



1 **Impacts of different types of El Niño events on water quality** 2 **over the Corn Belt, United States**

3 Pan Chen^{1,2}, Wenhong Li^{2,*} and Keqi He²

4 ¹ College of Water Resources Science and Engineering, Taiyuan University of Technology, Taiyuan 030024, China

5 ² Earth and Climate Sciences, Nicholas School of the Environment, Duke University, NC 27708, USA

6 *Correspondence to:* Wenhong Li (wenhong.li@duke.edu)

7 **Abstract.** The United States Corn Belt region, which primarily includes two large basins, namely, the Ohio-
8 Tennessee River Basin (OTRB) and the Upper Mississippi River Basin (UMRB), is responsible for the Gulf of
9 Mexico hypoxic zone. Climate patterns such as El Niño can affect the runoff and thus the water quality over the
10 Corn Belt. In this study, the impacts of eastern Pacific (EP) and central Pacific (CP) El Niños on water quality over
11 the Corn Belt region were analyzed using the Soil and Water Assessment Tool (SWAT) models. Our results
12 indicated that at the outlets, annual total nitrogen (TN) and total phosphorus (TP) loads decreased by 13.1% and 14.0%
13 at OTRB, 18.5% and 19.8% at UMRB, respectively, during the EP-El Niño years, whereas during the CP-El Niño
14 years, they increased by 3.3% and 4.6% at OTRB, 5.7% and 4.4% at UMRB, respectively. On the sub-basin scales,
15 more sub-basins showed negative (positive) anomalies of TN and TP during EP- (CP-) El Niño. Seasonal study
16 confirmed that water quality anomalies showed opposite patterns during EP- and CP-El Niño years. At the outlet of
17 OTRB, seasonal anomalies in nutrients matched the El Niño-Southern Oscillation (ENSO) phases, illustrating the
18 importance of climate variables associated with the two types of El Niño on water quality in the region. At the
19 UMRB, TN and TP were also influenced by agriculture activities within the region and their anomalies became
20 greater in the growing seasons during both EP- and CP-El Niño years. Quantitative analysis of precipitation,
21 temperature, and their effects on nutrients suggested that precipitation played a more important role than
22 temperature did in altering water quality in the Corn Belt region during both types of El Niño years. We also found
23 specific watersheds (located in Iowa, Illinois, Minnesota, Wisconsin, and Indiana) that faced the greatest increases in
24 TN and TP loads, were affected by both the precipitation and agricultural activities during the CP-El Niño years.
25 The information generated from this study may help proper decision-making for water environment protection over
26 the Corn Belt.

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32 1 Introduction

33 The Corn Belt region of the United States (U.S.) primarily includes the Ohio-Tennessee River Basin (OTRB) and
34 the Upper Mississippi River Basin (UMRB) (Kellner and Niyogi, 2015; Panagopoulos et al., 2017; Ting et al., 2021).
35 The Corn Belt is a very important area of the agricultural activity of the country, which is also responsible for the
36 Gulf of Mexico hypoxic zone (Panagopoulos et al., 2014, 2015; Rabalais et al., 2007). The required nutrient
37 reduction of the region to decrease hypoxia is the highest among all regions in the Mississippi River Basin
38 (Mississippi River/Gulf of Mexico Watershed Nutrient Task Force, 2011). Hence, water quality change of the Corn
39 Belt region has been receiving considerable attention.

40 The El Niño-Southern Oscillation (ENSO) is a coupled ocean-atmosphere phenomenon that occurs across the
41 tropical Pacific (Trenberth, 1997; Wang and Kumar, 2015). Precipitation and temperature are influenced by ENSO
42 in many places in the U.S., including the Corn Belt region (Gershunov, 1998; Lee et al., 2014; Thomson et al., 2003;
43 Wang and Asefa, 2018). For example, more frequent dry conditions were found east of the Ohio River in the early
44 spring of a decaying El Niño (Wang and Asefa, 2018). ENSO events could also have a significant impact on the
45 water quality of a basin through changing climate factors. Heavy or prolonged rains might contribute to the pollutant
46 loading from agricultural runoff (Paul et al., 1997). Temperature anomalies might change river water quality by
47 affecting evaporation and water temperature. Keener et al. (2010) showed that ENSO significantly altered water
48 flow and nitrate concentration in a southeastern U.S. basin. Sharma et al. (2012) found that worse water quality in
49 southeast Alabama was linked to the stream temperature anomalies during El Niño years. However, what are the
50 impacts of ENSO on water quality in the Corn Belt region have not been studied.

51 In recent years, studies have found two different types of El Niño events: eastern Pacific (EP) and central Pacific
52 (CP) El Niños (Larkin and Harrison, 2005; Li et al., 2011; Tan et al., 2020; Tang et al., 2016; Yeh et al., 2009). The
53 former has warmer sea surface temperature anomalies (SSTAs) in the Niño 3 region (5°N–5°S, 150°W–90°W),
54 while the latter, also called El Niño Modoki, is manifested by maximum SSTAs in the Niño 4 region (5°N–5°S,
55 160°E–150°W). The effects of these two types of El Niño on regional climate and runoff are different. For example,
56 reduced rainfall was found in the northern, central, and eastern parts of the Amazon during the EP-El Niño years
57 (EP-ENYs), while increased rainfall anomalies were observed in most of the Amazon during the CP-El Niño years
58 (CP-ENYs) (Li et al., 2011). CP-El Niños were more effective in causing drought conditions in India due to
59 atmospheric subsidence than EP-El Niños (Kumar et al., 2006). How these two types of El Niño affect the water
60 quality of the Corn Belt has not been studied. Over recent years, EP-El Niño has appeared less frequently, whereas
61 CP-El Niño has become more common (Kao and Yu, 2009; Yeh et al., 2009). In the future, CP-El Niño will likely
62 happen more frequently (Yeh et al., 2009; Yu et al., 2010). Understanding the impact of these different El Niño
63 events on water quality over the Corn Belt is of critical importance for the water quality management of streams and
64 rivers.

65 In this study, we used the Soil and Water Assessment Tool (SWAT) model to estimate the water quality changes
66 due to EP- and CP-El Niños over the Corn Belt. The SWAT model is widely used to assess climate change and
67 alternative land use/land management scenarios on runoff and nutrients in a basin (Afonso de Oliveira Serrão et al.,
68 2022; Chaplot et al., 2004; Chen et al., 2021; Johnson et al., 2015; Vaché et al., 2002; Zhang et al., 2020). The



69 detailed objectives of this study were to 1) analyze the impacts of the two types of El Niño on water quality in the
70 Corn Belt region, and 2) identify the main climate factors that affect the change in water quality due to these El
71 Niños. Water quality change associated with future El Niño change was also discussed. Such information is
72 particularly important to enable decision-makers to take timely actions to reduce nutrient loading under climate
73 change.

74 **2 Data and methods**

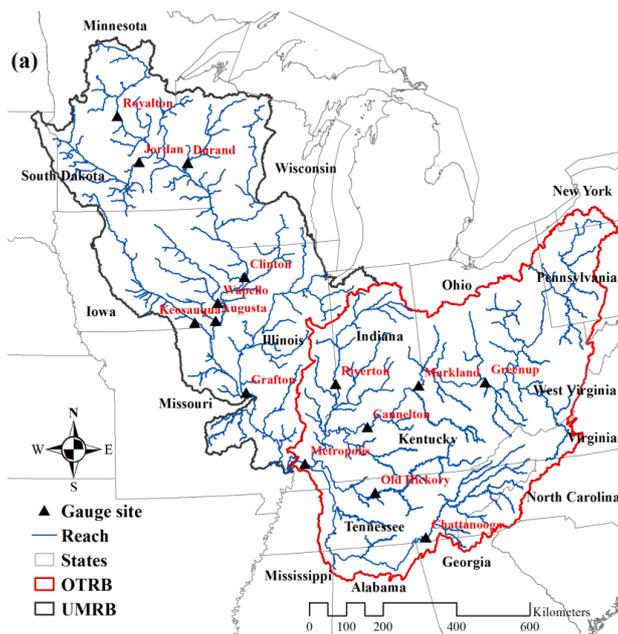
75 **2.1 Data**

76 The study area includes two large basins in the U.S., namely, OTRB (528,000 km²) and UMRB (492,000 km²), as
77 shown in Fig. 1. OTRB comprises a significant portion of Pennsylvania, Ohio, West Virginia, Indiana, Illinois,
78 Kentucky, and Tennessee (Fig. 1a). The amount of annual rainfall in OTRB was high with an average of nearly
79 1200 mm during 1975–2016. OTRB's slopes are steep, especially in the forested Tennessee basin with slopes
80 greater than 5% in most (60%) of the area. The primary land use types are 50% forest, 20% cropland, and 15%
81 pasture (Fig. 1b). The cropland is mainly grown with corn, soybean, and wheat (Santhi et al., 2006). UMRB mainly
82 includes five states: Iowa, Illinois, Missouri, Wisconsin, and Minnesota (Fig. 1a). The mean annual value of rainfall
83 in UMRB during 1975–2016 was 900 mm. UMRB is relatively flat and most of the basin (75%) has a slope lower
84 than 5%. Cropland is the major land use type of the basin (50%) and is primarily grown with corn and soybean (Fig.
85 1b).

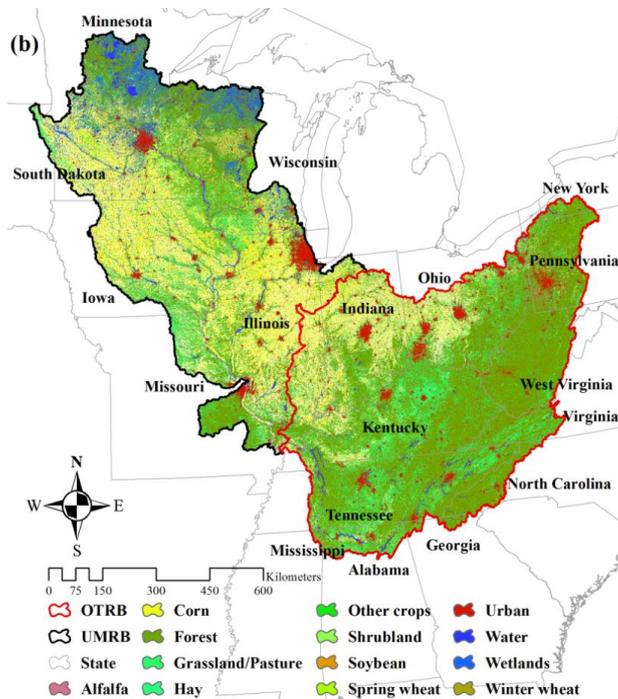
86 The weather data were obtained from 2,242 National Weather Service (NWS) stations in the study area. The
87 historical El Niño years were based on Table 1 in Li et al. (2011) and Table 2 in Ren et al. (2018). In summary, nine
88 EP-El Niño events (1976–1977, 1979–1980, 1982–1983, 1986–1987, 1987–1988, 1991–1992, 1997–1998, 2006–
89 2007, and 2015–2016) and six CP-El Niño events (1977–1978, 1990–1991, 1994–1995, 2002–2003, 2004–2005,
90 and 2009–2010) occurred during the study period (1975–2016). Significance was tested by the Monte Carlo method
91 (Levine and Wilks, 2000; Mo, 2010).

92 Available monthly streamflow and water quality data from 1975 to 2016 came from the 15 United States
93 Geological Survey (USGS) gaging stations in the study area. The final data used for calibration and validation of the
94 model were shown in Table 1.

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99 **Figure 1.** The Ohio-Tennessee River Basin (OTRB) and Upper Mississippi River Basin (UMRB) (a) available
 100 United States Geological Survey (USGS) gage sites (black triangles), reaches (light blue), and the two watersheds
 101 (heavy red: OTRB, heavy dark: UMRB) and (b) land use/land cover.



102 **Table 1.** Available periods of measured streamflow, total suspended sediment (TSS), total nitrogen (TN), and total
 103 phosphorus (TP) at 15 USGS gauge at the OTRB and UMRB

Site Name	Site Number	River Basin	Hydrologically Independent	Drainage (km ²)	Streamflow	TSS	TN	TP
Greenup	03216600		Yes	160,579	1975–2019	-	1975–2019	1975–2019
Markland	03277200		No	215,409	1975–2019	-	-	-
Riverton	03342000		Yes	34,087	1975–2019	-	-	-
Old hickory	03426310	OTRB	Yes	30,233	1988–2007	-	-	-
Cannelton	03303280		No	251,229	1975–2019	-	1975–2019	1975–2019
Metropolis	03611500		No	525,768	1975–2014	-	1975–2016	1975–2016
Chattanooga	03568000		Yes	55,426	1975–2008	-	-	-
Royalton	05267000		Yes	30,044	1975–2019	-	-	-
Jordan	05330000		Yes	41,958	1975–2019	-	-	-
Durand	05369500		Yes	23,336	1975–2019	-	-	-
Clinton	05420500		No	221,703	1975–2019	-	1975–2019	1975–2019
Augusta	05474000	UMRB	Yes	11,168	1975–2019	1975–2017	-	-
Wapello	05465500		Yes	32,375	1975–2019	1978–2017	1978–2019	1977–2019
Keosauqua	05490500		Yes	36,358	1975–2019	-	-	-
Grafton	05587450		No	443,665	1975–2019	1989–2017	1989–2019	1989–2019



104 **2.2 SWAT model description**

105 **2.2.1 Model description**

106 SWAT was developed by the U.S. Department of Agriculture Agricultural Research Services (Arnold et al., 1998). It has
107 been widely used in assessing the effects of climate and land use change on hydrological processes, sediment, and nutrients
108 in a basin (Neitsch et al., 2011; Pagliero et al., 2014; Yen et al., 2016). In the SWAT model, a basin is partitioned into sub-
109 basins, which are further divided into hydrological response units (HRUs) (Gassman et al., 2007; Neitsch et al., 2011;
110 Williams et al., 2008). Runoff, sediment, and nutrient loads are simulated for each HRU and then aggregated for the sub-
111 basins (Chen et al., 2021; Gassman et al., 2007; Neitsch et al., 2011). Hence, the pollution situation of each sub-basin during
112 different El Niño years could be obtained from this model.

113 **2.2.2 Model set-up and calibration**

114 In this study, 8-digit Hydrologic Unit Codes (HUC-8) defined by the USGS were selected as SWAT sub-basins. In total, the
115 OTRB and UMRB included 152 and 131 sub-basins, respectively. Flow paths between the sub-basins were determined using
116 the stream network of the National Hydrography Dataset Plus (NHDPlus) dataset developed by the USGS and U.S.
117 Environmental Protection Agency. Each of the sub-basins was further divided into several spatially uniform HRUs based on
118 land use, soil type, and slope (Chen et al., 2021; Neitsch et al., 2011). Thresholds of 0%, 10%, and 5% were used for land
119 use, soil, and slope, respectively, resulting in a total of 20,157 and 20,581 HRUs in the OTRB and UMRB. Then, point
120 sources (Schwarz et al., 2006), crop management (U.S. Department of Agriculture (USDA) - National Agricultural Statistics
121 Service (NASS), 2017), and tillage (Baker, 2011) dataset were incorporated to build the SWAT model (Chen et al., 2021).

122 SWAT Calibration and Uncertainty Programs (SWAT-CUP) with Sequential Uncertainty Fitting (SUFI-2) algorithm was
123 selected in this large-scale study to complete the calibration of the SWAT model (Abbaspour et al., 2012). The parameters of
124 water flow and water quality in OTRB and UMRB were selected based on a manual experimentation with SWAT parameters
125 and a literature review (Chen et al., 2021; Panagopoulos et al., 2014, 2015; Yen et al., 2016). The calibration steps followed
126 a recent study by Chen et al. (2021). The final parameters were shown in Table S1 and S2 in the supplementary materials.
127 The calibration results indicated that the SWAT model could rationally capture the observation (see Section 2.2.3).

128 **2.2.3 Model performance**

129 Overall, SWAT simulated the water flow of the OTRB and UMRB reasonably well in both calibration (1997–2016) and
130 validation (1975–1996) periods (Table 2). The coefficient of determination (R^2) and Nash-Sutcliffe efficiency (NSE) values
131 were larger than 0.5 for almost all the USGS gages except Chattanooga, and percent bias ($PBIAS$) values were all acceptable
132 during the calibration periods (1997–2016) based on the $\leq \pm 25\%$ deviation criterion (Moriassi et al., 2015). The validation
133 results also showed acceptable static values and in some cases were even better, such as Chattanooga and Grafton. Figures



134 S1 and S2 in the supplementary materials further demonstrated good agreement between the calculated and observed
 135 streamflow across OTRB and UMRB; particularly, most of the peaks and recession limbs were well matched in the
 136 simulations.

137 Modeled water quality was generally in agreement with the observations for the two river basins, as most of the *PBIAS*
 138 values were within bias criteria for sediment, TN, and TP during both the calibration and validation periods (Santhi et al.,
 139 2014) (Table 3). The only exception that did not meet the criteria was the sediment simulation at Augusta. The upstream
 140 drainage area of Augusta was relatively small (only around 3% of the UMRB), thus its influence on downstream sediment
 141 and pollutant transport downstream was minor (Fig. 1a). The sediment statistics in OTRB were not calculated here because
 142 of a lack of observations (Panagopoulos et al., 2015). Instead, we compared the simulated annual sediment at Metropolis
 143 with observations during 1975–2010 and found that the difference was also within the bias criteria for sediment
 144 (Panagopoulos et al., 2015; Santhi et al., 2014). Moreover, most of the *NSE* and R^2 values were positive and many R^2 values
 145 were greater than 0.5, indicating that the simulated water quality was reasonable (Chen et al., 2021). This finding could be
 146 further proved by the graphs of observed and simulated sediment and nutrients (Figs. S3 and S4).

147 **Table 2.** Monthly streamflow calibration and validation statistics.

Gauge Site	Calibration (1997–2016)			Validation (1975–1996)		
	R^2	<i>NSE</i>	<i>PBIAS</i>	R^2	<i>NSE</i>	<i>PBIAS</i>
Greenup	0.88	0.87	1.6	0.87	0.82	15.9
Markland	0.88	0.87	6.4	0.88	0.81	17.4
Riverton	0.84	0.83	−1.4	0.8	0.78	−7.4
Old Hickory	0.82	0.75	−4.7	0.81	0.79	7.8
Cannelton	0.89	0.87	9.6	0.9	0.84	15.9
Metropolis	0.84	0.83	4.1	0.88	0.8	15.8
Chattanooga	0.59	0.48	6.1	0.62	0.53	10.7
Royalton	0.63	0.56	−4.7	0.58	0.55	−0.5
Jordan	0.64	0.59	0.7	0.77	0.66	−22.8
Durand	0.65	0.5	−20.9	0.6	0.53	−10.7
Clinton	0.63	0.56	6.6	0.63	0.59	9.8
Augusta	0.75	0.7	−9.9	0.78	0.76	−9.9
Wapello	0.71	0.67	−2.3	0.76	0.76	−1
Keosauqua	0.76	0.73	5.8	0.8	0.78	14
Grafton	0.69	0.62	12.5	0.77	0.7	16.1

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Table 3. Monthly TSS, TN, and TP calibration and validation statistics.

Variable	Gauge Site	Calibration(1997–2016)			Validation(1975–1996)		
		R^2	NSE	$PBIAS$	R^2	NSE	$PBIAS$
TSS	Augusta	0.34	0.1	64.5	0.39	0.02	77.2
	Wapello	0.45	0.15	-21.5	0.51	0.48	13.6
	Grafton	0.43	0.22	0.7	0.26	0.07	13.9
TN	Greenup	0.57	0.23	-13.7	0.59	0.52	17
	Cannelton	0.63	0.59	8.3	0.59	0.46	26.9
	Metropolis	0.58	0.36	-9.2	0.57	0.51	14.9
	Clinton	0.43	-0.61	2.7	0.36	-0.25	9.4
	Wapello	0.51	0.05	1.8	0.44	0.32	17.6
	Grafton	0.54	0.15	2.2	0.53	0.08	2
TP	Greenup	0.56	0.46	-29.6	0.59	0.57	8.6
	Cannelton	0.56	0.48	21.2	0.44	0.35	28.1
	Metropolis	0.49	0.41	-8.6	0.45	0.38	-5.2
	Clinton	0.44	-0.59	-11.1	0.42	0.13	10.9
	Wapello	0.6	0.24	-19.1	0.55	0.29	-16.3
	Grafton	0.57	0.25	0.8	0.54	0.12	-14.8

155 **3 Results**

156 **3.1 Impacts of the EP- and CP-El Niños on the water quality in the Corn Belt**

157 **3.1.1 Annual composite**

158 (1) Water quality at the outlet

159 Table 4 lists the nutrient change during EP- and CP-ENYs at the outlet of OTRB and UMRB. Annual loads of TN and TP
 160 decreased during the EP-ENYs, while the pattern reversed during the CP-ENYs in the U.S. Corn Belt region. Specifically,
 161 compared to normal years, the TN and TP decreased by 13.1% (61,300 metric ton yr⁻¹, hereafter ton yr⁻¹) and 14.0% (7,300
 162 ton yr⁻¹) during EP-ENYs, respectively, whereas they increased by 3.3% (15,500 ton yr⁻¹) and 4.6% (2,400 ton yr⁻¹) during
 163 CP-ENYs at the outlet of the OTRB, respectively (Table 4). TN and TP at the outlet of the UMRB showed a similar pattern
 164 as that of the OTRB, decreasing (increasing) by 18.5% (5.7%) and 19.8% (4.4%) in EP (CP)-ENYs. Furthermore, EP-El
 165 Niños had a much greater impact on water quality than CP-El Niños at the outlets of OTRB and UMRB. The magnitudes of
 166 variation in both TN and TP during the EP-ENYs were three to four times greater than those during the CP-ENYs (Table 4).

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170 **Table 4.** Soil and Water Assessment Tool (SWAT) model’s estimates of annual mean TN and TP at the outlets of the OTRB
 171 and UMRB during all simulation year (1975–2016), eastern Pacific (EP)-El Niño years and central Pacific (CP)-El Niño
 172 years.

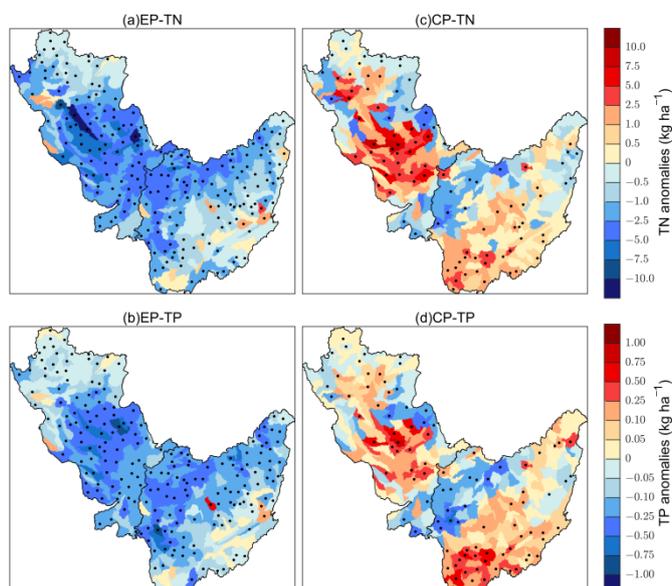
River Basin	Variable	Average	EP		CP	
			Anomaly	Percent (%)	Anomaly	Percent (%)
OTRB	TN ($\times 10^3$ ton yr ⁻¹)	469.0	-61.3	-13.1	15.5	3.3
	TP ($\times 10^3$ ton yr ⁻¹)	52.2	-7.3	-14.0	2.4	4.6
UMRB	TN ($\times 10^3$ ton yr ⁻¹)	526.2	-97.5	-18.5	29.9	5.7
	TP ($\times 10^3$ ton yr ⁻¹)	41.0	-8.1	-19.8	1.8	4.4

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174 (2) Water quality at the sub-basin scale

175 We analyzed water quality change on the sub-basin scale during EP-ENYs and CP-ENYs, respectively (Fig. 2). Anomalous
 176 patterns of water quality associated with El Niño events within the Corn Belt varied in space. Clearly, more sub-basins
 177 showed negative anomalies of TN and TP during EP-ENYs, whereas more sub-basins showed positive anomalies during CP-
 178 ENYs (Fig. 2). Specifically, during EP-ENYs, significant below-average TN and TP were found almost in the whole OTRB
 179 and UMRB with maximum reductions of TN and TP up to -11.7 and -0.9 kg ha⁻¹, respectively, which were of similar
 180 magnitudes to the mean values (12.7 kg of N ha⁻¹ and 1.0 kg of P ha⁻¹) in the Corn Belt region (Figs. 2a and 2b). During CP-
 181 ENYs, positive anomalies mainly occurred throughout the southern OTRB and UMRB. In the northern UMRB and OTRB,
 182 about 42.4% and 41.7% of sub-basins tended to have below-average TN and TP (Figs. 2c and 2d). These patterns coincided
 183 with the TN and TP changes at the outlets of the two basins, which could also explain why greater changes in TN and TP
 184 occurred in EP-ENYs than in CP-ENYs at the outlets of the OTRB and UMRB (Table 4).

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187 **Figure 2.** Composite results of annual TN and TP anomalies (unit: kg ha^{-1}) in EP-El Niño years (a and b) and in CP-El Niño
188 years (c and d) during the period of 1975–2016. Stippling denotes anomalies significantly different from zero at the 95%
189 confidence level based on the Monte Carlo test.

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191 3.1.2 Seasonal composite

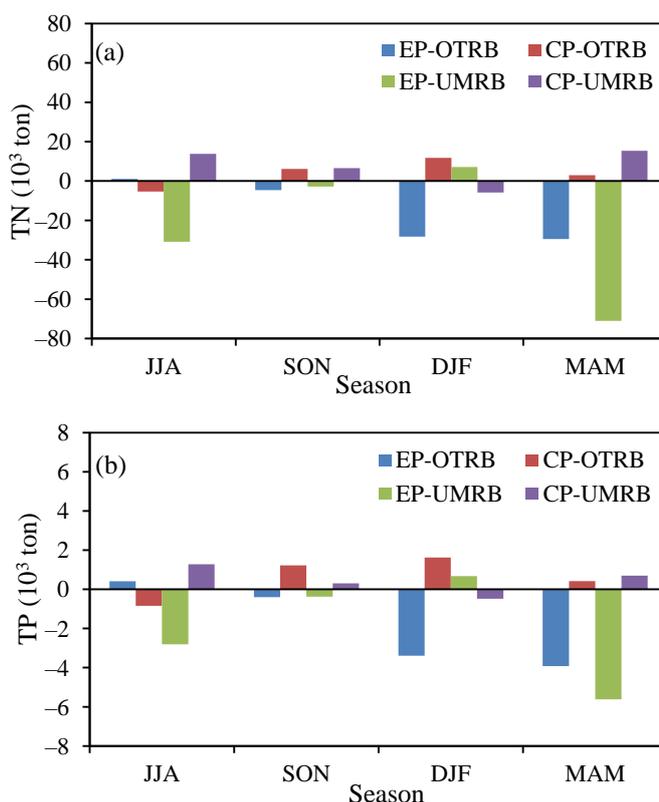
192 (1) Water quality at the outlet

193 Figure 3a showed TN anomalies at the OTRB and UMRB outlets in each season during EP- and CP-ENYs. At the outlet of
194 the OTRB, seasonal anomalies of the water quality reached the maximum when ENSO signals were the strongest (Fig. 3a).
195 El Niño usually developed in boreal summer (June-August, JJA) and autumn (September-November, SON), peaked in
196 winter (December of the current year and January and February of the following year, DJF), and decayed in spring (March-
197 May, MAM) (Trenberth, 1997; Li et al., 2011). Maximum changes in TN occurred in the winter and spring seasons during
198 EP-ENYs (decreased by 28.3×10^3 and 29.5×10^3 metric tons (hereafter ton), respectively, Fig. 3a). During CP-ENYs, TN
199 increased by 11.7×10^3 tons in winter, and did not change much in the rest of the three seasons (spring, summer, and autumn)
200 compared to that in the normal years (Fig. 3a). Seasonal TN anomalies at the outlet of the UMRB were different from those
201 of the OTRB. Figure 3a showed that TN decreased by 71.0×10^3 and 30.8×10^3 tons, respectively, in the spring and summer
202 seasons during EP-ENYs; but in winter and autumn, TN did not change much compared to normal years. Similarly, during
203 CP-ENYs, TN anomalies were greater in spring and summer although TN anomalies became positive during CP-ENYs,
204 different from TN changes during EP-ENYs (Fig. 3a).



205 Figure 3b demonstrated that the seasonal changes of TP during EP- and CP-ENYs were similar to those of TN in both
206 OTRB and UMRB during EP-ENYs. At the OTRB the magnitudes of TP reduction in boreal winter and spring were greater
207 than those in the summer and autumn seasons during EP-ENYs. During CP-ENYs, TP anomalies were greater in both
208 autumn and winter (Fig. 3b). This phenomenon was probably related to the different duration of the two types of El Niño.
209 The mean duration of EP-El Niño was about 15 months (Mo, 2010), its impact on water quality could last into the following
210 spring; while El Niño Modoki usually lasted for about eight months (Mo, 2010, Yu et al., 2010), therefore the impacts of the
211 CP-El Niño on water quality usually ended in the winter. Besides, at the UMRB maximum changes in TP occurred in the
212 spring and summer seasons during both EP- and CP-ENYs.

213 In summary, more seasons showed negative anomalies of water quality during EP-ENYs, whereas more seasons showed
214 positive anomalies during CP-ENYs (Fig. 3). Consequently, the annual TN and TP anomalies over the Corn Belt region
215 showed the opposite pattern during EP-ENYs and CP-ENYs (Table 4).



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218 **Figure 3.** Seasonal anomalies of (a) TN and (b) TP (unit: 10³ tons) in summer (June-August, JJA), autumn (September-
219 November, SON), winter (December of the current year and January and February of the following year, DJF) and spring
220 (March-May, MAM) at the outlets of the OTRB and UMRB during EP-El Niño years and CP-El Niño years.

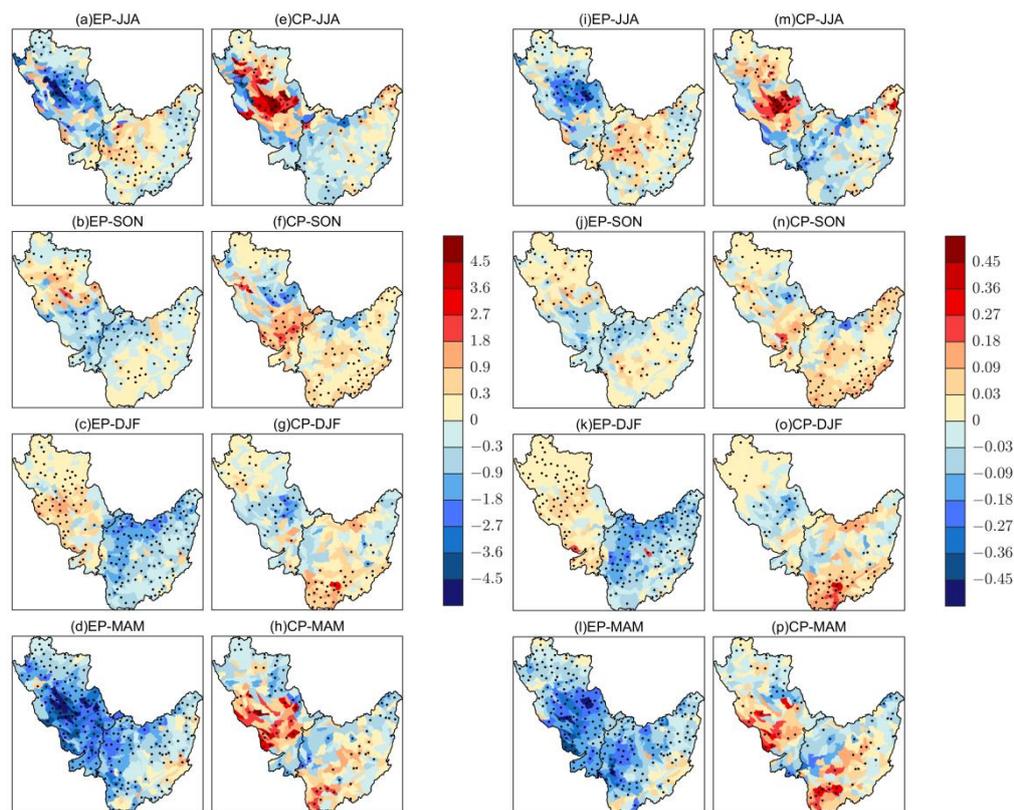
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223 (2) Water quality at the sub-basin scale

224 Water quality at the Corn Belt can vary in different locations/sub-basins and change between seasons. Figure 4 showed
225 spatial patterns of TN and TP anomalies at the OTRB and UMRB in each season during the CP- and EP-ENYs, separately.
226 EP-El Niño was characterized by negative TN and TP anomalies over most of the OTRB and UMRB for all seasons (Figs.
227 4a–4d and 4i–4l). Significant below-average TN and TP occurred in almost all of the UMRB and the eastern OTRB in the
228 summer when EP-El Niño was developing in the tropical Pacific, with maximum reductions up to -4.5 and -0.45 kg ha^{-1} ,
229 respectively (Figs. 4a and 4i). The negative water quality anomalies moved to the southern UMRB and northern OTRB in
230 the autumn (Figs. 4b and 4j). These negative anomalies further moved to the whole OTRB when EP-El Niño was mature in
231 the winter (Figs. 4c and 4k). This result generally agreed with previous findings of the El Niño impacts on the precipitation
232 in the study area, which indicated that the precipitation in the Ohio Valley was sensitive to El Niño events and showed
233 negative precipitation anomalies at EP-ENYs (Mo, 2010; Twine et al., 2005). In spring, severe TN and TP deficits were most
234 apparent almost all over the Corn Belt area (Figs. 4d and 4l).



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236 **Figure 4.** Seasonal composite of (a-h) TN and (i-p) TP anomalies (unit: kg ha^{-1}) in summer (JJA) (a and i), autumn (SON)
237 (b and j), winter (DJF) (c and k), and spring (MAM) (d and l) in EP-El Niño years during the period of 1975–2016; (e-h) and
238 (m-p) are the same as (a-d) and (i-l) but for CP-El Niño years. Stippling denotes anomalies significantly different from zero
239 at the 95% confidence level based on the Monte Carlo test.



240 In contrast, during CP-ENYs, positive TN and TP anomalies were scattered in most of the Corn Belt region with the
241 highest TN and TP anomalies increase up to 4.5 and 0.45 kg ha⁻¹, respectively (Figs. 4e–4h and 4m–4p). In summer,
242 northern UMRB and eastern OTRB were characterized by above-normal water quality (Figs. 4e and 4m). The positive TN
243 and TP anomalies moved to the southern UMRB and whole OTRB in the autumn (Figs. 4f and 4n). In winter, these positive
244 anomalies were concentrated in the southern OTRB and northern UMRB (Figs. 4g and 4o). In the spring, abnormally high
245 TN and TP mainly occurred in the southern part of the UMRB while most of the positive and negative anomalies in the
246 OTRB region were insignificant (Figs. 4h and 4p).

247 In conclusion, water quality anomalies showed opposite patterns during EP-ENYs and CP-ENYs on both annual and
248 seasonal time scales in the Corn Belt region. Furthermore, EP-El Niño seemed to have a greater and long-lasting impact on
249 TN and TP than CP-El Niño. Hence, treating the two as a single phenomenon was not appropriate when analyzing the
250 impacts of ENSO on water quality.

251 3.2 Possible climate reasons for the water quality change during CP- and EP-El Niño events

252 As precipitation and temperature usually respond differently to the two types of El Niño at different temporal and spatial
253 scales (Li et al., 2011; Tan et al., 2020), we hereafter analyzed these climate factors' impacts accordingly in the Corn Belt
254 region.

255 3.2.1 Precipitation

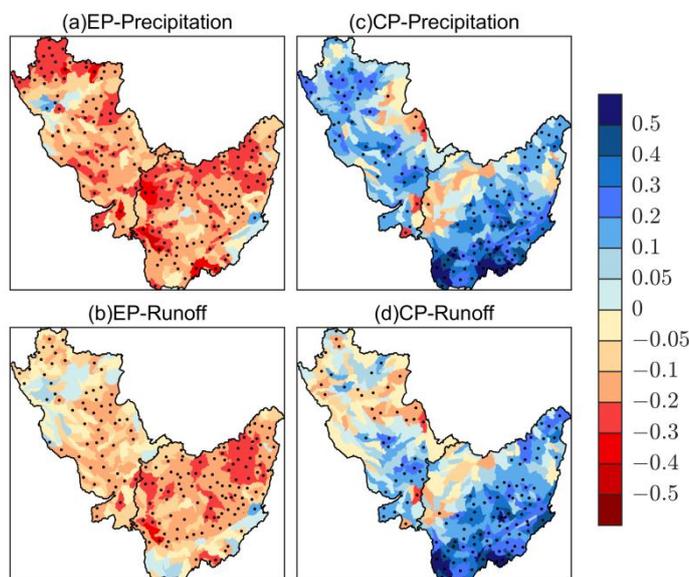
256 Decreased precipitation in EP-ENYs was one of the important reasons that improved water quality in the Corn Belt region.
257 This finding could be shown in the spatial patterns of annual precipitation (Fig. 5a), and TN and TP anomalies (Figs. 2a and
258 2b). For example, significantly below-normal precipitation occurred in much of the OTRB and UMRB (278 out of 283 sub-
259 basins) during EP-ENYs (Fig. 5a), thus TN and TP were reduced in most parts of the Corn Belt (Figs. 2a and 2b). During
260 CP-ENYs, precipitation anomalies became positive in 226 out of 283 sub-basins (Fig. 5c). Correspondingly, TN and TP
261 concentrations were elevated compared to normal years (Figs. 2c and 2d). Spatially, there were more sub-basins (98.2%) and
262 larger area with negative precipitation anomalies in EP-ENYs than positive anomalies (79.9% sub-basins) during CP-ENYs
263 in the Corn Belt, indicating that EP-El Niño tended to have a much wider impact on precipitation and thus water quality than
264 CP-El Niño.

265 To better understand how precipitation affected water quality in the Corn Belt, variations of annual runoff during EP-
266 ENYs and CP-ENYs were also discussed because nitrogen and phosphorus were transported by runoff (Neitsch et al., 2011),
267 and precipitation was a very important source of runoff (Gassman et al., 2007). The overall patterns of runoff anomalies
268 (Figs. 5b and 5d) were similar to those of precipitation anomalies (Figs. 5a and 5c) during the two types of El Niño with
269 pattern correlations being 0.79 at OTRB and 0.75 at UMRB in EP-ENYs, 0.96 and 0.77 in CP-ENYs, respectively.
270 Specifically, during EP-ENYs, negative annual runoff anomalies occurred in most of the Corn Belt region (Fig. 5b),
271 resulting in reduced TN and TP compared to normal years (Figs. 2a and 2b). In contrast, in CP-ENYs, positive runoff



272 anomalies were mainly concentrated in southern OTRB and UMRB (Fig. 5d), more nutrients were thus carried out by the
273 runoff associated with above-normal precipitation in the area (Figs. 2c and 2d). We also noticed that EP-El Niño had a wider
274 influence on runoff than CP-El Niño, as 71 more sub-basins (259 vs 188) and larger areas with significant runoff anomalies
275 were found in EP-ENYs than in CP-ENYs (Figs. 5b and 5d), generally consistent with the spatial patterns of precipitation
276 changes due to the El Niños (Figs. 5a and 5c). The phenomena could partly explain why EP-El Niños tended to have greater
277 impacts on the water quality than CP-El Niños at the outlets of the OTRB and UMRB.

278 Some differences also existed among precipitation, runoff, and nutrients. For precipitation and runoff, annual precipitation
279 anomalies were greater than runoff anomalies in the El Niño years. For example, precipitation anomalies of many sub-basins
280 in northern UMRB were stronger than -0.2 mm day^{-1} during EP-ENYs, but the magnitudes of runoff anomalies were
281 weaker than -0.2 mm day^{-1} (Figs. 5a and 5b). During CP-ENYs, most annual precipitation and runoff anomalies were
282 positive, but the magnitudes of precipitation anomalies were generally higher than that of runoff anomalies (Figs. 5c and 5d).
283 These findings suggested that the impact of El Niño on runoff was weakened by the land surface hydrological process. For
284 runoff and nutrients, changes of runoff in El Niño years were greater in the OTRB than in the UMRB (Figs. 5b and 5d),
285 whereas changes of TN and TP were smaller in the OTRB than in the UMRB during both EP-ENYs and CP-ENYs (Fig. 2).
286 The discrepancies between annual runoff and nutrient anomalies in the OTRB and UMRB may be related to the presence of
287 more cropland in the UMRB where more fertilizers were used (Chen et al., 2021, see Section 4.1 for details), hence a
288 relatively small change of runoff due to El Niño-induced precipitation could lead to large TN and TP variations.

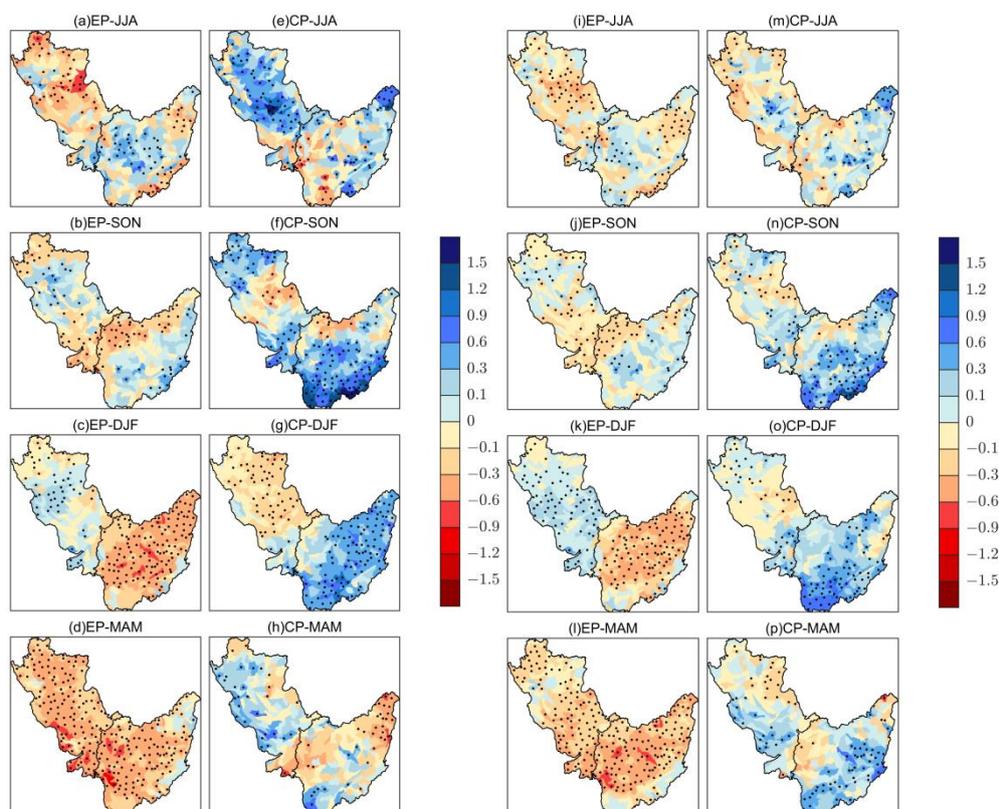


289
290 **Figure 5.** Composite of annual precipitation and runoff anomalies (unit: mm day^{-1}) for the two types of El Niño during the
291 period of 1975–2016 (a and b) in EP-El Niño years and (c and d) in CP-El Niño years. Stippling denotes anomalies
292 significantly different from zero at the 95% confidence level based on the Monte Carlo test.



293 Seasonal patterns of precipitation and runoff anomalies (Fig. 6) further proved the impacts of precipitation on water
294 quality through runoff during EP-ENYs and CP-ENYs. In summer, significantly below-normal precipitation occurred in
295 almost all of the UMRB when EP-El Niño was developing in the tropical Pacific, with maximum reductions of precipitation
296 up to -0.9 mm day^{-1} in the region (Fig. 6a). The runoff anomaly pattern was much the same as that of precipitation, but in a
297 weaker magnitude—less than -0.6 mm day^{-1} in UMRB (Fig. 6i). At the same time, TN and TP decreased in the area (Figs.
298 4a and 4i) because fewer nutrients were carried out by the reduced runoff in summer (Fig. 6i). Similar change patterns of
299 precipitation (Figs. 6b-6d), runoff (Figs. 6j-6l), and nutrients (Figs. 4b-4d and 4j-4l) could also be found in other seasons
300 during EP-ENYs. We noticed that negative runoff and precipitation anomalies reached their maximum in spring throughout
301 the Corn Belt, leading to better water quality in the region compared to the normal years. CP-El Niño events caused the
302 opposite patterns of seasonal precipitation (Figs. 6e-6h) and runoff anomalies (Figs. 6m-6p) in the Corn Belt region; TN and
303 TP thus increased in central and southern UMRB in the spring, summer, and autumn, and most of OTRB from autumn to
304 winter (Figs. 4e-4h and 4m-4p). Some differences also existed between precipitation and water quality in the UMRB,
305 especially in spring and summer. During these two seasons, the variation of the seasonal precipitation at each sub-basin was
306 relatively uniform in both EP-ENYs (Figs. 6a and 6d) and CP-ENYs (Figs. 6e and 6h), but nutrient variations were high in
307 some sub-basins of the UMRB (Figs. 4a, 4d, 4e, 4h, 4i, 4l, 4m, and 4p). This phenomenon suggested that water quality was
308 also influenced by local factors besides climate variables associated with El Niños.

309



310

311 **Figure 6.** Composite of seasonal (a-h) precipitation and (i-p) runoff anomalies (unit: mm day^{-1}) in JJA (a and i), SON (b and
312 j), DJF (c and k), and MAM (d and l) in EP-El Niño years during the period of 1975–2016; (e-h) and (m-p) are the same as
313 (a-d) and (i-l) but for CP-El Niño years. Stippling denotes anomalies significantly different from zero at the 95% confidence
314 level based on the Monte Carlo test.

315

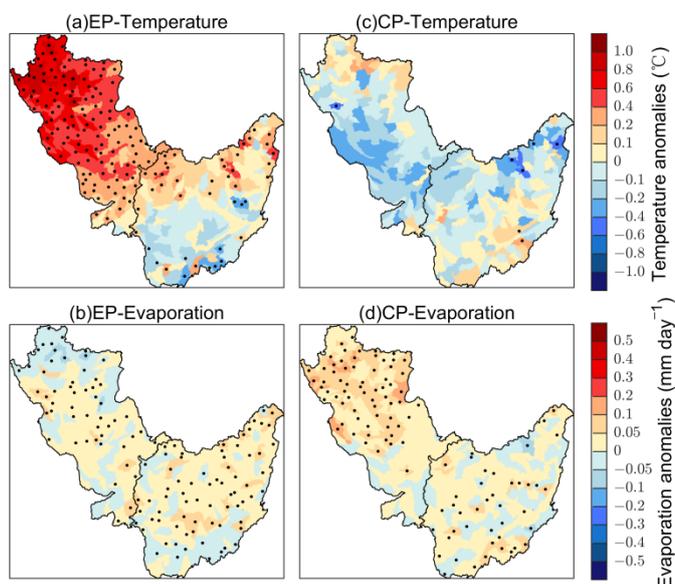
316 3.2.2 Temperature

317 Temperature changes have previously been found to affect water quality by changing the evaporation process of the water
318 cycle (Neitsch et al., 2011; Sun et al., 2011). Thus, how evaporation varied associated with temperature change during El
319 Niños in the Corn-Belt region was analyzed. Compared to normal years, the annual temperature increased over the UMRB,
320 but decreased in the OTRB, especially in the southern OTRB during EP-ENYs (Fig. 7a). Evaporation slightly increased in
321 most of OTRB (Fig. 7b), which did not share the same pattern with temperature change on the annual time scale (Fig. 7a).
322 This might be due to the fact that temperature directly affected potential evapotranspiration (Neitsch et al., 2011), the ability
323 of the atmosphere to remove water from the surface through both evaporation and transpiration, but the actual
324 evaporation/evapotranspiration was also related to other variables such as the amount of water available for evaporation
325 besides temperature. Enhanced evaporation further reduced runoff (Bales et al., 2017). Thus, decreased precipitation (Fig. 5a)



326 and enhanced evaporation (Fig. 7b) during the EP-ENYs would facilitate runoff decline and cause a much wider impact of
327 EP-El Niño events on water quality in the Corn Belt region (Fig. 2). During CP-ENYs, temperature decreased insignificantly
328 in most of the Corn Belt region (Fig. 7c). Evaporation increased in more sub-basins over the UMRB (Fig. 7d). The enhanced
329 evaporation (Fig. 7d) tended to offset, to some extent, the impact of higher than normal precipitation (Fig. 5c) on water
330 quality during CP-ENYs.

331 The impacts of the two climate factors, precipitation and temperature (through evaporation), on runoff were compared.
332 During the El Niño years, the magnitude of annual precipitation change was often greater than 0.1 mm day^{-1} (Figs. 5a and 5c)
333 while most of the annual evaporation varied between -0.05 and 0.05 mm day^{-1} (Figs. 7b and 7d). This suggested that both
334 precipitation and evaporation influence water quality through runoff, but precipitation seemed to play a more important role
335 in altering water quality over the Corn Belt region during El Niño years.



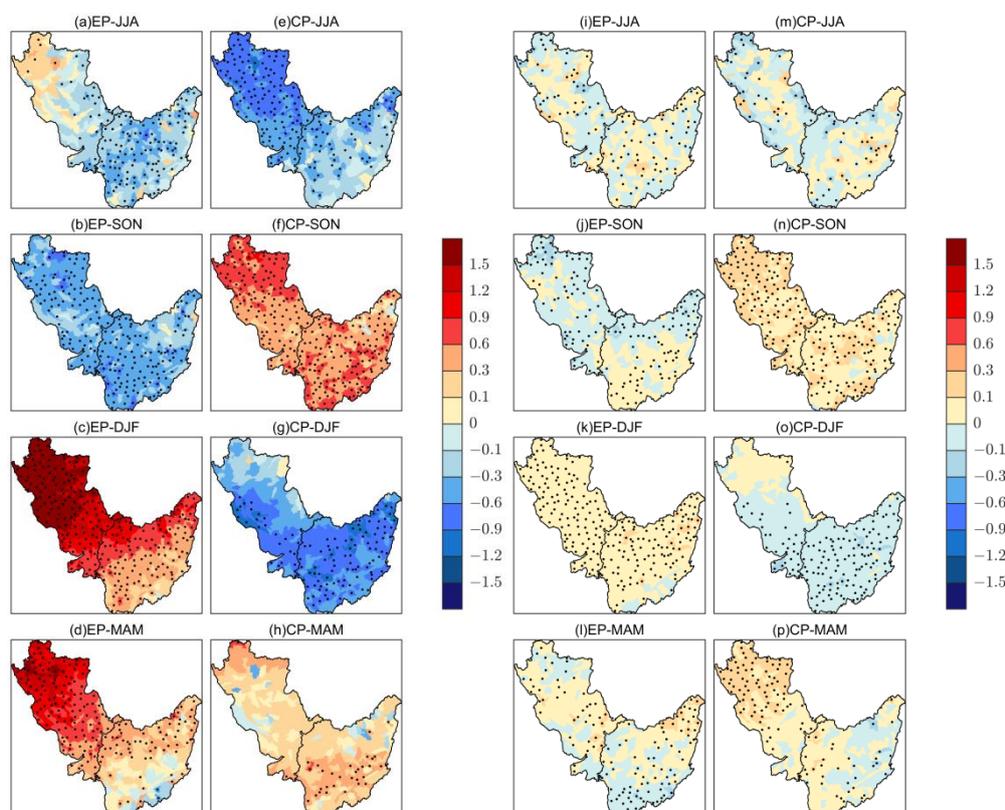
336
337 **Figure 7.** Composite results of annual average temperature (a and c, unit: °C) and evaporation (b and d, unit: mm day^{-1})
338 anomalies for EP-El Niño years (a and b) and CP-El Niño years (c and d), respectively. Stippling denotes anomalies
339 significantly different from zero at the 95% confidence level based on the Monte Carlo test.

340

341 Figure 8 showed seasonal patterns of temperature and evaporation anomalies during the CP- and EP-ENYs. In EP-ENYs,
342 significantly below-normal temperature occurred throughout the OTRB in the summer (Fig. 8a) and expanded to the entire
343 Corn Belt region in the autumn (Fig. 8b). In winter and spring, significantly positive temperature anomalies were shown in
344 the UMRB and most of OTRB (Figs. 8c and 8d). Corresponding to the seasonal temperature anomalies, evaporation varied
345 differently at different seasons (Figs. 8i-8l). In the summer and autumn, most sub-basins had negative evaporation anomalies
346 with decreased temperature (Figs. 8i and 8j); but in winter and spring, significantly above-normal evaporation occurred with
347 increased temperature in EP-ENYs (Figs. 8k and 8l). During CP-ENYs, the summer season was characterized by negative



348 temperature anomalies throughout the Corn Belt region, with the maximum anomalies up to -1.2 °C (Fig. 8e). Evaporation
349 anomalies became negative in most sub-basins (Fig. 8m). The temperature pattern was reversed in the autumn, with positive
350 temperature and evaporation anomalies over the entire region (Figs. 8f and 8n). In winter, temperature anomalies became
351 negative again (Fig. 8g), evaporation also reduced, with significantly negative anomalies in the OTRB and central and
352 southern UMRB (Fig. 8o). Temperature anomalies in spring were insignificant (Fig. 8h) most likely due to the short eight-
353 month duration of the CP-El Niño (Mo, 2010, Yu et al., 2010), although positive evaporation anomalies appeared in the
354 northern UMRB (Fig. 8p).



355
356 **Figure 8.** Composite results of seasonal temperature (a-h, unit: °C) and evaporation (i-p, unit: mm day^{-1}) anomalies in JJA (a
357 and i), SON (b and j), DJF (c and k), and MAM (d and l) in EP-El Niño years (a-d, i-l) and CP-El Niño years (e-h, m-p),
358 respectively. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte
359 Carlo test.

360



361 4 Discussion

362 4.1 Agricultural activity

363 This study focused on climate impact on water quality during EP-ENY and CP-ENY, we did not perform sensitivity
364 experiments on agriculture activities. In the SWAT model runs, the distribution of agriculture activities pattern (2008
365 Cropland Data Layers (CDL); USDA-NASS, 2016) was kept the same during El Niño and normal years. Corresponding to
366 the described agriculture activities, the corn growing areas, i.e., the southern UMRB and northern OTRB, usually produced
367 greater annual TN loads (>10 kg of N ha⁻¹) and TP loads (>1 kg of P ha⁻¹) (Fig. S5). During EP-ENYs, the nutrients were
368 largely reduced (>1 kg of N ha⁻¹ and >0.1 kg of P ha⁻¹) in the two corn-growing regions because of decreased precipitation
369 (<-0.1 mm day⁻¹) (Figs. 2a, 2b, and 5a). In CP-ENYs, the nutrient level increased in the southern UMRB (Figs. 2c and 2d)
370 since enhanced precipitation in CP-ENYs exacerbated the water quality in the area of heavy agriculture activities (Fig. 5c).
371 Water quality in the OTRB region showed different change patterns from the agriculture activities, i.e., TN and TP decreased
372 in the northern OTRB but increased in the southern OTRB (Figs. 2c and 2d). Such changes in water quality followed the
373 precipitation change in the OTRB (Fig. 5c), demonstrating that CP-El Niño-induced precipitation change played a more
374 important role in modulating water quality in OTRB.

375 On seasonal scales, changes in nutrients' magnitudes were stronger in spring and summer, especially in UMRB (Fig. 4).
376 This pattern was also related to the growth period of crops in the Corn Belt. The major crops here are corn and soybean,
377 which are often planted and fertilized in May and harvested in October. Hence, nutrients might be more likely to be removed
378 from the soil during spring and summer.

379 4.2 Equivalent impacts of CP- and EP-El Niño on water quality in specific watersheds

380 Section 3 discussed the impacts of El Niño on water quality at the outlets and sub-basin scales. At the outlets, EP-El Niño
381 had a much greater impact on TN and TP than CP-El Niño both at annual and seasonal time scales. But at the sub-basin scale,
382 CP- and EP-El Niño could have equivalent impacts on water quality in specific watersheds, predominantly in Iowa (IA),
383 Illinois (IL), Minnesota (MN), Wisconsin (WI), and Indiana (IN), which contributed the greatest amounts of nutrient change
384 to the whole basin loads (Fig. 2). Table 5 listed the top 10 HUC-8 sub-basins with the largest nutrient change during the two
385 types of El Niño years. In EP-ENYs, TN anomalies changed from 6.2 and 11.7 kg ha⁻¹ among the top 10 HUC-8 sub-basins;
386 while in CP-ENYs, TN anomalies changed from 5.6 and 9.3 kg ha⁻¹ (Table 5). These changes in TN during classic El Niño
387 and El Niño Modoki were comparable. Analysis of the top 10 HUC-8 sub-basins with the largest TP change during the EP-
388 and CP-El Niño illustrated similar results (not shown). These findings indicated that CP-El Niño could have comparable
389 impacts on TN and TP as EP-El Niño at the hot spot sub-basins although EP-El Niño had a much broader and longer impact
390 on water quality at the outlets.

391



392 **Table 5.** Information on the top 10 8-digit Hydrologic Unit Codes (HUC-8) sub-basins with greatest TN anomalies during
 393 EP- and CP-El Niño years.

El Niño Type	HUC-8	TN (kg ha ⁻¹)		Precipitation (mm yr ⁻¹)		Cropland Percentage (%)	States
		Average	Anomaly	Average	Anomaly		
EP	07080102	51.2	-11.7	912.9	-54.8	70.1	IA,MN
	07080201	33.7	-10.8	909.9	-51.1	73.5	IA,MN
	07090006	37.8	-10.4	935.2	-65.7	68.8	IL,WI
	07020011	28.3	-8.5	845.1	-84.0	78.6	MN
	07080202	27.4	-7.5	890.2	-32.9	72.1	IA,MN
	05120107	24.8	-7.2	1068.8	-102.2	73.8	IN
	07090007	31.2	-6.9	949.4	-87.6	78.9	IL
	07080204	33.0	-6.6	918.0	-69.4	76.0	IA
	07080208	27.0	-6.3	904.4	-62.1	59.9	IA
	07080209	25.6	-6.2	914.8	-94.9	58.4	IA
CP	07080103	26.2	9.3	948.0	142	70.2	IA
	07080208	27.0	9.1	904.4	135	59.9	IA
	07080206	21.9	9.1	923.1	128	63.3	IA
	07090007	31.2	7.3	949.4	62	78.9	IL
	07080203	24.2	6.8	855.4	124	68.1	IA,MN
	07080102	51.2	6.5	912.9	51	70.1	IA,MN
	07020007	16.5	6.1	778.0	69	73.2	MN
	07090006	37.8	5.9	935.2	18	68.8	IL,WI
	07060006	24.4	5.6	911.8	44	59.3	IA
	07130009	20.6	5.6	1005.0	124	80.9	IL

394

395 4.3 Future water quality change

396 Existing studies suggested that CP-El Niño episodes occurred more frequently in a warming climate (Yeh et al., 2009; Yu et
 397 al., 2010). We found that annual loads of TN and TP tend to increase in the Corn Belt region during CP-ENYs in the current
 398 climate. As CP-El Niño frequency increases in the future, TN and TP loads would likely increase over the Corn Belt region
 399 even under the same agricultural conditions, indicating a possible deterioration of water quality in this region when the
 400 climate warms.

401 The spatial patterns of TN and TP anomalies during CP-ENYs (Fig. S6) also suggested that specific watersheds,
 402 predominantly in southern UMRB and western OTRB, such as Iowa, Illinois, Wisconsin, Indiana, and Kentucky, will likely



403 experience the most increases of TN and TP loads in the future. Such information is critical to ensure proper decision-
404 making for watershed protection.

405 **4.4 Limitations and future work**

406 The model evaluation suggested that SWAT reasonably captured the hydrological and water quality behaviors in the Corn
407 Belt. However, this result could be influenced by the uncertainty of model simulations. For example, the limited number of
408 observation sites might bring uncertainties on the regional scales. The model was assessed by the best available observation
409 data, and the good agreement between the calculations and observations at 15 sites showed that the model reasonably
410 captured the changes in the water flow, TN, and TP. However, further assessment of the model is needed when more
411 observations are available.

412 The impacts of water temperature change on water quality during different El Niños were not analyzed in the study
413 because the influences of water temperature and runoff on water quality were coupled in the model. Numerical experiments
414 considering water temperature changes due to El Niños should be carried out and investigated in future studies.

415 In addition, as Chen et al. (2021) suggested that the CP-El Niño needed to be further classified into CP-I and CP-II types
416 due to the differences in sea surface temperature (SST) evolution patterns and climate impacts, the distribution of nutrients in
417 these two CP-El Niños might also differ and therefore are needed to be further studied in the future.

418 **5 Conclusions**

419 The impacts of EP- and CP-El Niños on water quality were investigated by using the SWAT model in the U.S. Corn Belt
420 region. Calibration and validation results indicated that the simulated streamflow and water quality generally agreed with
421 observations at most USGS gages. Then, the common features of the annual and seasonal loads of TN and TP for the EP-
422 and CP-El Niño events in the OTRB and UMRB were analyzed using a composite method based on the simulation results
423 during 1975–2016.

424 Annual composite results suggested that TN and TP loads decreased by 13.1% and 14.0% during EP-ENYs, respectively,
425 whereas they increased by 3.3% and 4.6% during CP-ENYs at the outlet of the OTRB, respectively. TN and TP also showed
426 a similar pattern at the outlet of the UMRB (18.5% and 19.8% reductions during EP-ENYs, and 5.7% and 4.4% increases
427 during CP-ENYs, respectively). Furthermore, more sub-basins showed negative annual anomalies of TN and TP during EP-
428 ENYs with maximum reductions up to -11.7 and -0.9 kg ha⁻¹, which were comparable to the normal year mean values (12.7
429 kg of N ha⁻¹ and 1.0 kg of P ha⁻¹), respectively, whereas they showed positive anomalies during CP-ENYs.

430 Seasonal composite results confirmed that water quality anomalies showed opposite patterns during EP-ENYs and CP-
431 ENYs and the changes in the water quality matched the ENSO phases at the outlet of the OTRB. Maximum reduction or
432 increase of the nutrients during EP-ENYs or CP-ENYs, respectively, occurred in winter, the peak season of El Niño. At the
433 outlet of the UMRB (corn-growing region), TN and TP anomalies were also influenced by agriculture activities and became



434 greater in spring and summer during both EP-ENYs and CP-ENYs, in consistent with the distribution of rainfall changes in
435 the basin. These results suggested that small changes in climate variables such as precipitation in the growing season could
436 have greater impacts on water quality in the UMRB during El Niño events.

437 Our analysis also found that at the outlets of UMRB and OTRB, EP-El Niño had a much greater impact on TN and TP
438 than CP-El Niño at both annual and seasonal time scales; but CP-El Niño could have comparable impacts on water quality as
439 EP-El Niño at specific watersheds or hot spots, predominantly in Iowa, Illinois, Minnesota, Wisconsin, and Indiana, which
440 contributed the greatest amounts of nutrient change to the whole basin loads.

441 Examination of the climate factors/processes on water quality change indicated that El Niño-induced precipitation and
442 temperature changes altered runoff and evaporation, and thus TN and TP in both UMRB and OTRB on annual and seasonal
443 time scales, as well as at the outlet and sub-basin scales. It is also found that nutrient levels were largely determined by
444 precipitation through runoff during both EP- and CP-ENYs, especially at the outlet, as the precipitation was a major source
445 of runoff, and nitrogen and phosphorus components were transported by runoff. At the sub-basin scale, water quality was
446 affected by the combination of precipitation and agricultural activities, especially in the UMRB during the growing season.

447 In the future when the climate continues to warm, the CP-El Niño episode is projected to occur more frequently, TN and
448 TP loads might increase in the Corn Belt region even under the same agricultural conditions, while water quality would
449 generally get better in EP-El Niño years. The findings from this study may help ensure proper decision-making for
450 watershed protection and possible ways to address anticipated water quality change associated with El Niño events.

451

452 **Data availability.** Available upon request.

453 **Author Contributions.** W.L. and P.C. designed the research; P.C. and K.H. performed the data collection and model
454 calculation; W.L. and P.C. contributed to the interpretation of the results; P.C. and W.L. wrote the manuscript.

455 **Competing Interests.** The authors declare no competing interests.

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465



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