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1 **Simulating the impact of the large-scale circulation on the 2-m**  
2 **temperature and precipitation climatology**

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8 **Abstract** The impact of the simulated large-scale atmospheric circulation on the regional  
9 climate is examined using the Weather Research and Forecasting (WRF) model as a regional  
10 climate model. The purpose is to understand the potential need for interior grid nudging for  
11 dynamical downscaling of global climate model (GCM) output for air quality applications  
12 under a changing climate. In this study we downscale the NCEP-Department of Energy  
13 Atmospheric Model Intercomparison Project (AMIP-II) Reanalysis using three continuous  
14 20-year WRF simulations: one simulation without interior grid nudging and two using dif-  
15 ferent interior grid nudging methods. The biases in 2-m temperature and precipitation for  
16 the simulation without interior grid nudging are unreasonably large with respect to the North  
17 American Regional Reanalysis (NARR) over the eastern half of the contiguous United States  
18 (CONUS) during the summer when air quality concerns are most relevant. This study ex-  
19 amines how these differences arise from errors in predicting the large-scale atmospheric  
20 circulation. It is demonstrated that the Bermuda high, which strongly influences the regional

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21 climate for much of the eastern half of the CONUS during the summer, is poorly simu-  
22 lated without interior grid nudging. In particular, two summers when the Bermuda high was  
23 west (1993) and east (2003) of its climatological position are chosen to illustrate problems  
24 in the large-scale atmospheric circulation anomalies. For both summers, WRF without in-  
25 terior grid nudging fails to simulate the placement of the upper-level anticyclonic (1993)  
26 and cyclonic (2003) circulation anomalies. The displacement of the large-scale circulation  
27 impacts the lower atmosphere moisture transport and precipitable water, affecting the con-  
28 vective environment and precipitation. Using interior grid nudging improves the large-scale  
29 circulation aloft and moisture transport/precipitable water anomalies, thereby improving the  
30 simulated 2-m temperature and precipitation. The results demonstrate that constraining the  
31 RCM to the large-scale features in the driving fields improves the overall accuracy of the  
32 simulated regional climate, and suggest that in the absence of such a constraint, the RCM  
33 will likely misrepresent important large-scale shifts in the atmospheric circulation under a  
34 future climate.

## 35 **1 Introduction**

36 Regional climate models (RCMs) are frequently used for dynamical downscaling of future  
37 climate projections from global climate models to develop regional climate change impact  
38 assessments and for climate change adaptation planning. For downscaling applications in  
39 which the RCM is forced by the global fields only via the lateral boundary conditions, the  
40 large-scale atmospheric circulation simulated by the RCM can diverge from that in the driv-  
41 ing fields. This is particularly true for large RCM domains when the synoptic forcing is  
42 relatively weak or in the tropics where there is localized convection (Wang et al. 2004).

43 Whether this divergence between the large-scale driving fields and the RCM solution is  
44 indicative of a problem or is a desired outcome of using a regional climate model is an open  
45 question, where the answer may depend on the specific application of interest (Giorgi 2006).  
46 Under one philosophical paradigm for dynamical downscaling, the RCM should be allowed  
47 as much freedom as possible to develop its own circulation in the interior of the modeling

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48 domain because of the potential for the RCM to provide added value. For some research ap-  
49 plications, such as process studies of the feedback of local- or regional-scale forcings on the  
50 large-scale dynamics, the deviation between the driving fields and the regional-scale fields  
51 is the intended focus of the research, and constraining the RCM in such instances would be  
52 undesirable (Lorenz and Jacob 2005; Inatsu and Kimoto 2009). A further consideration is  
53 that global climate models have their own biases (e.g. Pielke et al., in press), and it is pos-  
54 sible in some situations that the RCM can improve on the atmospheric circulation present  
55 in the driving fields, even at large scales (Veljovic et al. 2010). Also, constraining the RCM  
56 could have the unintended side effect of masking the RCM model biases (Christensen et al.  
57 2007).

58 An alternate philosophical paradigm of regional climate modeling is dynamical down-  
59 scaling where the RCM should resolve the mesoscale circulations while retaining the GCM  
60 resolved scales of motion (Grotch and MacCracken 1991; Jones et al. 1995; Laprise et al.  
61 2007). Various methods have been suggested to constrain the RCM to the input data: Kida  
62 et al. (1991), Waldron et al. (1996), von Storch et al. (2000) for spectral nudging methods; Lo  
63 et al. (2008) for frequent reinitialization; Bowden et al. (2012) for analysis nudging methods;  
64 Yhang and Hong (2011) for scale-selective bias correction. Recently, Bowden et al. (2012)  
65 conducted annual simulations using the Weather Research and Forecasting (WRF) model  
66 to show that persistent biases in simulated climatology can occur over large spatial regions  
67 in the absence of interior nudging, and that application of two different nudging techniques  
68 improved the accuracy of the downscaled climatology.

69 We extend the work of Bowden et al. (2012), which used annual simulations, by conduct-  
70 ing multi-decadal hindcast regional climate model simulations using a global reanalysis as  
71 the driving input fields. Using the multi-decadal simulations allows us to address the RCM's  
72 ability to retain the climatological large-scale atmospheric circulation without interior grid  
73 nudging. The goal is to investigate if inconsistencies in regional climatology, as represented  
74 by errors in 2-m temperature and precipitation, are associated with misrepresentation of the  
75 large-scale circulation. Only a few studies have used long continuous integrations, which  
76 are needed to reduce model internal variability (Alexandru et al. 2007; Lucas-Picher et al.

77 2008) to investigate the large-scale atmospheric circulation deviations within RCMs from  
78 the driving lateral boundary conditions. Sanchez-Gomez et al. (2009) addressed the problem  
79 of simulating the large-scale circulation for Europe with an ensemble of simulations from  
80 different RCMs. Using weather regimes, recurrent and spatially defined weather patterns  
81 (order of a few days to a few weeks), they found that the RCMs reproduced the weather  
82 regimes behavior in terms of composite pattern, mean frequency of occurrence and persis-  
83 tence reasonably well, indicating that the large-scale circulation was well represented within  
84 the RCMs. On the contrary, Yhang and Hong (2011) used a 26-year continuous integration  
85 to demonstrate problems in simulating large-scale atmospheric circulation and the resulting  
86 impact on the simulated precipitation. They found that using a scale-selective bias correction  
87 helped to reduce errors in the monsoon circulation, but there was no discernible advantage  
88 of using the scale-selective bias correction for precipitation. This study helps to provide fur-  
89 ther insight into the large-scale atmospheric circulation simulated within RCMs by showing  
90 robust examples of the impact of the large-scale atmospheric circulation on simulated 2-  
91 m temperature and precipitation. Additionally, this study is the first to compare the RCM  
92 simulated atmospheric large-scale circulation using two different interior grid nudging tech-  
93 niques.

94 The rest of this paper is organized as follows. The model setup and experiment design  
95 are described in Section 2. In Section 3 we evaluate the biases in monthly and regionally  
96 averaged quantities over the simulation period and identify summertime in the Southeastern  
97 United States as a season and region that, in the absence of interior nudging, is frequently  
98 simulated poorly. In Section 4 we relate the errors in simulated summer climatology in the  
99 Southeast to the large-scale atmospheric circulation. We conclude the paper with a concise  
100 summary and future research needs.

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## 101 2 Model Description and Experiment Design

102 The WRF model (Skamarock et al. 2008) is a fully compressible, non-hydrostatic model  
103 that uses a terrain-following vertical coordinate. In this study, WRF is run using a 34-layer  
104 configuration extending to a model top at 50 hPa. A two-way interactive nest is used with  
105 horizontal grid spacings of 108 km (81 x 51 grid points) covering most of North America  
106 and 36 km (187 x 85 grid points) over the contiguous United States (CONUS), as shown  
107 in Fig. 1. We use WRF version 3.2.1 with physics options including the Community At-  
108 mospheric Model for longwave and shortwave radiation (CAM; Collins et al. (2004)), the  
109 WRF single-moment six-class microphysics scheme (Hong and Lim 2006), the Grell ensem-  
110 ble convective parameterization (Grell and Dévényi 2002), the Yonsei University planetary  
111 boundary layer (PBL) scheme (Hong et al. 2006), and the Noah land-surface model (Chen  
112 and Dudhia 2001). The simulations use time-varying sea-surface temperature, sea ice, veg-  
113 etation fraction, and albedo.

114 WRF is used to downscale  $2.5^\circ \times 2.5^\circ$  analyses from the NCEP-Department of Energy  
115 Atmospheric Model Intercomparison Project (AMIP-II) Reanalysis (Kanamitsu et al. 2002)  
116 (hereafter, R-2) for the period 1988-2007. The model is initialized at 00 UTC 02 Dec 1987,  
117 allowed to spin up for one month, and integrated continuously to 00 UTC 01 Jan 2008. In  
118 these runs, the R-2 fields serve as proxies for data from a global climate model. However,  
119 the R-2 fields represent the best available representation of the meteorology that occurred  
120 at the  $2.5^\circ$  spatial scale for that historical period, so can be regarded as “perfect boundary  
121 conditions” (Christensen et al. 1997). The R-2 fields provide initial, lateral, and surface  
122 boundary conditions, and they serve as the constraints for interior nudging used in this paper.  
123 No observational data exogenous to the R-2 fields are assimilated for any of the simulations.

124 Two methods of interior grid nudging have been implemented in WRF. Both forms  
125 of interior nudging can reduce mean errors in regional climate modeling with WRF (e.g.,  
126 Lo et al. (2008); Bowden et al. (2012)). Analysis nudging (Stauffer and Seaman 1994) is  
127 theorized to be most useful when the input data fields are not significantly coarser than the  
128 model resolution. In analysis nudging, the prognostic equations are modified by adding a

129 non-physical term proportional to the difference between the model state and a value that  
130 is interpolated in time and space from the reference analysis. Spectral nudging (von Storch  
131 et al. 2000; Miguez-Macho et al. 2004) is attractive as a scale-selective interior constraint  
132 for dynamical downscaling because it is applied only to wavelengths longer than a specified  
133 threshold. In WRF, analysis nudging can be applied toward horizontal wind components,  
134 potential temperature, and water vapor mixing ratio, while spectral nudging is available  
135 for horizontal wind components, potential temperature, and total geopotential. When either  
136 analysis or spectral nudging is used, it is applied only above the PBL to maximize WRF's  
137 freedom to respond to mesoscale forcing within the PBL.

138 Three 20-year simulations are conducted. All simulations apply the R-2 boundary con-  
139 ditions using a 5-cell-width relaxation zone (Davies 1976). The first simulation, NN, uses  
140 no interior grid nudging; the second, AN, uses analysis nudging; and the third, SN, uses  
141 spectral nudging. The nudging coefficients and wave numbers used for analysis and spectral  
142 nudging are as specified in Table 1.

### 143 **3 Evaluation of Biases in Simulated Climatology**

144 Validating the atmospheric circulation in RCMs is difficult for large domains because atmo-  
145 spheric processes have non-uniform impacts on the regional climatology within the domain.  
146 Several approaches have been used to understand the large-scale circulation within RCMs.  
147 For instance, focus could be placed on large-scale circulation mechanisms that impact the  
148 regional climatology, e.g., atmospheric rivers and flooding for the Northwest United States  
149 (Leung and Qian 2009). Other approaches are mainly statistical, such as using cluster anal-  
150 ysis to group weather patterns based on the distribution of certain atmospheric variables  
151 (Robertson and Ghil 1999; Sanchez-Gomez et al. 2009).

152 Here, the analysis is focused on surface-based meteorology that is directly linked to the  
153 atmospheric large-scale circulation. For each month in the 20-year time series, we compute  
154 monthly- and area-averaged 2-m temperature and precipitation over land grid cells for six

155 regions of the CONUS, Fig. 1. These fields are chosen because of their fundamental impor-  
156 tance for climate change impact assessments. The 36-km WRF simulations are evaluated  
157 against the 32-km North American Regional Reanalysis (NARR; Mesinger et al. 2006),  
158 where the NARR data are bilinearly interpolated to the 36-km WRF domain. The NARR  
159 2-m temperature and precipitation data have been found to compare well with observations  
160 over land within the CONUS (Mesinger et al. 2006; Nigam and Ruiz-Barradas 2006) and  
161 have been used in several previous RCM model validation studies (Lo et al. 2008; Bukovsky  
162 and Karoly 2009; Bowden et al. 2012). Additionally, we calculate the area average differ-  
163 ence between NARR and R-2 over land because important biases between NARR and R-2  
164 will impact the nudged simulations.

165 The mean biases in 2-m temperature for the 20-year period are plotted by month in Fig.  
166 2 for each of the evaluation regions. With this model configuration, an overall cool bias ex-  
167 ists for all three simulations. The annual average bias over the CONUS is  $-2$  K for the NN  
168 simulation. For the Midwest, Northeast, and Southeast the mean error in the NN simulation  
169 typically exceeds  $-3$  K. Biases of this magnitude may pose a serious limitation for climate  
170 change impact assessments because regional climate change projections may have the same  
171 magnitude of change (Giorgi 2006); however, biases may not impact the climate change sig-  
172 nal if the model biases are conserved between current and future climates. The bias between  
173 R-2 and NARR is small over regions with large errors east of the Rockies in the NN simula-  
174 tion, further justifying nudging to R-2. Considering all regions, the largest average monthly  
175 error occurs in the Northeast during August, where the average bias is  $-5.2$  K. When either  
176 interior grid nudging technique is used, the mean annual bias across the CONUS improves  
177 to  $-1$  K. With the exception of the Southwest, both AN and SN reduce the mean regional  
178 2-m temperature error. Note that this region also has a large difference between NARR and  
179 R-2 demonstrating biases in the driving data impact the RCM bias. Specifically, notable  
180 differences between NARR and R-2 are found over regions with complex terrain.

181 Regionally averaged biases in the monthly accumulated precipitation are shown in Fig.  
182 3. Averaged over the CONUS, the NN simulation has an annual wet bias of about 12 mm  
183 month<sup>-1</sup>. However, there is a strong seasonal variation to the precipitation bias. The bias

184 decreases during the summer to late fall throughout the CONUS, becoming negative for all  
185 regions except for the Northwest, which mitigates the positive bias in the annual average.  
186 The AN simulation has the smallest bias of 9 mm month<sup>-1</sup> averaged over the CONUS, while  
187 the SN simulation has the largest wet bias of 21 mm month<sup>-1</sup>. Although SN is wetter than  
188 AN, the month-to-month bias is correlated between the AN and SN simulations, exceeding  
189 0.8 for all regions with the exception of the Northwest. The high correlation suggests that  
190 the two nudging techniques are behaving similarly. The difference in the magnitude of the  
191 precipitation bias between the nudged simulations may be because the water vapor mixing  
192 ratio is nudged in AN but not in SN. A notable difference in AN and SN relative to NN is  
193 the switch in sign of the bias over the Southeast extending into the Northeast region during  
194 the summer months (JJA). The AN and SN simulations have a wet bias during the summer  
195 for the Southeast with an average of 23 mm month<sup>-1</sup> compared to a dry bias exceeding 30  
196 mm month<sup>-1</sup> for the NN simulation. There is also a switch in sign of the bias for AN and  
197 SN during July and August compared to NN with a large positive precipitation bias for both  
198 AN and SN during the summer.

199 Next, for each region in the NN simulation we identify the months with absolute errors  
200 in the top 10% for the 20-year period (i.e., the 24 highest monthly errors) by boreal season  
201 (Figs. 4 and 5). The largest errors in 2-m temperature in NN most frequently occur during  
202 the summer in five of the six regions (Fig. 4). The incidences of large errors in 2-m temper-  
203 ature are greatest in the Northeast and Southeast (75% and 66%, respectively) which are the  
204 regions that are farthest from the inflow boundary.

205 Fig. 5 is similar to Fig. 4 but for precipitation, and it illustrates that the season during  
206 which the largest NN precipitation errors occur varies widely across the CONUS. For in-  
207 stance the Northwest region has 14 of the 24 (58%) largest biased months occurring during  
208 the boreal winter, while in the Southeast a plurality of the largest errors occur during the  
209 summer. The winter bias in the Northwest is clearly related to the seasonal cycle and when  
210 the majority of precipitation occurs. However, the Southeast has a more even climatological  
211 distribution of rainfall throughout the year.

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212 These results for the temperature and precipitation biases provide motivation to under-  
213 standing the extent to which errors in WRF are related to errors in simulating the atmo-  
214 spheric circulation over the eastern half of the U.S., in particular the Southeast, during the  
215 summer.

#### 216 **4 Evaluation of Atmospheric Circulation Errors**

217 Our focus for this dynamical downscaling research is assessing the impact of regional cli-  
218 mate change on air quality in the United States. Substantial errors in the NN simulation  
219 during the summer in the Southeast (when air quality is most problematic) present a signifi-  
220 cant problem and may adversely impact the reliability of the RCM output for the air quality  
221 application. The regional climate variability during the summertime over the Southeast is  
222 associated with several factors, including hurricanes (Liu and Fearn 2000), soil moisture  
223 (Koster et al. 2004), and atmospheric circulation anomalies associated with changes in sea  
224 surface temperatures (Wang and Enfield 2001; Seager et al. 2003). In particular, the at-  
225 mospheric circulation related to the position of the Bermuda high has a major impact on  
226 the regional climate and air quality for the Southeast. The position of the Bermuda high has  
227 shifted westward and become more intense in recent decades and is projected to shift further  
228 west and become more intense by GCMs as the climate warms (Li et al. 2010). RCMs that  
229 are unable to simulate the position and intensity of the Bermuda high under current climate  
230 are unlikely to properly simulate climate change impacts, such as for future air quality.

##### 231 **4.1 Bermuda High Index**

The location and intensity of the Bermuda high during the contemporary climate are exam-  
ined for the RCM simulations using the Bermuda High Index (BHI; Katz et al., 2003). The  
BHI measures the western extent of the Bermuda high by using the climatologically nor-  
malized difference in boreal summer (JJA) sea-level pressure between Bermuda and New

Orleans (Katz et al. 2003). Because Bermuda is located close to our lateral boundary, we adopt a modified approach using area averages for both regions. “Bermuda” is the region between 67°- 65°W and 32°- 34°N and “New Orleans” is the region between 91°- 90°W and 29°- 31°, Fig. 1. Positive and negative BHI values indicate that the Bermuda high is further east and west than normal, respectively. To calculate the BHI, the monthly sea-level pressure is first normalized at all grid points:

$$SLPnorm_{xy}(mon) = \frac{SLP_{xy} - \overline{SLP_{xy}}}{\sigma_{SLP_{xy}}} \quad (1)$$

The normalized monthly values are then averaged over JJA for the regions and subtracted to give the BHI:

$$BHI = \frac{1}{np} \sum_{i=1}^{np} \left[ \frac{1}{3} \sum_{t=1}^3 SLPnorm_{xy}(t) \right]_{be} - \frac{1}{np} \sum_{i=1}^{np} \left[ \frac{1}{3} \sum_{t=1}^3 SLPnorm_{xy}(t) \right]_{no} \quad (2)$$

232 where np is the number of grid points for Bermuda (be) and New Orleans (no), and the  
233 average is taken over three summer months.

234 The BHI is calculated for the -R2, NARR and the WRF simulations. In this analysis, the  
235 BHI quantifies WRF’s ability to properly simulate the Bermuda high intensity and location  
236 without and with interior nudging. In addition, the BHI is used to identify years when the  
237 Bermuda high is poorly simulated without interior grid nudging to understand how the errors  
238 in the large-scale circulation are related to errors in regional climate anomalies.

239 In Fig. 6, the BHI from the NARR, R-2, and the WRF simulations are compared to ex-  
240 amine WRF’s ability to capture the interannual variability in the intensity and position of the  
241 Bermuda high during the contemporary climate. The BHI correlation between NARR (R-2)  
242 and NN is 0.12 (0.11), while the correlation drastically improves to 0.98 (0.82) for both AN  
243 and SN, respectively. The poor correlation between the NARR data and NN suggests a de-  
244 ficiency in capturing the large-scale circulation, and it raises some questions. How does the  
245 misrepresentation of the Bermuda high impact the regional climate anomalies of interest to  
246 many end-user applications? How is the large-scale circulation different from the observa-  
247 tions when no nudging is used? Can interior grid nudging adjust the anomalous placement

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248 of the Bermuda high and the associated regional climate anomalies? To begin answering  
249 these questions, we use the BHI to identify two summers from the 20-year period when the  
250 Bermuda high was west/east of its climatological average position and poorly simulated in  
251 the NN simulation. Fig. 6 indicates that the most anomalous positions of the Bermuda high  
252 during this 20-year period are 1993 (west) and 2003 (east), which are both poorly repre-  
253 sented in the NN simulation. Below we discuss the temperature and precipitation anomalies  
254 and the corresponding large-scale atmospheric circulation from all simulations and observa-  
255 tions for 1993 and 2003. The anomalies for each model are relative to the 20-year average  
256 values (i.e., climatology) during the summer for that model.

#### 257 4.2 Temperature Anomalies

258 We first explore the impact of the placement of the Bermuda high on the regional cli-  
259 mate anomalies for 2-m temperature. During 1993 the observed BHI is negative (based  
260 on NARR), indicating a westward shift in the Bermuda high. This westward shift, centered  
261 closer to the eastern United States, favors warm anomalies for the Southeast, as shown in  
262 Fig. 7a. In the 1993 JJA observations, there is a corridor of warm anomalies ( $> 0.7$  K) ex-  
263 tending from Texas northeast into West Virginia. The warm anomalies over the Southeast  
264 are surrounded by  $-1.5$  K cool anomalies in the Midwest and northern Plains regions and as  
265 large as  $-1.0$  K over the Atlantic Ocean. An important signature in the temperature anoma-  
266 lies is their wavelength as indicated by their change in sign, which is on the order of 1000  
267 km. A wavelength with this magnitude indicates a shift in the synoptic scale atmospheric  
268 circulation. By contrast, in 2003 the BHI is positive, which indicates the center of the high  
269 is shifted east of its climatological average position. In 2003 the temperature anomalies are  
270 negative for most of the Southeast and the Midwest (Fig. 7e). The cool anomalies have ap-  
271 proximately the same magnitude as the warm anomalies in 1993,  $-0.7$  K. As in 1993, there  
272 is a signature of a shift in the synoptic circulation with warm anomalies to the west and east  
273 of the cool anomalies. Capturing the temperature anomalies in the eastern U.S. during 1993

274 and 2003 could indicate the model's overall ability to simulate the large-scale atmospheric  
275 circulation.

276 For JJA 1993 all three model runs correctly simulate a warm anomaly over much of the  
277 eastern half of the U.S., but the placement and magnitude of the anomalies differs between  
278 the simulations (Figs. 7b-7d). Tables 2 and 3 illustrate that both AN and SN improve the  
279 RMSE and pattern correlation over NN for the 1993 temperature anomalies. In particular,  
280 the temperature anomalies for 1993 in NN cover a much larger area that is centered much  
281 farther west towards southern Missouri and Illinois than in NARR, and they are warmer than  
282 observations by more than 1 K in some locations. Additionally, the temperature anomalies  
283 are of the opposite sign in some areas. Despite the disagreement in placement and sign, the  
284 warm anomalies are surrounded by anomalies of the opposite sign, as in the observations.  
285 The AN temperature anomalies are in better agreement with the observations than NN, but  
286 the warmest anomalies are in central Tennessee, west of the observations. However, the tran-  
287 sition from warm to cool anomalies, such as in central Missouri and the western Atlantic, is  
288 well simulated by AN. That transition is also well simulated in SN, but the magnitude of the  
289 warm anomalies is much larger than observed and even further west into northern Texas and  
290 Oklahoma for the SN simulation. Overall, the AN and SN 2-m temperature anomalies, and  
291 their gradients, suggest that the large-scale atmospheric circulation shift is well captured,  
292 but local processes are simulated differently between the two types of nudging techniques.

293 For 2003 and as in 1993, all three simulations correctly predict the sign of the anomaly  
294 over the eastern half of the U.S., but the placement and magnitude of the anomalies differ  
295 greatly (Figs. 7f-h). In the NN simulation, the strongest cool anomalies are farther north  
296 towards the Great Lakes and are much cooler, 0.5 K cooler than observations. The pattern  
297 correlation (Table 3) is only 0.27, indicating problems in simulating the placement of the  
298 temperature anomalies, while the RMSE of 1.1 K (Table 2) indicates problems in simulat-  
299 ing the magnitude of the anomalies. NN indicates a shift in the synoptic circulation, with  
300 warm anomalies surrounding the cooler anomalies over the eastern U.S.; however, there are  
301 large areas with differences in sign of the anomalies, such as the northern Midwest and cen-  
302 tral/northern Plains regions. In both AN and SN, the placement and magnitude of the cool

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303 anomalies is generally well simulated, with pattern correlation increasing to 0.74 and a de-  
304 crease in the RMSE by as much as 0.6 K. The warm anomalies over the Atlantic are also  
305 well captured in AN and SN, but the warm anomalies over the Plains are largely absent in  
306 both nudging cases. The absence of these warm anomalies, which cover a large area, may  
307 reflect reduced accuracy of the large-scale atmospheric circulation simulated by both AN  
308 and SN.

#### 309 4.3 Precipitation Anomalies

310 The observed negative precipitation anomalies over the Southeast for 1993 (Fig. 8a) are in-  
311 tuitively consistent with a westward shift in the Bermuda high towards the CONUS. The  
312 dry conditions extend from Texas across the Southeast and into the Northeast. The largest  
313 negative precipitation anomalies are centered over northern Georgia and western North Car-  
314 olina and coincide with some of the largest positive 2-m temperature anomalies. Consistent  
315 with the temperature field and a shift in the large-scale atmospheric circulation, there is a  
316 change in the sign of the precipitation anomalies towards the Midwest and northern Plains  
317 region. The year 1993 is well known for the devastating flooding that occurred over this  
318 region, as suggested by anomalies  $> 100 \text{ mm month}^{-1}$  (Fig. 8a). Trenberth and Guillemot  
319 (1996) discuss some of the large-scale circulation processes involved during the 1993 Mid-  
320 west flood, including a southward shift in the jet stream and strong moisture transport from  
321 the Gulf of Mexico. RCMs have been used to investigate the processes related to the 1993  
322 flood (Pal and Eltahir 2002) and as a benchmark for model performance (Anderson et al.  
323 2003). These studies have shown soil moisture and the timing of precipitation associated  
324 with mesoscale convective systems were important in simulating the 1993 flood. However,  
325 the RCM must also accurately simulate the large-scale circulation, which is responsible for  
326 moisture flux into this region. Accordingly, the summer of 1993 is ideal for relating prob-  
327 lems in the simulated temperature and precipitation anomalies to the large-scale circulation  
328 anomalies.

329 In 2003, with the Bermuda high east of its climatological position, positive precipitation  
330 anomalies are evident over the Southeast (Fig. 8e). The wet conditions extend as far north as  
331 Pennsylvania, with the largest positive precipitation anomalies concentrated along the Gulf  
332 Coast. There is also a dry bias in the central and northern Plains region. The 1993 and 2003  
333 precipitation anomalies are of the opposite sign, as with temperature, indicating a shift in  
334 the synoptic-scale atmospheric circulation. An exception is over the ocean where the NARR  
335 precipitation anomalies are of the same sign between 1993 and 2003, but the confidence in  
336 NARR precipitation is low over the ocean because there are few observations available for  
337 assimilation.

338 The 1993 precipitation anomalies for the NN, AN, and SN simulations are shown in Fig.  
339 8b, 8c, and 8d, respectively. Though NN indicates that summer 1993 is drier than average  
340 for the Southeast, the magnitude and extent of the Southeast drought are not captured. Fur-  
341 thermore, the precipitation anomalies in Texas, Florida, and Georgia have the wrong sign.  
342 Finally, the rainfall responsible for the Midwest flooding is poorly simulated, with the posi-  
343 tive precipitation anomalies in NN located in Minnesota and South Dakota, several hundred  
344 kilometers to the northwest of the observed location. The AN simulation improves the sig-  
345 nal of dry conditions relative to NN, see Tables 2 and 3, but the Southeast drought is more  
346 intense than observed and is located to the south and east of its observed position. The pre-  
347 cipitation anomalies associated with the Midwest flooding are well captured in AN, with  
348 the magnitude and location of the largest positive precipitation anomalies similar to obser-  
349 vations. In SN, the Southeast drought is more intense than observed for most locations,  
350 with the largest negative anomalies centered along the Gulf Coast. The 1993 flooding for  
351 the Midwest is also captured in SN, but the westward extent of the positive precipitation  
352 anomalies (towards Nebraska) is absent, and instead there is an eastward extension of the  
353 anomalies to Indiana and Ohio. Previous studies have demonstrated the importance of local  
354 processes such as evaporation and moisture flux (Pal and Eltahir, 2002), and perhaps the  
355 large-scale circulation contribute to these differences.

356 Precipitation anomalies for NN, AN, and SN during 2003 are shown in Fig. 8f, 8g, and  
357 8h, respectively. The NN 2003 precipitation anomaly is poorly simulated throughout the

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358 eastern half of the CONUS. Also, the precipitation anomaly for much of the Southeast is  
359 of opposite sign from the observations, and there are positive anomalies in the Midwest  
360 extending across central Illinois that are not observed. Both interior nudging techniques sig-  
361 nificantly improve the ability to simulate the precipitation anomalies including the RMSE  
362 and pattern correlation during 2003 (Tables 2 and 3). AN captures the wet conditions across  
363 the Southeast, but locally the precipitation anomalies are  $40 \text{ mm month}^{-1}$  larger than ob-  
364 served. However, the placement of the maximum precipitation anomalies along the Gulf  
365 Coast is well simulated, and AN also captures the transition to drier conditions into the  
366 Midwest and Plains regions. The SN simulation also captures wetter conditions, especially  
367 along the Gulf Coast, but the positive precipitation anomalies are stronger than observed  
368 along the eastern seaboard, from North Carolina to Connecticut. The SN simulation also  
369 does not capture the gradient to drier anomalies as well as AN does, with breaks in the nega-  
370 tive anomalies across Iowa and Nebraska that are absent from the observations. Overall, for  
371 1993 and 2003 the precipitation anomalies are best captured by AN, though SN provides a  
372 notable improvement over NN.

#### 373 4.4 Large-scale Atmospheric Circulation Anomalies

374 The large-scale atmospheric circulation associated with the Bermuda high consists of east-  
375 erly flow over the Caribbean and a southerly jet along the eastern flanks of the Sierra Madre  
376 Oriental range. This large-scale flow favors strong moisture transport into the eastern half  
377 of the CONUS during the summer and is well represented in both the R-2 and NARR data  
378 (Nigam and Ruiz-Barradas 2006). Anomalous placement of the Bermuda high adversely af-  
379 fects the large-scale atmospheric circulation and the corresponding regional climate anoma-  
380 lies. In this section, we examine the role of the large-scale atmospheric circulation with re-  
381 spect to the model simulated temperature and precipitation anomalies previously discussed.  
382 We investigate the large-scale circulation using the 500-hPa wind vector anomalies, 850-hPa  
383 moisture transport, and precipitable water anomalies for both the 1993 and 2003 summer  
384 seasons.

385 The summer was anomalously warm and dry in the Southeast during 1993, with cool  
386 anomalies to the west and east as shown earlier. The corresponding observed 500-hPa wind  
387 vector anomaly is shown in Fig. 9a. As anticipated from the temperature anomalies, there is  
388 a clear shift in the large-scale atmospheric circulation, with an anomalous anticyclonic cir-  
389 culation centered over northern Alabama and Mississippi. This anomalous anticyclonic cir-  
390 culation favors subsidence over the Southeast, consistent with the warm and dry anomalies  
391 and with a westward shift in the Bermuda high. Over the northern Atlantic and in northern  
392 Plains is a large anomalous cyclonic circulation consistent with the cold anomalies over the  
393 same regions. These large-scale atmospheric circulation anomalies are reversed in 2003 (Fig.  
394 9e), with an anomalous cyclonic circulation centered over northern Kentucky and southern  
395 Indiana and Ohio, consistent with the cooler and wetter conditions over this region and  
396 extending into the Southeast. An anomalous anticyclonic circulation is found off-shore cen-  
397 tered near the warmer anomalies. The reversal in the large-scale atmospheric circulation,  
398 temperature, and precipitation anomalies for the eastern half of the CONUS for 1993 and  
399 2003 can be used to understand the simulated large-scale atmospheric circulation anomalies  
400 and the potential need for interior nudging toward the driving fields.

401 The NN, AN, and SN 500-hPa wind vector anomalies for summer 1993 are shown in  
402 Fig. 9b, 9c, and 9d, respectively. All three simulations produce an anomalous anticyclonic  
403 circulation, but in NN it is centered over Kentucky, approximately 500 km to the northeast  
404 of the observations. This displacement of the large-scale atmospheric circulation by NN  
405 causes large errors in the regional climate anomalies, as the warmest temperature anomalies  
406 are located to the north of the observations. AN and SN simulate the anomalous anticyclone  
407 close to its observed location compared to NN and with similar strength, and accordingly  
408 better simulate the temperature anomalies. The RMSE of the wind speed anomalies (Table  
409 2) is reduced from  $0.7 \text{ ms}^{-1}$  to  $0.3 \text{ ms}^{-1}$  and pattern correlations (Table 3) increase from  
410 0.88 in NN to 0.98 for both AN and SN. Similar conclusions can be drawn from the 2003  
411 simulation. The 500-hPa wind anomalies for summer 2003, (Fig. 9f, 9g, and 9h) all depict  
412 an anomalous cyclone over the eastern half of the U.S. in agreement with the observations  
413 (Fig. 9e), but the location of the anomalous cyclone in NN was twice as strong as observed

414 and was centered approximately 500 km northeast of where it occurred in the observations.  
415 This is consistent with large errors in both the wind speed and direction for 2003 as the wind  
416 speed and wind direction errors are largest during 2003 for the NN simulation (see Tables 2  
417 and 3). The incorrect placement and strength of the cyclonic anomaly in NN leads to large  
418 errors in the regional placement of temperature anomalies for much of the eastern half of the  
419 U.S., with anomalies too cold in the Great Lakes area to the north and not cold enough over  
420 the Southeast. The AN and SN simulations improve the representation of the anomalous  
421 cyclone location and strength in 2003, with significant improvements in the wind speed and  
422 direction as seen in Tables 2 and 3, and consequently improve the simulated temperature  
423 anomalies. An exception is over the Great Lakes, where the temperature anomalies are larger  
424 than surrounding land areas. The improvement in the large-scale atmospheric circulation and  
425 the resulting impact on the regional climate anomalies with nudging complements previous  
426 studies that used shorter simulations (Castro et al. 2005; Miguez-Macho et al. 2005; Bowden  
427 et al. 2012). The results show that the choice of nudging technique is less important than the  
428 decision to use interior nudging.

429 To provide further insight into the precipitation anomalies, the 850-hPa moisture trans-  
430 port and precipitable water anomalies are shown for the summer of 1993 and 2003 (Fig.  
431 10). The observed precipitable water anomalies are as much as  $5 \text{ mm day}^{-1}$  for JJA over  
432 parts of the Midwest during 1993, and a large component of this moisture is due to trans-  
433 port from the Gulf of Mexico (Fig. 10a). Accurately modeling the anticyclonic anomaly  
434 over the Southeast must be complemented with correctly simulating the precipitable water  
435 anomalies in order to capture the observed precipitation anomalies. The precipitable water  
436 anomalies are positive over the western portions of the Southeast (Arkansas, Mississippi,  
437 and Alabama) and decrease toward the east (North Carolina, South Carolina and Georgia),  
438 which is consistent with the larger negative precipitation anomalies simulated in the east  
439 (Fig. 8a). The precipitation anomalies in 2003 result from a significantly different atmo-  
440 spheric circulation and provide additional evidence of the necessity for interior nudging  
441 to capture anomalies in both circulation and precipitation. During 2003 there is a stronger  
442 moisture flux component from the Gulf of Mexico for the Southeast (Fig. 10e), which gen-

443 erates positive precipitable water anomalies. This increase in moisture, in conjunction with  
444 an anomalous cyclonic circulation, contributes to the observed positive precipitation anoma-  
445 lies for the Southeast. The gradient in the precipitable water anomalies, from positive over  
446 the Southeast to negative over the Midwest and Plains, is consistent with the positive and  
447 negative precipitation anomalies for those respective regions.

448 The 1993 precipitable water anomalies and 850-hPa moisture transport for the NN, AN,  
449 and SN simulations are shown in Fig. 10b, 10c, and 10d, respectively. The NN anomalous  
450 low-level jet is consistent with observations except that the origin of the jet over the Gulf of  
451 Mexico has a stronger easterly component in NN. The difference in the moisture transport  
452 over the Gulf of Mexico in NN is a consequence of improperly simulating the large-scale  
453 anticyclonic anomaly over the Atlantic Ocean. The difference in the low-level circulation  
454 between NN and NARR results in maximizing moisture transport and convergence within  
455 the Southeast (northern Arkansas) in NN instead of in the Midwest. The observations in-  
456 dicate that the maximum precipitation anomaly coincides with the maximum precipitable  
457 water anomaly, but the maximum precipitation anomaly in NN is located much farther north  
458 than in the observations (Fig. 8). The AN and SN simulations improve the simulated east-  
459 erly component of the moisture transport associated with the low-level jet compared with  
460 NN during 1993. Improvements in the moisture transport lead to a concentration of moisture  
461 over the Midwest for both simulations that used interior nudging. AN provides a better esti-  
462 mate of the magnitude of the precipitable water anomalies and their placement than SN, with  
463 higher precipitable water amounts extending towards the Gulf Coast. That extension of the  
464 positive precipitable water anomalies explains differences between AN and SN precipitation  
465 anomalies, as SN is much drier along the Gulf Coast.

466 The simulated precipitable water anomalies and 850-hPa moisture transport for the NN,  
467 AN, and SN simulations during summer 2003 are shown in Fig. 10f, 10g, and 10h, respec-  
468 tively. The 2003 NN simulation does not capture the moisture flux anomaly from the Gulf  
469 of Mexico into the Southeast. This can be partly explained by the momentum transfer of  
470 the stronger upper-level cyclonic circulation, as seen at 500-hPa (Fig. 9f), to lower levels  
471 of the atmosphere favoring a more northerly wind component at lower levels. The northerly

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472 component cuts off moisture from the Gulf of Mexico, which is consistent with negative  
473 precipitable water and precipitation anomalies over the Southeast, which are opposite from  
474 the observed anomalies, reducing the pattern correlation and increasing the RMSE. Here,  
475 again, errors in modeling the anomalous large-scale circulation in NN adversely impact the  
476 regional climate anomalies. AN and SN both successfully simulate the low-level moisture  
477 transport from the Gulf Coast up the eastern seaboard, but the positive precipitable water  
478 anomalies in AN agree better with observations. Tables 2 and 3 illustrate that the precip-  
479 itable water anomalies are more than double in NN ( $1.7 \text{ mm day}^{-1}$ ) compared to AN ( $0.7$   
480  $\text{mm day}^{-1}$ ) with a pattern correlation increasing from 0.14 in NN to 0.87 in AN. Addition-  
481 ally, improvements in the precipitable water anomalies for AN compared to SN, as shown  
482 in Fig. 10 and Tables 2 and 3, suggests that nudging the moisture field may improve the  
483 accuracy of the simulated regional climate.

## 484 **5 Summary**

485 We examined the large-scale circulation in three continuous 20-year WRF simulations, one  
486 without interior grid nudging and two using different interior grid nudging methods. Ex-  
487 amining the large-scale circulation was motivated by our application of WRF to downscale  
488 GCM output to examine the impacts of air quality under a changing climate. Without in-  
489 terior grid nudging, WRF may be inadequate to simulate the placement of the resolved  
490 large-scale circulation as represented by the GCM. In particular, the bias in 2-m temperature  
491 and precipitation is typically larger during the summer when air quality concerns related to  
492 ozone are important. We investigated whether errors in predicting the large-scale circulation  
493 strongly contributed to the large summer bias at the surface. The Bermuda high was identi-  
494 fied as a large-scale circulation feature of interest because of its control on regional climate  
495 anomalies over the Southeast during the summer, its potential impact on air quality, and the  
496 observed/projected westward shift in the Bermuda high as the climate warms. This study  
497 illustrates problems that can arise in the large-scale circulation with weak constraint toward  
498 the driving fields.

499 The Bermuda high during the summer was first examined using the BHI to measure  
500 the intensity and anomalous placement of the Bermuda high. We found that the interannual  
501 variability in the intensity and placement of the Bermuda high is poorly simulated when no  
502 interior grid nudging is used. Both types of nudging drastically improved the representation  
503 of the BHI, which indicates that the large-scale circulation had been improved. Using the  
504 BHI, we identified two summers, 1993 and 2003, when the Bermuda high was anomalously  
505 west and east of its climatological position. For these events we examined the impact on re-  
506 gional climate anomalies of 2-m temperature and precipitation with respect to the large-scale  
507 circulation. The NN 500-hPa wind vector anomalies for both summers indicate problems in  
508 simulating the proper placement of the large-scale atmospheric circulation anomalies. In  
509 2003, there is an additional problem for NN as the anomalous circulation aloft is too strong,  
510 which may be transferring momentum to the lower atmosphere. This impacts the lower  
511 atmosphere by reducing the moisture transport and precipitable water affecting the convec-  
512 tive environment and precipitation. Both interior grid nudging strategies greatly improve the  
513 representation of the large-scale circulation aloft and moisture transport/precipitable water  
514 anomalies helping to improve the sign and spatial distribution of the simulated 2-m tempera-  
515 ture and precipitation anomalies. The results illustrate that weakly constraining the RCM to  
516 downscale GCM projections (as in NN) will likely misrepresent important large-scale shifts  
517 in the atmospheric circulation with respect to the Bermuda high and provide an unrealistic  
518 conceptual view of the regional climate change. Allowing the RCM large-scale circulation  
519 to deviate from the GCM should be avoided when faced with problems of modeling the  
520 large-scale circulation in the contemporary climate.

521 Although both nudging strategies result in improved simulation of large-scale circula-  
522 tion, there are differences in the regional climate anomalies for 2-m temperature and precipi-  
523 tation between the two nudging strategies. The differences in 2-m temperature and precipita-  
524 tion between AN and SN are generally local. The similarities in the large-scale environment  
525 indicate that local processes such as evaporation or cloud cover or embedded model bi-  
526 ases from the LSM or PBL physics schemes likely contribute to these differences. We are  
527 currently further investigating the role of local processes with particular interest in the im-

528 pact of nudging towards moisture. In addition, using the same modeling period as was used  
 529 here, Otte et al. (2012) showed that nudging improved the prediction of extremes. Overall,  
 530 these results suggest that more research is needed to further understand the impact of in-  
 531 terior grid nudging for mesoscale and local processes that are associated with added value  
 532 within RCMs. Regardless, using an interior constraint toward the driving model (such as  
 533 with nudging) is recommended to correctly simulate the large-scale circulation in the RCM.

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**Table 1** Nudging coefficients ( $s^{-1}$ ) and domain-relative wave numbers used for analysis and spectral nudging simulations. Time scales (h) that correspond to the nudging coefficients and length scales (km) that correspond to the wave numbers are in parentheses. Fields that are not applicable are indicated by -.

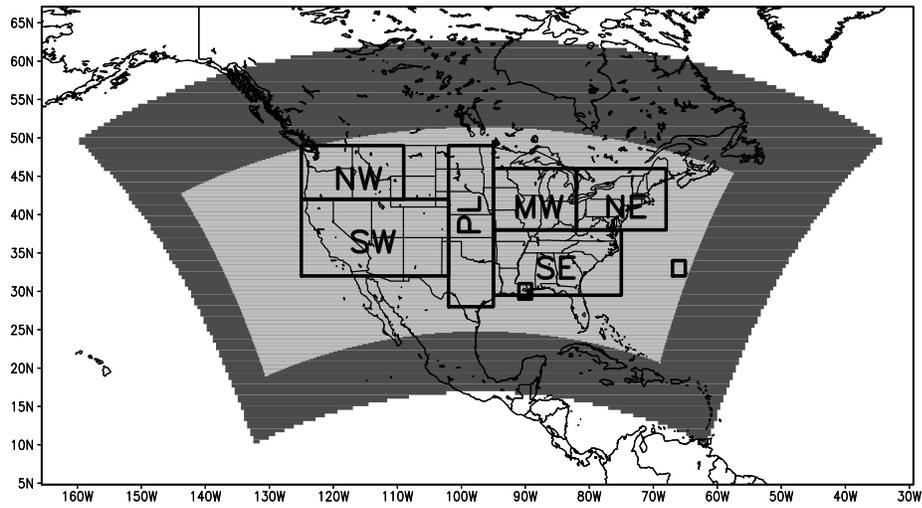
	Wind	Potential Temp.	Water Vapor Mixing Ratio	Geopotential	West-east wave number	South-north wave number
Analysis Nudging (108-km)	$3.0 \times 10^{-4}$ (0.9)	$3.0 \times 10^{-4}$ (0.9)	$4.5 \times 10^{-5}$ (0.9)	-	-	-
Analysis Nudging (36-km)	$1.0 \times 10^{-4}$ (2.8)	$1.0 \times 10^{-4}$ (2.8)	$1.0 \times 10^{-5}$ (27.8)	-	-	-
Spectral Nudging (108-km)	$3.0 \times 10^{-4}$ (0.9)	$3.0 \times 10^{-4}$ (0.9)	- -	$3.0 \times 10^{-4}$ (0.9)	5 (1728)	3 (1800)
Spectral Nudging (36-km)	$3.0 \times 10^{-4}$ (0.9)	$3.0 \times 10^{-4}$ (0.9)	- -	$3.0 \times 10^{-4}$ (0.9)	4 (1674)	2 (1512)

**Table 2** RMSE for 1993 and 2003 between NARR and WRF anomalies for 2-m temperature, precipitation, precipitable water, wind speed, and wind direction.

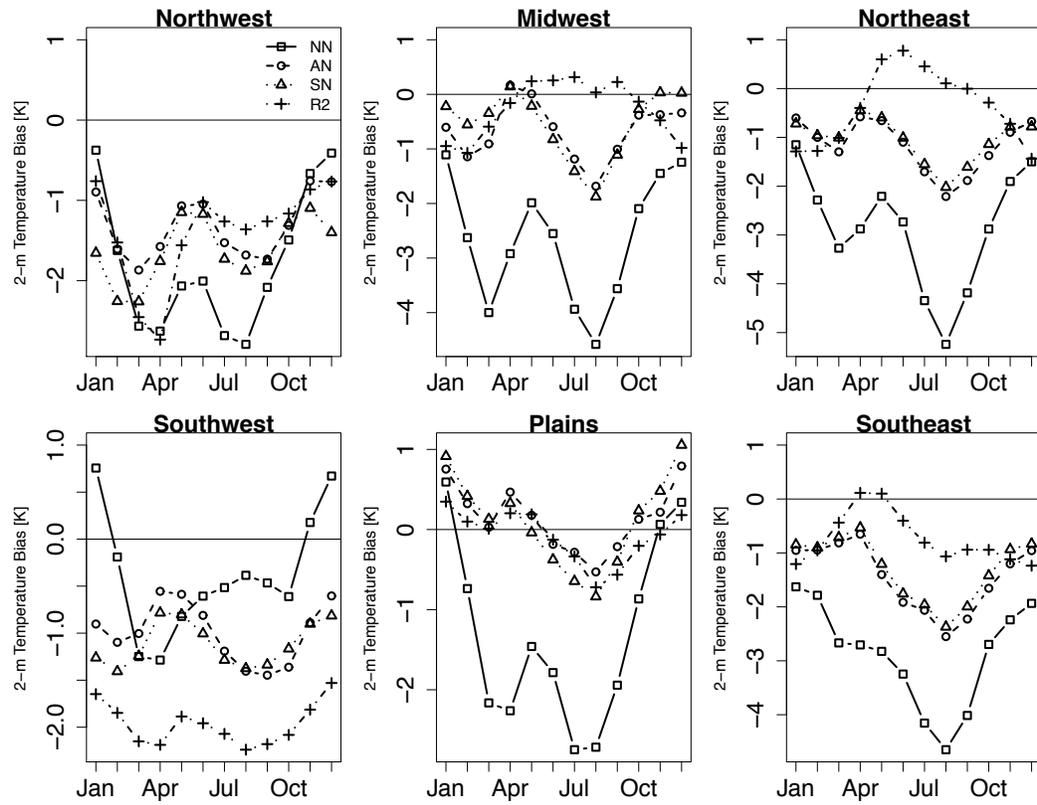
RMSE															
	T2 (K)			Pre (mm/day)			PWAT (mm/day)			Wspd (m/s)			Wdir (deg.)		
	NN	AN	SN	NN	AN	SN	NN	AN	SN	NN	AN	SN	NN	AN	SN
1993	0.7	0.3	0.5	1.4	1.0	1.2	0.8	0.5	0.7	0.7	0.3	0.3	90	39	34
2003	1.1	0.5	0.6	1.5	0.9	1.2	1.7	0.7	1.2	2.0	0.2	0.3	170	48	35

**Table 3** Pattern Correlation for 1993 and 2003 between NARR and WRF anomalies for 2-m temperature, precipitation, precipitable water, wind speed, and wind direction.

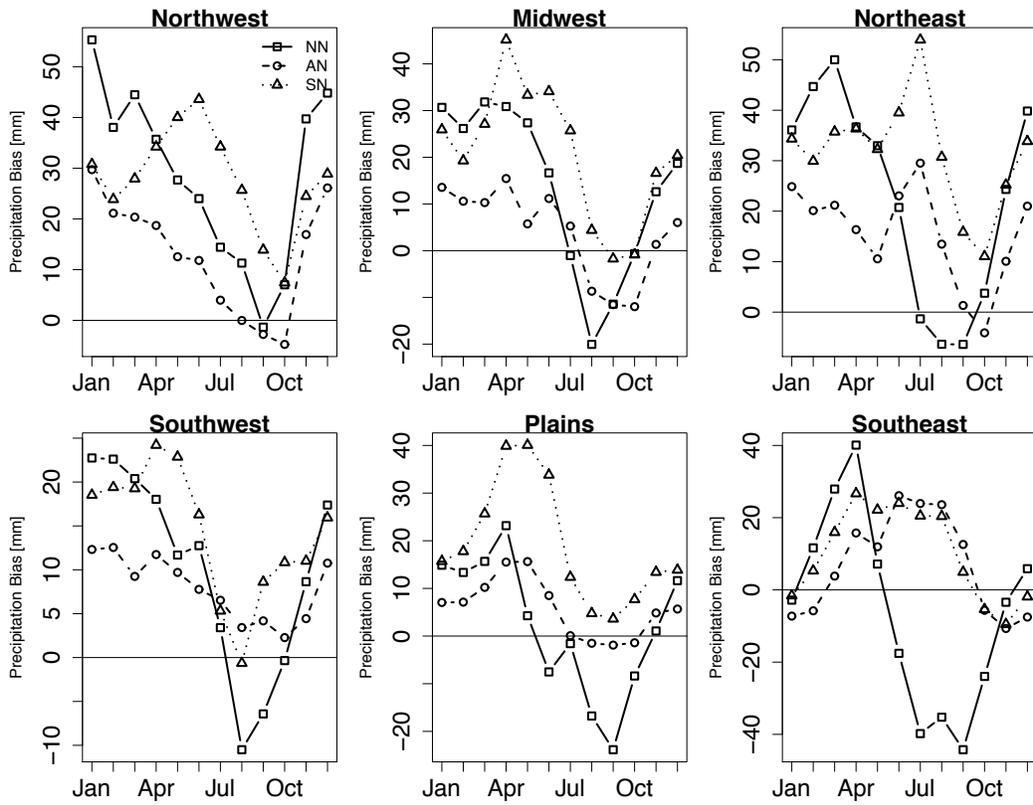
Pattern Correlation															
	T2			Pre			PWAT			Wspd			Wdir		
	NN	AN	SN	NN	AN	SN	NN	AN	SN	NN	AN	SN	NN	AN	SN
1993	0.82	0.96	0.92	0.35	0.85	0.66	0.81	0.93	0.88	0.88	0.98	0.98	0.47	0.91	0.93
2003	0.27	0.74	0.67	-0.19	0.60	0.40	0.14	0.87	0.70	0.29	0.96	0.93	0.05	0.88	0.85



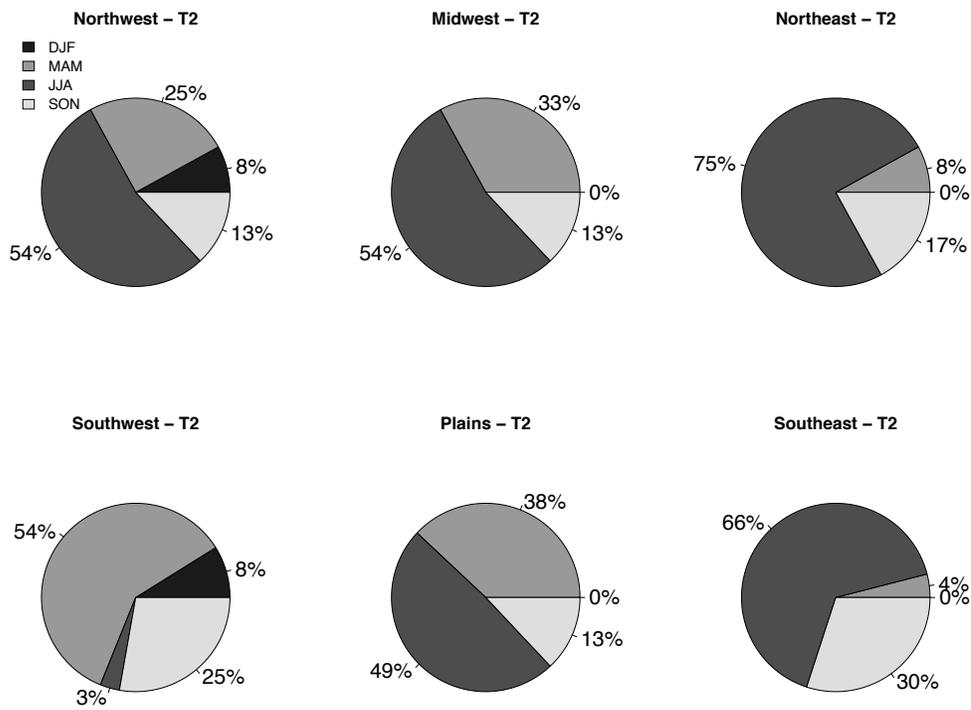
**Fig. 1** WRF outer (108-km) and inner (36-km) domains. Box regions used for model evaluation: Northwest (NW), Southwest (SW), Plains (PL), Midwest (MW), Southeast (SE), and Northeast (NE). Also shown are the boxes used to define Bermuda and New Orleans in calculating the BHI.



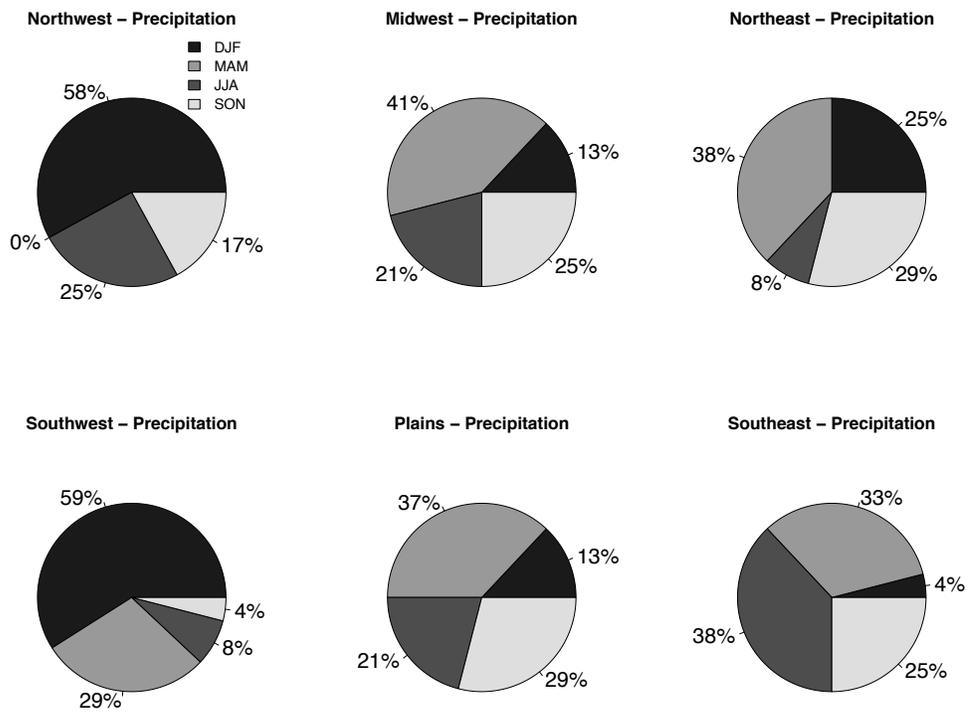
**Fig. 2** Mean monthly-averaged 2-m temperature bias (K) relative to NARR for each of the six verification regions shown in Fig. 1 for R-2 (plus - dot-dash), NN (square - solid), AN (circle - dash), and SN (triangle - dot)



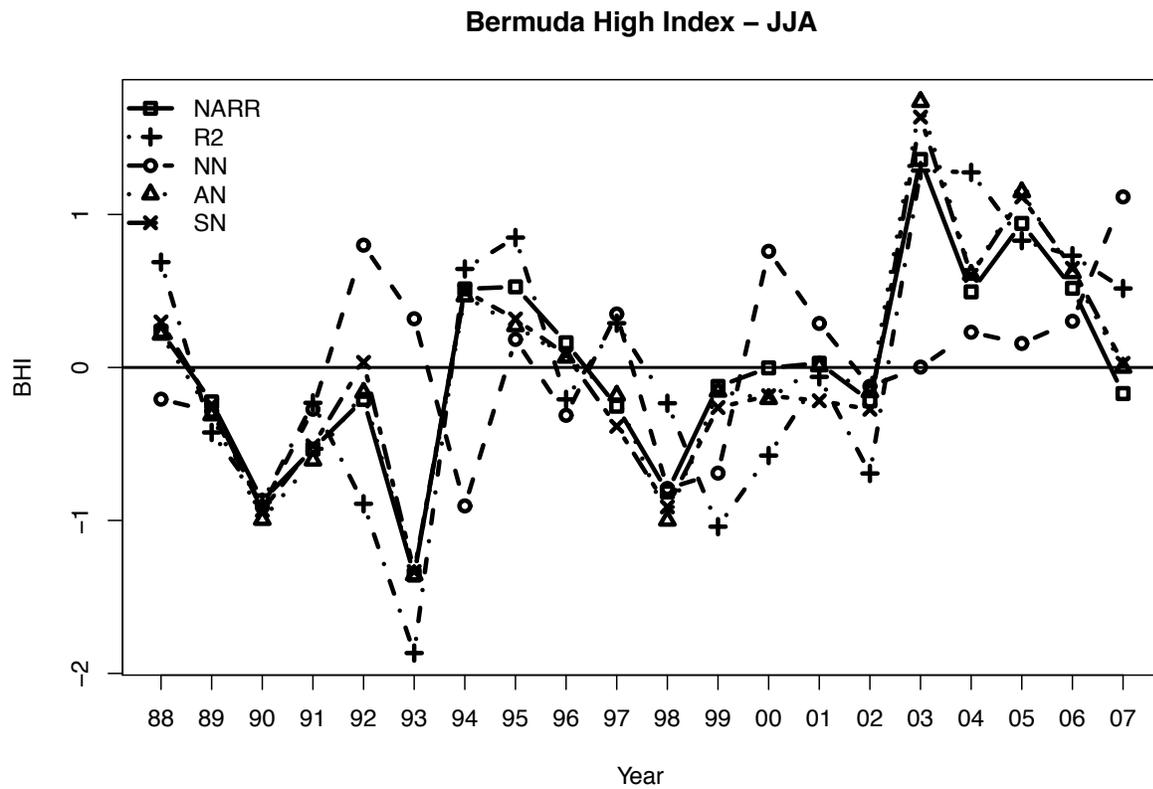
**Fig. 3** Mean monthly-averaged 2-m precipitation bias ( $\text{mm month}^{-1}$ ) relative to NARR for each of the six verification regions shown in Fig. 1 for NN (square-solid), AN (circle - dash), and SN (triangle - dotted)



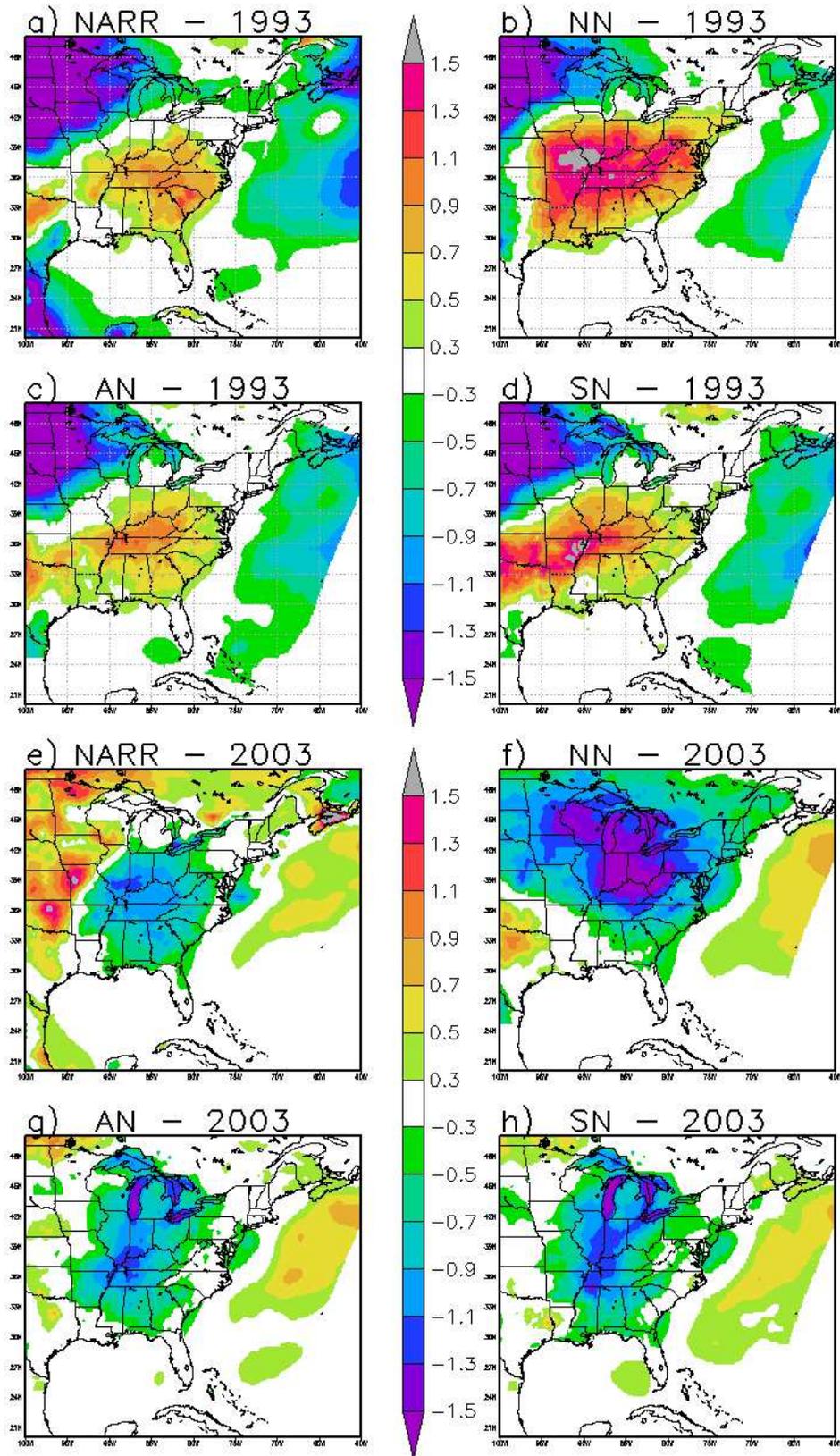
**Fig. 4** Seasonal distribution for NN of the top 10% highest errors in monthly-averaged temperature for each of the six regions shown in Figure 1. Each shade represents a different season



**Fig. 5** Seasonal distribution for NN of the top 10% highest errors in monthly-averaged precipitation for each of the six regions shown in Figure 1. Each shade represents a different season



**Fig. 6** Bermuda High Index calculated for the boreal summer season for NARR (square - solid), R-2 (plus - dot-dash), NN (circle - dash), AN (triangle - dot), and SN (x - dash-dot)



**Fig. 7** 2-m temperature anomaly (K) averaged for the summer season for 1993 (top 4 panels) and 2003 (bottom 4 panels). The panels are labeled a) NARR, b) NN, c) AN, d) SN, e) NARR, f) NN, g) AN, and h) SN

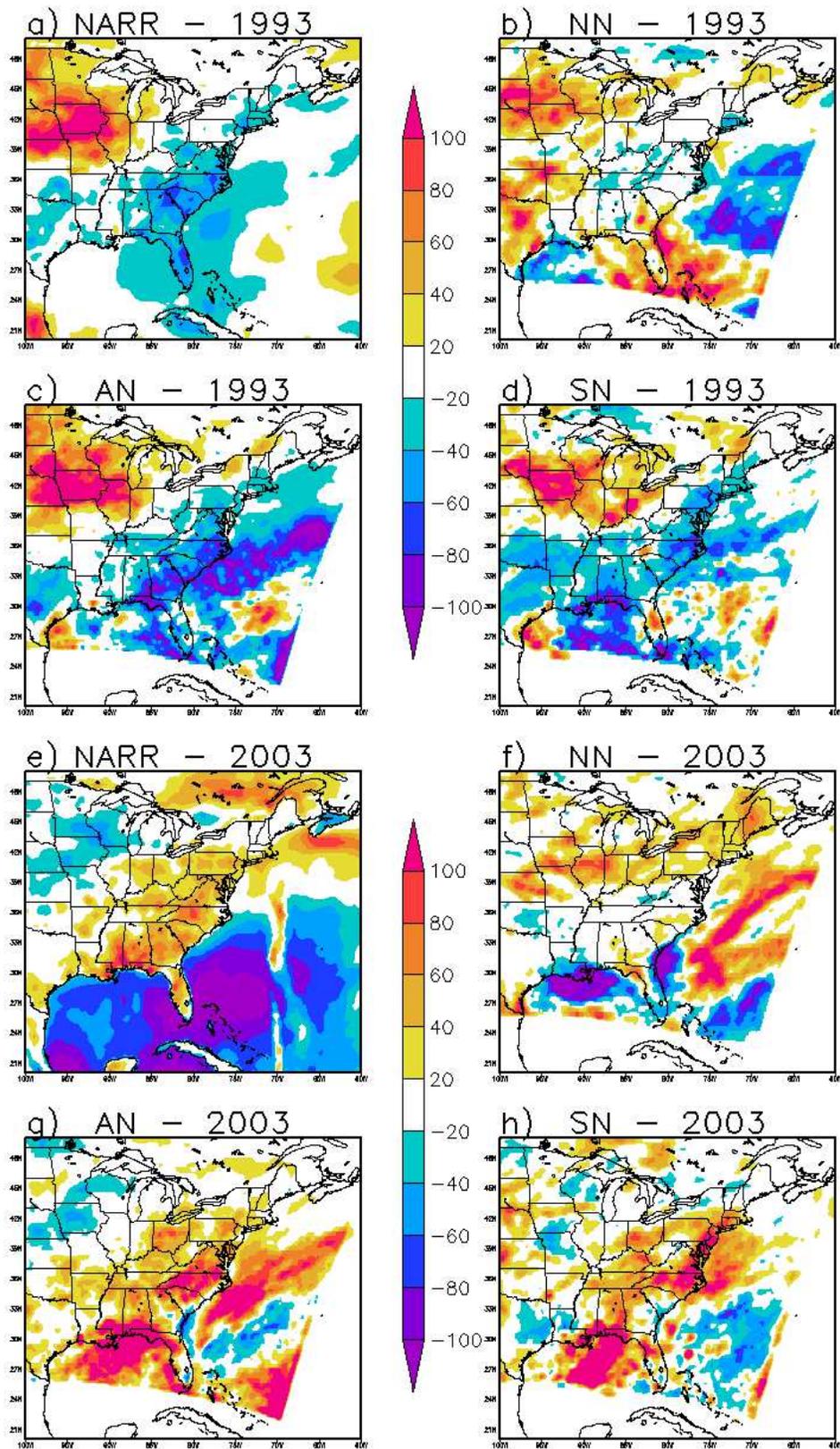


Fig. 8 Same as Fig. 7 but for precipitation anomaly (mm month<sup>-1</sup>)

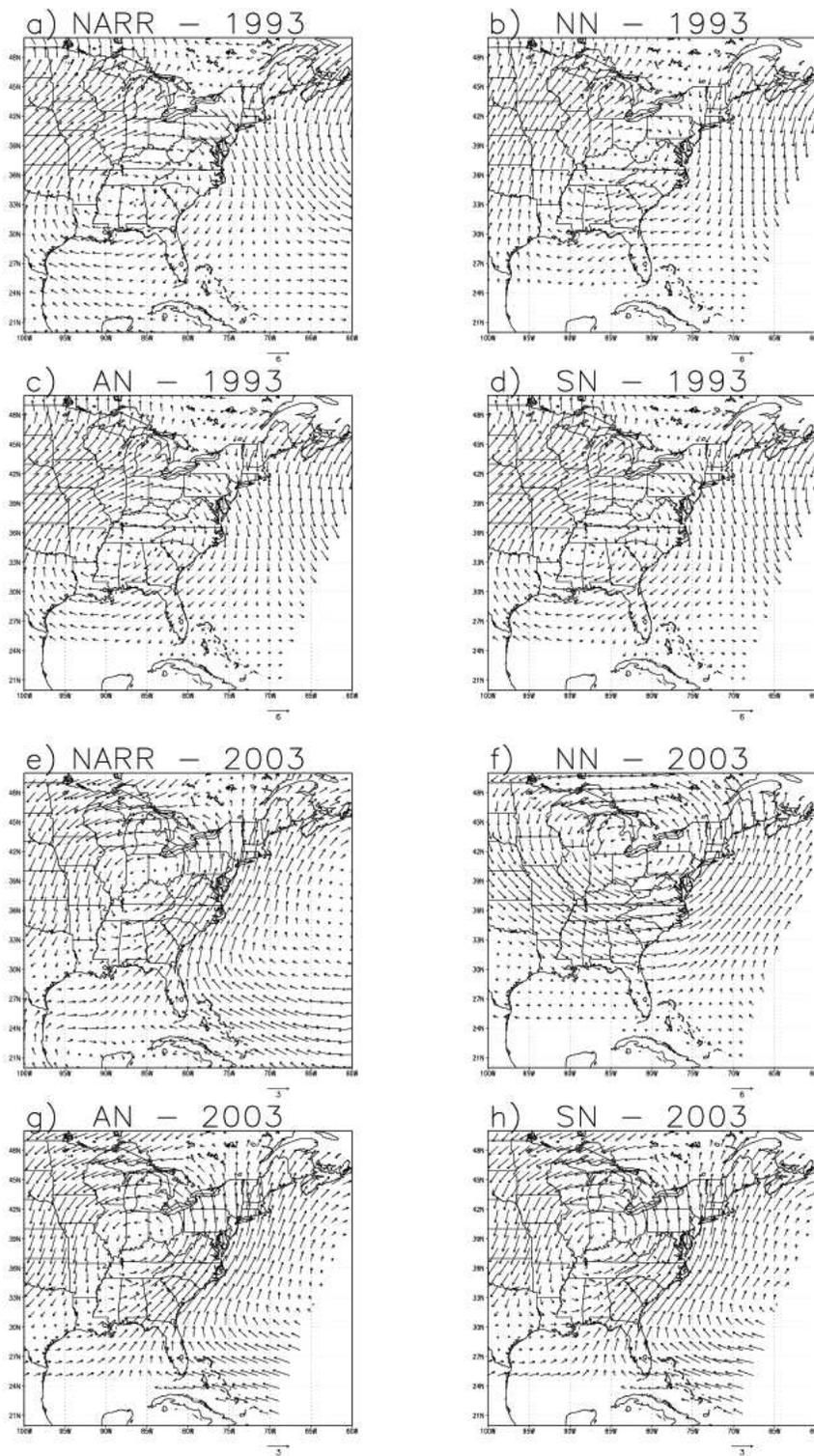
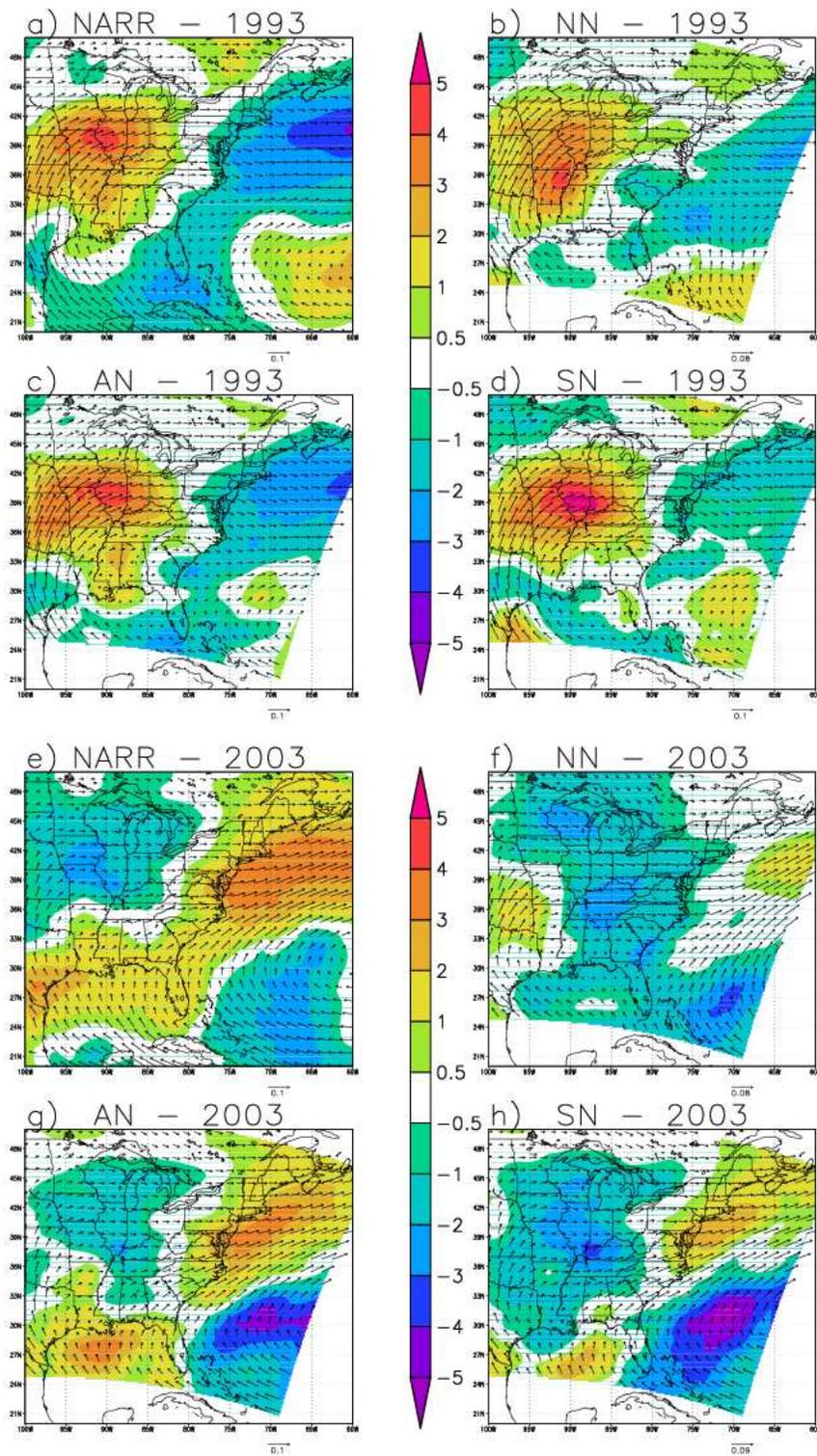


Fig. 9 Same as Fig. 7 but for 500-hPa wind vector anomalies ( $\text{m s}^{-1}$ )



**Fig. 10** Precipitable water anomaly ( $\text{mm day}^{-1}$ , shaded) with 850-hPa moisture transport anomaly ( $\text{m/s}$ ) for the summer season for 1993 (top 4 panels) and 2003 (bottom 4 panels). The panels are labeled a) NARR, b) NN, c) AN, d) SN, e) NARR, f) NN, g) AN, and h) SN