1	Spring Onset Variations and Trends in the Continental USA:				
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3	Past and Regional Assessment				
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5	Using Temperature-Based Indices				
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41 Abstract

42 Phenological data are simple yet sensitive indicators of the impacts of climate change on 43 ecosystems, but observations have not been made routinely or extensively enough to evaluate 44 spatial and temporal patterns across most continents, including North America. As an 45 alternative, many studies use easily calculated and weather-based algorithms linked to a specific 46 phenological response, for example the seasonal accumulation of growing degree-hours that 47 triggers the onset of leafout/flowering in many plants. The Spring Indices (SI, Schwartz et al. 48 2006) are one set of phenological models that have been successfully applied to evaluate 49 variations and trends in the onset of spring across the Northern Hemisphere's temperate regions. 50 To date, SI have been limited by only producing output in locations where the plants' SI were 51 designed to simulate grow successfully, principally where both the plants' chilling and growing-52 degree-hour requirements are met. In this paper, we consider an extended form of the Spring 53 Indices (abbreviated SI-x) that can be mapped from polar to subtropical latitudes by ignoring 54 chilling requirements while still retaining the utility and accuracy of the original SI (now 55 abbreviated SI-o). For the continental USA, SI-x variations from 1900-2010 show an abrupt and 56 sustained delay in spring onset of about 4-8 days around 1958 in parts of the Southeast and 57 southern Great Plains, and a comparable advance of 4-8 days in parts of the northern Great 58 Plains and the West. These conspicuous regional patterns are associated with the Pacific North 59 American (PNA) pattern, defined by anomalously high geopotential heights over the Northwest 60 and anomalously low ones in the Southeast. The SI-x are promising metrics for tracking regional 61 variations and trends in spring's onset, relating them to relevant ecological, hydrological and 62 socioeconomic phenomena, and exploring connections between atmospheric drivers and seasonal 63 timing, both in the past and elsewhere in the world.

64

## 64 Introduction

65 Phenology is the study of plant and animal life cycle events in relation to environmental drivers (especially weather and climate), and phenological data are simple, yet sensitive 66 67 indicators of the impacts of climate change on ecosystems (IPCC 2007). Phenological 68 measurements are made routinely and extensively in Europe (van Vliet et al. 2003), new national 69 networks were established recently in the USA (Betancourt et al. 2005) and Australia 70 (ClimateWatch 2012), and pleas are being made for developing networks in India and other 71 continents (Kushwaga and Singh 2008). Considerable challenges remain, however, in using 72 phenological data to assess the environmental impacts of climate variability and change across 73 most regions (Schwartz et al. 2006). Chief among these are the lack of: 1) historical and 74 contemporary phenological data in general; 2) long-term and replicated measurements of 75 different populations across the range of the target species; 3) coordination and standardization 76 among existing national phenological networks in terms of species and protocols; and 4) 77 worldwide phenological data sharing agreements.

78 Given these limitations, many researchers have used available phenological data to first 79 develop biologically-relevant algorithms for simulating "spring's onset", typically driven by 80 daily surface maximum-minimum air temperatures. Once tested and calibrated, such models 81 extend the possible spatial coverage and temporal range of phenological assessments of 82 environmental change, given the greater availability of meteorological data, both currently and in 83 the past. Now it would be fair at this point to ask "Why use phenological models instead of just 84 using the meteorological data alone for such assessments?" In order to answer this question, one 85 must consider that, when measuring environmental change, there are various levels of precision 86 related to the type of measure used, the length of time addressed, and the degree of spatial

87 aggregation. Let us consider changes in the start of the growing season for plants. Average 88 monthly (or seasonal) temperatures can give a general idea of the expected change at a specific 89 station and of the overall average change over a region. However, monthly values will not be as 90 responsive as a model designed to produce a precise output related to a specific phenological 91 response, for example the initiation of leafing and flowering. Also, if the phenological event 92 ranges over a broad geographic area or can be triggered by a brief period of extreme 93 temperatures, this may be poorly represented in general measures like average monthly or 94 seasonal temperature.

95 One set of phenological models that have been successfully applied to assess the impact 96 of environmental change on the onset of the spring season across temperate regions around the 97 Northern Hemisphere are the Spring Indices (SI, Schwartz et al. 2006; Ault et al. 2011; McCabe 98 et al. 2011). This suite of measures includes several sub-models and associated measures, all of 99 which can be calculated using daily maximum-minimum surface (shelter-height) temperatures 100 and station latitude. SI process weather data into a form comparable to the spring growth of 101 plants that are not water limited and are responsive to temperature increases (Schwartz et al. 102 2006).

103 SI were initially designed to simulate the growth of specific plants. As such, earlier 104 versions of SI do not produce output in locations where these plants do not grow successfully, 105 most specifically in areas where warm winter weather provides inadequate chilling (Schwartz et 106 al. 2006). Here, we explore development of an extended form of the Spring Indices (abbreviated 107 SI-x) that retains the utility and accuracy of the original SI (now abbreviated SI-o) while 108 allowing mapping into the subtropics. This permits assessment of spring onset variations and 109 trends in the Southeastern USA, particularly in reference to this region constituting a "warming

- 110 hole," where the secular trend during the past century has been towards later hard freezes
- 111 (Marino et al. 2011) and generally cooler springs and summers (Robinson et al. 2002; Pan et al.
- 112 2004; Kunkel et al. 2006; Wang et al. 2009; Meehl et al., in review).
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## 113 Data and Methodology

The meteorological stations used in this study came from the over 22,000 observation sites that record standard surface-level (1.5 m above the surface) daily maximum-minimum across the continental (lower-48) United States. The data were obtained from the National Climatic Data Center (NCDC) archives, covering the period of record for these stations through 2010. The final 716 station locations selected for inclusion in the analyses were those that had sufficient data to produce valid Spring Indices (SI) model output for at least twenty-five of thirty years over the 1981-2010 (30-year) period.

121 The methodology for producing the extended SI (SI-x) model output (SI first leaf date 122 and SI first bloom date) are the same as described in McCabe et al. (2011) for the original SI 123 models (SI-o) with the following exception. For SI-x first leaf calculation, no chilling hours are 124 accumulated, rather energy accumulation starts for all stations from the same fixed day, January 1<sup>st</sup>, each year. For the selected stations, SI-x first leaf and first bloom model dates were first 125 126 calculated for the station period of record. Next, from these yearly values, 30-year averages 127 ("normals") were calculated for the 1981-2010 period, and these normals were subsequently 128 used to turn the yearly SI-x output into departures-from-normal. These departures were used for 129 all subsequent analyses.

In addition, given the well-document performance of SI-o, SI-x output was compared to SI-o station output at all available stations where both model sets would produce output. The comparisons included Pearson's correlation, mean bias, and mean absolute differences. Lastly, for stations where cloned lilac (*Syringa chinensis* 'Red Rothomagensis' data—the main type of plants used in the original development of SI) were also available, these data were used to compare SI-x model and SI-x model accuracy.

136 The SI-x station departures from normal were accumulated, examined, and plotted over 137 the 1900-2010 period. This initial examination suggested that the time series was different in the 138 Southeastern United States (SEUS), than the rest of the continental USA (REST). Further, it 139 appeared that the decade from 1951-1960 was a pivotal period when broad changes appeared to 140 be taking place in the previous trends. To further explore these changes: 1) the temporal trends 141 were accumulated in two regions, the SEUS (defined as the area south of 37°N latitude and west 142 of 103°W longitude) and REST; and 2) changes in SI-x output values were compared by station 143 between the 1951-1960 and 2001-2010 periods, for all stations that had at least eight years of 144 valid output in both periods. 145 To assess the role of large-scale circulation anomalies on the timing of spring in the

SEUS time series, we correlated it with January and February 300mb heights from the National Center for Environmental Prediction's (NCEP) reanalysis data (Kalnay et al. 1996). These 300mb fields were computed using a numerical model of climate constrained by observational data from 1950 through 2010, and hence provide insight into the dynamical mechanisms responsible for interannual variability in SI-x. 151 Results

Table 1 shows the results of the comparison of SI-x and SI-o output. Both the first leaf and first bloom models are highly correlated, and the mean bias and mean absolute differences are around 2 days or smaller, with the first bloom models values closer to one day. The comparison of SI-x and SI-o model performance when compared to cloned lilac data are very similar, in terms of both bias and absolute errors. The error differences are 0.25 days or less, well within the 1-day resolution of model predictions.

The temporal trends in SI-x first leaf date are considerably different between the SEUS and REST for the first half of the 20<sup>th</sup> century (1900-1950), but begin to converge in the late-160 1950s (Fig. 1). By the 1980s the two regions seems to have come into phase. The spatial 161 coherence across the SEUS is considerable, and well shown by the station comparison between 162 the 1951-1960 and 2001-2010 periods (Fig. 2).

163 Correlations between SEUS and 300mb heights for January and February are shown in 164 Figure 3. Although correlations are stronger during January (Fig. 3a), the sign and geographic 165 pattern of the correlation fields are very similar for both months. Regions of negative correlation 166 (early SEUS spring with high 300mb heights) occur over the subtropical Pacific and southeastern 167 US, whereas negative correlations occur over northern North America. The pattern during both 168 months is highly reminiscent of the Pacific North American (PNA) stationary wave pattern in 169 mid-tropospheric flow (Wallace and Gutzler 1981). Leathers and Palecki (1992) attributed a 170 sharp increase in the PNA index to the dramatic decline in geopotential heights over SEUS in the 171 late 1950's, which accounts for winter/spring cooling and earlier onset in SEUS.

## 172 **Discussion**

173 Schwartz et al. (2006) and Parmesan (2007) have documented that SI spring onset and 174 phenological trends for comparable species (shrubs) are both moving earlier at rates of 175 approximately 1.1 to 1.2 days/decade on average at the hemispheric scale. With respect to this 176 hemispheric average, trends in the western USA are anomalously negative, while trends in the 177 southeastern USA are anomalously positive. Biological evidence for the dramatic advance in SI-178 SEUS around the late 1950's includes delayed seasonal flowering in many Florida plants, 179 inferred from herbarium specimens (Von Holle et al. 2010). 180 Previous studies have implicated large-scale atmospheric circulation patterns in driving 181 interannual variability and trends in the western USA (Ault et al. 2011; McCabe et al. 2011). In 182 particular, Ault et al. (2011) argued that the atmospheric trends towards an enhanced ridge over 183 western North America, with troughs over the subtropical Pacific and southeastern US, were 184 linked to a greater number of warm days earlier in the year and hence earlier spring. This pattern, 185 which resembles the positive phase of the PNA, would also be expected to generate a greater 186 number of outbreaks of cold air in the southeastern USA and consequently delays the onset of 187 spring in that region (Marino et al. 2011). The positive trend in the SEUS time series and the 188 correlation map in Figure 3 both support this explanation. Hence, the anomalous (and opposing) 189 trend in the western USA and southeastern USA are counterbalanced and linked by the same 190 large-scale mechanism.

The geographic pattern of southeastern USA stations where spring has been arriving later is consistent with the well-documented "warming hole" in the southwestern USA (Robinson et al. 2002; Pan et al. 2004; Kunkel et al. 2006; Wang et al. 2009; Meehl et al., in review). Recently, Meehl et al. (in review) have attributed this warming hole to decadal variability in the

195 Pacific Ocean, which induces atmospheric changes favoring a trough-ridge-trough (positive 196 PNA) structure that brings a greater number of cold outbreaks of air to the southwestern USA. 197 Using a coupled global climate model (GCM), the study further documents that the pattern of 198 Pacific decadal variability responsible for the warming hole in North America may be internally 199 generated, and therefore not directly linked to climate change (Meehl et al., in review). Because 200 the warming hole also evidently impacts seasonality, as shown here, projections of future 201 phenological change should take into account both the forced and natural sources of variability. 202 Future studies could use the new SI-x product to further explore the patterns of variations 203 and trends and natural low-frequency variability of spring's onset in other parts of the world or 204 over other time domains. For example, the springtime cooling trend in southwest China has been 205 attributed to the teleconnection between the winter North Atlantic Oscillation (NAO) and surface 206 air temperature over the lee side of the Tibetan Plateau (Li et al. 2005). 207 We emphasize that the calculation of SI-x only require daily maximum-minimum

207 We emphasize that the calculation of SFX only require daily maximum minimum
208 temperatures as input, and so they could be calculated from daily reanalysis data and GCM
209 output to develop a more refined dynamical explanation for the sources of spring onset
210 variability on interannual to centennial timescales. Such efforts would provide insight into the
211 sources of spring onset predictability, which in turn could be of critical importance to
212 agricultural and natural resource managers alike.

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265	Table 1: Comparison of Original Spring Indices (SI-o)						
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267	to Extended Spring Indices (81-x)						
268	and both to liles phonological data						
209	and both to mac phenological data						
270 271 272	Pearson's correlation	SI-x first leaf	date SI-x first b	loom date			
272 273 274	SI-o first leaf date	0.975	5				
275 276	SI-o first bloom date		0.9	95			
277 278 279	Mean Difference (days)	Bias differenc	ce Absolute d	ifference			
280 281	SI-x first leaf date	-1.4	2.5	5			
282 283	SI-x first bloom date	-0.7	1.1				
284 285	n of cases = 71,926						
286 287 288	Mean Error to Lilac (days)	Bias Error	Absolute Error	Error Difference			
289 290	SI-x first leaf date	-2.47	6.57	0.14			
291 292	SI-o first leaf date	-1.78	6.43				
293 294	SI-x first bloom date	-3.66	5.46	0.25			
295 296	SI-o first bloom date	-3.15	5.22				
297 298	n of cases = 830						
299							







