doi:10.2489/jswc.75.3.319

# Characterizing transport of natural and anthropogenic constituents in a long-term agricultural watershed in the northeastern United States

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**Abstract:** As recent technologies enable water samples to be collected at increasingly shorter time intervals, water quality data can more fully capture the range of conditions a stream experiences over time. Various metrics can be employed with the large, high-temporal resolution (i.e., subdaily) data sets to gain insights into the hydroclimatic and biogeochemical processes affecting chemical fate and transport. These insights can be helpful in understanding the extent to which anthropogenic activities have impacted the natural response of some constituents, such as nutrients and salts, in managed landscapes. Here, nearly four years (12,544 samples from 2015 to 2019) of water quality data for 12 constituents of interest were collected using three sampling strategies: (1) low-frequency sampling three times per week, (2) high-temporal resolution flow-paced sampling to capture stormflow, and (3) high-frequency time-paced sampling with a time interval of four hours. Seasonal trends were investigated to understand concentration variability over time, and concentration-discharge (C-Q) relationships were developed to categorize the transport dynamics of each constituent. Lorenz curves and Gini coefficients were employed to quantify the temporal inequality of the constituent loads discharged at the watershed outlet and understand the extents to which the transport behaviors of geogenic constituents and those affected by anthropogenic activities differed. Overall, the results suggested that nearly all of the geogenic constituents, plus nitrate-nitrogen (NO<sub>3</sub>-N) and sulfate (SO<sub>4</sub>-S), exhibited chemostatic dynamics with loads overwhelmingly controlled by flow variability, whereas aluminum (Al), iron (Fe), nitrite-N (NO,-N), and orthophosphate (PO<sub>4</sub>-P) exhibited episodic transport dynamics that were likely controlled by limited source availability. Since the transport of NO<sub>2</sub>-N was found to be similar to the transport of common geogenic constituents for the region, this suggests that decades of agricultural activities in the watershed have led to the emergence of legacy N sources, while the episodic dynamics observed for PO<sub>4</sub>-P suggest that best management practices appear to have prevented the emergence of P legacy sources.

**Key words:** agriculture—CEAP—concentration-discharge—geogenic constituents—legacy nutrients—surface water

The temporal resolutions over which surface water samples are collected can profoundly affect our understanding of how water quality changes over time, uncover the benefits or limitations of best management practices, and provide insight into whether or not legacy nutrient sources are present. Traditional water sampling strategies have been based on routine grab sampling methodologies that involve a person physically collecting water samples at

a point of interest at regular intervals of time (i.e., bi-weekly, monthly, or quarterly). More recent modifications of time-interval water quality sampling involve deployment of automated samplers that can be programmed to collect grab samples at shorter time intervals (i.e., hours, days, or weeks) based on time interval or flow stage, such that all filled bottles in the sampler can be physically retrieved in a single visit. If continued for sufficiently long periods of time (i.e., years to decades),

these sampling strategies can provide for a robust analysis of how baseflow water quality changes over time in response to changes in land use or land management.

In an effort to improve our understanding of the environmental benefits of past agricultural conservation practices and to maximize the contributions of current and future conservation practices, the USDA Natural Resources Conservation Service (NRCS) Conservation Effects Assessment Project (CEAP; USDA NRCS 2019) has promoted ongoing watershed monitoring and modeling efforts in CEAP watersheds and the more recently established research scenarios of the Long-term Agroecosystem Research Network (LTAR; USDA ARS 2019). In conjunction with these efforts, temporal sampling strategies have been evaluated for practicality and effectiveness at the plot (2.2  $\times$  10<sup>-4</sup> ha) and small watershed scale (<25 km<sup>2</sup>) (King and Harmel 2004; King et al. 2005). The most effective sampling strategies would capture the full range of a vast array of constituent concentrations and ensure both baseflow and stormflow dynamics are appropriately captured.

The temporal resolution at which samples are collected is of critical importance in development of accurate concentration-discharge (C-Q) relationships, which provide valuable insights into the extent to which hydrologic and biogeochemical variability are driving the transport dynamics of a constituent of interest (Basu et al. 2011; Thompson et al. 2011) and to calculate loads with minimal errors (Harmel et al. 2003). If a constituent of interest increases during surface runoff events, and high flow events are not adequately captured, then the true ranges of concentrations and loads are underestimated, with annual load underestimation being as high as 40% (Haggard et al. 2003; Ullrich and Volk 2010). However, storm-event sampling

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can generate data sets that bias high concentrations during periods of high flow, causing load overestimation (Robertson and Roerish 1999). In order to reduce bias in load calculations, concentration data that represent both stormflow and baseflow conditions in appropriate proportions are needed. Flowpaced sampling as opposed to time-paced can provide such a data set, but only if the volume of flow used to trigger the collection of a water quality sample adequately captures chemograph changes that may happen during a storm event (Harmel et al. 2003; Gall et al. 2010). Samples need to capture the rising and recession limbs of hydrographs, as well as the hydrograph peak, in order to prevent missing elevated concentrations for solutes that exhibit accretion dynamics and low concentrations for solutes that exhibit dilution dynamics.

Recent advances in data analysis techniques seek to leverage water quality data to construct C-Q relationships that can facilitate characterization of the transport dynamics of chemicals of interest (Godsey et al. 2009; Thompson et al. 2011; Gall et al. 2015; Musolff et al. 2015). One method utilizes the power function to relate concentration (C) and discharge (Q),  $C = aQ^b$ , where a and b are empirical constants (Haygarth et al. 2004; Vogel et al. 2005). Constituents originating from small, hydrologically unavailable or disconnected sources within a catchment typically exhibit high temporal variabilities in concentration when they reach the stream (Thompson et al. 2011), while catchments with large, hydrologically available or connected sources generate relatively constant biogeochemical responses, with low variability in concentrations observed at the catchment scale (Godsey et al. 2009). The former scenario is referred to as an "episodic" response, with high concentrations occurring during short periods of time (i.e., storm events), such that these short periods of time are disproportionately more important to overall transport behavior than other periods of time (i.e., baseflow conditions). The latter scenario produces C-Q relationships with  $b\sim0$ , such that the response is categorized as "chemostatic."

Chemostatic conditions are common for geogenic constituents, such as magnesium (Mg) and calcium (Ca), which weather from large geologic sources within the landscape (Godsey et al. 2009). However, the addition of anthropogenic sources of solutes that have low natural

sources in the landscape, such as fertilizer and manure inputs of nitrogen (N) and phosphorus (P), can create large legacy sources that overwhelm the natural variability, causing chemostatic responses to emerge (Basu et al. 2010, 2011; Thompson et al. 2011). When transport dynamics are chemostatic, hydrologic variability is the dominant control on solute export, while biogeochemical variability (i.e., source availability) is the dominant control for solutes exhibiting episodic behavior (Thompson et al. 2011). These responses are neither solute-specific nor catchment-specific, as catchments that exhibit chemostatic responses for some solutes may exhibit episodic responses for others and vice versa (Johnson et al. 1969; Mańczak and Florczyk 1971; Duffy and Cusumano 1998; Godsey et al. 2009; Basu et al. 2010; Thompson et al. 2011; Halliday et al. 2012). Additionally, these responses may be controlled by natural processes or result from long-term changes in intensive management practices.

Here, we leverage nearly four years of water quality data (12,544 total water samples across three sampling regimes) at a headwater watershed in northeastern United States to explore the effects of the temporal resolution at which the samples were collected on the development of C-Q relationships and annual load calculations. Twelve constituents of interest are included in the analysis: chloride (Cl), nitrate-N (NO<sub>2</sub>-N), nitrite-N (NO<sub>2</sub>-N), orthophosphate-P (PO<sub>4</sub>-P), sulfate (SO<sub>4</sub>-S), aluminum (Al), Ca, iron (Fe), potassium (K), Mg, sodium (Na), and silica (Si). These constituents represent a range of geogenic constituents expected to be driven largely by natural weathering processes and constituents whose natural biogeochemical cycles have been impacted by agricultural and other anthropogenic activities occurring in the watershed (e.g., fertilizer application and road salt). C-Q relationships were developed for each constituent using data collected with three sampling strategies: (1) three times per week; (2) high-temporal resolution flow-paced sampling at a stage change of 2.54 cm to capture stormflow; and (3) time-paced sampling with a time interval of four hours. These intervals were selected based on previous data collected from the compound weir at the site prior to the commencement of this experiment. Two years of rainfall and stream depth changes were used to calculate the average stream depth change (30.5 cm) during storm events. The threshold

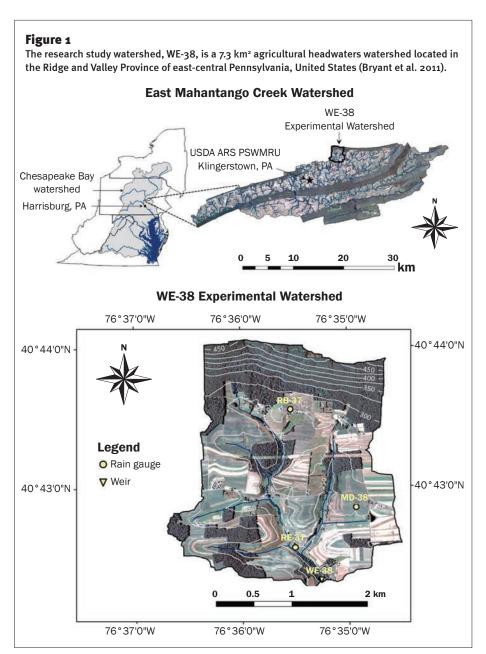
of 2.54 cm for triggering samples was chosen to best capture the rising and recession limbs of the hydrograph. For larger storm events, a four-hour interval was found to be frequent enough to capture the remainder of the hydrograph recession limb.

Results of the analyses presented here demonstrate the extent to which current management practices adopted within the watershed have been successful at improving or protecting water quality at the watershed scale. Additionally, this study highlights important implications for understanding flow conditions under which best management practices must be effective in order to meet water quality restoration goals for impaired water bodies, and provides insights into the extent to which nutrient legacy sources will need to be reduced before water quality improvements can be realized.

### **Materials and Methods**

Study Site. This research was conducted in a USDA Agricultural Research Service (ARS) experimental watershed known as WE-38. WE-38 is a 7.3 km<sup>2</sup> subcatchment within the Mahantango Creek (420 km²) watershed located in the Ridge and Valley Province of east-central Pennsylvania (figure 1). The climate in the watershed is temperate, with average annual precipitation of 1,080 mm and mean temperature of ~10°C (Bryant et al. 2011). Underlying geology is dominantly comprised of sandstone at higher elevations and shale and siltstone in valley bottoms (Gburek and Folmar 1999). The higher elevation soils are well drained, while the lower lands are typically poorly drained with perched water tables and frequently generate saturation excess runoff.

Within WE-38, about 39% of the land area is mature forest; 3% is permanent pasture; 2% is developed; and 55% is in silviculture, commercial vegetable production, or, most predominately, forage crop rotations. Agriculture is the dominant land use in the shale and siltstone portions of the watershed, while forests are dominant in the higher elevation sandstone regions (Schnabel et al. 1993). Field sizes average 1.1 ha, with a range of 0.01 ha to 14 ha, and have not varied substantially in size in the past 15 years (Veith et al. 2015). On the steep hillslopes, fields are narrow, farmed on-the-contour, and are in alternating rotations of corn (Zea mays L.) or soy (Glyzine max [L.] Merr.) with hay or alfalfa (Medicago sativa L.) so that



the hillslopes are essentially strip cropped. Primary agricultural activities in the watershed and surrounding regions include nonintensive beef, dairy, and swine farming plus supportive forage cropping. More details about specific agricultural practices are provided in Liu et al. (2017). Manure accounts for about 20% of total N and 30% of total P applications in the watershed, with about 50% of it being applied in the late fall and winter. The effects of the timing of manure application on water quality in the watershed are evaluated and discussed by Liu et al. (2017). Remaining agronomic requirements are met through chemical fertilizer applications, primarily applied in the spring.

Hydrologic Data Collection. Water level in the stream has been measured at five-minute intervals at the main outlet of the watershed (Buda et al. 2011) since 1968. The gauging station contains a compound weir, such that low flows are measured through a V-notch (90°) sharp-crested weir and high flows are measured through a 5:1 broad crested weir. The stage heights were monitored in separate stilling wells using a float and shaft encoder (Campbell Scientific CS410) connected to a datalogger (Campbell Scientific CR1000X, Logan, Utah).

Water Quality Sampling and Analysis. During the study period, water quality data were collected at the watershed outlet using three strategies: (1) low-frequency sampling

three times per week, (2) high-temporal resolution flow-paced sampling to capture stormflow, and (3) high-frequency timepaced sampling with a time interval of four hours. The samples collected using the first strategy were collected 130 m upstream of the weir manually by field technicians. Samples for the second sampling strategy were triggered by the datalogger when water level in the stream, as measured by the shaft encoders, changed by 2.54 cm. After the 2.54 cm rise in water level, samples were triggered every half hour until the storm receded. Samples for the third sampling strategy were triggered by the datalogger based on time (every four hours). The samples triggered by the datalogger were collected by an automated sampler (Hach AS950, Loveland, Colorado). All samples were collected in 500 mL high density polyethylene (HDPE) bottles and were handled as described in Church et al. (2011).

Upon collection from the field, each 500 mL sample was split into two portions and immediately transported on ice from Klingerstown, Pennsylvania, to University Park, Pennsylvania. One portion of each sample was left unfiltered and analyzed by inductively coupled plasma (ICP) for Al, Ca, Fe, K, Mg, Na, and Si. The remainder of each sample was passed through a 0.45 µm filter and analyzed for Cl, NO<sub>3</sub>-N, NO<sub>2</sub>-N, PO<sub>4</sub>-P, and SO<sub>4</sub><sup>2</sup>-S by ion chromatography (ICS-5000; Thermo-Fisher Dionex, Sunnyvale, California).

The method limit of detection (LOD) and quantification (LOQ) for each constituent of interest are 0.01 mg L<sup>-1</sup> and 0.1 mg L<sup>-1</sup>, respectively. The LOD was determined by a signal to noise ratio of 3 and the LOQ was determined by a signal to noise ratio of 10. Lab blanks were included in each instrument run to ensure no carryover between samples, and check standards were run every 40 samples. Further, one sample per instrument run was run three times to ensure the instrument was performing accurately. No gap filling was conducted, as the percentage of missing or erroneous data was small (<1%) for the flow data and constituents of interest.

**Development** and Interpretation of Concentration-Discharge Relationships. C-Q relationships using the empirical relationship  $C = aQ^b$  (Haygarth et al. 2004; Vogel et al. 2005) were developed for each of the constituents of interest for each of the sampling strategies. When plotted on a log-log scale, the slope of the trendline,  $\log(C) = b$ 

 $\times \log(Q) + a$ , can be used to determine the value of b for each relationship. The b value can then be used to characterize the C-Q pattern as accretion (b > 0), dilution (b < 0), or chemostatic ( $b \sim 0$ ). These patterns were used to provide insight into the transport dynamics exhibited by the constituents of interest throughout the study period.

It is possible for constituents with significant variability in their concentrations to show no relationship with flow (i.e.,  $b \sim 0$ ). Therefore, to further understand the role of hydrologic variability versus concentration variability in the transport of each of the constituents of interest, a ratio of the coefficient of variation of concentration (CVc) versus the coefficient of variation of discharge (CV<sub>Q</sub>) was employed following Thompson et al. (2011). The higher the ratio of CVc/ CVQ, the more concentration variability drives loads and transport from a watershed. However, when concentration variability is low, the variability in flow is much larger and the CVc/CVq, ratio is low. After analyzing solute and discharge data from nine experimental watersheds (31 subwatersheds) across the United States, Thompson et al. (2011) suggested that constituents having CVc/ CVQ values of 0.30 or less indicated chemostatic behavior, with hydrologic variability controlling constituent export dynamics. Constituents with CVc/CVQ values greater than 0.30 exhibit episodic transport, such that both flow variability and concentration variability play important roles in determining transport to the watershed outlet. The higher the value of CVc/CVo for a constituent, the more important that variations in concentration are to its export dynamics (Thompson et al. 2011; Gall et al. 2013; Musolff et al. 2015).

Flux and Load Calculations. For each sampling strategy, fluxes were calculated by multiplying constituent concentrations and the flowrate observed at the time the sample was collected. When concentrations were below the LOD, loads were considered to be zero. When concentrations were between the LOD and LOQ, concentrations equal to half the LOQ were used to calculate the flux. Fluxes were assumed to be constant for the time between sampling events and were utilized to calculate loads for each sampling strategy. Given the relatively high frequency of data collected, even for the lowest temporal resolution sampling strategy (every few days over a four-year period), the longest time between sampling periods was five days. Exceptions included approximately a month toward the end of 2017 when the weir was being cleaned and repaired and another month from the end of December of 2018 to early February of 2019 when the US government was shut down and no field or laboratory work could be conducted. Data were not extrapolated during these time periods and were excluded from the load calculations.

Temporal Inequality Calculations for Discharge and Loads. Lorenz inequality curves and their corresponding Gini coefficients were developed for loads calculated from the time-paced samples following methodology described by Jawtiz and Mitchell (2011) and Gall et al. (2013). While commonly used in economics to quantify wealth inequality, the method is gaining attraction in hydrologic and water quality research to quantify the temporal inequality of loads and to understand the relative importance of low flow and high flow events to the observed loads (Jawitz and Mitchell 2011; Gall et al. 2013; Masaki et al. 2014; Musolff et al. 2015; Opalinski et al. 2016). Lorenz inequality curves are generated by sorting the load data in ascending order and plotting the cumulative fraction of load observed over the study period versus the cumulative fraction of time over which the loads were exported.

The Gini coefficient is a formal measure of inequality that varies from 0 to 1 and is calculated as the ratio of the area between the line of equality and the Lorenz curve to the area under the line of equality (Jawitz and Mitchell 2011; Gall et al. 2013). A value of 0 indicates that all moments in time contribute equally to discharge and load (i.e., discharge and loads are constant). At the other extreme, a Gini coefficient value of 1 indicates perfect inequality, such that all of the discharge and solute loads are exported during one instant in time. The Gini coefficient values for constituent loads (GL) can be compared to the values for discharge (GQ). A ratio of  $GL/GQ \sim 1$  indicates that discharge is the dominant driver of the observed loads, while GL/GQ > 1 indicates that load inequality is much greater than discharge inequality. The value of GL/GQ is approximately 1 when  $b \sim 0$  and CVc/CVQ is low (i.e.,  $\sim 0.30$  or less). Constituents with b > 0 typically have high CVc/CVo ratios (>0.30) and GL/Go > 1, whereas constituents with b < 0 typically have  $G_L/G_Q < 1$ (Gall et al. 2013). A visual method of interpreting the results of the *b* and CVc/CVQ results is provided by Musolff et al. (2015). Therefore, each of these metrics can be used to characterize the transport dynamics of constituents of interest in a catchment.

### **Results and Discussion**

Each of the sampling strategies produced a different number of samples during the study period, with the three times per week strategy producing ~500 samples, the stormflow sampling producing nearly 5,000 samples, and the time-paced sampling (every four hours) producing more than 7,000 samples (table 1). The extent to which the ranges of flow were sampled with each sampling strategy varied widely (figure 2a). The flow duration curve for time-paced samples, which were collected every four hours, very closely matched the flow duration curve for the complete five-minute flow data over the study period (figure 2b). In contrast, flow duration curves for the samples collected three times per week underrepresented high flow events and overrepresented baseflow events. Conversely, the flow duration curves for stormflow sampling overrepresented the high flow events and underrepresented low flow events.

Summary of Water Quality Data. The percentage of samples with concentrations above the LOQ were generally above 95%, with NO2-N, PO4-P, Al, and Fe below the LOQ for the vast majority (>90%) of samples collected (table 1). The different sampling strategies tended to censor the data at approximately the same levels, with the percentage of samples from each sampling strategy below the LOD and LOQ remaining fairly consistent. Overall, the time-paced sampling strategy captured higher concentrations for the nutrients (N species, PO<sub>4</sub>-P, SO<sub>4</sub>-S, and K) and Cl, while the streamflow samples captured higher concentrations for all other constituents.

The concentrations observed during the study period exceeded the recommended levels for cattle drinking water use (20 mg N L<sup>-1</sup> for NO<sub>3</sub> + NO<sub>2</sub>-N and 0.3 mg L<sup>-1</sup> for Fe; Swistock 2015) in less than 1% of the samples collected. Calcium, Cl, Mg, Na, and SO<sub>4</sub>-S were below the recommended levels (500 mg L<sup>-1</sup>, 250 mg L<sup>-1</sup>, 500 mg L<sup>-1</sup>, 20 mg L<sup>-1</sup>, and 334 mg S L<sup>-1</sup> respectively; Swistock 2015) for the entire study period. Nitrate plus NO<sub>2</sub>-N exceeded the primary drinking water standard (10 mg N

 $L^{-1}$ ; USEPA 2019) in approximately 20% of the samples collected.

Seasonal trends were examined by combining data from all three sampling strategies into three-month subsets over the periods of record (figure 3). The highest concentrations for Al, Cl, Fe, Na, and Si were observed in the winter months, while the highest NO<sub>3</sub>-N concentrations were observed in the spring months; K, NO<sub>2</sub>-N, PO<sub>4</sub>-P, and SO<sub>4</sub>-S were highest in the summer, and Ca was highest in the fall. Application of road salts in the winter months likely contributed to elevated concentrations of Cl from January through March relative to concentrations observed in the summer and fall. Elevated levels of NO<sub>3</sub>-N observed during the spring and fall reflect the timing of manure and fertilizer applications to agricultural fields within the watershed. Although PO<sub>4</sub>-P was generally below 1 mg L<sup>-1</sup>, concentrations were greater than 3 mg P L<sup>-1</sup> for a several-week period at the end of June through the beginning of July of 2018 during a window of time that was identified by the National Climatic Data Center as being the wettest on record (NOAA 2019). Overall, concentrations for most of the constituents of interest were fairly consistent over time with similar averages across all seasons (figure 3), suggesting that their transport through the watershed likely depends heavily on hydrologic responses of this watershed, which are largely driven by variable source area saturation excess runoff (Buda et al. 2013).

Concentration-Discharge Relationships. C-Q relationships illustrate how well the different sampling strategies captured concentration ranges and variability of each constituent (figure 4), with the time-paced sampling strategy best representing the full range of observed flows while the three times per week strategy missed the highest flow events and the stormflow sampling missed low flow conditions. Slopes of the C-Q relationships generated by the timepaced sampling strategy were calculated to characterize the relationship between concentration and flow as a dilution pattern (b <0), an accretion pattern (b > 0), or independent of the flowrate  $(b \sim 0)$  (table 2).

Nearly all of the constituents of geologic origin (Ca, Cl, K, Mg, Na, and Si) exhibited weak dilution patterns or relatively chemostatic responses (*b* values between –0.09 and 0.06; table 2). These results are consistent with earlier monitoring studies conducted at

 Table 1

 Data summary for constituents of interest collected for each sampling strategy.

	3× weekly	Stormflow	Time-paced (4 h)
Constituent	n = 472	n = 4,968	n = 7,104
Chloride (mg L <sup>-1</sup> )	7.15 to 34.74	3.05 to 51.05	0.16 to 104.10
	460 > LOD	4,878 > LOD	6,779 > LOD
	460 > LOQ	4,877 > LOQ	6,777 > LOQ
Nitrate-N (mg L <sup>-1</sup> )	0.73 to 11.53	0.27 to 21.97	0.12 to 22.46
	472 > LOD	4,882 > LOD	6,787 > LOD
	472 > LOQ	4,882 > LOQ	6,781 > LOQ
Nitrite-N (mg L <sup>-1</sup> )	_	0.10 to 1.36	0.10 to 1.65
	0 > LOD	2,250 > LOD	2,771 > LOD
	0 > LOQ	454 > LOQ	577 > LOQ
Orthophosphate-P (mg L <sup>-1</sup> )	_	0.10 to 3.40	0.10 to 4.82
	4 > LOD	856 > LOD	706 > LOD
	0 > LOQ	104 > LOQ	72 > LOQ
Sulfate-S (mg L <sup>-1</sup> )	1.75 to 5.93	0.1 to 7.91	0.64 to 8.97
	460 > LOD	4,884 > LOD	6,757 > LOD
	460 > LOQ	4,881 > LOQ	6,753 > LOQ
		n = 3,590	n = 4,637
Aluminum (mg L <sup>-1</sup> )		0.10 to 3.80	0.11 to 1.04
		540 > LOD	92 > LOD
		288 > LOQ	37 > LOQ
Calcium (mg L <sup>-1</sup> )		5.74 to 23.10	8.15 to 22.12
		3,588 > LOD	4,635 > LOD
		3,588 > LOQ	4,635 > LOQ
Iron (mg L <sup>-1</sup> )		0.105 to 2.19	0.11 to 0.74
		1,062 > LOD	490 > LOD
		391 > LOQ	69 > LOQ
Potassium (mg L <sup>-1</sup> )		0.17 to 7.46	0.18 to 50.60
		3,590 > LOD	4,637 > LOD
		3,590 > LOQ	4,637 > LOQ
Magnesium (mg L <sup>-1</sup> )		1.62 to 7.66	2.69 to 8.06
		3,588 > LOD	4,635 > LOD
		3,588 > LOQ	4,635 > LOQ
Sodium (mg L <sup>-1</sup> )		0.54 to 21.60	0.55 to 18.90
		3,589 > LOD	4,637 > LOD
		3,589 > LOQ	4,636 > LOQ
Silica (mg L <sup>-1</sup> )		1.18 to 6.49	0.27 to 5.52
		3,588 > LOD	4,636 > LOD
		3,588 > LOQ	4,636 > LOQ

Notes: N = nitrogen. P = phosphorus. S = sulfur. LOD = limit of detection (0.01 mg  $L^{-1}$ ). LOQ = limit of quantification (0.1 mg  $L^{-1}$ ).

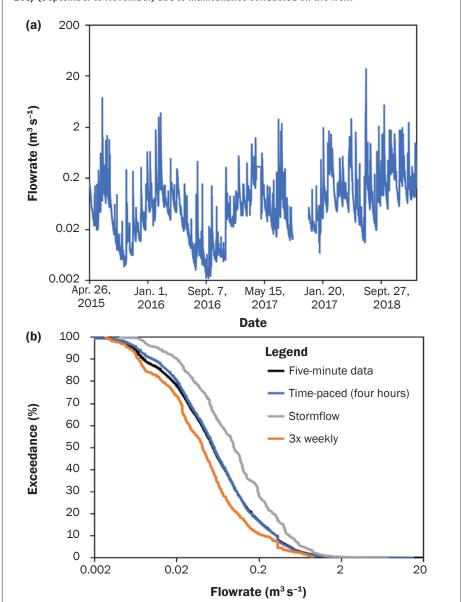
the study site, which found Ca, Cl, Mg, and Na to be among the dominant ions present in samples collected (Gburek and Folmar 1999) at the watershed outlet. Given that the dominant parent materials at the study site include sandstone, siltstone, and shale (Schnabel et al. 1993), the high occurrence of these constituents is expected. Additionally, nutrients such as SO<sub>4</sub>-S, NO<sub>3</sub>-N, and K exhibited concentrations that remained relatively constant over the full range of flowrates observed at

the site, likely due to anthropogenic fertilizer inputs to the landscape or atmospheric deposition that can overwhelm the natural variability associated with background levels of these constituents (Elkin et al. 2016; Basu et al. 2010).

In contrast, Al and Fe were typically below the LOD during low flow conditions, such that the C-Q relationships appear to generate chemostatic ( $b \sim 0$ ) patterns despite the low number of samples that were pres-

Figure 2

(a) Hydrograph for period of study (April of 2015 to February of 2019) using discharge measurements made at a five-minute interval; and (b) flow duration curves (FDCs) for flowrates observed during the study period at five-minute intervals and corresponding to samples collected during each of the three sampling strategies: (1) three times per week, (2) stormflow sampling, and (3) time-paced sampling (every four hours). Note that discharge data were not collected in fall of 2017 (September to November) due to maintenance conducted on the weir.



ent above the LOQ. The mobilization of Al and Fe in the landscape was likely episodic, as the *C*-Q relationships showed that the high concentrations of both Al and Fe were associated with high flow conditions (figure 4). Therefore, Al and Fe appeared to be dominantly transported during surface runoff events, although the source of these constituents to be mobilized with surface runoff was unclear. It is unlikely that Al and Fe were present in the parent

materials at substantial enough quantities or that the reduction-oxidation conditions that would promote mobilization occurred for long enough periods of time for chemostatic relationships between concentration and discharge to emerge. Rather, these results are consistent with the variable source area hydrology observed to occur in the watershed in which activation of runoff from the high-elevation areas occurs infrequently and

only during large storm events (Schnabel et al. 1993; Buda et al. 2009; Buda 2013).

CVc/CVq ratios (table 2) indicated that nearly all of the constituents of interest can be classified as exhibiting "chemostatic" conditions with nearly constant concentrations throughout the study period, as the ratios for nearly all of the constituents except PO<sub>4</sub>-P and Al were less than 0.30 (Thompson et al. 2011). For the constituents with low frequency of detection, we note that low CVc/ CVQ ratios do not imply chemostatic conditions as their infrequent detection causes transport dynamics to be episodic. Variability in flow was much greater than concentration variability for the majority of the constituents analyzed, suggesting that hydrologic variability overwhelmingly dominates their transport dynamics from the watershed. The low CVc/CVq ratio for NO<sub>2</sub>-N was consistent with results from other dominantly agricultural watersheds (Basu et al. 2010; Guan et al. 2011; Thompson et al. 2011; Gall et al. 2015; Zhang 2018), and suggests that legacy N sources may have built up in the landscape over decades of agricultural activities within this watershed. This is further supported by a change in the C-Q relationship for NO<sub>3</sub>-N, which used to exhibit a positive relationship between flow and discharge several decades ago (1984 to 1987), but now shows much less variability across flow rates, suggesting that groundwater concentrations have increased over the past 30 years of continued agricultural activities in the watershed (Schnabel et al. 1993).

Inequality in Exported Loads. Lorenz curves (figure 5) and corresponding GL values (table 2) further confirmed that for nearly all of the constituents of interest, flow variability was the dominant driver of the loads observed, with the temporal inequality of flow and most of the constituent loads nearly equal ( $GL/GQ \sim 1$ ; table 2). Consistent with the other metrics employed, GL/GQ ratios were higher for  $NO_2$ –N,  $PO_4$ –P, AI, and AI, indicating that the temporal inequality of loads for these constituents was higher than the temporal inequality of the flow.

Examining the temporal manner in which loads were cumulatively discharged from the watershed outlet (figure 6) provided insight into the "hot moments" that were most responsible for each of the constituents of interest. The patterns of cumulative load exported for NO<sub>3</sub>-N, Cl, and SO<sub>4</sub>-S most closely tracked the cumulative export of flow,

Figure 3 Boxplot by season for each constituent of interest. Concentrations represent values from the combined data set of all samples. Note: y-axis scales differ by constituent. No-1.5-1.0-1.0-0.5-(mg L<sup>-1</sup>) 80 60 40 ಽ 20 Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept PO<sub>4</sub>-P (mg L<sup>-1</sup>) -S (mg L-1) 4 AI (mg L<sup>-1</sup>) 6 3 4 2 1 Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. **Le (mg L-1**2.0
1.0
1.0
0.5 3a (mg L<sup>-1</sup>) **(mg l**-1 30 20 10 20 30 15 10 Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept 20 Na (mg L-1) Mg (mg  $\mathsf{L}^{-1}$ ) Si (mg L<sup>-1</sup>) 5 -4 -3 -5 4 2n Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept. Oct. to Dec. Jan. to Mar. Apr. to June July to Sept

which is consistent with the GL/GQ ratios all being close to 1.0, meaning that the temporal inequality in flow and loads for those constituents were nearly equal. In contrast, the highest PO<sub>4</sub>-P concentrations were observed from June 20 to July 10, 2018, during which ~40% of the PO<sub>4</sub>-P load observed over the entire study period was mobilized. May of 2018 had been extremely wet in the northeastern part of the United States, and a break in the record high rainfall events occurred early in June. This was likely the earliest farmers could get onto their fields to spread manure, and therefore the high concentrations observed later in the month were likely due to recent manure applications. The greatest load exported for NO2-N (~30% of the load exported during the study period) was during a two-week period in January of 2018 that likely consisted of a rain-on-snow melting event. Aluminum and Fe also exhibited several "hot moments" in winter and spring months, such that short periods of time in December of 2015, 2017, and February of 2018 each contributed to  $\sim$ 10% of the total loads for the entire study period (figure 6). The emergence of these hot moments for Al, Fe, NO $_2$ -N, and PO $_4$ -P were consistent with the episodic transport dynamics identified with the b and CVc/CVQ metrics (table 2).

Despite a significant agricultural presence in the watershed (55% of the land use), PO<sub>4</sub>-P does not exhibit chemostatic conditions, suggesting that the long-term agricultural activities in this watershed (since pre-1960s) have not overwhelmed the landscape's natural ability to serve as an effective biogeochemical filter for PO,-P. These results are in contrast to the chemostatic conditions that have been documented extensively for both N and P in intensively managed agricultural watersheds worldwide (Basu et al. 2010; Guan et al. 2011; Thompson et al. 2011; Haygarth et al. 2014; Gall et al. 2015; Dodd and Sharpley 2016; Miller et al. 2019). Significant efforts have been made by Northumberland County Conservation District and farmers within the watershed to adopt best management and conservation practices such as conservation tillage, contour cropping, fall manure application, and riparian buffers. Each of these best management practices has been adopted specifically to reduce sediment and P losses, lending evidence to suggest that best management practices aimed to reduce P accumulation in the watershed over the past 20+ years have been largely successful.

# **Summary and Conclusions**

More than 12,500 water samples analyzed for 12 constituents of natural and anthropogenic origin were collected with three sampling strategies and examined over a four-year period to characterize the transport dynamics of each constituent. Of the three sampling strategies—(1) three times weekly; (2) stormflow (triggered every half hour during storm events); and (3) time-paced (every four

Figure 4
Concentration-discharge (*C-Q*) relationships for each constituent of interest for each of the sampling strategies: (1) three times per week, (2) stormflow sampling, and (3) time-paced sampling (every four hours). Please note that the number of samples with concentrations above the method limit of quantification is less than 10% for nitrite-nitrogen (NO<sub>2</sub>-N) and less than 2% for orthophosphate (PO<sub>4</sub>-P), aluminum (Al), and iron (Fe) (see table 1).

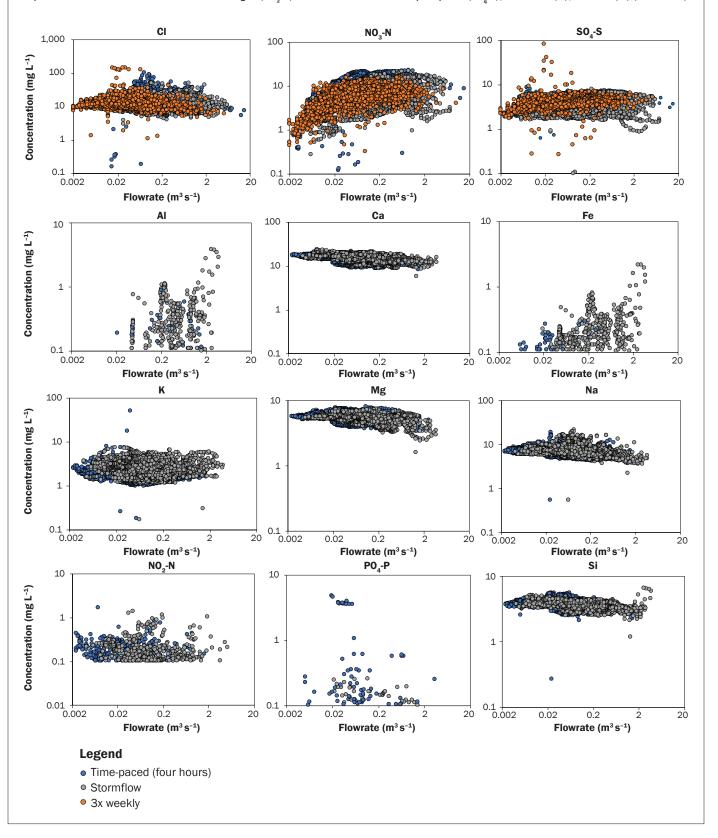


Table 2

Slopes of the concentration-discharge relationships (b), the ratio of the coefficient of variation of concentration (CVc) relative to the coefficient of variation of discharge (CVo), Gini coefficients for constituent loads (GL), and Gini coefficient ratios (GL/GO) for samples collected using the time-paced (four hours) sampling strategy. Only concentrations above the limit of quantification (LOQ) were used for these calculations.

Constituent	b	CVc/CVq	GL	GL/GQ
Chloride	0.06	0.15	0.65	1.00
Nitrate-N	0.26	0.16	0.69	1.07
Nitrite-N*	-0.05	0.32	0.97	1.51
Ortho-phosphate*	-0.24	0.58	1.00	1.55
Sulfate-S	0.13	0.11	0.66	1.03
Aluminum*	0.00	0.86	1.00	1.55
Calcium	-0.09	0.12	0.73	1.13
Iron*	0.10	0.35	1.00	1.55
Potassium	-0.07	0.34	0.75	1.17
Magnesium	-0.04	0.10	0.74	1.14
Sodium	-0.07	0.14	0.73	1.13
Silica	-0.06	0.10	0.73	1.13

Notes: N = nitrogen. S = sulfur.

hours)—the time-paced sampling resulted in ~7,000 water samples and captured the full range of flowrates observed during the study period. C-Q relationships, CVc/CVQ ratios, and GL/GQ ratios were employed to classify each constituent as exhibiting chemostatic, accretion, or dilution transport patterns.

Overall, the majority of the geogenic constituents of interest (Ca, Cl, Mg, Na, and Si) exhibited chemostatic transport behavior, with low variability in concentrations observed over the entire range of flows. Nitrate-N, SO<sub>4</sub>-S, and K also exhibited chemostatic dynamics with low concentration variability and high frequencies of detection, suggesting that anthropogenic inputs from fertilizer and manure (NO3-N and K) and from industry (SO<sub>4</sub>-S) have led to the emergence of legacy sources within the watershed that are enabling their transport to be chemostatic. Of particular interest is that PO<sub>4</sub>-P was present at detectable concentrations at the watershed outlet in a low percentage (<20%) of the samples collected and exhibited strong episodic transport dynamics as evidenced by high CVc/CVQ and GL/GQ ratios.

These results suggest that agricultural activities have led to the emergence of N but not P legacy sources within the watershed. Future adoption of best management practices is likely to continue to keep the watershed within its biogeochemical limits, with the watershed retaining its ability to act as an effective biogeochemical filter for P. However, a temporal lag in N load

reduction is likely to be observed given the extent of the N legacy sources already accumulated within the landscape. The results of the temporal inequality analysis suggest that NO<sub>2</sub>-N exhibits a moderate degree of temporal inequality, with a  $G_L$  value of  $\sim 0.7$ . To achieve load reduction goals consistent with those required by the Chesapeake Bay total maximum daily load (~25%), either low flow or high flow could be targeted with appropriate best management practices. To effectively treat low flow to achieve a 25% load reduction, the lowest 55% of the observed flowrates would need to be targeted. Conversely, a 25% load reduction goal could also be achieved by mitigating loads during the top 5% of the flowrates. Continued adoption of best management practices not specifically designed to treat either long durations of low flow or short periods of high flow events may not reduce NO<sub>3</sub>-N loads below current observed levels.

# **Acknowledgements**

The quantity and quality of these data are due to the unwavering technical support and troubleshooting of the Agricultural Research Service-Pasture Systems and Watershed Management Research Unit (ARS-PSWMRU) Klingerstown crew, who maintain the weirs and sampling equipment and collect the data, and to the chemists in the ARS-PSWMRU water analysis lab. Heather E. Preisendanz is supported, in part, by the Pennsylvania State University Institutes of Energy and the Environment and the USDA National Institute of Food and Agriculture Federal Appropriations under Project PEN04574 and Accession

number 1004448. This research and assessment was supported by the USDA Natural Resources Conservation Service Conservation Effects Assessment Project Watershed Assessment Studies and Agricultural Research Service National Program 211.

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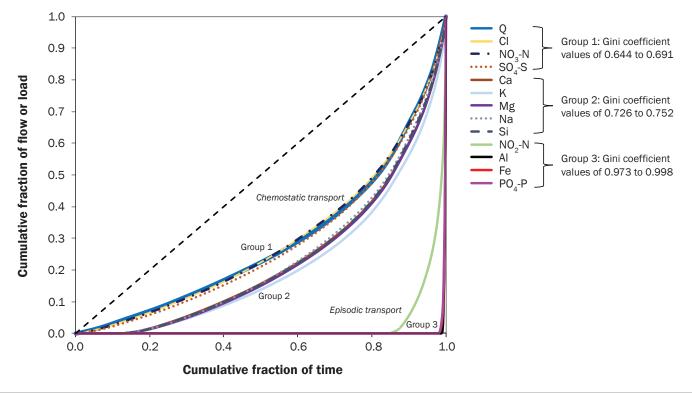
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<sup>\*</sup>For these constituents, <10% of the samples had concentrations above the LOQ (see table 1).

# Figure 5

Lorenz inequality curves for discharge and each of the constituents of interest for samples collected for the time-paced sampling strategy (every four hours). The further away from the 1:1 line that the Lorenz curves fall, the greater the temporal inequality of the constituent transport. When the constituent Lorenz curves overlap the discharge Lorenz curve, the transport dynamics are considered to be chemostatic (Group 1). Episodic transport occurs when high portions of load are exported during small portions of time (Group 3).



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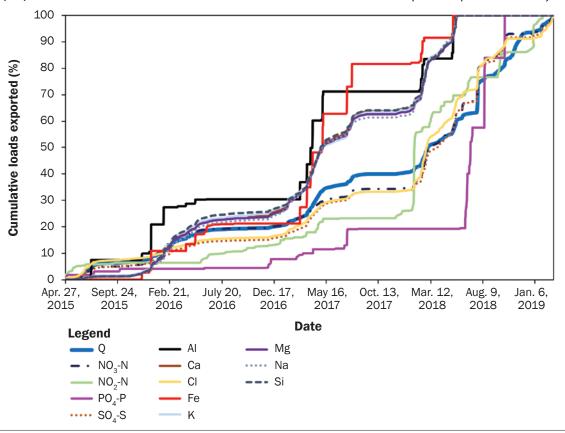
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Figure 6

Cumulative loads exported over the study period for each of the constituents of interest. Note that samples with concentrations below the limit of detection (LOD) were not considered to contribute to the loads. Loads were calculated from the time-paced samples collected every four hours.



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