

## RESEARCH ARTICLE

# Land cover, stream discharge, and wastewater effluent impacts on baseflow sediment and nutrient concentrations in SW Ohio streams

Rachel E. Spahr<sup>1</sup> | Jeffrey A. Lazar<sup>2</sup> | Bartosz P. Grudzinski<sup>2</sup>  | Thomas J. Fisher<sup>3</sup>

<sup>1</sup>Institute for the Environment and Sustainability, Miami University, Oxford, Ohio, USA

<sup>2</sup>Department of Geography, Miami University, Oxford, Ohio, USA

<sup>3</sup>Department of Statistics, Miami University, Oxford, Ohio, USA

## Correspondence

Bartosz P. Grudzinski, Department of Geography, Miami University, Oxford, OH, USA.

Email: [grudzibp@miamioh.edu](mailto:grudzibp@miamioh.edu)

## Abstract

Elevated nutrient and suspended sediment concentrations often result in negative environmental impacts within freshwater environments. Studies that directly compare suspended sediment and bioavailable nutrients between predominantly agricultural and predominantly urban watersheds during baseflow conditions are largely lacking. The purpose of this study was to determine the impacts of land cover, stream discharge, and wastewater treatment plant (WWTP) discharge on nutrient and sediment concentrations, across a large land cover gradient in Southwest Ohio streams. Weekly baseflow samples were collected from eight streams over 1 year from November, 2016 through November, 2017. Total suspended sediment, nitrate, and phosphate concentrations were measured. Results indicate that agricultural land cover and WWTPs increase nitrate and phosphate concentrations in the study area. Total suspended sediment and nitrate concentrations increased with discharge, and discharge was a relatively weak predictor of phosphate concentrations. Seasonal water quality trends varied by parameter and land use also had unique impacts on seasonal water quality trends. Results suggest that to improve water quality in the study area, efforts should focus on improving WWTP effluent treatment and agricultural land management.

## KEYWORDS

agriculture, baseflow, land use, nutrients, suspended sediment, urban, wastewater treatment, water quality

## 1 | INTRODUCTION

Streams are widely utilized as sources of water for consumption in urban areas (Padowski & Gorelick, 2014), as irrigation in agricultural regions (Hashemi et al., 2016), as habitat for wildlife (Raven et al., 2000), and as corridors for sediment and nutrient transfer from terrestrial environments (Glińska-Lewczuk et al., 2016). Expansion of

agricultural and urban land cover in the Midwest of the United States and throughout many areas of the world have significantly increased stream sediment and nutrient concentrations, fundamentally altering the chemical, physical, and biological properties of stream environments (Gallo et al., 2015; Padowski & Gorelick, 2014; Renwick et al., 2008; Riseng et al., 2011). Despite widespread degradation of water quality due to anthropogenic development (Giri et al., 2018),

This is an open access article under the terms of the [Creative Commons Attribution-NonCommercial-NoDerivs](https://creativecommons.org/licenses/by-nc-nd/4.0/) License, which permits use and distribution in any medium, provided the original work is properly cited, the use is non-commercial and no modifications or adaptations are made.

© 2024 The Authors. *River Research and Applications* published by John Wiley & Sons Ltd.

relative impacts between predominantly urban and predominantly agricultural land cover on stream sediment and nutrient concentrations remain poorly understood in the region and globally.

Sediment is considered the primary cause of water quality impairment in U.S. streams (Govenor et al., 2017). Excessive sediment inputs from developed watersheds can have negative impacts on biota by clogging stream bed habitat (Gayraud & Philippe, 2003), increasing abrasion on flora and fauna (Heatherly et al., 2007), and reducing overall biodiversity (Skarbovik et al., 2012). Excessive nutrient concentrations often lead to degradation of aquatic environments by promoting hypoxia and eutrophication (Brion et al., 2015; Dodds & Smith, 2016) and altering food web dynamics (Cashman et al., 2013; Sánchez-Pérez et al., 2009). Streams in the Midwest region of the United States contain some of the highest nutrient concentrations in the country (Bellmore et al., 2018). Most recently, excessive nutrient loading in the region has led to costly algal blooms in the Ohio River and Lake Erie (Daloğlu et al., 2012; Henson et al., 2018).

Agricultural practices such as tilling, crop harvesting, and fertilizer application often promote soil erosion and transport of sediment and nutrient-enriched water to nearby streams (Bates & Arbuckle Jr., 2017; Miller et al., 2011). In streams draining urban watersheds, sediment concentrations are often increased by construction in the watershed and destabilization of streambanks (e.g., through stream incision and widening) (Russell et al., 2017). Additionally, many urban areas contain aging sewage networks, which may lead to increased nutrient concentrations in adjacent streams (Ferreira et al., 2018). In watersheds with urban development and WWTPs, reducing nutrient concentrations in wastewater is an important component of water quality management (Yamashita & Yamamoto-Ikemoto, 2014). Nutrient-enriched effluent is particularly detrimental during baseflow periods (e.g., summer and fall in the U.S. Midwest), when it can account for a majority of stream discharge (Brion et al., 2015).

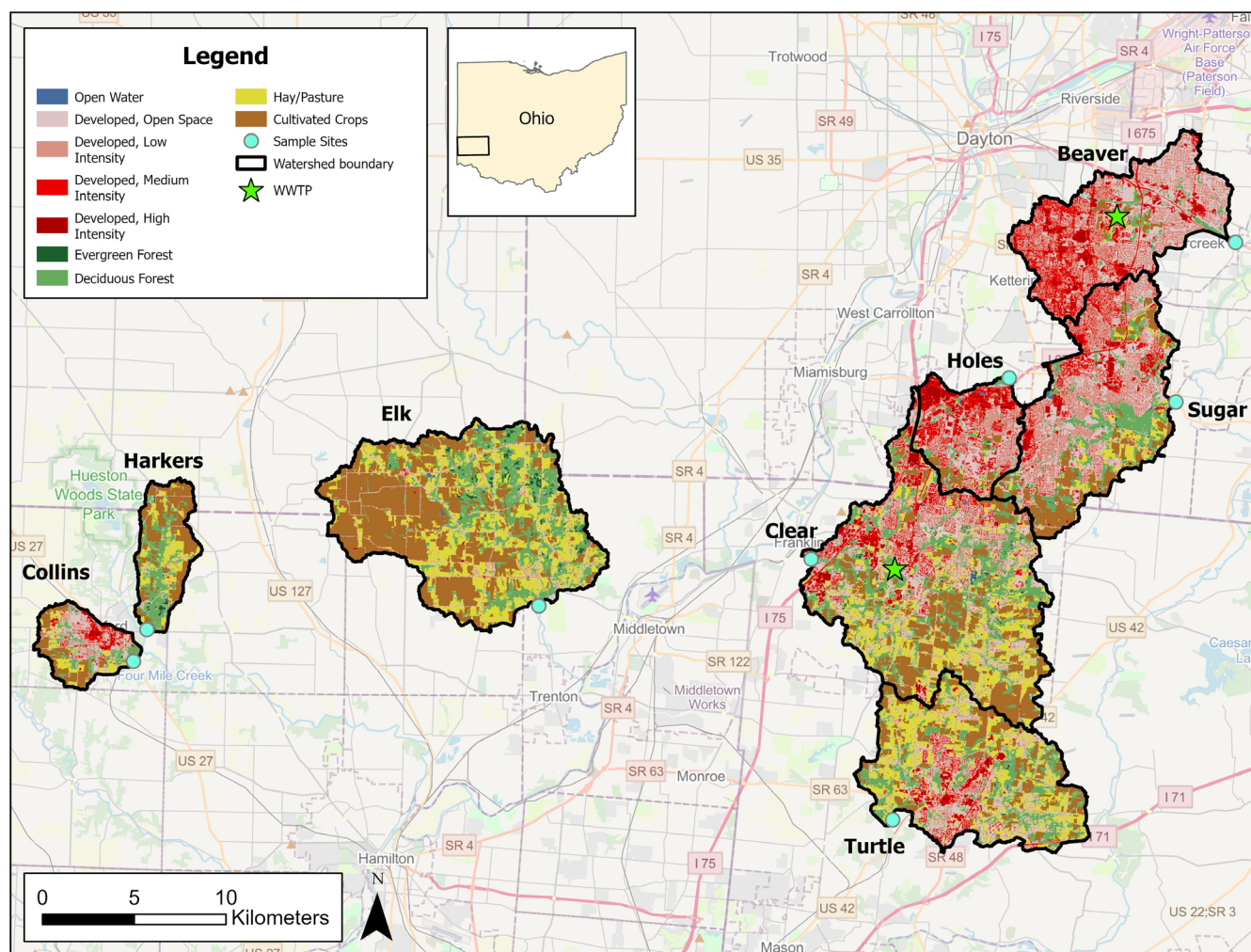
In addition to anthropogenic disturbances such as land cover modification and the development of WWTPs, sediment and nutrient concentrations are often influenced by streamflow (De Girolamo et al., 2015). During storm events, sediment and nutrient concentrations are often higher relative to baseflow conditions, due to increased runoff and sediment transport capacity (Pizarro et al., 2013). Yet, during baseflow periods, higher discharge may actually decrease suspended sediment concentrations, due to increased groundwater inputs of sediment-free water (Estrany et al., 2008). High discharge can also increase baseflow nutrient concentrations by increasing groundwater transport of nitrate and soluble phosphate through soils and into streams (Calhoun et al., 2002; Gallo et al., 2015). However, higher baseflow may also decrease nutrient concentrations through dilution, particularly in watersheds with high effluent discharge (Rodríguez Benítez et al., 2015).

Many studies have demonstrated that sediment and nutrient concentrations increase with land development (both agricultural and urban; Carpenter et al., 1998; Turner & Rabalais, 2003; Jin et al., 2020). Yet, most studies focus on either predominantly agricultural (Ford et al., 2018), predominantly urban (Hoellein et al., 2011), or mixed land cover watersheds (Richards et al., 2008). Currently, studies that directly

compare water quality between highly urbanized watersheds (>50%) and highly agricultural watersheds (>50%) are largely lacking. This makes prioritization of best management practices in mixed land cover watersheds difficult, particularly as seasonal patterns of baseflow water quality may vary across land cover types. Baseflow concentrations of bioavailable nutrients are particularly important in the region and globally as they have significant implications for harmful algal blooms (Ford et al., 2018; Shore et al., 2017). In this study, we aim to increase our understanding of impacts from land cover (agricultural vs. urban), the presence or absence of WWTPs, and stream discharge on stream total suspended sediment (TSS), nitrate ( $\text{NO}_3^-$ ), and phosphate ( $\text{PO}_4^{3-}$ ) concentrations within the Ohio River Valley. We also examine seasonal discharge and water quality dynamics between land cover types. We hypothesize that 1) agricultural land cover, rather than urban land cover, will be the primary driver of sediment and nutrient concentrations in the study streams, due to increased sediment and nutrient availability from exposed and fertilized soils; 2) WWTPs will significantly increase nutrient concentrations in the watersheds in which they are present; 3) sediment and nutrient concentrations will increase with baseflow stream discharge; and 4) sediment and nutrient concentrations will show greater seasonal variability in agricultural watersheds relative to urban watersheds.

## 2 | STUDY AREA

This study was conducted in Southwest Ohio located in the Eastern Corn Belt Ecoregion, where forested areas have largely been converted to urban or agricultural land cover (Figure 1). The two largest cities near the study area are Cincinnati to the south and Dayton to the north. One stream sampling point across eight watersheds was selected to attain a large land cover gradient (i.e., watershed sampling points are not nested and are thus independent of each other). In the eight study watersheds urban land cover (developed open space, low intensity, medium intensity, and high-intensity development) ranged from 7% to 92%, agricultural land cover (cultivated crops, pasture/hay) ranged from 4% to 71%, and forested land cover (deciduous) ranged from 4% to 29% (Figure 1; Table 1). Land cover within each watershed was determined based on the 2011 National Land Cover Database in ArcGIS (version 10.5). Vegetation in the region primarily consists of crops, mostly corn and soybeans, followed by deciduous forest cover (Rech et al., 2018). Corn is generally planted in May and harvested in mid-November, while soybeans are typically planted in early May and harvested in late October (USDA, 2010). Following cultivation, soils are left bare for about 5–6 months unless cover crops are planted. The average temperature in the study area is lowest in January ( $-1.1^\circ\text{C}$ ) and highest in July ( $24^\circ\text{C}$ ). The average annual precipitation is 1097 mm and average precipitation is lowest in February (68 mm) and highest in May (134 mm). Stream discharge is typically highest in early spring (March–May) and lowest in the fall (September–October). The underlying geology within the study area consists of limestones and shales from the Ordovician Period (Weiss & Sweet, 1964). Soils are primarily composed of silt and silty clay loams (Lerch et al., 1980).



**FIGURE 1** Watershed land cover, wastewater treatment plant (WWTP) locations, and sampling sites. [Color figure can be viewed at [wileyonlinelibrary.com](https://onlinelibrary.wiley.com/doi/10.1002/rm.4248)]

**TABLE 1** Land cover and wastewater treatment plant (WWTP) discharge within each study watershed.

Site	Watershed area (km <sup>2</sup> )	Agriculture (%)	Urban (%)	Forest (%)	WWTP (m <sup>3</sup> /day)
Beaver Creek	62.0	3.8	92.3	3.8	49,200
Holes Creek	29.9	6.6	83.6	9.3	NA
Sugar Creek	78.5	21.5	61.6	16.3	NA
Collins Creek	16.9	35.5	47.4	16.3	NA
Clear Creek	129.0	39.7	36.7	22.2	15,100
Turtle Creek	77.2	45.4	33.2	20.4	NA
Elk Creek	113.1	71.4	7.1	20.9	NA
Harker's Run	16.9	62.5	6.69	29.2	NA

Two of the study watersheds contain large WWTPs with National Pollutant Discharge Elimination System (NPDES) permits that allow for limited effluent discharge. The Springboro WWTP discharges into Clear Creek and has an average design flow of 15,142 m<sup>3</sup>/day (4,000,000 gal/day; Ohio EPA, 2017). The Eastern Regional Water Reclamation Facility discharges into Beaver Creek and has an average design flow of 49,210 m<sup>3</sup>/day (13,000,000 gal/

day). The WWTP is permitted to discharge a maximum TP load of 24.6 kg/day during the months of May through October and must limit TP discharge concentrations to 1.0 mg/L. Phosphorus concentrations during the winter have no maximum limits (Ohio EPA, 2013). Due to its smaller watershed area and greater WWTP discharge, Beaver Creek receives nearly 7 times more effluent per unit watershed area relative to Clear Creek.

### 3 | METHODS

#### 3.1 | Water sample collection and analysis

Baseflow water samples were collected weekly from each study watershed over a 1-year period from November 19, 2016 through November 11, 2017. During this period, water samples were collected for 50 of 52 weeks. We were unable to sample for 2 weeks due to logistical constraints. When storm events occurred, sampling was postponed typically for 24–48 h to mitigate the collection of water containing overland flow, however, it is possible that large precipitation events may have resulted in samples containing some overland flow. A total of 400 water samples were collected across all watersheds. Between 1% and 5% of collected samples, depending on parameter, were not included in the data analysis due to sample contamination in the field or lab ( $\text{NO}_3^-$ ,  $n = 379$ ;  $\text{PO}_4^{3-}$ ,  $n = 379$ ; TSS,  $n = 395$ ). Nutrient samples from the first two sampling days had to be discarded due to processing errors. These 2 days account for the majority ( $n = 16$  of 21) of excluded data. The few remaining samples that were not included in the analyses were not tied to any time period or study site.

During each sampling day, a 4 L pre-washed polyethylene bottle was triple rinsed with stream water and then filled from the thalweg of each stream. Once obtained, water samples were transported on ice in a cooler and subsequently refrigerated at 4°C until processing (within 48 hours). The TSS concentration (mg/L) for each sample was determined by filtering 2–3 L of collected stream water through pre-weighted type A/E glass micro-fiber filters (pore size of 1 µm), dehydrating sediment-laden filters for 48 h at 105°C, weighing dried filters on a microbalance (Mettler Toledo model XP6, Columbus, Ohio, USA), and dividing dried weight by the volume of water filtered. During sediment filtration, a 125 mL subsample of filtered water was collected and preserved with sulfuric acid for  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  analyses. Nutrient concentrations were measured with a Lachat Quickchem 8500 (series 2) auto-analyzer (Lachat Instruments, Loveland, Colorado, USA) following methods 10-107-04-1-A (for  $\text{NO}_3^-$ ; Wendt, 2000) and 10-115-01-1-Q (for  $\text{PO}_4^{3-}$ ; Diamond, 2007).

#### 3.2 | Discharge

During water sampling, the stream stage was recorded from a staff gauge that was installed at each of the eight study streams. Stream stages were converted to discharge with USGS WinXSPRO software (e.g., Cornwell et al., 2003; Sandercock & Hooke, 2010). To compute discharge, WinXSPRO requires the user to input each stream's cross-section profile, low and high flow slopes, and Manning's  $n$  value (Hardy et al., 2005). At each stream's sampling location, the channel cross-section was manually surveyed with a surveyor's level and stadia rod following Harrelson et al. (1994). Cross-section surveys were completed at ~30 cm increments with additional survey locations in areas of high topographic variability. Stream slope at low flow was

generally too small to accurately measure in the field and was determined by solving for  $S$ :

$$V = k/nR^{2/3}S^{1/2} \quad (1)$$

where  $V$  is velocity,  $k$  is a conversion factor,  $n$  is Manning's roughness coefficient,  $R$  is the hydraulic radius, and  $S$  is channel slope. In Equation (1),  $V$  was determined from manual discharge measurements that were completed in the field with a Flowtracker 2 velocity meter (SonTek, San Diego, California, USA; e.g., Lazar et al., 2019). Since measuring slope under high flow was not possible due to unwardable conditions, high flow slopes for each stream were determined based on slope measurements between geomorphic channel units (e.g., riffles) following Hardy et al. (2005). A Manning's  $n$  value was determined for each stream based on channel bed substrate, degree of channel irregularity, variation in channel cross-section, the presence-absence of obstructions, and amount of vegetation (Hardy et al., 2005). Following input of stream metrics, WinXSPRO calculated discharge values at stage increments of ~3 cm for each stream. Observed stages during sampling were matched with discharge calculations for each stream.

#### 3.3 | Statistical analyses

Linear mixed-effects models (LMMs) were utilized to determine if watershed land cover (as reported by the percentage of land that is agricultural; see Section 4 for details), stream discharge, and/or WWTP effluent discharge significantly influenced TSS,  $\text{NO}_3^-$ , and  $\text{PO}_4^{3-}$  concentrations. Stream discharge and WWTP discharge were scaled by watershed area to allow for cross-watershed comparisons. The structure of the model is

$$Y_{ij} = \beta_{0j} + \beta_{1j} \times \text{Discharge} + \gamma \times \text{WWTP} + \delta \times \text{LandUse} + \epsilon_{ij},$$

where  $Y_{ij}$  is the TSS,  $\text{NO}_3^-$ , and  $\text{PO}_4^{3-}$  concentration recorded at time point  $i$  in stream  $j$ . The fixed effects in the models were percent land cover and WWTP discharge (modeled by  $\delta$  and  $\gamma$ , respectively) while stream discharge was considered a random effect (the  $\beta_{1j}$  term), as streamflow can vary by site due to natural watershed variability and heterogeneity of precipitation characteristics between rain events. To account for any underlying different creek dynamics, the intercept (the  $\beta_{0j}$  term) is a random effect, and to model any potential temporal serial correlation in the data, an autoregressive model of order 1 (AR (1)) was fit on the underlying noise terms (the  $\epsilon_{ij}$  term). All response variables were log-transformed to satisfy the underlying statistical assumptions. Stream discharge and agricultural land cover percentage were log-transformed to improve the fitted model. Statistical models with various land covers (urban, agricultural, and forested) were compared with Akaike information criterion (AIC) and the model using agricultural land cover provided the best fit. The significance of coefficients ( $p$ -values) was utilized to determine which independent variables significantly impacted each dependent water quality variable where  $p$ -values

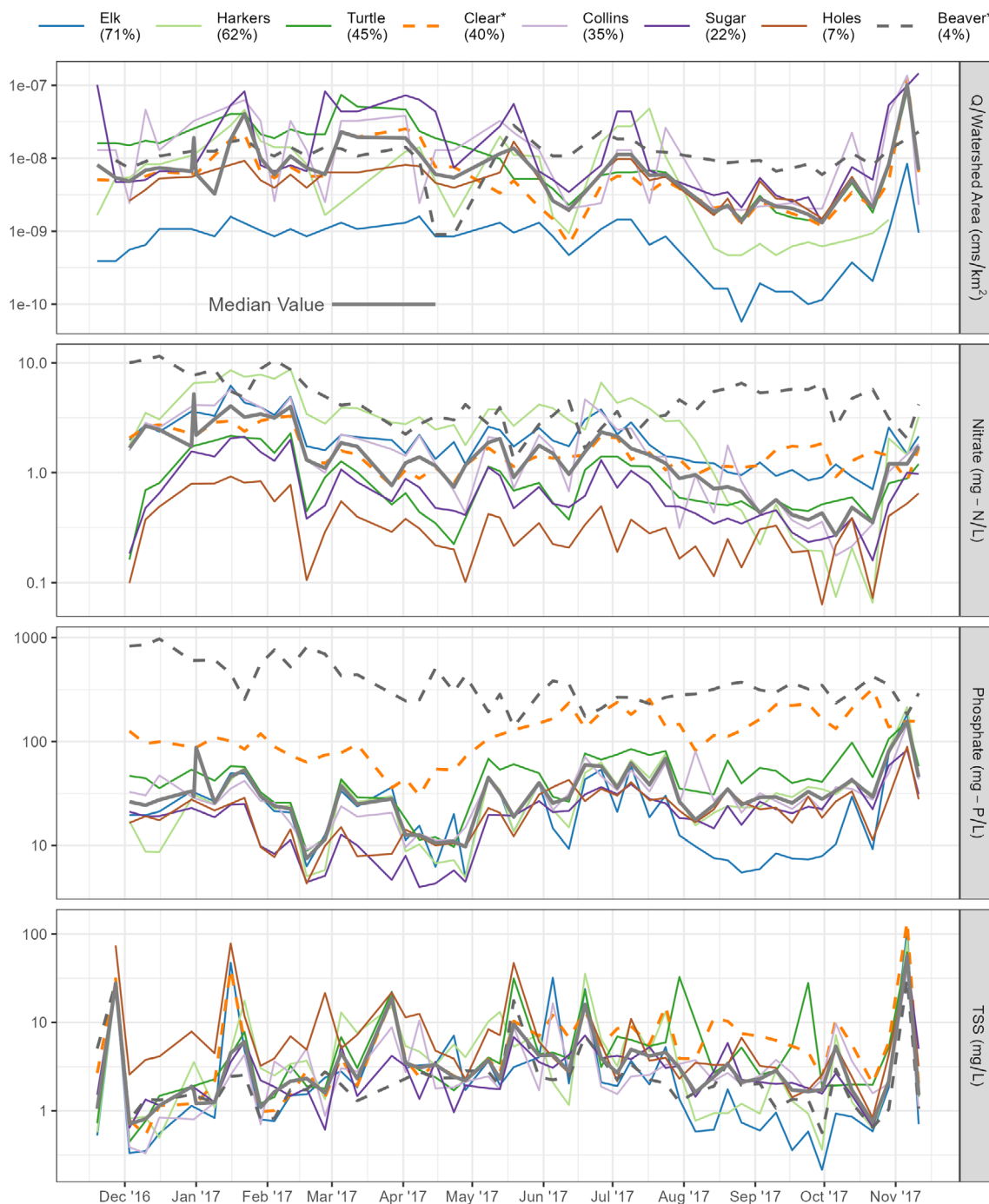


<0.05 are considered to be significant and  $p$ -values <0.10 are considered marginally significant. Pseudo- $R^2$  (hereafter referred to as partial  $R^2$  or  $R^2$ ) values were determined following Edwards et al. (2008) with the *r2glmm* package in R (e.g., Borzooei et al., 2019; Jaeger, 2017; Sarremejane et al., 2017). The relative influence of predictor variables within each model was then determined by comparing partial  $R^2$  values for variables that had a significant impact on each water quality parameter. Model fitting and analysis were completed using the *nlme* package (Pinheiro et al., 2018) within R (version 4.3.1, R Core Team, 2023).

## 4 | RESULTS

### 4.1 | Temporal discharge and water quality patterns

Over the course of the study period, discharge during baseflow sampling days followed typical patterns for the study area, with the highest baseflow discharge in the spring and lowest baseflow discharge in late summer and early fall (Figure 2). Discharge patterns during



**FIGURE 2** Observed discharge scaled by watershed area, nitrate ( $\text{NO}_3^-$ ), phosphate ( $\text{PO}_4^{3-}$ ), and total suspended sediment (TSS) concentrations from each sampling day (approximately  $n = 50$ ) across each study watershed ( $n = 8$ ), with median concentration displayed. Percent agricultural land cover is specified for each watershed at the top of the figure. [Color figure can be viewed at [wileyonlinelibrary.com](https://onlinelibrary.wiley.com/terms-and-conditions)]

baseflow sampling days were generally consistent across sites with some notable exceptions. For example, within the two most agricultural watersheds (Elk Creek and Harker's Run) decreases during the dry season appear to be more drastic relative to less agricultural watersheds. Beaver Creek has the lowest decrease during this period, likely due to consistent effluent discharge from the WWTP which increases overall discharge within the stream. Thus, discharge within Beaver Creek is more consistent throughout the year relative to other sites (Figure 2).

Nitrate concentrations were highest during the winter months and lowest during the fall. Nitrate concentrations increased from November through February, then decreased and varied through July and continued to decrease with more consistency through late October until again increasing through November (Figure 2). Within the two sites that contained WWTPs (Beaver Creek and Clear Creek),  $\text{NO}_3^-$  concentrations appear to decrease less during the fall and summer months than within the remaining sites. Phosphate concentrations are generally lowest during late winter and early spring months and are highest during summer and fall months. Phosphate concentrations were highly variable from the start of the study period through April, increased but remained variable through July, then decreased and remained relatively steady through mid-October, and increased again in November (Figure 2). Beaver Creek tends to deviate from the normal seasonal patterns of other sites and has the highest  $\text{PO}_4^{3-}$  concentrations in the winter. Clear Creek which has a lower amount of effluent discharge and a larger watershed area relative to Beaver Creek tends to follow seasonal  $\text{PO}_4^{3-}$  patterns consistent with other sites, although concentrations are higher throughout the year than in sites without WWTPs (Figure 2). Seasonal patterns in TSS are consistent across sites, with no apparent differences driven by land cover. TSS concentrations peak in the spring and early summer. Total suspended sediment concentrations varied throughout the year, but generally showed an increase from February through June and a decrease from August through October (Figure 2).

## 4.2 | Land cover

Agricultural land cover, rather than urban or forested land cover, were included in the statistical models for all response variables (TSS,  $\text{NO}_3^-$ , and  $\text{PO}_4^{3-}$ ), as AIC determined that it explained the greatest proportion of each model's variance (Table S1). Due to the strong inverse relationship between agricultural and urban land cover within the study watersheds (Figure S1), if the concentration of a water quality parameter increased with agricultural land cover,

it decreased with urban land cover. This does not indicate that urban land cover improves water quality, but rather, the observed patterns are due to a strong negative correlation between urban and agricultural land cover (i.e., if agricultural land cover increases in a watershed, then urban land cover will decrease). The correlation coefficient between urban and agricultural land cover is  $-0.9911$ . To reduce information redundancy (highly correlated variables do not necessarily add new information to a model and can make model interpretation difficult, see Harrison et al., 2018), we report findings with respect to agricultural land cover. However, it is important to note that the significance of independent variables does not change (to or from significance at  $p = 0.05$  or  $0.10$ ) when substituting urban land cover for agricultural land cover within the statistical model.

## 4.3 | Nitrate

The LMM model for  $\text{NO}_3^-$  was highly significant ( $p < 0.001$ ) and demonstrated that  $\text{NO}_3^-$  concentrations increased with agricultural land cover ( $p = 0.0038$ ), stream discharge ( $p = 0.036$ ), and WWTP discharge ( $p = 0.0011$ ) (Table 2; Figure 3). The model contained an  $R^2$  value of 0.70 and indicated that the strongest predictor of  $\text{NO}_3^-$  concentration was WWTP discharge ( $R^2 = 0.68$ ), followed by agricultural land cover ( $R^2 = 0.53$ ), and stream discharge ( $R^2 = 0.17$ ).

## 4.4 | Phosphate

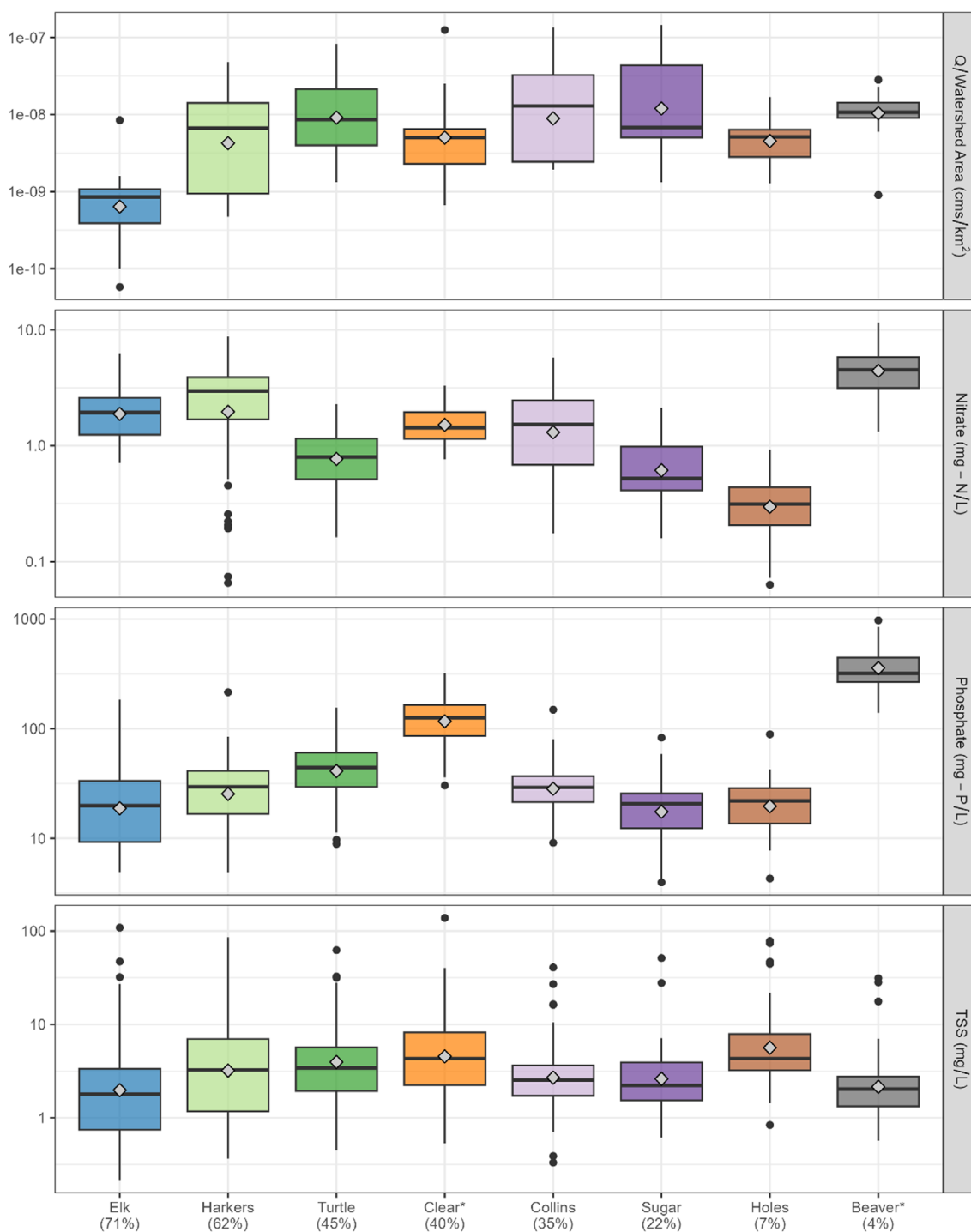
The LMM model for  $\text{PO}_4^{3-}$  was highly significant ( $p < 0.001$ ) and demonstrated that  $\text{PO}_4^{3-}$  concentrations increased with agricultural land cover ( $p = 0.069$ ), stream discharge ( $p = 0.059$ ), and WWTP discharge ( $p = 0.0037$ ) (Table 2; Figure 3). The model contained an  $R^2$  value of 0.80 and indicated that the strongest predictor of  $\text{PO}_4^{3-}$  concentration was WWTP discharge ( $R^2 = 0.79$ ), followed by agricultural land cover ( $R^2 = 0.40$ ), and stream discharge ( $R^2 = 0.22$ ).

## 4.5 | Total suspended sediment

The LMM model for TSS was significant ( $p = 0.039$ ) and demonstrated that TSS concentrations significantly increased with discharge ( $p = 0.0029$ ). Sediment concentrations were not significantly related to land cover or WWTP discharge ( $p > 0.10$ ) (Table 2; Figure 3). The model contained an  $R^2$  value of 0.19 and stream discharge was a weak predictor of sediment concentrations ( $R^2 = 0.16$ ).

Parameter	Y-intercept	Agriculture %	Stream Q ( $\text{m}^3/\text{s}/\text{km}^2$ )	WWTP Q ( $\text{m}^3/\text{s}/\text{km}^2$ )
$\text{NO}_3^-$ (mg/L)	1.0 (0.750)	0.0139 (0.003)	0.19 (0.090)	$5.8\text{e}-06$ ( $9\text{e}-07$ )
$\text{PO}_4^{3-}$ ( $\mu\text{g}/\text{L}$ )	2.8 (0.902)	0.0097 (0.004)	0.21 (0.108)	$7.0\text{e}-06$ ( $1\text{e}-07$ )
TSS (mg/L)	7.4 (2.172)	0.0019 (0.007)	0.75 (0.250)	$-2.5\text{e}-06$ ( $3\text{e}-08$ )

**TABLE 2** Y-intercepts and model coefficients for agricultural land cover, stream discharge (stream Q), and wastewater treatment plant discharge (WWTP Q). Standard errors of coefficients are in parenthesis.



**FIGURE 3** Distribution of discharge scaled by watershed area, nitrate ( $\text{NO}_3^-$ ), phosphate ( $\text{PO}_4^{3-}$ ), and total suspended sediment (TSS) concentrations within each stream. Sites (watersheds) are organized from the highest percentage of agricultural land cover (left) to the lowest percentage of agricultural land cover (right). \*Clear Creek and Beaver Creek contain WWTPs. Percent agricultural land cover is specified for each watershed at the bottom of the figure. [Color figure can be viewed at [wileyonlinelibrary.com](https://onlinelibrary.wiley.com/terms-and-conditions)]

## 5 | DISCUSSION

Baseflow nutrient concentrations in the study area were increased by agricultural land cover and WWTP discharge. Stream discharge also significantly increased  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  concentrations, but explained the

least amount of each model's variance. Suspended sediment concentrations increased with streamflow and were not significantly impacted by land cover or WWTP discharge. Although statistically significant, stream discharge was a weak predictor of TSS concentrations. Overall, WWTP discharge followed by agricultural land cover appears to be most

detrimental to water quality in the study region. Land cover and the presence of a WWTP in a watershed create distinct seasonal patterns in nutrient concentrations, but not sediment concentrations. Thus, watershed-specific best management practices will likely need to be developed for each watershed and will also depend on which water quality parameter is being prioritized for improvement and during which season. While we are not aware of any studies that have compared seasonal patterns of baseflow  $\text{NO}_3^-$  or  $\text{PO}_4^{3-}$  concentrations between various land cover types (e.g., majority agricultural vs. majority urban), Ford et al. (2018) examined total nitrogen and total phosphorus concentrations within an agricultural watershed. Our  $\text{PO}_4^{3-}$  seasonal patterns generally agree with their seasonal total phosphorus patterns with our high values generally occurring during the summer in agricultural watersheds. Furthermore, our  $\text{NO}_3^-$  concentrations reflect their total nitrogen concentrations within agricultural concentrations with peaks during winter months which coincide with periods of higher baseflow discharge. Higher  $\text{NO}_3^-$  concentrations during winter months are also likely influenced by lower microbial activity and denitrification rates during colder periods (Gervasio et al., 2022). Nitrate is highly soluble and easily transported through soils, thus higher baseflow values are likely to lead to increased connectivity with  $\text{NO}_3^-$  enriched groundwater (Ford et al., 2018). While stormflows are known to rapidly increase sediment concentrations, our results indicate that relatively small increases in baseflow discharge can also increase sediment concentrations within both agricultural and urban streams. Additional studies are needed that compare seasonal nutrient and sediment dynamics between predominantly urbanized watersheds and predominantly agricultural watersheds, particularly during baseflow conditions.

Nitrate and  $\text{PO}_4^{3-}$  concentrations increased with agricultural land cover within the study area. The use of fertilizers has significantly increased in the United States since the 1940s, particularly in the Midwest U.S. (Cao et al., 2018). Fertilizers, especially nitrogen, have been shown to increase baseflow stream nutrient concentrations, because they are prone to leaching from agricultural soils, into groundwater, and subsequently into streams (e.g., Chand et al., 2011; Collins et al., 2017; Stets et al., 2015). Surprisingly, the percentage of agricultural land cover in a watershed was not significantly related to baseflow TSS concentrations. Agricultural practices such as tilling and harvesting loosen soil making it readily available for transport during runoff events (Jones & Schilling, 2011). In regards to TSS, land cover may be a more important factor during stormflow conditions as sediment is predominantly transported to streams during overland flow (Koskela et al., 2018; Lazar et al., 2019).

Stream discharge may have only been a relatively weak predictor of  $\text{PO}_4^{3-}$  as concentrations were largely impacted by effluent inputs (Table 2). Higher streamflow often has a greater capacity to suspend and transport substrate (Lenhart et al., 2010), including sediment that contains bound nutrients (Harrington, 2014). The relationship between streamflow and TSS indicates that within the study area, higher baseflow rates are more likely to increase sediment transport capacity than to drive dilution. However, the overall low predictiveness of suspended sediment concentrations in relationship to discharge indicates that variables not measured in this study may also have a stronger effect on sediment transport during baseflow (e.g., underlying geology and stream

geomorphology; Bywater-Reyes et al., 2017, previous land use; Lizaga et al., 2021, and legacy effects; Jiang et al., 2020).

WWTP discharge increased  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  concentrations. Beaver Creek had the highest nutrient concentrations of all the sampled sites (Figure 3), likely due to the significant effluent discharge from the Eastern Regional WWTP. Beaver Creek is located in the most urban watershed within the study area and subsequently discharges the most wastewater effluent ( $>49,000 \text{ m}^3/\text{day}$ ). Furthermore, the watershed is relatively small ( $62 \text{ km}^2$ ) and only provides limited dilution potential (i.e., drains into a stream with low discharge). During sampling the scent of effluent was evident within Beaver Creek, particularly during summer days. In 2010, the United States Environmental Protection Agency conducted biological sampling and found that a longitudinal pattern of impact and recovery was evident in relation to the Springboro WWTP, and the macroinvertebrate community was rated as marginal downstream of the plant (Ohio EPA, 2012). Numerous studies outside of these watersheds have also found negative biological effects in relation to WWTP effluent (e.g., Berninger et al., 2014; Drury et al., 2013). This is especially concerning during dry periods when WWTP discharge can be the main source of flow in rivers and effluent is not effectively diluted. Without advanced treatment techniques that are effective at removing nutrients from effluent, significant  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  concentrations are likely to continue to be detrimental to local water bodies (e.g., Edlund et al., 2009; Volf et al., 2013).

There were several limitations that should be considered when interpreting the results of this study. First, only two of the watersheds contained WWTPs upstream of sampling sites. While WWTP discharge had a significant impact on nutrient concentrations, other WWTPs may have variable impacts due to differing treatment practices, discharge rates, and characteristics of receiving streams. Although the impact of the two WWTPs in this study should be cautiously interpreted due to limited replication, upstream-downstream sampling during the summer of 2017 (Figure S2) further supports that the WWTPs in the study area are driving the observed increases in nutrient concentrations. Specifically, the average  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  concentrations below the WWTPs were, respectively, 5 and 30 times higher downstream of the two WWTPs. Another limitation of this study is that there were no undisturbed, predominantly forested, watersheds to use as controls near the study area, as most of the land in Southwest Ohio is developed. Thus, we were unable to determine natural nutrient and sediment concentrations during the study period. Lastly, other factors such as fertilizer application rates and timing, crop types, use of soil conservation practices (no-till agriculture and cover crops), local variability in geology and stream geomorphology, and land use history may also have some influence on the sediment and nutrient concentrations observed in the study streams, but examining the impact of these factors was outside the scope of this study.

## 6 | CONCLUSION

Water quality in the study area during baseflow conditions is negatively impacted by agricultural land cover and WWTP effluent



discharge. WWTP discharge was the primary driver of nutrient concentrations but showed no impact on TSS concentrations. Agricultural land cover also significantly increased  $\text{NO}_3^-$  and  $\text{PO}_4^{3-}$  concentrations. Stream discharge increased all nutrient and suspended sediment concentrations, however, was a generally weaker predictor of nutrient concentrations. Agriculture along with the presence of WWTPs influence seasonal patterns of nutrient concentrations in the study area. Improved management practices are needed in the study area to bolster water quality by reducing nutrient concentrations that are sourced from WWTPs and agricultural land cover.

## ACKNOWLEDGMENTS

We would like to thank Jonathan Levy, Michael Vanni, and Amy Wolfe for suggestions on improving an earlier draft of this manuscript. We would also like to acknowledge Tera Ratliff and the Center for Aquatic and Watershed Sciences at Miami University for completing nutrient analyses.

## DATA AVAILABILITY STATEMENT

All raw data will be published within the “Mendeley Data” data repository <https://data.mendeley.com/> upon publication of the manuscript.

## ORCID

Bartosz P. Grudzinski  <https://orcid.org/0000-0002-4505-0824>

## REFERENCES

- Bates, H., & Arbuckle, J. G., Jr. (2017). Understanding predictors of nutrient management practice diversity in Midwestern agriculture. *Journal of Extension*, 55(6). <https://www.joe.org/joe/2017december/a5.php>
- Bellmore, R. A., Compton, J. E., Brooks, J. R., Fox, E. W., Hill, R. A., Sobota, D. J., Thornburgh, D. J., & Weber, M. H. (2018). Nitrogen inputs drive nitrogen concentrations in US streams and Rivers during summer low flow conditions. *Science of the Total Environment*, 639, 1349–1359. <https://doi.org/10.1016/j.scitotenv.2018.05.008>
- Berninger, J. P., Martinović-Weigelt, D., Garcia-Reyero, N., Escalon, L., Perkins, E. J., Ankley, G. T., & Villeneuve, D. L. (2014). Using transcriptomic tools to evaluate biological effects across effluent gradients at a diverse set of study sites in Minnesota, USA. *Environmental Science & Technology*, 48(4), 2404–2412. <https://doi.org/10.1021/ES4040254>
- Borzooei, S., Teegavarapu, R., Abolfathi, S., Amerlinck, Y., Nopens, I., & Zanetti, M. C. (2019). Data mining application in assessment of weather-based influent scenarios for a WWTP: Getting the Most out of plant historical data. *Water, Air, & Soil Pollution*, 230(1), 1–12. <https://doi.org/10.1007/S11270-018-4053-1>
- Brion, N., Verbanck, M. A., Bauwens, W., Elskens, M., Chen, M., & Servais, P. (2015). Assessing the impacts of wastewater treatment implementation on the water quality of a small Urban River over the past 40 years. *Environmental Science and Pollution Research*, 22(16), 12720–12736. <https://doi.org/10.1007/s11356-015-4493-8>
- Bywater-Reyes, S., Segura, C., & Bladon, K. D. (2017). Geology and geomorphology control suspended sediment yield and modulate increases following timber harvest in temperate headwater streams. *Journal of Hydrology*, 548, 754–769. <https://doi.org/10.1016/j.jhydrol.2017.03.048>
- Calhoun, F., Baker, D., & Slater, B. (2002). Soils, water quality, and watershed size: Interactions in the Maumee and Sandusky River basins of northwestern Ohio. *Journal of Environmental Quality*, 31(1), 47–53. <https://doi.org/10.2134/jeq2002.4700>
- Cao, P., Lu, C., & Yu, Z. (2018). Historical nitrogen fertilizer use in agricultural ecosystems of the contiguous United States during 1850–2015: Application rate, timing, and fertilizer types. *Earth System Science Data*, 10, 969–984. <https://doi.org/10.5194/essd-10-969-2018>
- Carpenter, S. R., Caraco, N. F., Correll, D. L., Howarth, R. W., Sharpley, A. N., & Smith, V. H. (1998). Nonpoint pollution of surface waters with phosphorus and nitrogen. *Ecological Applications*, 8(3), 559–568. <https://doi.org/10.2307/2641247>
- Cashman, M. J., Wehr, J. D., & Truhn, K. (2013). Elevated light and nutrients Alter the nutritional quality of stream Periphyton. *Freshwater Biology*, 58(7), 1447–1457. <https://doi.org/10.1111/fwb.12142>
- Chand, S., Ashif, M., Zargar, M. Y., & Ayub, B. M. (2011). Nitrate pollution: A menace to human, soil, water and plant. *Universal Journal of Environmental Research & Technology*, 1(1), 22–32.
- Collins, S., Singh, R., Rivas, A., Palmer, A., Horne, D., Manderson, A., Roygard, J., & Matthews, A. (2017). Transport and potential attenuation of nitrogen in shallow Groundwaters in the lower Rangitikei catchment, New Zealand. *Journal of Contaminant Hydrology*, 206, 55–66. <https://doi.org/10.1016/j.jconhyd.2017.10.002>
- Cornwell, K., Norsby, D., & Marston, R. (2003). Drainage, sediment transport, and denudation rates on the Nanga Parbat Himalaya, Pakistan. *Geomorphology*, 55(1), 25–43. [https://doi.org/10.1016/S0169-555X\(03\)00130-2](https://doi.org/10.1016/S0169-555X(03)00130-2)
- Daloğlu, I., Cho, K. H., & Scavia, D. (2012). Evaluating causes of trends in long-term dissolved reactive phosphorus loads to Lake Erie. *Environmental Science and Technology*, 46(19), 10660–10666. <https://doi.org/10.1021/es302315d>
- De Girolamo, A., Pappagallo, G., & Lo Porto, A. (2015). Temporal variability of suspended sediment transport and rating curves in a Mediterranean River basin: The Celone (SE Italy). *Catena*, 128, 135–143. <https://doi.org/10.1016/j.catena.2014.09.020>
- Diamond, D. (2007). QuikChem Method 10-115-01-1-Q Determination of Orthophosphate in Waters by Flow Injection Analysis. [https://www.uvm.edu/bwrl/lab\\_docs/protocols/OrthoP\\_Lachat.pdf](https://www.uvm.edu/bwrl/lab_docs/protocols/OrthoP_Lachat.pdf)
- Dodds, W. K., & Smith, V. H. (2016). Nitrogen, phosphorus, and eutrophication in streams. *Inland Waters*, 6(2), 155–164. <https://doi.org/10.5268/iw-6.2.909>
- Drury, B., Rosi-Marshall, E., & Kelly, J. J. (2013). Wastewater treatment effluent reduces the abundance and diversity of benthic bacterial communities in urban and suburban Rivers. *Applied & Environmental Microbiology*, 79(6), 1897–1905. <https://doi.org/10.1128/AEM.03527-12>
- Edlund, M. B., Triplett, L. D., Tomasek, M. D., & Bartilson, K. (2009). From paleo to policy: Partitioning the historical point and nonpoint phosphorus loads to the St. Croix River, Minnesota-Wisconsin, USA. *Journal of Paleolimnology*, 41(4), 679–689. <https://doi.org/10.1007/s10933-008-9288-1>
- Edwards, L. J., Muller, K. E., Wolfinger, R. D., Qaqish, B. F., & Schabenberger, O. (2008). An  $R^2$  statistic for fixed effects in the linear mixed model. *Statistics in Medicine*, 27(29), 6137–6157. <https://doi.org/10.1002/sim.3429>
- Estrany, J., Garcia, C., & Batalla, R. J. (2008). Groundwater control on the suspended sediment load in the Na Borges River, Mallorca, Spain. *Geomorphology*, 106(3–4), 292–303. <https://doi.org/10.1016/j.geomorph.2008.11.008>
- Ferreira, C. S., Walsh, R. P., & Ferreira, A. J. (2018). Degradation in urban areas. *Environmental Science & Health*, 5, 19–25. <https://doi.org/10.1016/j.coesh.2018.04.001>
- Ford, W. I., King, K., & Williams, M. R. (2018). Upland and in-stream controls on Baseflow nutrient dynamics in tile-drained agroecosystem watersheds. *Journal of Hydrology*, 556, 800–812. <https://doi.org/10.1016/j.jhydrol.2017.12.009>
- Gallo, E. L., Meixner, T., Aoubid, H., Lohse, K. A., & Brooks, P. D. (2015). Combined impact of catchment size, land cover, and precipitation on streamflow and Total dissolved nitrogen: A global comparative

- analysis. *Global Biogeochemical Cycles*, 29(7), 1109–1121. <https://doi.org/10.1002/2015GB005154>
- Gayraud, S., & Philippe, M. (2003). Influence of bed-sediment features on the interstitial habitat available for macroinvertebrates in 15 French streams. *International Review of Hydrobiology*, 88(1), 77–93. <https://doi.org/10.1002/IROH.200390007>
- Gervasio, M. P., Soana, E., Granata, T., Colombo, D., & Castaldelli, G. (2022). An unexpected negative feedback between climate change and eutrophication: Higher temperatures increase denitrification and buffer nitrogen loads in the Po River (northern Italy). *Environmental Research Letters*, 17(8), 084031. <https://doi.org/10.1088/1748-9326/ac8497>
- Giri, S., Qiu, Z., & Zhang, Z. (2018). Assessing the impacts of land use on downstream water quality using a hydrologically sensitive area concept. *Journal of Environmental Management*, 213, 309–319. <https://doi.org/10.1016/j.jenvman.2018.02.075>
- Glińska-Lewczuk, K., Golaś, I., Koc, J., Gotkowska-Plachta, A., Harnisz, M., & Rochwerger, A. (2016). The impact of urban areas on the water quality gradient along a Lowland River. *Environmental Monitoring and Assessment*, 188(11), 1–15. <https://doi.org/10.1007/S10661-016-5638-Z>
- Govenor, H., Krometis, L. A. H., & Hession, W. C. (2017). Invertebrate-based water quality impairments and associated stressors identified through the US clean water act. *Environmental Management*, 60(4), 598–614. <https://doi.org/10.1007/S00267-017-0907-3>
- Hardy, T. B., Panja, P., & Mathias, D. (2005). WinXSPRO: A channel cross section analyzer: User's manual. United States Department of Agriculture, Forest Service, Rocky Mountain Research Station general technical report RMRS-GTR-147. [https://www.fs.fed.us/rm/pubs/rmrs\\_gtr147.pdf](https://www.fs.fed.us/rm/pubs/rmrs_gtr147.pdf)
- Harrelson, C. C., Potyondy, J. P., & Rawlins, C. L. (1994). Stream Channel reference sites: An illustrated guide to field technique. United States Department of Agriculture, Forest Service, Rocky Mountain Research Station general technical report RM-245. [https://www.fs.fed.us/biology/nsaec/fishxing/fplibrary/Harrelson\\_1994\\_Stream\\_Channel\\_Reference\\_Sites\\_An\\_Illustrated.pdf](https://www.fs.fed.us/biology/nsaec/fishxing/fplibrary/Harrelson_1994_Stream_Channel_Reference_Sites_An_Illustrated.pdf)
- Harrington, S. T. (2014). Dissolved and particulate nutrient transport dynamics of a small Irish catchment: The river Owenabue. *Hydrology and Earth System Sciences*, 18, 2191–2200. <https://doi.org/10.5194/hess-18-2191-2014>
- Harrison, X. A., Donaldson, L., Correa-Cano, M. E., Evans, J., Fisher, D. N., Goodwin, C. E., Robinson, B. S., Hodgson, D. J., & Inger, R. (2018). A brief Introduction to mixed effects modelling and multi-model inference in ecology. *PeerJ*, 6, 1–32. <https://doi.org/10.7717/peerj.4794>
- Hashemi, F., Olesen, J. E., Dalggaard, T., & Børgesen, C. D. (2016). Review of scenario analyses to reduce agricultural nitrogen and phosphorus loading to the aquatic environment. *Science of the Total Environment*, 573, 608–626. <https://doi.org/10.1016/j.scitotenv.2016.08.141>
- Heatherly, T. I., Whiles, M. R., & Royer, T. V. (2007). Relationships between water quality, habitat quality, and macroinvertebrate assemblages in Illinois streams. *Journal of Environmental Quality*, 36(6), 1653–1660. <https://doi.org/10.2134/jeq2006.0521>
- Henson, M. W., Hanssen, J., Spooner, G., Fleming, P., Pukonen, M., Stahr, F., & Thrash, J. C. (2018). Nutrient dynamics and stream order influence microbial community patterns along a 2914 kilometer transect of the Mississippi River. *Limnology and Oceanography*, 63, 1837–1855. <https://doi.org/10.1002/lno.10811>
- Hoellein, T. J., Arango, C. P., & Zak, Y. (2011). Spatial variability in nutrient concentration and Biofilm nutrient limitation in an urban watershed. *Biogeochemistry*, 106(2), 265–280. <https://doi.org/10.1007/s10533-011-9631-x>
- Jaeger, B. (2017). r2glmm: Computes R Squared for Mixed (Multilevel) Models. R package version 0.1.2. <https://cran.r-project.org/web/packages/r2glmm/index.html>
- Jiang, G., Lutgen, A., Mattern, K., Sienkiewicz, N., Kan, J., & Inamdar, S. (2020). Streambank legacy sediment contributions to suspended sediment-bound nutrient yields from a mid-Atlantic, Piedmont watershed. *Journal of the American Water Resources Association*, 56(5), 820–841. <https://doi.org/10.1111/1752-1688.12855>
- Jin, G. Q., Xu, J., Mo, Y. M., Tang, H. W., Wei, T., Wang, T. G., & Li, L. (2020). Response to sediments and phosphorus to catchment characteristics and human activities under different rainfall patterns with Bayesian networks. *Journal of Hydrology*, 584, 124695. <https://doi.org/10.1016/j.jhydrol.2020.124695>
- Jones, C. S., & Schilling, K. E. (2011). From agricultural intensification to conservation: Sediment transport in the Raccoon River, Iowa, 1916–2009. *Journal of Environmental Quality*, 40(6), 1911–1923. <https://doi.org/10.2134/jeq2010.0507>
- Koskela, A. I., Fisher, T. R., Sutton, A. J., & Gustafson, A. B. (2018). Biogeochemical storm response in agricultural watersheds of the Choptank River basin, Delmarva Peninsula, USA. *Biogeochemistry*, 139(3), 215–239. <https://doi.org/10.1007/s10533-018-0464-8>
- Lazar, J. A., Spahr, R., Grudzinski, B. P., & Fisher, T. J. (2019). Land cover impacts on storm flow suspended solid and nutrient concentrations in Southwest Ohio streams. *Water Environment Research*, 91(6), 510–522. <https://doi.org/10.1002/wer.1054>
- Lenhart, C. F., Brooks, K. N., Heneley, D., & Magner, J. A. (2010). Spatial and temporal variation in suspended sediment, organic matter, and turbidity in a Minnesota Prairie River: Implications for TMDLs. *Environmental Monitoring and Assessment*, 165(1–4), 435–447. <https://doi.org/10.1007/s10661-009-0957-y>
- Lerch, N. K., Hale, W. F., & Lemaster, D. D. (1980). Soil Survey of Butler County, Ohio. United States Department of Agriculture Soil Conservation Service, 188. [https://www.nrcs.usda.gov/Internet/FSE\\_MANUSCRIPTS/ohio/butlerOH1980/butlerOH1980.pdf](https://www.nrcs.usda.gov/Internet/FSE_MANUSCRIPTS/ohio/butlerOH1980/butlerOH1980.pdf)
- Lizaga, I., Bode, S., Gaspar, L., Latorre, B., Boecks, P., & Navas, A. (2021). Legacy of historic land cover changes on sediment provenance tracked with isotopic tracers in a Mediterranean agroforestry catchment. *Journal of Environmental Management*, 288, 112291. <https://doi.org/10.1016/j.jenvman.2021.112291>
- Miller, J. D., Schoonover, J. E., Williard, K. J., & Hwang, C. R. (2011). Whole catchment land cover effects on water quality in the lower Kaskaskia River watershed. *Water Air and Soil Pollution*, 221(1), 337–350. <https://doi.org/10.1007/s11270-011-0794-9>
- Ohio EPA. (2012). Biological and water quality study of the lower great Miami River and select tributaries. Ohio Environmental Protection Agency biological and water quality reports EAS/2012-5-7. <https://www.epa.ohio.gov/portals/35/documents/GMR2012TSD.pdf>
- Ohio EPA. (2013). National Pollutant Discharge Elimination System (NPDES) Permit Program Fact Sheet Regarding an NPDES Permit To Discharge to Waters of the State of Ohio for Eastern Regional Water Reclamation Facility (WRF).
- Ohio EPA. (2017). National Pollutant Discharge Elimination System (NPDES) Permit Program Fact Sheet Regarding an NPDES Permit To Discharge to Waters of the State of Ohio for Springboro Wastewater Treatment Plant.
- Padowski, J. C., & Gorelick, S. M. (2014). Global analysis of urban surface water supply vulnerability. *Environmental Research Letters*, 9(10), 104004. <https://doi.org/10.1088/1748-9326/9/11/119501>
- Pinheiro, J., Bates, D., DebRoy, S., Sarkar, D., & R Core Team. (2018). nlme: Linear and Nonlinear Mixed Effects Models. R package version 3.1–137. <https://CRAN.R-project.org/package=nlme>
- Pizarro, J., Vergara, P. M., Morales, J. L., Rodríguez, J. A., & Vila, I. (2013). Influence of land use and climate on the load of suspended solids in catchments of Andean Rivers. *Environmental Monitoring and Assessment*, 186(2), 835–843. <https://doi.org/10.1007/s10661-013-3420-z>
- R Core Team. (2023). R: A language and environment for statistical computing. R Foundation for Statistical Computing. <https://www.R-project.org/>
- Raven, P. J., Naura, M., Holmes, N. T., & Dawson, F. H. (2000). Healthy river habitats fit for wildlife: Deriving the physical dimension. *Water*

- and *Environment Journal*, 14(4), 235–239. <https://doi.org/10.1111/j.1747-6593.2000.tb00255.x>
- Rech, A. J., Grudzinski, B., Renwick, W. H., Tenison, C. N., Jojola, M., Vanni, M. J., and Workman, R. (2018). Legacy deposits, milldams, water quality, and environmental change in the four Mile Creek watershed, southwestern Ohio. *Ancient Oceans, Orogenic Uplifts, and Glacial Ice: Geologic Crossroads in America's Heartland: Geological Society of America Field Guide* 51, 113–144. [https://doi.org/10.1130/2018.0051\(05\)](https://doi.org/10.1130/2018.0051(05))
- Renwick, W. H., Vanni, M. J., & Zhang, Q. (2008). Water quality trends and changing agricultural practices in a Midwest U.S. watershed, 1994–2006. *Journal of Environmental Quality*, 37(5), 1862–1874. <https://doi.org/10.2134/jeq2007.0401>
- Richards, R. P., Baker, D. B., Crumrine, J. P., Kramer, J. W., Ewing, D. E., & Merryfield, B. J. (2008). Thirty-year trends in suspended sediment in seven Lake Erie tributaries. *Journal of Environmental Quality*, 37(5), 1894–1908. <https://doi.org/10.2134/jeq2007.0590>
- Riseng, C. M., Wiley, J. M., Black, R. W., & Munn, M. D. (2011). Impacts of agricultural land use on biological integrity: A causal analysis. *Ecological Applications*, 21(8), 3128–3146. <https://doi.org/10.1890/11-0077.1>
- Rodríguez Benítez, A. J., Gómez, A. G., & Díaz, C. A. (2015). Definition of mixing zones in Rivers. *Environmental Fluid Mechanics*, 16(1), 209–244. <https://doi.org/10.1007/s10652-015-9425-0>
- Russell, K. L., Vietz, G. J., & Fletcher, T. D. (2017). Global sediment yields from urban and urbanizing watersheds. *Earth-Science Reviews*, 168, 73–80. <https://doi.org/10.1016/j.earscirev.2017.04.001>
- Sánchez-Pérez, J., Gerino, M., Sauvage, S., Dumas, P., Maneux, E., Julien, F., Winterton, P., & Vervier, P. (2009). Effects of wastewater treatment plant pollution on in-stream ecosystems functions in an agricultural watershed. *International Journal of Limnology*, 45(2), 79–92. <https://doi.org/10.1051/limn/2009011>
- Sandercock, P. J., & Hooke, J. M. (2010). Assessment of vegetation effects on hydraulics and of feedbacks on plant survival and zonation in ephemeral channels. *Hydrological Processes*, 24(6), 695–713. <https://doi.org/10.1002/hyp.7508>
- Sarremejane, R., Cañedo-Argüelles, M., Prat, N., Mykrä, H., Muotka, T., & Bonada, N. (2017). Do Metacommunities vary through time? Intermittent Rivers as model systems. *Journal of Biogeography*, 44(12), 2752–2763. <https://doi.org/10.1111/jbi.13077>
- Shore, M., Murphy, S., Mellander, P. E., Short, G., Melland, A. R., & Crockford, L. (2017). Influence of storm Flow Base flow phosphorus pressures on stream ecology in agricultural catchments. *Science of the Total Environment*, 590–591, 469–483. <https://doi.org/10.1016/j.scitotenv.2017.02.100>
- Skarbovik, E., Stalnacke, P., Bogen, J., & Bonsnes, T. (2012). Impact of sampling frequency on mean concentrations and estimated loads of suspended sediment in a Norwegian River: Implications for water management. *Science of the Total Environment*, 433, 462–471. <https://doi.org/10.1016/j.scitotenv.2012.06.072>
- Stets, E., Kelly, V., & Crawford, C. (2015). Regional and temporal differences in nitrate trends discerned from long-term water quality monitoring data. *Journal of the American Water Resources Association*, 51(5), 1394–1407. <https://doi.org/10.1111/1752-1688.12321>
- Turner, R. E., & Rabalais, N. N. (2003). Linking landscape and water quality in the Mississippi River basin for 200 years. *Bioscience*, 53(6), 563–572. [https://doi.org/10.1641/0006-3568\(2003\)053\[0563:LLAWQI\]2.0.CO;2](https://doi.org/10.1641/0006-3568(2003)053[0563:LLAWQI]2.0.CO;2)
- USDA. (2010). Field Crops: Usual Planting and Harvesting Dates. United States Department of Agriculture, National Agricultural Statistics Service, Agricultural Handbook 628. <https://downloads.usda.library.cornell.edu/usdaesmis/files/vm40xr56k/dv13zw65p/w955297d/planting-10-29-2010.pdf>
- Volf, G., Atanasova, N., Kompare, B., & Ožanić, N. (2013). Modeling nutrient loads to the northern Adriatic. *Journal of Hydrology*, 504(11), 182–193. <https://doi.org/10.1016/j.jhydrol.2013.09.044>
- Weiss, M. P., & Sweet, W. C. (1964). Kope formation (upper Ordovician): Ohio and Kentucky. *Science*, 145(3638), 1296–1302. <https://doi.org/10.1126/science.145.3638.1296>
- Wendt, K. (2000). QuikChem Method 10-107-04-1-A Determination of Nitrate/Nitrite in Surface and Wastewaters by Flow Injection Analysis. [https://www.uvm.edu/bwrl/lab\\_docs/protocols/Nitrate\\_water\\_lachat.pdf](https://www.uvm.edu/bwrl/lab_docs/protocols/Nitrate_water_lachat.pdf)
- Yamashita, T., & Yamamoto-Ikemoto, R. (2014). Nitrogen and phosphorus removal from wastewater treatment plant effluent via bacterial sulfate reduction in an anoxic bioreactor packed with wood and iron. *International Journal of Environmental Research and Public Health*, 11(9), 9835–9853. <https://doi.org/10.3390/ijerph110909835>

## SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

**How to cite this article:** Spahr, R. E., Lazar, J. A., Grudzinski, B. P., & Fisher, T. J. (2024). Land cover, stream discharge, and wastewater effluent impacts on baseflow sediment and nutrient concentrations in SW Ohio streams. *River Research and Applications*, 40(4), 497–507. <https://doi.org/10.1002/rra.4248>