Fox Illinois River Basin Watershed Calibration and Validation Datasets

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1. PROJECT BACKGROUND

The Department of Natural Resources (Department), together with many partners, is working to improve the surface water quality of tributaries, streams, rivers, and lakes within the Fox Illinois River Basin. To strengthen these ongoing efforts, the Department is developing a Total Maximum Daily Load (TMDL) for the river basin. The TMDL for this study area, referred to as the Fox Illinois River Basin (FOXIL) TMDL, will be a multi-year effort to address surface water quality impairments caused by phosphorus and total suspended solids. The TMDL study will provide a strategic framework and pollutant reduction goals for surface water quality improvement within the river basins.

The Fox Illinois River TMDL study area is located in southeastern Wisconsin. The study area includes the Fox River, the Des Plaines River, Nippersink Creek, North Mill Creek, and Channel Lake watersheds. The study area is primarily located in Racine, Kenosha, Walworth, and Waukesha counties. It is approximately bounded by Waukesha to the north, Lake Geneva to the southwest, and the western portions of Kenosha to the southeast. The TMDL study area covers approximately 1,060 square miles within Wisconsin, which is approximately 2 percent of the state. Within the study area, some lakes and streams are impaired (Wisconsin Department of Natural Resources, 2022), which means they are not meeting their water quality criteria. The extent of the TMDL and the waterbodies that are currently impaired are shown in Figure 1.1.

An important step in developing a TMDL is characterizing flows and loads at various locations in the watershed. The Department collected monitoring data in the project area from 2019 to 2022. These monitoring data were evaluated to develop estimates of continuous flow and estimates of daily load. The resulting datasets were used in the calibration and validation of a watershed model for the TMDL study area.

FIGURE 1.1 **Extent of Fox Illinois TMDL Study Area**

2. MONITORING SUMMARY

A monitoring plan for the Fox Illinois River Basin TMDL was implemented between December 2019 and May 2022. Monitoring was required to ensure adequate data were available for the calibration and validation of the watershed model. Water level, flow, and water chemistry data were collected at 13 sites during the monitoring period. A summary of the monitoring program is provided below, but a comprehensive report detailing the monitoring efforts are available in a separate report (Wisconsin Department of Natural Resources, 2023a).

2.1.1. Water Chemistry

Water chemistry data were collected at 13 locations in the study area. Water samples at five sites were collected by the Department, and samples at the remaining eight sites were collected by a private consultant, Cadmus. Water samples were evaluated for total phosphorus, orthophosphate, and total suspended solids. A list of the monitoring stations is provided in [Table 2.1,](#page-5-2) and the locations of the stations are displayed in [Figure 2.1.](#page-6-1) Results from the chemistry monitoring are available in a separate report (Wisconsin Department of Natural Resources, 2023a).

TABLE 2.1 **Fox Illinois River TMDL Chemistry Monitoring Sites**

Parameters: DOP = Dissolved orthophosphate, NH4 = Ammonium, NH3 = Nitrate, TN = Total Nitrogen, TP = Total Phosphorus, TSS = Total Suspended Solids

Project Boundary Washingto **TMDL Chemistry Monitoring** \bullet CADMUS **DNR** m. 683205 683096 10046937 10010534 6435 1001309 10032437 10040134 10053867 10029083 10012203 303066 20 Miles 10 WIDNR LISCS

FIGURE 2.1 **Fox Illinois River TMDL Chemistry Monitoring Locations**

2.1.2. Stage and Flow

Stage and flow monitoring data were also collected during the monitoring period. The Department collected periodic flow measurements and continuous stage data at five sites and periodic flow data at four sites. The sites with only flow measurements were located near gages maintained by the United States Geological Survey that had stage data available. A summary of the stage and flow monitoring sites is provided i[n Table 2.2,](#page-7-1) and the location of each site is provided in [Figure 2.2.](#page-7-0) Results from the stage and flow monitoring are available in a separate report (Wisconsin Department of Natural Resources, 2023a).

TABLE 2.2

FOX ILLINOIS RIVERS TMDL STAGE AND FLOW MONITORING SITES

FIGURE 2.2 **Fox Illinois River TMDL Stage and Flow Monitoring Locations**

3. CONTINUOUS FLOW ESTIMATION

Calibration of the watershed model developed for the FOXIL TMDL required continuous flow estimates. The stage and flow data summarized in the previous section were used to develop the continuous flow measurements. Rating curves—which relate flows to stage—were established at six sites in the study area. Sufficient stage data were not available at three sites, so an alternative approach was used that utilized linear relationships with flow measurements at nearby USGS gage stations (an enhancement to the area-weighted approach that can be used to estimate flows at different locations in a stream). The following sections describe the two methods for developing continuous flow estimates.

3.1. Stage-Discharge Relationships

One approach for producing estimates of continuous flow is the development of a stage-discharge relationships at each monitoring site. The stage-discharge relationship is expressed as an equation that estimates flow rate from stage. At six locations sufficient stage and discharge data were available to develop these relationships, and the methods are described in the following sections.

3.1.1. Development of Stage-Discharge Relationships

Continuous flow estimates were generated by developing stage-discharge relationships that relate periodic flow measurements to continuous stage measurements. Equation 3.1 provides the general form of the stage-discharge relationship (Hamilton, Watson, & Pike, 2019):

$$
Q = C_0 (H - e)^B
$$

Equation 3.1

where: Q: Discharge C0: Shape coefficient H: Stage e: Offset B: Shape exponent

The parameters in Equation 3.1 are linked to the physical features of a channel (Hamilton, Watson, & Pike, 2019). The coefficient (C_0) is the product of a coefficient describing the stage-area relationship and the flow resistance factor. The offset (e) is related to the stage at which flow is expected to be zero. The exponent (B) is the sum of a shape exponent in the stage-area relationship and a friction loss assumption. Since all of the parameters can be related to physical features, initial estimates of the coefficients can be derived based on the shape of the channel. [Figure 3.1](#page-9-0) is a table reproduced from Hamilton et al. (2019) that shows demonstrates the impact of channel shape on the stage-discharge equation.

Stage data were estimated from pressure transducers. The methods to convert pressure transducer data to stage data are described in the monitoring report for the FOXIL TMDL (Wisconsin Department of Natural Resources, 2023a). Sufficient continuous stage and discharge measurements were available at six monitoring locations. [Table 3.1](#page-9-1) lists the six stations and the source of the flow and stage measurements.

The channel shape at each of the locations with the flow data were estimated from data collected during the flow measurements. Flow measurements were collected with an Acoustic Doppler Current Profiler (ADCP), which provides data about depths along the width of the stream being measured. A channel shape at each of the sites were estimated from data collected on the day with the highest measured flow rate. The shapes of the cross-sections were used to ensure the exponent calculated for the stage-discharge relationship were physically reasonable. The cross-sections for each of the streams are provided in [Figure 3.2.](#page-9-2)

FIGURE 3.1

Link Between Channel Shape and Rating Curve Equations

Reproduced from Hamilton et al. (2019)

TABLE 3.1

Monitoring Locations Used for Stage-Discharge Relationships

FIGURE 3.2 **Cross-Sectional Areas at Measurement Locations**

Rating curves at each monitoring location were constructed in six steps:

- 1. *Identification of breakpoints:* A breakpoint is defined as a depth within the stream at which the control—in this case the channel cross section—encounters a notable change. For example, the shape of the cross-section at Honey Creek [\(Figure 3.2\)](#page-10-0) changed from a parabolic shape to a shape with near-vertical slopes at approximately 3 feet above the bottom of the channel. Since the rating curve equation was based on physical characteristics of the stream, the rating curve equation was likely different for depths below and above the breakpoint depth.
- 2. *Review of data for outliers:* Once breakpoints were established, datasets were reviewed for any potential outliers. Outliers were removed from the dataset for measurements where field staff indicated difficulties with the flow measurements collected.
- 3. *Transformation of data to log-log space:* Stage and discharge measurements for the remaining datapoints were converted to the natural log, which was based on recommendations from Department guidelines for developing rating curves (Wisconsin Department of Natural Resources, 2018).
- 4. *Calculation of linear regression statistics for log-transformed stage-discharge relationships:* The transformed dataset followed a linear trend, so statistics to determine linear fit were calculated. Linear regression statistics were calculated using the LINEST function in

Microsoft Excel (Microsoft Corporation, 2023) with the x values equal to the natural log of stage minus offset and the y values equal to the natural log of discharge.

- 5. *Adjustment of offset value to maximize coefficient of determination (R2):* The Solver tool in Excel was used to iterate on the value of the offset to maximize the coefficient of determination. Coefficient of determination indicates the strength of a linear relationship, and a value closer to one indicates a strong linear relationship.
- 6. *Review of stage-discharge parameters:* The values calculated from the Solver equation were reviewed to ensure they were reasonable. First, the offset was examined to ensure the estimated value represented a stage relatively close to the expected depth of zero-flow. Second, the exponent was compared to the stream cross-sections and the cross-sections in [Figure 3.1](#page-9-0) to ensure the values were reasonable.

The results of the rating curve estimates are summarized in [Table 3.2.](#page-11-1) The rating curves developed for the six sites and the observations are provided in [Appendix A.](#page-21-0) Overall, the relationship of the rating curves was strong for Honey Creek, Sugar Creek, the White River, and the higher flows of Muskego Lake Canal. The discharges predicted by the stage-discharge relationship for the Wind Lake outlet are close to the observed discharges; however, the rating curve overpredicts flow for the highest flow measured. Wind Lake Canal may be influenced by a backwater effect from the Fox River near Rochester, which may explain why the measured flow is lower than the predicted flow. Without additional data, however, it is not possible to verify the rating curve for higher flows. Thus, continuous flow estimates from at the Wind Lake outlet must be used with caution.

TABLE 3.2 **Parameters for Rating Curve Equations**

3.1.2. Adjustments for Muskego Lake and Wind Lake

During the monitoring period, periods of zero flow were observed in the Muskego Lake channel and at the Wind Lake outlet. Outflow from Wind Lake and Muskego Lake was actively managed during the monitoring period to ensure adequate levels in the lakes were maintained. During dry periods the crests of the respective dams were above the water level in the lakes and no flow was discharged downstream.

Since only periodic flow measurements were taken during the course of the monitoring period, periods of zero flow had to be estimated when developing the continuous flow estimates. The USGS gages at both Muskego Lake and Wind Lake contain stage data within the lake and in the downstream channel. The stage data in the channel downstream of the lake was used to develop the rating curves, but the stage data from the lake itself was useful for predicting periods of zero flow. To determine zero-flow periods, measured flows were plotted against the stage in the lake. From these plots, the approximate lake level maintained during different seasons was estimated. When measured lake level was less than the estimated maintained lake level, flow downstream was assumed to be zero. When measured lake level was greater than the estimated maintained lake level, the rating curve and the downstream stage data were used to estimate flows. A summary of the estimated maintained lake levels is provided in [Table 3.3.](#page-12-3)

TABLE 3.3 **Estimated Lake Level Maintained for Muskego and Wind Lakes**

When lake level measured at USGS gage maintained lake level, flow downstream is zero

3.1.3. Impacts of Ice on Flow Estimates

During the monitoring period, the streams being monitored were occasionally covered with ice. The presence of ice prevented flow data being collected. Ice also had an impact on stage measurements. Flows during periods when ice was present ('ice-on') could not be reliably predicted from the stagedischarge relationships, so no flow estimates are provided for these dates. These 'ice-on' periods were identified by reviewing the field data provided by the team conducting monitoring and by reviewing the stage data from the continuous level logger equipment.

3.1.4. Development of Continuous Flow Dataset

The rating curves and the information about lake level management were combined with measured continuous stage data to create continuous flow datasets. The estimated continuous flows compared to the measured flows are provided in [Appendix A.](#page-21-0) Time periods with no flow estimates are also shown in the figures. Periods without flow estimates generally align with dates when ice was present. Dates without estimates outside of expected ice-on periods correspond to times where stage estimates from the level loggers were unavailable or unreliable. Details about the operation of the monitoring equipment is provided in the report detailing the monitoring activities for the TMDL (Wisconsin Department of Natural Resources, 2023a).

3.2. FLOW ESTIMATION USING LINEAR REGRESSION WITH USGS GAGES

Continuous flow estimates at three sites were developed by comparing periodically measured flows with continuous flows at nearby USGS gages. A linear regression relationship was fit to these data to develop continuous flow estimates at the stations with only periodic flow measurements. The monitoring locations utilizing linear regression with USGS flow data are summarized in [Table 3.4](#page-12-4)

TABLE 3.4 **Monitoring Locations Utilizing Linear Regression with USGS Stations**

The linear regression method was used at these three locations because measured stage data prevented the development of reliable and robust stage-discharge relationships. At the Fox River at County Highway I, issues with the deployment of the level logger limited the length of the continuous stage records. Details about the challenges are provided in the monitoring report for this project (Wisconsin Department of Natural Resources, 2023a). Additionally, the Fox River at County Highway I is the location of a long-term trends water quality monitoring site operated by the Department. The long-term trends site contains a long-term dataset for water quality parameters. To best utilize these data, a long-term, continuous flow dataset was required to develop long-term load estimates. The load estimation procedure is described in detail in Section 4.

Stage data for the locations of the Fox River at Waterford and Rochester were available from the USGS; however, the stage gages are located upstream of the dams at Waterford and Rochester. The operation of the dams varies throughout the year, so stage alone is not a reliable predictor of flow. Due to these limitations, an accurate stage-discharge relationship could not be developed.

To estimate continuous flows at the three locations described above, periodic flow measurements were plotted against the flow measurements collected at the USGS gages on the same date. A linear regression relationship was fit to the plotted measurements to characterize the relationship between flows at the two locations. The equation from the linear regression is summarized in Equation 3.2. A summary of the linear regression relationships is provided i[n Table 3.5.](#page-13-1) A plot of the relationships are provided in [Appendix B.](#page-28-0)

$$
Q_{sta} = slp \times Q_{ISGS} + intercept
$$
 Equation 3.2

where:

O_{sta}: Discharge at the monitoring location slp: Slope of regression relationship QUSGS: Discharge at the USGS gage intercept: Intercept of the regression relationship

TABLE 3.5

Results of Linear Regression Equations for Estimating Flows

The linear regression method is an enhancement to the area-weighted approach that can be used to estimate flows at different locations in a stream. In the area-weighted approach, a ratio of the

upstream areas from a location with flow measurements and a location with no flow measurements is calculated. This ratio is multiplied by flows at the location with measured flows to estimate flows at the location without measured flows. The linear regression approach was used instead of the areaweighted method because using the available measured flow data provided a more accurate representation of estimated flows.

The results of the linear regression analysis defined by Equation 3.2 and summarized in [Table 3.5](#page-13-1) were applied to the continuous flow records at the respective USGS stations to develop a continuous flow estimate for each of the monitoring locations. The resulting continuous flow estimates and the observed flows are provided in [Appendix B.](#page-28-0)

4. DAILY FLUX (LOAD) ESTIMATION

Continuous flow measurements and periodic water quality monitoring data were used to establish daily flux estimates for fourteen locations in the study area. Daily flux was estimated using a modified version of the Fluxmaster and LOADEST methods developed by USGS (Schwarz, Hoos, Alexander, & Smith, 2006). The technique used to estimate daily flux was also used for previous TMDLs developed by the Department. This section of the report describes the methods used for the load estimation. The following text is reproduced from Appendix J of the Department's Northeast Lakeshore TMDL report (Wisconsin Department of Natural Resources, 2023b).

4.1. Site Specific Flux Models

Continuous daily fluxes were estimated for both TP and TSS at each site in the monitoring network. Flux computation was performed with a modified version of the methods that are associated with U.S. Geological Survey Fluxmaster and LOADEST software programs (Schwarz, Hoos, Alexander, & Smith, 2006). The purpose of these methods is to estimate constituent concentrations at a given site when water quality sampling frequency is insufficient for estimating continuous long-term flux. The methods are most effective for constituents that have a strong relationship with discharge and exhibit cyclic variation with season (e.g., sediment concentration is often greatest with snowmelt events in late Spring). Additionally, a time variable allows concentrations to vary, linearly or quadratically, over the sampling period.

4.2. Modifications to LOADEST Model

The first purpose of modifying the Fluxmaster/LOADEST method was to rectify issues with marginal sample sizes for most sites in the monitoring network. Since the development of these tools, new statistical methods have become available that allow model coefficients to vary by a grouping factor (e.g., a monitoring site). Using each monitoring site as a grouping factor, a single model per constituent can be fitted without the loss of degrees of freedom that would result from multiple independent models for each site using the regression methods implemented in Fluxmaster/LOADEST. The modeling framework chosen for model fitting was an implementation of linear mixed effects models in the R programming language for statistical computing (R Core Team, 2020) the library used in R software was the lme4 package (Bates, Maechler, Bolker, & Walker, 2015). Most of the same methods were used by adapting computer code from the rloadest R package (Runkel & De Cicco, 2017),except when model fitting was performed using lme4.

The second purpose of modifying the Fluxmaster/LOADEST method was based on initial findings that quickflow (the combination of surface runoff and shallow aquifer water yield) was a better predictor of TP and TSS than total discharge. To calculate quickflow for each site, a baseflow separation routine was applied based on wavelet transform (Nathan & McMahon, 1990) that is available in an R package called EcoHydRology (Fuka, Walter, Archibald, Steenhuis, & Easton, 2015).The recommended parameters (filter parameter = 0.925, passes = 3) were used. In model fitting for both TSS and TP, quickflow was always used instead of total discharge. All quickflow values were translate by a value of positive 0.01 to prevent the log transformation of zero values during dry periods when baseflow accounted for all flow. Constituent concentration models were fitted using quickflow, however flux estimates were calculated by multiplying concentration predictions by total discharge.

4.3. Model Selection

The rloadest R package provides a convenience function that fits 9 different models that are different permutations of discharge, season, and time as fixed-effect co-variates, then selects the best performing model as the one with the lowest AIC. Structuring these models as mixed-effect models using the lmer package in R allows more permutations of model coefficients. A mixed-effect model allows coefficients to vary by grouping factor, usually referred to as a "random effect". In a mixed-effects model, coefficients that do not vary by grouping factor are referred to as "fixed effects". Therefore, the 9 different models used by the rloadest package can be expanded to 30 with different permutations of discharge (i.e., quickflow), season, and time as both fixed effects and random effects. In cases where random effects were fitted, the coefficients were allowed to vary for each monitoring site. The intercept of all model permutations was allowed to vary by monitoring site.

Two models (one for TSS, and one for TP) were selected that predicted constituent concentrations across all monitoring sites. These models were selected by permuting through all combinations of fixed and random effects for quickflow, season, and time, then selecting those with the minimum AIC. The models selected for TSS and TP (model 6c) can be described by the following equation:

$$
\ln(\rho_m) = \beta_0 + [\beta_f] * \begin{bmatrix} \ln(Q_m) \\ \ln(Q_m^2) \\ \sin(2\pi T_m) \\ \cos(2\pi T_m) \end{bmatrix} + \gamma_0 + [\gamma_{f,m}] * \begin{bmatrix} \ln(Q_m) \\ \ln(Q_m^2) \\ \sin(2\pi T_m) \\ \cos(2\pi T_m) \end{bmatrix} + e_m \text{ Equation 4.1}
$$

In the above equation, m is a monitoring site, ρ is a matrix of TSS or TP concentrations, Q is a matrix of quickflow paired with ρ , T is a matrix of decimal time numbers paired with ρ , β is a fixed-effect coefficient, γ is a random-effect coefficient, and e is residual error.

Daily flux estimates were only generated for TSS at ten of the fourteen sites. Four of the sites with TSS data were located downstream of dams that impact the release of TSS. The models described above were tested using all fourteen sites, but an accurate model was not able to be established for the four sites downstream of dams. Despite the limitations with TSS data, accurate daily flux estimtes were able to be established for TP at all fourteen sites. The coefficients for both fixed effects are listed in [Table 4.1](#page-16-0) for TSS and [Table 4.2](#page-17-1) for TP.

TABLE 4.1

Coefficients for TSS Linear Mixed Effects Model

TABLE 4.2

Coefficients for TP Linear Mixed Effects Model

4.4. Flux Estimation

To estimate flux for a given day, the estimated constituent concentrations from Equation 4.1 must first be re-transformed from natural log to real space then multiplied by the average daily discharge for each day. In the process of transforming back to real space, systemic biases in the predictions can occur due to heteroscedasticity in the linear model. A common approach to reduce bias in estimates is to multiply each concentration prediction by a bias-correction factor (BCF). The BCF that is used in the LOADEST model is a function of the residual standard error (SE) of the mode (Runkel, Crawford, & Cohn, 2004)

$$
BCF = \exp\left(\frac{SE^2}{2}\right)
$$
 Equation 4.2

Residual error varies for each monitoring site, and therefore site-specific biases can be resolved by calculation a BCF for each monitoring site, m .

$$
BCF_m = \exp\left(\frac{SE_m^2}{2}\right)
$$
 Equation 4.3

The benefit of using a mixed-effect model is that coefficients are allowed to vary by grouping factors without the same reduction in degrees of freedom that would result in independent regression models fit for each group. However, there is some disagreement among statisticians about the interpretation of degrees of freedom associated with each group fit in a mixed-effects model. Therefore, the denominator in the equation used to calculate the SE of the residuals for each monitoring site, m , is subject to interpretation:

$$
SE = \sqrt{\frac{\Sigma_i e_{i,m}^2}{d.f.}}
$$
 Equation 4.4

Multiple calculations for degrees of freedom for each monitoring site were tested, including sample size, sample size minus the number of fixed effects, and sample size minus the number of the total of fixed and random effects. Simply using the sample size as the degrees of freedom for each monitoring site provided an appropriate balance of bias correction across sites. For those monitoring sites where the number of samples are limited, the flux predictions should be used with caution, paying closest attention to the overall bias of predictions.

4.5. Assessment of Fit

For each constituent—TSS and TP—a single mixed-effect model was fitted that included samples across all monitoring sites. However, the flux estimates and associated characterization of model fit were used site-by-site. Therefore, all observed (sample concentration multiplied by mean daily flow on the date the sample was taken) and simulated fluxes were first separated out by site before calculating performance statistics (i.e., error is characterized as e_m from Equation 4.1). Performance statistics for TSS and TP for each site are shown in [Table 4.3](#page-18-1) and [Table 4.4.](#page-18-3)

TABLE 4.3

Fit Statistics for TSS Mixed-Effects Model

TABLE 4.4

Fit Statistics for TP Mixed-Effects Model

In addition to quantitative performance statistics, visualizations can also aid in diagnosing problems in flux models (Hirsch, 2014). Systemic biases are apparent when model residuals are plotted against estimations and each of the covariates, discharge, time, and season. Biases can also be diagnosed when samples occur in a frequency that does not align with natural variation—boxplots of the variation between sample concentrations and estimates are useful for testing differences in these distributions. Similarly, these biases are apparent if boxplots are created for values of discharge on sampled days versus all daily discharge values. Simple scatterplots showing observed versus simulated for both concentration and flux, can also be useful. Reviewing these plots in a standardized format (Hirsch, 2014) is a quick way to assess an individual site-specific flux model. Diagnostic plots are provided in [Appendix C](#page-32-0) for TSS an[d Appendix D](#page-43-0) for TP.

In addition to diagnostic plots, time-series plots showing continuous daily flux estimates along with sampled flux (sample concentration multiplied by daily mean discharge), can reveal specific times when large errors occurred. The time-series plots for all TSS sites are provided in [Appendix E,](#page-57-0) and time-series plots for all TP sites are provided in [Appendix F.](#page-70-0)

5. REFERENCES

- Bates, D., Maechler, M., Bolker, B., & Walker, S. (2015). Fittling linear mixed-effects models using lme4. *Journal of Statistical Software, 67*(1), 1-48.
- Fuka, D., Walter, M., Archibald, J., Steenhuis, T., & Easton, Z. (2015). *EcoHydRology: A community modeling foundation for Eco-Hydrology.*
- Hamilton, S., Watson, M., & Pike, R. (2019). The role of the hydrographer in rating curve development. *Confluence: Journal of Watershed Science and Management, 3*(1), 1-15.
- Hirsch, R. (2014). Large biases in regression-based constituent flux estimates: causes and diagnostic tools. *Journal of the American Water Resources Association, 50*(6), 1401-1424.

Microsoft Corporation. (2023). Microsoft Excel.

- Nathan, R., & McMahon, T. (1990). Evaluation of automated techniques for base flow and recession analysis. *Water Resources Research, 26*(7), 1465-1473.
- R Core Team. (2020). R: A language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing. Retrieved from https://www.R-project.org/
- Runkel, R., & De Cicco, L. (2017). rloadest: river load estimation.
- Runkel, R., Crawford, C., & Cohn, T. (2004). *Load Estimator (LOADEST): A FORTRAN program for estimating constituent loads in streams and rivers.* Reston, VA: U.S. Geological Survey.
- Schwarz, G., Hoos, A., Alexander, R., & Smith, R. (2006). *The SPARROW surface water quality model: theory, application, and user documentation.* Reston, VA: United States Geological Society.
- Wisconsin Department of Natural Resources. (2018). *Guidelines for Developing Flow Rating Curves.* Madison, WI.
- Wisconsin Department of Natural Resources. (2022). Water Condition Lists. Madison, WI. Retrieved from https://dnr.wisconsin.gov/topic/SurfaceWater/ConditionsList.html
- Wisconsin Department of Natural Resources. (2023a). *Monitoring Results for the Fox Illinois River Basin TMDL.* Madison, WI. Retrieved from https://dnr.wisconsin.gov/sites/default/files/topic/TMDLs/FOXILTMDL-MonitoringResults.pdf
- Wisconsin Department of Natural Resources. (2023b). *The Northeast Lakeshore TMDL: Total Maximum Daily Load for Total Phosphorus and Total Suspended Solidsd.* Madison, WI. Retrieved from https://dnr.wisconsin.gov/sites/default/files/topic/TMDLs/NEL_TMDL_report_approved.pdf

APPENDIX A

RATING CURVES AND CONTINUOUS FLOW **ESTIMTES**

FIGURE A.1 **Rating Curve for Fox River at ES**

FIGURE A.2 **Rating Curve for Honey Creek at Academy Road**

FIGURE A.3 **Rating Curve for Sugar Creek at Potter Road**

FIGURE A.4 **Rating Curve for White River at Hwy 36**

FIGURE A.5 **Rating Curve for Muskego Canal at Muskego Dam Road**

FIGURE A.6 **Rating Curve for Wind Lake Outlet at South Wind Lake Road**

FIGURE A.7 **Flow Estimates for Fox River at ES**

FIGURE A.8 **Flow Estimates for Honey Creek at Academy Road**

FIGURE A.9 **Flow Estimates for Sugar Creek at Potter Road**

FIGURE A.10 **Flow Estimates for White River at Hwy 36**

FIGURE A.11 **Flow Estimates for Muskego Canal at Muskego Dam Road**

FIGURE A.12 **Flow Estimates for Wind Lake Outlet at South Wind Lake Road**

A.6

APPENDIX B

LINEAR REGRESSION RELATOINSHIPS AND CONTINUOUS FLOW ESTIMATES

FIGURE B.1 **Relationship between Fox River at Cth I and USGS 05543830**

FIGURE B.2 **Relationship between Fox River at Waterford and USGS 05545750**

B.1

FIGURE B.3 **Relationship between Fox River at Rochester and USGS 05545750**

FIGURE B.4 **Continuous Flow Estimates for Fox River at Cth I**

FIGURE B.5 **Continuous Flow Estimates for Fox River at Rochester**

FIGURE B.6 **Continuous Flow Estimates for Fox River at Rochester**

APPENDIX C

DIAGNOSTIC PLOTS FOR TSS LOAD ESTIMATION

C.8

APPENDIX D

DIAGNOSTIC PLOTS FOR TP LOAD ESTIMATION

D.8

APPENDIX E

TIME SERIES PLOTS FOR TSS LOAD ESTIMATION

DOY

DOY

DOY

 $E.7$

DOY

DOY

DOY

APPENDIX F

TIME SERIES PLOTS FOR TP LOAD ESTIMATION

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DOY

F.4

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