree-ring reconstructions or orthogonality constraint of EOF analysis. 14 Instrumental records suggest that the 1950s drought is the most severe and prolonged in 15 the Southwest for the 20th century (Fye et al., 2003). The NADA PC1 suggests that this 16 drought is not unprecedented in terms of both severity and duration for the past seven 17 centuries. For example, the droughts in the 1850s-60s and 1660s are comparable to the 18 1950s drought in severity, and the droughts in the 1360s-1390s and 1560s-1580s had 19 persisted much longer than the 1950s drought (Fig. 5C). In contrast, the early 20th century 20 pluvial recorded by instrumental data is unmatched in both severity and duration for the past 21 seven centuries, despite that several severe and sustained pluvials took place, such as those 22 in the 1300s-1330s and 1820s-30s. These findings from NADA PC1 are consistent with 23 previous studies that use either NADA but with different analytical techniques, or other

1 independent datasets (e.g., Meko et al., 1995; Fye et al., 2003; Woodhouse et al., 2005;

2 Herweijer et al., 2007; Stahle et al., 2007, 2011).

3 Positive scores of NADA PC2 indicate concomitant pluvials in the PNW and droughts in 4 the TexMex region, and vice versa for negative scores (Fig. 5D). Instrumental data indicate 5 that the 1930s drought is pronounced in the Pacific Northwest for the 20th century (Fye et 6 al., 2003). The NADA PC2 agrees with instrumental data, and provides a longer perspective 7 that the 1930s drought is the worst in the Pacific Northwest for the past seven centuries, 8 although its duration has been exceeded by the 1430s-80s and 1650s-60s drought. Likewise, 9 the early 20th century pluvial is unprecedented in severity for the past seven centuries, 10 though its duration has been surpassed by pluvials around the 1300s-40s and 1590s-1620s. 11 Moreover, a comparison of NADA PC2 with NADA PC1 indicate that many pluvials, such 12 as those around the 1300s-30s, 1590s-1620s, 1820s-30s, 1900s-20s, and many droughts, 13 such as those around the 1430s-70s, 1750s, and 1930s, occurred concomitantly over the 14 Southwest and the Pacific Northwest, whereas opposite moisture conditions took place over 15 the two regions in a few other periods such as the 1410s, 1480s, 1560s-80s, 1950s, and 16 1990s. These variations in spatially coherent moisture conditions underline marked changes 17 in atmospheric circulations over western NA. As will be discussed in the following section, 18 variations in the spatial coherence of moisture conditions across western NA are mainly due to the shifts in wintertime ENSO teleconnections over the PNW/TexMex region. 19

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21 **4. Summer droughts in relation to ENSO phases**

Among others, remote SST forcing and regional land-air interactions are two important processes that determine summer moisture condition in a region. SST anomalies in remote

1 oceans can excite or shift large-scale atmospheric circulation patterns, thereby causing 2 precipitation and temperature anomalies that favor the occurrence of droughts or pluvials in 3 different regions (Hoerling and Kumar, 2003; Deser et al., 2010). The land-air interactions 4 involve changes in boundary conditions such as soil moisture and surface albedo, and 5 promote additional precipitation and temperature anomalies that often feedback positively 6 on regional moisture condition (Eltahir, 1998; Pal and Eltahir, 2001; Notaro and Zarrin, 7 2011). These two processes are intrinsically related, but their relative importance on 8 modulating moisture condition might differ by region. Therefore, in order to pinpoint the 9 influence of ENSO on summer droughts over MA and NA, we will first examine how 10 summer droughts in each region are related to remote SST anomalies (in this section), and 11 then analyze how they are related to regional precipitation and temperature change (in the 12 next section). 13 The MADA PC1 is most significantly and positively correlated with east-central tropical 14 Pacific SSTs in prior winter (DJF), but their correlations weaken in spring to summer and

vanish thereafter (Fig. 6A-F). The progressive weakening of their correlations in antecedent to current seasons indicates that summer drought anomalies over Central/South Asia are most related to the peak phase of El Niño events that develop in the spring to summer of the preceding year and decay in current spring to summer. The positive correlations indicate the occurrence of concomitant pluvials in central Asia and droughts in South Asia when eastcentral tropical Pacific SSTs are warmer than normal.

In contrast, the MADA PC2 has no significant correlations with east-central tropical
Pacific SSTs in prior winter (Fig. 6G). Instead, they are significantly correlated from current
spring to the following winter, with the highest correlations in August-October (ASO) (Fig.

1	6H-L). This seasonality of correlations indicates that summer drought anomalies over the
2	Maritime Continent are most related to the developing phase of El Niño events that peak in
3	the following winter. The negative correlations indicate that droughts occur over the
4	Maritime Continent when east-central tropical Pacific SSTs are warmer than normal.
5	Similar to MADA PC1, the NADA PC1 is also most significantly and positively
6	correlated with east-central tropical Pacific SSTs in prior winter (DJF), and their correlations
7	weaken in spring to summer and vanish thereafter (Fig. 7A-F). This indicates that summer
8	drought anomalies over Southwest NA are most related to the peak phase of the preceding
9	El Niño event, with wet conditions over the Southwest when east-central tropical Pacific
10	SSTs are warmer than normal.
11	The NADA PC2 is also most strongly correlated with east-central tropical Pacific SSTs
12	in prior winter (DJF), but the magnitude of correlations is much lower than that for NADA
13	PC1 (Fig. 7G-L). Thus summer drought anomalies over the PNW/TexMex region are
14	weakly related to the peak phase of the preceding El Niño event. The negative correlations
15	indicate the occurrence of concomitant droughts in the PNW and pluvials in the TexMex
16	region when east-central tropical Pacific SSTs are warmer than normal.
17	Compared to the other three modes, the correlations between NADA PC2 and tropical
18	Pacific SSTs are significant but much weaker, indicating that the ENSO-drought
19	associations might vary over time. We test this possibility by calculating the running
20	correlations between DJF tropical SSTs and each drought mode for successive 31-year
21	intervals during 1870-2005. The results indicate that the ENSO-drought teleconnection is
22	highly persistent over Central/South Asia, the Maritime Continent and Southwest NA during
23	the instrumental period (figures not shown). However, the ENSO-drought teleconnection

1 has varied substantially over the past 136 years over the PNW/TexMex region, with strong 2 association around 1900-30 and weak association around 1960-90 (Supplementary Fig. 1). 3 These results indicate the non-stationary association between ENSO and summer drought 4 anomalies over the PNW/TexMex region. 5 The association with different ENSO phase for each drought pattern is further assessed 6 by calculating their lead-lag correlations with ENSO indices. Because of the weak and 7 unstable correlations between NADA PC2 and tropical Pacific SSTs, we only calculated 48-8 month lead-lag correlations between the other three PCs and the Niño3/4 indices for their 9 common period 1870-2005 (Fig. 8). The results indicate that the correlations of each drought 10 pattern with the Niño3/4 indices exhibit similar seasonal evolution, albeit subtle differences 11 exist in the relative magnitude. For MADA PC1, the highest correlations are positive, and 12 are found when ENSO leads by 6-7 months summer droughts over Central/South Asia (Fig. 13 8A). For MADA PC2, the highest correlations are negative, and are found in the season 14 when ENSO lags by 0-2 months behind summer droughts over the Maritime Continent (Fig. 15 8B). Similar to MADA PC1, the highest correlations for NADA PC1 are positive, and are 16 found when ENSO leads by 6-7 months summer droughts over Southwest NA (Fig. 8C). 17 Taken together, the above results indicate that different El Niño phases are embedded in 18 MADA and NADA. Summer droughts over Central/South Asia, Southwest NA, and the 19 PNW/TexMex region are most related to the peak phase of the preceding El Niño event, 20 whereas those over the Maritime Continent are most related to the developing phase of the 21 following El Niño event. Moreover, the ENSO-drought teleconnections are highly persistent 22 over Central/South Asia, the Maritime Continent, and Southwest NA, but they have varied 23 substantially over the PNW/TexMex region, at least for 1870-2005.

2	5. Summer droughts in relation to precipitation/temperature anomalies
3	Section 4 established that summer droughts over Central/South Asia, the Maritime
4	Continent and Southwest NA are related to different ENSO phases. In this section we
5	analyze how summer droughts in these regions are related to current and antecedent
6	precipitation and temperature anomalies, and discuss physical processes that imprint ENSO
7	effects on local climate/moisture anomalies.
8	
9	5.1. Central/South Asia
10	The MADA PC1 is positively (negatively) correlated with precipitation over Central
11	(South) Asia from prior winter to current summer, but their correlations vanish in the
12	following autumn to winter (Fig. 9A-F). Consistent correlation patterns are found with the
13	observed summer PDSI (Supplementary Fig. 2), suggesting that tree-rings over
14	Central/South Asia have faithfully reproduced the seasonality and the strength of the
15	observed correlation patterns without significant amplification or dampening effect.
16	Correlations of MADA PC1 with temperature are broadly consistent with that of
17	precipitation, albeit with reversed signs (Fig. 9G-L). Strong correlations of MADA PC1 with
18	precipitation and temperature are both found from April-June (AMJ) to June-August (JJA).
19	In Central Asia, precipitation concentrates in spring to summer, and is light in
20	autumn/winter period (Fig. 9A-F). The season of peak rainfall differs spatially over
21	Central/South Asia, but is generally found in AMJ to JJA (Fig. 9C and 9D). These results
22	indicate that summer moisture over Central/South Asia is tuned mainly to precipitation and
23	temperature anomalies in the major wet season.

1	The above results indicate that both precipitation and temperature are important for
2	summer drought anomalies over Central/South Asia. This raises a question of whether they
3	contribute independently, or one is simply a covariate of another. To examine whether
4	precipitation and temperature are independent casual factors for summer droughts, we
5	calculated their point-by-point correlations in space, with a focus on the season when they
6	are strongly related to summer droughts. The results indicate that precipitation and
7	temperature in AMJ are closely related to each other over Central/South Asia
8	(Supplementary Fig. 4A), such that their anomalies are largely due to the same forcing.
9	From late spring through summer over transition zones between wet and arid regions, soil
10	moisture controls surface evaporation. The strong soil moisture feedback during this time
11	makes surface air temperature sensitive to antecedent and current precipitation (Guo et al.
12	2012), and surface cooling occurs in response to excessive antecedent and current
13	precipitation, consistent with our results in Fig. 9.
14	The MADA PC1 is highly correlated with Niño3/4 SST 6-7 months in advance
15	(equivalent to the preceding NDJ), but their correlations are weak at 0-2 months lead (Fig.
16	8A). This is because ENSO peaks in NDJ and decays substantially by AMJ/JJA (Fig. 1).
17	Thus the high correlation between MADA PC1 and precipitation/temperature in AMJ/JJA
18	indicates that the ENSO influence on summer droughts over Central/South Asia is likely
19	indirect, mediated by SST anomalies outside the Niño3/4 region. The delayed Indian Ocean
20	warming in response to El Niño is a good candidate. Recent studies indicate that the Indian
21	Ocean SST warming persists well into summer (Du et al., 2009), and is influential on
22	climate of the northwest Pacific and East Asia (Xie et al., 2009; Du et al., 2011; Huang et al.,
23	2010). Our results suggest that this Indian Ocean capacitor effect extends to Central Asia (Li

1	et al., 2010; Fang et al., 2011). Analysis to be presented elsewhere implicates orographic			
2	effects by the Tibetan Plateau in response to circulation anomalies associated with the			
3	tropical tropospheric warming (Y. Kosaka, personal communications). Over South Asia, the			
4	correlations of MADA PC1 are much higher with air temperature (~0.6, Fig. 9I) than with			
5	precipitation (~-0.3, Fig. 9C), indicating the importance of the persistent SST warming over			
6	the North Indian Ocean and South China Sea for summer dry conditions. Together, these			
7	results indicate that although correlated with the peak phase ENSO in the preceding winter,			
8	the MADA PC1 actually records tropical-wide SST conditions during the ENSO-decay			
9	spring/summer, not just over the equatorial Pacific but including the Indian Ocean and South			
10	China Sea.			
11				
12	5.2. Maritime Continent			
13	Correlations of MADA PC2 with precipitation in antecedent seasons are generally weak			
14	over the Maritime Continent (Fig. 10A-C). Instead, they are strongly and positively			
15	correlated in current summer and thereafter (Fig. 10D-F). The seasonality of MADA PC2			
16	correlations with temperature is broadly consistent with that of precipitation (Fig. 10G-L),			
17	and is also consistent with local SST anomalies over the North Indian Ocean and South			
18	China Sea (Fig. 6G-L). The highest correlations of MADA PC2 with precipitation and			
19	temperature are both found in current summer (JJA) (Fig. 10D and 10J). In JJA, the			
20	correlations between precipitation and temperature are generally insignificant			
21	(Supplementary Fig. 4B), suggesting that they may contribute independently to regional			
22	summer drought anomalies.			
23	For the Maritime Continent, the major rainy season differs spatially, but peak rainfall			

generally occurs in DJF when the Southern Hemisphere summer monsoon prevails (Haylock
and McBride, 2001; Hendon, 2003; Chang et al., 2005), with JJA the dry season for most of
the region (Fig. 10A-F). The MADA PC2 is highly correlated with Niño3/4 SST in summer
to autumn (Fig. 8B). Therefore, the peak correlation between MADA PC2 and precipitation
in JJA indicates the direct influence of the developing phase of ENSO on summer droughts
over the Maritime Continent.

7 The Maritime Continent is located in the ascending limb of Walker circulation. In peak 8 wet season (DJF), the prevailing Southern Hemisphere summer monsoon is perturbed by 9 local complex topography, such that rainfall anomalies are not well organized in space over 10 the region (Haylock and McBride, 2001; Hendon, 2003; Chang et al., 2005). In contrast, 11 rainfall variability in summer to autumn occurs in response to atmospheric circulation 12 anomalies associated with ENSO, with high spatial coherence that allows soil moisture to 13 develop analogous patterns over a large region. The highest correlations between MADA 14 PC2 and ENSO are found in ASO (Fig. 6K), consistent with the fact that the ENSO 15 influence on rainfall over the Maritime Continent is most significant in boreal autumn 16 (Haylock and McBride, 2001; Hendon, 2003; Chang et al., 2005).

17

18 5.3. Southwest North America

19 The NADA PC1 is positively correlated with precipitation over Southwest NA. The 20 correlations peak in prior winter (DJF; ~0.5), decay toward current summer, and vanish in 21 the following autumn to winter (Fig. 11A-F). Consistent correlation patterns are found with 22 the observed summer PDSI (Supplementary Fig. 3), suggesting that tree-rings have 23 faithfully reproduced the seasonality of the observed correlation patterns. By contrast, the

correlations of NADA PC1 with temperature are negative, and peak in AMJ to JJA (~-0.5)
 instead of DJF (Fig. 11G-L).

3	The above results indicate that both precipitation and temperature are important for			
4	summer droughts over Southwest NA, albeit in different seasons. Because winter/spring			
5	precipitation may be stored as snowpack and soil moisture that carry over into subsequent			
6	seasons (Pal and Eltahir, 2001; Wu and Kinter III, 2009; Notaro and Zarrin, 2011), surface			
7	air temperature (SAT) response to precipitation may be lagged by several months. Therefore,			
8	we examined the correlations between December-May precipitation and JJA SAT over the			
9	Southwest. The results indicate that their correlations are generally insignificant			
10	(Supplementary Fig. 4C), suggesting that they may contribute independently to summer			
11	drought anomalies over the region.			
12	Both observed summer PDSI and tree-ring reconstructions cannot resolve summer			
13	monsoon rainfall over Southwest NA, and instead they are mainly tuned to winter/spring			
14	precipitation (Supplementary Fig. 3). Probable reasons include the coarse resolution of the			
15	PDSI grid, high evaporation loss in summer, and the nature of higher inter-annual rainfall			
16	variability in winter than in summer (St George et al., 2010). Regardless, NADA is capable			
17	of tracking winter precipitation that is most strongly related to El Niño variability. These			
18	results are consistent with the fact that ENSO teleconnections via the Pacific-North America			
19	(PNA) pattern are strongest in boreal winter (Trenberth et al., 1998; Alexander et al., 2002;			
20	Johnson and Feldstein, 2010).			
21	Because of weak correlations between December-May precipitation and JJA SAT, the			

22 possibility of direct teleconnections from the tropical Pacific to the Southwest during the

23 ENSO decay spring/summer cannot be completely ruled out and needs further investigation.

1	Precipitation correlations with NADA PC1 remain high (>0.3) in the southern Rockies in			
2	AMJ (Fig. 11C), and are marginally significant (~0.3) on the Great Plains in JJA (Fig. 11D).			
3	Further studies are necessary to determine the relative importance of direct tropical			
4	teleconnection versus soil moisture feedback for precipitation anomalies in spring/summer			
5	of the ENSO decay year.			
6				
7	6. Covariability between summer droughts over MA and NA			
8	Recent studies indicate that there exists concomitant summer drought variability			
9	between MA and NA (e.g., Lau and Weng, 2002; Hoerling and Kumar, 2003; Seager et al.,			
10	2005; Zhao et al., 2011). However, these studies are based on instrumental or reanalysis data,			
11	too short to characterize their long-term relationship. Here we employ MADA and NADA to			
12	investigate covariability of the dominant drought patterns over MA and NA for the past			
13	seven centuries, and discuss their relationship in the context of ENSO forcing.			
14	The concurrent correlations of MADA PC2 with the other three PCs over 1300-2005 are			
15	nearly zero, largely because MADA PC2 is related to the developing phase of the following			
16	El Niño event, whereas the other three PCs are related to the peak phase of the preceding El			
17	Niño event (section 4). In light of this, we shifted MADA PC2 one year forward so that it			
18	correlates with the same ENSO events as other PCs. Accordingly, common period for their			
19	cross-correlations is adjusted to 1301-2005.			
20	NADA PC1 is positively correlated with MADA PC1 and negatively with MADA PC2			
21	(Table 1), suggesting that there are concomitant pluvials over Central Asia/Southwest NA			
22	and droughts over the Maritime Continent/South Asia when east-central tropical Pacific			
23	SSTs are warmer than normal. However, as revealed by their 31-year running correlations,			

1	the covariability of summer droughts between MA and Southwest NA is not always
2	significant over the past seven centuries, with marked modulation at interdecadal timescales
3	(Fig. 12A). A comparison with the reconstructed ENSO variance series (Fig. 12B; Li et al.,
4	2013) indicates that their covariability is generally strong (weak) when the ENSO variance
5	is high (low), with each high/low-coherence epoch lasting for several decades. Therefore,
6	the covariability of summer droughts between MA and Southwest NA is likely modulated
7	by interdecadal changes in ENSO variance and associated teleconnections.
8	In contrast, the correlations of NADA PC2 with MADA PC1/PC2 are generally weak
9	(Table 1), so are their 31-year running correlations (results not shown). These results suggest
10	that there are no strong covariability between summer droughts in MA and the
11	PNW/TexMex region over the past seven centuries, despite a few coincident droughts and
12	pluvials in individual years or short intervals (Hoerling and Kumar, 2003; Cook et al., 2010;
13	Zhao et al., 2011). These results are consistent with our early findings that the ENSO
14	teleconnections are stationary over MA but non-stationary over the PNW/TexMex region
15	(Li et al., 2013). Therefore, the concurrent drought anomalies in MA and the PNW/TexMex
16	region are non-stationary and may only occur when the ENSO teleconnections are strong in
17	both regions.
18	
19	7. Summary

We have performed a pioneering study to examine ENSO phases embedded in summer droughts over MA and NA by employing the tree-ring based MADA and NADA. Our results indicate that summer droughts over MA and NA display distinct modes, each responding to a distinct phase of ENSO. For Central/South Asia, summer droughts are most

1	correlated with tropical Pacific SSTs at the peak phase of ENSO in prior winter (DJF).			
2	Regional precipitation and temperature anomalies occur one or two seasons later in late			
3	spring to early summer. The time delay indicates that the ENSO effects are indirect and are			
4	probably due to the Indian Ocean capacitor effect. For the Maritime Continent, the			
5	developing phase ENSO affects summer droughts most, through concurrent precipitation			
6	and temperature anomalies. For Southwest NA, the peak phase ENSO in prior winter has the			
7	most significant influence on summer droughts through winter precipitation. For the			
8	PNW/TexMex region, the peak phase ENSO in prior winter has a weak influence on			
9	summer droughts but its influence on winter precipitation is non-stationary. The ENSO			
10	effects persist through spring/summer over most of western NA, owing to the varied			
11	contributions of direct tropical teleconnection and regional snowpack-soil moisture feedback			
12	in ENSO decay seasons. Further studies are necessary to determine their relative importance			
13	over the region.			
14	The above results illustrate the importance of determining the time when precipitation			
15	and temperature imprint on summer droughts. For example, although summer droughts over			
16	Central/South Asia and Southwest NA are both most related to the peak phase of ENSO in			
17	the preceding winter, the timing of strongest precipitation anomalies leads us to conclude			
18	that those in Southwest NA record ENSO teleconnections at the peak phase of ENSO, while			
19	those in Central/South Asia represent tropical-wide SST conditions at the decay phase of			
20	ENSO.			
21	Our findings reported here provide a framework for future efforts aiming at			
22	reconstructing ENSO variability more accurately. For example, tree-rings in Southwest NA			
23	might be more useful for reconstructing the peak phase ENSO variability, whereas those in			

1	the Maritime Continent might be more appropriate for reconstructing the developing phase
2	ENSO variability. If tree-rings in MA and NA are to be jointly used in studying ENSO it is
3	important to shift by one year those from the Maritime Continent as they respond to the
4	ENSO events that develop in the year of tree-ring formation, whereas those in Central/South
5	Asia and Southwest NA respond to the ENSO events that decay in the year of tree-ring
6	formation. In light of this, more assessment of ENSO phases encoded in tree-rings and
7	other proxies around the world is necessary and shall lead to more skillful multi-proxy
8	reconstructions that will improve our understanding of the complex ENSO system.
9	
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10 11 12 13 14 15 16 17	Acknowledgements We acknowledge the researchers who have contributed their tree-ring data for the development of MADA and NADA. This research was funded by the National Science Foundation, the National Basic Research Program of China (2012CB955600), the National Oceanic and Atmospheric Administration, the Japan Agency for Marine-Earth Science and Technology, and the Hui Oi-Chow Trust Fund. This is an International Pacific Research Center/School of Ocean and Earth Science and Technology Contribution (######) and a Lamont–Doherty Earth Observatory Contribution (#####)

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Figure Captions:

2	Fig. 1. Composite mean and standard deviation of monthly Niño3.4 indices, based on SST			
3	anomalies for 20 observed El Niño events during 1950-2011. El Niño events are			
4	defined according to the criteria that maximum warming in the Niño3.4 region			
5	during the months of November-January (NDJ) exceeds 0.5°C. Anomalies are			
6	relative to the 1971-2000 base period.			
7	Fig. 2. Map of Asia and North America showing the locations of the reconstructed PDSI			
8	grid points in MADA and NADA.			
9	Fig. 3. Number (A) and percentage (B) of reconstructed grip points available back in time i			
10	MADA and NADA.			
11	Fig. 4. Spatial pattern of the first two EOFs of MADA and NADA.			
12	Fig. 5. Principal components of the first and second EOF mode for MADA and NADA			
13	during 1300-2005. Bold line denotes a 15-year low-pass filter for each series.			
14	Fig. 6. Spatial correlations with global SSTs for MADA PC1 (A-F) and MADA PC2 (G-L)			
15	during 1870-2005. Left (right) box in each panel denotes the Niño4 (Niño3) region,			
16	respectively. The correlation coefficient at the 0.05 significance level is about 0.17,			
17	based on a two-tailed student's t-test.			
18	Fig. 7. Same as in Fig. 6, but for NADA PC1 (A-F) and NADA PC2 (G-L).			
19	Fig. 8. Lead-lag correlations with Niño3 (blue) and Niño4 (red) index for MADA PC1 (A),			
20	MADA PC2 (B), and NADA PC1 (C). The common period for calculating the			
21	correlations is 1870-2005. The dashed lines denote the 0.05 significance level, based			
22	on a two-tailed student's t-test.			

1	Fig. 9. Spatial correlations of MADA PC1 with seasonal precipitation (A-F) and temperature			
2	(G-L) during 1951-2005. Contours overlapped in A-F denote the ratio of seasonal to			
3	annual rainfall over the region. The correlation coefficient at the 0.05 significance			
4	level is about 0.27, based on a two-tailed student's t-test.			
5	Fig. 10. Same as in Fig. 9, but for MADA PC2.			
6	Fig. 11. Same as in Fig. 9, but for NADA PC1.			
7	Fig. 12. Running 31-year Spearman's rank correlations of NADA PC1 with MADA			
8	PC1/PC2 over 1301-2005. Note that the y-axis for MADA PC2 was flipped in order			
9	to make its correlations visually comparable to those of MADA PC1. The dashed			
10	line denotes the 0.05 significance level, based on a two-tailed student's t-test.			

3 Table 1. Cross correlations of the MADA and NADA PCs over the common period 1301-

4 2005. Bold values are significant at the 0.05 levels, based on a two-tailed student's t-test.

	MADA PC1	MADA PC2	NADA PC1	NADA PC2
MADA PC1	1.00			
MADA PC2	-0.33	1.00		
NADA PC1	0.18	-0.22	1.00	
NADA PC2	-0.10	0.06	0.00	1.00











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Figure 4

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- 1 Figure 6





Figure 7



- 1 Figure 8









