

The Circumglobal North American wave pattern and its relation to North American cold events

Nili Harnik^{1,2*}, Gabriele Messori^{2,3}, Rodrigo Caballero^{2,4}, Steven B. Feldstein⁵

¹Department of Geophysics and Planetary Sciences, Tel Aviv University, Tel Aviv, Israel

²Department of Meteorology, Stockholm University, Stockholm, Sweden

³Met Office Hadley Centre, Exeter, UK

⁴Bolin center for Climate Research, Stockholm University, Stockholm, Sweden

⁵Pennsylvania State University, PA, USA

Key Points:

- North American cold events are associated with the Circumglobal Teleconnection Pattern 2
- PNA events are related to Circumglobal Teleconnection Pattern 1 and are in quadrature with pattern 2
- The peak cold lags the upper level North American wave by 3 days, and an Asian wave by about 1 week

*Department of Geosciences, Tel Aviv University, Tel Aviv, Israel, currently on sabbatical at the Meteorology Department, Stockholm University

Corresponding author: Nili Harnik, harnik@tau.ac.il

Abstract

Extreme large scale North American cold events are associated with strong undulations in the tropospheric jet stream which bring cold polar air southward over the continent. Here we propose that these jet undulations are associated with the North American part of the Circumglobal Teleconnection Pattern - a pair of zonally oriented waves of zonal wavenumber 5 which are in zonal quadrature with each other. While the PNA is associated with the first circumglobal wave pattern, North American extreme cold events are associated with the second pattern. The 300–*hPa* meridional wind and surface temperature anomalies associated with the Circumglobal North American wave packet are similar to those associated with the strongest Eastern US cold events. Both types of events are associated with a wave packet propagating all the way from Asia over the Pacific and across North America, with anomalous cold temperatures spreading south-eastwards from Canada over the continent.

1 Introduction

The tropospheric polar vortex has gained unprecedented media attention during the past few winters, when large undulations in the jet stream formed over North America along with extreme winter conditions. This was most notable during the winter of 2013-14 [e.g. *Wallace et al.*, 2014; *Baxter and Nigam*, 2015; *Davies*, 2015; *Lee et al.*, 2015; *Yu and Zhang*, 2015; *Waugh et al.*, 2016; *Watson et al.*, 2016], but also during winter 2014-15 and February 2016¹). The occurrence of such severe cold events during recent years which are globally amongst the warmest on record points to the differences between regional and global climate change; specifically, it highlights the need to fully understand the physical drivers of the events and their representation in climate models in order to robustly predict possible future changes in their frequency and intensity.

The dynamical drivers of cold events over North America have been studied extensively, using a diverse range of definitions for the cold events themselves [e.g. *Konrad*, 1996; *Walsh et al.*, 2001; *Portis et al.*, 2006; *Loikith and Broccoli*, 2012; *Grotjahn et al.*, 2015; *Messori et al.*, 2016, and references therein]. A robust finding of these studies is the association of North American cold events with large scale circulation anomalies which advect very cold air equatorward, typically from the northwest. A common finding is also that these large scale circulation anomalies are similar in scale and structure to the Pacific/North American (PNA) pattern, but the two patterns are zonally shifted, and thus project weakly onto each other [e.g. *Walsh et al.* [2001]; *Cellitti et al.* [2006]; *Grotjahn et al.* [2015]; *Messori et al.* [2016] also *Davies*, 2015, for winter 2013-14]. On the other hand, *Linkin and Nigam* [2008] noted a relation between North American cold air outbreaks and the North Pacific Oscillation [NPO, *Walker and Bliss*, 1932] and its associated upper level West Pacific (WP) teleconnection pattern.

Here we propose that the large scale circulation anomalies which drive cold events over North America east of the Rockies are associated with the Circumglobal Teleconnection Pattern (CTP) introduced by *Branstator* [2002]. *Branstator* [2002] obtained the CTP from the first two EOFs of monthly mean seasonal anomalies of Dec-Feb 300–*hPa* nondivergent meridional wind. The two patterns together represent a zonally oriented medium scale wave train of zonal wave 5 with arbitrary zonal phasing. *Feldstein and Dayan* [2008] showed that on daily time scales, these patterns appear as localized wave packets which propagate downstream with an eastward group speed and near zero phase speed [see also *Watanabe*, 2004; *Yuan et al.*, 2011]. These yield a circumglobal wave pattern when averaged over time [*Feldstein and Dayan*, 2008].

The zonal scale of the CTP is similar to the jet stream undulations as seen in the 2013-14 winter (the longitudinal span of the North American landmass is roughly zonal wavenumber 5), and we may expect its quasi-stationary nature to allow for persistent advection of cold

¹ See http://www.cpc.ncep.noaa.gov/products/CDB/CDB_Archive.html/CDB_archive.shtml

63 polar air across the continent. It is therefore plausible that the CTP may play a role in driv-
 64 ing North American extreme cold events. The CTP has been associated with extreme weather
 65 during summer [Schubert *et al.*, 2011; Teng *et al.*, 2013], and with winter precipitation over
 66 Israel [Feldstein and Dayan, 2008], but has not been studied in the context of North Amer-
 67 ican winter weather. Recently, Messori *et al.* [2016] composited extreme Eastern US cold events,
 68 and showed a zonally oriented wave pattern with a wavelength corresponding roughly to zonal
 69 wave number 5 and a phasing similar to the second CTP pattern, which propagates from the
 70 Pacific over North America to the Atlantic (see their Figure 2). In this paper, we explicitly ex-
 71 amine the relation of this wave pattern to the second CTP, by comparing surface temperature
 72 and upper level flow fields during times when the flow projects strongly onto the North Amer-
 73 ican part of the CTP, and during extreme cold events over the eastern US. The linking of the
 74 two phenomena has possible implications for predictability [Teng *et al.*, 2013; Grazzini and
 75 Vitart, 2015], which we will briefly explore by examining the upstream origins and precur-
 76 sors of the North American part of the CTP.

77 2 Data and analysis methods

78 We use daily anomalies, defined by removing a smoothed daily climatology (using a 21
 79 day running mean). The events are calculated using daily mean and monthly mean gridded
 80 fields from ERA-Interim [Dee *et al.*, 2011] for 1979-2014. Much of the analysis for the CTP
 81 based events was repeated using NCEP I reanalysis [Kalnay *et al.*, 1996] from 1958-2015 and
 82 except where noted, the results are similar.

83 The statistical significance of the composites is calculated both using a 500 member boot
 84 strap method and a sign-test which indicates where a certain percentage of composite mem-
 85 bers have the same sign as the composite mean. The chances for a given percentage of events
 86 to have the same sign as the composite mean is determined using a binomial formula, assum-
 87 ing equal chances for positive and negative anomalies. Spatial correlations between two pat-
 88 terns are calculated after weighting each by square root of cosine latitude, and the statistical
 89 significance is estimated by correlating one of the patterns with the corresponding field from
 90 1000 randomly chosen days. We define several kinds of events, as follows:

91 2.1 The Circumglobal North American (CNA) wave pattern

92 To obtain a regional CTP index, we begin by calculating the global CTP patterns using
 93 a method similar to *Branstator* [2002]. The global CTPs are the first two EOFs of the win-
 94 ter anomalies of monthly mean 300-hPa meridional wind, calculated by removing each sea-
 95 son's Dec-Feb mean fields from the monthly mean meridional wind fields. A square root of
 96 cosine latitude weighting is used for the analysis (so that the variance is weighted by a co-
 97 sine latitude). Unlike *Branstator* [2002], we use the full meridional wind, rather than non di-
 98 vergent meridional wind, since the two give very similar results. We note that without remov-
 99 ing the seasonal mean anomalies, the first EOF has a planetary scale, and using daily fields
 100 gives a pair of zonal wavenumber 6 patterns for the first EOFs .

101 Using monthly ERA-Interim data, the first two EOFs (shading, Figure 1) are a pair of
 102 quasi zonal wave number 5 patterns which explain 13.3 and 11.5 percent of the variance, and
 103 are not well separated according to the *North et al.* [1982] criterion, suggesting that they span
 104 a continuum of zonal wave number 5 patterns with arbitrary phase. The robustness of these
 105 patterns is examined using NCEP and discussed in the supplementary information (Figure S1).
 106 Following *Feldstein and Dayan* [2008], we obtain daily regional CTP indices by projecting the
 107 daily 300-hPa meridional wind anomalies onto specific zonal sectors of the global CTPs
 108 (see supplementary information for details). For the North American region, we choose the
 109 domain 180–324°E, 10–85°N (see the boxes in Figure 2c,d), where the longitudinal span
 110 of 144° was chosen to capture two full wavelengths of zonal wave 5. We then project the daily
 111 data onto the second CTP pattern in this region. We refer to this pattern as the Circumglobal
 112 North American (CNA) pattern, and to its normalized projection as the CNA index.

To define positive CNA events, we find the days on which the 5-day running mean of the CNA index exceeds its mean by one standard deviation. We define the peak value to be the event center, with events being separated by at least one day in which the index drops below one half a standard deviation above the mean. We find 78 positive events for ERA-Interim (and 123 events for NCEP). We do not impose any additional time separation, but find that the vast majority of events are separated by 7 or more days.

2.2 The Circumglobal Eurasian wave pattern

To examine the possibility of precursor patterns, we repeat the procedure done for CNA events, using a Eurasian region: $0 - 144^{\circ}E$, $10 - 85^{\circ}N$ (see the box in Figure 4a), and define events in a similar manner. We find 79 events using ERA-Interim and 139 events in NCEP.

2.3 Eastern United States Cold (USC) events

We rank Eastern United States Cold (USC) events based on the area weighted 2-meter temperature anomaly over the region $100 - 70^{\circ}W$, $30 - 45^{\circ}N$ (see box in Figure 2f). This is the same region used by *Messori et al.* [2016]. The normalized area-weighted temperature time series is referred to as the USC index. To choose USC events we smooth the temperature anomalies using a 5-day running mean, and choose the days with the largest negative temperature anomaly. We discard events which are closer than 7 days to another event, keeping the colder event of the two. For comparison with the CNA event composites, we choose the 78 coldest events, which consist of an area-weighted anomaly below $-4.7K$.

2.4 Daily Pacific North American (PNA) pattern

We calculate a daily Pacific North American (PNA) index by combining the standardised 500 hPa geopotential height anomalies at (20N, 160W), (45N, 165W), (55N, 115W) and (30N, 85W), following *Wallace and Gutzler* [1981] and *Cellitti et al.* [2006], and normalizing the resulting index to have unit standard deviation. Positive events are defined by the index exceeding 1.0.

3 Results

Figure 1 shows the first two CTP patterns of anomalous $300-hPa$ meridional wind, along with the anomalous $300-hPa$ meridional wind composites for day 0 positive PNA events and day -3 USC events. We note that the first CTP in its positive phase corresponds to anomalous poleward flow over Alaska, with anomalous equatorward flow on its sides. This pattern fits the positive phase of the PNA (compare the shading and contours in the left panel). The meridional wind anomaly associated with the coldest 78 USC events, on the other hand is in phase with the second CTP pattern (compare shading and contours in the right panel). This motivates us to further examine the relation between USC events and times when the flow projects strongly onto the second CTP over North America (which we have defined as CNA events). Before doing so, we note that the two CTPs are roughly in quadrature over the Pacific-North American region. Since the PNA projects strongly onto the first CTP while USC events project onto the second CTP, we expect the simultaneous correlation between the PNA and USC indices to be weak [as was found by *Cellitti et al.*, 2006; *Grotjahn et al.*, 2015; *Messori et al.*, 2016]. Indeed, the correlation between the PNA and USC indices reaches only -0.26 (at lag 0), compared to a correlation of 0.73 between the PNA and an index similar to the CNA but based on the first, rather than the second CTP.

Figure 2 shows the time lagged composites of $300-hPa$ meridional wind anomaly overlain on the full CTP EOF2, for CNA and USC events. The black boxes show the projection region used for defining the CNA events. The highest spatial correlation between the CNA and USC composites, with a value of 0.86 (statistically significant above the 99.9% level), is found

159 when USC events lag the CNA events by 3 days, thus we show the composites with a 3 day
 160 offset. We see a very clear similarity between the meridional wind pattern for both types of
 161 events. The patterns show a localized wavy pattern, with the wave packet amplitude propa-
 162 gating in time from Asia, over the Pacific and onto North America, but individual positive and
 163 negative centers being quasi-stationary. In both cases, we see a precursor wave packet over
 164 Asia with a clear branch along the subtropical jet and another more northward branch. The
 165 wave packets span roughly 3 full wavelengths at CNA day 0, their amplitude peaks shift east-
 166 ward with a group speed of about 20° longitude per day corresponding to about one week to
 167 travel from the Western Pacific to the Atlantic, and they have an almost zero eastward phase
 168 speed. This yields an equatorward flow over the northeastern North American continent, which
 169 is strong and significant for more than one week, making these quasi-stationary wave pack-
 170 ets efficient in driving temperature anomalies.

171 Figure 3 shows the surface temperature composites for both kinds of events (right and
 172 left columns), with the USC fields shifted 3 days earlier than the CNA events. We see a strong
 173 similarity between the two types of events with the peak cold anomaly covering most of North
 174 America and a warm anomaly over Alaska and the Bering sea, at USC lag 0, CNA lag 3. At
 175 these times, both anomalies reach their coldest values over North America, and the spatial cor-
 176 relation between the two fields is maximal (0.85, statistically significant above the 99.9% level).
 177 The time evolution is also quite similar, with the peak USC cold anomaly starting from the
 178 northwest, spreading over most of the continent and shifting towards the east coast. The CNA
 179 cold anomalies do not shift meridionally in time as much as the USC events, and they are pre-
 180 ceded by stronger anomalies over the Pacific Ocean and over Asia at early time lags, suggest-
 181 ing the Asian wave packet precursor temperature signal is more robust for CNA events. We
 182 note that the NCEP 2-meter temperature, which is derived differently from ERA Interim, shows
 183 slightly weaker and shorter lasting cold anomalies for the CNA events (Figure S2), and the
 184 ocean 2-meter temperature anomalies are absent.

185 Also shown (middle column) is the number of CNA composite members which have the
 186 same sign as the composite mean at each grid point (the *hit-rate*). Much of the statistically
 187 significant cold anomalies occur in more than 65% of the CNA events, and at the peak cold
 188 anomaly, in more than 75% and even 85% of the events (the chances of this happening ran-
 189 domly are much less than 1%). Of the 76 CNA events (we exclude two because at lag -7 they
 190 overlap with the peak of the preceding event), 30 are followed within 7 days by one of the
 191 78 coldest spells identified here, exceeding the 95% statistical significance level obtained from
 192 random sampling. It is found that 44 of the events are followed by a cold event with temper-
 193 atures below the 5th percentile of the full wintertime distribution, again above the 95% sta-
 194 tistical significance level. The time correlation between the daily Dec-Feb (DJF) USC and CNA
 195 indices peaks when the US cold conditions lag the CNA pattern by 3 days, and is -0.45 (sta-
 196 tistically significant at the 99.99% level, assuming a conservative 3 degrees of freedom per win-
 197 ter). We further find that 36.9% of the days on which the USC index is less than -1.0 are pre-
 198 ceded 3 days earlier by a CNA index above 1.0 standard deviation, and 62% are preceded by
 199 a CNA projection above 0.5 standard deviations (see the conditional CNA distribution for all
 200 days vs strongly negative USC days in Figure S3). During the 2013-14 winter there was a suc-
 201 cession of periods with a strong positive projection onto the CNA pattern, followed by extreme
 202 cold anomalies and negative USC values (Figure S4). Explicitly, the DJF mean normalised USC
 203 index was -0.6 with 37 days (41%) with a USC index below -1.0 standard deviation, compared
 204 to 5.9% of all 1980-2014 winter (DJF) days. The DJF mean normalised CNA index was 0.74
 205 with 32 days (36%) with a CNA index larger than 1.0 standard deviation, compared to 5.3%
 206 of all 1980-2014 winter days.

207 The precursor Asian wave packet found in Figure 2 suggests a predictability pathway
 208 on a 7-10 day time scale for eastern North American cold events. To check this further we ex-
 209 amine composites based on Eurasian CTP2 events, defined by the black rectangular region marked
 210 in the top left panel of Figure 4. We see a clear Eurasian wave packet at day zero, with cor-
 211 responding statistically significant temperature anomalies being cold over the Middle East and

212 Siberia, and warm over Central Asia (not shown). There is a clear downstream group prop-
 213 agation over the Pacific and onto North America, leading to persistent and significant northerly
 214 flow anomalies over eastern North America, in phase with CTP2. Consistently, after lag 7 we
 215 see a gradual development of significant cold anomalies over the eastern US, reaching a peak
 216 around day 11, in some regions occurring in more than 67% of composite members (these anoma-
 217 lies are also found in NCEP though in a smaller more poleward region, Figure S5). We fur-
 218 ther count how many Eurasian wave packet events were followed by one of the 78 coldest USC
 219 events, 8-12 days later. This choice of range of lags is made to in order to include both day
 220 9, when the spatial correlation with the corresponding surface temperature anomalies over the
 221 CNA area is highest (0.7, statistically significant at the 99% level), and day 11, when the strongest
 222 temperature anomalies (Figure 4f) take place. We find this happens for 9 of the 77 Eurasian
 223 wave packets (excluding two which overlap at lag +11 of the peak of the preceding event),
 224 which is exactly the 90th percent statistical significance level.

225 4 Discussion

226 We have shown that a large number of the extreme North American cold events have
 227 been driven by jet stream undulations similar to the CNA pattern. The statistical link between
 228 the CNA and North American cold events is significant and robust, though not all days with
 229 a strong projection onto the CNA lead to an extreme USC event a few days later, and not all
 230 extreme USC events are preceded by a strong projection onto the CNA. It does point, how-
 231 ever, to a potential predictability pathway on 7-10 days, resulting from downstream propaga-
 232 tion of the waves from Asia, and identifies potential precursor patterns. *Grazzini and Vitart [2015]*
 233 used an objective wave packet tracking algorithm and a forecast verification database and found
 234 increased predictability for long lasting wave packets originating in the West Pacific. They did
 235 not differentiate, however, between synoptic scale propagating waves and medium scale quasi
 236 stationary waves like the CTP, but it is very probable that these different types of wave pack-
 237 ets affect predictability differently.

238 *Bao and Wallace [2015]* recently performed a cluster analysis of Northern Hemisphere
 239 10 day low pass filtered 500 *hPa* geopotential heights. They found four reproducible patterns,
 240 three of which they related to the NAO and PNA phases. Their second pattern, which was sug-
 241 gestive of Alaska blocking with a downstream wave train extension over North America, is
 242 similar to the positive CNA pattern. More examination is required to establish this connec-
 243 tion robustly, but it suggests that the CNA may be a dominant recurring pattern of variabil-
 244 ity of the Northern Hemisphere wintertime flow.

245 An explicit examination of the 2013-14 winter shows that the projection onto the pos-
 246 itive phase of the CNA was strong during many of the days that winter, suggesting that the
 247 cold events were associated with this flow pattern. This is consistent with *Wang et al. [2014]*
 248 and *Davies [2015]* who also noted a wave train propagating from the Pacific, resulting in a
 249 strong trough-ridge anomaly over the Gulf of Alaska-Great Lakes. *Linkin and Nigam [2008]*
 250 related the NPO/WP pattern to North American cold air outbreaks, and *Baxter and Nigam [2015]*
 251 and *Yu and Zhang [2015]* associated it with the winter of 2013-14. A comparison of the NPO/WP
 252 pattern [see e.g. Figure 3a of *Linkin and Nigam, 2008*] with the CNA suggests that they project
 253 strongly onto each other in some regions but the two are not the same.

254 A few recent studies of the extreme North American winter of 2013-14 have suggested
 255 that the anomalous ridge-trough pattern was forced by the anomalous Pacific SST anomalies
 256 during that winter [*Wang et al., 2014; Baxter and Nigam, 2015; Hartmann, 2015; Lee et al.,*
 257 *2015; Seager et al., 2015; Yu and Zhang, 2015; Watson et al., 2016*]. *Watson et al. [2016]* fur-
 258 ther obtained an improved forecast of the anomalous circulation over North America when the
 259 tropics were relaxed to ERA Interim observations, suggesting the forcing of divergent flow by
 260 SST anomalies [e.g. via excitation of Rossby wave sources *Sardeshmukh and Hoskins, 1988*]
 261 plays a central role. This raises the possibility that the CNA is excited or enhanced by SST

262 anomalies and points to the need for a better understanding of CNA drivers and of its possi-
 263 ble interactions with SSTs and climate change.

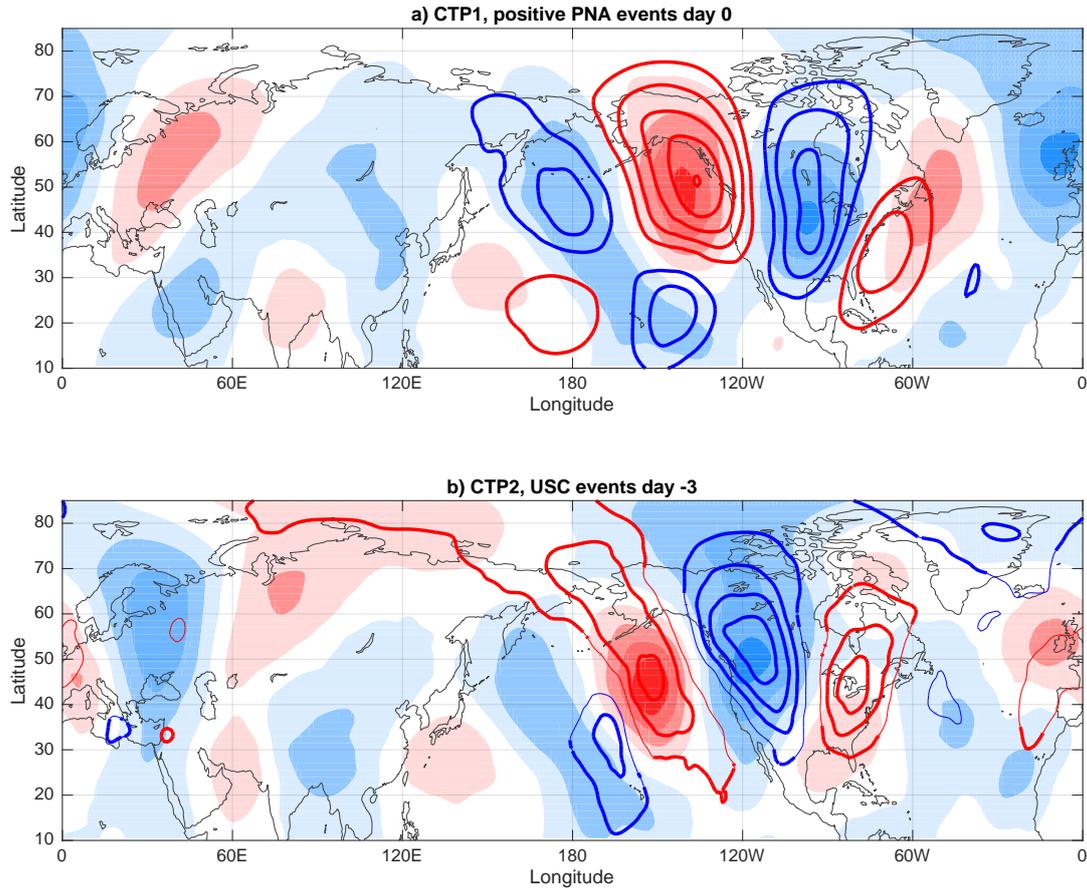
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 270 available MATLAB based tool for retrieval, analysis and visualisation of geophysical data (<http://www.goat-geo.org>). ECMWF ERA-Interim data used in this study have been obtained from the ECMWF
 271 Data Server http://data-portal.ecmwf.int/data/d/interim_daily/.
 272

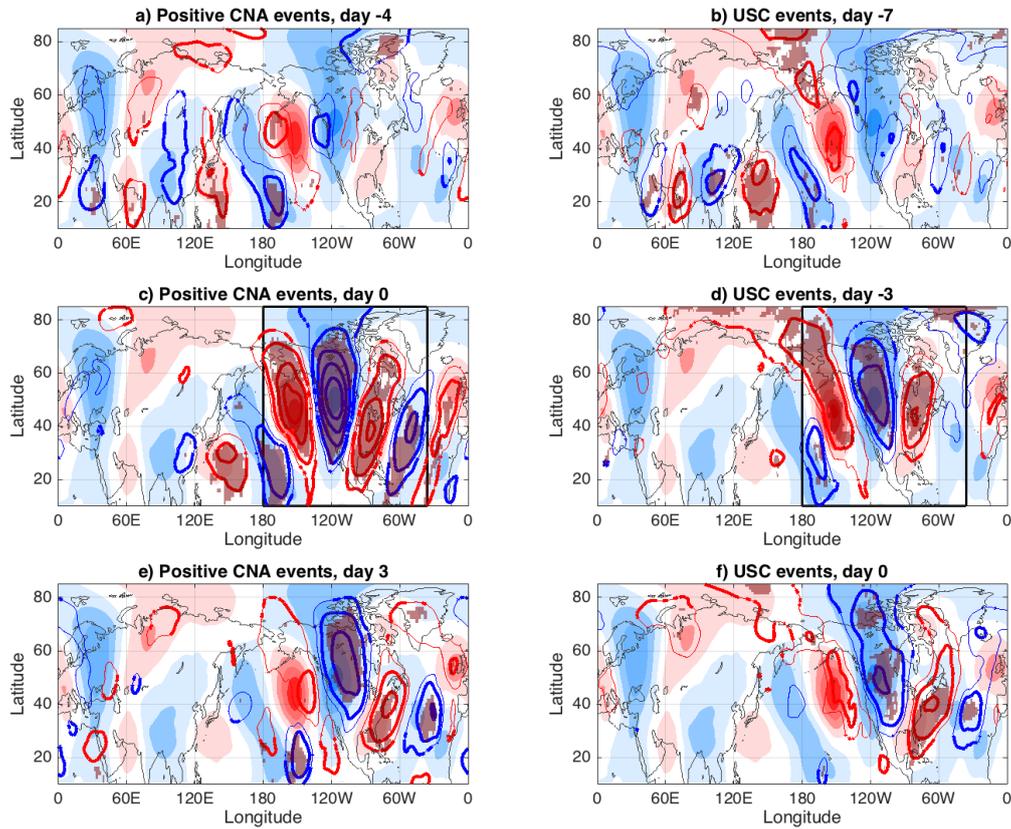
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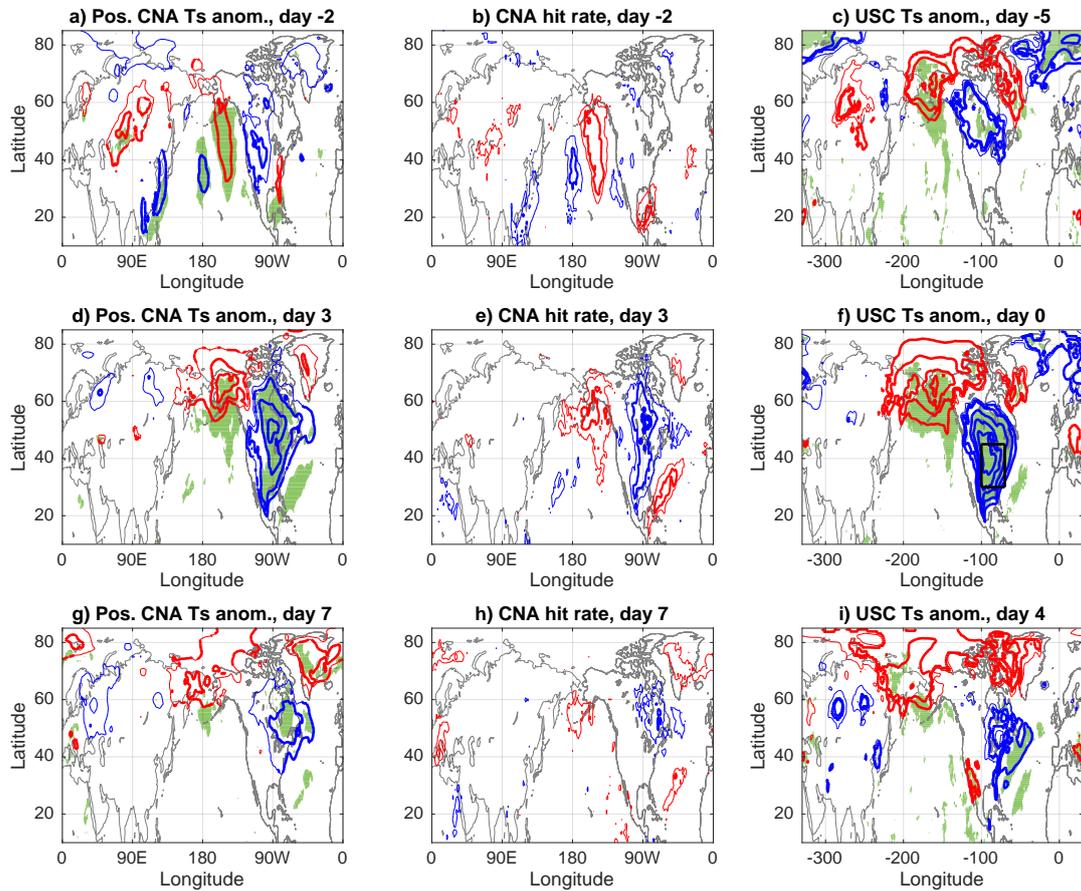
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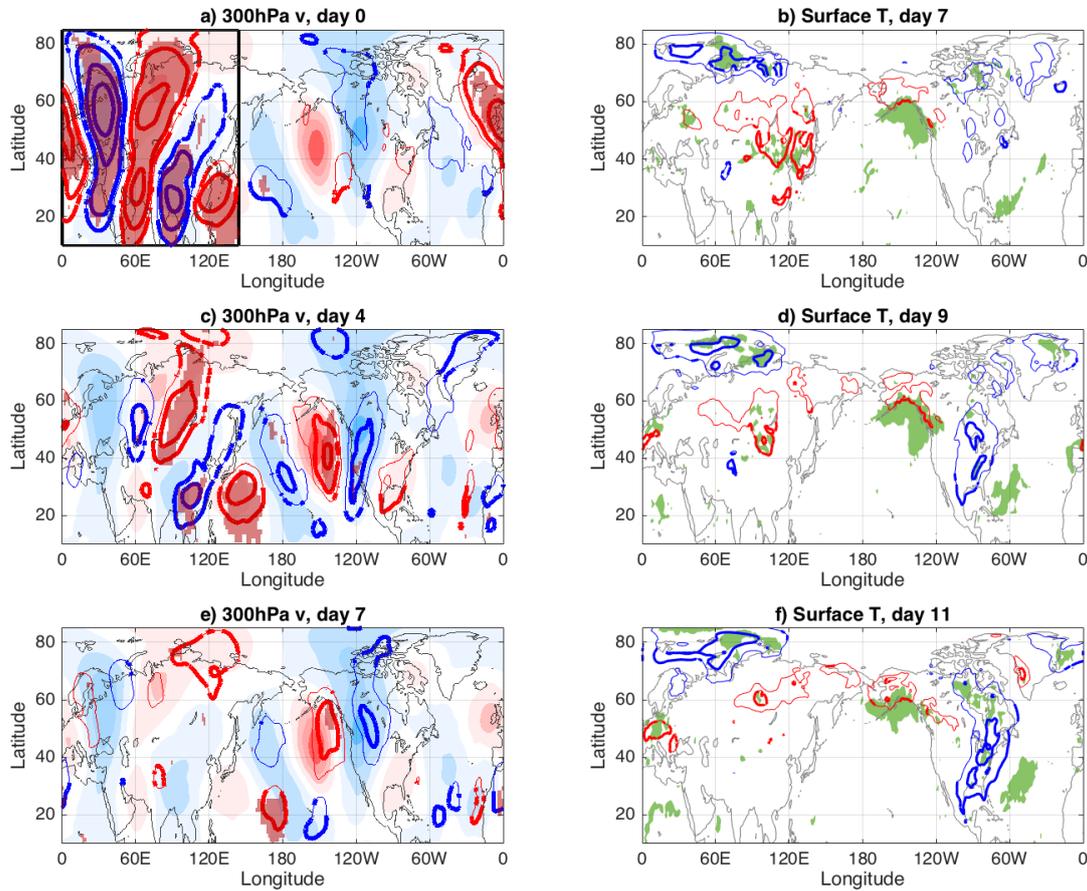
361 **Figure 1.** a) The first EOF of monthly mean seasonal anomalies of 300 – *hPa* meridional winds (shading)
 362 and the meridional wind composite of positive PNA events (red-blue contours) using Dec-Feb data. b) The
 363 second EOF of monthly mean seasonal anomalies of 300 – *hPa* meridional winds (shading) and the merid-
 364 ional wind composite of Eastern US Cold events (78 coldest events, red-blue contours). Contour interval is
 365 3m/sec, red is positive and blue negative, thick lines are statistically significant at the 95% level.



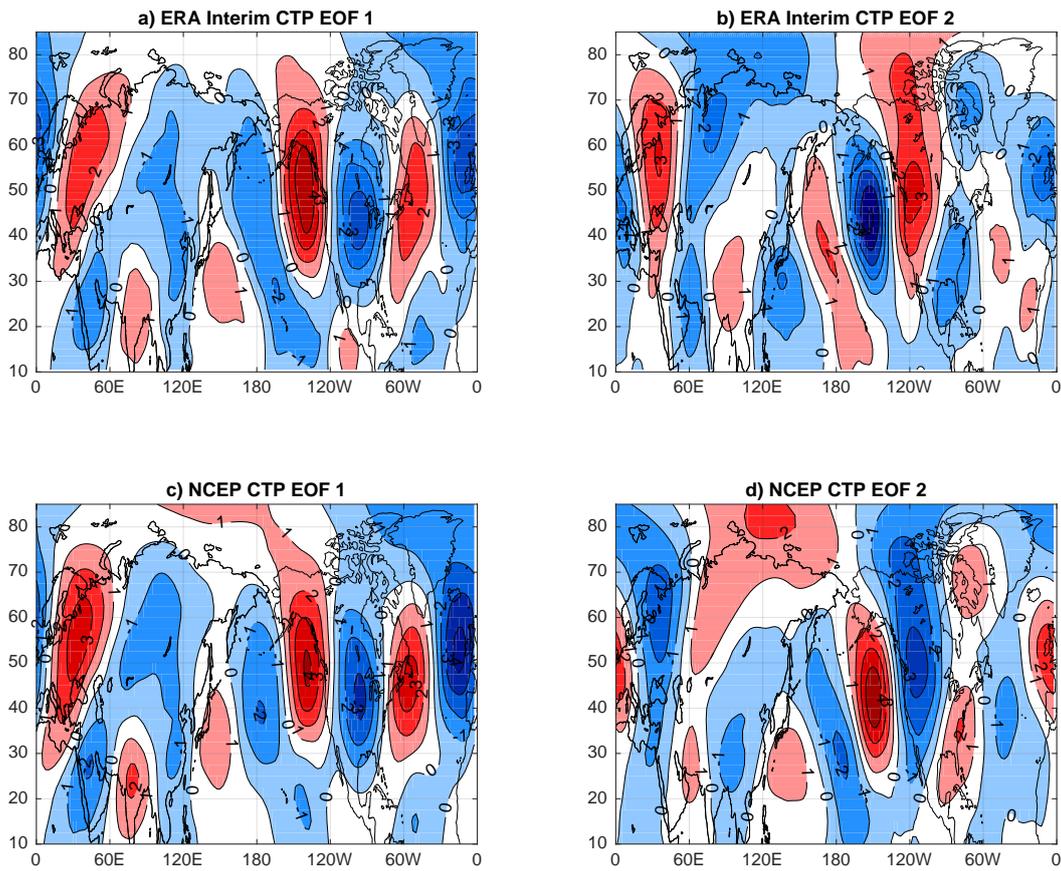
366 **Figure 2.** Time lagged composite of 300 – *hPa* meridional wind anomalies (contours) for a,c,d) positive
 367 CNA events at lags -4, 0 and 3 days; b,d,f) USC events at lags -7, -3 and 0; all plotted over the 300 – *hPa*
 368 meridional wind anomalies corresponding to the second CTP pattern (shading). The projection region used
 369 to define the CNA events is marked by the black rectangle in panels c,d. Contour interval is $5m/sec$, with
 370 the $\pm 2.5m/sec$ contours added. For contours and shading, red is positive, blue is negative. Thick contours
 371 mark 95% statistical significance, green shading marks regions where 67% of the composite members have
 372 the same sign of the composite (chances of this happening randomly are well below the 5%).



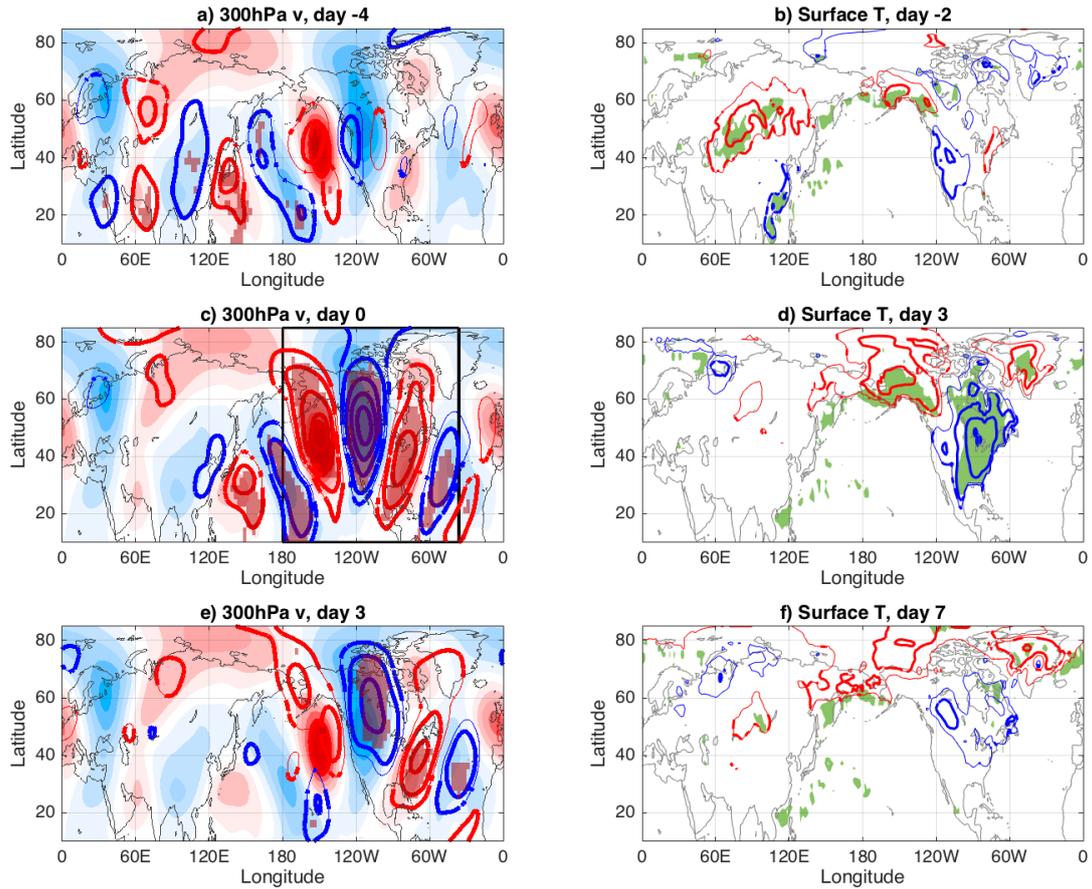
373 **Figure 3.** a,d,g) Time lagged composite of surface temperature anomalies (contours, red positive blue
 374 negative) for positive CNA events at lags -2, 3 and 7 days. Contour interval is $2K$ with the $\pm 1K$ contours
 375 added. Thick contours mark values statistically significant at the 95% level, darker shadings mark regions
 376 where 67% of the composite members have the same sign of the composite itself (chances of this happening
 377 randomly are well below 5%). b,e,h) The percent of positive CNA composite members (in a,d,g respectively)
 378 which have the same sign as the composite. Contours mark 65%, 75%, and 85% (marked by increasing thick-
 379 ness), with red/blue marking regions where the composite mean is positive/negative. c,f,i) same as a,d,g but
 380 for USC events but at lags -5, 0 and 4 days; The black rectangle in panel f shows the region used to define the
 381 USC events.



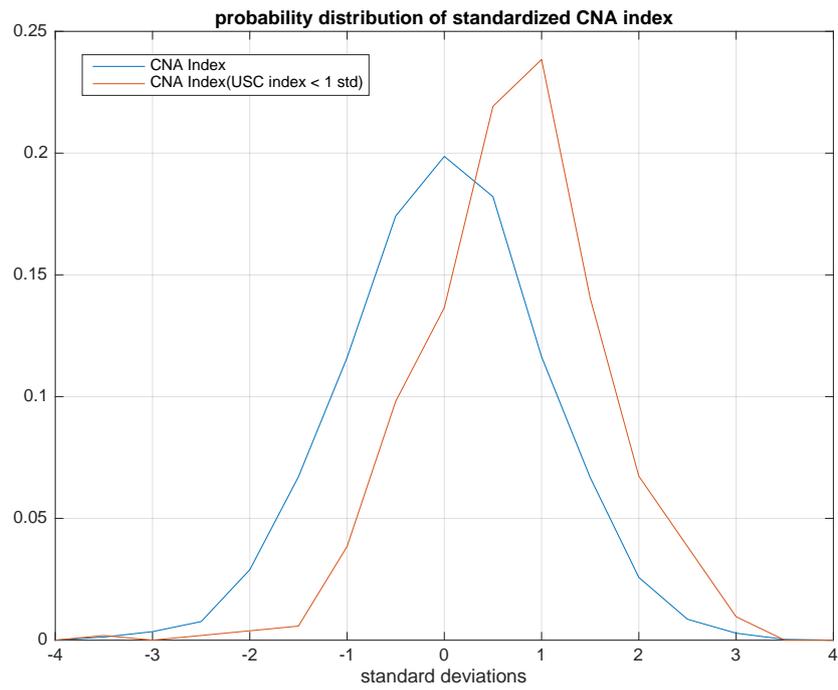
382 **Figure 4.** Precursor wave packet time lagged composites of: a,c,e) 300 – hPa meridional wind anomalies
 383 at days 0, 4, 7 (contours, interval 5m/sec with the ± 2.5 m/sec contours added); b,d,f) surface temperature
 384 anomalies at lags 7, 9, and 11 days (contours, interval 2m/sec with the ± 1 m/sec contours added). The com-
 385 posites are for events when the 300 – hPa meridional wind projects maximally onto the Euro-Asian sector
 386 (marked by the black rectangle in panel a), using 79 events. Also shown in panels (a,c,e) are the 300 – hPa
 387 meridional wind anomalies corresponding to the second CTP pattern. For contours and shadings, red is posi-
 388 tive and blue is negative. Thick contours mark values statistically significant at the 95% level, and the darker
 389 shading on a,c,e and green shading on b,d,f mark regions where 67% of the composite members have the
 390 same sign of the composite itself (chances of this happening randomly are well below 5%).



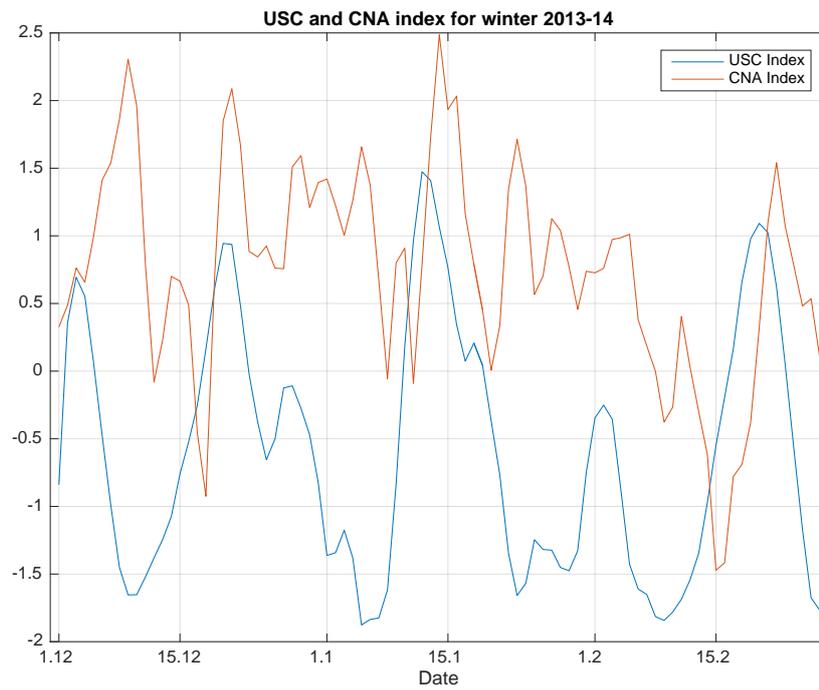
391 **Figure S1.** The first two CTP EOF patterns using ERA-Interim (top row) and NCEP (bottom row). EOF 1
 392 (left) and EOF 2 (right) in ERA-Interim explain $13.3 \pm 1.8\%$ and $11.5 \pm 1.6\%$ of the variance. These EOFs
 393 in NCEP explain $13.6 \pm 1.5\%$ and $10.8 \pm 1.2\%$ of the variance



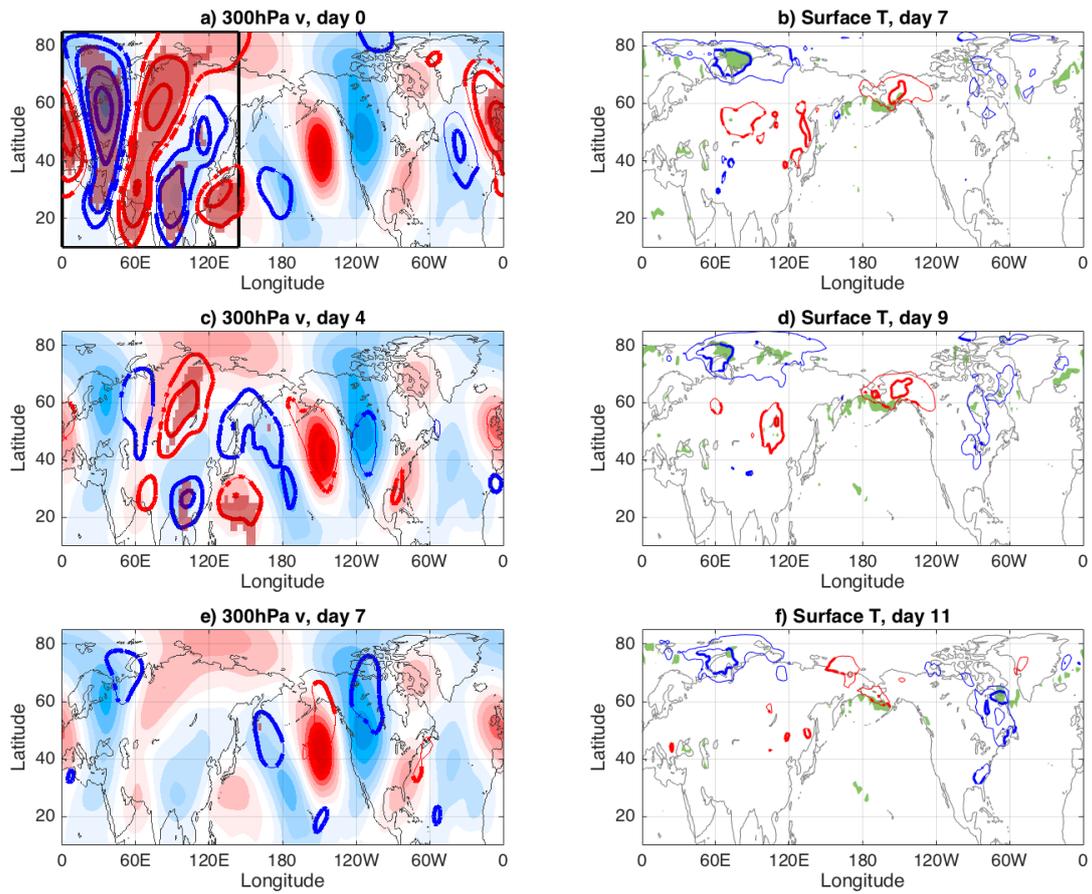
394 **Figure S2.** As in the left columns of figures 2 and 3 but using NCEP data. Note the absence of the
 395 ocean surface air temperature anomalies found for ERA-Interim (Figure 2a,c,e), despite the similarities of
 396 $300 - hPa$ v between the two data sets (compare to Figure 3a,c,e). Since there are 123 events for this data
 397 set, the chances to randomly get 67% of the composite members with the same sign of the composite itself are
 398 0.03%.



399 **Figure S3.** The probability distribution function (pdf) of the Dec-Feb standardised CNA index values for
400 all days (blue) and for days which are followed 3 days later by a USC index below -1 standard deviations
401 (red).



402 **Figure S4.** The daily CNA (red) and USC (blue) standardised index time series for the winter of 2013-14
403 (1 Dec 2013 - 28 Feb 2014). We subtracted the mean Dec-Feb value between Dec 1st 1979 and Feb 28th
404 2014, and divided by the standard deviation.



405 **Figure S5.** As in figure 4 but using NCEP data. Note again the absence of the ocean surface air temperature
 406 anomalies found for ERA-Interim and the smaller eastern North America 2-meter temperature anomaly at
 407 lag 11 days. The composites are based on 139 events, and the chances to randomly get 67% of the composite
 408 members with the same sign of the composite itself are less than 0.01%.