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

Pan Chen[1,2], Wenhong Li[2,*] and Keqi He[2]

[1] College of Water Resources Science and Engineering, Taiyuan University of Technology, Taiyuan 030024, China
[2] Earth and Climate Sciences, Nicholas School of the Environment, Duke University, NC 27708, USA

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

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

1 Introduction

The Corn Belt region of the United States (U.S.) primarily includes the Ohio-Tennessee River Basin (OTRB) and the Upper Mississippi River Basin (UMRB) (Kellner and Niyogi, 2015; Panagopoulos et al., 2017; Ting et al., 2021). The Corn Belt is a very important area of the agricultural activity of the country, as 75% of the corn and 60% of the soybean produced in the U.S. are grown in the region (Thaler et al., 2021). The region's agricultural activities such as fertilizers contribute to the increase of nitrogen and phosphorus levels, which are responsible for the Gulf of Mexico hypoxic zone (Panagopoulos et al., 2014, 2015; Rabalais et al., 2007). The required nutrient reduction of the Corn Belt to decrease hypoxia is the highest among all regions in the Mississippi River Basin (Mississippi River/Gulf of Mexico Watershed Nutrient Task Force, 2011). Hence, water quality changes in the Corn Belt region have been receiving considerable attention.

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

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

In this study, we used the Soil and Water Assessment Tool (SWAT) model to estimate the water quality changes due to EP- and CP-El Niños over the Corn Belt. The SWAT model is widely used to assess climate change and alternative land use/land management scenarios on runoff and nutrients in a basin (Afonso de Oliveira Serrão et al., 2022; Chaplot et al., 2004; Chen et al., 2021; Johnson et al., 2015; Vaché et al., 2002; Zhang et al., 2020). The detailed objectives of this study were to 1) analyze the impacts of the two types of El Niño on water quality in the Corn Belt region, and 2) identify the main climate factors that affect the change in water quality due to these El Niños. Water quality change associated with future El Niño change was also discussed. Such information is particularly important to enable decision-makers to take timely actions to reduce nutrient loading under climate change.

2 Data and methods

2.1 Data

The study area includes two large basins in the U.S., namely, OTRB (528,000 km^2) and UMRB (492,000 km^2), as shown in Fig. 1. OTRB comprises a significant portion of Pennsylvania, Ohio, West Virginia, Indiana, Illinois, Kentucky, and Tennessee (Fig. 1a). The amount of annual rainfall in OTRB was high with an average of nearly 1200 mm during 1975–2016. OTRB's slopes are steep, especially in the forested Tennessee basin with slopes greater than 5% in most (60%) of the area. The primary land use types are 50% forest, 20% cropland, and 15% pasture (Fig. 1b). The cropland is mainly grown with corn, soybean, and wheat (Santhi et al., 2006). UMRB mainly includes five states: Iowa, Illinois, Missouri, Wisconsin, and Minnesota (Fig. 1a). The mean annual value of rainfall in UMRB during 1975–2016 was 900 mm. UMRB is relatively flat and most of the basin (75%) has a slope lower than 5%. Cropland is the major land use type of the basin (50%) and is primarily grown with corn and soybean (Fig. 1b).

The weather data, including precipitation and temperature, were obtained from 2,242 National Weather Service (NWS) stations in the study area. The 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 EP-El Niño events (1976–1977, 1979–1980, 1982–1983, 1986–1987, 1987–1988, 1991–1992, 1997–1998, 2006–2007, and 2015–2016) and six CP-El Niño events (1977–1978, 1990–1991, 1994–1995, 2002–2003, 2004–2005, and 2009–2010) occurred during the study period (1975–2016).

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

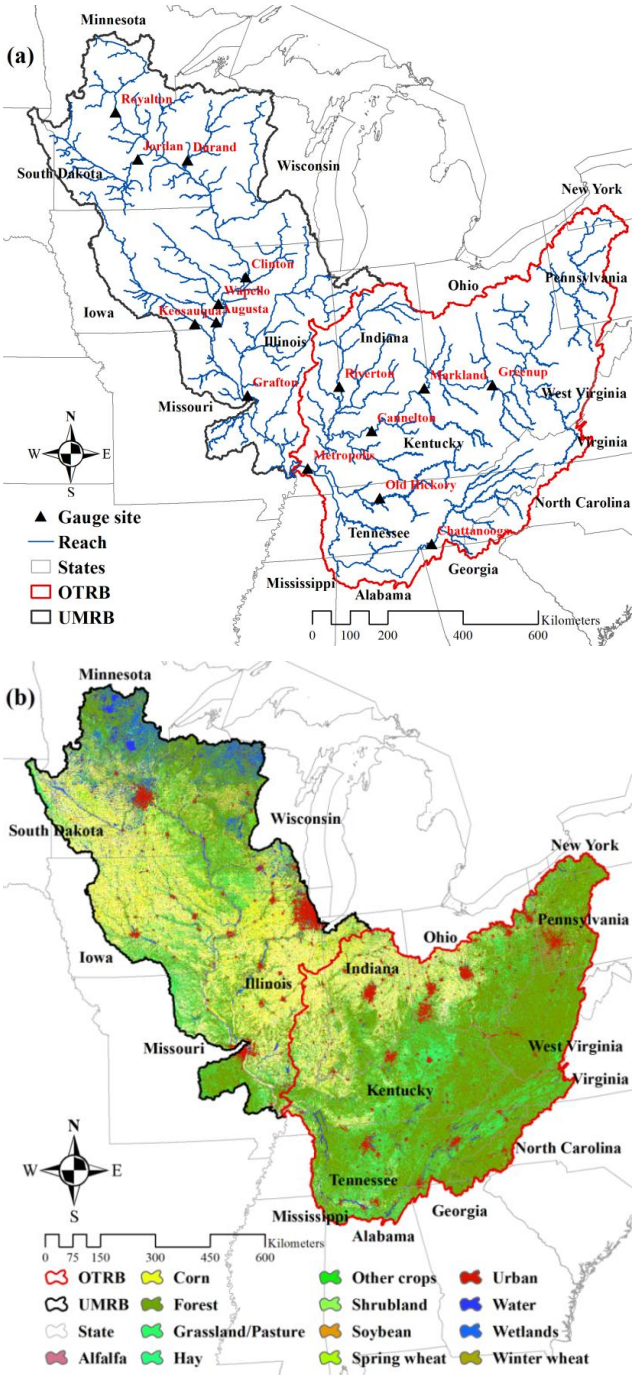


Figure 1. The Ohio-Tennessee River Basin (OTRB) and Upper Mississippi River Basin (UMRB) (a) available United States Geological Survey (USGS) gage sites (black triangles), reaches (light blue), and the two watersheds (heavy red: OTRB, heavy dark: UMRB) and (b) land use/land cover.

Table 1. Available periods of measured streamflow, total suspended sediment (TSS), total nitrogen (TN), and total 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	OTRB	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		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	UMRB	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		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

Statistical significance of precipitation, temperature, runoff, evaporation, TN, or TP anomalies in El Niño years was tested by the Monte Carlo method (Mo, 2010). The underlying concept of the method is to use randomness to solve problems that might be deterministic in principle (Wilks, 1995). Taking TN as an example, in order to test whether TN anomalies in EP-El Niño years were significantly different from those in normal years, we first composited (i.e., averaged) TN anomalies for the nine EP-El Niño years. The composite analysis is a useful technique to determine some of the basic structural characteristics of a climatological phenomenon, such as El Niño which occur over time. We then randomly selected nine years out of 1975-

2016 (i.e., keeping the same number of years as the EP-El Niño years) and averaged/composited TN anomalies for the nine randomized years as the first sample. The process was repeated 500 times. These composite samples were used to generate a distribution corresponding to the null hypothesis, against which we could evaluate whether TN anomalies during EP-El Niño were significantly different from those in normal years at a 95% confidence level. Similarly, significance levels of the composite results of precipitation, temperature, runoff, evaporation, and TP anomalies in EP-El Niño years could be determined. Such a method has been widely used in climate-related studies (Laken and Čalogović, 2013; Mo, 2010; Sanchez and Karnauskas, 2021) due to its robustness.

2.2 SWAT model

2.2.1 Model description

SWAT was developed by the U.S. Department of Agriculture Agricultural Research Services (Arnold et al., 1998). It has been widely used in assessing the effects of climate and land use change on hydrological processes, sediment, and nutrients 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-basins, which are further divided into hydrological response units (HRUs) (Gassman et al., 2007; Neitsch et al., 2011; Williams et al., 2008). Runoff, sediment, and nutrient loads are simulated for each HRU and then aggregated for the sub-basins (Chen et al., 2021; Gassman et al., 2007; Neitsch et al., 2011). Thus, the spatial resolution of the model is measured by the number of HRUs and sub-basins. The pollution situation of each sub-basin during different El Niño years could be obtained from this model. The model was calculated on a daily time scale, and the results were analyzed on a monthly time scale.

2.2.2 Model set-up and calibration

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

SWAT Calibration and Uncertainty Programs (SWAT-CUP) with Sequential Uncertainty Fitting (SUFI-2) algorithm was selected in this large-scale study to complete the calibration of the SWAT model (Abbaspour et al., 2012). The parameters of water flow and water quality in OTRB and UMRB were selected based on a manual experimentation with SWAT parameters and a literature review (Chen et al., 2021; Panagopoulos et al., 2014, 2015; Yen et al., 2016). The calibration steps followed

a recent study by Chen et al. (2021). The final parameters were shown in Table S1 and S2 in the supplementary materials. The calibration results indicated that the SWAT model could rationally capture the observation (see Section 2.2.3).

2.2.3 Model performance

Overall, SWAT simulated the water flow of the OTRB and UMRB reasonably well in both calibration (1997–2016) and validation (1975–1996) periods (Table 2). The coefficient of determination (R^2) and Nash-Sutcliffe efficiency (*NSE*) values were larger than 0.5 for almost all the USGS gages except Chattanooga, and percent bias (*PBIAS*) values were all acceptable during the calibration periods (1997–2016) based on the ≤ ±25% deviation criterion (Moriasi et al., 2015). The validation results also showed acceptable static values and in some cases were even better, such as Chattanooga and Grafton. Figures S1 and S2 in the supplementary materials further demonstrated good agreement between the calculated and observed streamflow across OTRB and UMRB; particularly, most of the peaks and recession limbs were well matched in the simulations.

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

Table 2. Monthly streamflow calibration and validation statistics.

Gauge Site	Calibration (1997−2016)			Validation (1975−1996)		
	R^2	*NSE*	*PBIAS*	R^2	*NSE*	*PBIAS*
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

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

3 Results

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

3.1.1 Annual composite

(1) Water quality at the outlet

Figure 2 showed the nutrient change during EP- and CP-ENYs at the outlet of OTRB and UMRB. Annual loads of TN and TP decreased during the EP-ENYs, while the pattern reversed during the CP-ENYs in the U.S. Corn Belt region. Specifically, 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 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 CP-ENYs at the outlet of the OTRB, respectively (Fig. 2). TN and TP at the outlet of the UMRB showed a similar pattern as that of the OTRB, decreasing (increasing) by 18.5% (5.7%) and 19.8% (4.4%) in EP (CP)-ENYs. Furthermore, EP-El Niños had a much greater impact on water quality than CP-El Niños at the outlets of OTRB and UMRB. The magnitudes of variation in both TN and TP during the EP-ENYs were three to four times greater than those during the CP-ENYs (Fig. 2).

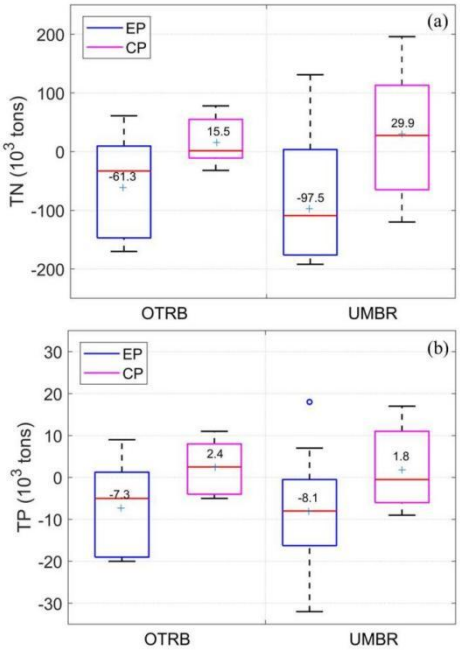


Figure 2. Box plots of annual (a) TN and (b) TP anomalies (unit: 10^3 tons) at the outlets of the OTRB and UMRB during EP-El Niño years and CP-El Niño years, respectively. The green plus (+), red solid horizontal line, box, and whisker ends indicate the mean, median, 25th and 75th percentile, and the 10th and 90th percentile, respectively. The data points outside the ranges are shown in hollow dots.

(2) Water quality at the sub-basin scale

We analyzed water quality change on the sub-basin scale during EP-ENYs and CP-ENYs, respectively (Fig. 3). Anomalous patterns of water quality associated with El Niño events within the Corn Belt varied in space. Clearly, more sub-basins showed negative anomalies of TN and TP during EP-ENYs, whereas more sub-basins showed positive anomalies during CP-ENYs (Fig. 3). Specifically, during EP-ENYs, significant below-average TN and TP were found almost in the whole OTRB and UMRB with maximum reductions of TN and TP up to −11.7 and −0.9 kg ha^{-1}, respectively, which were of similar 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. 3a and 3b). During CP-ENYs, positive anomalies mainly occurred throughout the southern OTRB and UMRB. In the northern UMRB and OTRB, about 42.4% and 41.7% of sub-basins tended to have below-average TN and TP (Figs. 3c and 3d). These patterns coincided with the TN and TP changes at the outlets of the two basins, which could also explain why greater changes in TN and TP occurred in EP-ENYs than in CP-ENYs at the outlets of the OTRB and UMRB (Fig. 2).

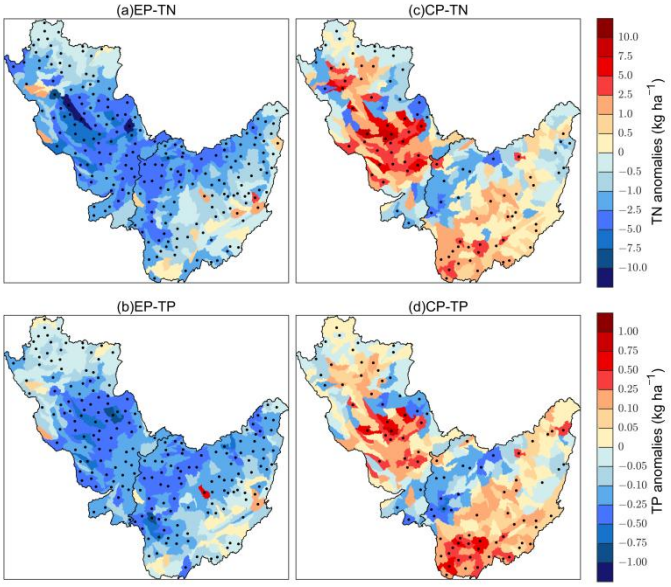


Figure 3. Composite results of annual TN and TP anomalies (unit: kg ha^{-1}) in EP-El Niño years (a and b) and in CP-El Niño years (c and d) during the period of 1975–2016. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

3.1.2 Seasonal composite

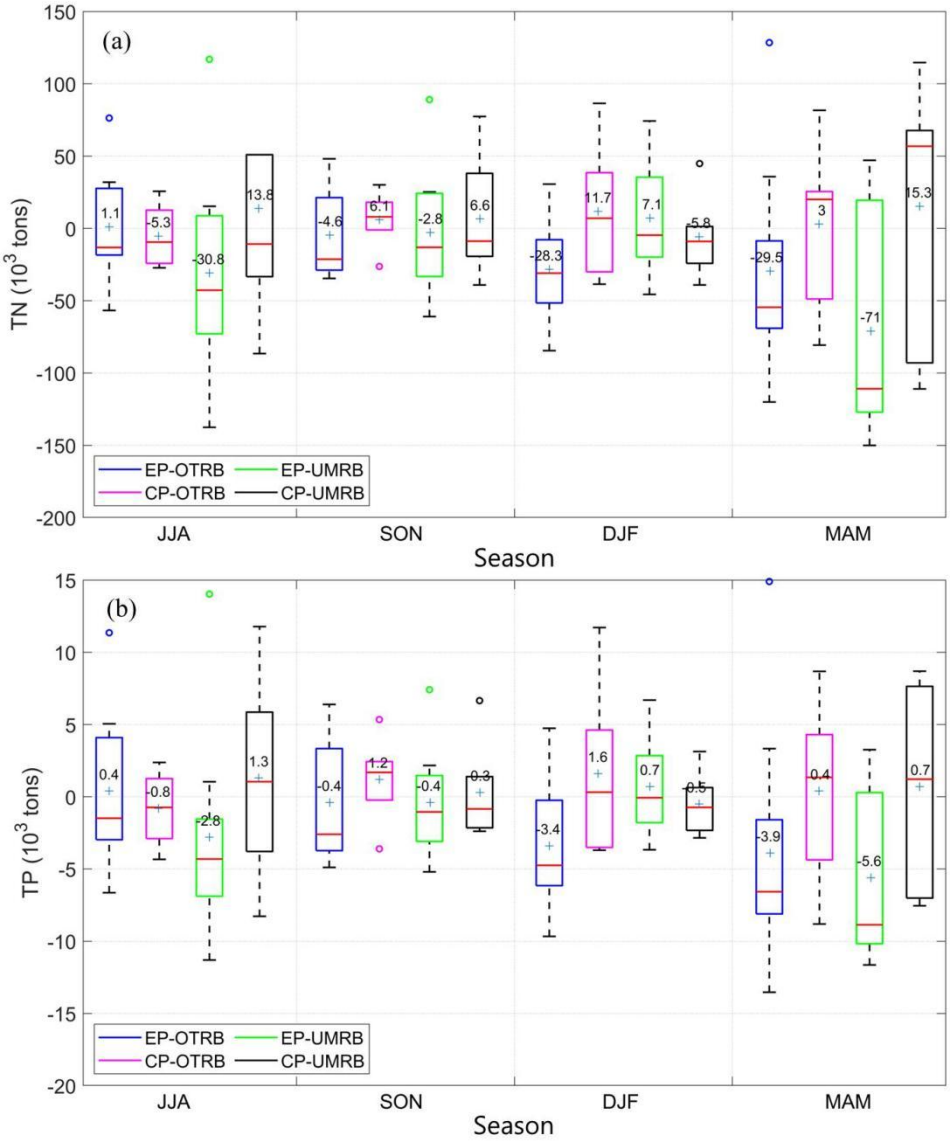
(1) Water quality at the outlet

Figure 4a showed TN anomalies at the OTRB and UMRB outlets in each season during EP- and CP-ENYs. At the outlet of the OTRB, seasonal anomalies of the water quality reached the maximum when ENSO signals were the strongest (Fig. 4a).

El Niño usually developed in boreal summer (June-August, JJA) and autumn (September-November, SON), peaked in winter (December of the current year and January and February of the following year, DJF), and decayed in spring (March-May, MAM) (Trenberth, 1997; Li et al., 2011). Maximum changes in TN occurred in the winter and spring seasons during EP-ENYs (decreased by 28.3×10^3 and 29.5×10^3 metric tons (hereafter ton), respectively, Fig. 4a). During CP-ENYs, TN 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) compared to that in the normal years (Fig. 4a). Seasonal TN anomalies at the outlet of the UMRB were different from those of the OTRB. Figure 4a showed that TN decreased by 71.0×10^3 and 30.8×10^3 tons, respectively, in the spring and summer seasons during EP-ENYs; but in winter and autumn, TN did not change much compared to normal years. Similarly, during CP-ENYs, TN anomalies were greater in spring and summer although TN anomalies became positive during CP-ENYs, different from TN changes during EP-ENYs (Fig. 4a).

Figure 4b demonstrated that the seasonal changes of TP during EP- and CP-ENYs were similar to those of TN in both OTRB and UMRB during EP-ENYs. At the OTRB the magnitudes of TP reduction in boreal winter and spring were greater than those in the summer and autumn seasons during EP-ENYs. During CP-ENYs, TP anomalies were greater in both autumn and winter (Fig. 4b). This phenomenon was probably related to the different duration of the two types of El Niño. The mean duration of EP-El Niño was about 15 months (Mo, 2010), its impact on water quality could last into the following spring; while El Niño Modoki usually lasted for about eight months (Mo, 2010, Yu et al., 2010), therefore the impacts of the CP-El Niño on water quality usually ended in the winter. Besides, at the UMRB maximum changes in TP occurred in the spring and summer seasons during both EP- and CP-ENYs.

In summary, more seasons showed negative anomalies of water quality during EP-ENYs, whereas more seasons showed positive anomalies during CP-ENYs (Fig. 4). Consequently, the annual TN and TP anomalies over the Corn Belt region showed the opposite pattern during EP-ENYs and CP-ENYs (Fig.2).



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230 **Figure 4.** Same as Fig. 2 but for seasonal scales, i.e., summer (June-August, JJA), autumn (September-November,
231 SON), winter (December of the current year and January and February of the following year, DJF), and spring (March-
232 May, MAM).

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

234 Water quality at the Corn Belt can vary in different locations/sub-basins and change between seasons. Figure 5 showed
235 spatial patterns of TN and TP anomalies at the OTRB and UMRB in each season during the CP- and EP-ENYs, separately.
236 EP-El Niño was characterized by negative TN and TP anomalies over most of the OTRB and UMRB for all seasons (Figs.

5a–5d and 5i–5l). Significant below-average TN and TP occurred in almost all of the UMRB and the eastern OTRB in the summer when EP-El Niño was developing in the tropical Pacific, with maximum reductions up to −4.5 and −0.45 kg ha^{-1}, respectively (Figs. 5a and 5i). The negative water quality anomalies moved to the southern UMRB and northern OTRB in the autumn (Figs. 5b and 5j). These negative anomalies further moved to the whole OTRB when EP-El Niño was mature in the winter (Figs. 5c and 5k). This result generally agreed with previous findings of the El Niño impacts on the precipitation in the study area, which indicated that the precipitation in the Ohio Valley was sensitive to El Niño events and showed negative precipitation anomalies at EP-ENYs (Mo, 2010; Twine et al., 2005). In spring, severe TN and TP deficits were most apparent almost all over the Corn Belt area (Figs. 5d and 5l).

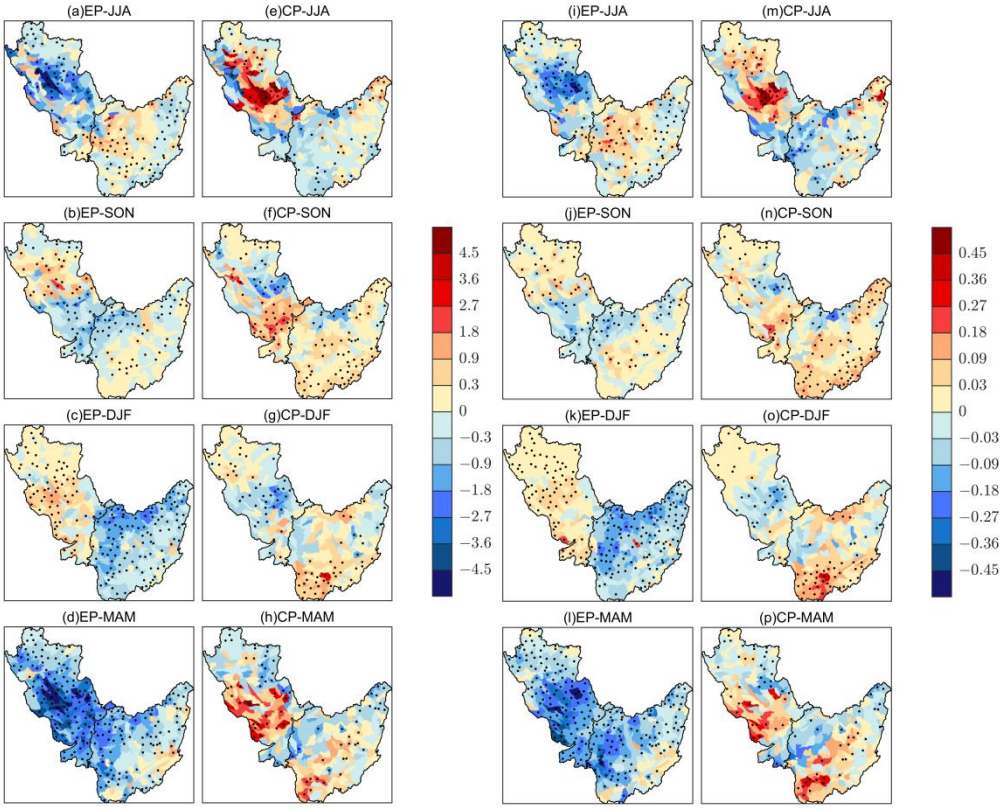

Figure 5. Seasonal composite of (a-h) TN and (i-p) TP anomalies (unit: kg ha^{-1}) in summer (JJA) (a and i), autumn (SON) (b and j), winter (DJF) (c and k), and spring (MAM) (d and l) in EP-El Niño years during the period of 1975–2016; (e-h) and (m-p) are the same as (a-d) and (i-l) but for CP-El Niño years. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

In contrast, during CP-ENYs, positive TN and TP anomalies were scattered in most of the Corn Belt region with the highest TN and TP anomalies increase up to 4.5 and 0.45 kg ha^{-1}, respectively (Figs. 5e–5h and 5m–5p). In summer, northern UMRB and eastern OTRB were characterized by above-normal water quality (Figs. 5e and 5m). The positive TN and TP anomalies moved to the southern UMRB and whole OTRB in the autumn (Figs. 5f and 5n). In winter, these positive

anomalies were concentrated in the southern OTRB and northern UMRB (Figs. 5g and 5o). In the spring, abnormally high TN and TP mainly occurred in the southern part of the UMRB while most of the positive and negative anomalies in the OTRB region were insignificant (Figs. 5h and 5p).

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

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

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

3.2.1 Precipitation

Decreased precipitation in EP-ENYs was one of the important reasons that improved water quality in the Corn Belt region. This finding could be shown in the spatial patterns of annual precipitation (Fig. 6a), and TN and TP anomalies (Figs. 3a and 3b). For example, significantly below-normal precipitation occurred in much of the OTRB and UMRB (278 out of 283 sub-basins) during EP-ENYs (Fig. 6a), thus TN and TP were reduced in most parts of the Corn Belt (Figs. 3a and 3b). During CP-ENYs, precipitation anomalies became positive in 226 out of 283 sub-basins (Fig. 6c). Correspondingly, TN and TP concentrations were elevated compared to normal years (Figs. 3c and 3d). Spatially, there were more sub-basins (98.2%) and larger area with negative precipitation anomalies in EP-ENYs than positive anomalies (79.9% sub-basins) during CP-ENYs in the Corn Belt, indicating that EP-El Niño tended to have a much wider impact on precipitation and thus water quality than CP-El Niño.

To better understand how precipitation affected water quality in the Corn Belt, variations of annual runoff during EP-ENYs and CP-ENYs were also discussed because nitrogen and phosphorus were transported by runoff (Neitsch et al., 2011), and precipitation was a very important source of runoff (Gassman et al., 2007). The overall patterns of runoff anomalies (Figs. 6b and 6d) were similar to those of precipitation anomalies (Figs. 6a and 6c) during the two types of El Niño with pattern correlations being 0.79 at OTRB and 0.75 at UMRB in EP-ENYs, 0.96 and 0.77 in CP-ENYs, respectively. Specifically, during EP-ENYs, negative annual runoff anomalies occurred in most of the Corn Belt region (Fig. 6b), resulting in reduced TN and TP compared to normal years (Figs. 3a and 3b). In contrast, in CP-ENYs, positive runoff anomalies were mainly concentrated in southern OTRB and UMRB (Fig. 6d), more nutrients were thus carried out by the runoff associated with above-normal precipitation in the area (Figs. 3c and 3d). We also noticed that EP-El Niño had a wider influence on runoff than CP-El Niño, as 71 more sub-basins (259 vs 188) and larger areas with significant runoff anomalies were found in EP-ENYs than in CP-ENYs (Figs. 6b and 6d), generally consistent with the spatial patterns of precipitation

changes due to the El Niños (Figs. 6a and 6c). The phenomena could partly explain why EP-El Niños tended to have greater impacts on the water quality than CP-El Niños at the outlets of the OTRB and UMRB.

Some differences also existed among precipitation, runoff, and nutrients. For precipitation and runoff, annual precipitation anomalies were greater than runoff anomalies in the El Niño years. For example, precipitation anomalies of many sub-basins in northern UMRB were stronger than −0.2 mm day^{-1} during EP-ENYs, but the magnitudes of runoff anomalies were weaker than −0.2 mm day^{-1} (Figs. 6a and 6b). During CP-ENYs, most annual precipitation and runoff anomalies were positive, but the magnitudes of precipitation anomalies were generally higher than that of runoff anomalies (Figs. 6c and 6d). These findings suggested that the impact of El Niño on runoff was weakened by the land surface hydrological process. For runoff and nutrients, changes of runoff in El Niño years were greater in the OTRB than in the UMRB (Figs. 6b and 6d), whereas changes of TN and TP were smaller in the OTRB than in the UMRB during both EP-ENYs and CP-ENYs (Fig. 3). The discrepancies between annual runoff and nutrient anomalies in the OTRB and UMRB may be related to the presence of more cropland in the UMRB where more fertilizers were used (Chen et al., 2021, see Section 4.1 for details), hence a relatively small change of runoff due to El Niño-induced precipitation could lead to large TN and TP variations.

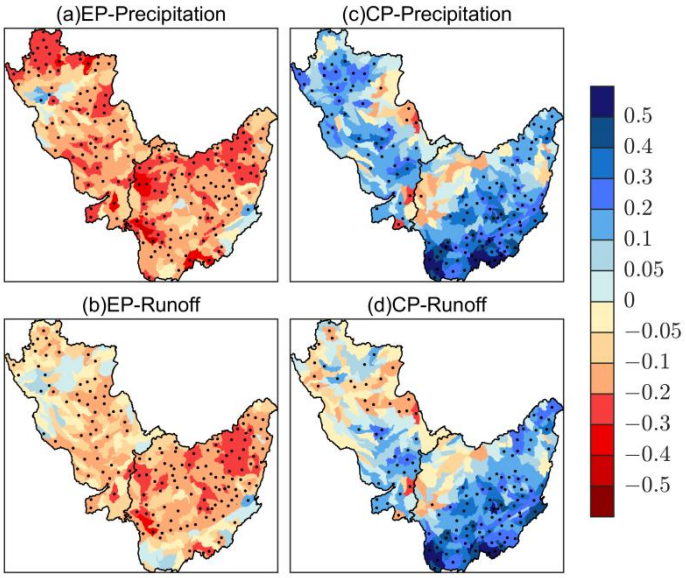

Figure 6. Composite of annual precipitation and runoff anomalies (unit: mm day^{-1}) for the two types of El Niño during the period of 1975–2016 (a and b) in EP-El Niño years and (c and d) in CP-El Niño years. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

Seasonal patterns of precipitation and runoff anomalies (Fig. 7) further proved the impacts of precipitation on water quality through runoff during EP-ENYs and CP-ENYs. In summer, significantly below-normal precipitation occurred in almost all of the UMRB when EP-El Niño was developing in the tropical Pacific, with maximum reductions of precipitation up to −0.9 mm day^{-1} in the region (Fig. 7a). The runoff anomaly pattern was much the same as that of precipitation, but in a

weaker magnitude—less than −0.6 mm day^{-1} in UMRB (Fig. 7i). At the same time, TN and TP decreased in the area (Figs. 5a and 5i) because fewer nutrients were carried out by the reduced runoff in summer (Fig. 7i). Similar change patterns of precipitation (Figs. 7b-7d), runoff (Figs. 7j-7l), and nutrients (Figs. 5b-5d and 5j-5l) could also be found in other seasons during EP-ENYs. We noticed that negative runoff and precipitation anomalies reached their maximum in spring throughout the Corn Belt, leading to better water quality in the region compared to the normal years. CP-El Niño events caused the opposite patterns of seasonal precipitation (Figs. 7e-7h) and runoff anomalies (Figs. 7m-7p) in the Corn Belt region; TN and TP thus increased in central and southern UMRB in the spring, summer, and autumn, and most of OTRB from autumn to winter (Figs. 5e-5h and 5m-5p). Some differences also existed between precipitation and water quality in the UMRB, especially in spring and summer. During these two seasons, the variation of the seasonal precipitation at each sub-basin was relatively uniform in both EP-ENYs (Figs. 7a and 7d) and CP-ENYs (Figs. 7e and 7h), but nutrient variations were high in some sub-basins of the UMRB (Figs. 5a, 5d, 5e, 5h, 5i, 5l, 5m, and 5p). This phenomenon suggested that water quality was also influenced by local factors besides climate variables associated with El Niños.

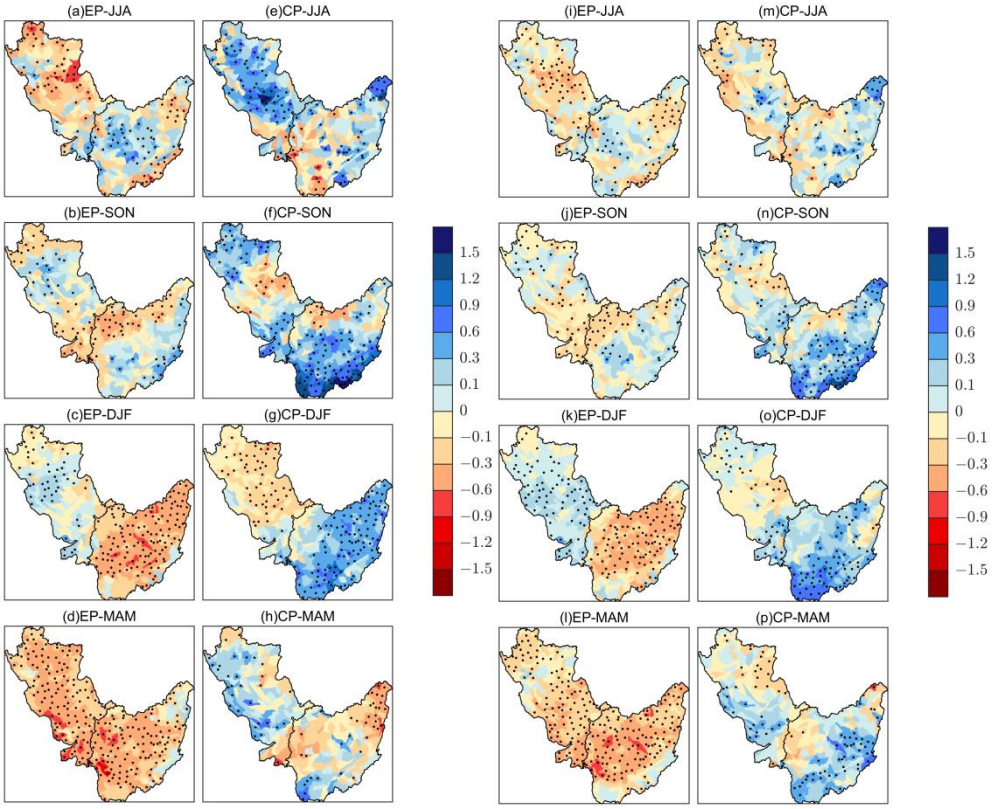

Figure 7. Composite of seasonal (a-h) precipitation and (i-p) runoff anomalies (unit: mm day^{-1}) in JJA (a and i), SON (b and j), DJF (c and k), and MAM (d and l) in EP-El Niño years during the period of 1975−2016; (e-h) and (m-p) are the same as (a-d) and (i-l) but for CP-El Niño years. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

3.2.2 Temperature

Temperature changes have previously been found to affect water quality by changing the evaporation process of the water cycle (Neitsch et al., 2011 Sun et al., 2011). Thus, how evaporation varied associated with temperature change during El Niños in the Corn-Belt region was analyzed. Compared to normal years, the annual temperature increased over the UMRB, but decreased in the OTRB, especially in the southern OTRB during EP-ENYs (Fig. 8a). Evaporation slightly increased in most of OTRB (Fig. 8b), which did not share the same pattern with temperature change on the annual time scale (Fig. 8a). This might be due to the fact that temperature directly affected potential evapotranspiration (Neitsch et al., 2011), the ability of the atmosphere to remove water from the surface through both evaporation and transpiration, but the actual evaporation/evapotranspiration was also related to other variables such as the amount of water available for evaporation besides temperature. Enhanced evaporation further reduced runoff (Bales et al., 2017). Thus, decreased precipitation (Fig. 6a) and enhanced evaporation (Fig. 8b) during the EP-ENYs would facilitate runoff decline and cause a much wider impact of EP-El Niño events on water quality in the Corn Belt region (Fig. 3). During CP-ENYs, temperature decreased insignificantly in most of the Corn Belt region (Fig. 8c). Evaporation increased in more sub-basins over the UMRB (Fig. 7d). The enhanced evaporation (Fig. 8d) tended to offset, to some extent, the impact of higher than normal precipitation (Fig. 6c) on water quality during CP-ENYs.

The impacts of the two climate factors, precipitation and temperature (through evaporation), on runoff were compared. During the El Niño years, the magnitude of annual precipitation change was often greater than 0.1 mm day^{-1} (Figs. 6a and 6c) while most of the annual evaporation varied between -0.05 and 0.05 mm day^{-1} (Figs. 8b and 8d). This suggested that both precipitation and evaporation influence water quality through runoff, but precipitation seemed to play a more important role in altering water quality over the Corn Belt region during El Niño years.

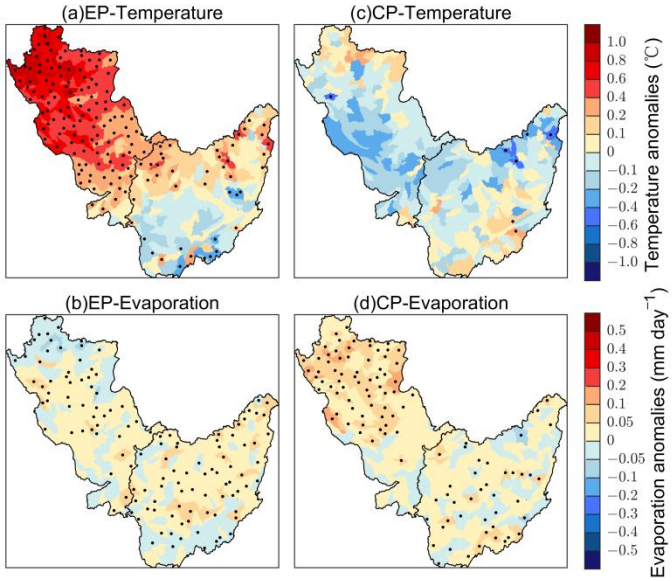

Figure 8. Composite results of annual average temperature (a and c, unit: ℃) and evaporation (b and d, unit: mm day^{-1}) anomalies for EP-El Niño years (a and b) and CP-El Niño years (c and d), respectively. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

Figure 9 showed seasonal patterns of temperature and evaporation anomalies during the CP- and EP-ENYs. In EP-ENYs, significantly below-normal temperature occurred throughout the OTRB in the summer (Fig. 9a) and expanded to the entire Corn Belt region in the autumn (Fig. 9b). In winter and spring, significantly positive temperature anomalies were shown in the UMRB and most of OTRB (Figs. 9c and 9d). Corresponding to the seasonal temperature anomalies, evaporation varied differently at different seasons (Figs. 9i-9l). In the summer and autumn, most sub-basins had negative evaporation anomalies with decreased temperature (Figs. 9i and 9j); but in winter and spring, significantly above-normal evaporation occurred with increased temperature in EP-ENYs (Figs. 9k and 9l). During CP-ENYs, the summer season was characterized by negative temperature anomalies throughout the Corn Belt region, with the maximum anomalies up to −1.2 °C (Fig. 9e). Evaporation anomalies became negative in most sub-basins (Fig. 9m). The temperature pattern was reversed in the autumn, with positive temperature and evaporation anomalies over the entire region (Figs. 9f and 9n). In winter, temperature anomalies became negative again (Fig. 9g), evaporation also reduced, with significantly negative anomalies in the OTRB and central and southern UMRB (Fig. 9o). Temperature anomalies in spring were insignificant (Fig. 9h) most likely due to the short eight-month duration of the CP-El Niño (Mo, 2010, Yu et al., 2010), although positive evaporation anomalies appeared in the northern UMRB (Fig. 9p).

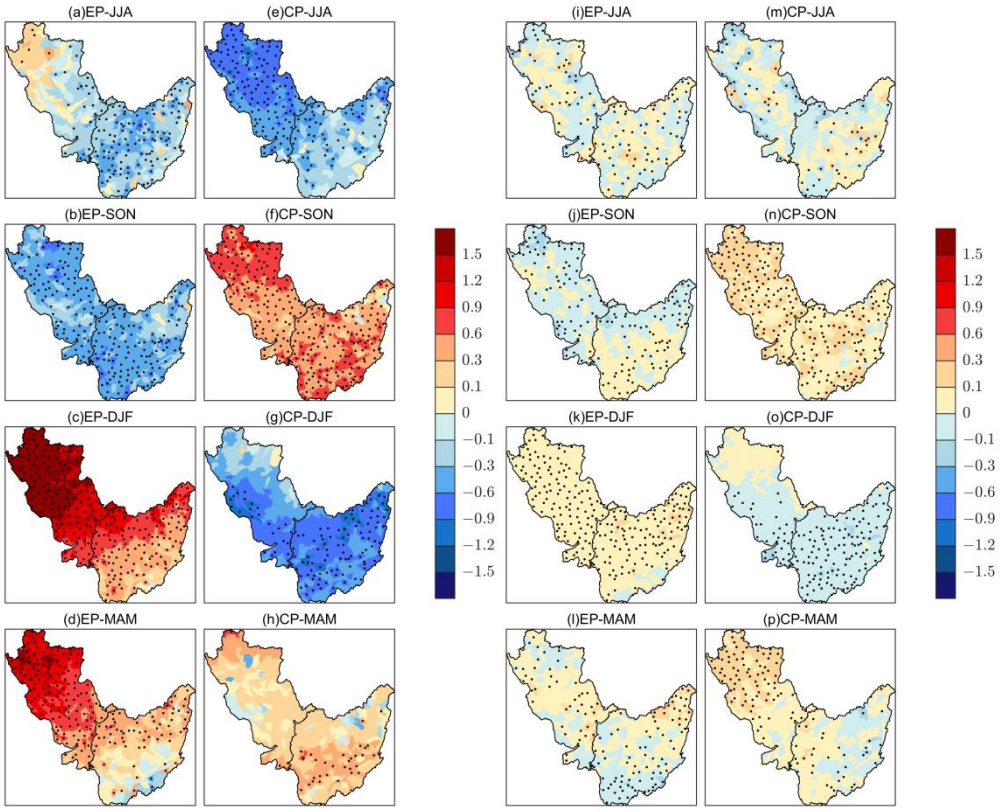

Figure 9. Composite results of seasonal temperature (a-h, unit: ℃) and evaporation (i-p, unit: mm day^{-1}) anomalies in JJA (a and i), SON (b and j), DJF (c and k), and MAM (d and l) in EP-El Niño years (a-d, i-l) and CP-El Niño years (e-h, m-p), respectively. Stippling denotes anomalies significantly different from zero at the 95% confidence level based on the Monte Carlo test.

4 Discussion

4.1 Agricultural activity

This study focused on climate impact on water quality during EP-ENY and CP-ENY, we did not perform sensitivity experiments on agriculture activities. In the SWAT model runs, the distribution of agriculture activities pattern (2008 Cropland Data Layers (CDL); USDA-NASS, 2016) was kept the same during El Niño and normal years. Corresponding to the described agriculture activities, the corn growing areas, i.e., the southern UMRB and northern OTRB, usually produced 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 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 (< −0.1 mm day^{-1}) (Figs. 3a, 3b, and 6a). In CP-ENYs, the nutrient level increased in the southern UMRB (Figs. 3c and 3d) since enhanced precipitation in CP-ENYs exacerbated the water quality in the area of heavy agriculture activities (Fig. 6c).

Water quality in the OTRB region showed different change patterns from the agriculture activities, i.e., TN and TP decreased in the northern OTRB but increased in the southern OTRB (Figs. 3c and 3d). Such changes in water quality followed the precipitation change in the OTRB (Fig. 6c), demonstrating that CP-El Niño-induced precipitation change played a more important role in modulating water quality in OTRB.

On seasonal scales, changes in nutrients' magnitudes were stronger in spring and summer, especially in UMRB (Fig. 5). The heavy loading of nutrients was related to the agriculture activities during the growth period of crops in the Corn Belt. The major crops here are corn and soybean, which are often planted and fertilized in May and harvested in October (Chiang et al., 2014). Hence, the higher nutrient levels were likely associated with the removal of fertilizers from the soil during spring and summer.

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

Section 3 discussed the impacts of El Niño on water quality at the outlets and sub-basin scales. At the outlets, EP-El Niño 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, CP- and EP-El Niño could have equivalent impacts on water quality in specific watersheds, predominantly in Iowa (IA), Illinois (IL), Minnesota (MN), Wisconsin (WI), and Indiana (IN), which contributed the greatest amounts of nutrient change to the whole basin loads (Fig. 3). Table 4 listed the top 10 HUC-8 sub-basins with the largest nutrient change during the two 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; while in CP-ENYs, TN anomalies changed from 5.6 and 9.3 kg ha^{-1} (Table 4). These changes in TN during classic El Niño and El Niño Modoki were comparable. Analysis of the top 10 HUC-8 sub-basins with the largest TP change during the EP- and CP-El Niño illustrated similar results (not shown). These findings indicated that CP-El Niño could have comparable 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 on water quality at the outlets.

Table 4. Information on the top 10 8-digit Hydrologic Unit Codes (HUC-8) sub-basins with greatest TN anomalies during 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

4.3 Biogeochemical process variation due to temperature change

The effect of temperature on water quality through affecting evaporation and runoff has been analyzed in Section 3.2.2. In fact, the temperature can also affect water quality through some biochemical processes of nutrients (Neitsch et al., 2011). In order to analyze the biogeochemical process variations due to temperature changes during EP- and CP-El Niños, new analyses on nitrogen and phosphorus components, such as nitrate, organic nitrogen, soluble phosphorus, mineral phosphorus, and organic phosphorus had been carried out. Results from the analyses demonstrated that compared to precipitation, temperature plays a secondary role in altering nutrient levels through biogeochemical processes. Taking nitrate as an example, we showed the composite results of annual and seasonal nitrate anomalies (Figs. S6 and S7), respectively, during EP- and CP-El Niños. Figures S6 and S7 indicated that the pattern of nitrate was more similar to that of precipitation (Figs. 6a, 6c, and 7a-7h) but different from that of temperature (Figs. 8a, 8c, and 9a-9h) in the Corn Belt region during El Niños.

This could be further confirmed by the pattern correlation results. The correlation coefficients of annual nitrate and precipitation were 0.47, 0.36, 0.22, and 0.39, respectively, at OTRB and UMRB during EP- and CP-El Niños. The correlation coefficients between nitrate and temperature were relatively small (the coefficients were -0.15, 0.08, 0.30, and -0.31, respectively). The coefficient values altered between positive and negative at the two basins during EP- and CP-El Niños. The inconsistent relationships between nitrate and temperature were mainly because the nitrate content could vary through nitrification, mineralization, denitrification, and plant uptake processes (Neitsch et al., 2011). When the temperature rises, the former two processes increase nitrate content, but the latter two decrease nitrate content. Thus, the final sign of the correlation coefficient between nitrate and temperature really depends on the dominant processes. Similar results were also found at seasonal scales (not shown). These results indicated that nitrate variations were dominated by precipitation variations in the two basins during EP- and CP-El Niños, instead of temperature impacts on the biogeochemical processes. Similar results were also found for other nutrient components, such as organic nitrogen, soluble phosphorus, mineral phosphorus, and organic phosphorus (not shown).

4.4 Future water quality change

Existing studies suggested that CP-El Niño episodes occurred more frequently in a warming climate (Yeh et al., 2009; Yu et 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 climate. As CP-El Niño frequency increases in the future, TN and TP loads would likely increase over the Corn Belt region even under the same agricultural conditions, indicating a possible deterioration of water quality in this region when the climate warms.

The spatial patterns of TN and TP anomalies during CP-ENYs (Fig. S8) also suggested that specific watersheds, predominantly in southern UMRB and western OTRB, such as Iowa, Illinois, Wisconsin, Indiana, and Kentucky, will likely experience the most increases of TN and TP loads in the future. Such information is critical to ensure proper decision-making for watershed protection.

4.5 Limitations and future work

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

The impacts of irrigation on runoff and nutrient levels were not analyzed in the study due to a lack of irrigation data over the Corn Belt. Existing documents suggest that vast acreages of corn and soybeans are watered by center pivot irrigation in the region, which uses an apparatus that sprays water across a field with a 75−90% efficiency, thus irrigation water mostly

infiltrates into the soil (Grassini et al., 2011; 2014; Green et al., 2018). In other words, precipitation likely plays a dominant role in runoff; we thus focus on the impact of precipitation on runoff and water quality in the study. Further discussions on irrigation are needed once detailed irrigation data are available.

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 due to the differences in sea surface temperature (SST) evolution patterns and climate impacts, the distribution of nutrients in these two CP-El Niños might also differ and therefore are needed to be further studied in the future.

5 Conclusions

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

Annual composite results suggested that TN and TP loads decreased by 13.1% and 14.0% during EP-ENYs, respectively, whereas they increased by 3.3% and 4.6% during CP-ENYs at the outlet of the OTRB, respectively. TN and TP also showed 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 during CP-ENYs, respectively). Furthermore, more sub-basins showed negative annual anomalies of TN and TP during EP-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 kg of N ha^{-1} and 1.0 kg of P ha^{-1}), respectively, whereas they showed positive anomalies during CP-ENYs.

Seasonal composite results confirmed that water quality anomalies showed opposite patterns during EP-ENYs and CP-ENYs and the changes in the water quality matched the ENSO phases at the outlet of the OTRB. Maximum reduction or increase of the nutrients during EP-ENYs or CP-ENYs, respectively, occurred in winter, the peak season of El Niño. At the outlet of the UMRB (corn-growing region), TN and TP anomalies were also influenced by agriculture activities and became greater in spring and summer during both EP-ENYs and CP-ENYs, in consistent with the distribution of rainfall changes in the basin. These results suggested that small changes in climate variables such as precipitation in the growing season could have greater impacts on water quality in the UMRB during El Niño events.

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

Examination of the climate factors/processes on water quality change indicated that El Niño-induced precipitation and temperature changes altered runoff and evaporation, and thus TN and TP in both UMRB and OTRB on annual and seasonal time scales, as well as at the outlet and sub-basin scales. It is also found that nutrient levels were largely determined by

precipitation through runoff during both EP- and CP-ENYs, especially at the outlet, as the precipitation was a major source of runoff, and nitrogen and phosphorus components were transported by runoff. At the sub-basin scale, water quality was affected by the combination of precipitation and agricultural activities, especially in the UMRB during the growing season.

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

Data availability. Available upon request.

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

Competing Interests. The authors declare no competing interests.

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Acknowledgments. We thank Dr. Yongping Yuan for the suggestions on simulations of water quality in the OTRB and UMRB. This research was supported by the Shanxi Province Science Foundation for Youths, China (No. 202103021223106, 202103021223079 and 20210302124458), the Research Project Supported by Shanxi Scholarship Council of China (No. 2022-080 and 2021-051), the Shanxi Province Water Science and Technology Research and Promotion Project, China (No. 2022GM023 and 2022GM017) the National Key Research and Development Program, China (No. 2018YFC0406406), the China Scholarship Council (File No. 201806935023), the Institute-Local Cooperation Project of the Chinese Academy of Engineering (No. 2020SX8), the Government Financial Grants Project, China (No. ZNGZ2015-036), the Key Research and Development Program (Social Development Field) of Shanxi Province, China (No. 201903D321052), and the Natural Science Foundation of Shanxi Province, China (No. 201901D111059 and 201901D111060).

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