



Supplementary Materials for

Legacy nitrogen may prevent achievement of water quality goals in the Gulf of Mexico

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Materials and Methods

S1.0 Relationship Between NO₃-N and Total N Loading

Based on a multi-model analysis of the relationship between N loading from the Mississippi River Basin and the areal extent of the hypoxic zone, it has been suggested that an approximately 60% decrease in total N loading may be necessary to meet policy goals (6). In our simulations, however, we focus only on prediction of changing NO₃-N loads based on the assumption that total N trajectories will mirror temporal changes in NO₃-N loads. We believe this assumption to be justifiable based on our analysis of 1980-2016 Mississippi N data, which shows that NO₃-N consistently accounts for 65% of total N ($\pm 6\%$) and thus demonstrates a strong linear relationship ($R^2=0.83$) with total N (Fig. S2).

S2.0 The ELEMeNT Model

The ELEMeNT modeling framework couples source-zone N dynamics (the root zone) with a travel time-based approach that accounts for both transport and transformation of N along groundwater pathways, as has been described in detail in Van Meter et al (4). The model was created to explicitly account for changing stocks of N within the soil profile and in groundwater pools. Model development was based on the principle that watershed nutrient dynamics at any given time are a function of both current and historical land use and management. ELEMeNT is the only watershed model that can directly simulate the impacts of watershed-scale nutrient legacies.

S2.1 *Source Zone Dynamics*

To capture long-term patterns of land-use (LU) and management, the model segments the watershed into s distinct units corresponding to distinct land-use trajectories, with each being stored within a 2D land-use array, $LU(s,t)$ representing distributions of land use over time, t .

$$LU(s,t) = \begin{cases} 2 & s \leq A_{crop}(t) & \text{cropland} \\ 1 & A_{crop}(t) < s \leq [A_{crop}(t) + A_{past}(t)] & \text{pastureland} \\ 0 & s > [A_{crop}(t) + A_{past}(t)] & \text{other} \end{cases}$$

(1)

where A_{crop} and A_{past} correspond to watershed-scale percent cropland and percent pastureland, respectively (4).

S1.2.2 *Source Function*

40 Each distinct LU trajectory has a corresponding $J_s(s,t)$, which describes the N mass
 41 leaching from the source zone (vadose zone) at any time t . Accordingly, the source
 42 function, which is the mass of N leaching from the source zone to groundwater, is
 43 quantified as the sum of source-zone values across the distribution of all LU trajectories:
 44

$$J_{s_{vzhd}}(t) = \sum_{s=1}^{1000} J_s(s,t) \quad (2)$$

45
 46
 47 Within the soil profile, we consider that all of the annual N surplus cycles through
 48 either active or protected soil organic N (SON) pools, with inorganic N then being
 49 produced as a function of SON mineralization. Additional details regarding source-zone
 50 dynamics can be found in Van Meter et al (4).
 51

52 *S2.3 Catchment N loading*

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 54 ELEMENNT couples a travel time-based approach with source-zone dynamics to
 55 predict N loading at the catchment outlet. This approach entails conceptualization of all
 56 points on the landscape as corresponding to individual stream tubes, each with its unique
 57 travel time to the stream network (4, 32). The resulting distribution of travel times then
 58 controls NO_3 -N mass fluxes at the outlet, $M_{out}(t)$, as described in the following equation:
 59

$$M_{out}(t) = \int_0^{\infty} J_s(t-\tau) f(\tau) e^{-\gamma\tau} d\tau + (1-k_h)W(t) \quad (3)$$

60
 61 where $J_s(t-\tau)$ is the contaminant input function describing the mass flux of N from the
 62 unsaturated zone to groundwater, as discussed in S2.2, γ is the denitrification rate
 63 constant, and $W(t)$ is human N consumption and k_h the denitrification rate constant for
 64 human waste. For additional details, see Van Meter et al (4).
 65
 66

67 S3.0 N Surplus Calculations and Data Sources

68
 69 A surface N balance approach was used to calculate annual N surplus values, which
 70 are the primary input term to the model's source zone function (4, 33). Inputs were
 71 considered to include N fertilizer application, biological N fixation, livestock manure N,
 72 and atmospheric deposition, with crop N uptake as the major output. For further details
 73 regarding mass balance calculations and data sources, see Van Meter et al (4).
 74
 75

76 S4.0 Uncertainty Analysis, Sensitivity Analysis, and Model Calibration/Validation

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 78 The ELEMENNT model developed for the Mississippi River Basin was subjected to
 79 rigorous evaluation to ensure that model predictions were robust. The evaluation process
 80 included uncertainty analysis for calculated N surplus values, sensitivity analysis for

81 model parameters, and a calibration/validation approach which included use of mean
82 absolute error values for both measured soil organic N (SON) data and N loading data at
83 the catchment outlet as objective functions. A summary of our approach is provided
84 below, and additional details can be found in Van Meter et al (4).

86 *S4.1 Uncertainty Analysis for N Surplus Trajectories*

88 Monte Carlo simulations were used to characterize uncertainty related to the calculated N
89 surplus trajectories for the watershed. For each parameter used in the mass balance
90 calculations, we assumed a normal distribution of values with a coefficient of variation
91 (CV) value of 0.3. We then carried out 1000 simulations to obtain time-varying median
92 and interquartile range values for the N surplus trajectories for the period 1800-2014.

94 *S4.2 Sensitivity Analysis*

96 For the sensitivity analysis, we utilized the Latin hypercube sampling technique, a form
97 of stratified Monte Carlo sampling (34), to generate 1000 parameter sets from designated
98 ranges of values based on literature review and data specific to the MRB. Model
99 simulations were run with these parameter sets, and the evaluated output variables
100 included not only annual N loading at the catchment outlet, but median soil organic N
101 values for soil across the MRB (1950-2015) and soil organic N accumulation (1980-
102 2010). The results of the sensitivity analysis showed that the primary parameters
103 impacting 1950-2010 median SON levels were the mineralization rate constants and
104 humification coefficients for non-cultivated and cultivated land. N loading at the
105 catchment was found to be particularly sensitive to denitrification rate constants in soil
106 and groundwater and the mean travel time through hydrologic pathways, including the
107 shallow subsurface, deeper groundwater, and surface water.

109 *S4.3 Model Calibration*

111 For the model calibration, we selected parameters for optimization based on the results of
112 the sensitivity analysis, as described in *S4.2*. The model was then calibrated to optimize
113 simulation of (1) current levels of SON, based on USDA gridded soil survey data, and (2)
114 N loading at the catchment outlet, as quantified from USGS water quality data
115 (Mississippi River near St. Francisville, Louisiana, and Atchafalaya River at Melville,
116 Louisiana) and discharge data (Mississippi River at Tarbert Landing, Old River Outflow
117 Channel near Knox Landing, Atchafalaya River at Simmesport, Ohio River at
118 Metropolis, and Mississippi River at Thebes). To assess the goodness of fit of the
119 observed data with predicted value, we used as our objective functions mean absolute
120 error values for both stream NO₃-N loading and current soil SON data. Optimization was
121 carried out iteratively, leading to a final selection of the top-performing 10% of parameter
122 sets. Performance was based on goodness of fit to the specified objective function.
123 Median parameter values were extracted for all relevant parameters. For watershed NO₃-
124 N loads, model calibration was carried out for the years 1979-2013, and the period 1955-
125 1970 was used for validation.

127 Measured data for the period 1979-2013 indicate median $\text{NO}_3\text{-N}$ loading of 3.3 kg
128 $\text{ha}^{-1} \text{ yr}^{-1}$ (980 kt yr^{-1}) compared with modeled values of 3.2 kg $\text{ha}^{-1} \text{ yr}^{-1}$, a difference of
129 only 1.3%. For the 1955-1970 validation period, median measured loading values for the
130 watershed were 1.2 kg $\text{ha}^{-1} \text{ yr}^{-1}$, while modeled median values were 1.1 kg $\text{ha}^{-1} \text{ yr}^{-1}$, a
131 statistically insignificant difference.

132

133 S5.0 Future Simulations

134

135 In the main text, we report the results of our core simulations, which assume
136 constant inputs of N from atmospheric deposition and wastewater treatment plants across
137 the study period. We have also performed a set of secondary simulations which assume
138 predicted changes in these values. Details regarding the two different simulation sets are
139 described below.

140

141 *S5.1 Core Simulations*

142

143 In the main text, we present the results of our core simulation scenarios for the
144 MRB. In the business-as-usual (BAU) scenario, we assume that agricultural management
145 practices remain constant; as a result, watershed N surplus values remain at 2016 levels
146 throughout the simulation period. For the other scenarios, we assume simple percent
147 decreases in the agricultural N surplus (ANS) in relation to mean 2010-2015 surplus
148 values. As an example, we have determined that from 2010-2015, the mean N surplus for
149 cropland throughout the MRB was approximately 48.9 kg/ha (2). Under the 25%
150 reduction, the ANS value is reduced to 36.7 kg/ha, and under the 100% reduction
151 scenario it becomes zero. We consider the ANS to include inputs and outputs specific to
152 crop and livestock production. Accordingly, it does not include atmospheric N
153 deposition or wastewater N. Under these scenarios, the changes in watershed
154 management are considered to be immediate, with a start year of 2017.

155

156 In all of the core simulations we assume that atmospheric N deposition and
157 wastewater N inputs remain constant over the simulation period. This simplified
158 approach, though not our best prediction of future dynamics, allows us to more clearly
159 evaluate the role of legacy N in driving changes in N loading without the addition of
160 other complicating factors. We have, however, also run a set of simulations that do
161 include such changes, as described in S5.2.

162

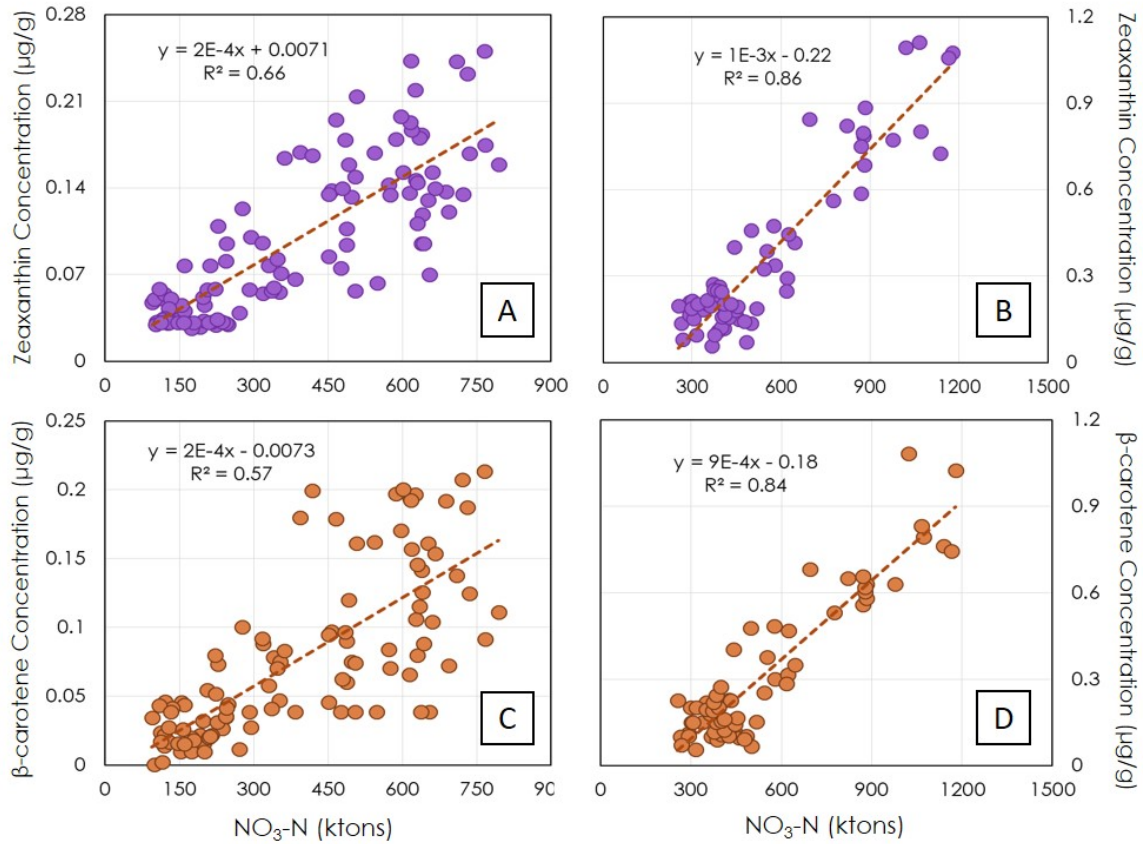
163 *S5.2 Secondary Simulations*

164

165 In our secondary simulations, we run the BAU and N surplus reduction scenarios, as
166 described in S3.1, but also consider both projected increases in human population and
167 projected decreases in atmospheric N deposition. Population trajectories were obtained
168 from the U.S. Census Bureau, which predicts a 22% increase in population by the year
169 2060. Trends in atmospheric deposition assume a 22% decrease in N deposition by the
170 year 2022 in response to implementation of tighter air quality standards (12).

171

172 As seen in Fig. S3, the trends in N loading are quite similar to those under the core
173 simulation scenarios, as shown in Fig. 2 of the main text, but the actual reductions in
174 loading that can be achieved are somewhat greater. Under the secondary BAU scenario,
175 a 16% reduction in N loading is achieved by 2050, as compared with the 11% achieved
176 under the core BAU scenario. Similarly, a 100% reduction in the agricultural N surplus
177 under the secondary scenario provides a 57% reduction in loading by 2050, as compared
178 with a 55% reduction under the core scenario. Under both core and secondary scenarios,
179 however, the lag time to achieving new steady-state loading levels is still approximately
180 30 years, and it is clear that legacy N still has a strong influence on N loading.
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182



184

185 **Fig. S1.**

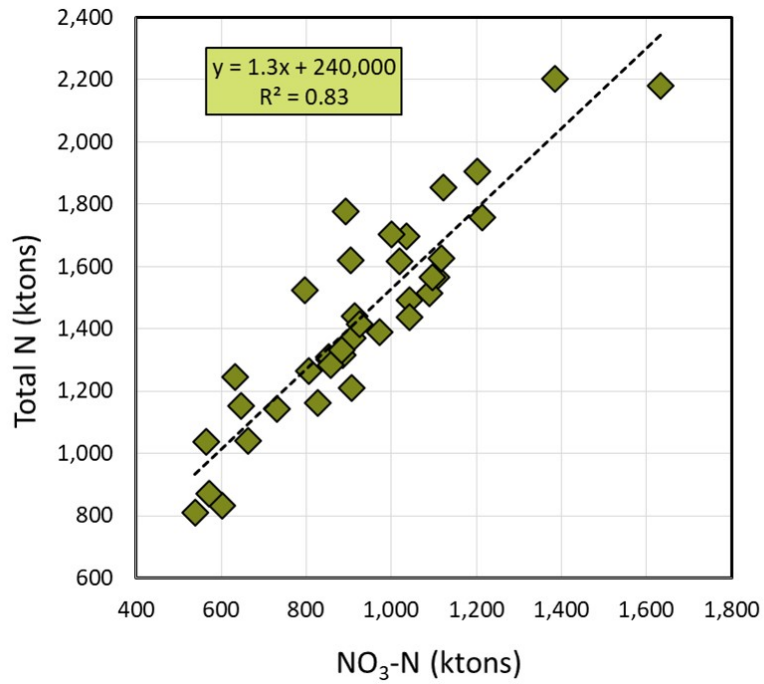
186 Linear correlations between predicted nitrate loads and chloropigment (zeaxanthin and β-

187 carotene) concentrations in sediment cores obtained from the northern Gulf of Mexico

188 (1). Panels (A) and (C) represent the period from 1820-1930, and panels (B) and (D) that

189 from 1930-1998. All relationships are significant ($p < 0.001$)

190

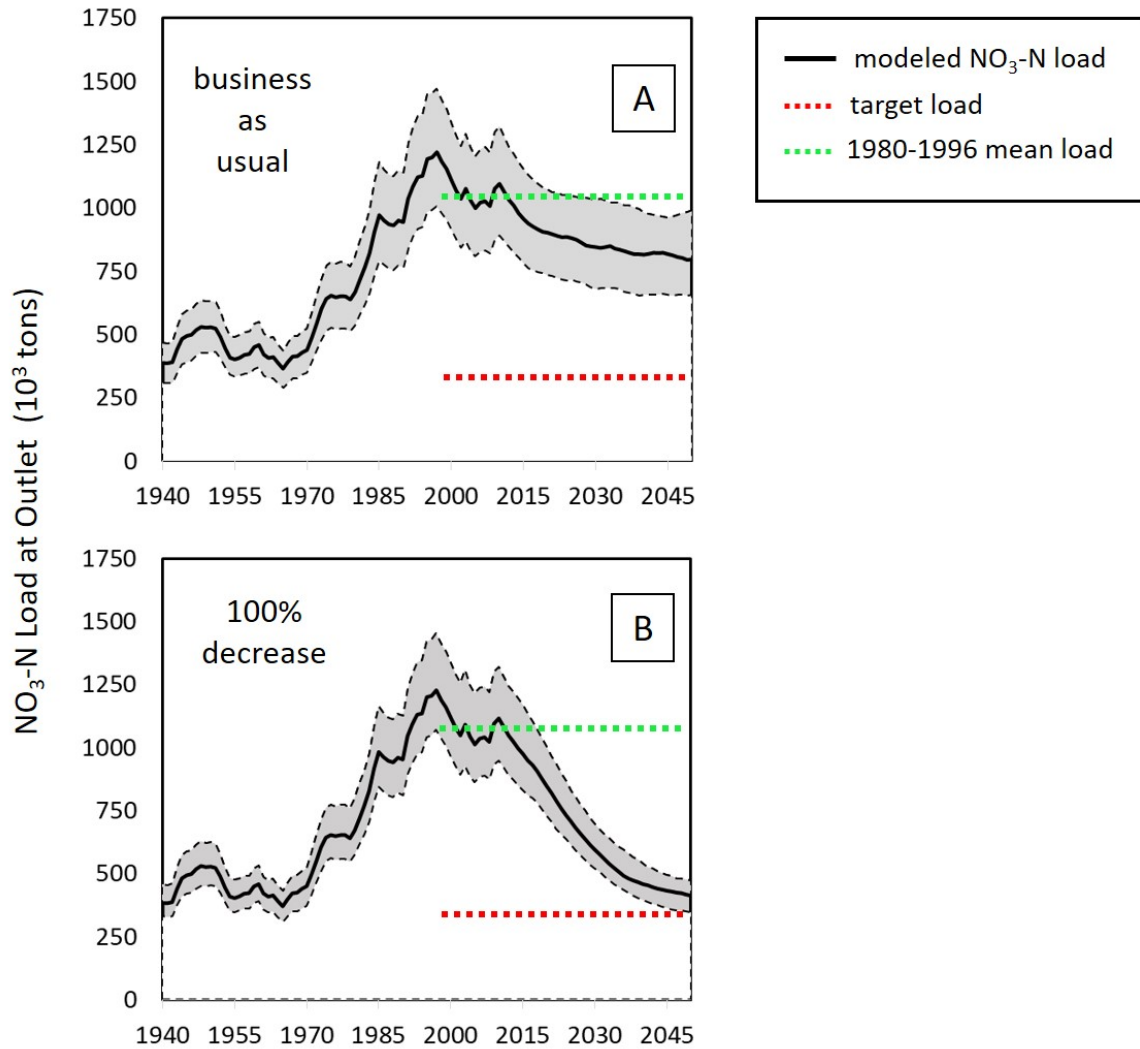


192

193 **Fig. S2**

194 Strongly linear relationship between NO₃-N and total N loads in the Mississippi River
195 Basin, 1980-2016. Loads represent the sum of the mainstem Mississippi River and the
196 Atchafalaya River

197



198

199 **Fig. S3**

200 Catchment-scale NO₃-N loading for the Mississippi River Basin, 1940-2050, under two
 201 future scenarios: (A) business as usual and (B) a 100% decrease in the agricultural N
 202 surplus. The green dashed lines represent mean N loading for the period 1980-1996, and
 203 the red dashed lines represent target N loading to achieve an areal reduction in the
 204 summer Gulf of Mexico hypoxic zone to 5,000 km². In these scenarios, we take into
 205 account predicted increases in population as well as decreases in atmospheric N
 206 deposition. Under these scenarios, our results suggest that reductions in N loading from
 207 16-57% can be achieved.

208

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