

An Integrated Approach for Predicting Nutrient Transports from the Land to the Ocean

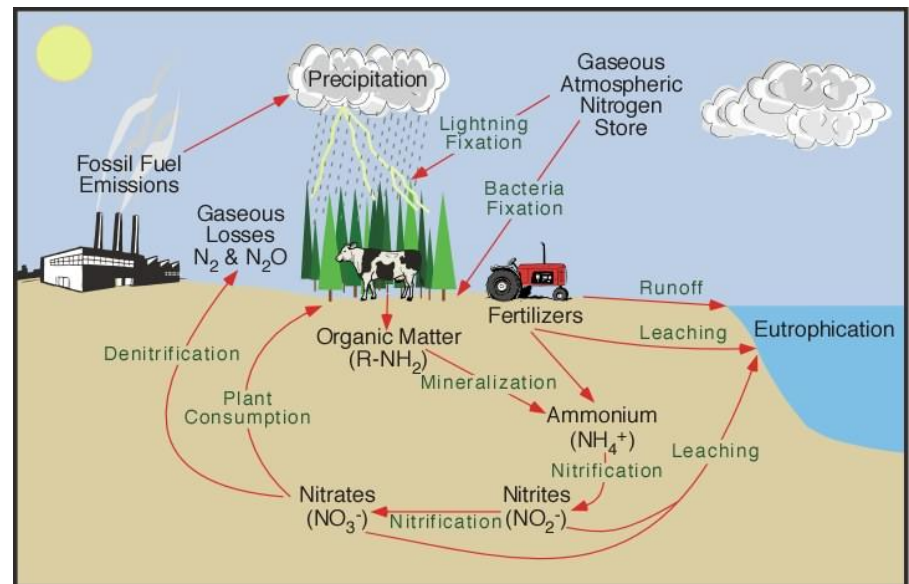
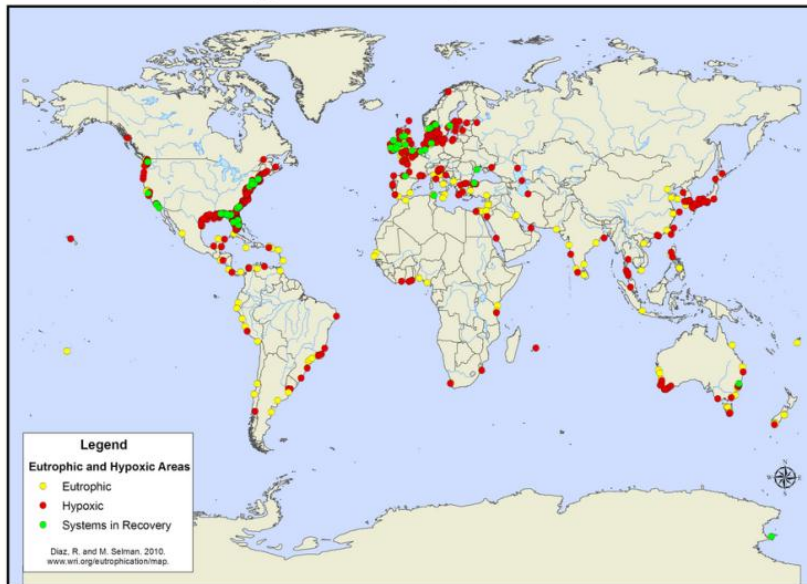
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Issues in Riverine Nutrient Export Research

- Lacking integrated climate, extreme weather, land surface, river flow, biogeochemistry, and ecological models

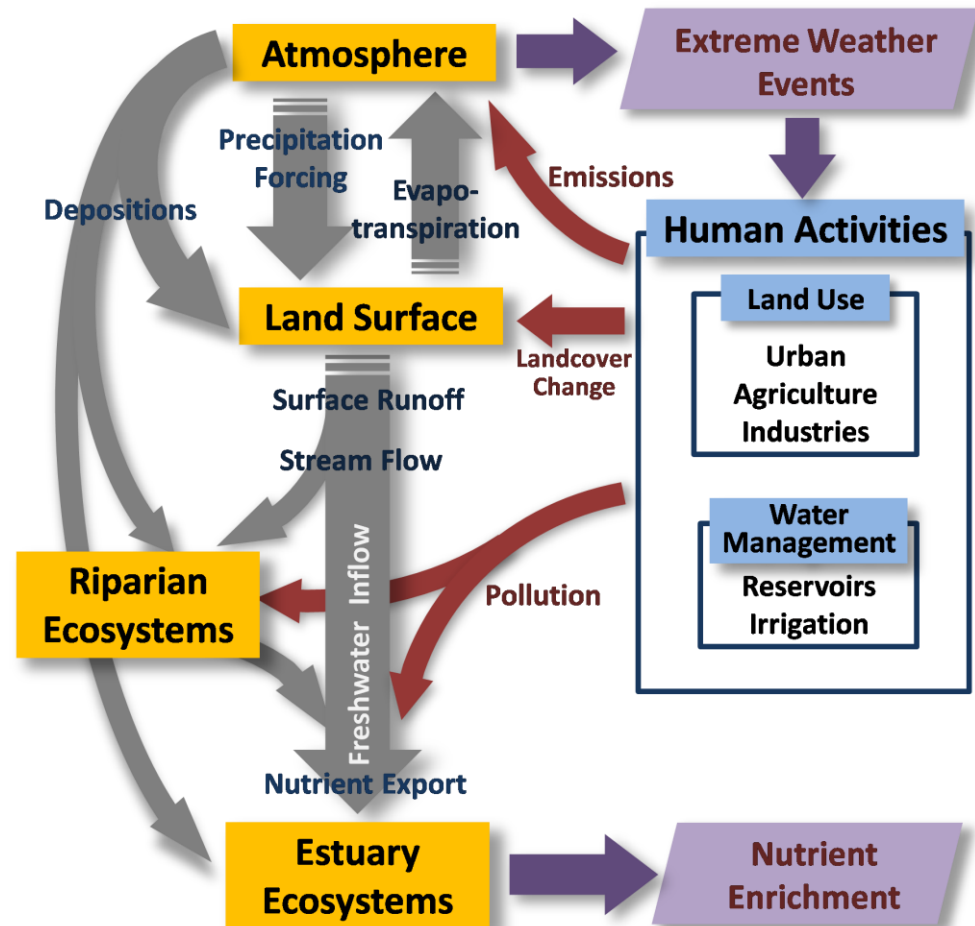
World Hypoxic and Eutrophic Coastal Areas



<http://www.wri.org/map/world-hypoxic-and-eutrophic-coastal-areas>

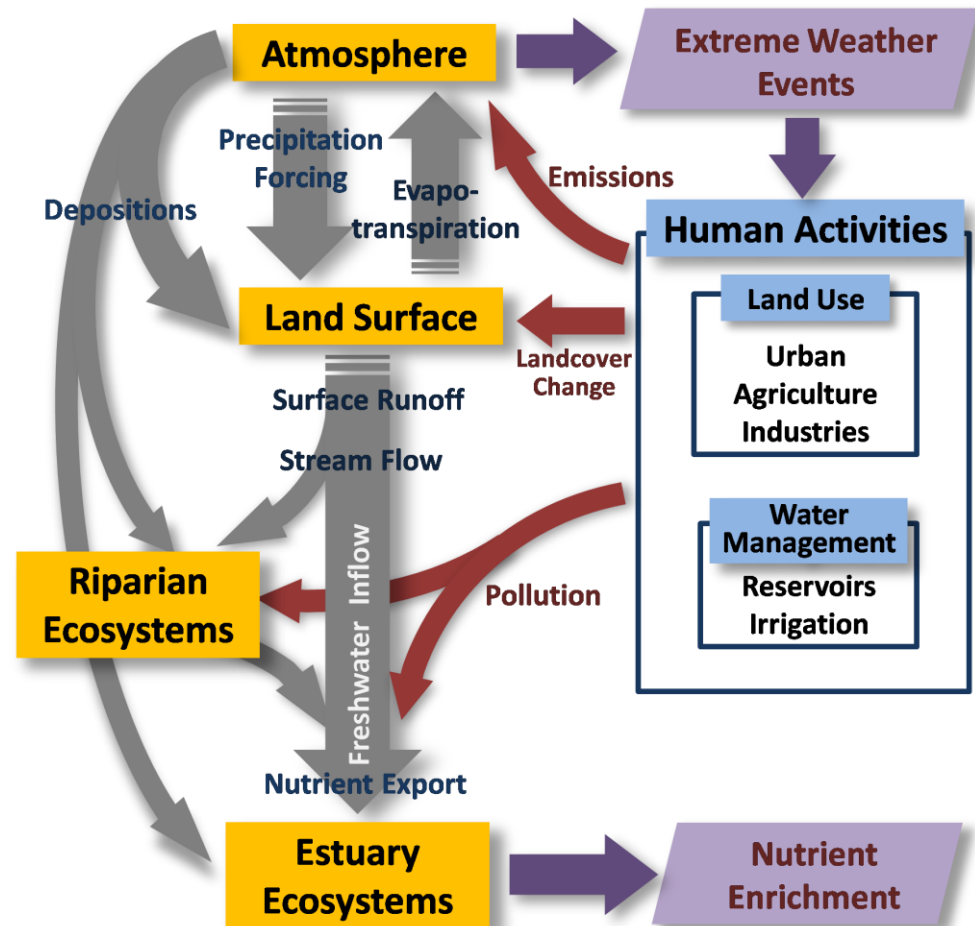
How do we deal with these issues?

- Develop an integrated approach
 - United States NASA Interdisciplinary Research in Earth Science (IDS)

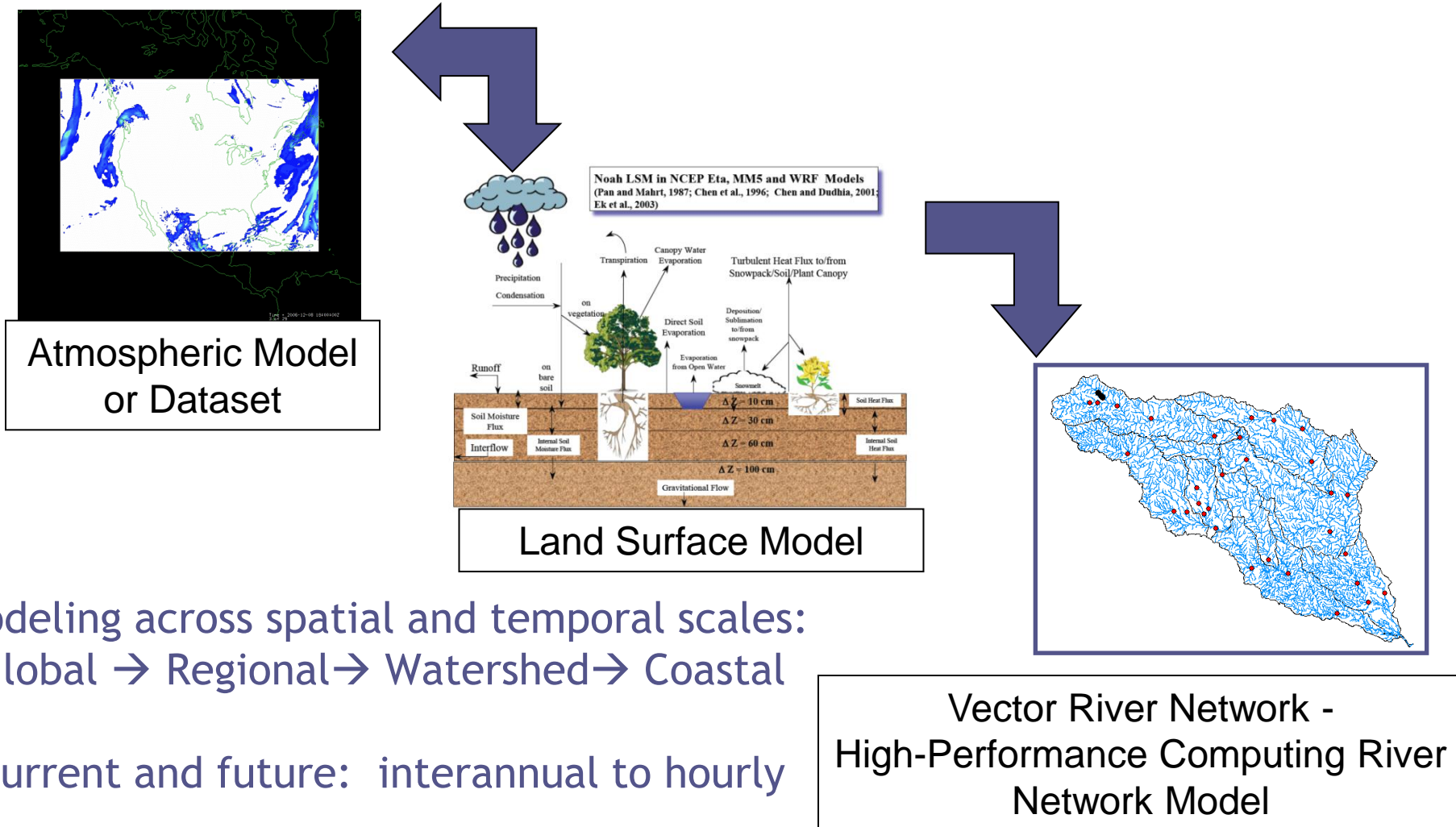


Integrated Approach

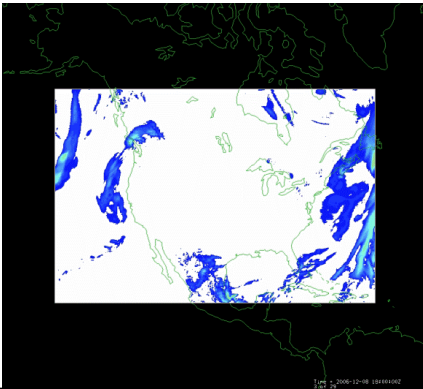
- Global Climate Model
 - (CESM)
- Regional Forecast Model
 - (WRF)
- Land Surface Model
 - (Noah-MP)
- River Flow Model
 - (RAPID)
- Simple estuary model



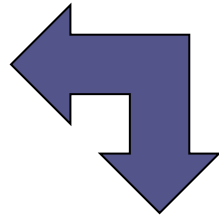
Framework for calculation



Framework for calculation



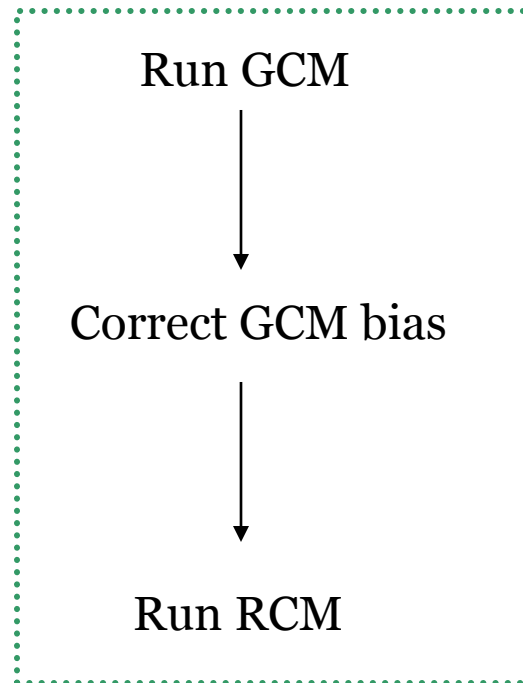
Atmospheric Model
or Dataset



Modeling across spatial and temporal scales:
Global → Regional

The New Dynamic Downscaling (NDD) method

- Central Idea
 - Correct climatological mean bias in GCM outputs



Methodology of GCM bias correction

$$CAM = \overline{CAM} + CAM'$$

$$NNRP = \overline{NNRP} + NNRP'$$

- Bias correction 1:

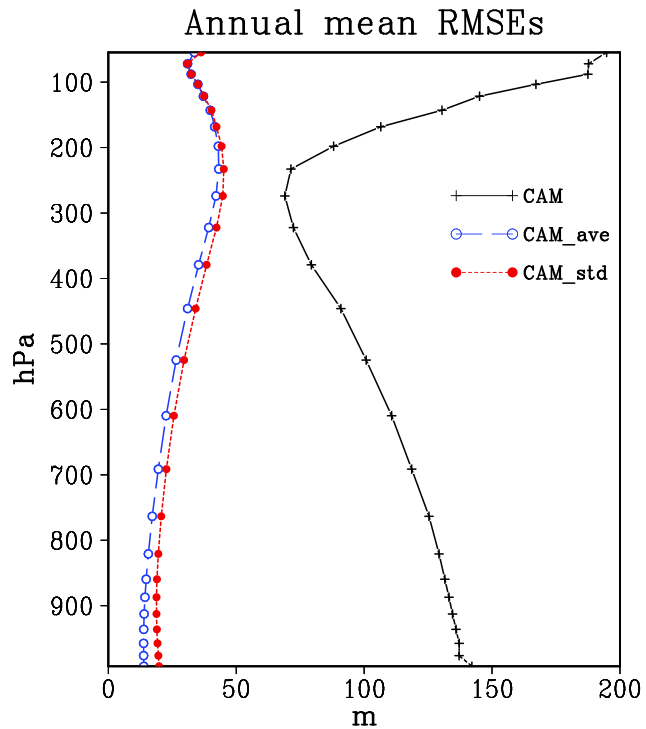
$$CAMbc1 = NNRP + CAM'$$

- Bias correction 2:

D : standard deviation

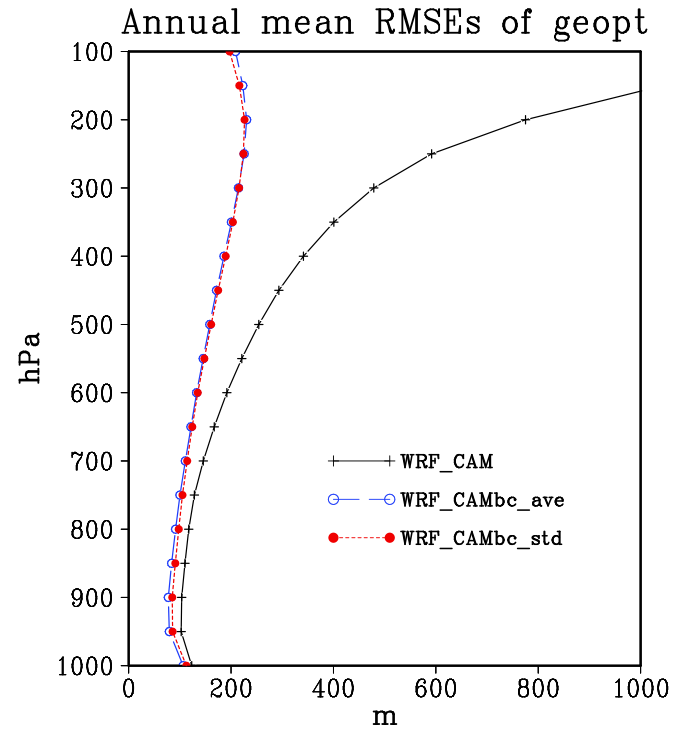
$$CAMbc2 = \overline{NNRP} + CAM' \cdot \frac{D_{NNRP}}{D_{CAM}}$$

Annual mean RMSEs in GCM and RCM



RMSE in GCM

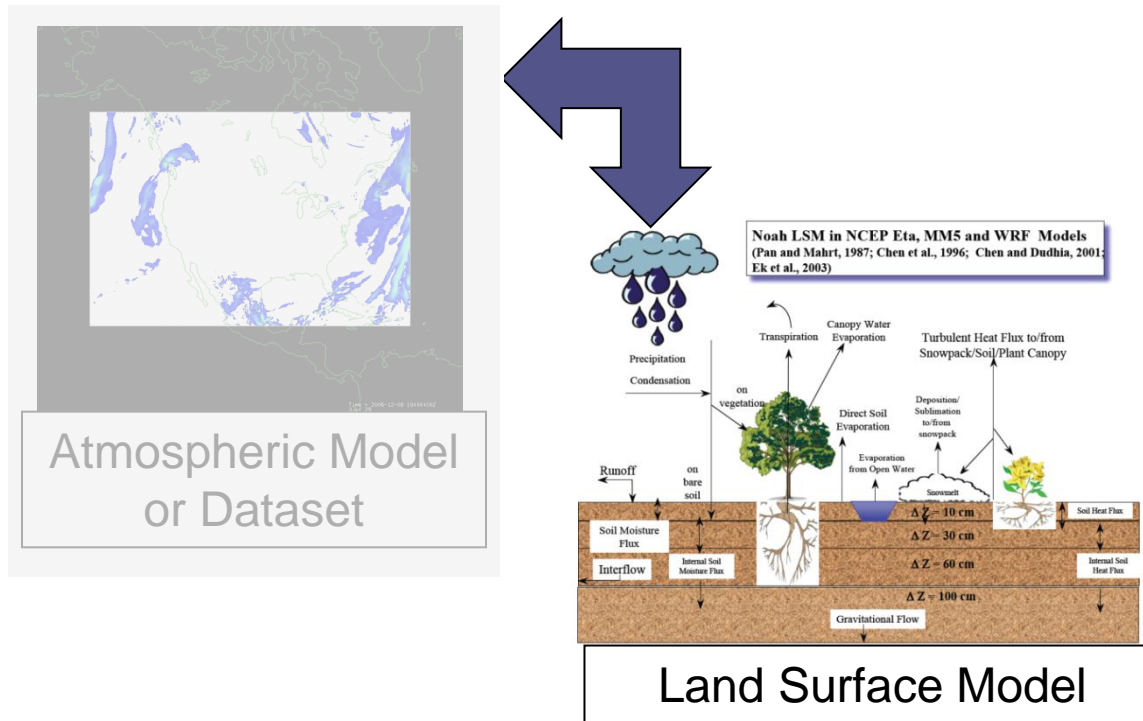
Before downscaling



RMSE in RCM

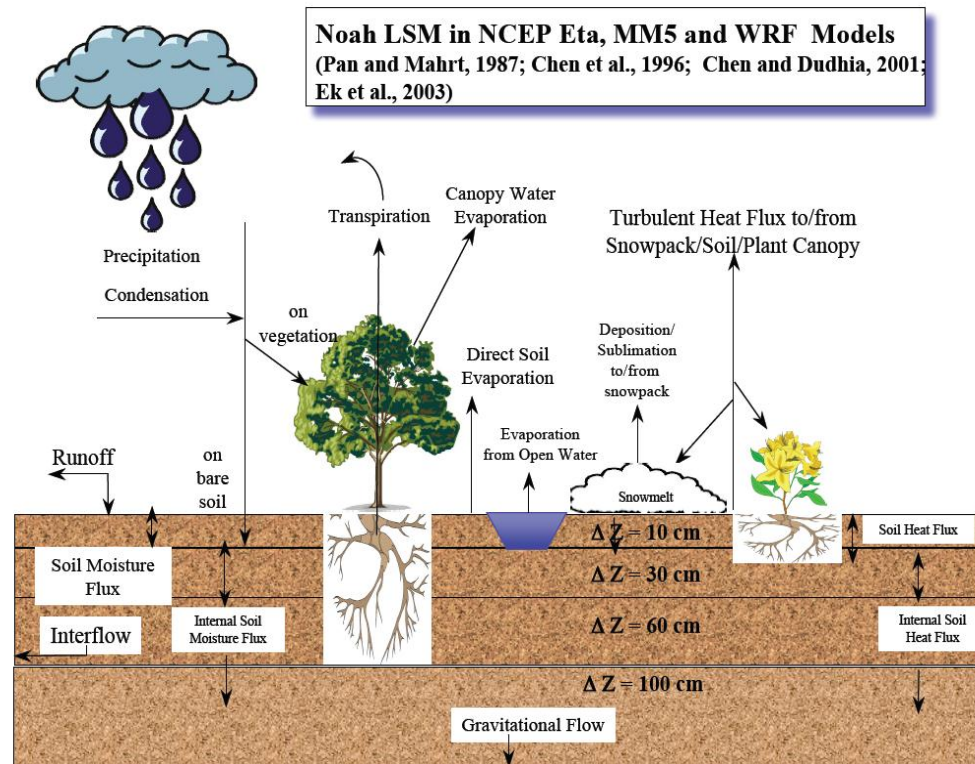
After downscaling

Framework for calculation



Modeling across spatial and temporal scales:
Global \rightarrow Regional \rightarrow Watershed

Noah land surface model with multi-physics options



Noah LSM with multi-physics options

1. Leaf area index (prescribed; predicted)
2. Turbulent transfer (Noah; NCAR LSM)
3. Soil moisture stress factor for transpiration (Noah; BATS; CLM)
4. Canopy stomatal resistance (Jarvis; Ball-Berry)
5. Snow surface albedo (BATS; CLASS)
6. Frozen soil permeability (Noah; Niu and Yang, 2006)
7. Supercooled liquid water (Noah; Niu and Yang, 2006)
8. Radiation transfer:
 - Modified two-stream:
 - Gap = F (3D structure; solar zenith angle; ...) \leq 1-GVF
 - Two-stream applied to the entire grid cell: Gap = 0
 - Two-stream applied to fractional vegetated area: Gap = 1-GVF
9. Partitioning of precipitation to snowfall and rainfall (CLM; Noah)
10. Runoff and groundwater:
 - TOPMODEL with groundwater
 - TOPMODEL with an equilibrium water table (Chen&Kumar,2001)
 - Original Noah scheme
 - BATS surface runoff and free drainage

More to be added

Niu et al. (2011)

Collaborators: Yang, Niu (UT), Chen (NCAR), Ek/Mitchell (NCEP/NOAA), and others

Maximum Number of Combinations

1. Leaf area index (prescribed; predicted) 2
2. Turbulent transfer (Noah; NCAR LSM) 2
3. Soil moisture stress factor for transpiration (Noah; BATS; CLM) 4
4. Canopy stomatal resistance (Jarvis; Ball-Berry) 2
5. Snow surface albedo (BATS; CLASS) 2
6. Frozen soil permeability (Noah; Niu and Yang, 2006) 2
7. Supercooled liquid water (Noah; Niu and Yang, 2006) 2
8. Radiation transfer: 3

Modified two-stream:

Gap = F (3D structure; solar zenith angle; ...) \leq 1-GVF

Two-stream applied to the entire grid cell: Gap = 0

Two-stream applied to fractional vegetated area: Gap = 1-GVF

9. Partitioning of precipitation to snowfall and rainfall (CLM; Noah) 2
10. Runoff and groundwater: 4

TOPMODEL with groundwater

TOPMODEL with an equilibrium water table (Chen&Kumar,2001)

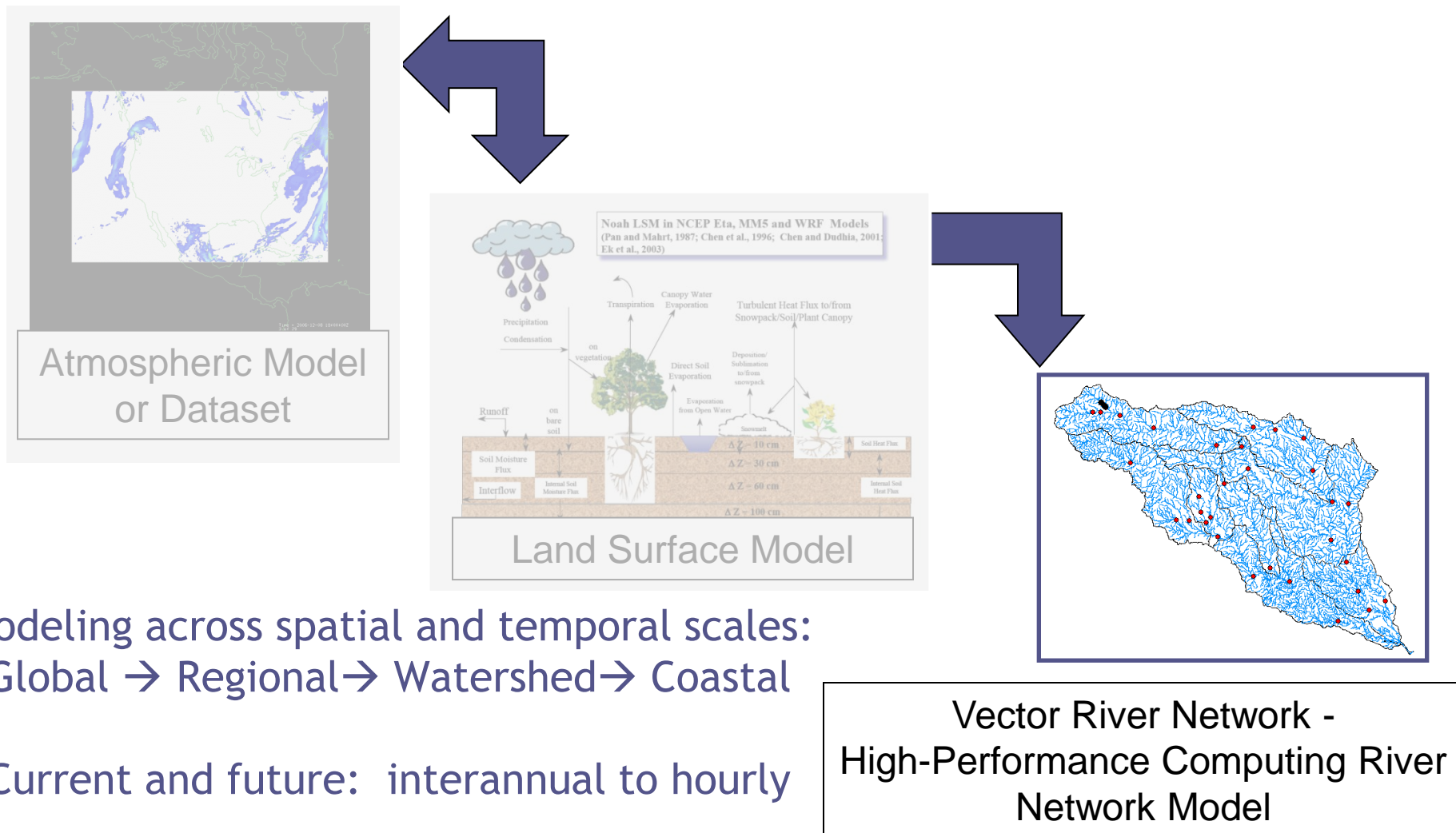
Original Noah scheme

BATS surface runoff and free drainage

$2 \times 2 \times 3 \times 2 \times 2 \times 2 \times 2 \times 3 \times 2 \times 4 = 4608$ combinations

Process understanding, probabilistic forecasting, quantifying uncertainties

Framework for calculation

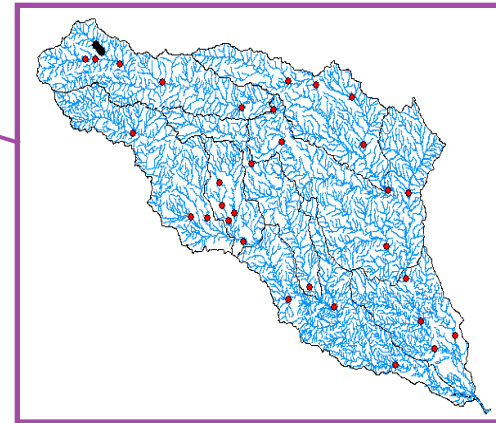
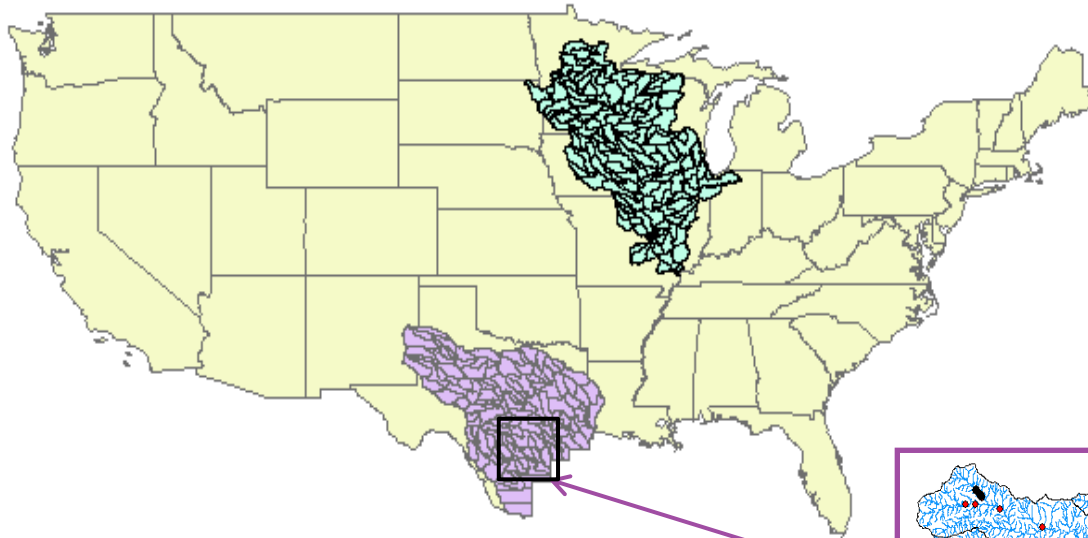


Modeling across spatial and temporal scales:
Global → Regional → Watershed → Coastal

Current and future: interannual to hourly

Vector River Network -
High-Performance Computing River
Network Model

River network modeling



RAPID

- Uses mapped rivers
- Uses high-performance parallel computing
- Computes everywhere including ungaged locations

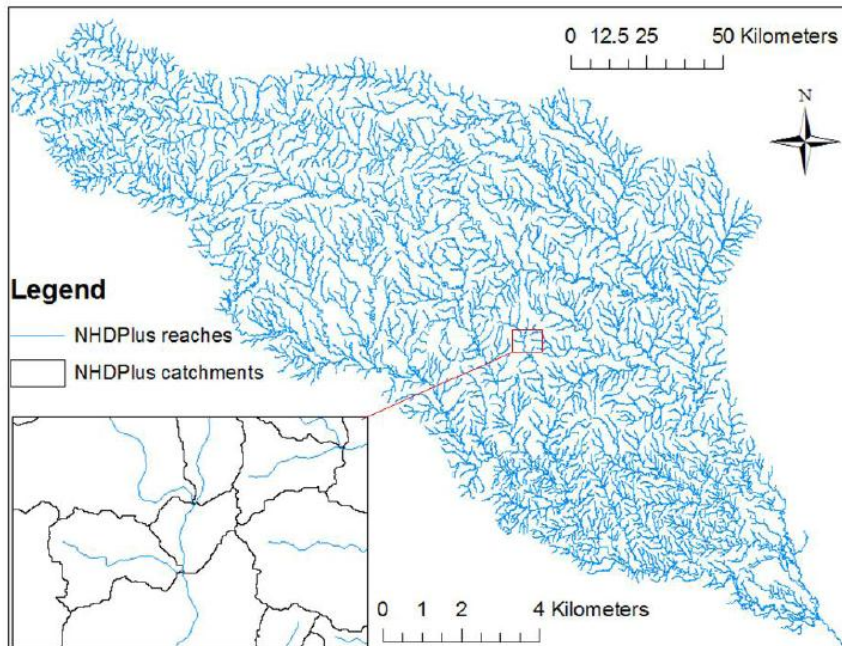
River Network Model: RAPID

Routing Application for Parallel computation of Discharge

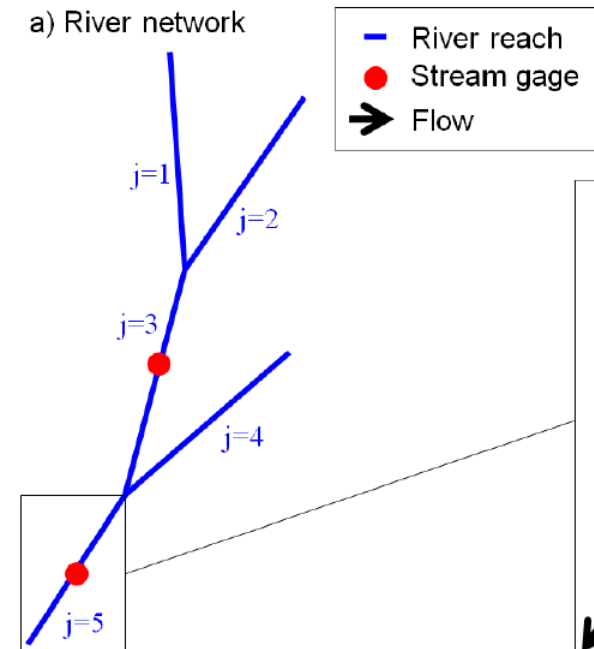
Based on Muskingum
Method

$$(I - C_1 \cdot N) \cdot Q(t + \Delta t) = C_1 \cdot Q^e(t) + C_2 \cdot [N \cdot Q(t) + Q^e(t)] + C_3 \cdot Q(t)$$

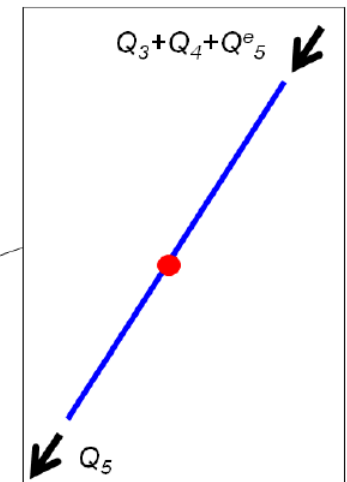
$$V(t + \Delta t) = V(t) + [N \cdot Q(t) + Q^e(t)] \cdot \Delta t - Q(t) \cdot \Delta t$$



a) River network

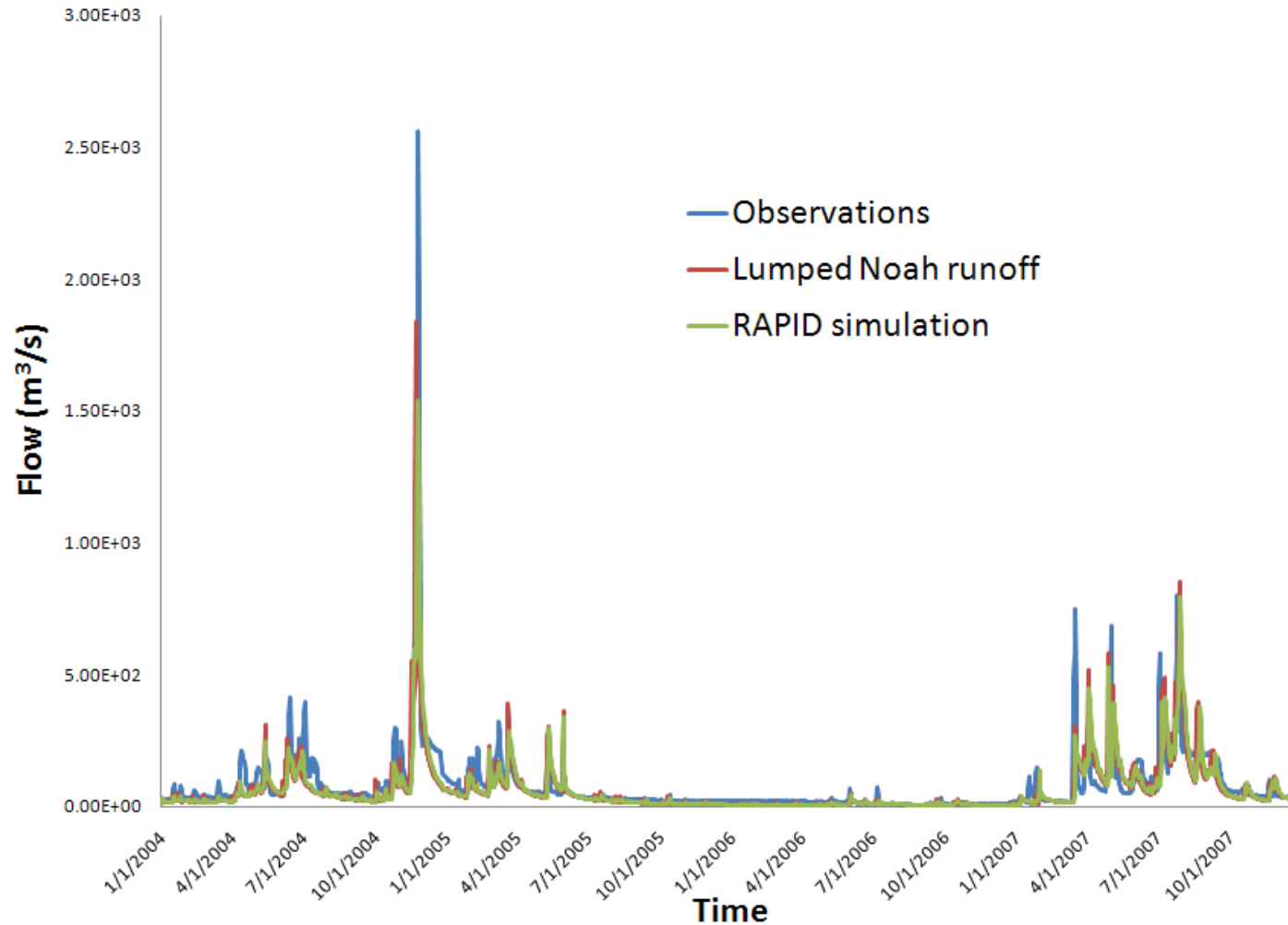


b) Single reach



RAPID and Noah-MP Performance Results

Guadalupe River nr Victoria



Texas Rivers Draining to the Gulf of Mexico

- 01/01/2004 – 12/31/2007 every 3 hours
 - 4-km grid
 - NARR meteorological forcing + NEXRAD rainfall
 - Noah-MP runoff → RAPID routing
-
- facilitate modeling of nutrient loading, transport, and export to coastal waters

<http://www.geo.utexas.edu/scientist/david/rapid.htm>

Thanks to Cedric David, Bryan Hong, David Maidment, Ben Hodges, Ahmad Tavakoly, and Adam Kubach of Texas Advanced Computing Center

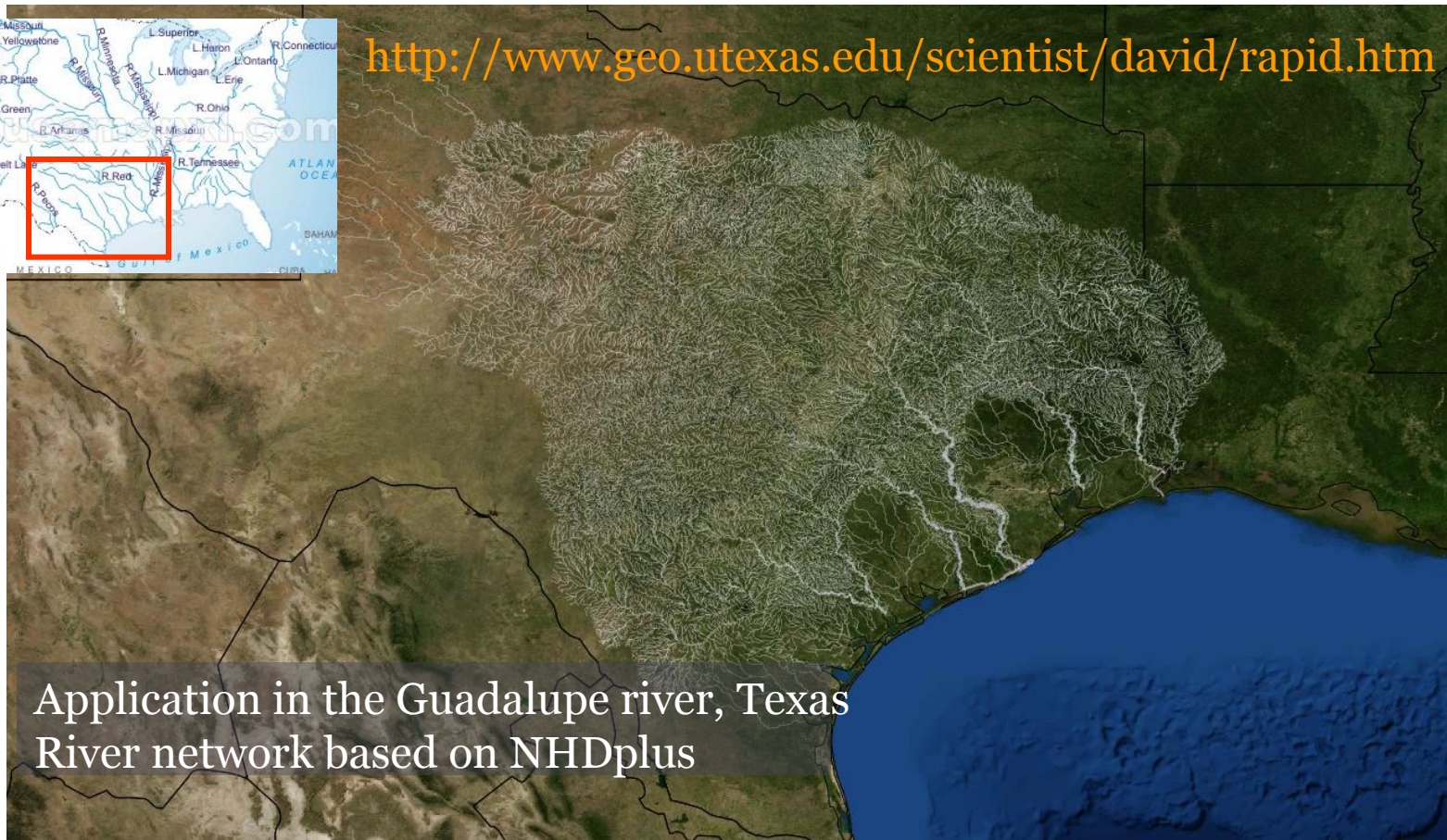
(David et al., 2011, HP, JHM)

RAPID Routing model

- adapted to large scale basin with high spatial resolution
- few parameters, inversion process included
- numerical efficiency (parallel computation)

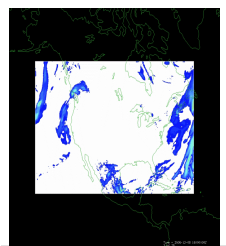


<http://www.geo.utexas.edu/scientist/david/rapid.htm>

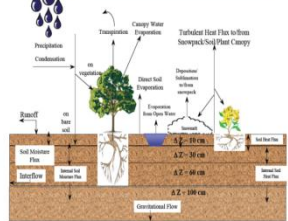
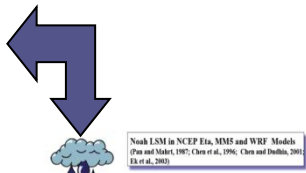


Application in the Guadalupe river, Texas
River network based on NHDplus

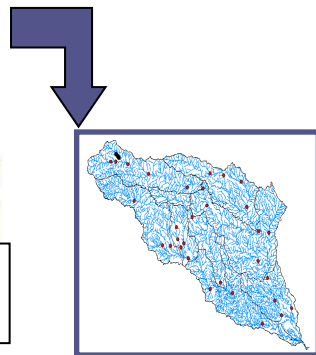
Framework; what is missing?



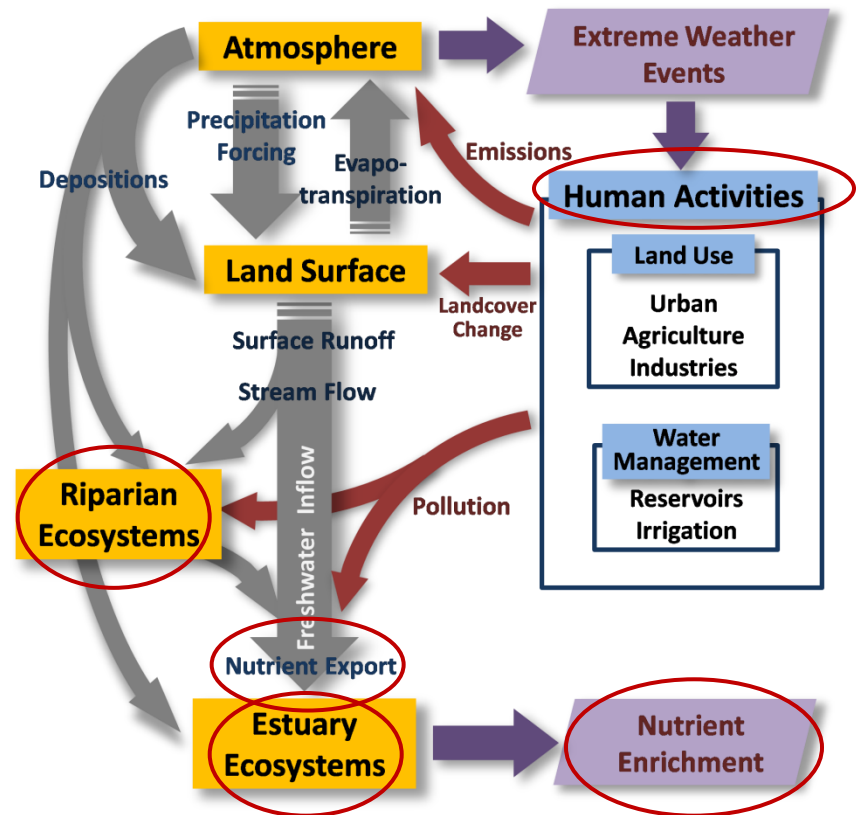
Atmospheric Model or Dataset



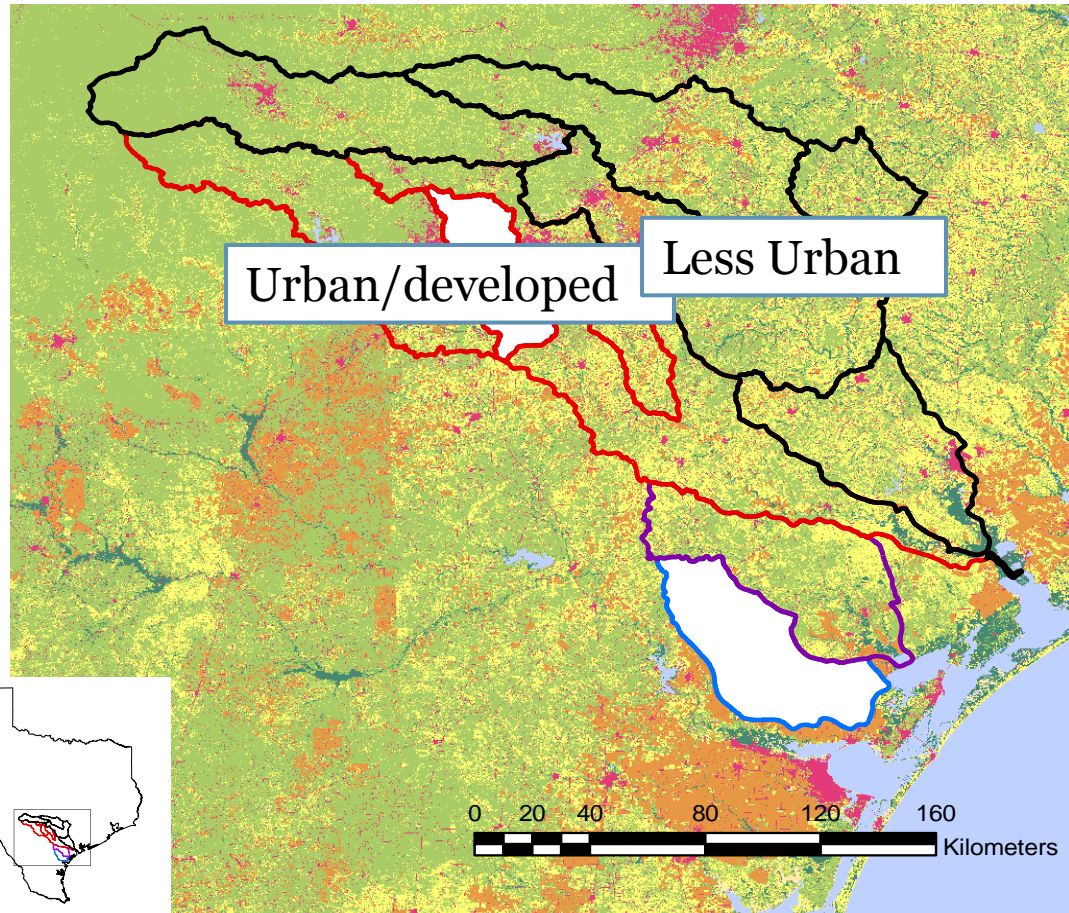
Land Surface Model



Vector River Network - High-Performance Computing River Network Model



San Antonio, Guadalupe, Mission, and Aransas Rivers



Legend

- Guadalupe Less Urban/developed
- San Antonio Urban/developed
- Mission
- Aransas

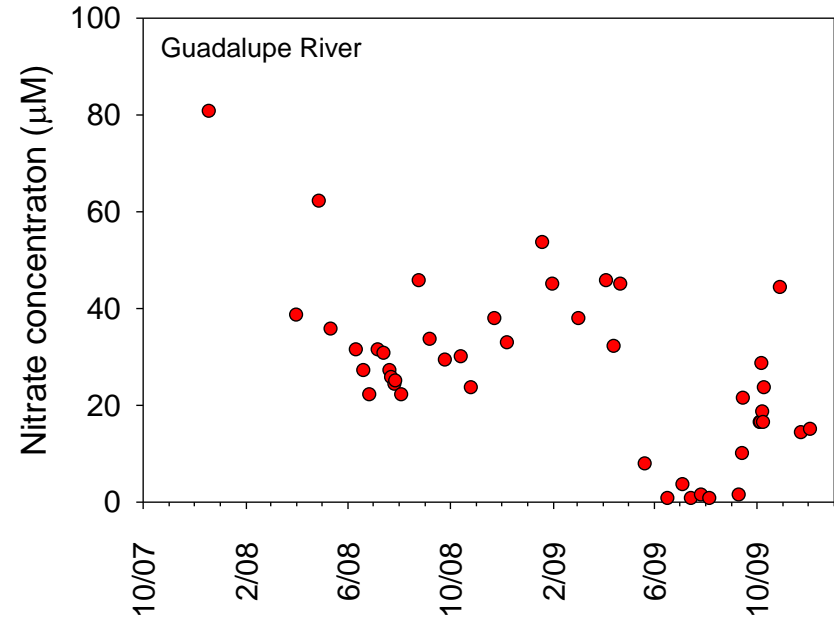
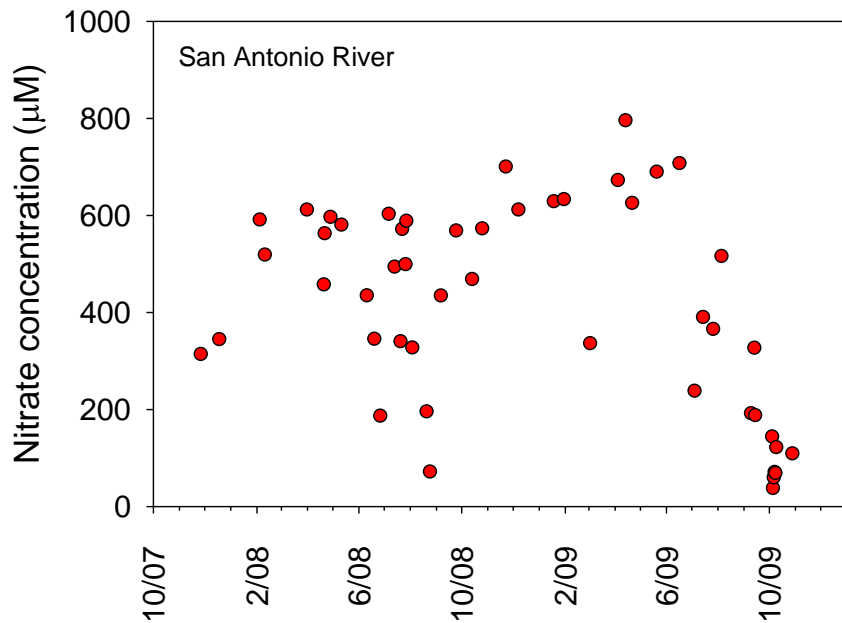
Land Use/Land Cover

- Open Water
- Developed
- Rock/Sand/Clay
- Forest/Shrub
- Grassland/Pasture
- Cultivated Crops
- Wetland

Observations and Chemistry Sampling

- Sampling targeted to high flow events
 - potential for high nutrient export
- Stream Gauge data:
 - Taken from Texas Commission of Environmental Quality (TCEQ)
 - Gauge data collected at constant time step
 - Taken from University of Texas Marine Science Institute (UTMSI)
 - Gauge data collected during high flow events

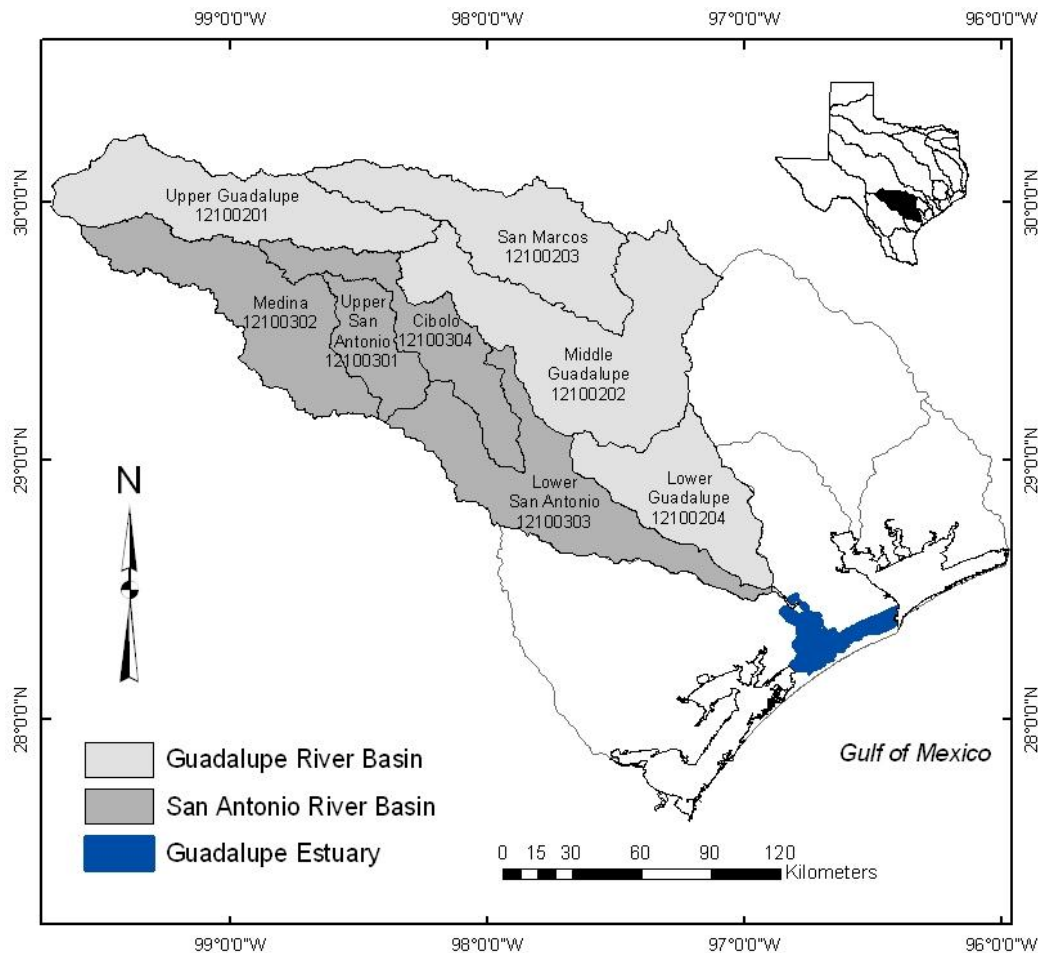
Stream Gauge Nitrate Concentration: Urban vs. Less Urban



• Urban/Developed Location

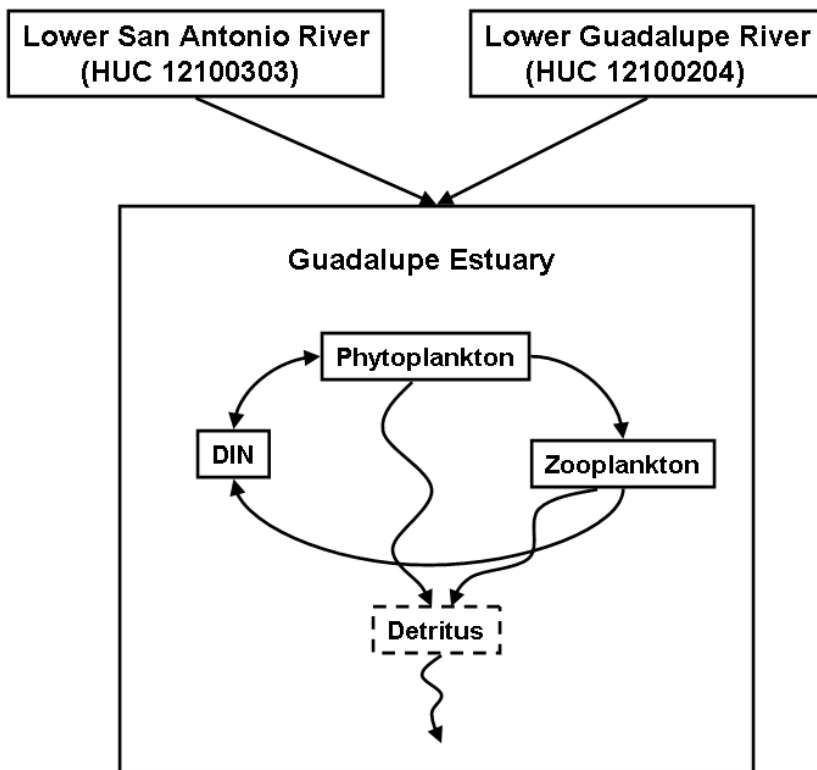
• Less Urban Location

Estuary Model Nutrient Transport Study



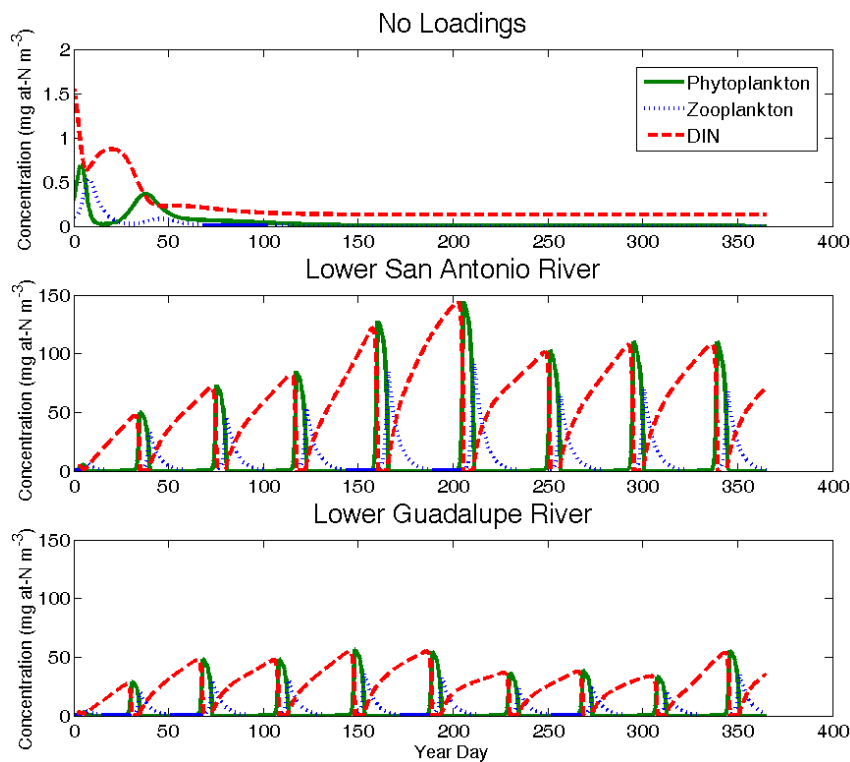
- Same River Basins
 - Guadalupe (Less Urban)
 - San Antonio (Urban/developed)
- Four HUCs in each basin
- Guadalupe Estuary
 - Centrally located along Texas coast
 - Microtidal
 - Small bay area but large watershed relative to other Texas systems

Generic Ecosystem Model (3 components with 2 boundary conditions)



- Mass-balance model
- Two boundaries: LGRW & LSRW
- Three components: Nutrient (DIN) – Phytoplankton – Zooplankton
- Re-mineralization and implicit sinking (or horizontal exchange) were assumed to be 50%, respectively
- $\Delta=1$ hr & RK 4th order scheme

Generic Ecosystem Model Results



- No Loadings (both boundary conditions shut down)
- Lower San Antonio River (Urban/developed region)
- Lower Guadalupe River (Less Urban Region)

Source: Arismendez et al. (2009) Ecol. Informatics 4: 243-253

Generic Ecosystem Model Conclusions and Discussion

- Estuary response differs with respect to varying nutrient concentrations.
- Lower San Antonio River (Urban/developed region) is delivering more nutrients and driving greater ranges of ecological response than the Lower Guadalupe River (Less Urban region).
- Increases in nutrient concentrations due to human alterations of the landscape may result in future eutrophic conditions in the Guadalupe Estuary.

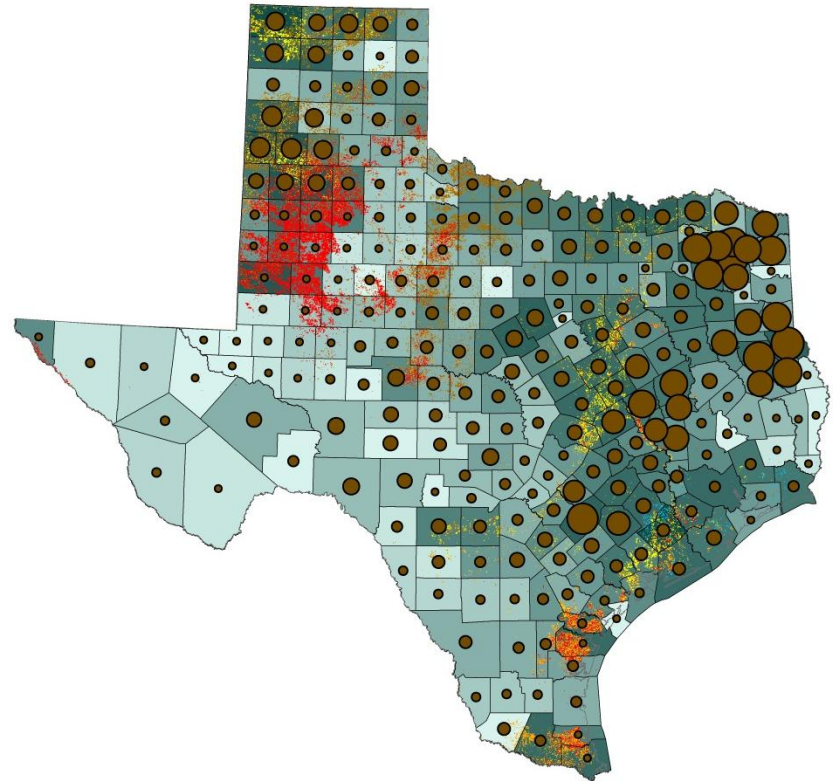
Improving on Nutrient Loading

- Developing a Comprehensive Nitrogen Budget for Texas
 - Agriculture Sources
 - Crop fixation, Livestock, and Fertilizer application
 - Atmospheric
 - Dry and wet deposition

Quantification of Sources

Legend

All_animal_density07

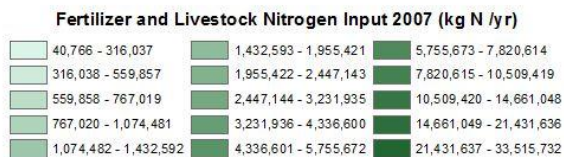
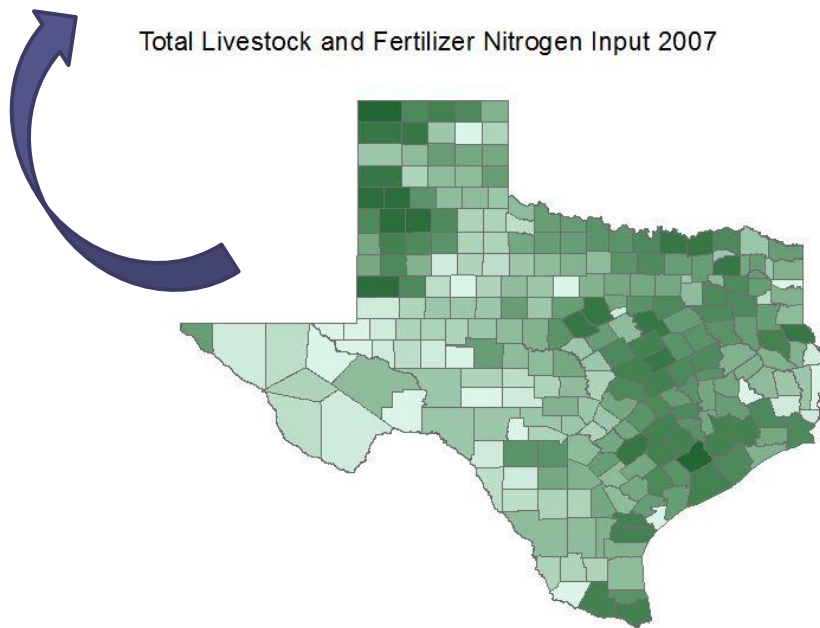


Conclusions

- Predicting nutrient transport from land to coast requires an integrated approach
- Improvement of atmosphere, land, and river flow modeling has lead to better prediction of nutrient fluxes
- Understanding the full pathways of nutrients, with enhanced modeling techniques, will lead to better understanding of sources and solutions

Future Work

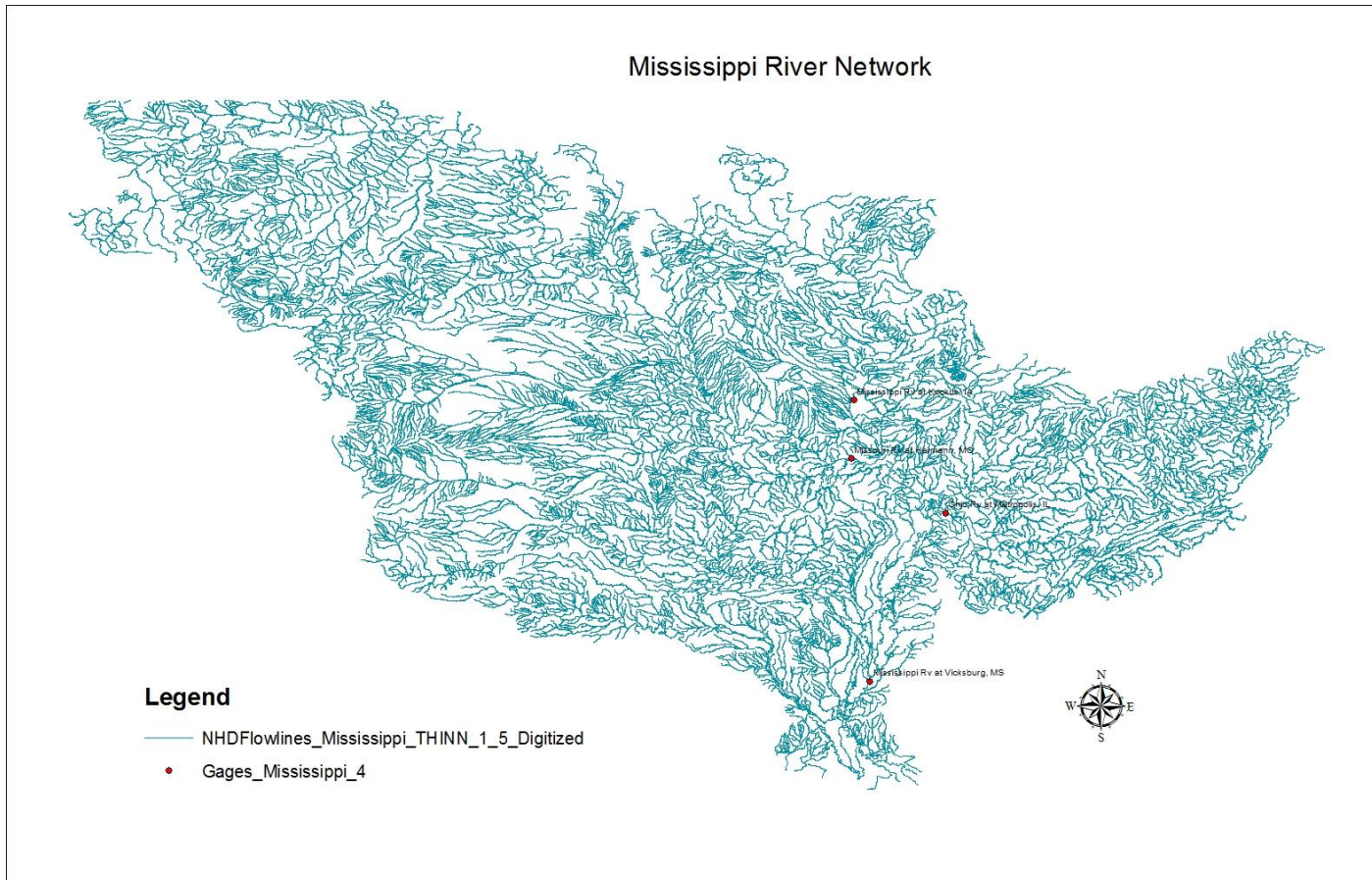
- Land Surface model with leaching (Noah-MP), coupled with regional weather model (WRF)



RAPID

Future Work

- Expansion beyond the Texas Regional Domain



Thank You

- **Zong-Liang Yang** (PI)
- David Maidment (co-I), Paul Montagna (co-PI), James McClelland (co-I), Hongjie Xie (co-PI), Wei Min Hao (Co-PI)
- Guo-Yue Niu, Xiaoyan Jiang, Seungbum Hong, Cédric David, Hae-Cheol Kim, Sandra Arismendez, Rae Mooney, Patty Garlough, Rachel Mills, Beibei Yu, Ling Lu, Almoutaz El Hassan, Zhongfeng Xu, Xitian Cai, Ahmad Tavakoly
- Funded by NASA Interdisciplinary Research in Earth Sciences (IDS)



DON: concentration-runoff relationships

