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#### 21 Abstract

22 This study analyzes the association between wintertime high wind events (HWEs) in the 23 northeast United States and extratropical cyclones. Sustained wind maxima in the Daily 24 Summary Data from the National Climatic Data Center's Integrated Surface Database are 25 analyzed for 1979-2012. For each station, a Generalized Pareto Distribution (GPD) is fit to the 26 upper tail of the daily maximum wind speed data, and probabilistic return levels at 1, 3 and 5-27 years are derived. Wind events meeting the return level criteria are termed HWEs. The HWEs 28 occurring on the same day are grouped into simultaneous wind exceedance dates, termed multi-29 station events. In a separate analysis, extratropical cyclones are tracked using the ECMWF ERA-30 Interim reanalysis. The multi-station events are associated with the extratropical cyclone tracks 31 based on cyclone proximity on the day of the event. The multi-station wind events are found to be most often associated with cyclones travelling from southwest to northeast, originating west 32 33 of the Appalachian Mountains. To quantify the relative frequency of the strong wind associated 34 cyclones, the full set of northeastern cyclone tracks are separated based on their path, using a 35 crosshairs algorithm designed for this region. The tracks separate into an evenly distributed set of 36 four pathways approaching the northeast US: from the due west, from the southwest, nor'easters, 37 and storms starting off coast, north of the Carolinas. Using the frequency of the tracks in each of 38 the pathways, it is shown that the storms associated with multi-station wind events are most 39 likely to approach the northeast US from the southwest.

## **1. Introduction**

43	A series of recent, costly weather disasters has led to an increased interest in
44	understanding and quantifying severe weather events (e.g., Vose et al. 2014; Kunkel et al. 2013).
45	For the northeast region of the United States (US), the most frequent cause of extreme
46	wintertime weather is extratropical cyclones, which can create damage through their
47	precipitation (Kunkel et al., 2012) and their winds (Ashley and Black, 2008). In view of this, the
48	study herein seeks to understand the connection between strong wintertime surface wind events
49	and extratropical cyclones in the Northeast United States.
50	Extratropical cyclones can approach the northeast US from the west, from the southwest
51	and from the south, the latter of which are referred to as nor'easters. Several aspects of these
52	wintertime storms have been discussed in the scientific literature. Miller (1946) separated
53	nor'easters based on their genesis regions, drawing a distinction between those that originate
54	over the Gulf of Mexico and those that develop over the Atlantic. Reitan (1974) estimated the
55	most frequent paths of storms for 1951-1970, distinguishing paths for storms over the northeast
56	US as: from the west, from the southwest, from the southeast and over the ocean (Fig. 12a in
57	Reitan (1974)). Hirsch et al. (2001) developed a climatology of east coast winter storms, and
58	included a strong wind threshold in their criteria for defining the storms. Dolan and Davis (1992)
59	show that nor'easters tend to cause strong beach erosion events due to the westward direction of
60	the winds poleward of the storm center, while Bernhardt and DeGaetano (2012) report on how
61	the North Atlantic Oscillation and El Nino-Southern Oscillation relate to the storms that cause
62	storm surge. However, less attention has been given to storms causing strong wind events over
63	land in the Northeast US.

64	Vose et al. (2014) review the trends in wind events in the US, and find that available
65	surface datasets and reanalysis products disagree on the sign of the trend (their Fig. 3, and Pryor
66	et al. 2009). Similarly, Knox et al. (2011) review the current understanding of non-convective
67	wind events, and suggest that there is some debate regarding the mechanisms causing high-wind
68	events in extratropical cyclones. For instance, some case studies suggest that downward
69	momentum mixing associated with tropopause folds may be responsible for high-wind events
70	(e.g. Iacopelli and Knox 2001; Browning 2004), while other case studies find a key forcing from
71	isallobaric winds (e.g., Durkee et al. 2012). However, the sting jet events discussed in Browning
72	(2004) are rare, and the work on case studies over land (Fink et al., 2009; Gatzen et al., 2011;
73	Durkee et al. 2012; Ludwig et al., 2015) suggest a more prominent role for ageostrophic fluxes.
74	Studies of strong surface wind in regions of the northeast United States have examined
75	the most likely wind direction during an event. For instance, Niziol and Paone (2000) used
76	station winds in western New York to show that the winds tend to be directed from the southwest
77	to northeast during the strong events. For the Great Lakes region, Lacke et al. (2007) found a
78	similar southwesterly propensity for the wind direction of strong, non-convective events
79	(identified using weather reports) in which they defined strong events using the National
80	Weather Service (NWS) criteria for high-wind watch or warning (sustained winds greater or
81	equal to 18 ms <sup>-1</sup> for 1 hour or a gust greater or equal to 26 ms <sup>-1</sup> for any duration). Lacke et al.
82	(2007) also found that the non-convective high wind events occur slightly more often in March
83	and April, as compared to November-February. Most recently, Pryor et al. (2014) found spatial
84	coherence over distances of up to 1000km in strong surface wind events, which, as they point
85	out, implies synoptic systems create the wind events.

86 For Europe, far more attention has been given to windstorms in the literature, with studies 87 that examine surface observations (Seregina et al., 2014) and reanalysis (Pinto et al., 2007; 88 Leckebusch et al., 2008; Donat et al., 2010; Nissen et al., 2010; Pfahl, 2014; Roberts et al., 89 2014), global climate models (Knippertz et al., 2000; Della-Marta and Pinto, 2009), case-studies 90 (Fink et al., 2009; Gatzen et al., 2011; Ludwig et al., 2014) as well statistical models (Schwierz 91 et al., 2010; Haas and Pinto, 2012; Born et al. 2012; Pinto et al. 2012). Leckebusch et al. (2008) 92 developed a method for identifying windstorms in gridded data, termed "footprinting". The 93 technique detects winds that exceed a local threshold and then looks for spatial clusters of 94 exceedances and tracks the clusters in time. Using this method, Leckebusch et al. (2008) 95 established that high wind events associated with extratropical cyclones tend to occur to the 96 south/southeast of the cyclone center, either along the cold front or slightly ahead of it. Nissen et 97 al. (2010) used the same technique to show that a similar spatial arrangement exists for high 98 wind events over the Mediterranean. These results for Europe, coupled with the work in the 99 northeast US (Niziol and Paone, 2000; Lacke et al., 2007) suggest that associating extratropical 100 cyclones with high wind events in the northeast US should identify a predominance of storms 101 with their centers to the north/northwest of the wind events.

With this in mind, the present study will examine northeast US strong wind events and associate them with extratropical cyclone tracks. A goal of this work is to test if the results from Europe, that the location of the strongest winds occur southeast of the storm center, apply in the Northeast US. We analyze station based wind data from the Daily Summaries of the National Oceanic and Atmospheric Administration (NOAA) Integrated Surface Database (Smith et al. 2011), a quality-controlled, surface-station dataset. To maximize the likelihood of studying extratropical cyclones, we only examine winds that occur in December through February (DJF).

109	Our analysis begins with an examination of high wind events in the northeast US and
110	then turns its focus to those storms identified as creating the strong wind events. To categorize
111	the strong wind events, this study uses a probabilistic approach, following Della-Marta and Pinto
112	(2009). Once identified strong wind events are associated with extratropical cyclone tracks, as
113	for example in Yarnal (Chapter 6 (1993)), to identify the pathway of the storms that are
114	associated with strong winds in the northeast US. After identifying the most likely pathway for
115	the storms, we test the robustness of the pathway results.
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117	2. Data and Methods
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119	2.1 Data
120	This study uses the Daily Summary Data from NOAA's Integrated Surface Database
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122 123 124 125 126	(ISD). The ISD consists of global, synoptic observations compiled from surface weather observation stations, ranging from airports to military bases. The Daily Summary dataset is a quality-controlled subset of the ISD provided by NOAA. The key variable we examine is the sustained wind maximum, which NOAA defines as the daily maximum of the 2-minute averages from each hourly observation reported for the day (personal communication, Mark Lackey, NOAA). Here we refer to this variable as MAX. We focus the analysis on the sustained wind
<ol> <li>122</li> <li>123</li> <li>124</li> <li>125</li> <li>126</li> <li>127</li> </ol>	(ISD). The ISD consists of global, synoptic observations compiled from surface weather observation stations, ranging from airports to military bases. The Daily Summary dataset is a quality-controlled subset of the ISD provided by NOAA. The key variable we examine is the sustained wind maximum, which NOAA defines as the daily maximum of the 2-minute averages from each hourly observation reported for the day (personal communication, Mark Lackey, NOAA). Here we refer to this variable as MAX. We focus the analysis on the sustained wind maximum rather than the wind gust because the MAX data are more frequently available for our
<ol> <li>122</li> <li>123</li> <li>124</li> <li>125</li> <li>126</li> <li>127</li> <li>128</li> </ol>	(ISD). The ISD consists of global, synoptic observations compiled from surface weather observation stations, ranging from airports to military bases. The Daily Summary dataset is a quality-controlled subset of the ISD provided by NOAA. The key variable we examine is the sustained wind maximum, which NOAA defines as the daily maximum of the 2-minute averages from each hourly observation reported for the day (personal communication, Mark Lackey, NOAA). Here we refer to this variable as MAX. We focus the analysis on the sustained wind maximum rather than the wind gust because the MAX data are more frequently available for our study period and region. We also use the daily mean wind speed (MEAN), defined as the 24-

130 reported in whole knots, which results in the data being quantized (with an approximate interval

of 0.5 ms<sup>-1</sup>) rather than continuous (Pryor et al. 2009). We note that the Daily Summary dataset
does not include wind direction, and therefore it is not considered in this study.

133 Our analysis focuses on the Northeast Region as defined by NOAA, which consists of 12 134 states: West Virginia, Maryland, Pennsylvania, Delaware, New Jersey, New York, Connecticut, 135 Rhode Island, Massachusetts, Vermont, New Hampshire, Maine. For these states, we use all of 136 the ISD stations for which at least 80% of MAX data are reported during DJF for the period from 137 January 1979 to December 2012, which yields 49 stations (Fig. 1a). We choose January 1979 as 138 the start date for our analysis because it coincides with the beginning of the reanalysis data used 139 to identify extratropical cyclones (see Section 2.3). A table that lists all station names, locations 140 and the percentage of data available is provided in the supplementary material (Supplemental 141 Table S1).

142 We choose a cut-off of 80% data coverage to establish broad station coverage over the 143 entire study region, which allows for a synoptic scale analysis. To test that this amount of data 144 coverage yields robust results, we performed two sensitivity analyses: (1) we repeated the main 145 analysis reported in Section 3 using only stations with 90% or more data coverage, (2) we tested 146 if missing data at a given station occurs more often when a high wind event occurs at one or 147 multiple other stations within 250 km. Neither analysis indicated a systematic bias, suggesting 148 that this set of 49 stations provides a representative synoptic view for winds in the northeast US. 149 Before analyzing the data, we took additional steps to address other potential biases. 150 First, we removed any sustained wind maximum data for which the concurrent mean wind speed 151 data are zero (dubious data). Second, any sustained wind maxima that were found to be 152 suspiciously larger than the concurrent mean wind for that day have been removed. To 153 accomplish this, we define a new variable,  $\eta$ , for each station *i*:

154 
$$\eta_i(t) = \frac{MAX_i(t) - MEAN_i(t)}{\sum_{j=1}^{N} MAX_j(t) - MEAN_j(t)}.$$
 (1)

In the denominator, we average over the N stations within 250 km of station *i*, not including station *i*. If  $\eta$  is large, then the difference between the MAX and MEAN at station *i* is large, as compared to the difference between MAX and MEAN for the surrounding stations. We chose to remove any data for which  $\eta$  was larger than 4, which led to a removal of overall less than 0.002% of the original data, or 172 total data points.

160 The data removed using the  $\eta$  threshold, are, by definition of  $\eta$ , isolated winds events. 161 However, some of the data removed are strong winds, which might suggest this method is 162 removing important data. However, 168 of the 172 WMAX data removed using  $\eta$  occur prior to 163 Jan 1, 1999 (Supplemental Figure S1). This date corresponds to the near completion of the 164 transition to the ASOS observing systems (McKee et al. 2000), which meant the majority of the 165 manual reporting was replaced by electronic reporting. McKee et al. (2000) note that the speed 166 and direction were similar for manual and ASOS, but there were issues with the gust 167 measurements, due to differences in the measurement-averaging window of the devices. Hayes 168 and Kuhl (1995) note a difference in the reporting of peak wind events, due to differences in 169 thresholds for defining peak winds. These biases would not affect our results, because we do not 170 focus on gusts or the count of peak wind reports. On the other hand, the fact that such a high 171 percentage of data identified using  $\eta$  occurred prior to 1999 suggests that the data removed 172 because  $\eta > 4$  may indeed be erroneous. For our purposes of associating multi-station wind 173 events with extratropical cyclones, the removal of the data with large  $\eta$  is justified.

174

#### 175 **2.2 Identifying High Wind Events**

176 The classification of high wind events (HWEs) that will be utilized in this study is a 177 probabilistic approach following statistical extreme value theory (EVT) (e.g., Coles, 2001; Coles 178 and Pericchi, 2003; Davison and Smith, 1990). For the identification of HWEs we use a peak-179 over-threshold (POT) model for MAX, based on the generalized Pareto distribution (GPD). 180 Asymptotic arguments (e.g., Pickands, 1975) justify the use of the GPD for modeling 181 exceedances over a high (enough) threshold because the GPD is the limiting distribution of a 182 normalized exceedance over a threshold as the threshold approaches the maximum of the 183 distribution (e.g., Coles, 2001). The GPD is defined as:

184 
$$F(x) = 1 - \left[1 + \xi \frac{x - \mu}{\sigma}\right]^{-\frac{1}{\xi}}, \sigma > 0, x > \mu, 1 + \xi \frac{x - \mu}{\sigma} > 0, \qquad (2)$$

185 where x are daily data (here MAX), u is the threshold value and  $\sigma$  and  $\xi$  are the scale (a measure 186 of the spread of the distribution of x) and shape parameter (which is determining the shape of the 187 distribution, rather than shifting it as u does or shrinking/stretching it as  $\sigma$  does), respectively. In 188 the GPD framework an essential step is to determine a threshold *u* for which the asymptotic GPD 189 approximation holds. Threshold choice involves a trade-off between bias and variance as: (i) a 190 too high threshold will reduce the number of exceedances and increase the estimation variance; 191 while (ii) a too low threshold will induce a bias as the GPD will poorly fit the exceedances. 192 In this study we use the POT-package (Ribatet, 2007) within R for the EVT analysis. In 193 this package the GPD parameters ( $\sigma$  and  $\xi$ ) are computed by maximum-likelihood estimation. 194 Evaluation of the GPD fit at the 49 northeast US sites considered here show that the 97-th 195 quantile provides a suitable threshold choice at all sites, satisfying the trade-off between bias and 196 variance. Supplemental Figure S2 provides an exemplary comparison of results from GPD fits at 197 selected sites with too high and too low threshold values.

Supplemental Figure S3 shows the probability density functions (PDFs) of wintertime MAX from the 49 northeast US sites in the ISD that fulfill the data selection criteria outlined in Section 2.1. The PDFs are asymmetric with heavier upper tails. We note that a similar skew was found in the PDFs of surface winds using the entire year, rather than DJF (He et al. 2010), and similar statistics were found in Pryor et al. (2014). Figure S3 also shows that the winds exceeding the threshold for a high wind watch or warning for the NWS (days with MAX > 18 ms<sup>-1</sup>) represent the upper end of the PDF range and occur very rarely.

205 Next we analyze the winds from two stations to illustrate why we have chosen to use 206 probabilistic statistics. The top panels of Figure 2 show the observed MAX (y-axis) versus the 207 estimate from a Gaussian fit (x-axis) for two selected (and representative) sites in the northeast 208 US: Bridgeport, Connecticut (left column) and Elkins-Randolph County, West Virginia (right 209 column). The figures confirm that the tails of MAX are non-Gaussian (i.e., data from a Gaussian 210 distribution would lie close to the diagonal 1:1 line). The grey-hashed boxes in the top panels (a 211 and b) in Figure 2 give the data range at the two selected sites beyond the 97-th quantile. The 212 middle panels (c and d) of Figure 2 (which are a zoom-in on the grey, hashed boxes of the top 213 panels) show observed (y-axis) versus GPD-fitted (x-axis) MAX. Comparing the top and middle 214 panels in Figure 2 shows that the GPD provides a better fit compared to a Gaussian distribution. After fitting the GPD ( $F_{\xi \mu \sigma}$ ), we calculate the empirical return level (R<sup>T</sup>) as: 215

216

217 
$$R_{T} = F_{\xi,\mu,\sigma}^{-1} \left( 1 - \frac{1}{T} \right).$$
(3)

Return levels are of practical interest because they describe the probability of exceeding a value x within a time window T. The bottom panels of Figure 2 show return level plots for the two selected sites. Thus, for example, MAX > 18 ms<sup>-1</sup> at the Bridgeport site would have a

probabilistic 5-year return level, while at the Elkins-Randolph County site it would have aprobabilistic return level of more than 20 years.

223 For the purpose of this study we choose to use 1-year, 3-year and 5-year return levels to 224 define HWEs. The reason being twofold: (i) return levels accurately capture the tail properties of 225 MAX; (ii) they provide a comparable standardized metric for MAX across individual sites. 226 Using HWEs at each station, we identify simultaneous exceedances of multiple station return 227 levels (hereafter, multi-station events) by finding all HWEs that occur on the same date +/- 1 228 day. The window of +/-1 day accounts for the possibility that a storm caused HWEs on either 229 side of 0Z (i.e., two different days in the daily summary), and the possibility of the same storm 230 transiting the study region over a 2-day period. We define the center of a multi-station event as 231 the average of latitude and longitude positions of the stations reporting the event.

232

233 **2.3 Extratropical Cyclone Association** 

234 Extratropical cyclones are identified by tracking their low-pressure centers, using 6-235 hourly sea level pressure (SLP) fields from the European Center for Medium Range Forecasts 236 (ECMWF) ERA-Interim reanalysis (Dee et al. 2011). ERA-Interim has been shown to compare 237 favorably with other reanalysis data for cyclone tracking (Hodges et al. 2011). To account for 238 possible biases in the trackers (e.g. Neu et al., 2013), we performed our analysis using two 239 separate cyclone-tracking algorithms: TRACK (Hodges, 1999) and the MAP Climatology for 240 Midlatitude Storminess (Bauer and Del Genio 2006). Despite major differences in the design of 241 the tracking algorithms, we found similar results in the wind analysis for both. Therefore we 242 present in the remainder of the paper only results based on the Hodges tracking scheme.

243 For the track database, we include tracks that last for at least 48-hours and travel at least 244 1000 km, which allows focusing on mobile synoptic systems. Figure 1b shows the track density 245 for all storms that pass through a box over the northeast region (Fig. 1b, black, dashed box). The 246 box used is sufficiently larger than the region of the stations so that the storm set includes all 247 storms that might influence the area. The track density is a count of the tracks per 2° by 2° grid 248 box per winter (DJF). The pattern shows a maximum over the Gulf Stream and a secondary 249 maximum over the Great Lakes, in good agreement with the pattern reported for east coast 250 wintertime storms in previous work (Hirsch et al. 2001). For DJF, from 1979-2012, for tracks 251 passing through the box in Figure 1b, we find a total of 1034 storms. 252 To associate the cyclone tracks with multi-station wind events, we require that the

cyclone center be within 1500 km of the geographical center of the event (see end of Section 2.2 for the definition of a center of a multi-station event). We have tested other radii, i.e., 1000 km, and found that the smaller distance excludes obvious storms. Since the track data are 6-hourly, while the station data are daily, we consider any cyclone that is within 1500 km at the time of the event  $\pm$  12 hours. For the multi-station events that occur on a single day, we use 12Z for that day. For the events that span two days, 0Z on the latter day is used.

In the cases in which multiple storms are found in proximity (in time and space) of the wind event, wind direction data from the ERA-Interim reanalysis are used to identify the most likely related storm. For this, first the area average of the 925-hPa zonal and meridional winds over a 5° by 5° region centered on the multi-station event is calculated. Second, wind direction is calculated from the area-averaged winds. If the wind direction has a northerly component, we retain the cyclones east of the station event (i.e., the winds are part of the back-end of the storm), and vice-versa for winds with a southerly component. For the rare case that there are still

266 multiple storms that fulfill the selection criteria, the storm that is closest in space to the wind267 event is kept.

268

#### 269 **3. Results**

#### 270 3.1 Extratropical Cyclone Tracks for Multi-Station HWEs

271 The HWEs during DJF in the northeast United States are defined by identifying wind 272 events that exceed the station-specific 1, 3 and 5-yr return levels (Table 1). We then find the 273 dates on which multiple stations have HWEs, hereafter, multi-station events. Table 1 shows the 274 results for exceedances of the 1-, 3-, and 5-year return levels, with the number of events 275 occurring simultaneously at multiple stations decreasing as the number of stations increases, 276 though not monotonically. The analysis that follows will mainly focus on multi-station events for 277 which 3 or more stations exceed their 3-year return levels. There are 52 of these events (i.e., 278 13+8+6+8+17, using the data on the 3-year return level row in Table 1). Analysis will also be 279 carried out on multi-station events for which 5 or more stations exceed their 5-year return 280 periods, for which there are: 15 (i.e., 6+4+5, using the data on the 5-year return level row in 281 Table 1) events.

Isolated events are defined as the dates for which only one station exceeds the given return level and these occur most frequently. As shown in Column 3 of Table 1, the occurrence of isolated events greatly reduces if 1-year return levels for surrounding stations are considered. For example, if the simultaneous exceedances of 5-year return levels are considered, then 85 single-station exceedances of the 5-year levels are found. However, if we consider simultaneous exceedances of 1-year and 5-year levels, the number of single-station exceedances of the 5-year levels drops significantly, down to 28. For reference, the dates for multi-station events defined as

5 or more stations with exceedances of the 5-year levels are listed in Supplemental Table S2.
Some of these storms were deadly (see for instance, Asuma 2010).

Using the extratropical cyclone association technique described in Section 2.3, we associate each multi-station event with a cyclone track, when possible. Figure 3 shows examples of this for multi-station events in which 3 stations simultaneously experienced winds that exceeded their 3-year return levels (Table 1). In this case, cyclones were associated with 11 of the 13 multi-station events. The figure shows that in some cases a multi-station event is based on 3 stations in close proximity (e.g., Dec 21, 2012 in Fig. 3b), while in other cases the stations are spread across the region (e.g., Dec 29, 1994 in Fig. 3a).

298 Next, we examine the associated tracks when using different thresholds to define a multi-299 station event (Fig. 4). Figure 4a shows the tracks for all events for which there are at least 5 300 stations at which the wind exceeded the station's 1-year return level. There are 102 multi-station 301 events that fit the definition and for 82 of these events an associated cyclone is identified. In this 302 case, no preferred path is obvious, perhaps due to the large number of tracks included in the plot. 303 Figure 4b shows the paths for multi-station events defined as exceedances of the 3-year return 304 level at 3 or more stations. There are 52 events that fit this definition, and for 44 of these events 305 an associated cyclone is identified. In this case, there appears to be more storms that arrive in the 306 northeast region from the west or southwest. Figures 4c and 4d show results for more stringent 307 definitions of multi-station events, and a higher percentage of events are associated with a 308 cyclone track that arrives in the northeast from the southwest. For the multi-station events 309 defined as 5 or more stations exceeding their 3-year return level, 31 events are found with 26 310 associated cyclone tracks. For the multi-station events defined at 5 or more stations exceeding 311 their 5-year return level, 15 events are found with 13 associated cyclones identified.

312 Figure 5 shows, for different thresholds used to define a multi-station event, the location 313 of the storm center (in red) and the location of the average of the latitude and longitudes for the 314 stations with HWEs in the event (in blue). For each of these definitions, the majority of the storm 315 centers are north or northwest of the stations experiencing a wind event, suggesting that the 316 winds are in the south/southeast quadrant of the cyclones. Consistent with this result, a 317 composite of the SLP field for the study domain (using ERA-Interim reanalysis) on the day of 318 the multi-station events, based on 5 or more stations exceeding the 5-year return levels, also 319 shows the storm center north of our study region (Fig. 6). The SLP contours further suggest that 320 the winds are directed from the southwest to the northeast, which is in agreement with the 321 individual station studies of Niziol and Paone (2000) and Lacke et al. (2007).

322

#### 323 3.2 Quantifying the Preferred Extratropical Cyclone Path

The qualitative results from the previous section show a preference for the multi-station events being caused by storms approaching from the southwest. Next, we quantify this preference by examining the relative occurrence of strong wind associated storms arriving from different directions. To do this, a new methodology for separating the cyclone tracks based on their initial locations and paths is presented. Then the technique is applied to all cyclone tracks in the northeast US and to the tracks associated with multi-station events.

Motivated by the track separation presented in Reitan (1974), we have designed an analysis aimed at separating the cyclone tracks into those that take a zonal path towards the northeast US, those that arrive from the SW, and those that move northward along the coast. The analysis utilizes knowledge of the tracks initial development region and their trajectory across the northeast US. We use a reference frame centered at the geometric average of the latitude and

335 longitude positions of the 49 weather stations to draw a crosshairs based on fixed lines of latitude

336 (41.37° N), and longitude (75.06° W), which hereinafter are referred to as latFIX and lonFIX for

337 simplicity. The storms are then separated into four groups:

338 (1) fromNW: tracks that begin northwest of the intersection and remain north of latFIX.

- 339 (2) fromSW: tracks that begin southwest of the intersection and either remain in that
- 340 quadrant or cross latFIX traveling north to the west of lonFIX.

341 (3) fromSE: tracks that cross lonFIX traveling east to the south of latFIX.

342 (4) overOCEAN: tracks that remain east of lonFIX or cross lonFIX traveling west.

343 We note that many of the storms in the fromSE and overOCEAN tracks could be considered

344 nor'easter's based on the wind pattern they generate when passing the northeast US. However,

345 the classification used here does not include nor'easters as an individual category, because the 346 paths have been separated based on their origin.

Panels (a)-(d) in Figure 7 show the track density (using the same procedure as in Fig. 1b) for the full storm set, based on these categories. For this separation, we find that if we consider all events there is a relatively equal number of tracks per characteristic path (Table 2). To test the sensitivity of the separation in respect to the values of lonFIX and latFIX, we repeat the analysis, shifting the location of the reference frame center by one degree in each direction (Table 2). As expected the results show that counts change with shifts, however this does not result in any drastic changes.

Next the track separation technique is used to parse the tracks associated with the multistation events. For this analysis, we use the tracks found based on events for which the winds exceed the 3-year return level at 3 or more stations (i.e., Fig. 4b). Panels (e)-(h) in Figure 7 show these tracks separated into the characteristic pathways, with the counts as follows: fromNW (7), 358 fromSW (27), fromSE (9), overOCEAN (1). Using the number of total storms per characteristic 359 track found (given in Table 2), the relative frequency of storms causing multi-station events per 360 characteristic path is calculated. For fromSW the value is 10.5%, which is at least three times 361 greater than any of the frequencies for the other pathways. Furthermore, given that each of the 4 362 pathways have nearly the same number of tracks when all of the extratropical cyclones are 363 considered (Table 2), we can use binomial probabilities to test the significance of the strong wind 364 path result. In particular, if we consider this a Bernoulli Experiment and use the binomial 365 distribution to test the likelihood of 27 of the 44 events coming from one pathway. The 366 probability is less than 1 in a million.

367 To conclude this section, we discuss our choice for extratropical pathway separation. The 368 crosshairs separation technique used is subjective and based on prior understanding of the likely 369 pathways that storms take to arrive in the northeast US (e.g., Reitan 1974). In an attempt to make 370 a more objective track separation the tracks separated were also using hierarchical clustering 371 (Ward 1963), a technique that has been previously applied to atmospheric circulation regimes 372 (e.g. Casola and Wallace 2007). The clustering analysis resulted in a similar set of final clusters, 373 i.e. the characteristics paths, as those we found using the crosshairs. However, the number of 374 tracks per final cluster was very sensitive to the geographical extent of the tracks that was fed 375 into the clustering algorithm. The clustering algorithm does not provide a simple mechanism for 376 showing the sensitivity of the track separation to slight changes in the method, as we did here for 377 the crosshairs method with Table 2. This led us to conclude that our technique, though 378 subjective, offers the simplest and most easily reproducible method for separating the tracks. 379

380 3.3 Robustness of the Preferred Extratropical Cyclone Path

This section details two analyses designed to test the robustness of the preferred pathway result. First, the sensitivity of the pathway analysis to the geographical density of the surface stations is evaluated. Second, we test if the pathway analysis is sensitive to the number of stations within range of the cyclone winds.

385 To test if the existence of a denser concentration of stations along the coast versus inland 386 (see Fig. 1a) creates a bias, we repeat the storm association analysis using a subset of stations 387 that are more evenly spaced. To this aim we retain only one station separated by a 100-km 388 radius, which results in a subset of 23 stations (Fig. 1a; yellow crosses). Using the 23-station 389 subset, we find 27 multi-station events defined based on at least a 3-year return level at 3 or more 390 stations (as opposed to the 52 multi-station events found using the full set). For these 27 events, 391 we find 23 associated tracks, and the track separation of the storms again results in fromSW 392 again being the most likely pathway (Supplemental Fig. S4). For further sensitivity analysis we 393 repeated the analysis using radii of 50 km and 150 km (to create more regularly spaced station 394 data sets) and found results consistent with those presented based on the 100-km radius. Thus, 395 the results show that the geographical density of stations does not affect our results

Given the location of the stations relative to the paths of the cyclones centers, one could argue that the fromSW pathway being the most likely to cause multi-station events is a result of there being more stations within range of the cyclone winds that take this path. To test this hypothesis, we repeat the track association analysis using HWEs identified in the wind field in the ERA-Interim reanalysis using a fixed location. The idea behind this analysis is to utilize the temporal and spatial continuity of the reanalysis data in order to identify high wind events similar to the scale found using the multi-station approach at a single, fixed location.

403 For the region within 77.5° W to 70° W by 40° N to 43° N (red box in Fig. 8a), the 3 404 strongest values of the, 925-hPa daily-averaged windspeed from ERA-Interim are identified and 405 averaged to a single value. Then the DJF values in the resulting time-series are fit to a GPD. 406 Because the 925-hPa daily-averaged wind speed represents a smoother distribution with less 407 striking extremes compared to the ISD observations, we focus on shorter return levels (i.e., 1 408 year or above) to establish robust statistics. We identify the high-wind events as those that 409 exceed the 1-year return level and then isolate events that are at least 3 days apart (to remove the 410 chance of double counting a storm). If multiple exceedances of the 1-year return level occur 411 within 3 days, the strongest event is used. These HWEs are then associated with extratropical 412 cyclones using the method described in Section 2.3.

413 Figure 8a shows the tracks associated with 925-hPa HWEs using the black box shown in 414 the figure. In this case, as for the ISD multi-station events, most of the tracks travel from the 415 southwest. To test if this characteristic pathway is caused by coastline geometry or topography, 416 we repeat the analysis using two other boxes at the same latitude, east of the first box (Fig. 8b, 417 8c). In these cases we test for associated storms using a set of cyclone tracks that includes more 418 storms over the ocean, which are not necessarily included in the original set of 1034 storms. 419 Once again, the tracks that create high wind events for each region tend to be those that approach 420 the box from the southwest. These results suggest that the identification of the fromSW pathway 421 in the station analysis is unlikely to be based on which track passed over the most stations. These 422 results also have implications for the cause of the fromSW pathway being the dominant track for 423 wind events in the Northeast, related to the location within cyclones where the strongest winds 424 occur. This is discussed in Section 4.

#### 426 3.4 Geographical Distribution of High Wind Return Levels

427 For the EVT-based HWEs, we also examine the geographical climatology of events in 428 the northeast US by plotting the average wind speeds for the 1-, 3- and 5-year events, per station 429 (Fig. 9). The three panels show that the average strength of HWEs at stations near the Great 430 Lakes and stations along the coast are usually larger than those for inland stations. Given the 431 results from Figure 9, we use the geographical locations of the stations to create four subsets of 432 sites for the northeast: Great Lakes, Inlands, Near-Coast and At-Coast (Fig. 9, and Table S1 in 433 the supplemental material for each station's designation). Station designation is defined in the 434 following way. The Great Lakes stations are all stations within 100 km of any Great Lake. The 435 At-Coast stations are all stations within 40 km of the coastline, while all stations between 100 436 km and 40 km from the coastline are classified as Near-Coast. We then calculate average wind 437 speeds for the 1-, 3- and 5- year events for each of the subsets. The results show that winds are 438 stronger near the Great Lakes and at the Coast. A detailed summary is presented in Table 3 439 serving as a first-order benchmark for the strength of wintertime high wind events in these 440 regions of the northeast. We note that the distances used to separate the data are arbitrary and 441 chosen to simplify the presentation in Table 3.

442

#### 443 **4. Discussion**

The analysis reveals that storms taking a path from the southwest towards the northeast region are most likely to cause multi-station strong winds events in the region (Fig. 4). It appears that this is a result of the south by southeast quadrant of these storms being more likely to pass over the stations as compared to any of the other paths, as evidenced by the fixed location analysis of the reanalysis winds (Fig. 8). Figure 8 also shows that if we consider a region farther

449 east, the dominant storms would be those from the nor'easters or fromSW categories for our 450 1034 storms. This, again, is because the east by southeast quadrant of those storms would be 451 more likely to pass over that region. As such, our work does not imply that the from SW storms 452 create stronger winds than storms from the other groups, but that the strong winds generated by 453 the storms taking the from SW path are most likely to occur over the northeast US. This is 454 consistent with an analysis of strong wind producing storms over western Europe (Ulbrich et al. 455 2001; Leckebusch et al., 2008; Nissen et al., 2010; Pfahl 2014), which find the cyclone centers 456 tend to be north of the wind events, and the wind events tend to be in the warm sector near the 457 cold front, or just behind the cold front. The locations of the strong winds relative to the storm 458 center are also in accord with composite views of winds within extratropical cyclones (e.g., 459 Bengtsson et al. 2009; Catto et al. 2010; Booth et al. 2013).

460 We also tested for a relationship between the strength of the wind events and the strength 461 of the storms, based on the storm-centered SLP gradient (gradSLP), for each storm at the time of 462 the wind event. In an analysis of the set of multi-station events for which the 3-year return levels 463 are exceeded by 3 or more stations, we calculate the station-averaged windspeed and gradSLP 464 for the associated storms. However, no correlation between the gradSLP and surface station 465 winds for the multi-station events was found. This null result is somewhat expected. The SLP 466 gradient provides a proxy for the geostrophic forcing of the surface winds, however, as shown in 467 Fink et al. (2009) and Durkee et al. (2012), the surface winds also contain ageostrophic 468 components. Because the strong winds occur in the proximity of the cold front of the storms, it is 469 also possible that momentum mixing associated with convection also provides an ageostrophic 470 forcing for the surface winds.

471

472 **5. Summary** 

473 This study identified historical strong wintertime surface wind events in the northeast US 474 using station data. We applied methods from statistical extreme value theory to calculate 475 probabilistic 1-, 3- and 5-year return levels for surface weather stations and linked events that 476 occurred on the same date to identify multi-station events. Using these multi-station strong wind 477 events, the associated extratropical cyclones were identified. The main finding of the presented 478 study is that storms approaching the region from the southwest are most likely to be associated 479 with strong surface winds. Results of a track separation analysis of all cyclone tracks for 1979-480 2012 show that, a storm causing strong surface winds is more likely to approach from the 481 southwest than any other direction.

482 Our findings regarding the strongest winds within the warm sector support and expand on 483 results from multiple studies over Europe (e.g., Leckebusch et al. 2008 and Nissen et al. 2010). 484 In particular, the present study confirms that for the northeast US, the Leckebusch et al. (2008) 485 results regarding the relative location of the winds within the cyclone is the key for 486 understanding the locations at which cyclones creates strong winds. Additionally, we here 487 utilized a new technique to identify strong synoptic wind events using station data: our multi-488 station event approach. This technique is unique from the wind footprinting analysis Leckebusch 489 et al. (2008) and Nissen et al. (2010) applied to reanalysis winds. Therefore, the consistent 490 results regarding the associated cyclones suggest that both methods (ours using surface 491 observations and theirs using reanalysis winds) are capable of identifying strong synoptic wind 492 storms. Future work will directly compare the two techniques.

493 To conclude, we discuss some of the implications of our results for storm impacts. First,
494 if we consider storm impacts in the current climate, we can conclude that the extratropical

495 cyclones that are associated with the strongest wind events over land most frequently are not the 496 same as those that cause storm surge (i.e., Nor'easters), as reported in Dolan and Davis (1992). 497 Next, if we consider storm impacts in a warmer world, the implications of our work suggest that 498 projecting changes in surface wind events will depend in the foremost on the track of the 499 cyclones. Based on the study of Colle et al. (2013), global climate models (GCMs) project an 500 increase cyclone tracks over the coastline and slightly inland. Based on our results, this suggests 501 a possible increase in strong wind events if the GCM projected track changes are correct. 502 503 Acknowledgements

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#### Tables

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|--|

Table 1: Count of HWEs and multi-station events										
Determ		Isolated		Multi-station events on the same day by number of stations <sup>a</sup>						f stations <sup>a</sup>
Level (years)	Total HWEs	Events (isolated at 1-yr RL)		Two	Three	Four	Five	Six	Seven or more	Max # of stations
1	1621	172		58	42	24	18	22	62	29
3	490	116 (47)		27	13	8	6	8	17	14
5	289	85 (28)		16	11	10	6	4	5	8

<sup>a</sup> For each of the return levels, the count of multi-station events per number of stations does not monotonically decrease as the number of station increases. It does have a downward tendency, however, it also has a long tail, as indicated by the last column.

## 

Table 2: Track counts per characteristic paths vs location of crosshairs         Distribution         Distr <tr< td=""></tr<>					
co	ounts are listed as: from	NW/IromSW/IromSE/0	overUCEAN		
	<b>40.4</b> °N	41.4°N	42.4°N		
283.8°W	239/254/240/301	221/256/256/301	201/255/277/301		
284.8°W	245/259/264/266	225/258/285/266	205/263/300/266		
285.8°W	249/265/279/243	228/260/303/243	207/267/317/243		

### 

Table 3: Average Strength of Wind Events by region MIN/MEAN/MAX (ms-1)					
	Great Lake	Inland	Near- Coast	At-Coast	
1 4 2	12.3	10.8	11.3	13.3	
1- to 3-	16	14.3	14.6	15.6	
year KL	19.6	17	17.5	18	
2	16.5	12.8	12.4	15.3	
3- to 3-	18	15.6	16.2	16.8	
yeai KL	20.1	17.5	18	19.6	

Table 2	2: Track counts per char	acteristic paths vs loca	tion of crossha
co	ounts are listed as: from	NW/fromSW/fromSE/o	overOCEAN
	<b>40.4</b> °N	41.4°N	42.4°N
283.8°W	239/254/240/301	221/256/256/301	201/255/27
284.8°W	245/259/264/266	225/258/285/266	205/263/30
285.8°W	249/265/279/243	228/260/303/243	207/267/31

### 716 **Figure Caption List**

717

718	Figure 1: Stations (a) and track density (b). In (a) Locations of ISD stations in NOAA Northeast
719	Region with at least 80% MAX data for DJF for 1979-2012 are shown. Color of stations
720	corresponds to percentage of data available. Yellow x's show stations used for repeated analysis
721	in which a set of more evenly spaced staions was used (i.e., 1 site within 100 km radius, see text
722	for further explanation). In (b) track density for extratropical cyclones in DJF, based on tracks
723	from the TRACK algorithm. Units: count per winter (CPW). Contour interval is 2.5 CPW.
724	Thicker contours show 5 CPW and 10 CPW. Black box shows region through which all tracks
725	must travel to be included in database.
726	
727	Figure 2: (a) Quantile-Quantile (QQ) plot comparing observed MAX (m/s) at Bridgeport with a
728	least-square fitted Gaussian. (b) As (a) but for Elkins-Randolph County. (c) QQ-plot comparing
729	observed MAX from Bridgeport with GPD-fitted MAX, (d) as (c) but for Elkins-Randolph
730	County. (e) Return level plot for Bridgeport from the fitted GPD in (c), (f) as (e) but for Elkins-
731	Randolph County. Grey Hashed boxes in (a) and (b) mark the data range above the 97-th
732	quantile at each site. Orange dashed lines mark the NWS threshold for a high wind watch or
733	warning (i.e., 18 ms <sup>-1</sup> ) in all panels. Secondary axis in (a) and (b) show corresponding mean
734	values (M) and standard deviations ( $\sigma$ ).
735	
736	Figure 3: Multi-station events and associated tracks examples: multi-station events for which the

vinds exceed 3-year return levels at exactly 3 stations. 13 multi-station events were identified for

this criterion. For 11 of these events, an associated extratropical cyclone is identified. Cyclone

tracks are the lines; station locations are the dots. The associated tracks and stations are given in the same color. The green dot on each track shows the location of storm at date of multi-station event. The legend shows the full date extent of each track and date of multi-station event in parentheses. For the Dec 4, 1990 case, there are two stations nearly overlapping in the NYC region.

744

Figure 4: Track associated with multi-station events based on different criteria: (a) 1-year return level (RL) at 5 or more station; track count 84 (total events: 102), (b) 3-year RL at 3 or more stations; track count 44 (total events: 52), (c) 3-year RL at 5 or more stations; track count 26 (total event 31), (d) 5-year RL at 5 or more stations; track count 13 (total events 15). Track count gives the number of associated tracks and total events gives the number of multi-station events identified for each specified criterion.

751

Figure 5: Location of cyclone centers (in blue) and geographical average location of associated stations (in red) during multi-station events with: (a) 3-year return level (RL) at 3 or more stations, (b) 3-year RL at 5 or more stations, (c) 5-year RL at 5 or more stations. Dashed black lines connect station center to associated storm center. For reference, the black circle shows a distance of 1000 km from the geographical center of all of the stations.

757

Figure 6: Composite for multi-station events. Contours show SLP (hPa), shading shows wind
speed at 925 hPa (ms<sup>-1</sup>). Multi-station events here are defined as HWEs exceeding the 5-yr return
level at 5 or more stations.

762	Figure 7: Separating tracks based on characteristic pathways: (a-d) track density for all tracks
763	and (e-h) track paths for storms associated with multi-station events. Pathway names: (a)
764	fromNW, (b) overOCEAN (c) fromSW and (d) fromSE. Contour interval in (a-d): thin lines:
765	1.25 counts per winter, thick lines: 2.5 counts per winter. For storms associated with multi-
766	station events, track count per path: (e) 7, (f) 1, (g) 27, (h) 8. Multi-station events defined here
767	as: 3 or more stations exceeding their 3-year return level. Dashed lines show crosshairs
768	designated by the geometric mean latitude and longitude of the stations.
769	
770	Figure 8: Cyclone track association for area average of 925-hPa reanalysis winds in black boxes:
771	Latitude range for all boxes: $40^{\circ}$ N – $43^{\circ}$ N. Longitude ranges: (a) 77.5°W-70°W, (b) 67.5°W –
772	$60^{\circ}$ W, and (c) $57.5^{\circ}$ W – $50^{\circ}$ W. Red line indicates the cyclone tracks, blue dot marks location of
773	cyclone at time of association with high wind event for the area-averaged wind in the box.
774	
775	Figure 9: (a) 1-year MAX return level on site basis; (b)-(c) as (a) but for 3-year and 5-year return
776	levels.
777	

#### 779 Figures



781 Figure 1: Stations (a) and track density (b). In (a) Locations of ISD stations in NOAA Northeast Region with at least 80% MAX data for DJF for 1979-2012 are shown. Color of stations 782 783 corresponds to percentage of data available. Yellow x's show stations used for repeated analysis 784 in which a set of more evenly spaced staions was used (i.e., 1 site within 100 km radius, see text 785 for further explanation). In (b) track density for extratropical cyclones in DJF, based on tracks 786 from the TRACK algorithm. Units: count per winter (CPW). Contour interval is 2.5 CPW. 787 Thicker contours show 5 CPW and 10 CPW. Black box shows region through which all tracks 788 must travel to be included in database. 789



Figure 2: (a) Quantile-Quantile (QQ) plot comparing observed MAX (m/s) at Bridgeport with a 793 794 least-square fitted Gaussian. (b) As (a) but for Elkins-Randolph County. (c) QQ-plot comparing 795 observed MAX from Bridgeport with GPD-fitted MAX, (d) as (c) but for Elkins-Randolph 796 County. (e) Return level plot for Bridgeport from the fitted GPD in (c), (f) as (e) but for Elkins-797 Randolph County. Grey Hashed boxes in (a) and (b) mark the data range above the 97-th 798 quantile at each site. Orange dashed lines mark the NWS threshold for a high wind watch or warning (i.e., 18 ms<sup>-1</sup>) in all panels. Secondary axis in (a) and (b) show corresponding mean 799 800 values (M) and standard deviations ( $\sigma$ ).







804 Figure 3: Multi-station events and associated tracks examples: multi-station events for which the 805 winds exceed 3-year return levels at exactly 3 stations. 13 multi-station events were identified for 806 this criterion. For 11 of these events, an associated extratropical cyclone is identified. Cyclone 807 tracks are the lines; station locations are the dots. The associated tracks and stations are given in 808 the same color. The green dot on each track shows the location of storm at date of multi-station 809 event. The legend shows the full date extent of each track and date of multi-station event in 810 parentheses. For the Dec 4, 1990 case, there are two stations nearly overlapping in the NYC 811 region.



Figure 4: Track associated with multi-station events based on different criteria: (a) 1-year return
level (RL) at 5 or more station; track count 84 (total events: 102), (b) 3-year RL at 3 or more
stations; track count 44 (total events: 52), (c) 3-year RL at 5 or more stations; track count 26
(total event 31), (d) 5-year RL at 5 or more stations; track count 13 (total events 15). Track count
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Figure 5: Location of cyclone centers (in blue) and geographical average location of associated
stations (in red) during multi-station events with: (a) 3-year return level (RL) at 3 or more
stations, (b) 3-year RL at 5 or more stations, (c) 5-year RL at 5 or more stations. Dashed black

- 828 lines connect station center to associated storm center. For reference, the black circle shows a
- 829 distance of 1000 km from the geographical center of all of the stations.
- 830
- 831



833 Figure 6: Composite for multi-station events. Contours show SLP (hPa), shading shows wind



835 level at 5 or more stations.

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Figure 7: Separating tracks based on characteristic pathways: (a-d) track density for all tracks and (e-h) track paths for storms associated with multi-station events. Pathway names: (a) fromNW, (b) overOCEAN (c) fromSW and (d) fromSE. Contour interval in (a-d): thin lines: 1.25 counts per winter, thick lines: 2.5 counts per winter. For storms associated with multistation events, track count per path: (e) 7, (f) 1, (g) 27, (h) 9. Multi-station events defined here as: 3 or more stations exceeding their 3-year return level. Dashed lines show crosshairs



846 Figure 8: Cyclone track association for area average of 925-hPa reanalysis winds in black boxes:

- 847 Latitude range for all boxes:  $40^{\circ}N 43^{\circ}N$ . Longitude ranges: (a)  $77.5^{\circ}W-70^{\circ}W$ , (b)  $67.5^{\circ}W-$
- $60^{\circ}$ W, and (c)  $57.5^{\circ}$ W  $50^{\circ}$ W. Red line indicates the cyclone tracks, blue dot marks location of
- 849 cyclone at time of association with high wind event for the area-averaged wind in the box.
- 850



853 Figure 9: (a) 1-year MAX return level on site basis; (b)-(c) as (a) but for 3-year and 5-year return

levels.