



Tropical and Subtropical North Atlantic Vertical Wind Shear and Seasonal

Tropical Cyclone Activity

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ABSTRACT

7 Given recent insights into the role of anticyclonic Rossby wave breaking (AWB) in driving
8 subseasonal and seasonal North Atlantic tropical cyclone (TC) activity, this study further examines
9 tropical versus subtropical impacts on TC activity by considering large-scale influences on boreal
10 summer tropical zonal vertical wind shear (VWS) variability, a key predictor of seasonal TC
11 activity. Through an empirical orthogonal function analysis, it is shown that subtropical AWB
12 activity drives the second mode of variability in tropical zonal VWS, while El Niño-Southern
13 Oscillation (ENSO) primarily drives the leading mode of variability. Linear regressions of the four
14 leading principal components against tropical North Atlantic zonal VWS and accumulated cyclone
15 energy show that, while the leading mode holds much of the regression strength, some improvement
16 can be achieved with the addition of the second and third modes. Furthermore, an index of AWB-
17 associated VWS anomalies, a proxy for AWB impacts on the large-scale environment, may be a
18 better indicator of summertime VWS anomalies. The utilization of this index may be used to better
19 understand AWB's contribution to seasonal TC activity.

20 **1. Introduction**

21 While forecast schemes predicting North Atlantic basin seasonal tropical cyclone (TC) activity
22 have skill (Klotzbach et al. 2017), there remain gaps in our current understanding of the large-scale
23 mechanisms that influence the overall activity within the ocean basin. A key influence on TC activity
24 is vertical wind shear (VWS), which acts to disrupt the TC circulation and reduce overall activity
25 (Nolan and McGauley 2012). Until recently, the scientific literature indicated that the drivers
26 of VWS on seasonal time-scales were predominantly tropical in origin with El Niño-Southern
27 Oscillation (ENSO) being one of the leading drivers of interannual zonal VWS variability within
28 the tropical North Atlantic (Goldenberg and Shapiro 1996; Wang 2004; Aiyyer and Thorncroft
29 2006). However, extratropical sources of shear resulting from Rossby wave breaking (RWB) have
30 recently been identified as an important dynamical influence on TC activity (Zhang et al. 2016;
31 Papin 2017). In this study, we examine the physical drivers of VWS variability in the tropical North
32 Atlantic, with the aim of improving our understanding of both tropical and extratropical sources of
33 VWS and their impacts on TC activity.

34 A common mechanism proposed for ENSO forcing of the Atlantic tropical circulation is through
35 its modulation of the Walker circulation over the Pacific and North Atlantic basins (Wang 2004).
36 Anomalously warm SSTs in the equatorial central and eastern Pacific associated with El Niño
37 result in an eastward-shift of the Walker Circulation from the tropical western Pacific towards the
38 International Date Line with associated anomalous downdrafts over the tropical Atlantic. This shift
39 produces anomalous westerly upper-level winds and an associated increase in westerly VWS that
40 inhibits TC development across the tropical North Atlantic MDR (Gray 1984; Zhang et al. 2016,
41 2017), generally defined as the region between 10°N - 20°N and 80°W - 20°W . The reverse effect
42 occurs with an anomalously cold equatorial central and eastern Pacific (e.g., La Niña).

43 Atlantic SST variability is also a major driver of tropical North Atlantic VWS (Chelliah and Bell
44 2004; Chiang and Vimont 2004; Kossin and Vimont 2007; Klotzbach and Gray 2008). The Atlantic
45 Meridional Mode (AMM) is the leading dynamical mode in tropical Atlantic climate variability,
46 identified from a maximum covariance analysis of Atlantic sea surface temperature (SST) and
47 10-m winds (Chiang and Vimont 2004). The AMM is known to drive both interannual and decadal
48 variability of vertical wind shear through its modulation of meridional gradients in SSTs and
49 subsequent shifts in the Inter-tropical Convergence Zone (ITCZ) and the Hadley circulation. A
50 positive phase of the AMM is associated with an anomalously warm northern tropical Atlantic
51 relative to the southern tropical Atlantic. When the AMM is positive, the ITCZ and ascending arm
52 of the Hadley cell shifts further north, reducing upper-level westerly winds throughout the Atlantic
53 MDR. With a negative phase of the AMM, the ascending branch of the Hadley circulation shifts
54 further south, increasing upper-level westerlies and thus increasing westerly shear over the MDR.

55 In contrast to the tropical forcing of shear by ENSO and the AMM, the mid-latitude forcing arises
56 primarily from RWB events characterized by an irreversible deformation of potential vorticity (PV)
57 contours on an isentropic surface forced by a strong temperature or pressure gradient (McIntyre
58 and Palmer 1983; Strong and Magnusdottir 2008; Papin 2017). RWB events often result in the
59 exchange of air between the drier midlatitudes and moister tropics via intrusions of equatorward
60 high PV midlatitude tropospheric air and poleward low PV tropical air (Postel and Hitchman 1999).
61 RWB activity also shows a strong inter-relationship with seasonal variations in both tropical North
62 Atlantic VWS and TC activity (Papin 2017; Zhang et al. 2017). Anticyclonic RWB (AWB) occurs
63 more frequently in the summer than in the winter over the North Atlantic basin and has been
64 examined in recent studies with a focus on Rossby wave breaking dynamics and their relationship
65 with large-scale climate phenomena (Homeyer and Bowman 2013; Papin 2017; Zhang et al. 2016,
66 2017; Zavadoff and Kirtman 2019).

67 Seasonal TC forecast verifications have previously highlighted the role of midlatitude interactions
68 in suppressing TC development, such as the 2007 North Atlantic hurricane season (Klotzbach et al.
69 2007) and the 2013 season (Klotzbach and Gray 2013; Saunders and Lea 2014). Zhang et al. (2016)
70 indicated that enhanced RWB frequency in 2013 was associated with strong VWS and reduced
71 precipitable water within the North Atlantic MDR that effectively suppressed TC development. As
72 with the 2007 hurricane season, strong suppression in 2013 occurred despite otherwise favorable
73 environmental conditions, such as anomalously warm tropical Atlantic sea surface temperatures
74 (SSTs) and a persistent ENSO-neutral phase (Klotzbach and Gray 2013). However, modulations
75 in RWB-induced shear occur on much faster timescales than SST or ENSO, with RWB events and
76 their impacts typically lasting no more than a couple of days (Li et al. 2018).

77 Another driver of VWS and TC variability within the tropical North Atlantic is atmospheric
78 variability closely tied to African Sahel rainfall dynamics (Landsea and Gray 1992; Aiyyer and
79 Thorncroft 2006). Earlier seasonal forecast schemes included African Sahel rainfall as a robust
80 predictor of TC activity (Gray et al. 1994). While Sahelian rainfall variability and its impacts
81 are not fully understood (e.g. Janicot et al. (1998), Janicot et al. (2001)), it is generally held
82 that Sahelian rainfall has a positive correlation with North Atlantic TC activity and a negative
83 correlation with VWS (Landsea and Gray 1992). Enhanced convection over the Western Sahel
84 results in upper-tropospheric easterly wind anomalies and a reduction of the climatological westerly
85 shear within the NA region, promoting both TC development and intensification (Landsea and
86 Gray 1992). Chelliah and Bell (2004) also suggest that VWS may be driven by a stationary wave
87 response in the upper-level winds to anomalous heating from West African convection. Earlier
88 work identified the effect of both ENSO and Sahel rainfall on the variability of the North Atlantic
89 tropical circulation (Goldenberg and Shapiro 1996; Aiyyer and Thorncroft 2006) but did not fully
90 explore extratropical influences on the variability. Dunion (2011) highlighted that midlatitude dry-

91 air intrusions comprised a small but significant percentage of tropical North Atlantic atmospheric
92 soundings and suggested that dry-air intrusions from the subtropics were a persistent feature within
93 the tropical circulation.

94 In the analysis that follows, we further examine the respective contributions of tropical and
95 extratropical factors on tropical North Atlantic VWS variability. We will examine the physical
96 drivers of tropical VWS variability in the North Atlantic and propose a way of quantifying how
97 much extratropical AWB activity contributes to seasonal VWS. We will show that VWS anomalies
98 associated with AWB explain a significant portion of tropical VWS variability and may lend
99 another perspective to the predictability of the seasonal TC environment. The paper is organized
100 as follows: Section 2 outlines the data and procedures used to characterize the different drivers of
101 VWS variability. Section 3 characterizes mean spatial VWS variability in high versus low seasons
102 of ENSO and RWB activity. Section 4 outlines the different spatial and temporal effects of tropical
103 versus subtropical drivers of seasonal tropical VWS anomaly. Section 5 provides a discussion of
104 the results outlined in Section 4 and summarizes the implications for improved seasonal predictions
105 of North Atlantic TC activity.

106 **2. Data and Methods**

107 *a. Data*

108 All atmospheric field data employed in this study are sourced from the European Center for
109 Medium-Range Weather Forecasts (ECMWF) ERA-Interim Reanalysis (Berrisford et al. 2011;
110 Dee et al. 2011). Monthly fields are gridded with a resolution of $0.75^\circ \times 0.75^\circ$ and extend from
111 January 1979 to August 2019. Six-hourly datasets in which AWB-induced anomalies are identified
112 are gridded with a resolution of $2.5^\circ \times 2.5^\circ$ to remove small-scale disturbances (Postel and Hitchman

113 1999; Zhang et al. 2016; Papin 2017). A lowpass-filtered version of the data that suppresses cycles
114 less than four months is subtracted from the original data to remove signals on longer timescales, as
115 suggested by Wang et al. (2010). The seasonal cycle is then removed by subtracting the 1981-2010
116 monthly climatology to obtain seasonal anomaly fields.

117 Monthly SST anomalies from 1982 to 2016 are derived from the National Oceanic and At-
118 mospheric Administration's Optimum Interpolation Sea Surface Temperature version 2 (NOAA
119 OI-SSTv2) dataset (Reynolds et al. 2002). Monthly Sahel precipitation indices are taken from the
120 Joint Institute of the Study of the Atmosphere and Ocean (JISAO) archive (Mitchell 2013). The
121 analysis regions over which the climate indices are averaged are outlined in Table 1.

122 Tropical North Atlantic VWS is defined here as the difference between the 200hPa and 850hPa
123 zonal wind fields over the region 10°N - 30°N and 90°W - 20°W and are standardized over the period
124 1981-2010. We have chosen to use a domain larger than the canonical MDR to ensure that
125 the impacts of large-scale drivers are fully captured within the spatiotemporal analysis, as will be
126 discussed further in Sections 4 and 5. For this study, we only utilize the zonal component of VWS as
127 much of the observed variability in VWS is zonally modulated (Thorncroft et al. 1993; Aiyyer and
128 Thorncroft 2006; Nolan and McGauley 2012). Furthermore, changes in the circulation due to RWB
129 are associated with variations in the strength and position of the subtropical upper-tropospheric
130 westerly jet (Homeyer and Bowman 2013).

131 The Accumulated Cyclone Energy (ACE) index is used to represent overall seasonal TC activity
132 and is defined as the sum of the squares of the six-hourly maximum wind speed for each tropical
133 and sub-tropical cyclone where they at least possess one-minute maximum sustained winds of 34
134 knots (Bell et al. 2000). The ACE index from July-September was calculated from the National
135 Hurricane Center's best track database (HURDAT2; Landsea and Franklin (2013)). Note that we
136 assessed VWS and AWB activity over the shorter July-September (JAS) season, in contrast to the

137 overall hurricane season from June-November as in Papin (2017) or July-October as in Zhang et al.
138 (2017). We chose July-September as both VWS and AWB activity show distinct summertime
139 peaks during these months. To focus on the impacts of tropical VWS, only TCs that were named
140 south of 35°N were considered in the calculation of ACE.

141 *b. Climate Indices*

142 Table 1 presents the climate indices used to investigate the physical mechanisms driving the
143 leading modes of tropical zonal VWS variability. Based on previous studies (for example Aiyyer
144 and Thorncroft (2006)), we analyze the correlations with indices representing ENSO, Sahelian
145 rainfall, the AMM and AWB. Though the Walker Circulation is closely associated with ENSO in
146 the tropical North Atlantic atmospheric circulation, we examine it as a separate index in order to
147 determine the extent to which it is correlated with tropical zonal VWS variability. For Atlantic SST
148 variability, Chiang and Vimont (2004)'s AMM index is applied. The AMM index is calculated
149 using detrended and smoothed SSTs and low-level winds within the region 21°S - 32°N and 74°W -
150 15°E . Once ENSO variability is removed, a maximum covariance analysis is applied to obtain
151 the AMM index. Chiang and Vimont (2004)'s method is used extensively within the scientific
152 literature (Kossin and Vimont 2007; Vimont and Kossin 2007). The strength of the Pacific Walker
153 cell is used as a proxy for the influence of the Walker circulation over the North Atlantic region
154 and is defined as the difference between the 500hPa vertical velocity averaged over the equatorial
155 eastern Pacific (5°S - 5°N and 160°W - 120°W) and the equatorial western Pacific (5°S - 5°N and
156 120°E - 160°E) (Wang 2004). All indices are standardized over 1981-2010, except for SST indices
157 which are standardized over 1982-2010.

158 *c. Detecting AWB*

159 To detect AWB activity, we employ a simplified potential vorticity streamer (PVS) detection
160 algorithm based on the technique outlined by Papin (2017), where the PVS intensity index is
161 calculated. AWB-associated PVSs are detected along the 2-PVU ($1 \text{ PVU} \equiv 10^{-6} \text{ K m}^2 \text{ s}^{-1} \text{ kg}^{-1}$)
162 contour in the 350-K potential vorticity field using the following steps:

- 163 • The algorithm detects more than two consecutive points along the 2-PVU contour with an
164 eastward (west to east) PV gradient ($\frac{\partial PV}{\partial x} > 0$) and a reversal in the poleward meridional PV
165 gradient ($\frac{\partial PV}{\partial y} > 0$). This defines the upstream edge of a PV tongue. Similarly, the downstream
166 edge may be identified with conditions ($\frac{\partial PV}{\partial x} < 0$) and ($\frac{\partial PV}{\partial y} < 0$).
- 167 • Once the points outlining the PV tongue are identified, a line connects the two end points to
168 capture the PVS polygon and as much of the PVS area as possible. We do not assess the PVS
169 area for a perimeter distance threshold or 3:1 aspect ratio of PV tongues as is done by Papin
170 (2017). This may lead to slight differences in the results obtained.
- 171 • The PVS intensity is then found by calculating the standardized PV anomaly relative to a
172 six-hourly climatological mean integrated across all grid points within the PVS polygon and
173 then integrated over time.
- 174 • The VWS anomaly along both the upstream and downstream edges of the detected PV tongue
175 are collected.

176 For the years 1979-2016, our detection algorithm finds 32,014 AWB events in the ERA-Interim
177 dataset. This is somewhat high compared with Papin (2017)'s total of 21,149 between 1979-2015.
178 However, for the July-September focus of our study, the average number of events is 337, similar
179 to Papin (2017)'s value of 355 for the same season. The climatology for the total number of PV

180 streamers detected each month (not shown) is comparable to the climatological mean intensity of
181 AWB activity over the North Atlantic region.

182 Papin (2017) pointed out that the detection algorithms are sensitive not only to the reanalysis
183 dataset used, but also to the method in which the PVS area is detected. The detection method
184 and the PVS area captured varies across studies (Postel and Hitchman 1999; Abatzoglou and
185 Magnusdottir 2006; Wernli and Sprenger 2007; Barnes and Hartmann 2012; Kunz et al. 2015).
186 However, the AWB climatology shown in Fig. 1 is in good agreement with previous assessments
187 of AWB variability over the North Atlantic region (Postel and Hitchman 1999; Abatzoglou and
188 Magnusdottir 2006; Papin 2017; Zhang et al. 2017). The index also correlates positively with
189 ENSO and negatively with North Atlantic ACE. There are some differences in the years detected
190 with above-normal or below-normal AWB activity compared with the results of both Papin (2017)
191 and Zhang et al. (2017). The differences may be due to the season chosen for our assessment. The
192 results also show a sensitivity to the domain chosen. Similar to Zhang et al. (2017), we restrict our
193 detection region to 20° - 40° N and 100° - 5° W, where AWB is most frequent, to better quantify the
194 effect of AWB activity on the tropical circulation. While Zhang et al. (2017) varies the northern
195 boundary of the detection domain, we opt to set the northern boundary at 40° N.

196 *d. EOF Analysis*

197 To analyze the various modes of variability in tropical North Atlantic VWS, the leading empirical
198 orthogonal functions (EOFs) are calculated via an eigenanalysis of the covariance matrix for July-
199 September anomalous tropical North Atlantic VWS over the domain 10° - 30° N and 90° - 20° W. The
200 VWS data is standardized prior to the EOF calculation; only the annual cycle is removed. The first
201 four leading modes are retained for analysis as will be discussed in Section 4 and are regressed
202 against global VWS anomalies to assess possible remote versus local forcing on tropical North

203 Atlantic VWS. The Pearson correlation coefficients between the principal components derived from
204 the EOF analysis and the climate indices outlined in Table 1 are calculated and used to assess how
205 large-scale subtropical forcing differs from large-scale tropical forcing of tropical North Atlantic
206 VWS.

207 **3. Tropical North Atlantic VWS variability relative to AWB**

208 *a. VWS and AWB Climatology*

209 Figure 1 shows a comparison of the 1979-2016 monthly climatology of mean zonal VWS and
210 AWB activity over the tropical North Atlantic region. The VWS climatology (where positive values
211 indicate westerly zonal shear and negative values indicate easterly zonal shear) exhibits maximum
212 westerly shear in January-February. From March onwards, there is a steady decline in westerly
213 shear in conjunction with the onset of the North Atlantic hurricane season in June. This westerly
214 shear reaches its climatological minimum in July-September, as observed in previous studies of
215 mean shear within the North Atlantic MDR (Gray 1968; Aiyyer and Thorncroft 2006). Figure
216 2 shows a spatial plot of the 1979-2016 July-September mean zonal VWS across the Atlantic
217 region. There is strong mean westerly shear cutting through the North Atlantic, stretching from the
218 Caribbean northeast to the subtropical northeastern Atlantic. The strong westerly shear is flanked
219 by strong easterly shear south of 10°N and weaker westerly shear just north of 25°N in the western
220 subtropical North Atlantic, similar to the observations of Gray (1968) for the mean boreal summer
221 shear.

222 Collocated with the peak weakening of westerly shear during July-September is a peak occurrence
223 in North Atlantic AWB activity identified along the +2-PVU contour on the 350K isentrope. This
224 is consistent with previous studies that identified peak anomalies on or around 350K (Postel and

225 Hitchman 1999; Abatzoglou and Magnusdottir 2006; Homeyer and Bowman 2013). Homeyer and
226 Bowman (2013) and Kunz et al. (2015) further explained that equatorward AWB tends to occur
227 in regions of weak mean westerly winds resulting from a weakening and a poleward shift of the
228 subtropical jet over the North Atlantic. The increase in AWB activity is also collocated with mean
229 northerly meridional shear (not shown).

230 In order to maximize the relationship with tropical VWS and subtropical AWB activity, we focus
231 on the July-September season since this is when AWB shows a distinct peak. In the subsection
232 below, we compare the dynamical variations in July-September VWS influenced by ENSO with
233 variations associated with AWB activity.

234 *b. Spatial variability in VWS composites*

235 Figure 3 shows composites of July-September North Atlantic VWS anomalies in years associated
236 with the 12 warmest El Niño events versus the 13 years with the most intense AWB activity
237 (AWB(+)). For AWB, our choice of years are based on the PVSI index outlined in Section 2c;
238 the years chosen are consistent with those identified by Zhang et al. (2016) and Papin (2017).
239 During the warmest El Niño seasons (shown in Fig. 3a), VWS is enhanced over the Atlantic MDR.
240 A similar pattern in the VWS anomaly field is observed in years of intense AWB (Fig. 3b). The
241 difference between the composites (Fig. 3c) shows that ENSO dominates tropical VWS variability.
242 The most significant differences between AWB and ENSO on VWS (assessed using the signed
243 rank test) occur between 20°-35°N.

244 Based on separate EOF analyses of mean July, August and September VWS anomalies, we find
245 that both July and September have AWB patterns for their second leading monthly modes, similar
246 to those for August. This is consistent with the findings of Zhang et al. (2016) who conducted a
247 similar EOF analysis for the North Atlantic environment in August. We further suggest that the

248 effects of AWB on the Atlantic environment and consequently, TC activity, may also be observed
249 not only in August but in July and September as well. The August composites (not shown) show
250 similar features to the July-September composites, with a pronounced stretch of westerly shear
251 associated with anomalously warm SSTs driven by El Niño. The anomalously strong subtropical
252 jet stream, just north of the African Sahel, is also more evident in the August composites. However,
253 there are few areas with significant differences between the August El Niño and AWB composites
254 within the North Atlantic region.

255 In Figure 4, the shear anomaly composite for the 12 coldest La Niña years is compared with a
256 composite of the 13 years with the least intense AWB activity. Figure 4 composites show a similar
257 but opposite effect on VWS from those in Fig. 3. As expected, there is very little difference between
258 the La Niña and AWB(-) composites shown in Fig. 4c. Unlike the El Niño-AWB(+) composites,
259 the most significant vertical wind shear difference in the La Niña-AWB(-) composite occurs in the
260 North Atlantic north of 35°N.

261 There is a great deal of overlap between extreme ENSO seasons and AWB activity, as indicated
262 by the years used to create the composites. Further attempts to separate the two effects indicate
263 that the impacts of tropical and subtropical drivers are not completely separable due to strong
264 tropical-subtropical teleconnections, for example, the AMM-NAO relationship (Grossmann and
265 Klotzbach 2009). The VWS field minus the regressed influences of ENSO and the AMM show no
266 significant correlation with the JAS PVS index (not shown). Similarly, a frequency separation by
267 applying a highpass filter on the frequency of days shows little evidence of the characteristic wave
268 breaking pattern.

269 This suggests that the mechanisms of the two drivers are related, and that ENSO may drive part
270 of AWB variability (Lau and Nath 1996; Martius et al. 2008) or that the effects of ENSO and its
271 teleconnections partially overshadow the effects of other possible drivers. Our AWB activity index

272 has a correlation of 0.14 with the Niño-3.4 index. Zhang et al. (2016) and Papin (2017) showed
273 a correlation of ~ -0.3 between ENSO and AWB activity. While ENSO and AWB activity are
274 not strongly correlated, ENSO dominates VWS variability and masks the effect of AWB on VWS
275 anomalies in both Figs. 3 and 4. Also, significant differences in the composites are observed within
276 the Niño-4 region associated with warm-pool (WP) ENSO events (Ashok et al. 2007). This may
277 suggest that WP ENSO has a role to play in driving North Atlantic AWB and its effects on tropical
278 VWS. In Section 4, we discuss the results of an EOF analysis of mean tropical VWS anomalies
279 to further explore the different effects of ENSO and AWB on VWS in the North Atlantic region.
280 Note that for the remainder of the study, the abbreviation "EOF" refers to the spatial patterns of
281 the leading modes observed in Fig. 6, while "PC" refer to the principal components or temporal
282 variations associated with each EOF mode, shown in Fig. 7.

283 **4. Modulation of VWS by AWB in EOFs**

284 *a. Eigenanalysis of tropical North Atlantic VWS anomaly*

285 Figure 5 illustrates the variance explained by the first 20 EOFs of July-September tropical North
286 Atlantic zonal VWS. Most of the structured variability in tropical VWS can be accounted for in
287 the first two EOFs which together explain $\sim 59\%$ of the explained variance. While EOFs 3 and 4
288 show some continuity with the tail end of the spectrum (explaining 12% and 8% of the variance,
289 respectively), we believe that EOFs 3 and 4 are sufficiently separated from the remaining EOFs to
290 have some physical significance in explaining VWS variability. Based on the criteria outlined by
291 North et al. (1982), we opt to retain the first four EOFs that together account for 79% of the total
292 variance in zonal VWS in the tropical North Atlantic.

293 Figure 6 shows a regression of the first four PCs onto July-September global zonal VWS anoma-
294 lies. The first leading mode of variability (EOF1) shown in Fig. 6a accounts for 36% of the
295 observed variance. The strongest spatial signal, that extends well outside of the North Atlantic
296 region, is mostly confined to the tropical belt and exhibits a tongue-like feature within the Niño-3
297 region reminiscent of the ENSO signal exhibited in Figs. 3a and 4a and the tropical interannual
298 mode examined in Chelliah and Bell (2004). Correlations with SST anomalies within the four
299 ENSO regions indicate a strong association of temporal variations of EOF1 (PC1) with ENSO
300 variations as shown in Table 2. We note that PC1 shows a significant correlation with all indices
301 used in the analyses. This may be due to PC1 accounting for most of the structured variance
302 in tropical VWS. Variations in both the Walker Circulation and Sahel rainfall are known to have
303 teleconnections with equatorial Pacific and Atlantic SST variability (Janicot et al. 1998, 2001).
304 Of the four ENSO indices, the July-September Niño-3.4 index has the strongest correlation with
305 PC1 ($r = 0.73$). PC1 also shows a significant correlation of $r = -0.53$ with the Atlantic Meridional
306 Mode (AMM). Based on Fisher's r-to-z transformation (Lee and Preacher 2013), the correlation
307 between Niño-3.4 and PC1 is significantly higher than the correlation between the AMM and PC1.
308 This result is expected as ENSO is the dominant driver of tropical interannual variability, and the
309 influence of the AMM-VWS relationship is less dominant at high-frequency timescales (Chelliah
310 and Bell 2004; Vimont and Kossin 2007).

311 Figure 6b displays the second leading mode (EOF2) that accounts for 23% of the structured
312 variance. The EOF2 pattern shows a locally confined lobe of westerly shear (positive zonal
313 anomalies) sandwiched between regions of easterly shear (negative zonal anomalies) stretching
314 across the North Atlantic region. EOF2 exhibits features associated with active subtropical wave
315 breaking that have been identified in zonal anomalies by Homeyer and Bowman (2013) and Zhang
316 et al. (2016). In contrast to PC1, correlations with the tropical indices outlined in Table 2 are

317 substantially reduced for PC2, further suggesting that the driver of the second mode of variability
318 is subtropical in nature. AWB activity spurs anomalous easterly shear over the northernmost
319 section of the tropical Atlantic with westerly VWS in the southernmost section of the Atlantic
320 MDR as shown in Fig. 3b. Therefore, AWB(+) years are indicated by strong positive anomalies
321 within the PC2 time series. It is also notable that AWB-associated shear shows a correlation of $r =$
322 0.43 with PC1. We expect that PC1 will capture much of the variability in tropical VWS including
323 impacts from AWB activity. ENSO and the AMM may also be indirect drivers of AWB activity
324 by modulating large temperature gradients and driving large-scale features such as the Walker
325 and Hadley Circulations and thereby triggering AWB (Matthews and Kiladis 1999; Papin 2017;
326 Zavadoff and Kirtman 2019; Zhang and Wang 2019).

327 EOF3 in Fig. 6c shows a region with strong westerly VWS anomalies, flanked to the north
328 and south by anomalous easterly shear in the equatorial eastern Pacific region, indicative of an
329 anticyclonic circulation. This strong westerly shear anomaly extends into the western North
330 Atlantic region, while the eastern North Atlantic region is affected by anomalous easterly shear.
331 We hypothesize that the EOF3 pattern is related to the tropical North Atlantic VWS's response to
332 variations in the Walker Circulation. The third principal component (PC3) has a correlation of -0.34
333 with the Walker Circulation index (see Table 2). Also, Arkin (1982) found that the anticyclonic
334 pattern shown in EOF3 was generally associated with a warm ENSO phase and a weak phase of the
335 Walker Circulation. PC3 also shows a moderate relationship with North Atlantic SSTs (Fig. 8c),
336 which is not surprising given its relationship to anomalous variations in the Walker Circulation
337 (Fig. 7c). We further point out that for PC3, the North Atlantic drives the pressure gradient shown
338 in Fig. 9c and may indicate the possible role of the state of the equatorial Pacific relative to the
339 tropical North Atlantic in modulating North Atlantic VWS variability.

340 The fourth leading mode (EOF4) shown in Fig. 6d features a tongue of anomalous zonal shear
341 extending from the African Sahel into the tropical North Atlantic MDR. Figure 6d shows easterly
342 shear with a wetter-than-normal Sahel, consistent with the studies of Karnauskas and Li (2016) and
343 Dunion (2011) that highlighted the role of Sahel dynamics in influencing the tropical North Atlantic
344 environment. EOF4 explains 8% of the variability in zonal shear in the North Atlantic region.
345 African Sahel rainfall may likely contribute more to variations in meridional shear as rainfall over
346 Africa is often modulated by a north-south shift of the Intertropical Convergence Zone (ITCZ).
347 A wetter-than-normal Sahel induces anomalously easterly 200-hPa winds (resulting in anomalous
348 easterly shear) while drought-like conditions over the Sahel favor westerly 200-hPa zonal wind and
349 shear anomalies (Zhang and Delworth 2006). Therefore, tropical VWS has an inverse relationship
350 with Sahel rainfall ($r = -0.41$, as shown in Table 2). This is mirrored in correlations between PC4
351 and Sahel rainfall ($r = -0.33$).

352 We note here that the AWB signal featured in EOF2 does not appear in EOF analyses of VWS
353 averaged over the Atlantic MDR generally defined by the area 10°N - 20°N and 80°W - 20°W . In the
354 EOFs formed from MDR VWS, more than 50% of the explained variance is accounted for by the
355 leading mode of variability and ENSO. African Sahel rainfall is also identified as a major driver,
356 consistent with Goldenberg and Shapiro (1996) and Aiyyer and Thorncroft (2006). Our results
357 suggest that variability between 20°N - 30°N is necessary to fully capture the impacts of AWB on
358 VWS.

359 *b. AWB-associated VWS anomaly (AWB-VWS) as a predictor of summertime shear*

360 Various studies show that the environmental impact of AWB events can be observed in the
361 modulation of deep-layer VWS (Zhang et al. (2016); Papin (2017); Li et al. (2018)). We create
362 an index of the sum of the shear anomalies collected along the downstream edge of potential

363 vorticity streamers, which is the edge most associated with an increase in westerly shear in the
364 North Atlantic MDR (Papin 2017; Zhang et al. 2017). The index, referred to as AWB-VWS, shows
365 AWB(+) years to have positive (westerly) anomalies within the North Atlantic, while AWB(-) years
366 show negative (easterly) anomalies indicating a weak downstream anticyclonic circulation. The
367 AWB-VWS index shows a strong positive correlation with tropical VWS and its first and second
368 modes of variability, consistent with our dynamical explanations of EOF1 and EOF2. As shown in
369 Table 2, AWB-VWS has a 0.57 correlation with tropical North Atlantic VWS anomalies and a 0.53
370 correlation with PC2. Therefore, EOF2 is strongly influenced by the large easterly shear anomalies
371 lining the northern edge of the tropical North Atlantic region (see Fig. 6b).

372 Table 3 outlines the correlations of the time series of PCs1-4 and the AWB-VWS index with
373 seasonal tropical North Atlantic shear and ACE south of 35°N. As expected, PC1 shows the
374 strongest correlations with JAS VWS, but also shows strong correlations with shear prior to and
375 after the shear's peak in the JAS season. Correlations with PC2 are weak prior to JAS and show little
376 correlation with tropical VWS in SON. This is also reflected in correlations with the AWB-VWS
377 index and suggests that AWB activity has the greatest impact on contemporaneous seasonal shear
378 anomalies. This relationship was also observed for subseasonal modulations of shear by AWB
379 activity (Li et al. 2018). Compared to PC2, the AWB-VWS index shows a stronger correlation with
380 shear outside of the contemporaneous shear season. Correlations with PC3 and PC4 are weaker
381 than the PC1, PC2 and AWB-VWS indices. PC3 has stronger correlations with shear in the ASO
382 and SON seasons, while PC4 has weak seasonal correlations.

383 While the second leading mode of variability only accounts for 23% of the explained variance,
384 previous studies (Papin 2017; Zhang et al. 2016, 2017) have already indicated the likely influence
385 of AWB-associated VWS on seasonal TC activity. We expect that a better understanding of the
386 contribution of each driver to summertime VWS variability will ultimately improve our under-

387 standing of the drivers of seasonal TC activity in the North Atlantic. Tropical zonal VWS has a
388 correlation of -0.43 with July-September ACE (see Table 3). We further examine the ability of the
389 AWB-VWS index as a predictor of TC seasonal variability in Tables 4 and 5. Table 4 shows the
390 average number of various metrics of TC activity for years corresponding to the five highest and
391 lowest values for PC1 and PC2. As expected, changes in PC1 show a stronger change in the TC
392 metrics. The PC2-TC relationship is less consistent with only modest changes in the TC metrics.

393 We further quantify the contribution of PC2 and AWB-VWS to a statistical linear prediction
394 model for ACE based on zonal VWS. The four leading modes are regressed against ACE using
395 stepwise regression, and the contribution of each mode to the regression strength is outlined in
396 Table 5. A major contribution to the regression is indicated by a decrease in the root mean squared
397 error (RMSE), an increase in the variance explained (r^2) and an increase in the significance of
398 the variance or F-statistic of the linear combination. PC1 holds most of the regression strength
399 and is capable of being a stand-alone predictor. The addition of PCs 2 and 3 further lowers the
400 RMSE, and improves the variance explained, though changes to the F-statistic are modest. The F-
401 statistic weakens with the addition of PC4. The AWB-VWS index shows similar skill at simulating
402 ACE compared to the PC1-PC2 model combination. While the contribution may be modest, the
403 inclusion of AWB impacts on shear into statistical schemes for operational seasonal prediction
404 shows improvement in accounting for years driven predominantly by AWB. In future research, we
405 will analyze the inclusion of the AWB-VWS index in Colorado State University's statistical TC
406 forecast scheme.

407 Table 5 also shows that PC3 contributes a significant increase to the regression strength against
408 the July-September ACE index, increasing the variance explained from 27% to 44%. Following
409 a similar stepwise regression approach as described above, with PC3 as the only predictor, PC3
410 explains ~ 15% of the observed variance with a regression strength of 7.3. Based on these values,

411 the Walker circulation-associated PC3 may not be the most suitable stand-alone predictor for ACE,
412 but it does seem to explain an important portion of variance not already covered by PCs 1 and 2. In
413 contrast to the other PCs, PC3 has a strong tropical eastern Atlantic signal (Fig. 6c), possibly due
414 to SLP variations over the tropical South Atlantic (Fig. 9c). The regression statistics suggest that
415 the PC-1+2+3 combination is a stronger representation of the overall impact of VWS on seasonal
416 TC variability.

417 **5. Discussion and Conclusions**

418 In this study, both tropical and extratropical contributions to the variability of seasonal 200-850
419 hPa zonal vertical wind shear in the tropical North Atlantic region were identified using compositing
420 and EOF analysis. Major findings of this analysis include:

- 421 1. The first leading mode of variability in tropical North Atlantic zonal VWS accounts for 36%
422 of the structured variance and is driven by interannual variations in ENSO and the AMM,
423 suggesting that tropical sources of shear are the dominant contributor to VWS.
- 424 2. Anti-cyclonic wave breaking (AWB) activity is shown to be associated with the second EOF
425 mode and accounts for 23% of the structured variance. While not as strong as ENSO, this
426 extratropical source of shear is a significant contributor to VWS variability and TC activity.
- 427 3. The third leading mode is associated with the pressure gradient likely modulated by the Walker
428 Circulation, accounting for 12% of tropical VWS.
- 429 4. African Sahel rainfall is associated with the fourth mode of variability in high-frequency
430 variations of tropical North Atlantic zonal VWS, accounting for 8% of the structured variance.

431 While the leading EOF modes are by design orthogonal, there remains some shared physical
432 relationships between the drivers of each of the four leading modes. As observed in Section 3

433 above, the influence of AWB on VWS activity is difficult to characterize due to ENSO's strong
434 influence on the tropical North Atlantic region (see Figs. 3 and 4). One limitation of the study
435 was the inability to separately analyze the contributions of AWB to tropical VWS variability due
436 to substantial overlap with strong ENSO events. Even if the threshold definitions of ENSO and
437 AWB events are relaxed, there are still not enough samples of anomalous AWB years with neutral
438 ENSO conditions or anomalous ENSO years with neutral AWB conditions to fully differentiate
439 their impacts on the observed environment. The overlap raises a question about how much ENSO
440 imprints on AWB variability and the subsequent relationship with seasonal TC activity. Therefore,
441 a complete separation of the tropical and subtropical influences has not been achieved. The
442 dominance of ENSO may be a key reason for the inability to observe significant impacts of AWB
443 on seasonal TC activity during all years with anomalous AWB, as observed by Li et al. (2018).

444 The results presented in Section 4 suggest that the tropical sources of VWS from ENSO are
445 indeed dominant, but that extratropical sources of VWS from AWB are an important contribution
446 to tropical and subtropical VWS variability and TC activity. The present analysis shows that the
447 second mode of variability of tropical North Atlantic VWS may be attributed to subtropical AWB
448 activity (Galarneau et al. 2015; Zhang et al. 2017). Our results are consistent with Zhang et al.
449 (2016) and Zhang et al. (2017), who previously highlighted the dynamical role of AWB in driving
450 seasonal TC variability through modulations of VWS. The current study adds to their findings
451 by quantifying the AWB impact on the tropical North Atlantic VWS relative to ENSO's strong
452 influence on the tropical North Atlantic summertime circulation.

453 Another key result is the overshadowing of an AWB signal in the leading modes of interannual
454 variability when restricting analyses of unfiltered tropical North Atlantic VWS to only the main
455 development region (MDR). The MDR does not fully encompass the spatiotemporal patterns
456 driving VWS within the North Atlantic region where TCs are prevalent. The MDR domain is

457 large enough to incorporate the impacts of both ENSO and the African Sahel on VWS (Aiyyer and
458 Thorncroft 2006), but not the impacts of AWB. Extending the domain to 30°N better captures the
459 full extent of the leading modes of VWS variability that impact TCs.

460 This study has focused on zonal VWS due to the fact that meridional VWS variability accounts
461 for a smaller portion of the overall horizontal shear variability. There may be impacts from large-
462 scale variations in the Atlantic Multidecadal Mode (Patricola et al. 2016) that modulate meridional
463 variations in VWS over the wider North Atlantic (Vimont and Kossin 2007). Further analysis of
464 meridional shear variability is recommended for future work.

465 While recent studies have improved our understanding of the variability of both deep-layer shear
466 and AWB activity, their impacts on TC activity are not well-documented and warrant further
467 research. By examining the deep-layer shear directly, we take the first step to assessing the
468 predictability of VWS and quantifying the response of the environment, and consequently TC
469 activity, with respect to each large-scale driver. VWS anomalies induced by AWB activity (AWB-
470 VWS) may be a better indicator of AWB's impacts on VWS compared to directly using the second
471 leading EOF mode as an index. The AWB-VWS index calculated has a significant correlation with
472 both tropical North Atlantic VWS ($r = 0.57$) and seasonal ACE ($r = -0.50$), suggesting its possible
473 use as a predictor in seasonal TC forecasting. The results of this study highlight the importance
474 of AWB on seasonal North Atlantic TC activity in the summertime, in addition to the impacts of
475 ENSO, the Walker circulation, and Sahel rainfall. Future work will further explore the dynamical
476 relationships between large-scale atmospheric and oceanic drivers of tropical climate and VWS
477 variability, and how AWB variability may be further incorporated into forecasts of seasonal TC
478 activity.

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630 **LIST OF TABLES**

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649 TABLE 1. Regions used to define the four ENSO indices (Niño-1+2, Niño-3, Niño-3.4, and Niño-4), the African
 650 Sahel index, the anticyclonic Rossby wave breaking (AWB) index, the Atlantic Meridional Mode (AMM), and
 651 the Walker Circulation index. All indices are standardized relative to the 1981-2010 base period.

<i>Index</i>	<i>Region</i>	<i>Reference</i>
Niño-1+2	10° S-0°, 90° W-80° W	NOAA (2017)
Niño-3	5° S-5° N, 150° W-90° W	NOAA (2017)
Niño-3.4	5° S-5° N, 170° W-120° W	NOAA (2017)
Niño-4	5° S-5° N, 160° E-150° W	NOAA (2017)
Sahel	10° N-20° N, 20° W-10° E	Mitchell (2013)
AWB	20° N-40° N, 100° W-5° W	Zhang et al. (2016); Papin (2017)
AMM	21° S-32° N, 74° W-15° E	Zhang et al. (2016); Chiang and Vimont (2004)
Walker Index	equatorial eastern Pacific: 5° S-5° N, 160° W-120° W equatorial western Pacific: 5° S-5° N, 120° E-160° E	Wang (2004)

652 TABLE 2. Pearson correlation between the four leading principal component (PC) time series of tropical North
653 Atlantic VWS with ENSO, PVSI, Sahel, and AMM indices. Correlations statistically significant (based on the
654 two-tailed p-value) are highlighted in italics. The strongest correlation with each PC is highlighted in bold.

<i>Index</i>	<i>VWS</i>	<i>PC1</i>	<i>PC2</i>	<i>PC3</i>	<i>PC4</i>
Niño-1+2	<i>0.46</i>	<i>0.54</i>	-0.13	0.42	-0.11
Niño-3	<i>0.65</i>	<i>0.69</i>	-0.02	<i>0.37</i>	-0.09
Niño-3.4	<i>0.73</i>	0.73	0.10	0.29	-0.09
Niño-4	<i>0.73</i>	<i>0.63</i>	0.27	<i>0.35</i>	-0.09
Walker Index	<i>-0.38</i>	<i>-0.48</i>	0.07	<i>-0.34</i>	0.04
Sahel	<i>-0.41</i>	<i>-0.56</i>	0.07	0.28	-0.33
AMM	<i>-0.49</i>	<i>-0.53</i>	-0.20	-0.12	0.17
AWB-VWS	<i>0.57</i>	<i>0.43</i>	0.53	-0.01	-0.21

655 TABLE 3. Correlations of the four principal components (PC1, PC2, PC3, PC4) and the AWB-VWS index
 656 with seasonal tropical North Atlantic VWS and North Atlantic ACE south of 35°N. Correlations statistically
 657 significant at the 95% level are highlighted in bold.

	<i>PC1</i>	<i>PC2</i>	<i>PC3</i>	<i>PC4</i>	AWB-VWS
JJA VWS	0.82	0.35	-0.13	-0.08	0.48
JAS VWS	0.86	0.40	-0.14	-0.16	0.57
ASO VWS	0.72	0.35	-0.31	-0.08	0.53
SON VWS	0.45	0.09	-0.40	-0.23	0.18
JJA ACE	-0.30	-0.29	-0.23	0.26	-0.49
JAS ACE	-0.46	-0.32	-0.42	0.17	-0.50
ASO ACE	-0.56	-0.26	-0.32	0.16	-0.58
SON ACE	-0.65	-0.21	-0.32	0.11	-0.53

658 TABLE 4. The average number of named storms, hurricanes, major hurricanes and ACE in the five years with
 659 the highest (lowest) standardized values for PC1 and PC2.

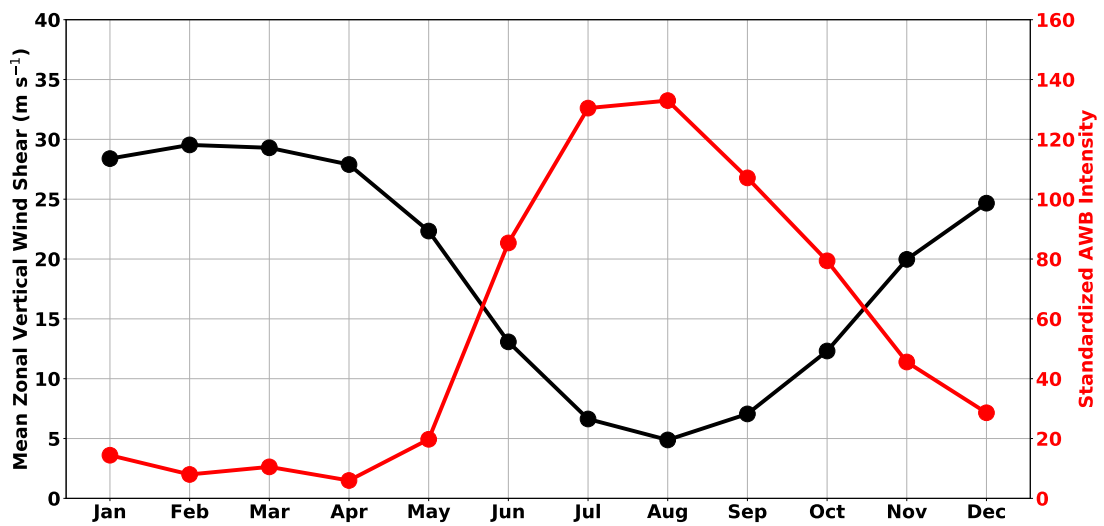
	PC1-top five	PC1-bottom five	PC2-top five	PC2-bottom five
Named storms	9	14	13	12
Hurricanes	3	8	5	6
Major hurricanes	2	4	1	3
ACE	44.3	142.5	72.4	80.0

660 TABLE 5. The root mean squared error (RMSE), variance explained (r^2), and F-statistic associated with the
661 stepwise regression of VWS, AWB-VWS and the four leading modes against the July-September ACE index.

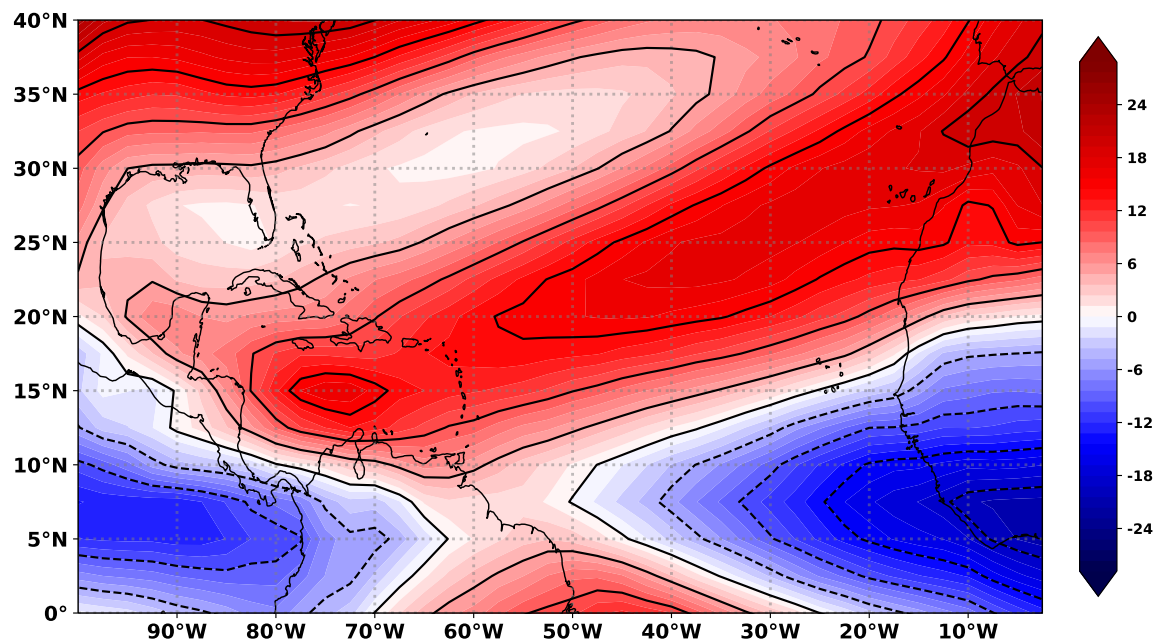
Combination	RMSE	r^2	F
VWS only	14.30	0.17	8.61
AWB-VWS only	13.74	0.24	12.37
PC 1 only	14.35	0.17	8.33
PCs 1+2	13.36	0.27	8.09
PCs 1+2+3	11.75	0.44	10.70
PCs 1+2+3+4	11.88	0.43	7.91

LIST OF FIGURES

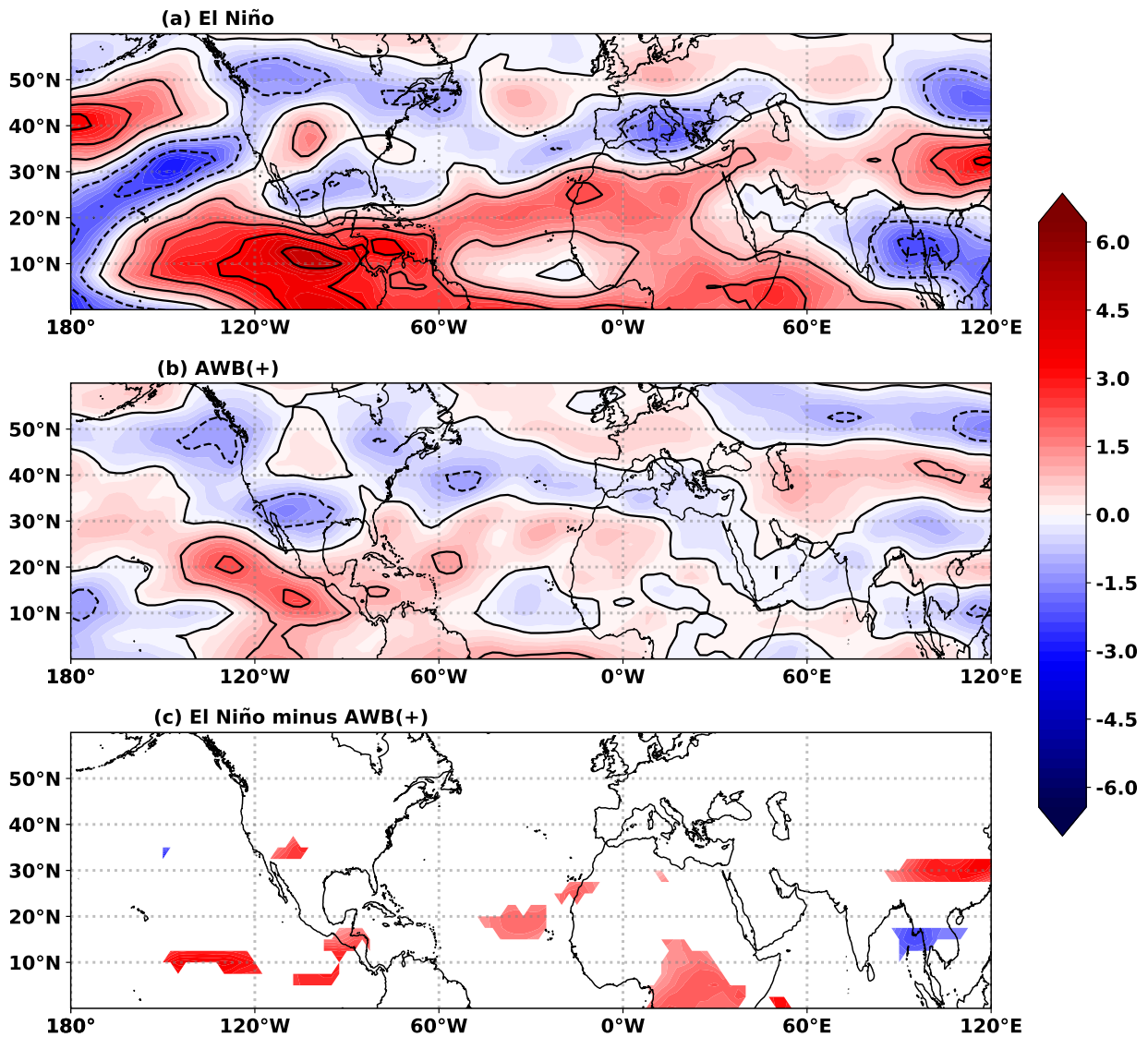
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680	Fig. 5.	Spectrum of the covariance matrix showing the percentage variance explained by the first 20 EOFs with the bars representing the 95% error bounds for each EOF. The error distribution of each EOF is calculated using North et al. (1982)'s "rule of thumb". 40
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686	Fig. 7.	The first four principal components (PC) of July-September tropical North Atlantic zonal VWS anomalies. The PCs are expressed as a unit variance from the zero-mean. Each PC (in black) is plotted along with the July-September climate index (in blue) showing the largest correlation with the PC. PCs 3 and 4 are multiplied by a factor of -1 to better highlight the correlation between the corresponding climate index. The Walker circulation and Sahel rainfall indices are standardized to be comparable to PCs 3 and 4. The blue dashed lines indicate ± 1 standard deviation. 42
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693	Fig. 8.	Spatial Pearson correlations of the first four principal components (PC1, PC2, PC3 and PC4) of tropical North Atlantic VWS with global mean sea surface temperatures. Colored shading indicates correlations statistically significant at the 95% level. 43
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696	Fig. 9.	As in Figure 8, but for global mean sea level pressure. Colored shading indicates correlations statistically significant at the 95% level. 44
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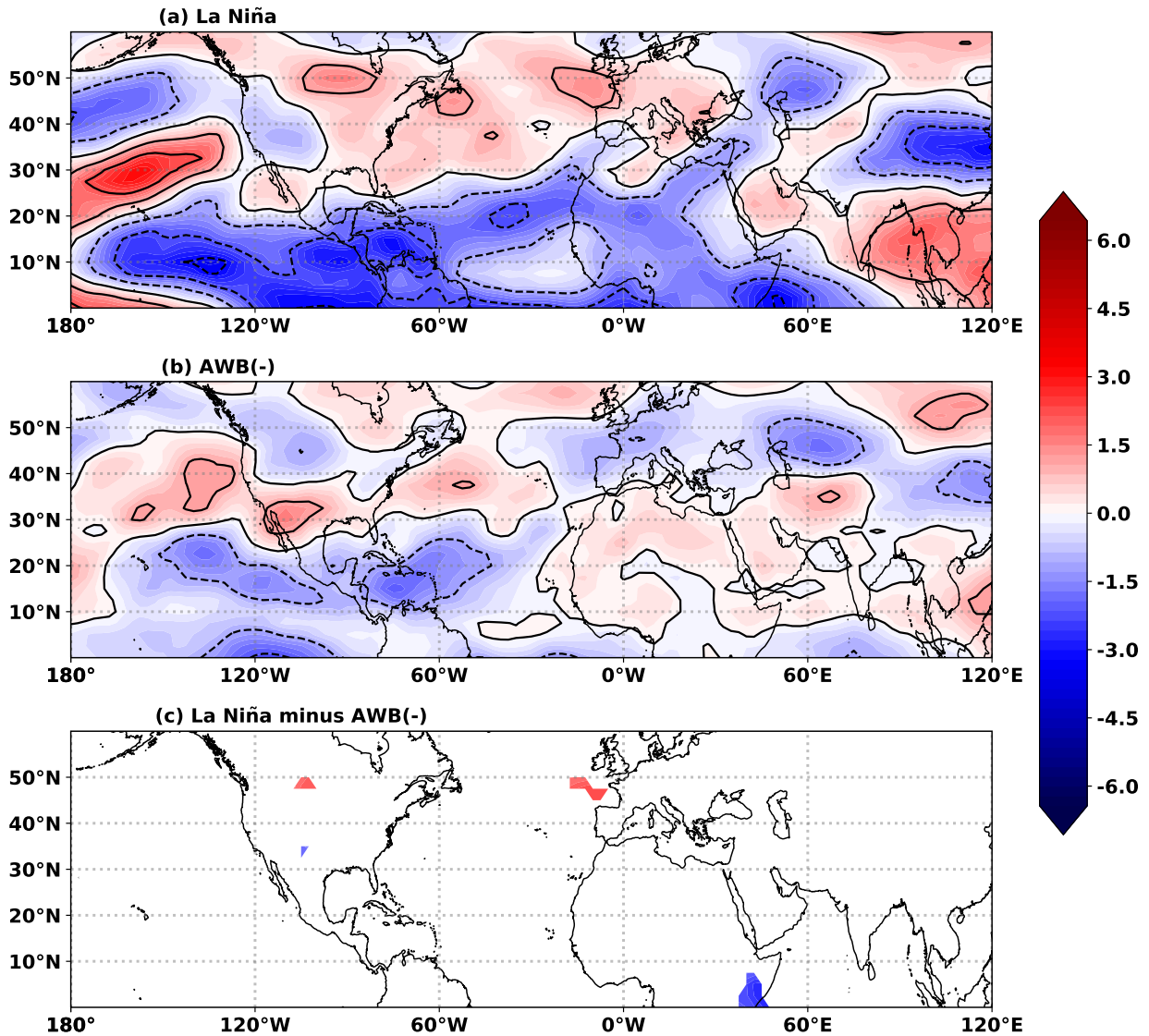
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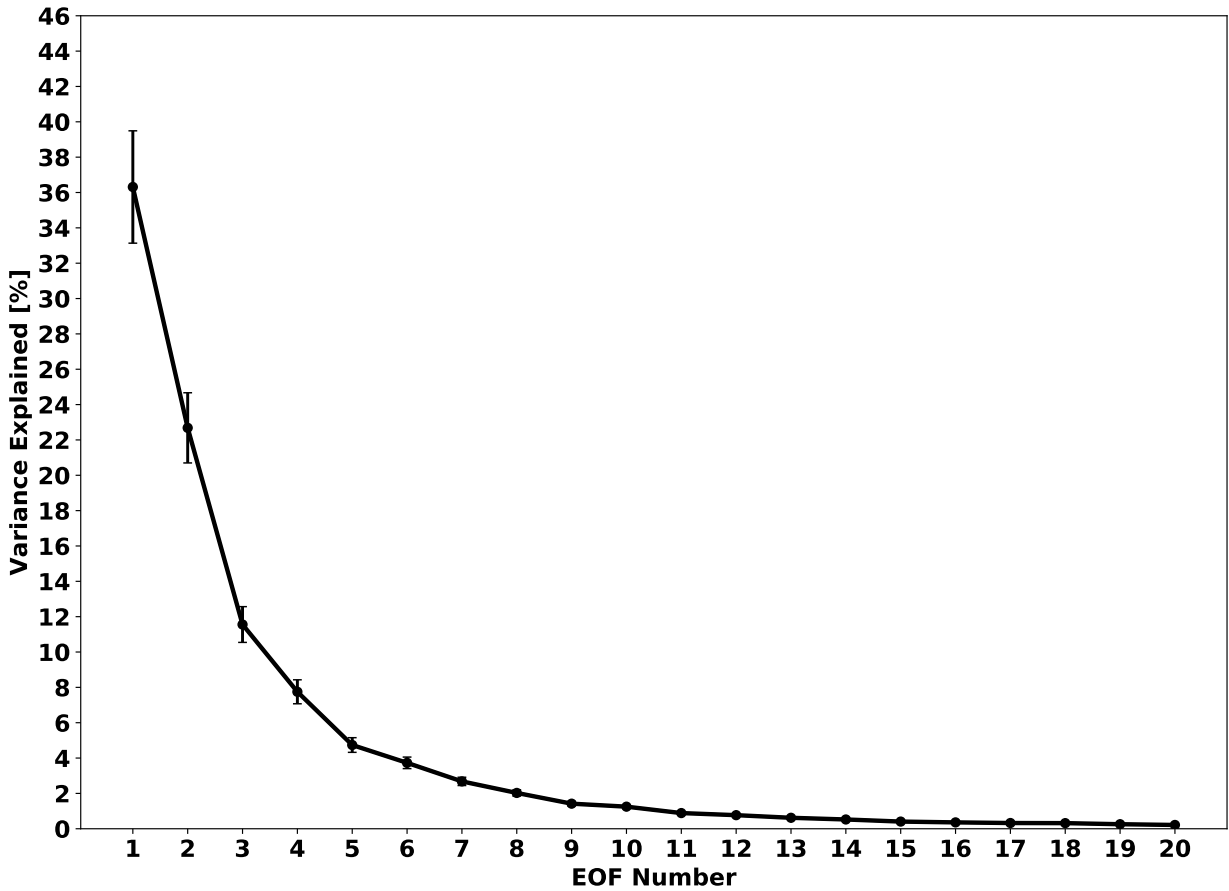
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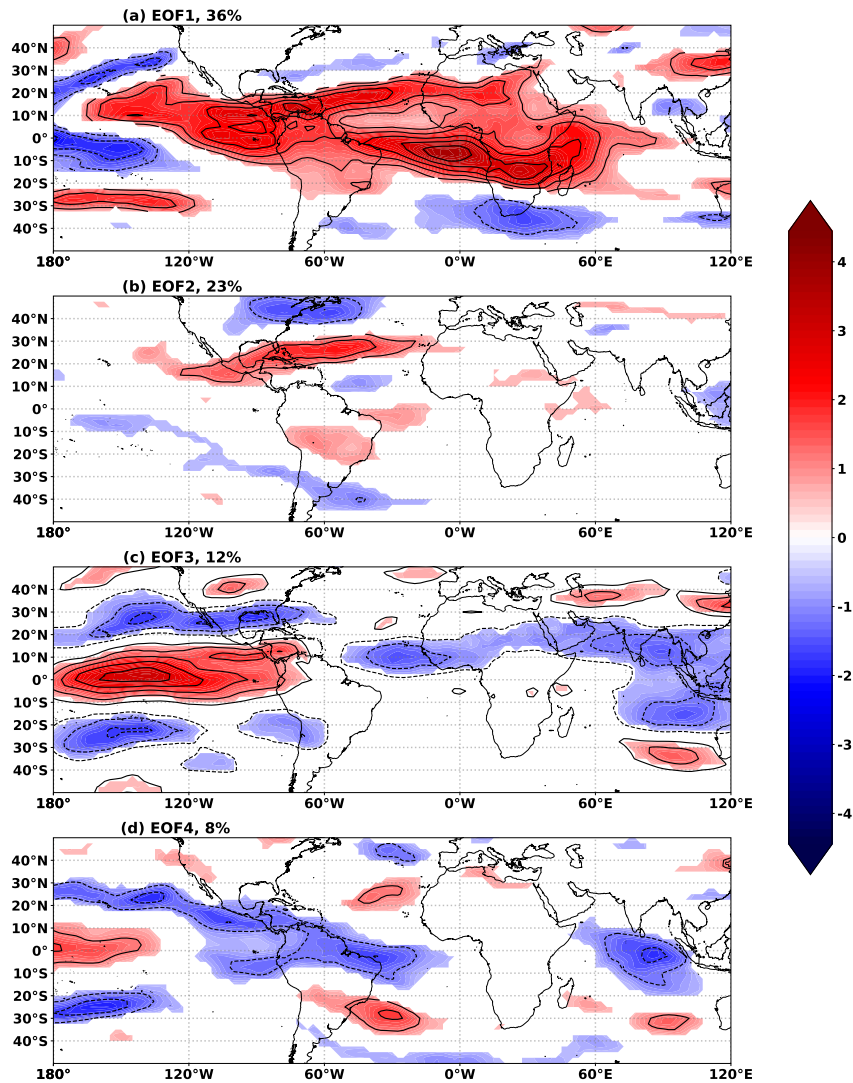
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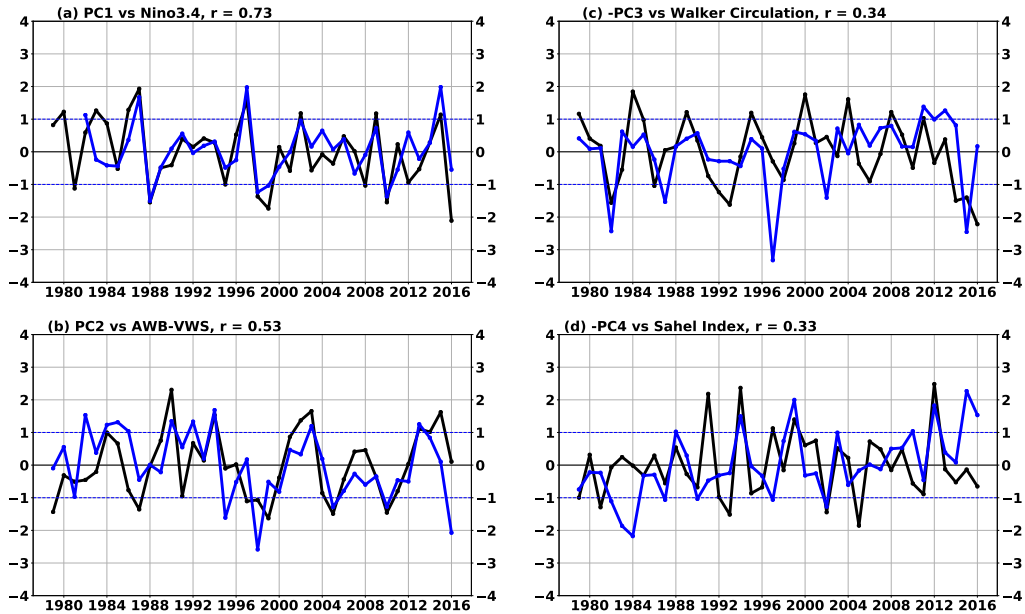
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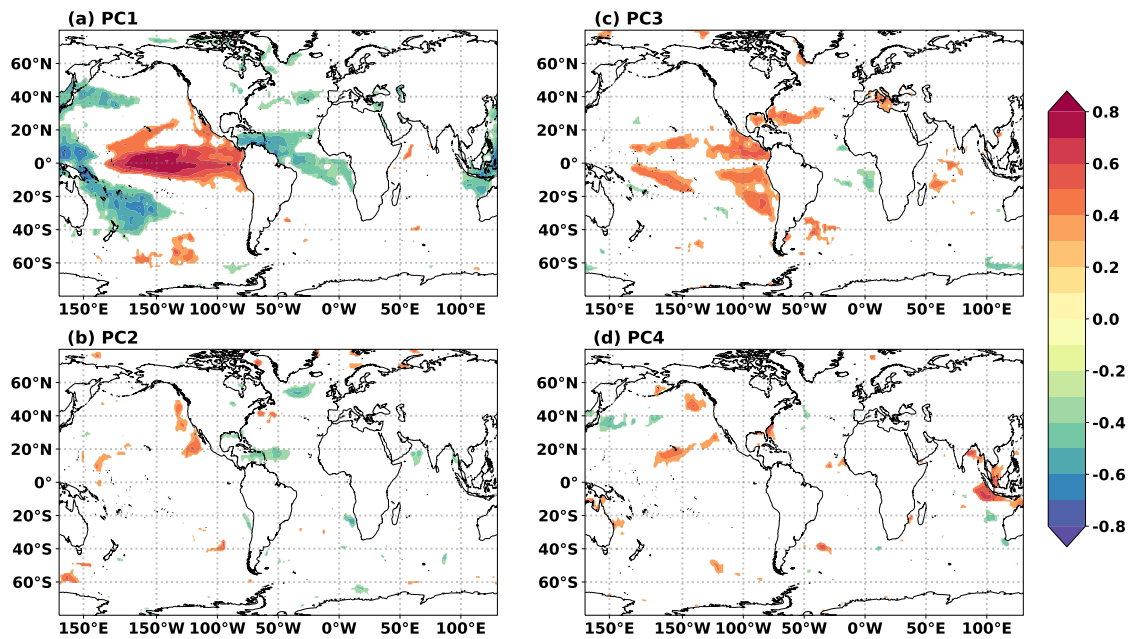
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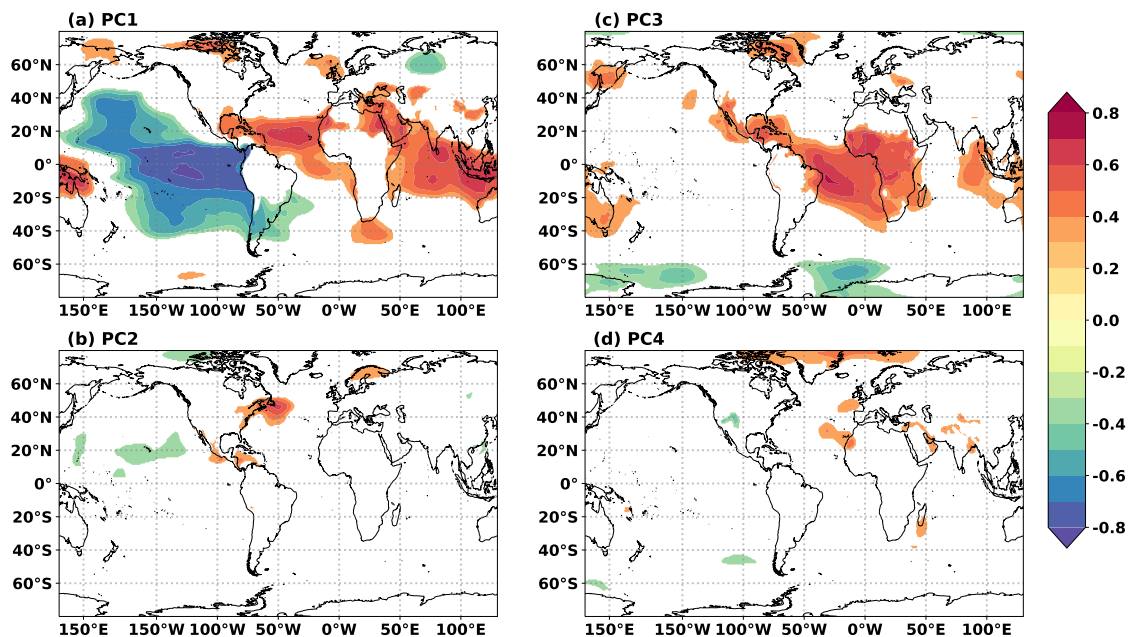
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