

# Appendix 16. Model Archival Summary for Nitrate plus Nitrite Concentration at U.S. Geological Survey Site 06892350, Kansas River at De Soto, Kansas, during June 2013 through September 2019

This model archival summary summarizes the nitrate plus nitrite ( $\text{NO}_x$ ; U.S. Geological Survey [USGS] parameter code 00631) concentration model developed to compute 15-minute  $\text{NO}_x$  concentrations from June 2013 onward. This model supersedes all previous models.

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## Site and Model Information

Site number: 06892350

Site name: Kansas River at De Soto, Kansas

Location: Lat 38°59'00", long 94°57'52" referenced to North American Datum of 1927, in NE 1/4 SE 1/4 SE 1/4 sec.28, T.12 S., R.22 E., Leavenworth County, Kans., hydrologic unit 10270104.

Equipment: A YSI 6600 water-quality monitor equipped with sensors for water temperature, specific conductance, dissolved oxygen, pH, and turbidity was installed from August 2012 through June 2014. A Xylem YSI EXO2 water-quality monitor equipped with sensors for water temperature, specific conductance, dissolved oxygen, pH, turbidity, and chlorophyll and phycocyanin fluorescence was installed during June 2014 through September 2019. A Hach Nitratax plus sc sensor ( $\text{sNO}_x$ ; 5-millimeter path length) that monitors ultraviolet (UV) nitrate concentrations was installed from June 2013 through September 2019. The monitors were housed in side-by-side 4-inch-diameter galvanized steel pipes. Readings from the water-quality and nitrate plus nitrite monitors were recorded every 15 minutes and transmitted by way of satellite, hourly.

Date model was created: March 31, 2020

Model calibration data period: June 17, 2013, through September 24, 2019

Model application date: June 17, 2013, onward

## Model-Calibration Dataset

All data were collected using USGS protocols (Wagner and others, 2006; U.S. Geological Survey, variously dated) and are stored in the National Water Information System (U.S. Geological Survey, 2020) database and available to the public. Ordinary least squares analysis was used to develop regression models using R programming language (R Core Team, 2020). Potential explanatory variables that were evaluated individually and in combination included streamflow, water temperature, specific conductance, dissolved oxygen, pH, turbidity, chlorophyll and phycocyanin fluorescence, and  $\text{sNO}_x$ . The maximum time span between two continuous data points used for interpolation was 2 hours (in order to preserve the sample dataset, field monitor averages obtained during sample collection were used for model development data if no continuous data were available or if gaps larger than 1 hour in the continuous data record resulted in missing interpolated data). Seasonal components (sine and cosine variables) were also evaluated as potential explanatory variables.

The final selected regression model was based on 84 concurrent measurements of  $\text{NO}_x$  concentration and sensor-measured  $\text{sNO}_x$  during June 17, 2013, through September 24, 2019. Samples were collected throughout the range of continuously observed hydrologic conditions. Twenty samples had concentrations below laboratory detection limits; therefore, a Tobit regression model was developed to compute estimates of linear regression model parameters using the absolute maximum likelihood estimation approach (Cohen, 1950; Hald, 1949; Helsel and others, 2020; Tobin, 1958). Summary statistics and the complete model-calibration dataset are provided below. Potential outliers were identified using the methods described in Rasmussen and others (2009). Additionally, outlier test criteria, including leverage and Cook's distance (Cook's D; Cook, 1977), were used to estimate potential outlier influence on the final Tobit regression model. One of the  $\text{NO}_x$  results, from May 9, 2019, was deemed an outlier and removed from the model calibration dataset. This sample was collected during 3 hours of rapidly changing conditions due to a runoff event and may not have been representative of the associated sample time. All other potential outliers were not found to have errors associated with collection, processing, or analysis and were therefore considered valid.

This model is specific to the Kansas River at De Soto, Kans., during this study period and cannot be applied to data collected from other sites on the Kansas River or data collected from other waterbodies.

## Nitrate plus Nitrite Sampling Details

Cross-section samples typically were collected either from the downstream side of the bridge or instream within 100 feet of the

bridge. The equal-width-increment collection method was used (although multiple vertical, single vertical, and grab samples were occasionally collected), and samples typically were composited for analysis (U.S. Geological Survey, variously dated). During July 2012 through June 2017, cross-section samples were collected every 2 weeks during March through October, once a month during November through February, and during selected reservoir release and runoff events. During July 2017 through September 2019, cross-section samples were collected on a monthly to bimonthly basis, depending on flow conditions. A FISP US DH–81, DH–95, D–95, D–96a, or D–96 depth integrating sampler was used. Additional detail on sample collection is available in Foster and Graham (2016) and Graham and others (2018). Samples were analyzed for NO<sub>x</sub> concentration at the USGS National Water Quality Laboratory in Lakewood, Colorado.

## Model Development

Discretely collected NO<sub>x</sub> was related to sensor-measured sNO<sub>x</sub> and other continuous sensor-measured data using stepwise regression analysis in R programming language (R Core Team, 2020). The distribution of residuals was examined for normality, and the plots of residuals (the difference between the measured and computed values) were examined for homoscedasticity (departures from zero did not change substantially over the range of computed values). Previously published explanatory variables were also strongly considered for continuity.

23.8 percent of the model-calibration dataset were censored results (less than the minimum reporting level). Tobit regression models were developed using absolute maximum likelihood estimation methods to relate discretely collected NO<sub>x</sub> concentration to sensor-measured sNO<sub>x</sub>. Tobit model parameter estimates were calculated using the *smwrQW* (v0.7.9) package in R programming language (R Core Team, 2020).

sNO<sub>x</sub> was selected as a good surrogate for NO<sub>x</sub> based on residual plots, pseudocoeficient of determination (pseudo-*R*<sup>2</sup>), and estimated residual standard error. Values for all the aforementioned statistics were computed and are included below along with all relevant sample data and additional statistical information.

## Model Summary

The following is a summary of final Tobit regression analysis for NO<sub>x</sub> concentration at USGS site 06892350:

NO<sub>x</sub> concentration-based model:

$$NO_x = 1.02 \times sNO_x - 0.325$$

where

NO<sub>x</sub> = nitrate plus nitrite concentration, in milligrams per liter; and

sNO<sub>x</sub> = sensor-measured nitrate plus nitrite, in milligrams per liter.

sNO<sub>x</sub> makes physical and statistical sense as an explanatory variable for NO<sub>x</sub>.

## Previous Models

Start Year	End Year	Model Equation	Reference
1999	2003	NO <sub>x</sub> = 0.914logQ - 0.541logWT - 2.04  where Q = streamflow, in cubic feet per second; and, WT = water temperature, in degrees Celsius	Rasmussen and others (2005)

## Model Statistics, Data, and Plots

### Model

$$\text{NO}_x = + 1.02 * \text{sNO}_x - 0.325$$

Computation method: Absolute Maximum Likelihood Estimation (AMLE)

### Variable Summary Statistics

	NO <sub>x</sub>	sNO <sub>x</sub>
Minimum	<0.01	0.068
1st Quartile	0.034	0.373
Median	0.776	1.130
Mean	0.667	1.024
3rd Quartile	1.27	1.519
Maximum	2.20	2.259

### Basic Model Statistics

Estimated residual standard error (unbiased)	0.2209
Number of observations	84
Number censored	20 (23.8 percent)
Log-likelihood (model)	-3.677
Log-likelihood (intercept only)	-146
Chi-square	284.7
Degrees of freedom	1
p-value	<0.0001
Pseudo-R-squared	0.8975
Akaike Information Criterion	13.35
Bayesian Information Criterion	20.65

### Explanatory Variables

Coefficients:

	Estimate	Std. Error	z-score	p-value
(Intercept)	-0.3246	0.05456	-5.95	0
sNO <sub>x</sub>	1.0160	0.04269	23.80	0

### Outlier Test Criteria

Leverage	Cook's D
0.03571	0.69897

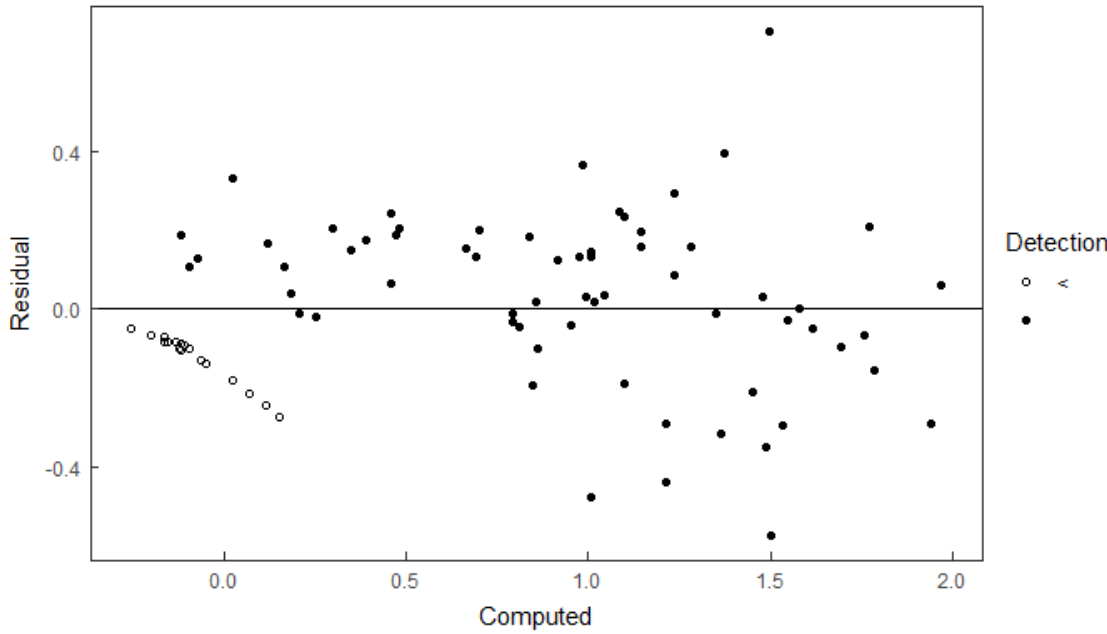
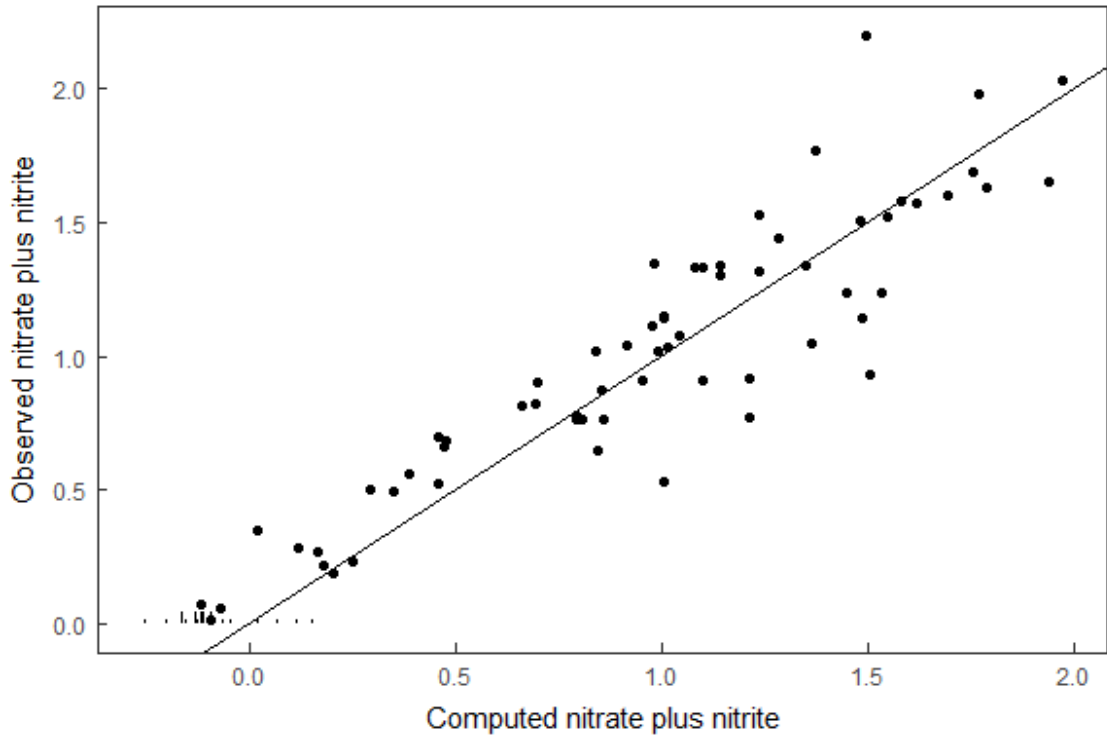
### Flagged Observations

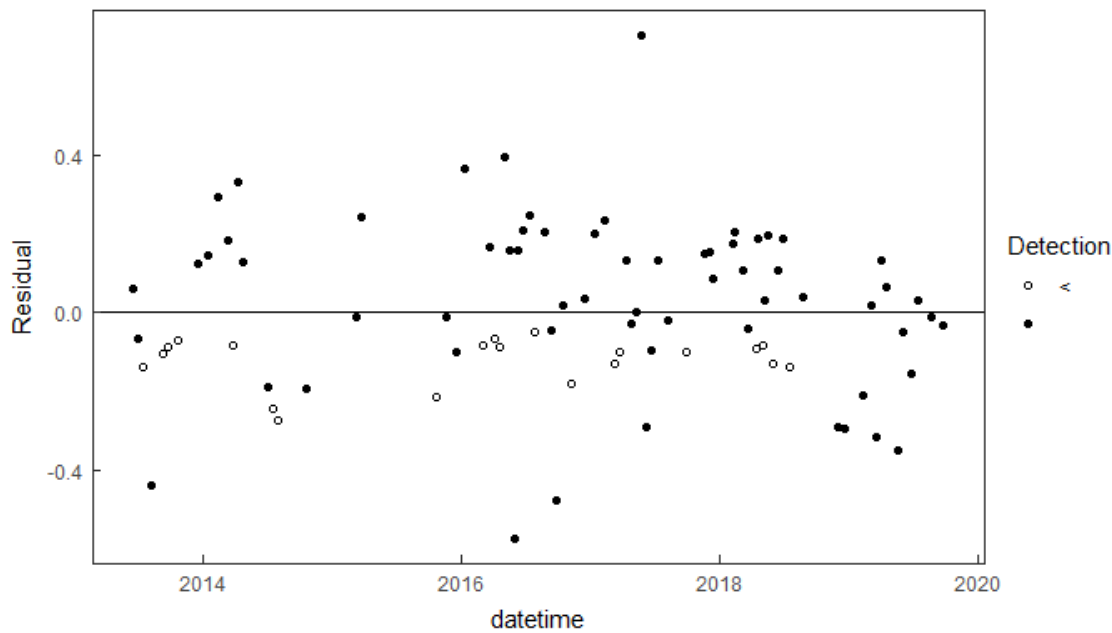
Observations exceeding at least one test criterion						
	NO <sub>x</sub>	ycen	yhat	resids	leverage	cooksD
1	2.03	FALSE	1.9705	0.05953	0.05688	0.002323
2	1.69	FALSE	1.7563	-0.06626	0.04284	0.002104
27	0.01	TRUE	-0.2022	-0.06676	0.03596	0.001767
33	1.98	FALSE	1.7716	0.20844	0.04375	0.021311
35	0.01	TRUE	-0.2558	-0.04822	0.03885	0.001002
50	1.65	FALSE	1.9399	-0.28987	0.05472	0.052742
51	1.60	FALSE	1.6951	-0.09506	0.03931	0.003945
81	1.63	FALSE	1.7869	-0.15686	0.04468	0.012350

95 Percent Confidence Interval

	2.5 %	97.5 %
(Intercept)	-0.4315937	-0.2177029
sNO <sub>x</sub>	0.9323271	1.0996520

Plots





## Model-Calibration Dataset

	datetime	NO <sub>x</sub>	sNO <sub>x</sub>	Computed_NOx
1	2013-06-17 12:30:00	2.03	2.2590	1.9705
2	2013-07-01 13:20:00	1.69	2.0482	1.7563
3	2013-07-15 10:40:00	<0.01	0.2711	-0.0492
4	2013-08-05 11:30:00	0.772	1.5135	1.2131
5	2013-09-09 10:30:00	<0.01	0.2033	-0.1180
6	2013-09-23 09:40:00	<0.04	0.2033	-0.1180
7	2013-10-21 10:30:00	<0.04	0.1581	-0.1639
8	2013-12-16 08:00:00	1.04	1.2199	0.9148
9	2014-01-13 08:00:00	1.15	1.3102	1.0066
10	2014-02-10 07:40:00	1.53	1.5361	1.2361
11	2014-03-10 13:40:00	1.02	1.1446	0.8383
12	2014-03-24 12:50:00	<0.04	0.1883	-0.1333
13	2014-04-07 11:40:00	0.35	0.3388	0.0197
14	2014-04-21 12:00:00	0.057	0.2485	-0.0721
15	2014-06-30 12:20:00	0.91	1.4006	1.0984
16	2014-07-14 12:50:00	<0.01	0.4292	0.1115
17	2014-07-28 13:50:00	<0.01	0.4669	0.1497
18	2014-10-20 14:00:00	0.65	1.1521	0.8459
19	2015-03-09 12:20:00	0.78	1.0994	0.7924
20	2015-03-23 10:50:00	0.696	0.7681	0.4557
21	2015-10-19 09:30:00	<0.01	0.3840	0.0656
22	2015-11-16 09:10:00	0.19	0.5196	0.2033
23	2015-12-14 15:10:00	0.76	1.1671	0.8612
24	2016-01-11 10:10:00	1.35	1.2876	0.9836
25	2016-03-03 09:00:00	<0.01	0.1581	-0.1639
26	2016-03-21 10:50:00	0.285	0.4367	0.1191
27	2016-04-04 09:20:00	<0.01	0.1205	-0.2022
28	2016-04-18 09:30:00	<0.04	0.2033	-0.1180
29	2016-05-02 09:30:00	1.77	1.6717	1.3738
30	2016-05-16 09:10:00	1.44	1.5813	1.2820
31	2016-05-28 13:20:00	0.93	1.7997	1.5039
32	2016-06-06 11:30:00	1.3	1.4458	1.1443
33	2016-06-20 08:20:00	1.98	2.0632	1.7716
34	2016-07-11 08:40:00	1.33	1.3855	1.0831

35	2016-07-25	08:30:00	<0.01	0.0678	-0.2557
36	2016-08-22	09:00:00	0.5	0.6099	0.2951
37	2016-09-12	08:50:00	0.76	1.1144	0.8077
38	2016-09-26	10:30:00	0.53	1.3102	1.0066
39	2016-10-11	09:20:00	0.87	1.1596	0.8536
40	2016-11-07	08:40:00	<0.01	0.3388	0.0197
41	2016-12-12	09:50:00	1.08	1.3479	1.0448
42	2017-01-09	09:50:00	0.9	1.0090	0.7006
43	2017-02-06	10:00:00	1.33	1.4006	1.0984
44	2017-03-06	10:40:00	<0.01	0.2560	-0.0645
45	2017-03-20	10:30:00	<0.04	0.2259	-0.0951
46	2017-04-10	09:20:00	1.11	1.2801	0.9760
47	2017-04-25	10:10:00	1.52	1.8449	1.5498
48	2017-05-08	09:20:00	1.58	1.8750	1.5804
49	2017-05-22	09:20:00	2.2	1.7921	1.4962
50	2017-06-05	08:50:00	1.65	2.2289	1.9399
51	2017-06-19	09:10:00	1.6	1.9879	1.6951
52	2017-07-10	09:10:00	0.825	1.0015	0.6929
53	2017-08-07	09:30:00	0.23	0.5647	0.2492
54	2017-09-26	10:20:00	<0.01	0.1958	-0.1257
55	2017-11-17	14:40:00	0.496	0.6626	0.3486
56	2017-12-01	09:00:00	0.814	0.9714	0.6623
57	2017-12-11	10:30:00	1.32	1.5361	1.2361
58	2018-02-05	10:00:00	0.56	0.7003	0.3869
59	2018-02-09	09:30:00	0.684	0.7906	0.4787
60	2018-03-05	10:10:00	0.27	0.4819	0.1650
61	2018-03-20	08:20:00	0.91	1.2575	0.9530
62	2018-04-11	13:00:00	<0.04	0.2090	-0.1123
63	2018-04-16	09:30:00	0.07	0.2033	-0.1180
64	2018-04-30	09:20:00	<0.01	0.1657	-0.1563
65	2018-05-07	10:40:00	1.02	1.2952	0.9913
66	2018-05-15	09:30:00	1.34	1.4458	1.1443
67	2018-05-29	08:40:00	<0.01	0.2560	-0.0645
68	2018-06-11	10:00:00	0.01	0.2259	-0.0951
69	2018-06-25	07:20:00	0.66	0.7831	0.4710
70	2018-07-16	11:00:00	<0.01	0.2711	-0.0492
71	2018-08-21	10:40:00	0.22	0.4970	0.1803
72	2018-11-29	12:20:00	0.92	1.5135	1.2131
73	2018-12-18	10:40:00	1.24	1.8298	1.5345
74	2019-02-06	10:20:00	1.24	1.7470	1.4503
75	2019-03-06	07:50:00	1.03	1.3177	1.0142
76	2019-03-19	09:20:00	1.05	1.6641	1.3662
77	2019-04-01	12:00:00	1.14	1.3102	1.0066
78	2019-04-16	09:30:00	0.52	0.7681	0.4557
79	2019-05-20	12:30:00	1.14	1.7846	1.4886
80	2019-06-03	10:20:00	1.57	1.9126	1.6186
81	2019-06-26	12:30:00	1.63	2.0783	1.7869
82	2019-07-16	10:10:00	1.51	1.7771	1.4809
83	2019-08-20	09:20:00	1.34	1.6491	1.3509
84	2019-09-24	09:50:00	0.76	1.0994	0.7924

## Definitions

**Cook's D:** Cook's distance (Helsel and others, 2020).

**Leverage:** An outlier's measure in the x direction (Helsel and others, 2020).

**NO<sub>x</sub>:** Inorganic nitrogen (nitrate and nitrite), in milligrams per liter as nitrogen (00631).

**p-value:** The probability that the independent variable has no effect on the dependent variable (Helsel and others, 2020).

**Pseudo-R-squared:** Pseudocoeficient of determination. An estimation of the proportion of variance in the response variable explained by the model (McKelvey and Zavoina, 1975).

**sNO<sub>x</sub>:** Sensor measured inorganic nitrogen (nitrate and nitrite), in milligrams per liter as nitrogen (99133).

**z-score:** The estimated coefficient divided by its associated standard error (Helsel and others, 2020).

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