- **Capturing Interactions between Nitrogen and Hydrological** 1 **Cycles under Historical Climate and Land Use:** 2 Susquehanna Watershed Analysis with the GFDL Land 3 **Model LM3-TAN** 4 5 M. Lee<sup>1</sup>, S. Malyshev<sup>2</sup>, E. Shevliakova<sup>2</sup>, P. C. D. Milly<sup>3</sup>, and P. R. Jaffé<sup>1</sup> 6 7 8 [1] {Department of Civil and Environmental Engineering, Princeton University, Princeton, NJ, 9 USA} [2]{Princeton Environmental Institute, Princeton University, NOAA/Geophysical Fluid 10 11 Dynamics Laboratory affiliate, Princeton, NJ, USA} [3]{U. S. Geological Survey and NOAA/Geophysical Fluid Dynamics Laboratory, Princeton, 12 NJ, USA} 13 14 Correspondence to: M. Lee (minjinl@princeton.edu)
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# 16 Abstract

17 We developed a process model LM3-TAN to assess the combined effects of direct human influences and climate change on Terrestrial and Aquatic Nitrogen (TAN) cycling. The model 18 19 was developed by expanding NOAA's Geophysical Fluid Dynamics Laboratory land model 20 LM3V-N of coupled terrestrial carbon and nitrogen (C-N) cycling and including new N cycling processes and inputs such as a soil denitrification, point N sources to streams (i.e. 21 22 sewage), and stream transport and microbial processes. Because the model integrates 23 ecological, hydrological, and biogeochemical processes, it captures key controls of transport 24 and fate of N in the vegetation-soil-river system in a comprehensive and consistent 25 framework which is responsive to climatic variations and land use changes. We applied the model at 1/8 degree resolution for a study of the Susquehanna River basin. We simulated with 26 27 LM3-TAN stream dissolved organic-N, ammonium-N, and nitrate-N loads throughout the 28 river network, and we evaluated the modeled loads for 1986-2005 using data from 16

1 monitoring stations as well as a reported budget for the entire basin. By accounting for inter-2 annual hydrologic variability, the model was able to capture inter-annual variations of stream N loadings. While the model was calibrated with the stream N loads only at the last 3 downstream SRBC station Marietta (40.02' N, 76.32' W), it captured the N loads well at 4 5 multiple locations within the basin with different climate regimes, land use types, and associated N sources and transformations in the sub-basins. Furthermore, the calculated and 6 7 previously reported N budgets agreed well at the level of the whole Susquehanna watershed. 8 Here we illustrate how point and non-point N sources contributed to the various ecosystems are stored, lost, and exported via the river. Local analysis for 6 sub-basins showed combined 9 10 effects of land use and climate on soil denitrification rates, with the highest rates in the Lower 11 Susquehanna Sub-basin (extensive agriculture; Atlantic coastal climate) and the lowest rates 12 in the West Branch Susquehanna Sub-basin (mostly forest; Great Lakes and Midwest 13 climate). In the re-growing secondary forests, most of the N from non-point sources was 14 stored in the vegetation and soil, but in the agricultural lands most N inputs were removed by 15 soil denitrification indicating that anthropogenic N applications could drive substantial 16 increase of  $N_2O$  emission, an intermediate of the denitrification process.

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#### 18 **1** Introduction

19 Biologically available nitrogen (N) in terrestrial ecosystems has significantly increased via 20 anthropogenic nutrient inputs: artificial fertilizer, cultivation of N fixing crops, and fossil fuel 21 consumption (Galloway et al., 2004; 2008). This increase has caused acidification and N 22 saturation in some terrestrial and aquatic ecosystems (Henriksen and Brakke, 1988; Kelly et al., 1990; Murdoch and Stoddard, 1992; Howarth, 2002). N-saturated soils and streams are 23 24 also major sources of nitrous oxide (N2O) emissions, which is a potent greenhouse gas 25 (Albritton et al., 1994). Other concerns include severe water-quality problems associated with 26 cultural eutrophication, which results in harmful algal blooms and hypoxia in rivers, lakes, 27 estuaries, and coastal zone ecosystems (Smith, 2003; Smith et al., 2006). Climate change and 28 variability also affect water quality through the distribution of high and low flow extremes (Scavia et al., 2002; Howarth et al., 2006). It is generally accepted that microbial processes 29 related to the N cycle are strongly influenced by abiotic factors, and warm or wet climate 30 31 provides favorable environments for certain groups of bacterial activities. Quantification and 1 management of the diverse and coupled effects of human activity and climate change on N 2 cycling requires a comprehensive model of the relevant coupled processes that can support 3 the design of optimal nutrient loading controls to maintain desirable water quality and 4 terrestrial ecosystem integrity.

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To characterize implications of human and climate driven perturbation in the earth N cycling 6 7 and its implication for water and air quality, the next-generation of N cycling models need to 8 (1) account for regional and local changes in terrestrial and aquatic ecosystem structure and 9 functioning, (2) represent in a consistent manner emissions and transformation of N to air, 10 rivers and coasts, and (3) be global in extent and integrated with climate and earth system models. Previously, none of the existing models addressed the above 3 challenges. Here we 11 12 present a novel modeling framework capable of addressing these challenges and, prior to its global application, we evaluate this modeling framework in the Susquehanna River Basin 13 14 whose sub-basins vary in climate, land use, and associated N sources and transformations, 15 with the detailed dataset of observations.

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17 There has been keen interest and progress in modeling the N cycle in terrestrial ecosystems. 18 However, in most models vegetation and land use type distribution are prescribed and do not 19 change in time. Modeling studies with EPIC, ANIMO, and CENTURY/DAYCENT typically 20 prescribe crop distribution and simulate crop production and related nutrient and carbon (C) 21 cycling (Sharpley and Williams, 1990; Parton et al., 1993; Williams et al., 1995; Kroes and 22 Roelsma, 1998; Del Grosso et al., 2009). Because these models do not simulate decadal-tocentury changes in vegetation structure (e.g. forest regrowth after harvesting) they are likely 23 24 to overlook changes in the storage of N in vegetation. Furthermore, during wood harvesting 25 and forest clearing for agriculture, biomass residue is an important additional input to the soil organic C-N pools. Such additional N inputs lead to additional N inorganic loads. In addition, 26 27 many regional models (e.g. EPIC, ANIMO), which have been applied to far smaller basins 28 compared to the Susquehanna watershed, often use basin-specific parameters for mineralization, nitrification, and denitrification, which complicates their application on a 29 global scale for decadal-to-century scale studies. LM3-TAN is capable of describing N 30 31 dynamics with a universal parameter set – the same parameters for all of the sub-basins within an area of 71,220 km<sup>2</sup> and time periods for this simulation. LM3-TAN is among very few modeling frameworks (e.g. CLM-CN: Thornton et al 2007; CLM4MOD: Thomas et al., 2013) that can be used as a component of an ESM – that is it is capable to represent sub-diurnal exchanges of moisture, energy, C and N species among land-atmosphere. Unlike CLM4MOD, LM3-TAN simulates water quality in the rivers and nutrient loadings to the coastal environment.

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8 Contrary to the simulations of land models limited to the terrestrial component, most 9 watershed models do estimate stream N concentrations and loads, but they simplify or neglect many key mechanisms describing terrestrial N dynamics (e.g., vegetation and land-use 10 dynamics, interactive C-N feedbacks on vegetation and soil microbial processes; phenological 11 12 leaf drop and its contribution to soil organic matter pools). INCA-N and SWAT are widely 13 used Geographic Information System-based watershed models (Wade et al., 2002; Schilling 14 and Wolter, 2009). However, when it comes to large scale applications, because these models 15 are semi-distributed, they are less capable of representing spatial variability, requiring users to 16 define the number and sizes of sub-basins, in which land use and all of the processes for each 17 land use are assumed to be homogeneous and needed to be defined individually. This limits their ability to analyze complex land-use management scenarios. In this class of models, 18 19 RHESSys is one of a few models with an ecology component that can be used to investigate 20 interactions between ecosystems and hydrological processes according to climate variability 21 (Tague and Band, 2004; Beckers et al., 2009). However, like the most models, because these 22 models do not simulate vegetation and land use type distribution, a specific parameter set that 23 describes typical soil, vegetation, and land use characteristics has to be developed using its 24 special module when a study site requires different vegetation or soil types from its default 25 application. This explains why RHESSys has only been applied to very small or sub sections 26 of catchments (Band et al., 2001; Tague and Band, 2004).

27

Given the current lack of models that link terrestrial C-N cycling, long-term vegetation and land-use dynamics to N loads and concentrations in streams, accounting for different N species, the goal of this research was to build a model to simulate stream N loads that is based on a global-scale terrestrial and N-enabled land model, followed by its testing on a large and 1 complex watershed, for which many years of stream discharge and stream N data are 2 available. For this purpose and to assess the combined effects of direct human influences and 3 climate change on Terrestrial and Aquatic Nitrogen (TAN) cycling, we developed a process 4 model LM3-TAN. The new features include integrated effects of point and non-point sources 5 on river N loads, a soil denitrification module, and stream microbial processes.

6

7 We applied LM3-TAN to the Susquehanna River basin, the largest of the watersheds in the 8 northeastern U.S., draining an area of 71,220 square kilometers, at the resolution of 1/8 9 degree. The model was evaluated using 20 year (1986-2005) data of stream ammonium 10  $(NH_4^+)$  and dissolved organic N loads as well as stream nitrate  $(NO_3^-)$  N loads from 16 11 monitoring stations. For each of 6 sub-basins, we conducted local analysis to assess combined 12 effects of land use and climate on the soil denitrification. We then built up a N budget and 13 compared it with the corresponding reported budget to better understand how point and non-14 point N sources contributed to the various ecosystems are stored, lost, and exported via the river at the level of the whole Susquehanna watershed. Although there are several parameters 15 16 that required calibration by fitting simulated to reported stream N loads, these parameters are 17 used universally for the entire basin where climate, soil, vegetation, and land use 18 characteristics vary. Efforts have been made in the development of this model to limit the 19 number of calibrated parameters.

20

# 21 2 Model Description

## 22 2.1 Overview

LM3V-TAN is an expansion of earlier GFDL land models, beginning with LM3V of 23 24 Shevliakova et al (2009), which describes vegetation and C dynamics. LM3-TAN was expanded to include vegetation- and soil-N dynamics from LM3V-N (Gerber et al., 2010), 25 26 new soil physics and hydrology from LM3 (Milly et al, 2014), and N cycling processes 27 described here. LM3 was used as a component of the GFDL Earth System Models (Dunne et al, 2012) and included several enhancements, such as vertically resolved soil physics and 28 29 hydrology and explicit river dynamics and physics. LM3-TAN includes soil denitrification and transport and chemistry of N cycle in rivers. This version of the model allows more 30

1 complete tracking of N through the soil-river continuum. In this section, we first summarize

key features of the model, and then we describe the newest N cycling features.

- 2
- 3

LM3V simulates distribution of five vegetation functional types (C3 and C4 grasses, and 4 5 temperate-deciduous, tropical, and cold-evergreen trees) on the basis of total biomass and prevailing climate conditions. The model tracks hundreds of years of land use change using 6 7 global land use transition scenarios that were historically reconstructed by combining 8 satellite-based contemporary patterns of agriculture with historical data on agriculture and 9 population (Hurtt et al., 2006). The four land use types are natural vegetation (land 10 undisturbed by human activities), secondary vegetation (land formerly disturbed by human activities), cropland, and pasture. The model is spatially distributed, and each grid cell 11 12 consists of up to 15 tiles: 1 natural vegetation, 1 cropland, 1 pasture, and 1 to 12 secondary vegetation tiles representing unique disturbance histories (i.e. de/reforestation, agricultural 13 14 practice change). Exchanges of water, energy, and between land and atmosphere are computed with a time step of 30 minutes. Atmospheric and terrestrial reservoirs include C 15 16 pools in vegetation (leaves, fine roots, sapwood, heartwood, and labile C storage), soil (fast and slow), and anthropogenic storage. The C pools in the vegetation are updated on a daily 17 18 time step to account for vegetation growth and allocation, leaf drop and display, and natural 19 mortality and fire. The soil C, which is supplied by the vegetation both naturally and during 20 land-use conversion, is stored in two pools with different turnover times.

21

# 22 2.2 Coupled C-N dynamics in vegetation and soil

The previous two soil C pools in LM3V were divided into four pools (fast and slow litter, and 23 24 slow and passive soil organic matter) in LM3V-N. Each C pool in the vegetation and soil was paired with a respective N compartment using pool-specific C:N ratios. The decomposition 25 processes release biologically available forms of N  $(NO_3^- - N; NH_4^+ - N)$ . This allows 26 27 simulating N limitation on plant growth and biological N fixation as well as N feedbacks on organic matter decomposition and stabilization. Inorganic N is removed by sorption to soil 28 particles, plant uptake, immobilization into long-lived organic compounds, and hydrological 29 30 leaching, while organic N is lost through fire, hydrological leaching, and mineralization. Loss of nitrate N by soil denitrification was not differentiated from the hydrological nitrate-N
 leaching in LM3V-N.

3

# 4 2.3 Improved soil and river physics and hydrology

5 LM3 introduced vertically distributed soil-water, soil-ice and temperature profiles extending many meters below the surface, but with high resolution (thinnest layer 0.02 m) near the 6 7 surface. Water (potentially) discharges laterally from each soil layer to the local river reach. 8 Each horizontal grid cell of the model contains only one river reach, and each reach 9 discharges to another reach in the downstream grid cell, following a network that ultimately discharges to the ocean; the sub-grid-scale stream network is ignored. Relations among 10 11 discharge, storage, velocity, width, and depth in each reach are specified according to Leopold 12 and Maddock (1953).

13

#### 14 **2.4** Synthesis and Extension of Earlier Developments

For this study, we first combined the lumped N model LM3V-N with the distributed physics of LM3. To complete the N mass balance, we next added a soil denitrification module. Finally, we added stream transport and microbial processes to track the fate of soil N leaching and resolve N dynamics in the aquatic ecosystem. Each of these steps is described below. Figure 1 shows stores and fluxes of N in the resultant model, along with relevant processes. Newly introduced or adjusted parameters from the earlier developments are summarized in Table 1 and variables are listed in Table 2.

22

### 23 **2.4.1** Merging lumped N model with distributed physical model

To account for dependence of processes in the lumped soil C and N pools upon the vertically resolved physical states of the soil (temperature and water content), the latter were vertically averaged with an exponentially decaying weight function of depth (e-folding depth of 10 m). Leaching of any mobile constituent was defined as the product of a concentration and the sum

1 of lateral and vertical discharge from the soil layer between the surface and a depth of 10 m. The concentration of available N was calculated as dividing available N contents by the 2 3 effective soil depth which was approximated assuming C weight content 3.4% and average soil density 1500 kg/m<sup>3</sup>. The available N refers to the N contents reduced by buffering 4 5 factors which represent processes such as sorption to soil particles. To compensate for many 6 processes that were not accounted for in the model, calibration factors for each N species 7 were introduced to slow down overall N movement from the soil to the stream. These factors 8 include impacts of soil microbes, which are able to take up and incorporate all N forms  $(NO_3^- - N, NH_4^+ - N, DON)$  with a much greater capacity than plant uptake (Nordin et al., 9 10 2004). The nitrate calibration factor also accounts for storage in groundwater since nitrate (the 11 primary form of N in ground water) can persist for decades at high levels with increasing N 12 applications. This is further explained by Bachman et al. (1998) which reported that 17 to 80 percent of the N delivered to streams of the Chesapeake Bay watershed was through ground 13 14 water. Furthermore, the lumped single-layer N sub-model bypasses most of the vertically 15 distributed hydrologic system, and the soil N leaching based on the average water drainage is transferred directly from the N layer into the stream. These calibration factors were fit to 16 17 match inter-annual variations of reported and simulated stream N loads to make up for this 18 modeling approach as well as the unresolved processes that might cause inter-annual stream 19 N loads more sensitive to climate variability than those in reality. Considering its importance 20 in groundwater, a relatively larger size of the nitrate N factor is expected. The need to 21 incorporate these calibration factors, which are at the present basin specific, indicates that 22 future improvements to LM3-TAN should focus on resolving these processes (i.e. N cycle in 23 microbes, reservoirs, and vertically distributed soil layers). Dissolved organic, ammonium, 24 and nitrate N leaching from the soil are described as:

25 
$$L_{DON} = \frac{D_s}{\rho_w r_{DOM}} \left[ N_{DON,av} \right] = \frac{D_s}{\rho_w r_{DOM}} \left( \frac{f_{LF} N_{LF} + f_{LS} N_{LS} + f_{SS} N_{SS}}{b_{DOM} h_s} \right)$$
(1)

26 
$$L_{NH_4^+} = \frac{D_s}{\rho_w r_{NH_4^+}} \left[ N_{NH_4^+,av} \right] = \frac{D_s}{\rho_w r_{NH_4^+}} \left( \frac{N_{NH_4^+}}{b_{NH_4^+} h_s} \right)$$
 (2)

1 
$$L_{NO_{5}^{-}} = \frac{D_{s}}{\rho_{w} r_{NO_{5}^{-}}} \left[ N_{NO_{5}^{-},av} \right] = \frac{D_{s}}{\rho_{w} r_{NO_{5}^{-}}} \left( \frac{N_{NO_{5}^{-}}}{b_{NO_{5}^{-}} h_{s}} \right)$$
 (3)

$$2 h_s = \frac{c_{LF} + c_{LS} + c_{SS}}{r_c \rho_s} (4)$$

where,  $L_{DON}$ ,  $L_{NH_4^+}$ , and  $L_{NO_8^-}$  are the dissolved organic, ammonium, and nitrate N leaching 3 from the soil  $(kg/m^2 s)$ ;  $D_s$  is the water drainage from the active soil layer  $(kg/m^2 s)$ ;  $\rho_w$  is 4 the water density (1000 kg/m<sup>3</sup>);  $r_{DOM}$ ,  $r_{NH_4^+}$ , and  $r_{NO_5^-}$  are dissolved organic matter, 5 ammonium, and nitrate N calibration factors;  $[N_{DON,av}], [N_{NH_4^+,av}]$ , and  $[N_{NO_8^-,av}]$  are the 6 7 concentration of available N in dissolved organic, ammonium, and nitrate N pools  $(kg/m^3)$ ;  $N_{LF}$ ,  $N_{LS}$ , and  $N_{SS}$  are the fast litter, slow litter, and slow soil N contents (kg/m<sup>2</sup>);  $f_{LF}$ ,  $f_{LS}$ , 8 and  $f_{SS}$  are the fractions of soluble organic N in the fast litter, slow litter, and slow soil N 9 pools;  $N_{NH_4^+}$  and  $N_{NO_5^-}$  are the soil ammonium and nitrate N contents (kg/m<sup>2</sup>);  $b_{DOM}$ ,  $b_{NH_4^+}$ , 10 and  $b_{NO_s}$  are dissolved organic matter, ammonium, and nitrate N buffering factors due to 11 12 sorption to soil particles;  $h_s$  is the effective soil depth (m);  $r_c$  is the C weight content (3.4%);  $\rho_s$  is the average soil density (1500 kg/m<sup>3</sup>);  $C_{LF}$ ,  $C_{LS}$ , and  $C_{SS}$  are the fast litter, slow litter, 13 and slow soil C contents  $(kg/m^2)$ . 14

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#### 16 **2.4.2 Denitrification in soil**

17 Denitrification is a process that reduces nitrate or nitrite to gaseous forms (e.g., NO,  $N_2O$ ,  $N_2$ ) 18 in anaerobic conditions, where the oxidized N species serve as a terminal electron acceptor in 19 metabolism by soil denitrifying bacteria. The rate of denitrification generally depends on soil 20 nitrate content or concentration, soil water content (a surrogate for oxygen content), and soil temperature. Because soil nitrate contents are relatively low and limiting under natural conditions, we used a first-order loss function with respect to soil nitrate N content, with adjustments for the influence of soil water content and temperature to simulate soil denitrification rate:

$$5 \quad D_N = f_S f_T k_{denitr} N_{NO_B^-} \tag{5}$$

6 where,  $D_N$  is the soil denitrification rate  $(kg/m^2 yr)$ ;  $f_S$  is a soil water content reduction 7 function;  $f_T$  is a soil temperature reduction function;  $k_{denitr}$  is a first-order denitrification 8 coefficient (1/yr);  $N_{NO_5^-}$  is the soil nitrate N content  $(kg/m^2)$ .

9 
$$f_T = Q_{10}^{(T-T_T)/T_p}$$
 (6)

$$10 f_{S} = \begin{cases} S_{min} & S < S_{t} \\ \left(\frac{S-S_{t}}{S_{max}-S_{t}}\right)^{w} & S_{t} \leq S \leq S_{max} \\ S_{max} & S_{max} < S \end{cases}$$
(7)

11 where, *T* is the soil temperature (°C);  $T_r$  is a reference temperature (°C);  $T_p$  is a parameter;  $Q_{10}$ 12 is a factor change in rate with a 10 degree change in temperature; *S* is the soil water content; 13  $S_t$  is a threshold soil water content;  $S_{max}$  is the maximum soil water content;  $S_{min}$  is the 14 minimum soil water content; *w* is an empirical constant.

15

Heinen (2006) tabulates reported values of the various parameters introduced above. Figure 2 shows the effects of the reduction functions on the soil denitrification rate that were applied in diverse models as well as LM3-TAN. Figure 2a shows how fast soil nitrate N content is reduced to half of the initial amount depending on the different first-order denitrification coefficients. As temperature increases the bacterial activities increase exponentially (Fig. 2b). Soil denitrification occurs and increases non-linearly only if soil water content exceeds a certain threshold point due to enhanced anaerobic bacterial activity (Fig. 2c). The soil water

1 content reduction function for other microbial processes (e.g., mineralization, nitrification) 2 used in LM3-TAN is also shown in Fig. 2d. Because  $k_{denitr}$  is by far the most widely used 3 parameter of these, with reported values ranging over three orders of magnitude, our strategy 4 was to fix the other parameters using reported values, and to calibrate the model by determining  $k_{denitr}$  within the bounds reported in the literatures. Because soil denitrification 5 and nitrate-N leaching are competing sinks of nitrate N in the soil, soil denitrification 6 increases as soil nitrate-N leaching or stream nitrate-N load decreases; thus  $k_{denitr}$  was fit to 7 8 match reported and simulated stream nitrate-N loads.

9

10 The wide ranges of the functions discussed above are mostly driven by the dependencies of the parameters on specific regions (with different soil properties, vegetation, land use, etc.). 11 12 Given a number of proposed individual functions, it seems that there is no universal process 13 module to simulate soil denitrification. Because such reduction functions display a diversity 14 of shapes as ecosystems are modeled over a range of climate patterns, vegetation type, and 15 land use practices, soil denitrification on a large scale cannot be modeled without proper adjustments that compensate the site-specific properties. This explains why only a few studies 16 have applied models to watersheds larger than 1000 km<sup>2</sup> despite the diversity of existing 17 18 dynamic N models and why semi-distributed models often parameterize these individual 19 functions for each of sub-basins in large-scale applications. We hypothesize that LM3-TAN's integrated modeling framework, which is capable of simulating long-term vegetation 20 functional type and land use change as a function of changes in  $CO_2$ , climate, and human 21 22 influences, allows us to use a universal parameter set to simulate soil denitrification for each 23 of the distinct sub-basins. Still, care has to be taken when applying the model to other 24 watersheds that may be very different in terms of soil and climate properties from the 25 Susquehanna watershed. Furthermore, because soil denitrification becomes zero-order in 26 extreme nitrate rich environment, instead of using the first-order loss function for all of the 27 land use types, using a Monod function for agricultural land use may help LM3-TAN's global 28 application where N loadings would vary widely.

#### 1 **2.4.3** Microbial processes in rivers

2 Despite its importance to water quality, processes that control N removal from water bodies 3 are rarely resolved in watershed scale models, due to both uncertainties in measurement techniques and lack of measurements. To date, none of studies focusing on river 4 5 denitrification rate is based on measurements of an entire river network, but rather only on the 6 data from low order streams or individual catchments. Here we applied a non-linear 7 regression function based on the LINX (Lotic Intersite Nitrogen experiment; Mulholland et 8 al., 2008) reach-scale measurements that correlates river denitrification rate with nitrate-N 9 concentration and river depth to estimate the reaction rate constant of river denitrification for 10 each reach (Alexander et al., 2009). River denitrification happens mainly in the benthic and/or 11 hyporheic zones. Therefore, a river denitrification rate that is inversely proportional to the 12 river depth accounts for the ratio of water column to benthic area. The measured reaction rate constants vary from 0.034 to 117 (1/day), and we chose the median value 0.53 (1/day) as the 13 14 minimum reaction rate constant of river denitrification. Equation 12 indicates that the reaction 15 rate constant decreases with increase in nitrate-N concentration and river depth, since both  $b_1$ and  $b_2$  are negative, and it increases with temperature. Reaction rate constants for river 16

17 mineralization and nitrification were calibrated to match stream N loads.

18

Figure 1 shows structure of the river component. Each reach directly receives N from point sources (e.g., sewage and waste water discharge) and indirectly receives N from non-point sources (e.g., atmospheric deposition, fertilizer, manure, and legume applications) via soil leaching. The N loads in a reach are routed downstream with the water as following.

23 
$$\frac{dR_{DON}}{dt} = F_{DON}^{in} - F_{DON}^{out} + L_{DON} + P_{DON} - f_T' k_{min}' R_{DON}$$
(8)

24 
$$\frac{dR_{NH_{4}^{+}}}{dt} = F_{NH_{4}^{+}}^{in} - F_{NH_{4}^{+}}^{out} + L_{NH_{4}^{+}} + P_{NH_{4}^{+}} + f_{T}^{'}k_{min}^{'}R_{DON} - f_{T}^{'}k_{nitr}^{'}R_{NH_{4}^{+}}$$
(9)

25 
$$\frac{dR_{NO_{\rm g}}}{dt} = F_{NO_{\rm g}}^{in} - F_{NO_{\rm g}}^{out} + L_{NO_{\rm g}} + P_{NO_{\rm g}} + f_T' k_{nitr}' R_{NH_4} - f_T' k_{denitr}' R_{NO_{\rm g}}^{-}$$
(10)

26 
$$f_T' = T_p'^{(T'-T_p')}$$
 (11)

1 
$$k'_{denitr} = max \{ k'_{denitr,min}, C_{d,s} (b_0 C_{NO_3^{-b_1}} H^{b_2} c^t) \}$$
 (12)

where, i is DON,  $NH_4^+$ , or  $NO_3^-$ ;  $R_i$  is the river N (kg/m<sup>2</sup>);  $F_i^{in}$  and  $F_i^{out}$  is the inflow and 2 outflow of the river N (kg/m<sup>2</sup> s);  $L_i$  is the N leaching from the soil (kg/m<sup>2</sup> s);  $P_i$  is the N 3 point source  $(kg/m^2 s)$ ;  $f_T'$  is the stream temperature reduction function; T' is the water 4 temperature (°C);  $T_{p}$  is the reference water temperature (°C);  $T_{p}$  is a parameter;  $k_{min}$ ,  $k_{nitr}$ , 5 and  $k'_{denitr}$  are the reaction rate constants for river mineralization, nitrification, and 6 denitrification (1/s);  $k'_{denitr,min}$  is the minimum reaction rate constant of river denitrification 7 (1/s);  $C_{NO_{s}}$  is the nitrate N concentration (µmol N/l); *H* is the river depth (m);  $b_{0}$ ,  $b_{1}$ , and  $b_{2}$ 8 are the constants;  $c^{t}$  is the log re-transform bias correction factor;  $C_{d,s}$  is a unit-conversion 9 10 constant.

11

### 12 3 Study site

13 The Susquehanna River basin, where nearly four million people live, is the largest of the 14 watersheds in the northeastern U.S. and drains an area of 71,220 square kilometers, 15 contributing two-thirds of the annual N load to the Chesapeake Bay (Fig. 3). The basin includes 2,293 lakes, reservoirs, and ponds (322 km<sup>2</sup>) as well as 50,190 kilometers of rivers 16 17 and streams. The main stem of the Susquehanna River originates at Otsego Lake, N.Y., and flows about 750 kilometers through New York, Pennsylvania, and Maryland to the 18 19 Chesapeake Bay at Havre de Grace, Md. The Susquehanna Large River Assessment Project 20 reported that only 6.9 percent of water quality values exceeded their standards, but the 21 majority of theses exceedances were for nutrients (e.g. TN, TP) (Hoffman, 2009), explaining 22 why the Chesapeake Bay suffers from nutrient enrichment problems and hypoxia.

The reported 2000 land use is about 63 percent forest or wooded, 19 percent crop land, 7 1 2 percent pasture, 9 percent urban, and 2 percent water. The Upper Susquehanna River flows through mostly forested and agricultural land with some small communities and one larger 3 population center, then confluences with the Chemung River at Sayre, Pa. The West Branch 4 5 Susquehanna Sub-basin is mostly woods and grasslands. The Middle Susquehanna River, from the confluences with the Chemung River at Sayre, Pa. to the confluences with the West 6 7 Branch Susquehanna River at Sunbury, Pa., flows along very diverse land use. The Lower 8 Susquehanna Sub-basin contains extensive agriculture and several large population centers. 9 The other major urban areas are found within the Juniata Sub-basin (Hoffman, 2008).

10

The geology of the watershed is mainly clastic sedimentary rock of sandstone and shale. 11 12 Elevations vary from 30 meters at the Chesapeake Bay in Maryland to 955 meters in central New York State (McGonigal, 2011). The Great Lakes and Midwest climate exert influence 13 14 over the Upper Susquehanna, Chemung, and West Branch Susquehanna Sub-basin, whereas the Atlantic coastal climate affect on the other portions of the watershed. The basin has 15 16 experienced severe droughts about once every decade, and the worst droughts occurred in 17 1930, 1939 and 1964. The basin is also one of the most flood-prone watersheds in the nation 18 with frequent and localized flash floods every year. The worst recorded flooding in the basin 19 happened in 1972 as a result of tropical storm Agnes.

20

## 21 **4** Stream Sampling Description

22 Stream discharge data are provided by the network of stream gauges operated by the U.S. 23 Geological Survey (USGS), which collects and summarizes time series data to derive annual, 24 monthly, and daily stream discharge and statistics (Fig. 3). Chemical constituents of the 25 basin's water were monitored by the USGS and Susquehanna River Basin Commission 26 (SRBC). One USGS and six SRBC long-term nutrient monitoring sites monitored since 1985 27 and 9 newly introduced SRBC sites monitored since either 2004 or 2005 to present (Table 3; 28 Fig. 3; McGonigal, 2011; USGS, 2014) were chosen for model evaluation. The 16 sites vary in sub-basin area and land use. Among the USGS and SRBC sites, the Conowingo and 29 30 Marietta sites on the main channel of the Susquehanna River have the largest sub-basin areas 31 respectively (70,189 and 67,314  $\text{km}^2$ ). The sub-basin of the Conestoga site contains extensive

agriculture (48%) and the most populated urban land use with several large population centers 1 (24%) within a very small area (1,217 km<sup>2</sup>). The West Branch River flows mostly along 2 3 woods and grasslands to the Lewisburg site. The long-term sites have collected two samples 4 per month. Additional samplings are made during seasonal storm conditions. The collected 5 water samples are analyzed for various N species: dissolved N (DN), dissolved nitrite and 6 nitrate (DNO<sub>23</sub>), dissolved ammonia (DNH<sub>3</sub>), dissolved organic N (DON), and dissolved 7 ammonia and organic N (DKN) in milligrams per liter. In addition, annual, seasonal, and 8 monthly loads are computed by the Minimum Variance Unbiased Estimator (ESTIMATOR; 9 SRBC, 2006; USGS, 2014). River temperatures were reported when the samplings were collected for the chemical analysis of stream waters. 10

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# 12 5 Anthropogenic N sources

13 Anthropogenic N data over the two decades (1985-2005) were provided by the Chesapeake Community Modeling Program (CCMP). Atmospheric deposition data were provided by the 14 15 county-based land segments. Fertilizer, manure, and legume applications as well as combined 16 sewer overflows (CSOs) were provided by the land-river segments of the GIS-based Phase 17 5.3 Community Watershed Model (USEPA, 2010a). The atmospheric deposition data were 18 calculated by the Chesapeake Bay Program (CBP) Airshed Model, which is a combination of 19 a regression model of wet deposition (Grimm and Lynch, 2005) and the Community 20 Multiscale Air Quality Model (CMAQ) that estimates dry deposition (Dennis et al., 2007; 21 Hameedi et al., 2007). The fertilizer, manure, and legume data were estimated for the years of 22 1985, 1987, 1992, 1997, 2002, and 2005 by the Scenario Builder Version 2.2, a process based 23 model that is designed to use agricultural censuses as a main input data (USEPA, 2010b). The 24 agricultural censuses were produced by the United States Department of Agriculture National 25 Agricultural Statistics Service (NASS) and include data of animal populations, farms, 26 agricultural land areas, and crop yields. The point sources were estimated by 42 CSO 27 communities within the Susquehanna basin, using either various versions of EPA's Storm 28 Water Management Model (SWMM) or spatial data collected as a result of a direct survey of 29 the communities (USEPA, 2010a). The detailed data description can be found in the Phase 5.3 30 Community Watershed Model documentation (USEPA, 2010a).

Over the two decades, the total N sources decreased by about 20 percent. The atmospheric deposition was predominantly nitrate-N accounting for about 69 percent; ammonium-N 27 percent; organic-N 4 percent. The sum of the fertilizer, manure, and legume applications consisted of 49 percent of ammonium-N, followed by 37 percent organic-N, and 14 percent nitrate-N. Especially, the ammonium-N and organic-N loads had considerable variability across the spatial domain because they were strongly influenced by local emissions from the extensive agricultural areas.

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10 Figure 4 shows spatial distribution maps of the applied anthropogenic N sources, which were calculated as a spatial resolution of  $0.125^{\circ}$  by  $0.125^{\circ}$  and a temporal resolution of one year. 11 12 For each grid cell, which consists of up to 15 land-use tiles, atmospheric depositions (nitrate-13 N, ammonium-N, and organic-N) were applied to all of the land tiles, and fertilizer, manure, 14 and legume applications (nitrate-N, ammonium-N, and organic-N) were applied only to the 15 crop land tiles. Combined sewer overflows (nitrate-N, ammonium-N, and organic-N) were directly applied to the river reaches. The 20-year (1986-2005) average non-point and point N 16 17 sources for the six sub-basins are summarized in the Table 4. The thick solid arrows in Fig. 1 18 depict fluxes of each of N species for the anthropogenic N sources to the corresponding 19 terrestrial and river pools respectively.

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# 21 6 Model forcing and simulations

The model was implemented with a spatial resolution of  $0.125^{\circ}$  by  $0.125^{\circ}$  with time increments of 30 minutes. The model was forced using reported hydrological data cycled over a horizon of 61 years (1948-2008) to perform long-term simulations. The data include precipitation, specific humidity, air temperature, surface pressure, wind speed, and short and long wave downward radiation with a spatial resolution of 1° by 1°on timescales of 3 hours (Sheffield et al., 2006). Land-use change was simulated from 1704 to 2005 using a scenario of land use transitions (Hurtt et al., 2006). Preindustrial CO<sub>2</sub> concentration assumed as 286 ppm was applied from 1704 to 1799, and changes in  $CO_2$  concentrations were applied from 1800 to 2005 using an reported data from NOAA's Earth System Research Laboratory. For 250 years 3 (1704-1953), the estimated preindustrial N deposition (Dentener and Crutzen, 1944; Green et 4 al., 2004; Gerber et al., 2010) was applied as a uniform annual rate. We then applied the 5 reported 1985's anthropogenic N data from 1954 to 1984, and reported annual anthropogenic 6 N data from 1985 to 2005.

7

## 8 7 Result and Discussion

### 9 7.1 Evaluation of stream waters and N loads

We simulated with LM3-TAN stream dissolved organic-N, ammonium-N, and nitrate-N loads throughout the river network. The model was calibrated by comparing the modeled stream N loads with the corresponding reported N loads at the last downstream SRBC station Marietta, in which contributions of the entire watershed to the stream flows and N loads can be assessed. Thus, temporal evaluation of the stream discharges and N loads for the period 1987-2005 was focused on at the Marietta station. River data from the 16 monitoring stations (1986-2005) were also used to evaluate spatial stream discharges and N loads.

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18 Using global hydrological data and a universal parameter set for the entire watershed, the 19 model produced reasonable temporal patterns of annual stream discharge. The simulated 20 stream discharges were in a good agreement with the reported values in dry years and periods 21 (July to September), but under-estimated stream discharges in wet years and periods (March 22 to May). Overall, although the 19-year average simulated discharge was about 28% lower 23 than the corresponding reported value, their linear and rank correlations were significantly 24 high (Table 5), implying that the bias was systemic and accounted for in the calibration of the 25 N species.

26

27 Due to their complex physical and biogeochemical interactions with soil particles and soil 28 organic matter, simulating reactive transport of ammonium and dissolved organic N is far 29 more challenging than simulating nitrate N transport. For example, the correlation at Marietta

between stream discharge and nitrate N load ( $R^2 = 0.98$ ) was significantly higher than that for dissolved organic N ( $R^2 = 0.48$ ) or for ammonium N ( $R^2 = 0.85$ ) loads, implying that in 2 3 addition to the hydrological processes governing soil N transport to rivers, terrestrial physical 4 and microbial processes (e.g., sorption to soil particles, organic matter decomposition and 5 stabilization) have to be accounted for when estimating stream ammonium and dissolved 6 organic N loads. This, plus the fact that the highest component in the overall stream N load is 7 nitrate N, explains why existing watershed models have focused on stream nitrate N loads, 8 and neglected ammonium and dissolved organic N loads. Within the LM3-TAN's integrated 9 modeling framework, we estimated all of the N species for the entire drainage network.

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At Marietta, 19 year average simulated stream dissolved-N (-0.5%), nitrate-N (-0.2%), 11 12 ammonium-N (+4.7%), and dissolved organic-N (-2.6%) loads were close to the 13 corresponding reported values. Both of the simulated and monitored dissolved-N loads consisted of predominantly nitrate N (79%), followed by dissolved organic-N (18%), and 14 15 ammonium-N (3%). The model also produced reasonable temporal patterns of annual 16 dissolved-N (r = 0.7), nitrate-N (r = 0.6), ammonium-N (r = 0.7), and dissolved organic-N (r = 0.7) 17 0.6) loads (Fig. 5; Table 5). At Conowingo, 20 year average simulated nitrate-N load agreed 18 well with the corresponding reported value (-3.7%), but the model, which doesn't have lakes or reservoirs, fails to capture inter-annual variations of the loads (r = 0.2), which are affected 19 20 by the reservoir system between the Marietta and Conowingo monitoring sites (Fig. 5).

21

22 and  $kg/km^2$  yr (normalized by its sub-basin area summarized in Table 3). Among the 6 long-23 24 term monitoring sites, the highest and lowest amount of river N loads were reported and 25 simulated at the Marietta and Conestoga sites respectively (Fig. 6a). This finding is consistent 26 with the general view that the amount of stream N loads is proportional to size of the basin 27 area. A very high N flux was reported at the Conestoga site (Fig. 6b), which can be explained 28 by its sub-basin's extensive agriculture and urban land use. Because the West Branch 29 Susquehanna is dominated mostly by woods and grasslands, the Lewisburg site had the lowest

Simulated and reported dissolved -N loads were graphed in different units: millions of kg/vr

N flux. The model also captured the stream N loads at the 15 monitoring sites well (Fig. 6c and 6d). These results attest to the model ability to correctly simulate the stream N loads for the entire basin based on the climate as well as land use and the corresponding N sources and transformations in the sub-basins.

5

# 6 7.2 Spatial distribution of stream N load and soil denitrification rate

7 Observation of the spatial distribution of the river N load (Fig. 7) and soil denitrification rate 8 (Fig. 8d) helps to identify the extent of the terrestrial and aquatic N pollution across the basin. 9 A large amount of N is exported via the main stem of the Susquehanna River as well as its 10 three major tributaries, where many small-order streams converge. The N loads in the streams 11 increase gradually from the headwaters to the watershed outlet, implying that the N loads to 12 the rivers exceed N removal mechanisms within the rivers. Although stream N loads are in general higher in the larger rivers, at the Lower Susquehanna sub-basin, high N loads are 13 14 present even in small-order streams due to the extensive agricultural land use.

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16 Figure 8 presents 20 year average (1986-2005) simulated soil water content, temperature, 17 nitrate-N content, and denitrification rate, and these for each of 6 sub-basins as well as the 18 corresponding sub-basin area, non-point and point N sources are summarized in Table 4. An 19 analysis for the 6 sub-basins shows that the combined effects of land use and climate on the 20 soil denitrification rate, which were the highest in the Lower Susquehanna Sub-basin 21 (extensive agriculture; Atlantic coastal climate) and the lowest in the West Branch 22 Susquehanna Sub-basin (mostly forest; Great Lakes and Midwest climate). These results show that the most significant soil denitrification is associated with extensive agricultural land 23 use (non-point sources). The calculated  $\mathbb{R}^2$  statistic between the monthly soil denitrification 24

rate and soil water content ( $\mathbb{R}^2 = 0.51$ ) was significantly higher than that for soil temperature or soil nitrate N content, implying that the soil water content played the greatest role in the soil denitrification process among the three factors. This is because the soil denitrification occurred and increased non-linearly only when the soil water content exceeded the threshold point ( $S_t = 0.577$ ). The significant effect of the soil water content on the soil denitrification is further illustrated in the upper east side of the Upper Susquehanna Sub-basin, where
 extremely low soil water content (Fig. 8a) impeded the overall soil denitrification process
 (Fig. 8d).

4

# 5 7.3 N budget

As a further means of evaluating the model output, we compared the simulated N budget for the period 1988-1992 to the budget constructed by Boyer et al. (2002), Seitzinger et al., (2002), and Van Breemen et al., (2002) for the same period (Fig. 9). Overall, reasonable agreements were found between these two budgets. Total N inputs to the whole basin were reported as 4,774 (kg/km<sup>2</sup> yr; atmospheric deposition + fertilizer + forest and agricultural N fixation + net N import in feed and food), while we applied 4,443 of N (kg/km<sup>2</sup> yr;

12 atmospheric deposition + fertilizer + manure + legume + sewage) using the data sources 13 provided by CCMP (USEPA 2010a). The simulated soil denitrification (-4%), harvest rates (+7%), river export (-1%), and river denitrification (-5%) agreed well with the corresponding 14 15 reported values. To investigate the importance of N removal within rivers, we ran an experiment in which the reaction rate constant for river denitrification was set to zero. We 16 17 then compared N loads within the rivers with and without river denitrification. Figure 10 18 shows a spatial map of the difference in N loads between these simulations, which represents 19 the river N removal. A large amount of N was removed along the main stem of the 20 Susquehanna River as well as its three major tributaries, implying that the N removal 21 increases gradually as distance from the headwaters increases. About 28 percent of the N that 22 enters to the rivers was removed by river denitrification.

23

For the entire basin, we divided the simulated land use into either agricultural land (cropland and pasture) or secondary forest (land formerly disturbed by human activities). We then graphed simplified N budgets for each land use (Fig. 9c). The reported agricultural land use was 29 percent (Fig. 9a), whereas the model simulated 24 percent of cropland and pasture (Fig. 9b). In the secondary forest land, most of the applied N (43%) was stored in the terrestrial system (vegetation and soil pools), whereas the highest proportion of the applied N was removed by soil denitrification (44%) in the agricultural land. These results imply that

1 applications of artificial N to agricultural lands can result in considerable soil denitrification 2 rates, and thus significant increase of  $N_2O$  production. This is evident when comparing maps 3 of the applied fertilizer, manure, and legume N applications (Fig. 4c and 4d) and the 4 simulated soil denitrification (Figure 8d) that corresponds well, especially in the Lower 5 Susquehanna Sub-basin with extensive agricultural land use. Even if there are some 6 discrepancies between these two budgets, we can conclude that the reactive transport of N 7 from the terrestrial to aquatic ecosystems was appropriately simulated by the model, 8 providing suitable descriptive information for the entire drainage network.

9

# 10 8 Conclusions

Results of our study show that LM3-TAN captures well the key mechanisms that control N
dynamics in the climate-plant-soil-river system. Specifically, we demonstrate:

On a sub-basin scale with different climate and land-use regimes, the LM3-TAN properly simulates terrestrial N cycling, including effects of long-term vegetation dynamics, land-use changes, and hydrological cycles. The interaction among those three processes allow LM3-TAN to capture soil C-N organic matter and mineral N transformations as well as soil emissions of nitrate-N and leaching of dissolved organic, ammonium, and nitrate N.

The ability to capture N soil budget and losses then enables LM3-TAN to consistently
 characterize trends and variability in riverine N inputs and exports of ammonium,
 dissolved organic, and nitrate N with explicit representation of their transformations and
 transport in rivers.

In the re-growing secondary forests, a large fraction of the N from atmospheric deposition has been stored in the vegetation and soil, but in the agricultural lands most N inputs were removed by soil denitrification indicating that anthropogenic N inputs could drive substantial increase of N<sub>2</sub>O emission, an intermediate of the denitrification process.

LM3-TAN captures effects of long-term trends and variability of hydrological cycles
 (e.g., precipitation, soil water content, stream discharge) on N cycling in vegetation-soil river system, and thus resolves inter-annual variations of stream N loadings caused by
 climate variability.

- The model results suggest that the soil denitrification is most sensitive to soil water
  variations.
- Among the 6 sub-basins, the soil denitrification rate was the highest in the Lower
  Susquehanna Sub-basin with the most intensive land-use non-point N sources as well as
  with the warmest and wettest soils, attributed to the Atlantic coastal climate.
- Even though the N denitrification and riverine biogeochemistry N modules were
   calibrated only at the last downstream station Marietta, application of the universal
   parameters at the entire watershed produced simulations which compared well at other
   observational stations. The applicability of the universal parameters in other watersheds
   is a subject of the future research.
- This study shows that linking terrestrial N and C cycling, long-term land-use and vegetation dynamics, and hydro-climate variations to N loads and concentrations in streams, provides an effective and consistent framework for analysis of the surface water N processes and water quality for large watersheds and basins.

## 16 Acknowledgements

We thank Krista A. Dunne from the United States Geological Survey for help and advice in
using and interpreting the hydrologic module of LM3-TAN, including creation of the 1/8
degree river network dataset.

20

We thank Joel Blomquist from the United States Geological Survey for assisting with
interpretation of N load data at the USGS Conowingo sampling station. We thank Kevin
McGonigal from the Susquehanna River Basin Commission for providing us with the various
N load data.

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Support for M. Lee was provided by a Fulbright Scholarship, by the Princeton Environmental
Institute at Princeton University through the Mary and Randall Hack '69 Research Fund, and
by the Korean National Institute of Environmental Research.

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- 24

1 Table 1. Newly introduced or adjusted parameters from the earlier developments.

Parameter	Description	Value Unit		Reference or Rationale						
Parameters in the Land Component Equations										
$b_{DOM},$ $b_{NH_4^+}, b_{NO_5^-}$	buffering factors for DOM, ammonium-N, nitrate-N	3, 5, 1	unitless	Leadly et al., 1997; Neff and Asner, 2001						
f <sub>DOM</sub>	fraction of litter soil decomposition that becomes potential DOM (Gerber et al., 2010)	0.034	unitless	calibrated to match stream DON loads; Gerber et al., 2010						
k <sub>denitr</sub>	first-order denitrification coefficient	6.5	1/yr	Heinen, 2006						
$r_{\rm DOM}, r_{\rm NH_4^+},$ $r_{\rm NO_5^-}$	calibration factors for DOM, ammonium-N, nitrate-N	10, 20, 100	unitless	calibrated to match inter-annual variations of stream N loads						
$q_{max}$	transfer fractions form slow litter to slow soil (Gerber et al., 2010)	0.6	unitless	Parton et al., 1993; Bolker et al., 1998; Gerber et al., 2010						
q <sub>sp</sub>	transfer fractions form slow litter to passive soil (Gerber et al., 2010)	0.004	unitless	Parton et al., 1993; Bolker et al., 1998; Gerber et al., 2010						
S <sub>min</sub>	minimum soil water content	0	unitless	Bril et al., 1994; Heinen, 2006						
S <sub>max</sub>	maximum soil water content	1	unitless	Bril et al., 1994; Heinen, 2006						
S <sub>t</sub>	threshold soil water content	0.577	unitless	Bril et al., 1994; Heinen, 2006						
W	empirical constant	2	unitless	Bril et al., 1994; Heinen, 2006						
T <sub>p</sub>	parameter	10	unitless	Sogn and Abrahamsen, 1997; Johnsson et al., 1987; Heinen, 2006						
$T_r$	reference temperature	15	°C	Sogn and Abrahamsen, 1997; Johnsson et al., 1987; Heinen, 2006						

<i>Q</i> <sub>10</sub>	factor change in rate with a 10 degree change in temperature	2	unitless	Sogn and Abrahamsen, 1997; Johnsson et al., 1987; Heinen, 2006	
	Parameters in the Ri	ver Componen	t Equation	ns	
$b_0, b_1, b_2$	constants	0.559, -0.478, -0.612	unitless	Alexander et al., 2009	
ct	log re-transform bias correction factor	1.90	unitless	Alexander et al., 2009	
$k^{'}_{denitr,min}$	minimum reaction rate constant of river denitrification	0.53/86400	1/s	Alexander et al., 2009	
C <sub>d,s</sub>	unit-conversion constant	1/86400	day/s	conversion from 1/day to 1/s	
$k^{'}_{min},k^{'}_{nitr}$	reaction rate constants for river mineralization, nitrification	0.11/86400, 0.51/86400	1/s	calibrated to match stream N loads	
$T_{p}^{'}$	parameter	1.047	unitless	Wade et al., 2002	
$T_r^{'}$	reference water temperature	20	°C	Wade et al., 2002	

- 1 Table 2. Definition of prognostic (PV) and diagnostic (DV) variables and inputs/forcings (IF)
- 2 used in the equations.

Vegetation and Soil Equations							
$C_{LF}, C_{LS}, C_{SS}$	PV   fast litter, slow litter, slow soil C contents		kg/m <sup>2</sup>				
$D_N$	DV	soil denitrification rate	kg/m² yr				
$D_s$	DV	water drainage from active soil layer	kg/m² s				
$f_{LF}, f_{LS}, f_{SS}$	PV	fractions of soluble organic N in the fast litter, slow litter, slow soil N pools (Gerber et al., 2010)	unitless				
fs	PV	soil water content reduction function	unitless				
$f_T$	PV	soil temperature reduction function	unitless				
$h_s$	PV	effective soil depth	m				
$L_{\rm DON}, L_{\rm NH_4^+}, L_{\rm NO_5^-}$	PV	soil leaching for DON, ammonium-N, nitrate-N	kg/m² s				
$\begin{split} \begin{bmatrix} N_{DON,av} \end{bmatrix}, \begin{bmatrix} N_{NH_4^+,av} \end{bmatrix}, \\ \begin{bmatrix} N_{NO_8^-,av} \end{bmatrix} \end{split}$	PV	concentration of available N in DOM, ammonium-N, nitrate-N pools	kg/m <sup>3</sup>				
$N_{LF}, N_{LS}, N_{SS}$	PV	fast litter, slow litter, slow soil N contents	kg/m²				
$N_{NH_4^+}, N_{NO_5^-}$	PV	soil ammonium-N, nitrate-N contents	kg/m <sup>2</sup>				
S	PV	soil water content	unitless				
Т	T PV soil temperature		°C				
	1	<b>River Equations</b>					
C <sub>NOs</sub>	PV	/ nitrate-N concentration µ					
$f_T'$	PV	stream temperature reduction function	unitless				
$F_{DON}^{in},F_{NH_4}^{in},F_{NO_5}^{in}$	$_{N}, F_{NH_{4}^{+}}^{in}, F_{NO_{5}^{-}}^{in}$ DV river inflow of DON, ammonium-N, nitrate-N		kg/m² s				

$F_{DON}^{out},F_{NH_4}^{out},F_{NO_8}^{out}$	DV	river outflow of the DON, ammonium-N, nitrate-N	kg/m² s
Н	IF	river depth	m
$k_{\scriptscriptstyle denitr}'$	PV	reaction rate constant for river denitrification	1/s
$P_{DON}, P_{NH_4^+}, P_{NO_5^-}$	IF	IF point sources of DIN, ammonium-N, nitrate-N	
$R_{DON}, R_{NH_4^+}, R_{NO_5^-}$	DV	DON, ammonium-N, nitrate-N in rivers	
Τ'	PV	water temperature	°C

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- 1 Table 3. Susquehanna River Basin Geographic Statistics for the USGS and SRBC nutrient
- 2 monitoring sites (McGonigal, 2011; USGS, 2014).

		Sub basin	2000 Land Use Percentages					
Site Location	Waterbody	Area,	Water/		Agricu	ltural		
		km <sup>2</sup>	Wetland	Urban	Cropland	Pasture	Forest	Other
		7 Long-	term Sites					
Towanda, 1989~	Susquehanna	20,194	2	5	17	5	71	0
Danville, 1985~	Susquehanna	29,060	2	6	16	5	70	1
Lewisburg, 1985~	W B Susque	17,734	1	5	8	2	84	0
Newport, 1985~	Juniata	8,687	1	6	14	4	74	1
Marietta, 1987~	Susquehanna	67,314	2	7	14	5	72	0
Conestoga, 1985~	Conestoga	1,217	1	24	12	36	26	1
Conowingo, 1985~	Susquehanna	70,189	2	9	7	19	63	0
		9 Newly Int	troduced S	ites			1	
Conklin, 2005~	Susquehanna	5,778	3	3	18	4	71	1
Smithboro, 2004~	Susquehanna	11,989	3	5	17	5	70	0
Campbell, 2005~	Cohocton	1,217	3	4	13	6	74	0
Chemung, 2004~	Chemung	6,488	2	5	15	5	73	0
Wilkes-Barre, 2004~	Susquehanna	25,785	2	6	16	5	71	0
Karthaus, 2004~	W B Susque	3,785	1	6	11	1	80	1
Castanea, 2004~	Bald Eagle	1,087	1	8	11	3	76	1
Saxton, 2004~	Raystown B Juni	1,957	< 0.5	6	18	5	71	0
Manchester, 2004~	W Conewago	1,320	2	13	12	36	36	1

- 1 Table 4. Sub-basin area, 20 year (1986-2005) average applied non-point and point N sources,
- 2 and simulated soil water content, temperature, nitrate-N content, and denitrification rate (% of
- 3 the non-point N sources) for each of 6 sub-basins.

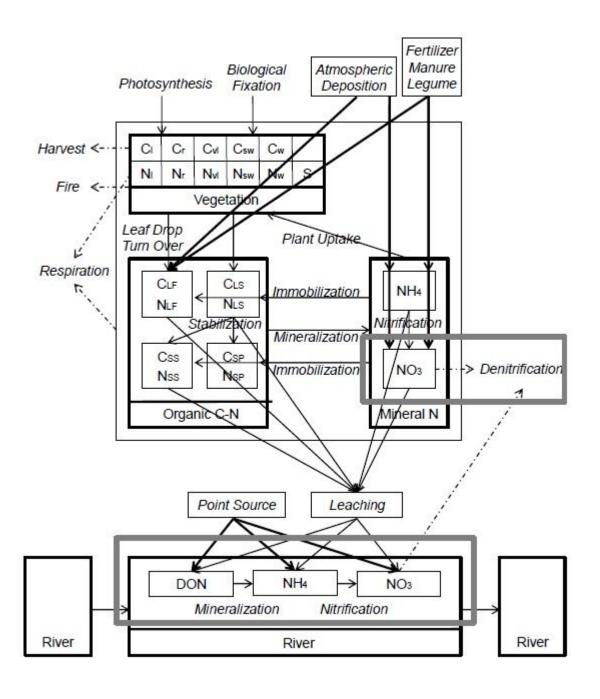
6 Sub- basins	Basin Area, km <sup>2</sup>	Non-point N Sources kg/km <sup>2</sup> yr	Point N Sources kg/km <sup>2</sup> yr	Soil Water Content	Soil Temp. C	Soil Nitr. N kg/km <sup>2</sup>	Soil Denitr. kg/km <sup>2</sup> yr
Upper Susquehanna	14,126	3,315	40	0.439	8.65	12,713	1,213 (37%)
Chemung	6,731	2,962	76	0.454	8.60	9,888	916 (31%)
Middle Susquehanna	9,847	3165	331	0.459	9.39	11,599	1,142 (36%)
West Branch Susquehanna	18,447	3,163	70	0.458	9.24	11,746	959 (30%)
Juniata	8,686	4,553	41	0.480	10.58	17,002	1,538 (34%)
Lower Susquehanna	16,070	6,098	163	0.463	10.27	27,358	2,717 (45%)

Table 5. Temporal evaluation of the annual stream discharges and N loads for the period 

1987-2005 at Marietta. If a p-value is smaller than 0.05, the correlation between the modeled

and reported data is significantly different from zero. 

		Discharge	DN	Nit. N	Amm. N	DON
R <sup>2</sup>		0.6	0.5	0.4	0.5	0.4
	Pearson's linear	0.7	0.7	0.6	0.7	0.6
Corr.	(p-value)	(< 0.0001)	(< 0.0001)	(0.0044)	(< 0.0001)	(0.0064)
Coef	Spearman's rho	0.7	0.7	0.6	0.6	0.6
	(p-value)	(0.0011)	(< 0.0001)	(0.0056)	(0.0099)	(0.0160)



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Figure 1. Structure of LM3-TAN. Two thick boxes show the incorporated denitrification module in the terrestrial component and stream microbial processes in the river component. The river systems are a series of continuously stirred tank reactors (CSTR) that simulate stream mineralization, nitrification, and denitrification. The other boxes show major C and N pools in vegetation (leaves, fine roots, labile, sapwood, heartwood, and N buffer storage), soil (fast and slow little, slow and passive soil, mineral N), and river (organic and mineral N). The arrows depict fluxes of anthropogenic N sources (thick solid), C-N organic compounds and

- 1 mineral N (thin solid) with associated processes (italic), and C and N lost to the atmosphere or
- 2 anthropogenic pool (dashed).

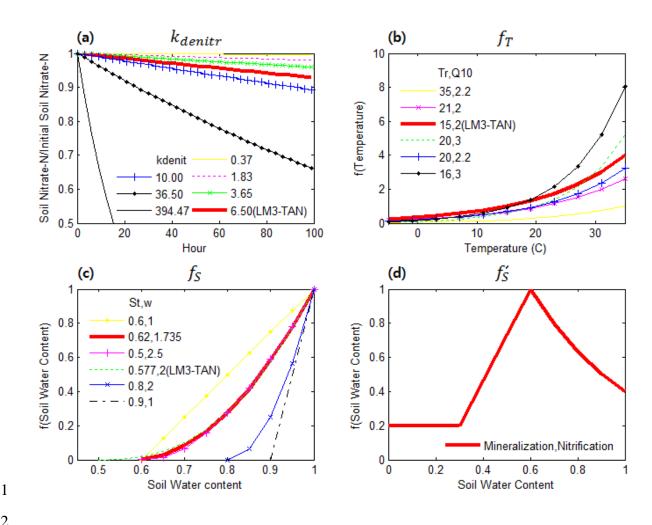




Figure 2. Overview of the denitrification module. Effects of first-order denitrification coefficient (a), soil temperature reduction function (b), soil water content reduction function (c) on soil denitrification rate; soil water content reduction function for mineralization and nitrification (d). The curves were produced using the Table 3, 6, and 7 in Heinen (2006).

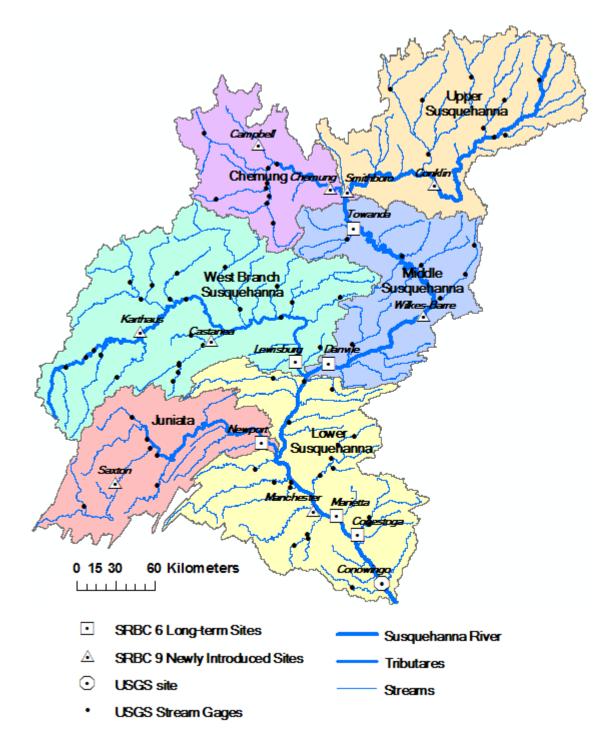




Figure 3. Map of the Susquehanna watershed, showing 6 major sub-basins, main stem of the
Susquehanna River, major tributaries (Chemung, West Branch Susquehanna, and Juniata
River), streams, and the location of USGS stream gauges and USGS and SRBC nutrient
monitoring sites.

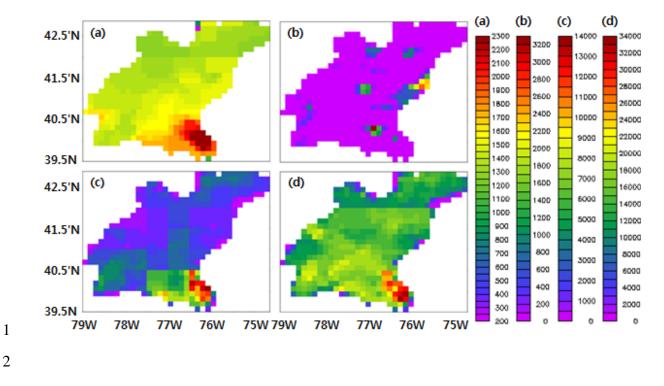


Figure 4. Spatial distribution maps of the applied 20 year (1986-2005) average anthropogenic
N sources: atmospheric deposition (kg/km<sup>2</sup> year) (a), combined sewer overflow
(kg/km<sup>2</sup> year) (b), and fertilizer, manure, and legume applications (kg/km<sup>2</sup> year) (c) and
(kg/crop land km<sup>2</sup> year) (d).

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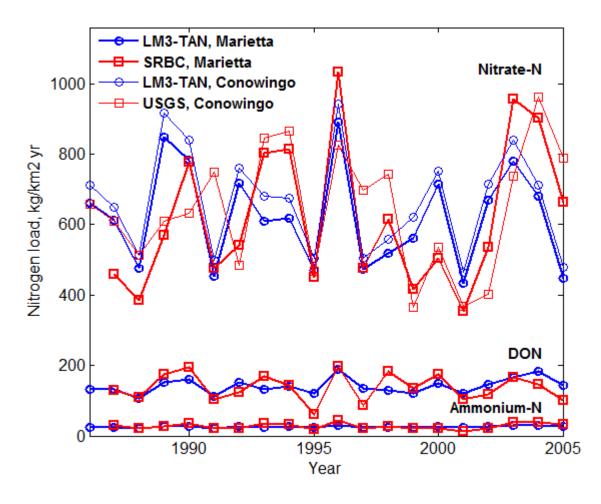
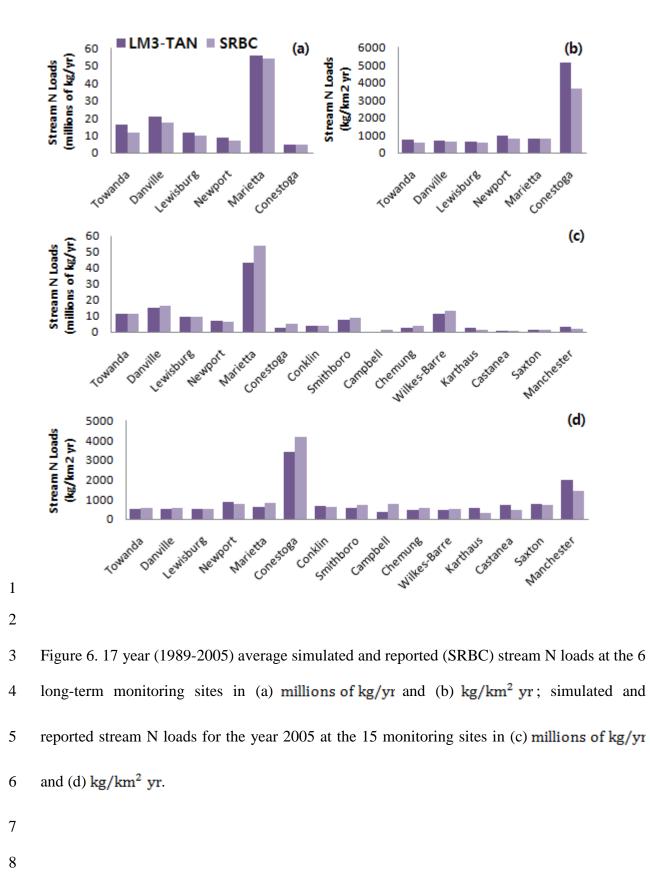


Figure 5. 20 years (1986-2005) of the simulated stream N loads (normalized by sub-basin
areas) at Marietta and Conowingo and the corresponding reported data from SRBC and USGS.

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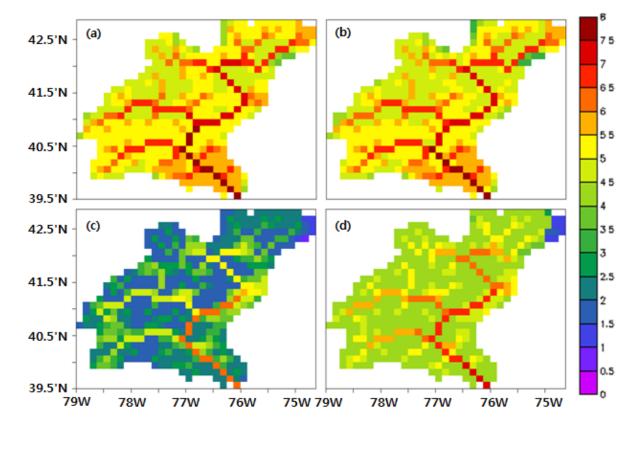
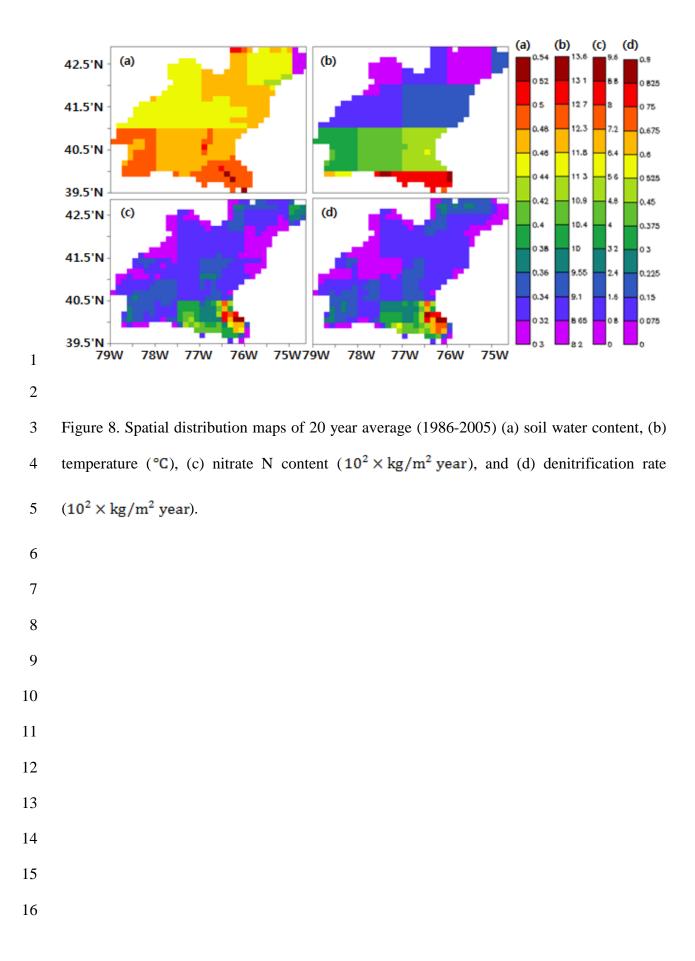


Figure 7. Spatial distribution maps of 20 year (1986-2005) average simulated stream (a)
dissolved N, (b) nitrate N, (c) ammonium N, and (d) dissolved organic N loads,
log (kg/year).



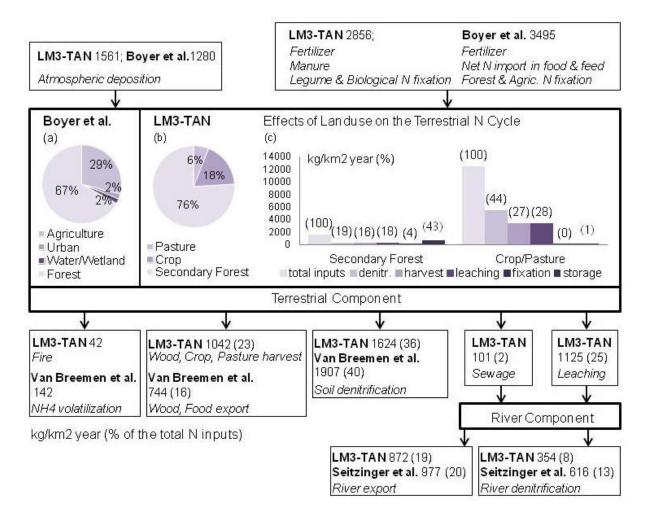
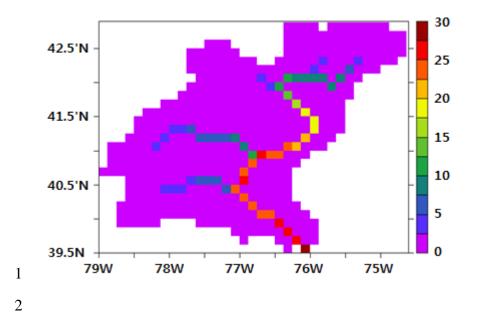


Figure 9. Comparison between the calculated and reported budgets of N sources, retention,
lost, transport, and river export at the level of the whole Susquehanna watershed for the
period 1988 to 1992 (Boyer et al., 2002; Breemen et al., 2002; Seitzinger et al., 2002;
USEPA, 2010a).

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3 Figure 10. N removal by river denitrification (%) = (river N load with " $k'_{denitr} = 0$ " - river N

4 load with "estimated  $k'_{denitr}$ ") / river N load with " $k'_{denitr} = 0$ "× 100.