

Intercomparison of Hypoxia Models for the Northern Gulf of Mexico and Nutrient Load Scenarios

PIs

Katja Fennel (lead, Dal)

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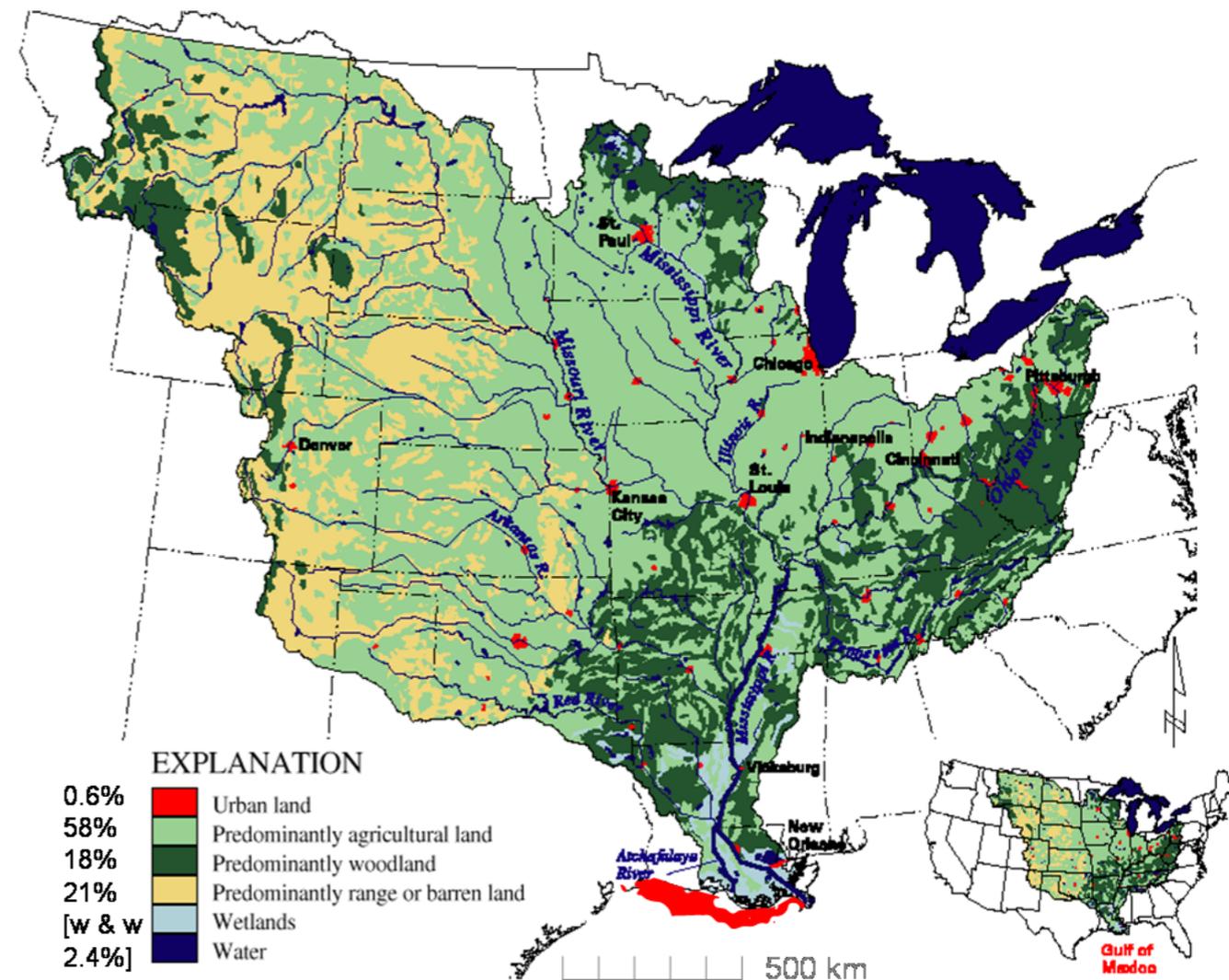
Partners

Frank Aikman (CSDL)

Jiangtao Xu (CSDL)

John Lehrter (EPA)

Mike Murrell (EPA)



Applications

Federal Entities/Partners

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- **NOAA - Ecological Roadmap**
- **Hypoxia Task Force** (states, federal agencies incl. NOAA, EPA, and tribes) - **Action Plan**
- **EPA - Clean Water Act**

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- **short-term hypoxia forecasts**
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Skillful application need good process understanding. Black box is likely not providing useful predictive capabilities with known error statistics.

Practical considerations for operational use (computational cost, robustness, data needs). Different models will be appropriate for the different uses.

COMT SH models

- **ROMS small domain**
 - simple hypoxia model
 - full bgc model
- **ROMS large domain**
 - simple hypoxia model
 - full bgc model
 - quasi-operational physical model
- **FVCOM**
 - simple hypoxia model
 - WASP model
- **NCOM**
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 - GEMS model

research mode,
hindcasts only,
run at Dal



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short-term
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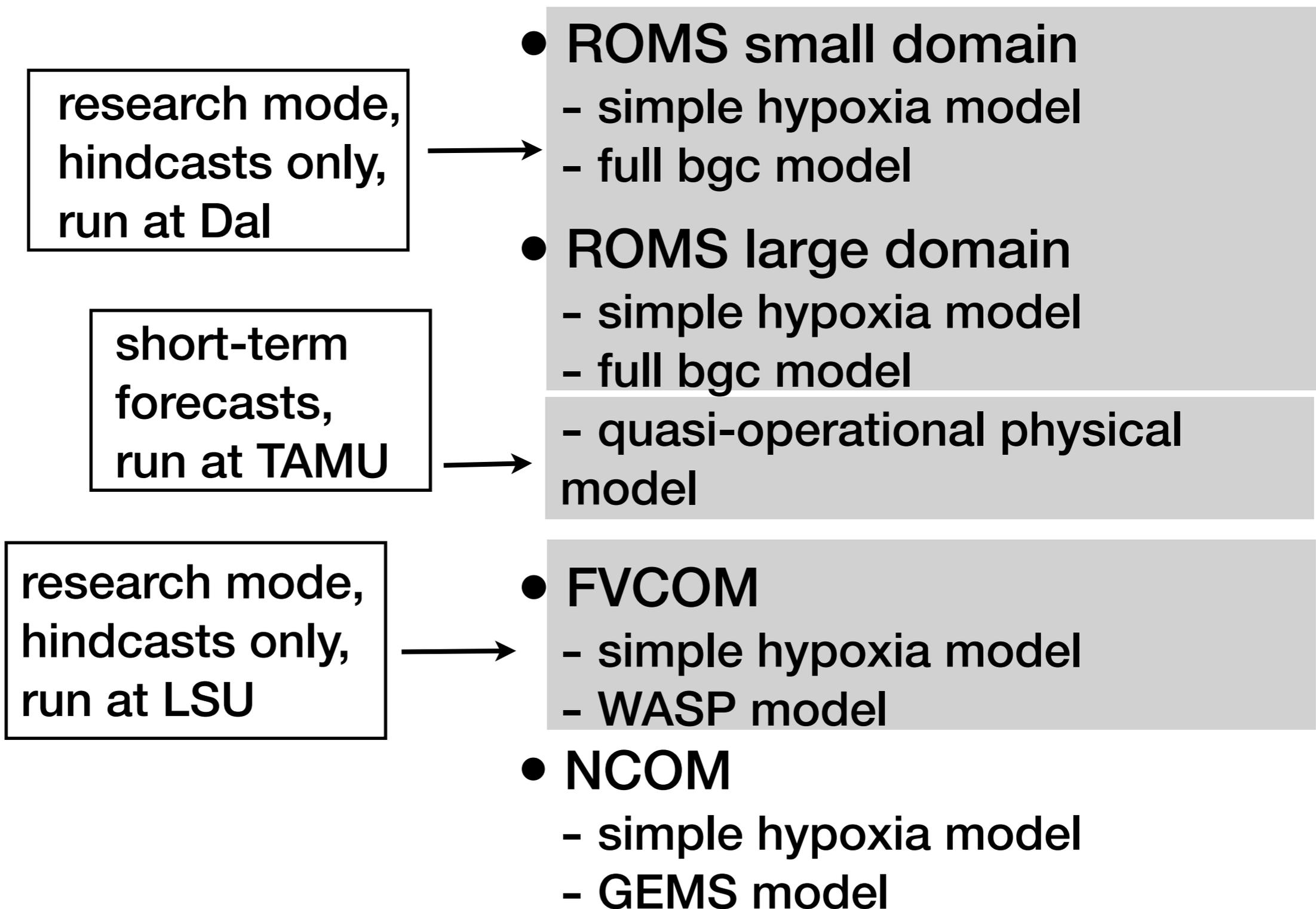
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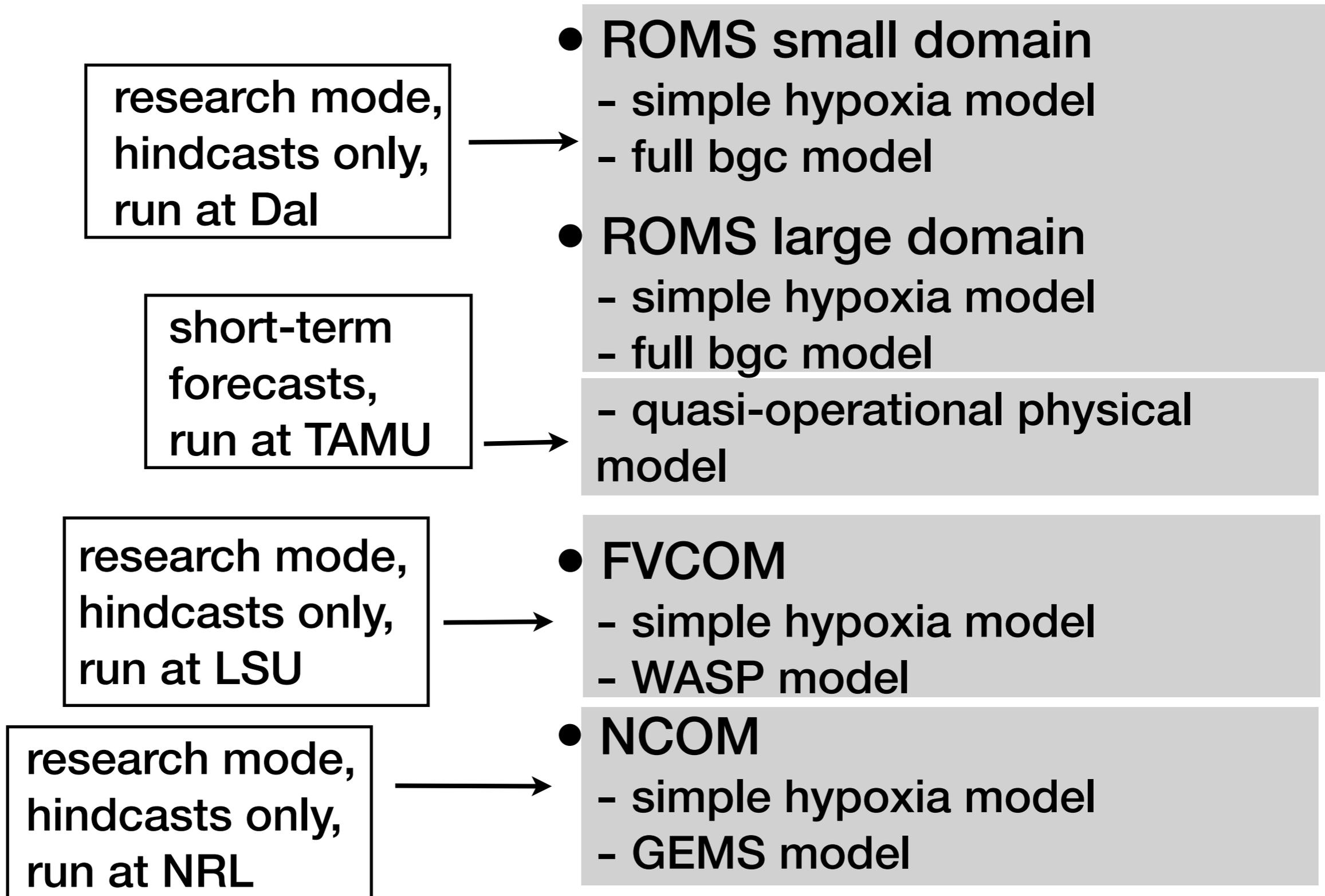
- FVCOM
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COMT SH models



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17

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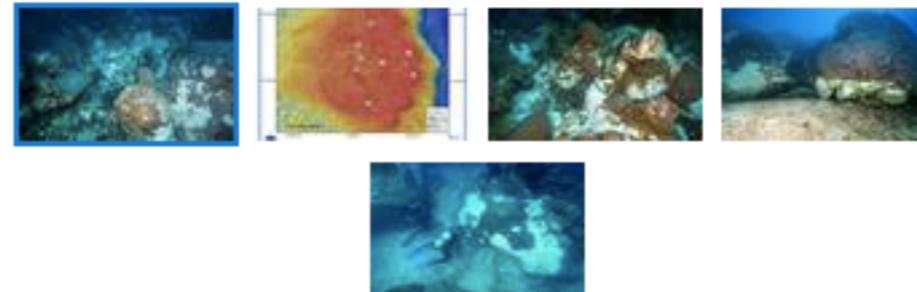
Posted by Eric Haun

August 12, 2016

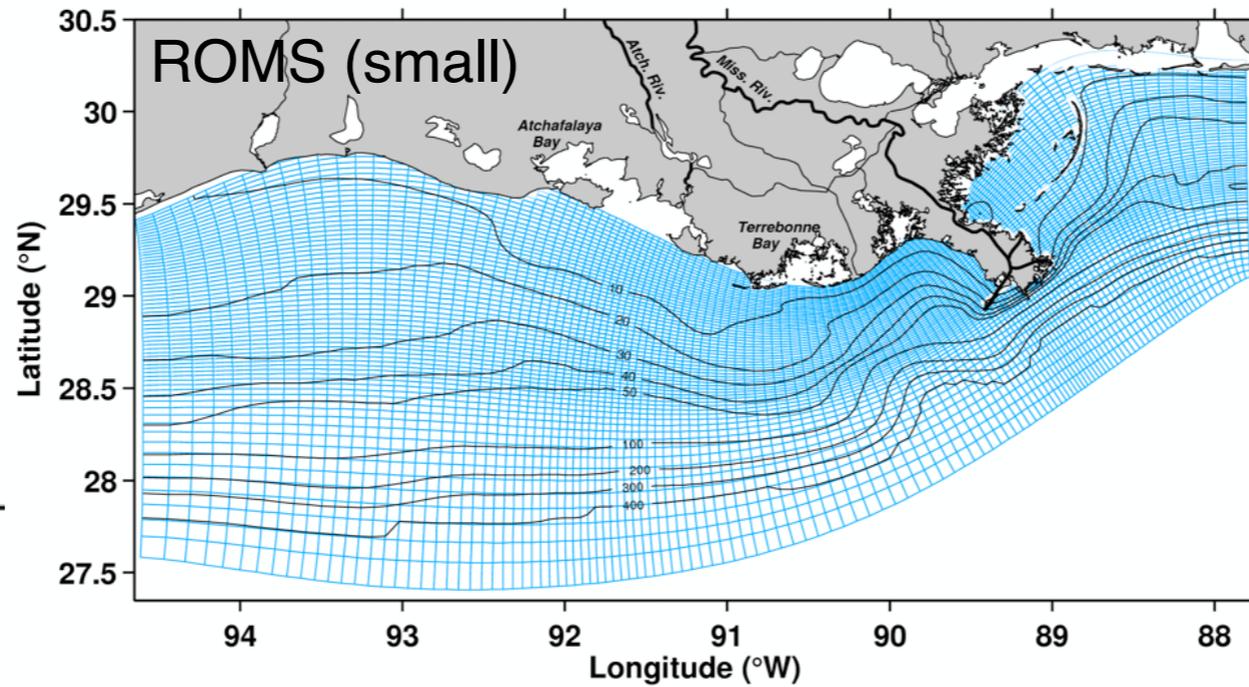
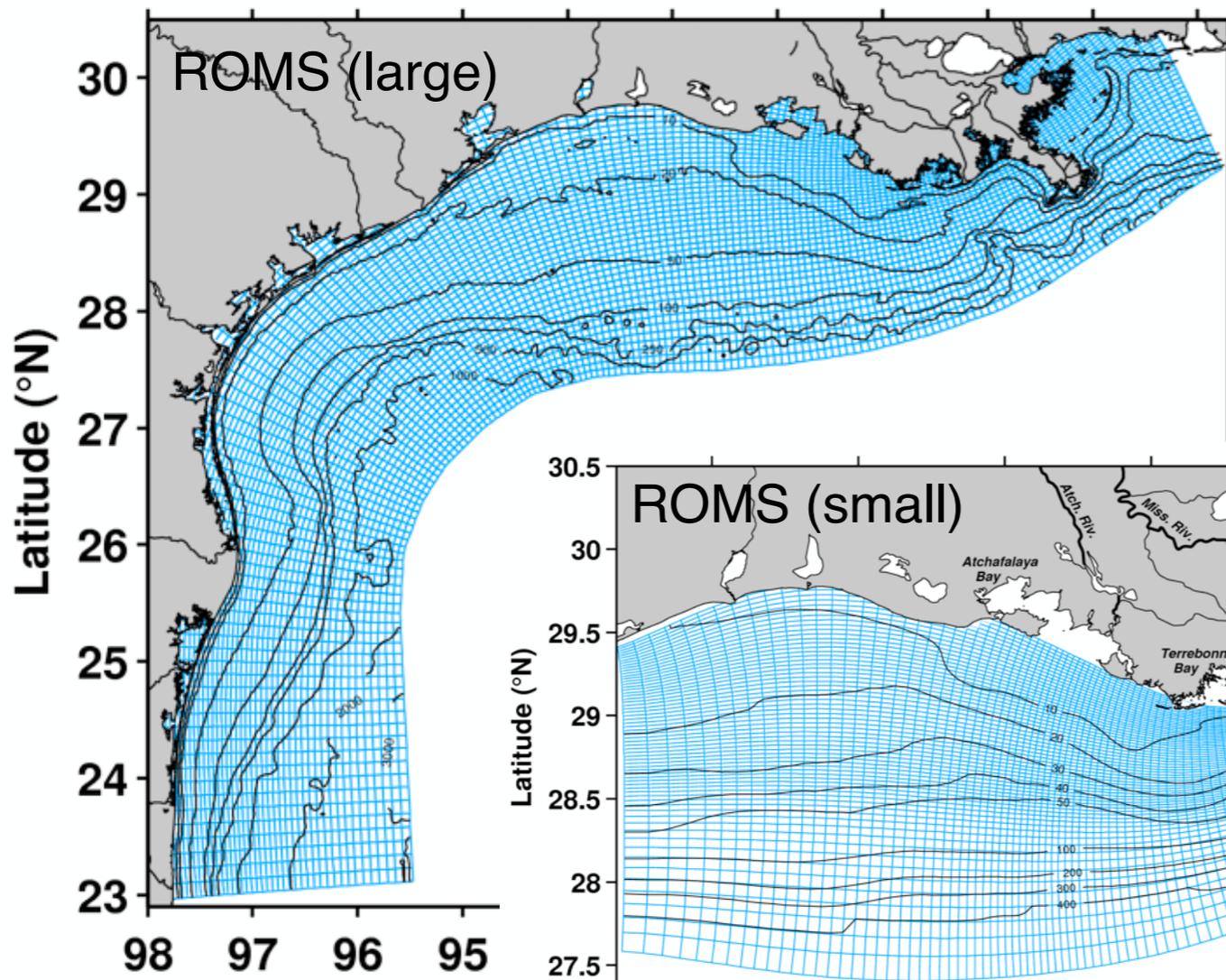
Why is Coral Dying at East Flower Garden Bank?

Normally the reefs of Flower Garden Banks National Marine Sanctuary (FGBNMS) are considered to be the healthiest in the region, but now scientists from around the world are trying to figure out what's behind a mysterious event in the area that's killed thousands of coral colonies and associated reef invertebrates.

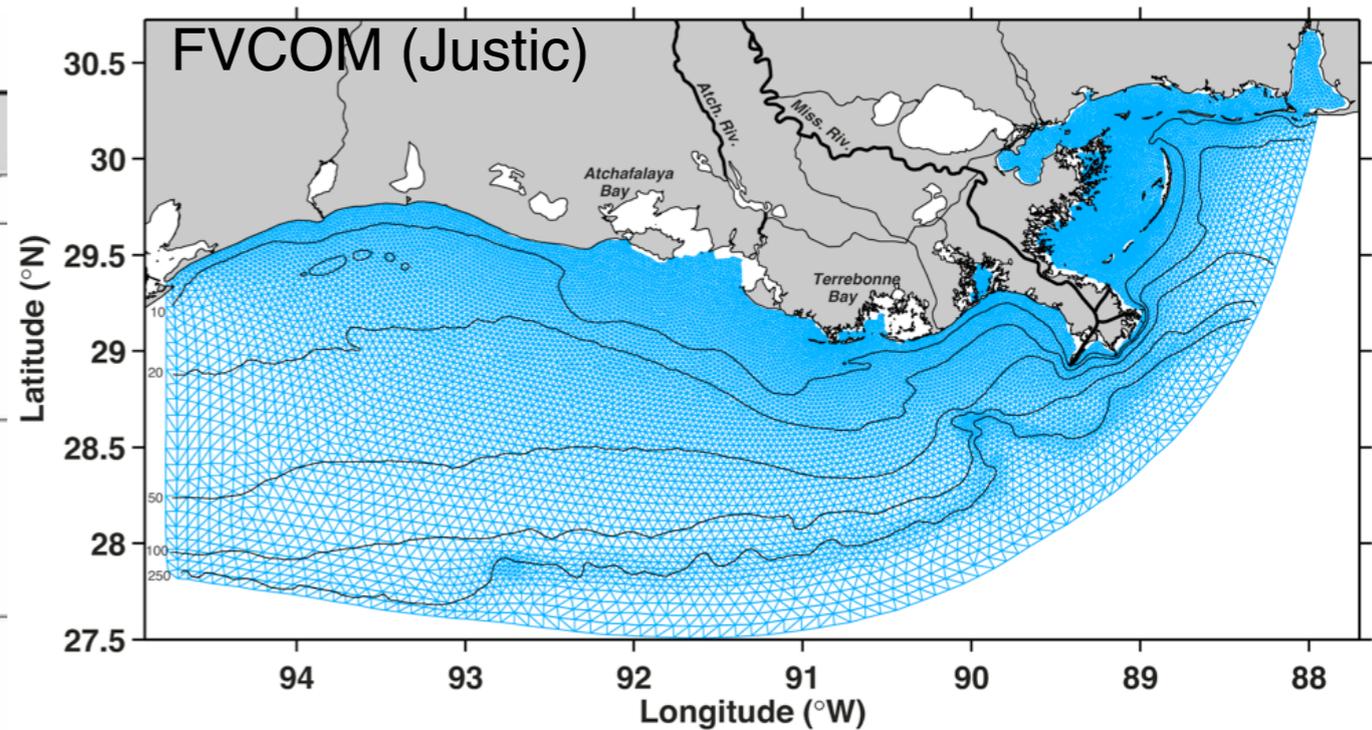
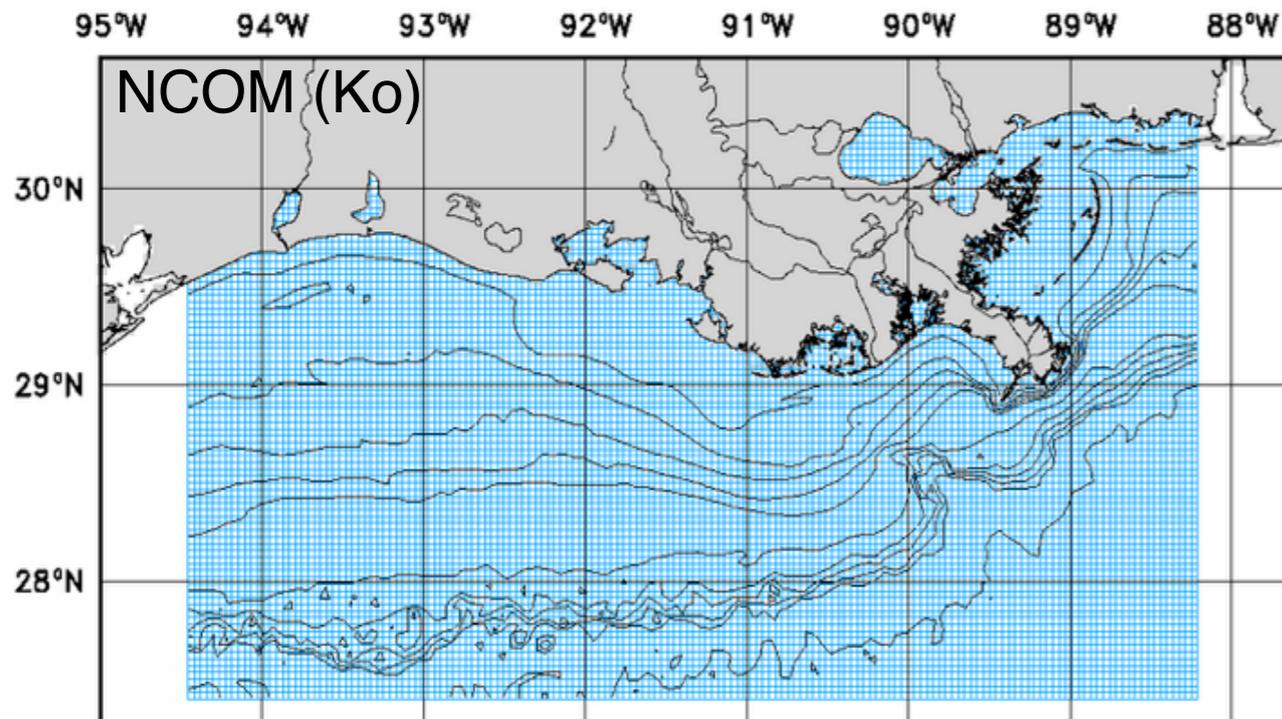
On July 25, sport divers on the M/V Fling reported green, hazy water, huge patches of ugly white mats on corals and sponges, and dead animals littering the bottom at East Flower Garden Bank, buoy #4. The charter captain notified FGBNMS and the U.S. Bureau of Ocean Energy Management (BOEM) researchers,

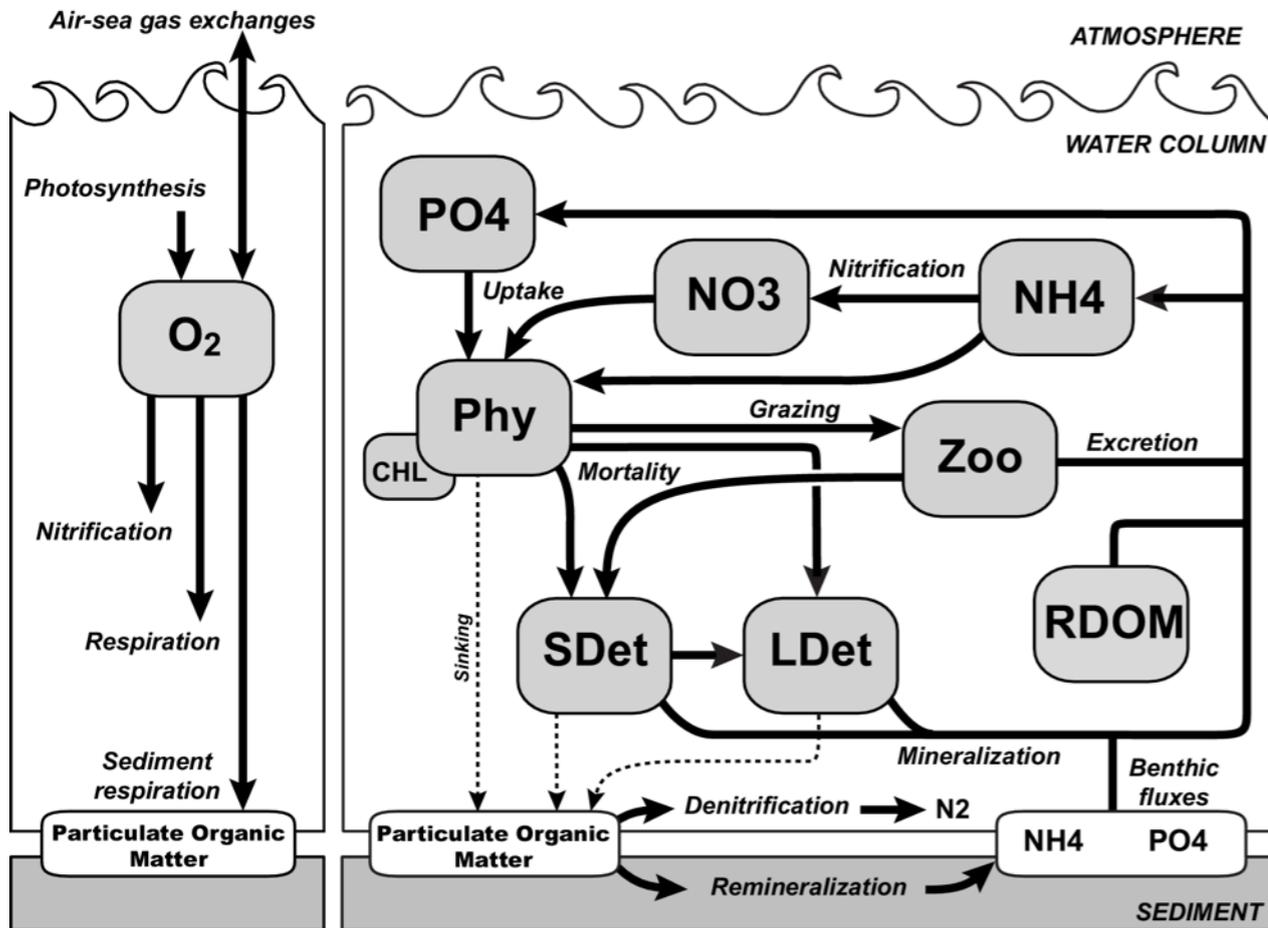
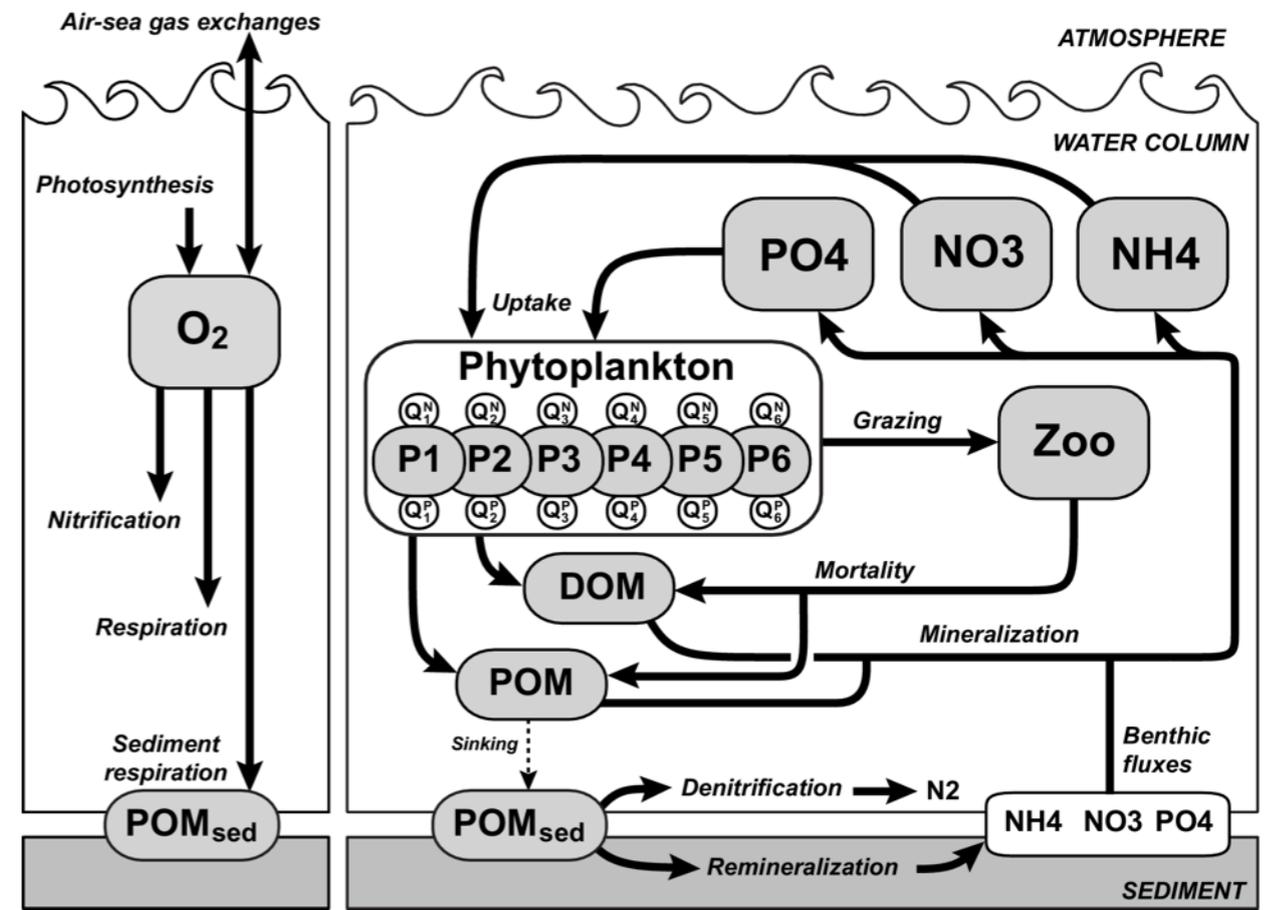
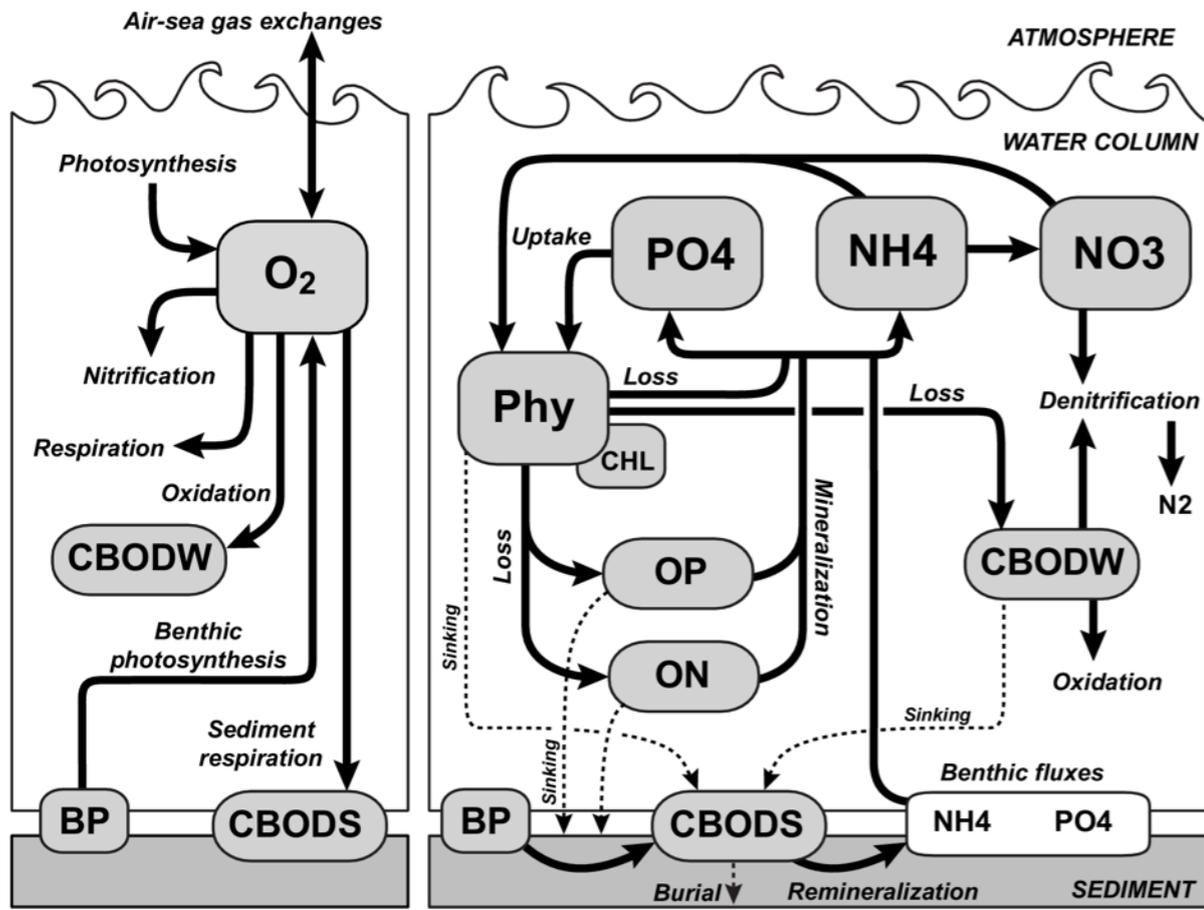






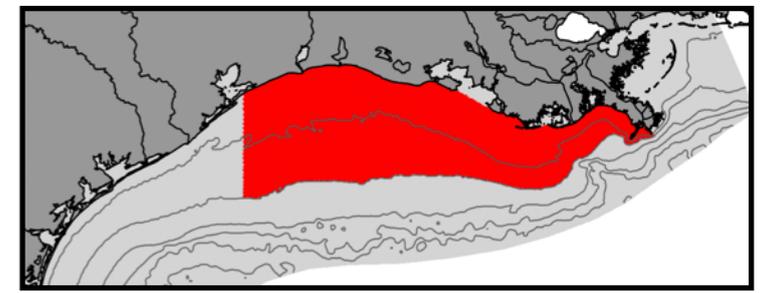
- ROMS (large and small domain)
- FVCOM
- NCOM



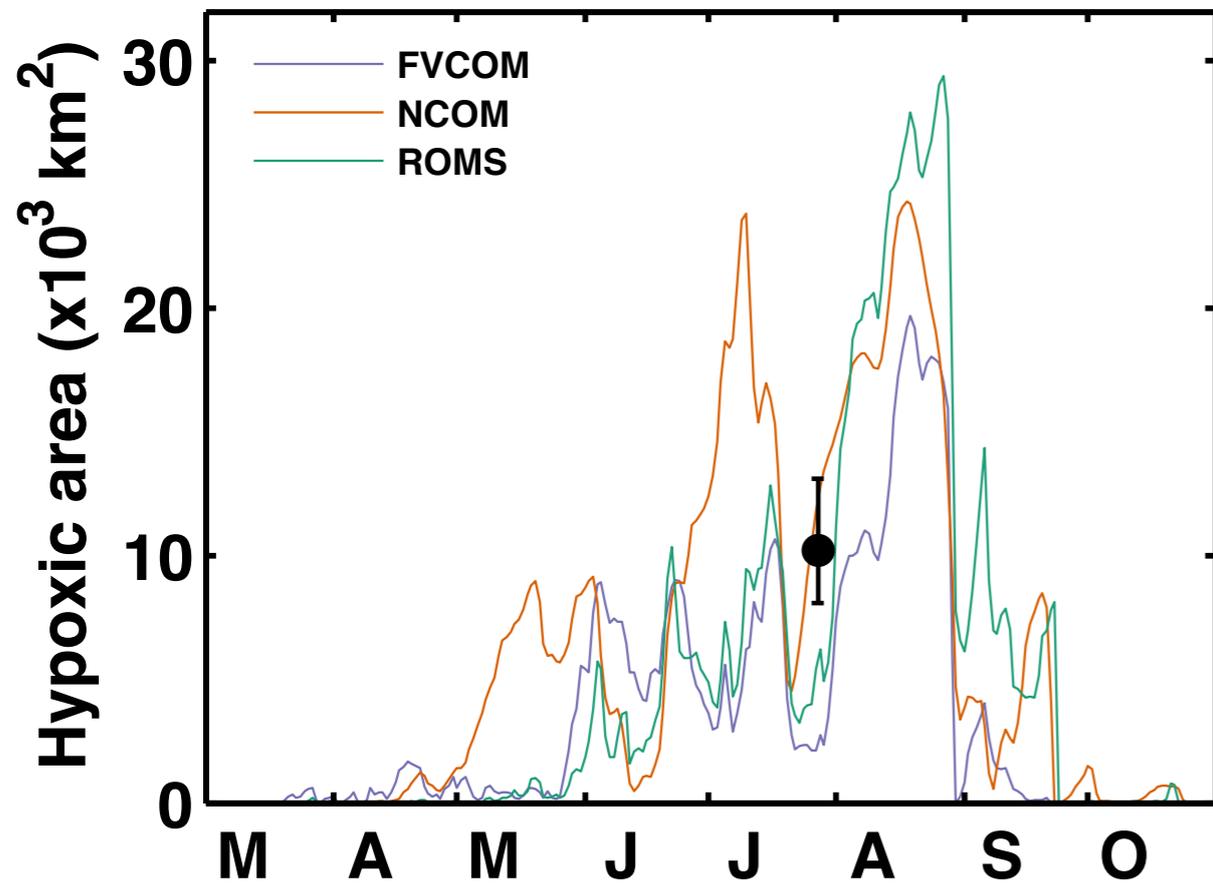


- **FVCOM (WASP)**
- **NCOM (GEM)**
- **ROMS**

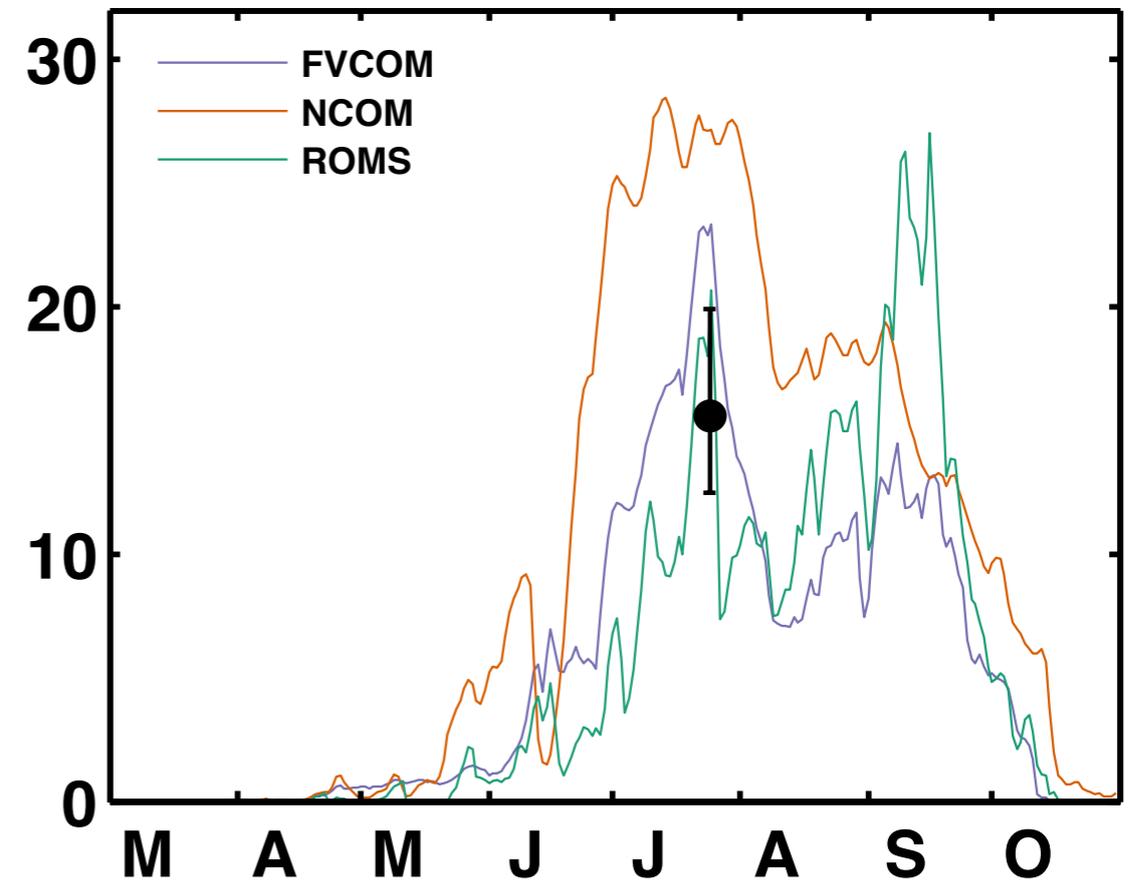
Hypoxic area: full biogeochemical models



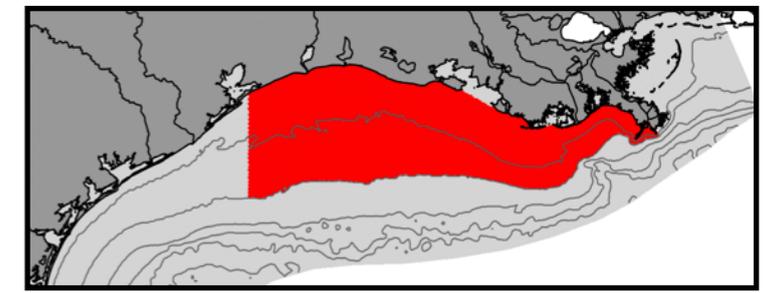
2005



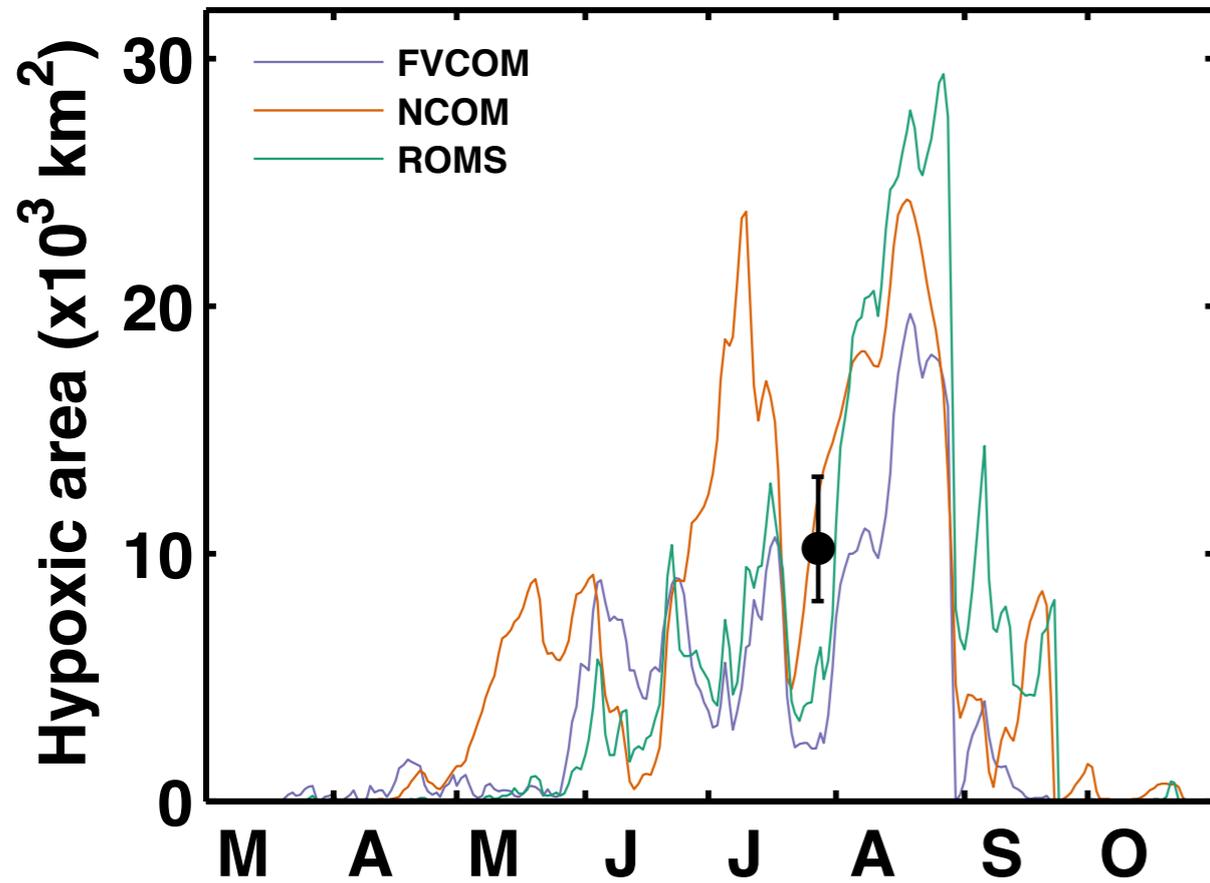
2006



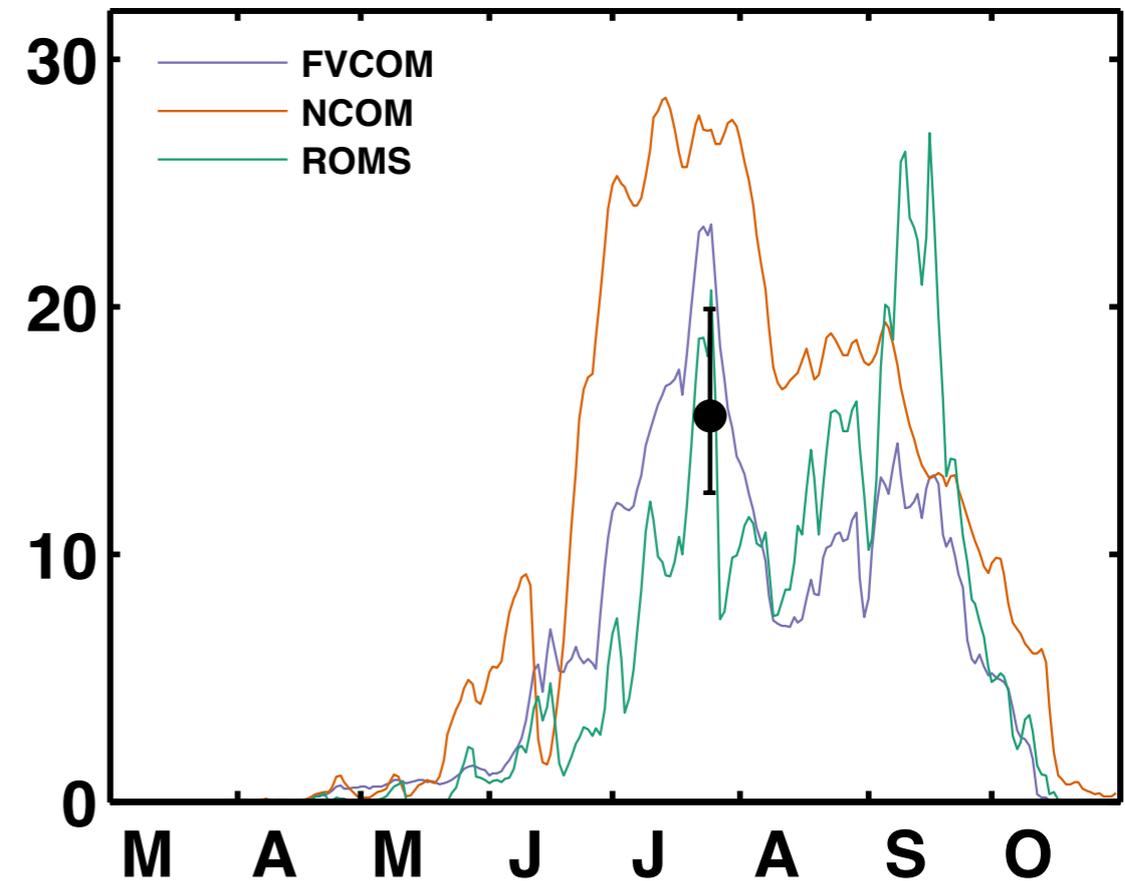
Hypoxic area: full biogeochemical models



2005



2006



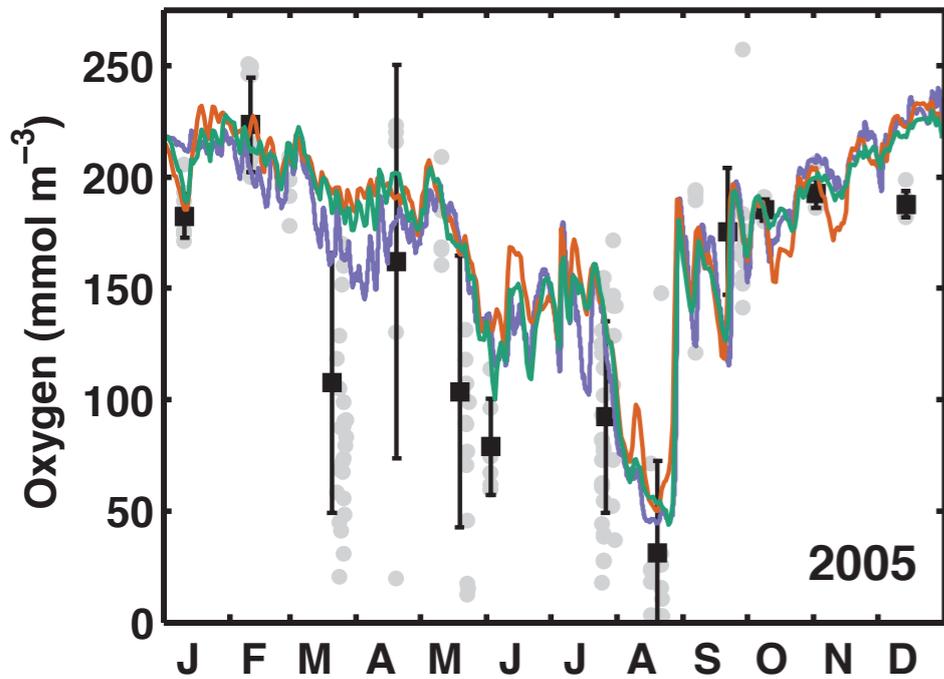
Differences in hypoxia predictions could be due to differences in model physics and/or biology.

Need to disentangle both effects.

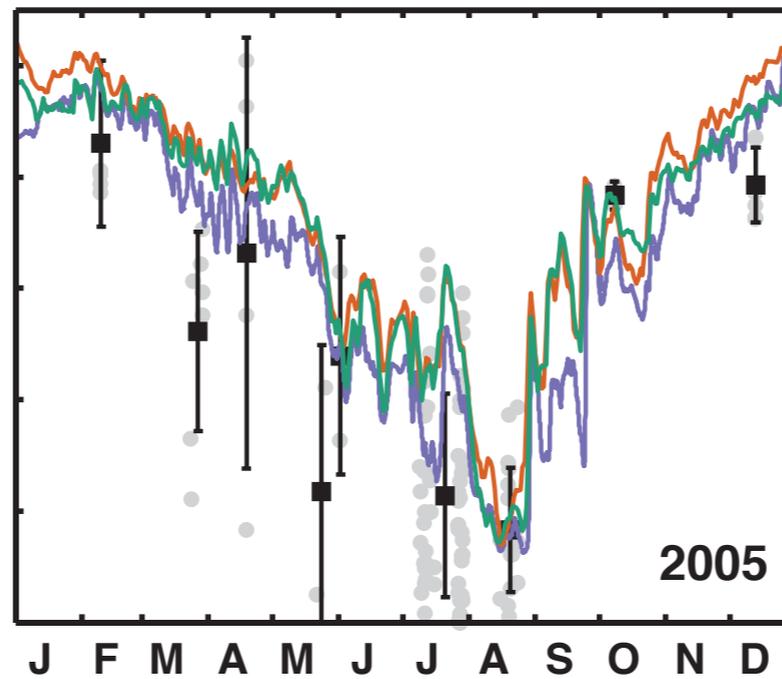
**Using simple oxygen parameterization by Yu et al. (JGR 2015)
(same in all models) includes air-sea gas exchange, water column
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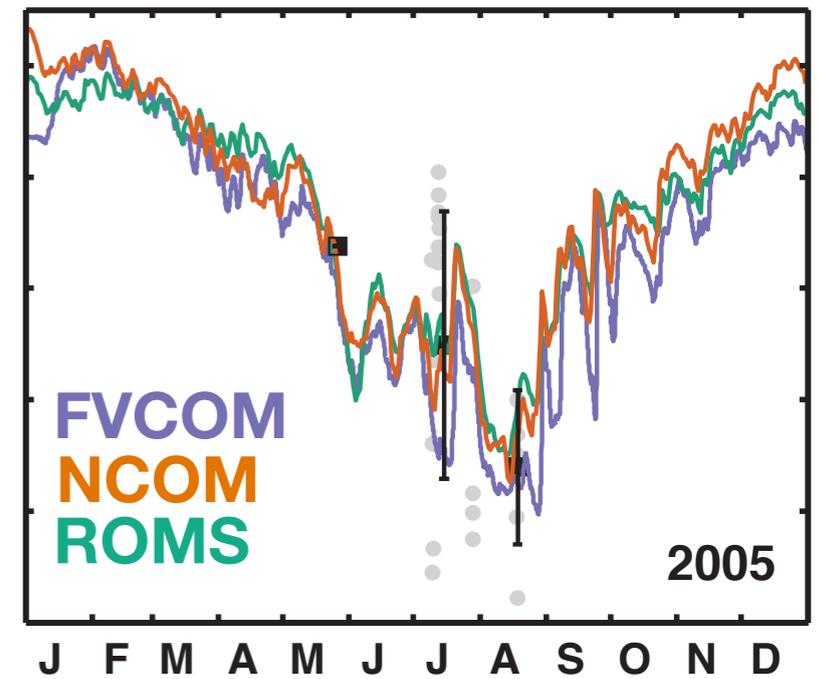
Mississippi (z<20m)



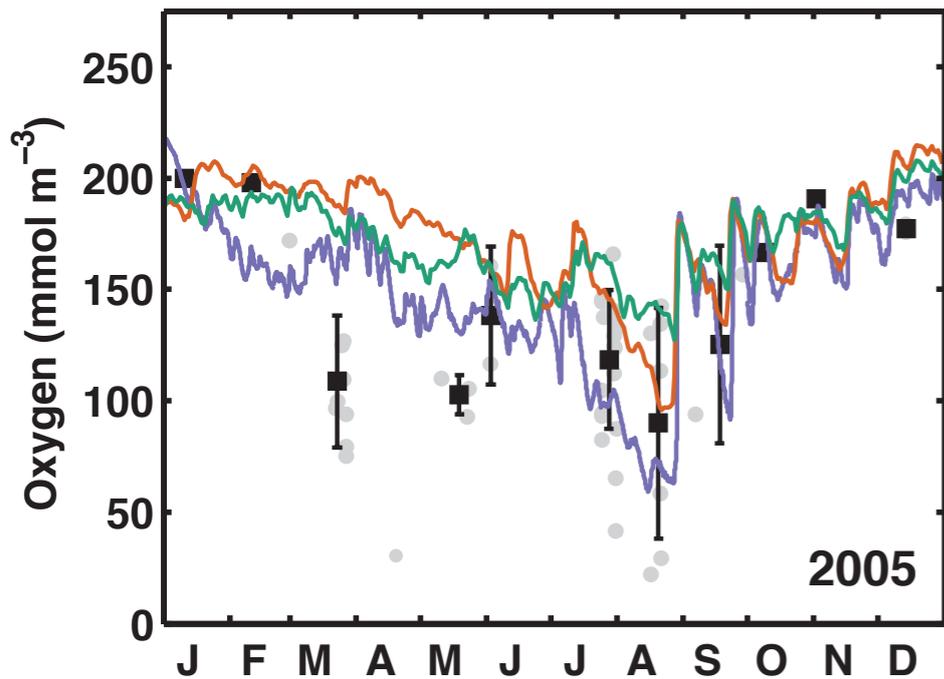
Atchafalaya (z<20m)



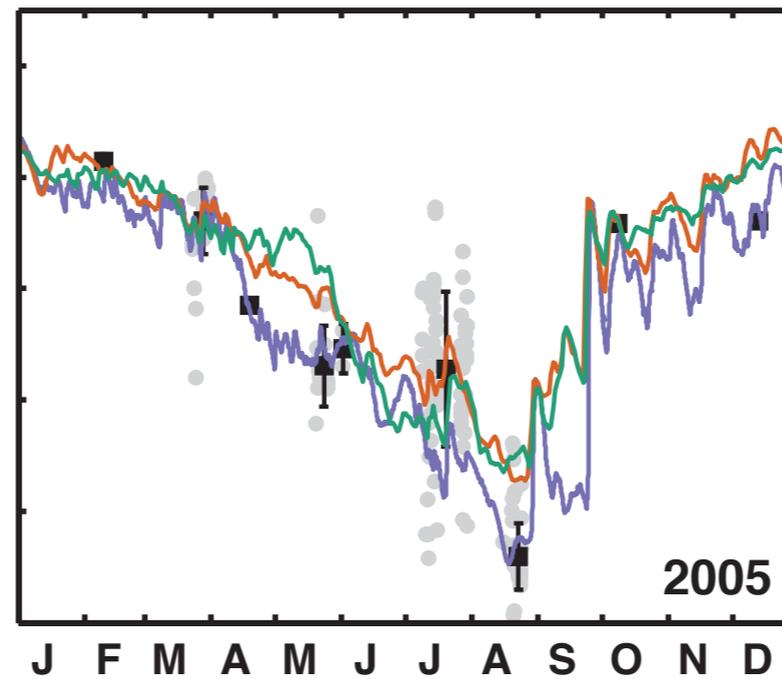
Far Field (z<20m)



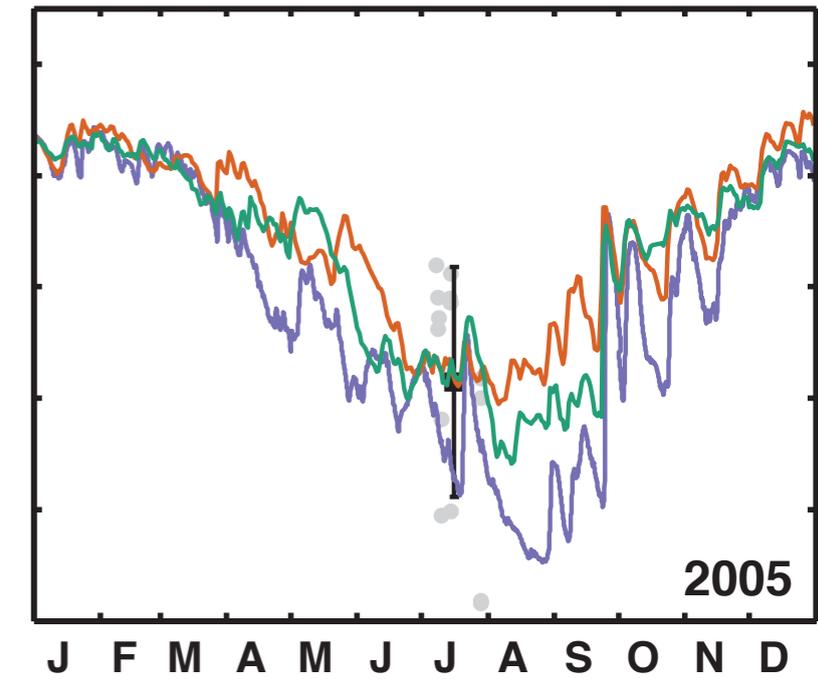
Mississippi (z>20m)



Atchafalaya (z>20m)



Far Field (z>20m)



Find the odd one out!

oxygen
consumption
in the water
column

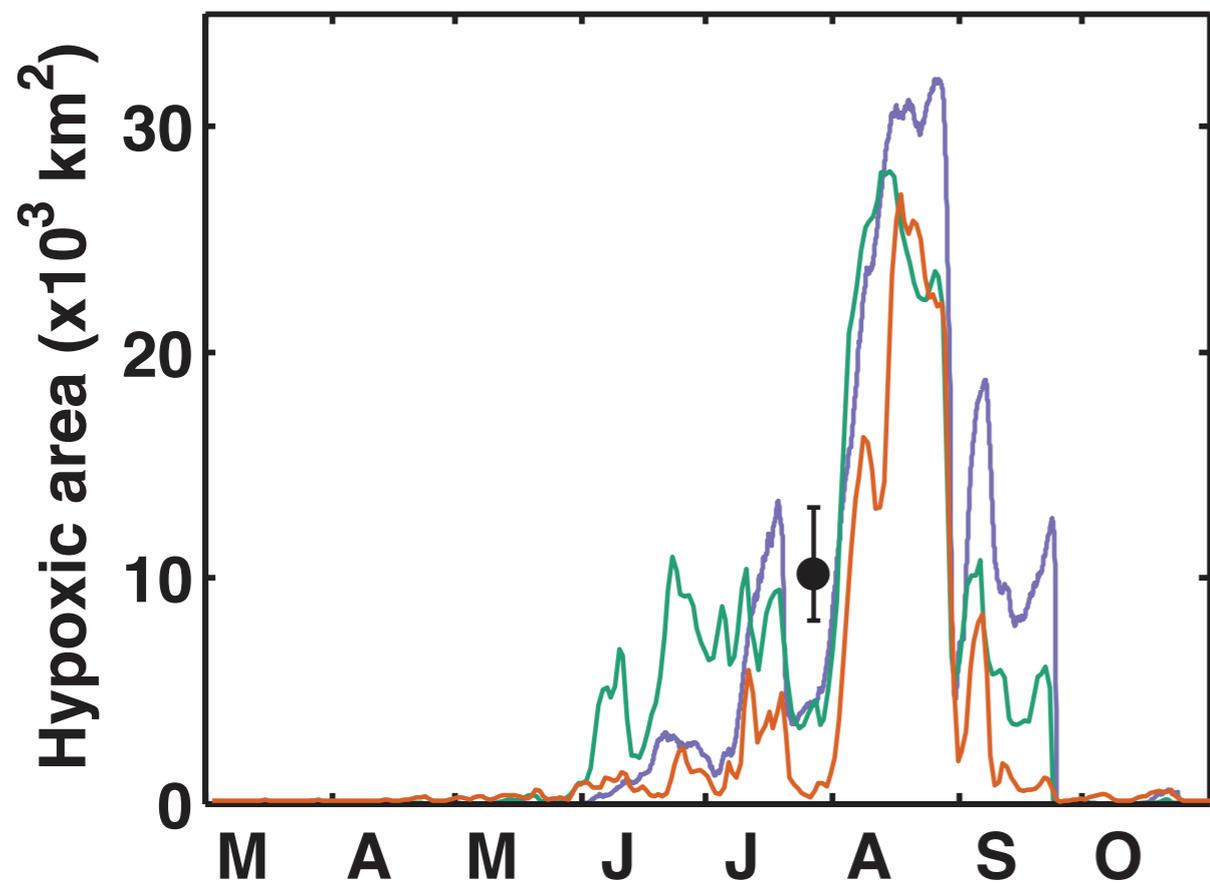
vertical
stratification

oxygen
consumption by
the sediment

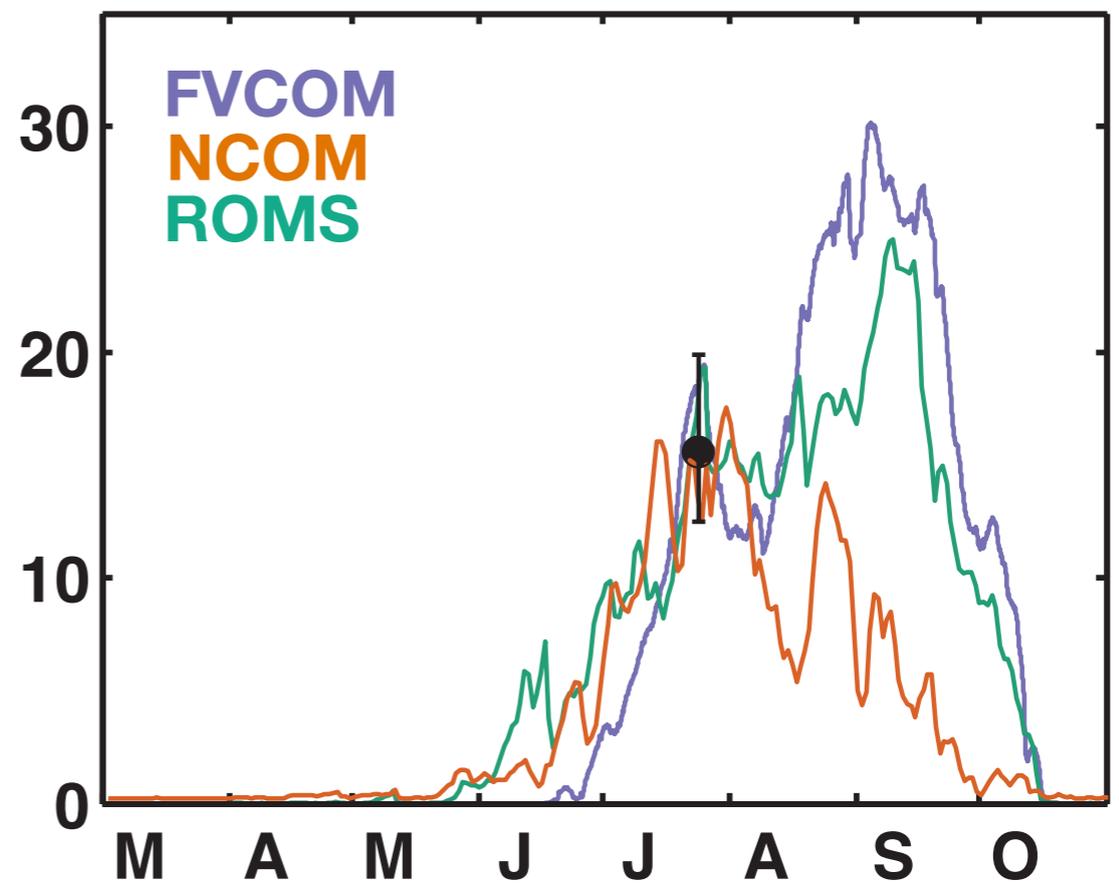
bottom drag
parameter

vertical
attenuation
of shortwave
radiation

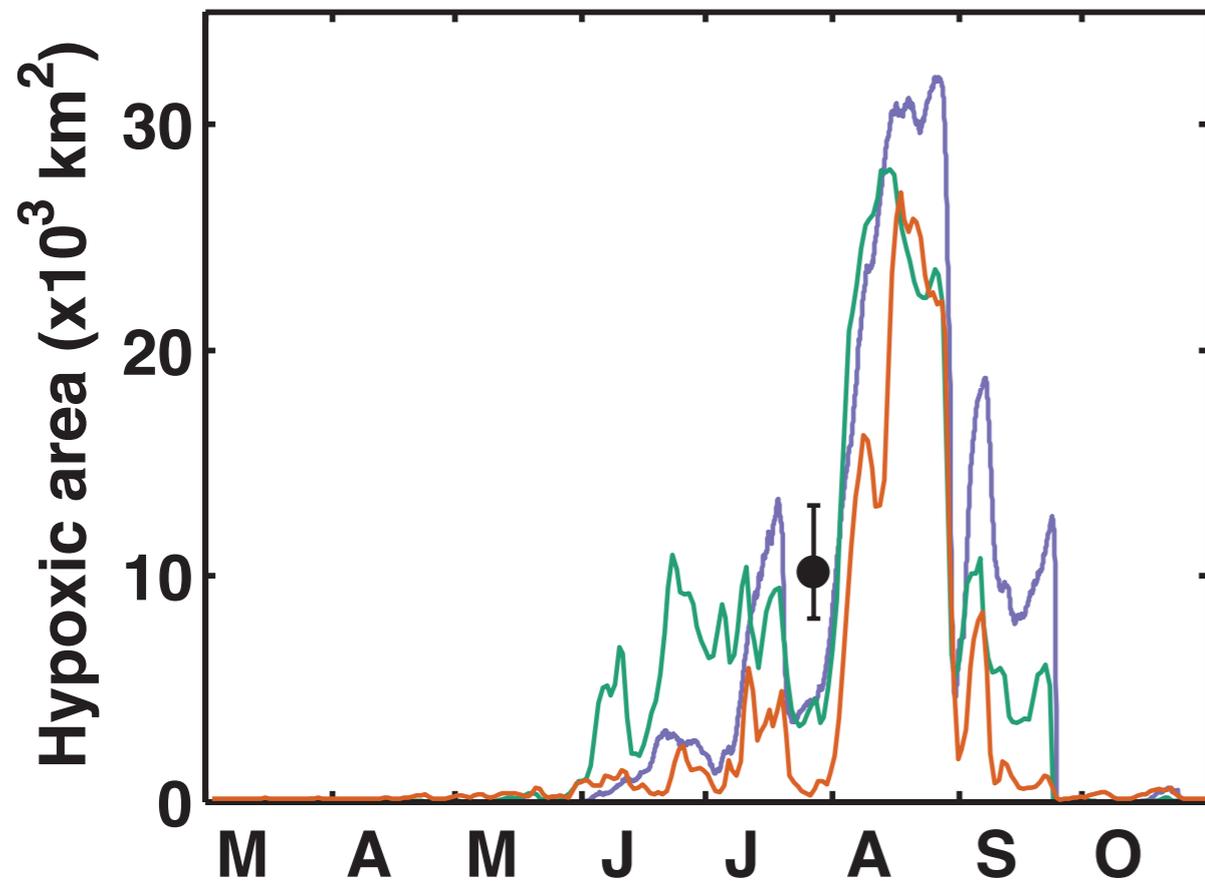
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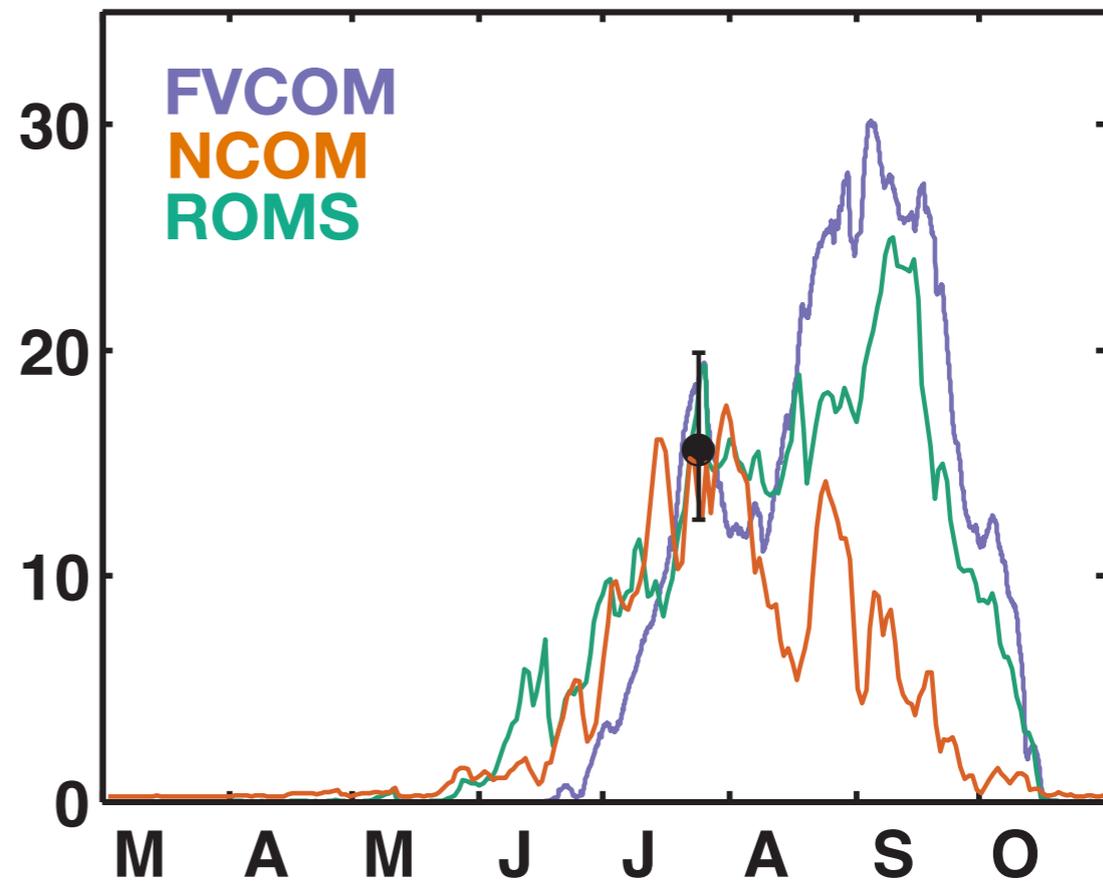
2006



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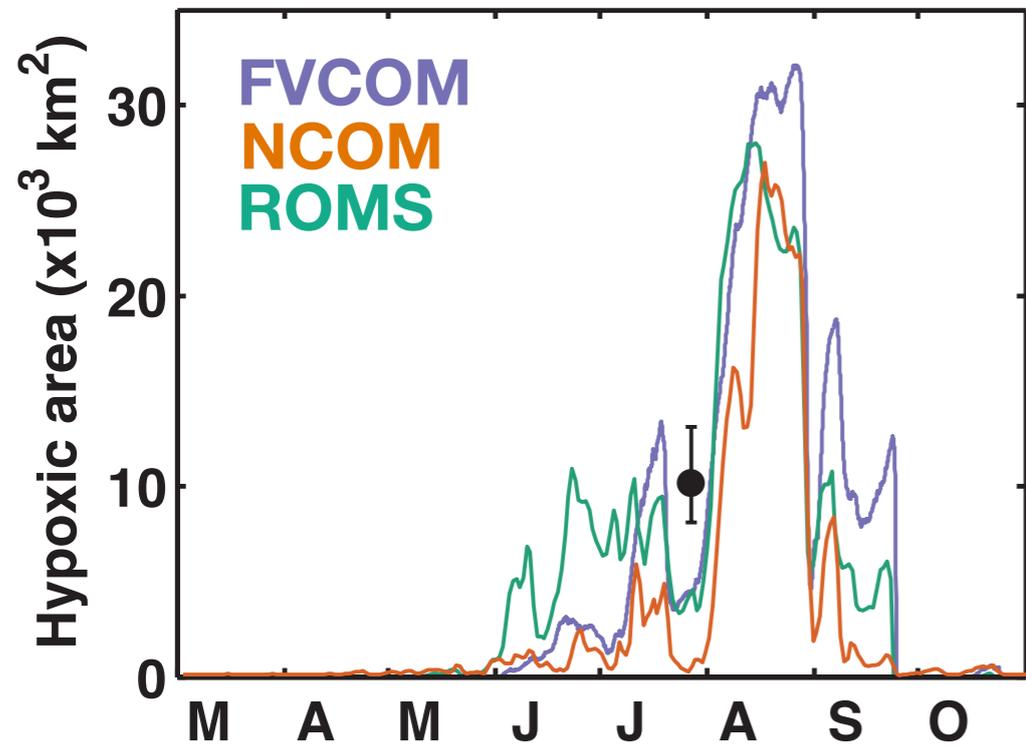
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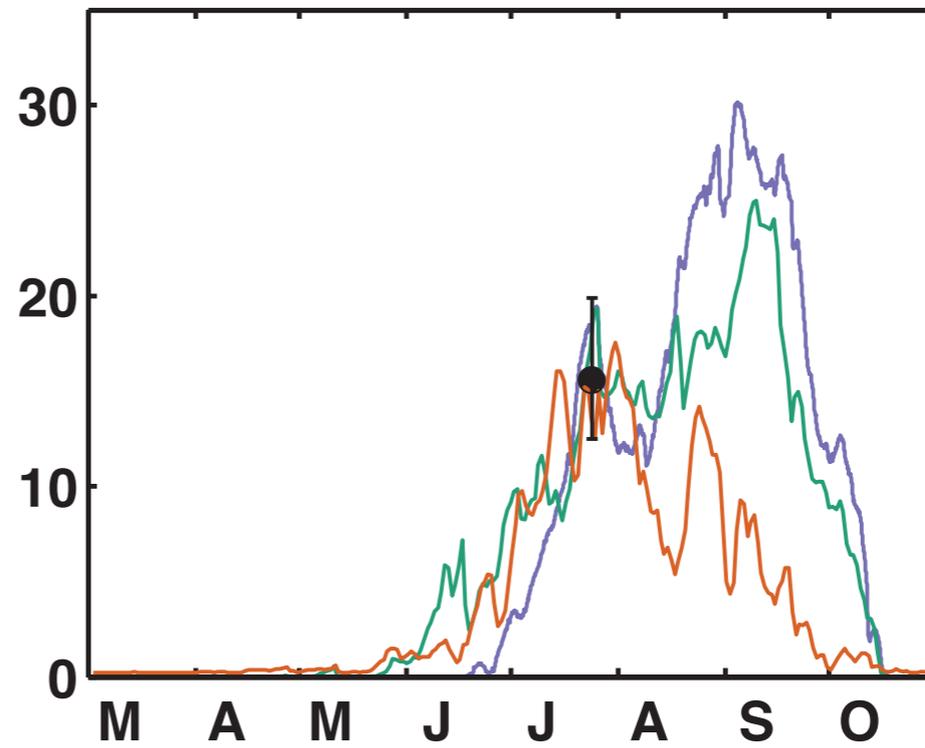
Oxygen concentration is controlled by the balance of **oxygen supply** and **oxygen consumption**.

In these simple hypoxia models water column respiration is equal among all models, but **sediment oxygen consumption** depends on bottom water temperature.

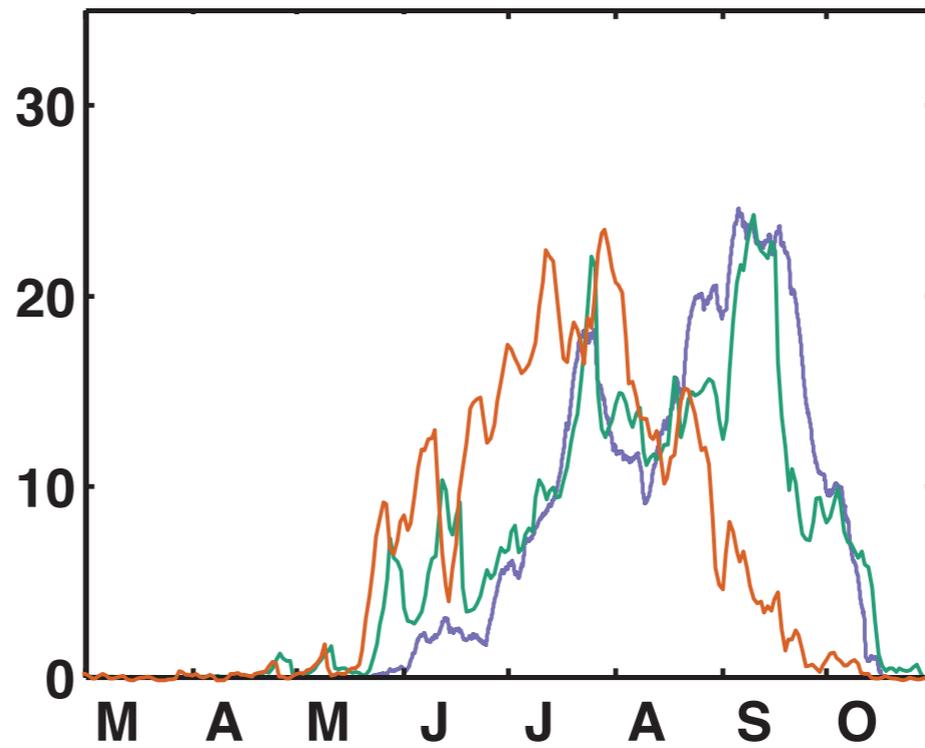
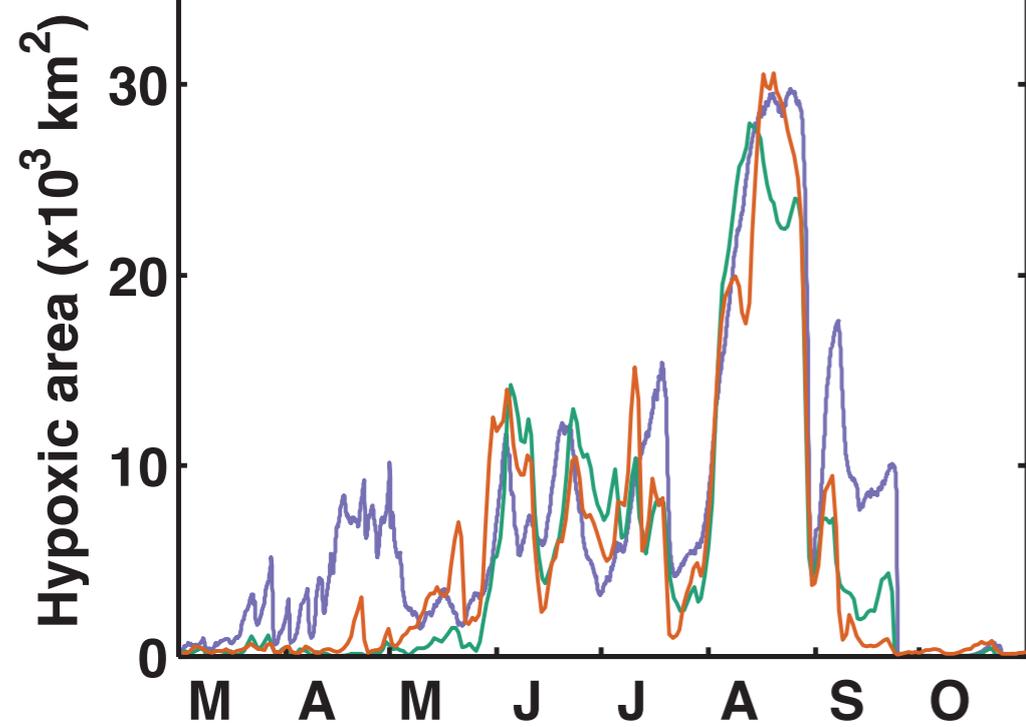
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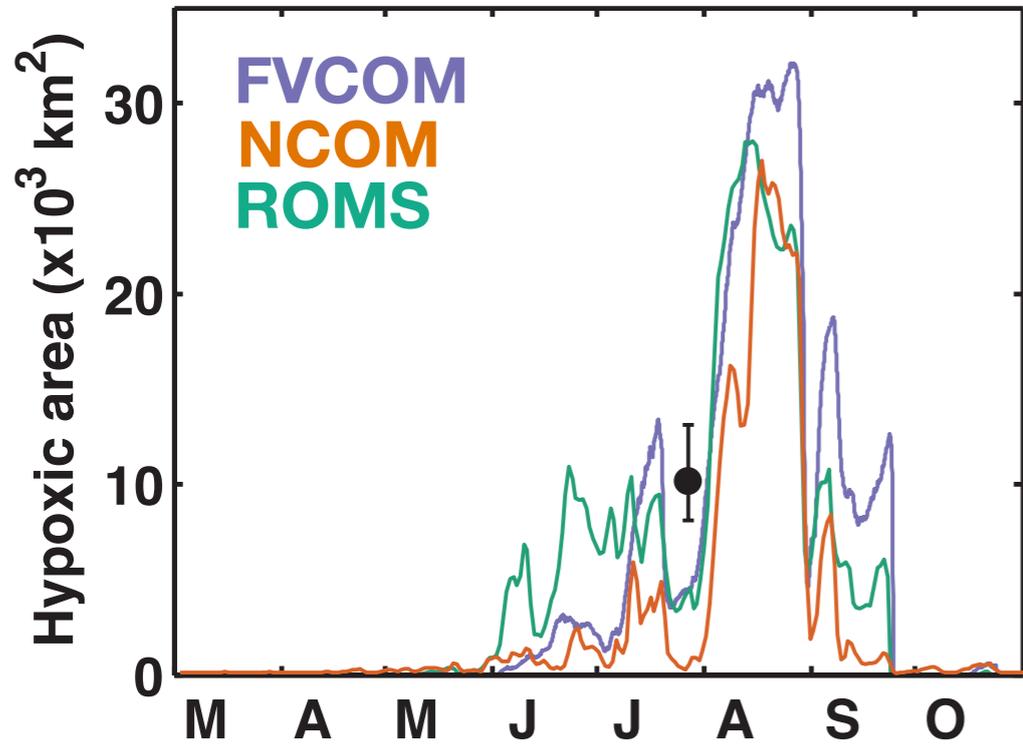


**Simple
model**

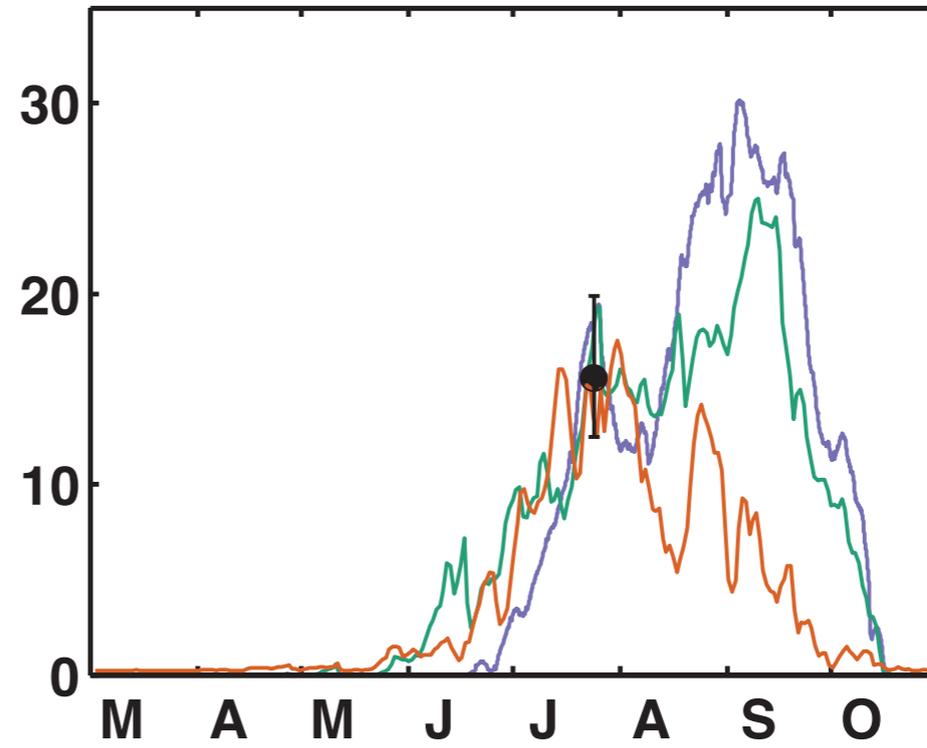


**Diagnostic
runs**

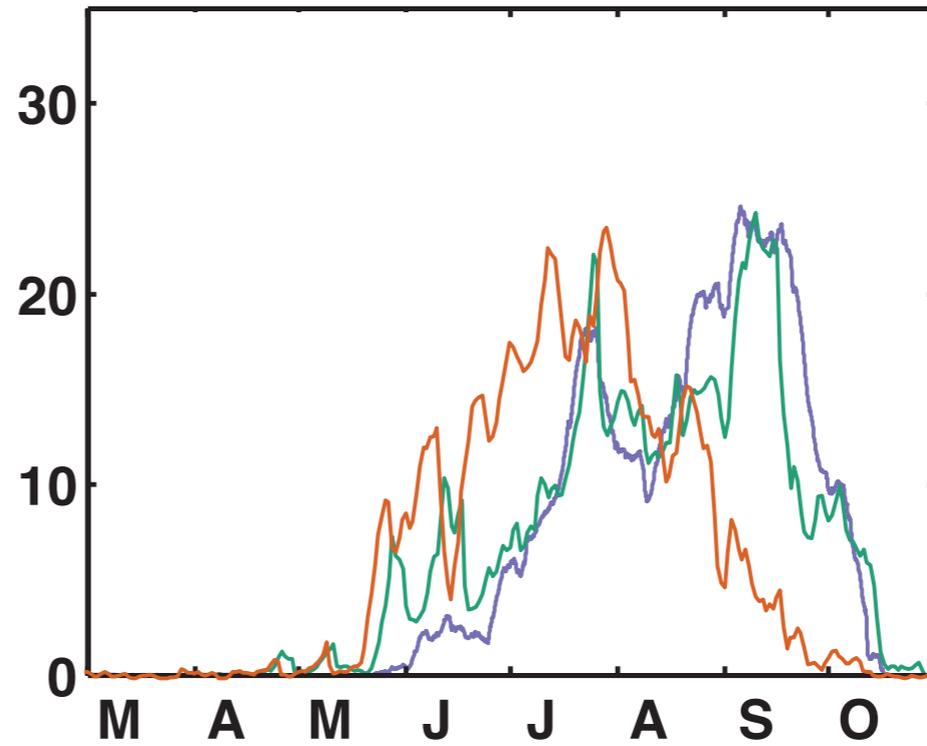
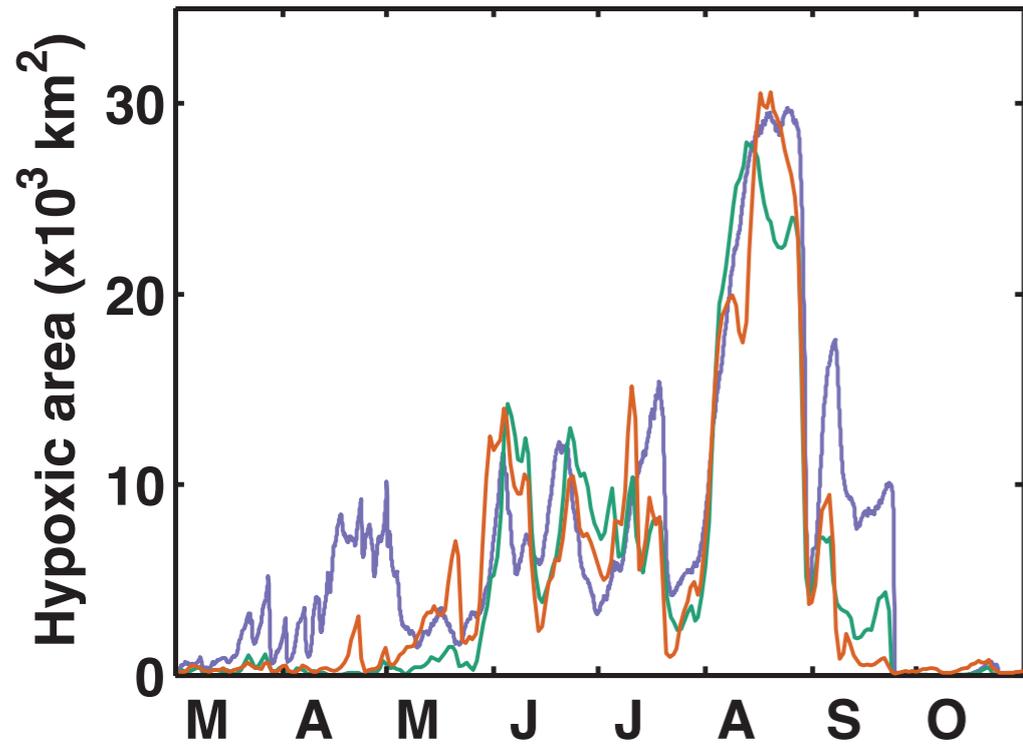
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2006

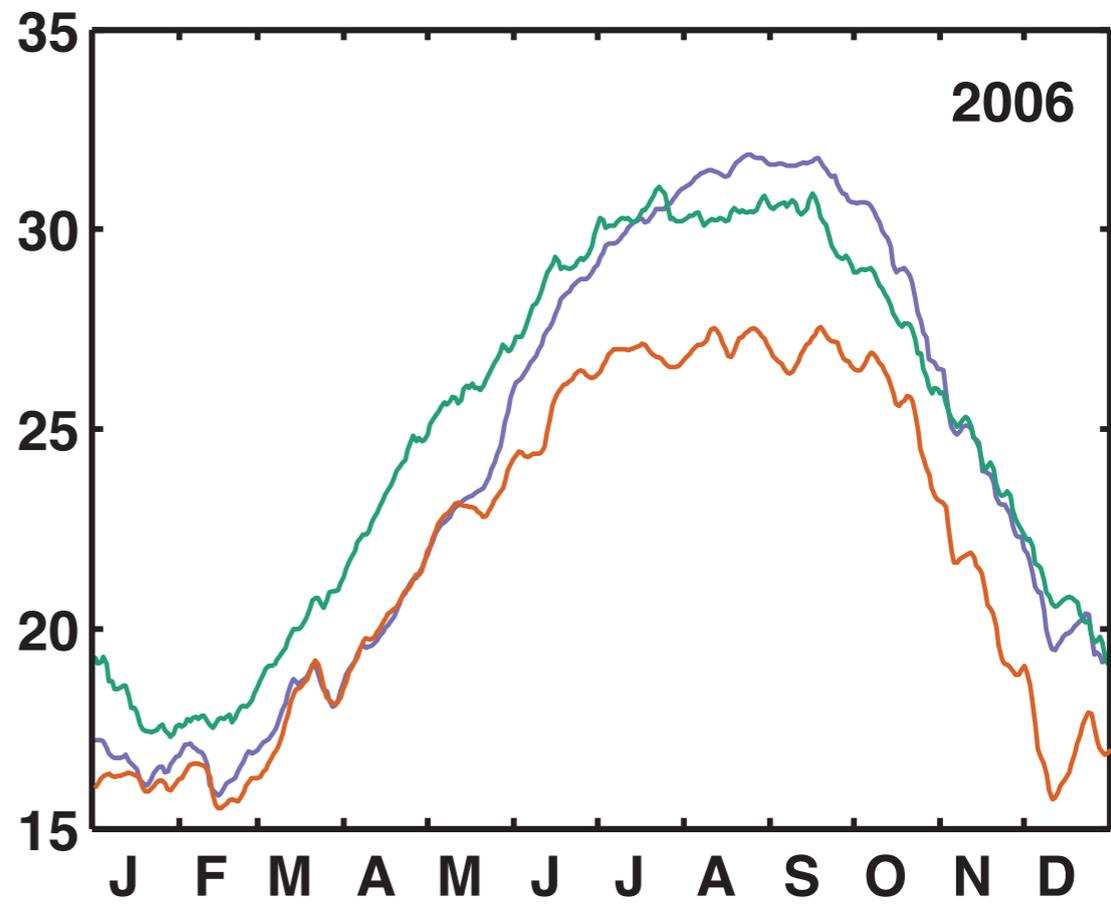
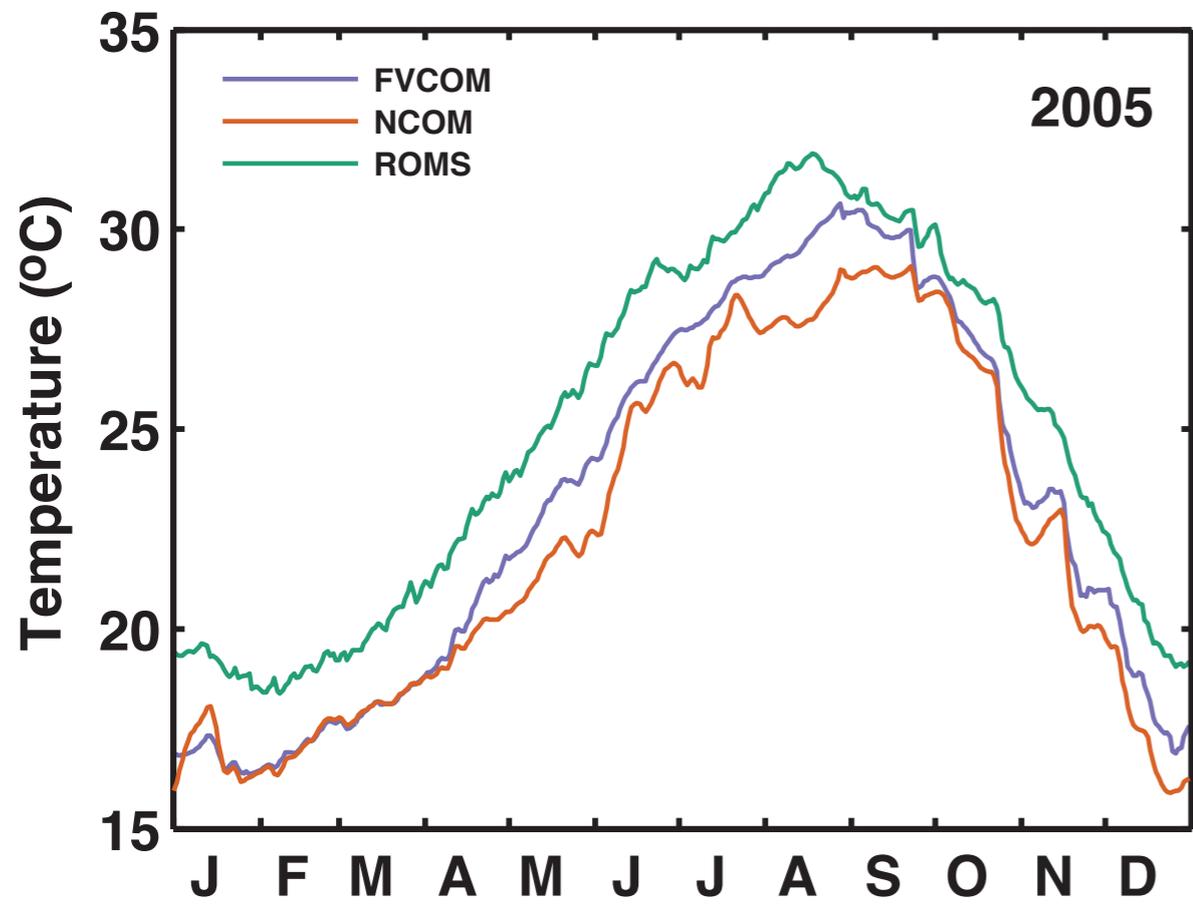


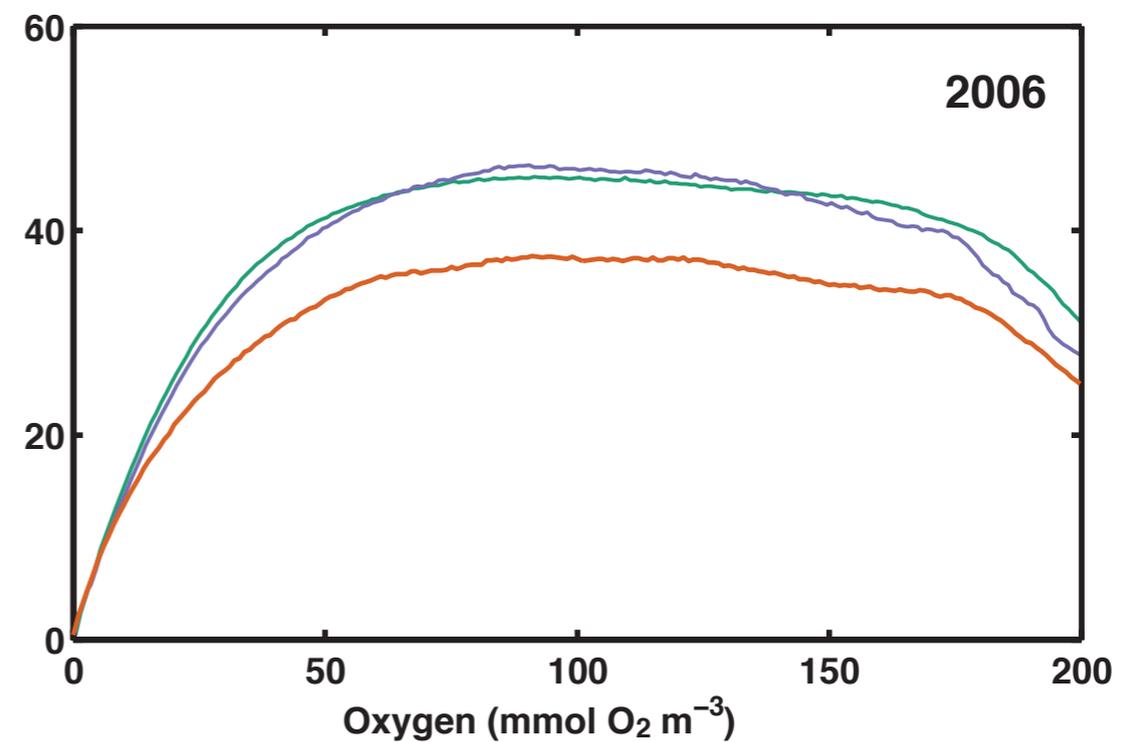
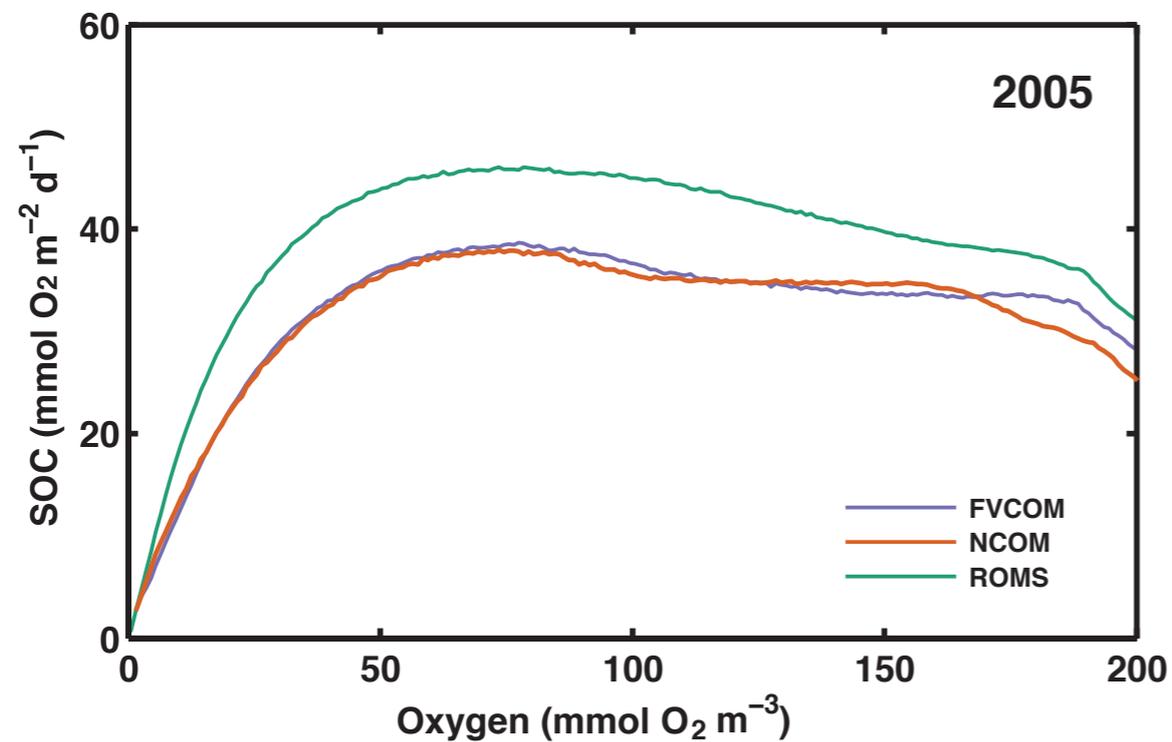
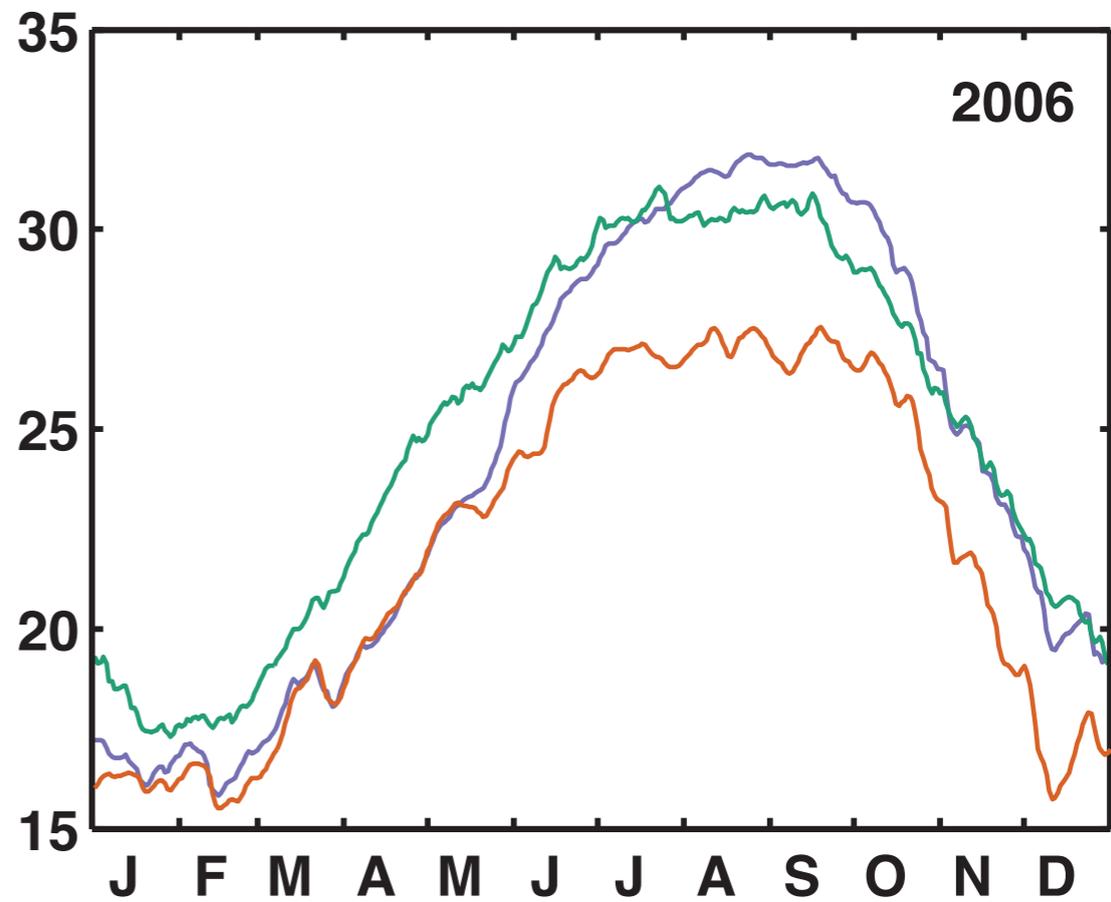
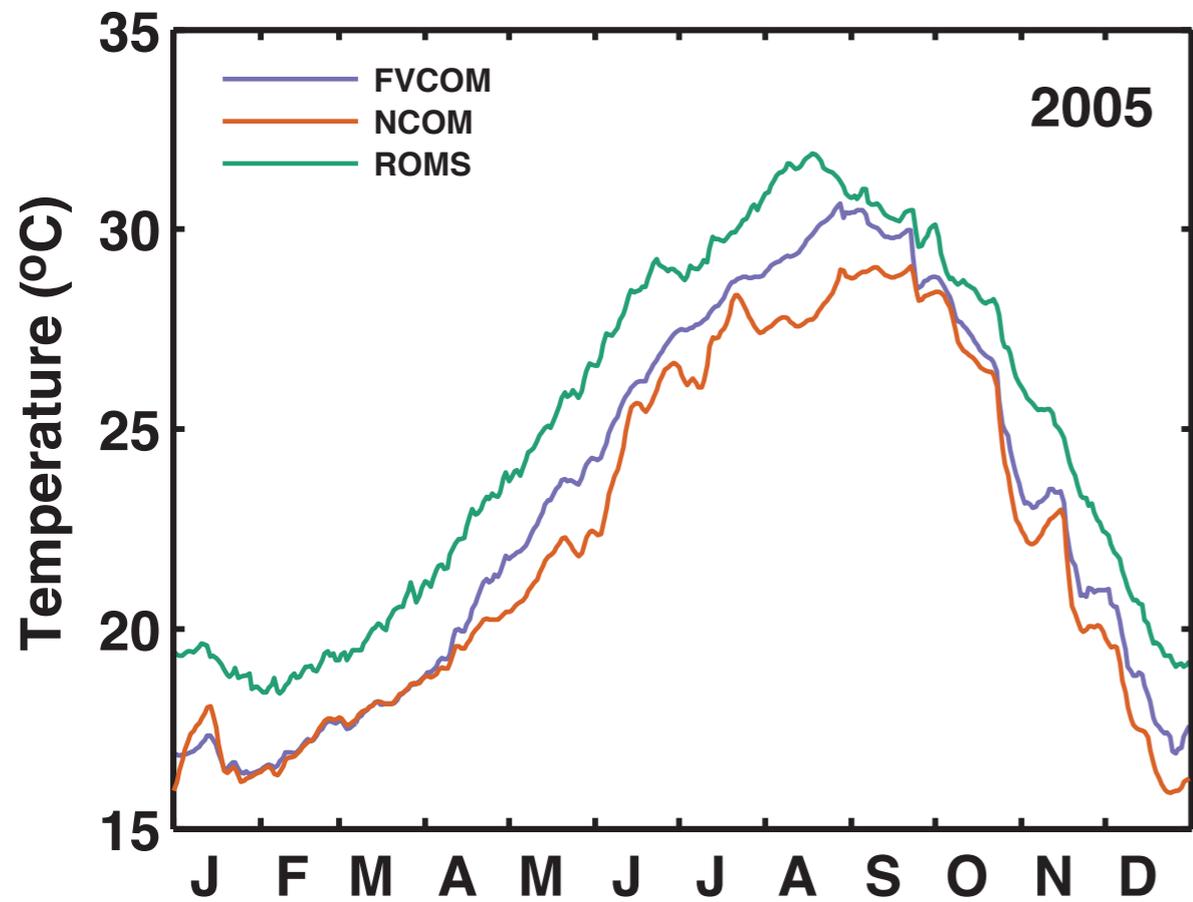
Simple model

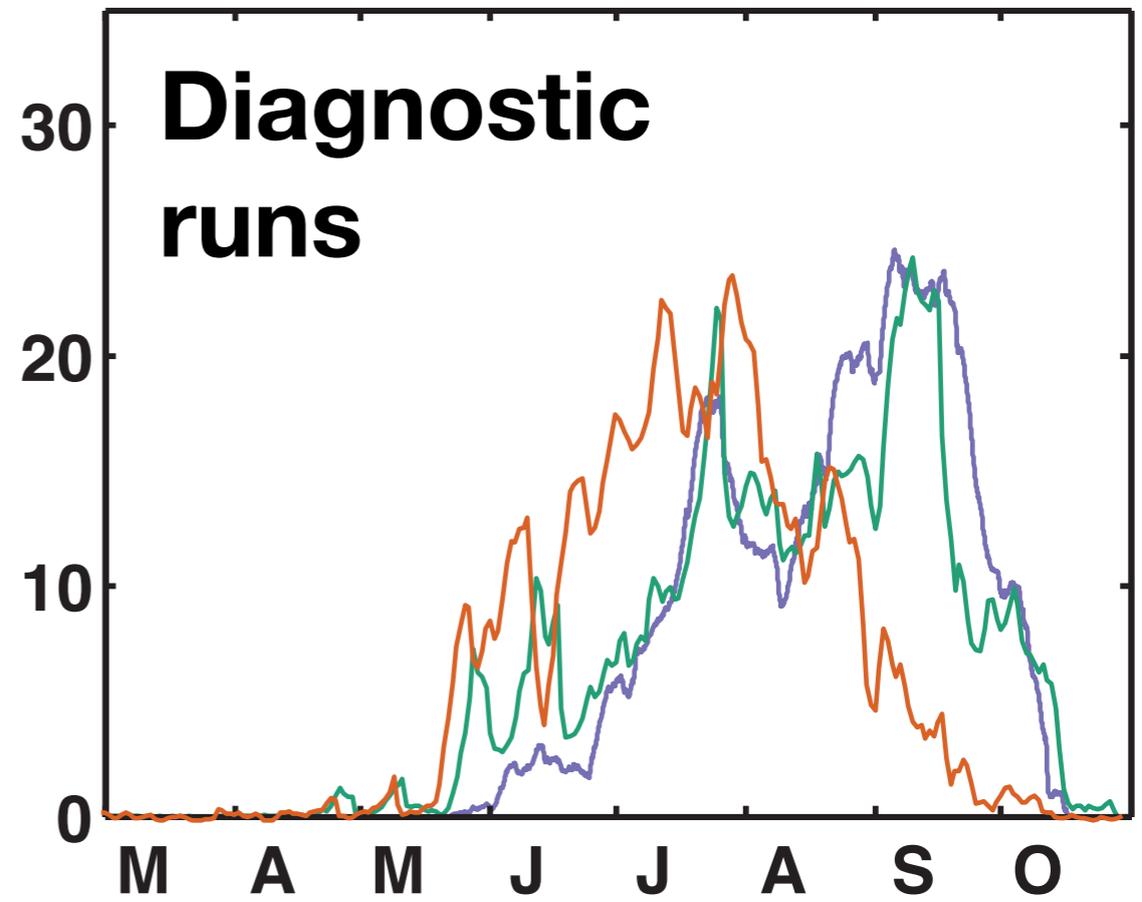
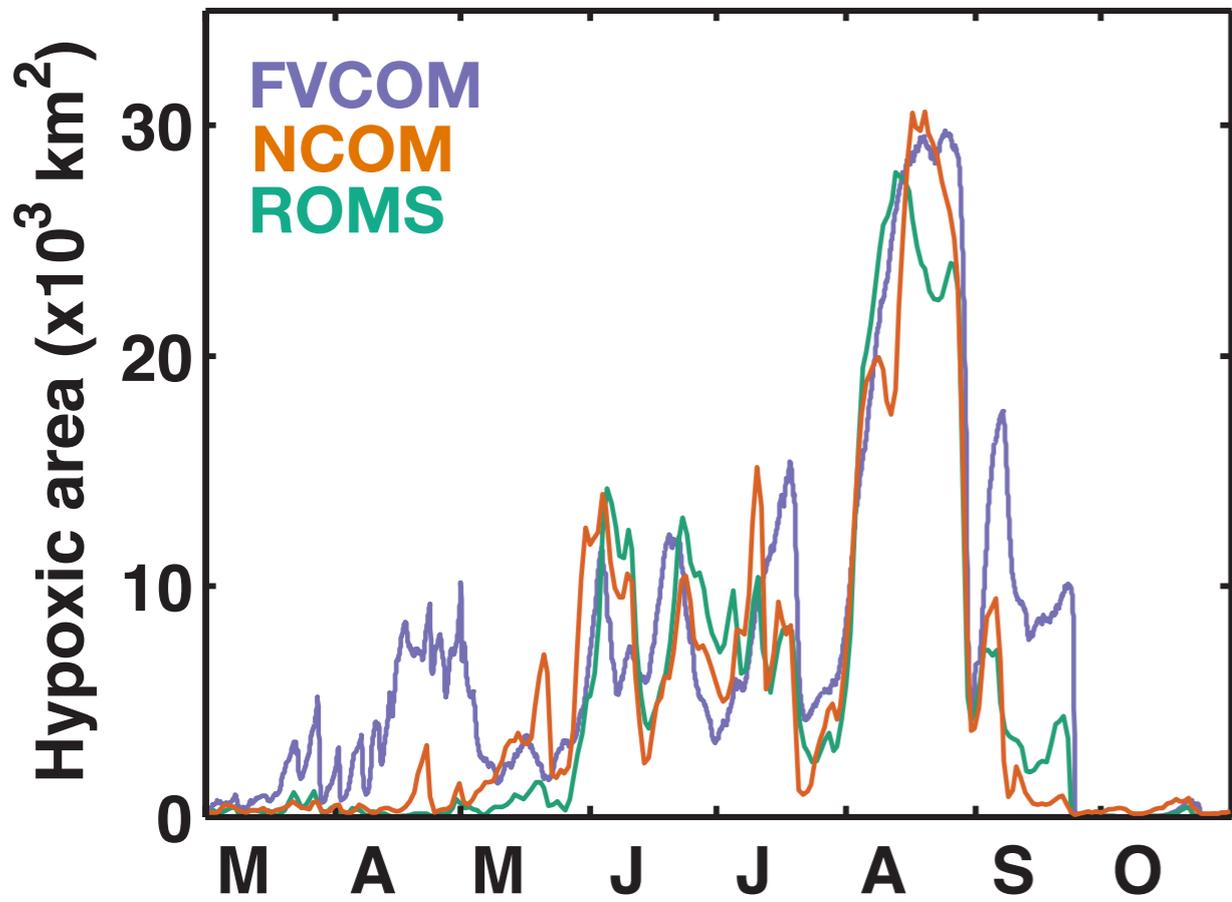


Diagnostic runs

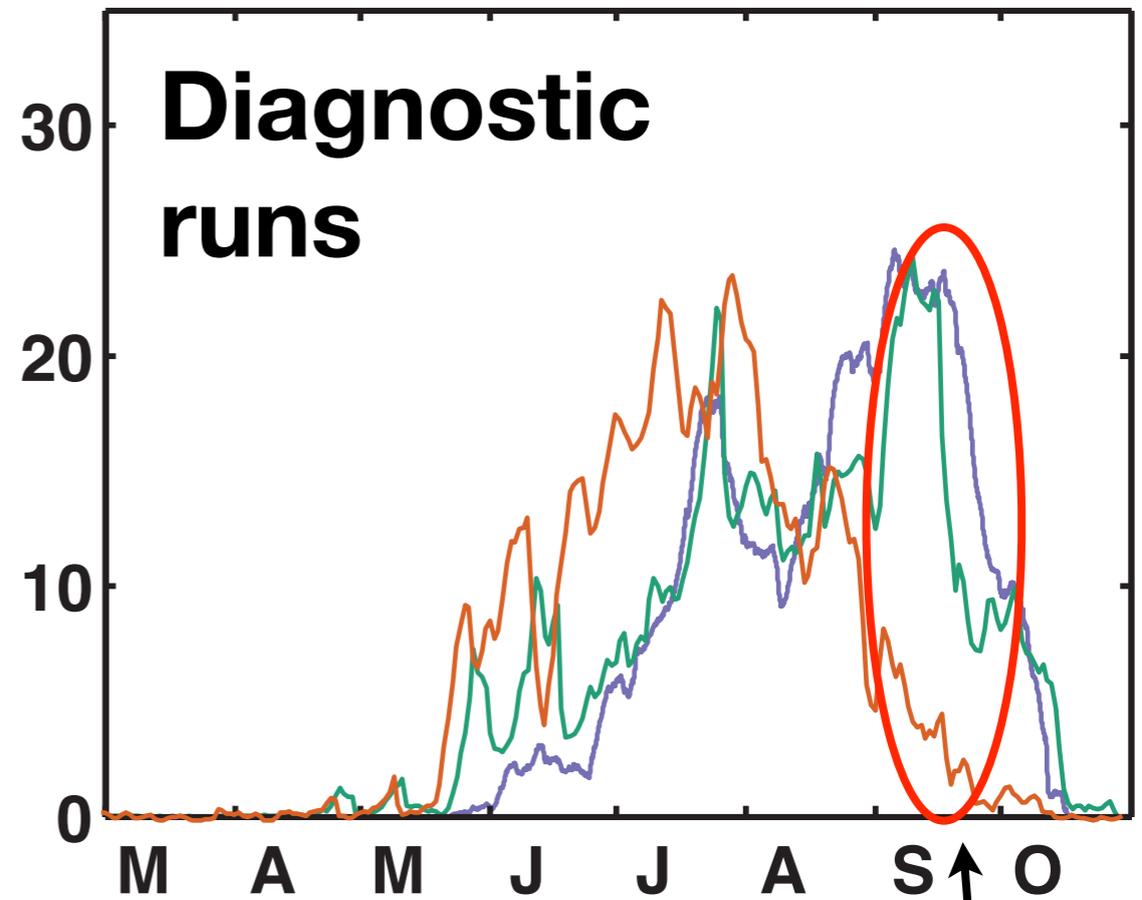
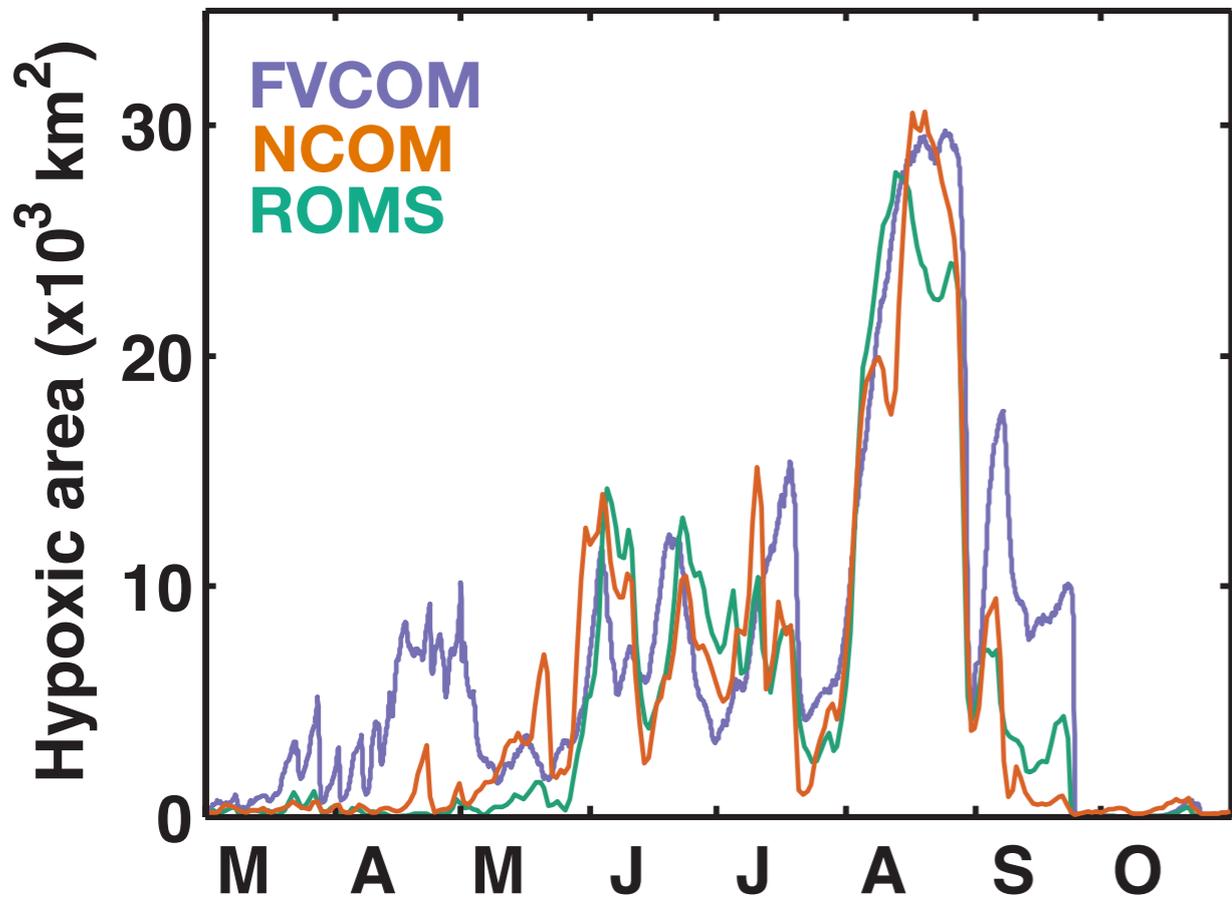
1. Effects of bottom water temperature





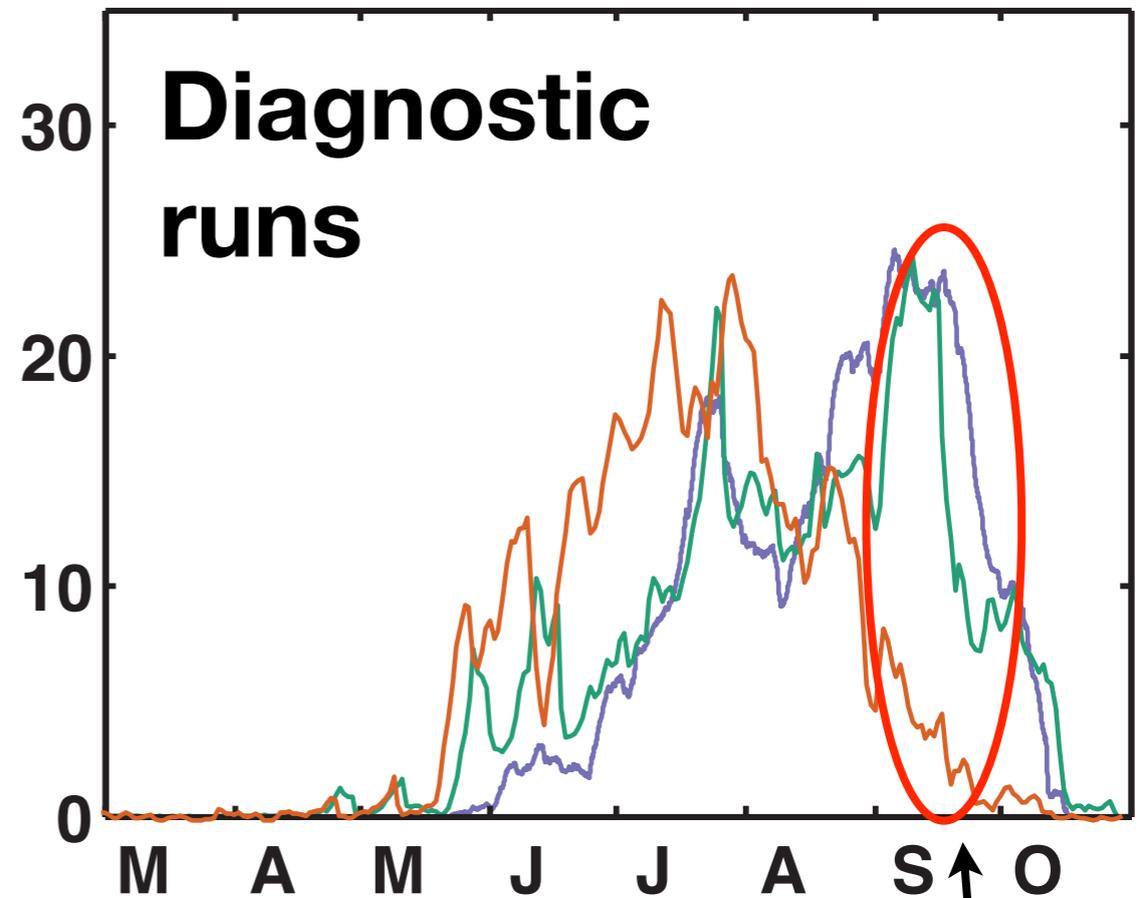
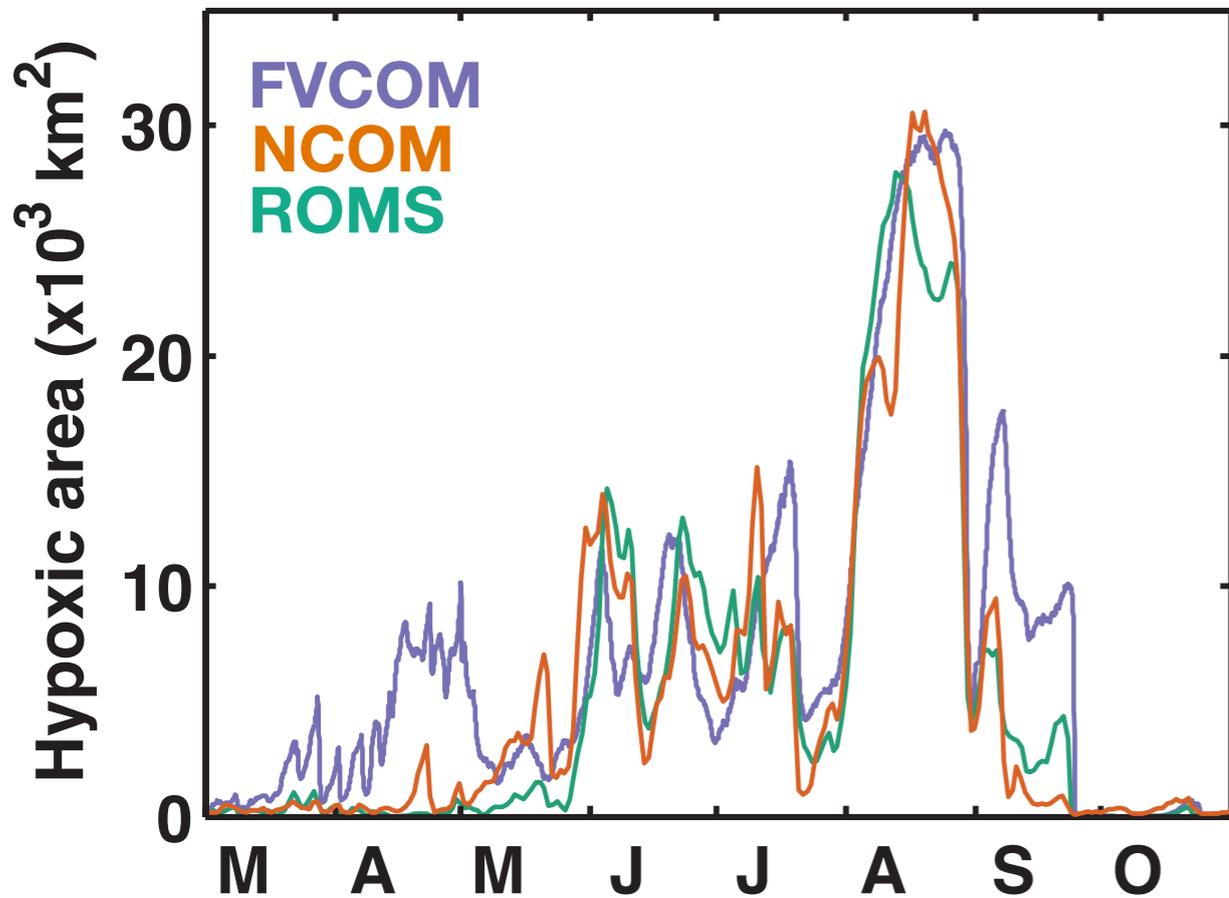


2. Effects of oxygen supply



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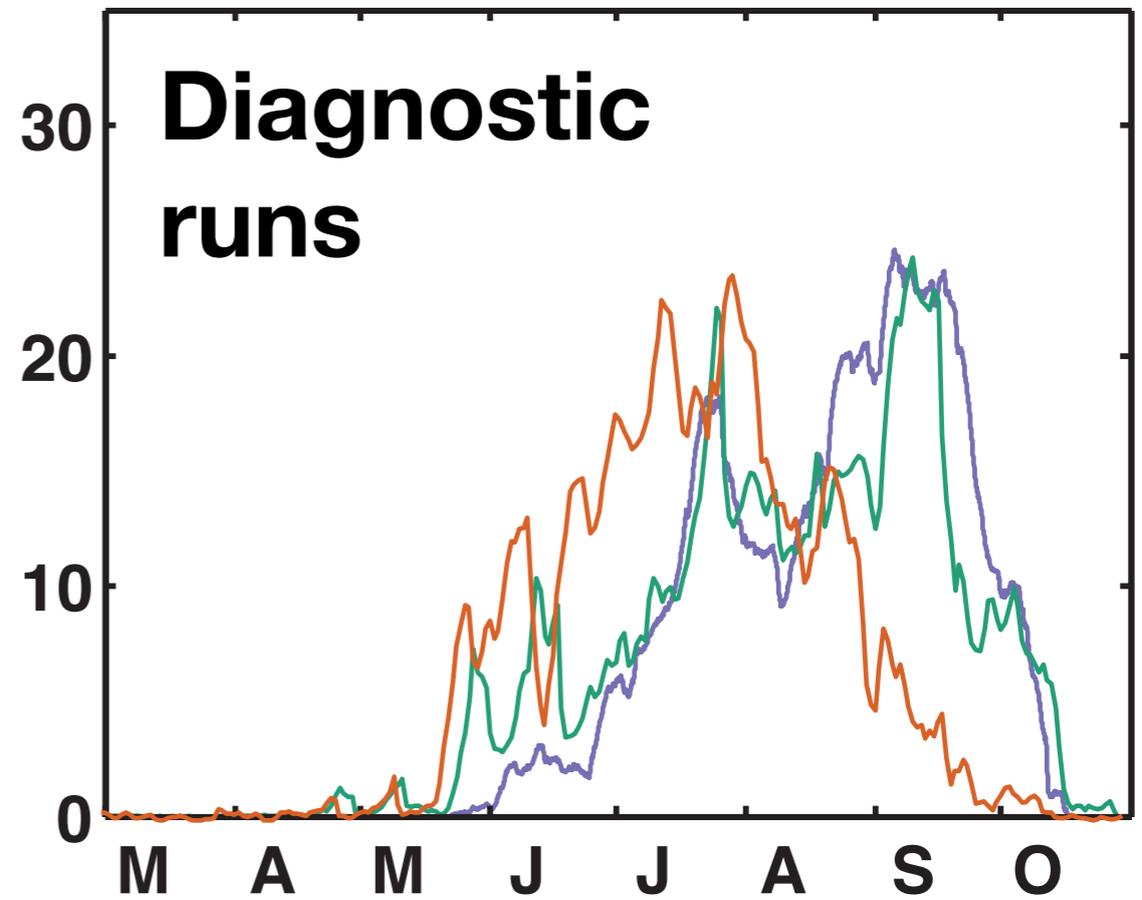
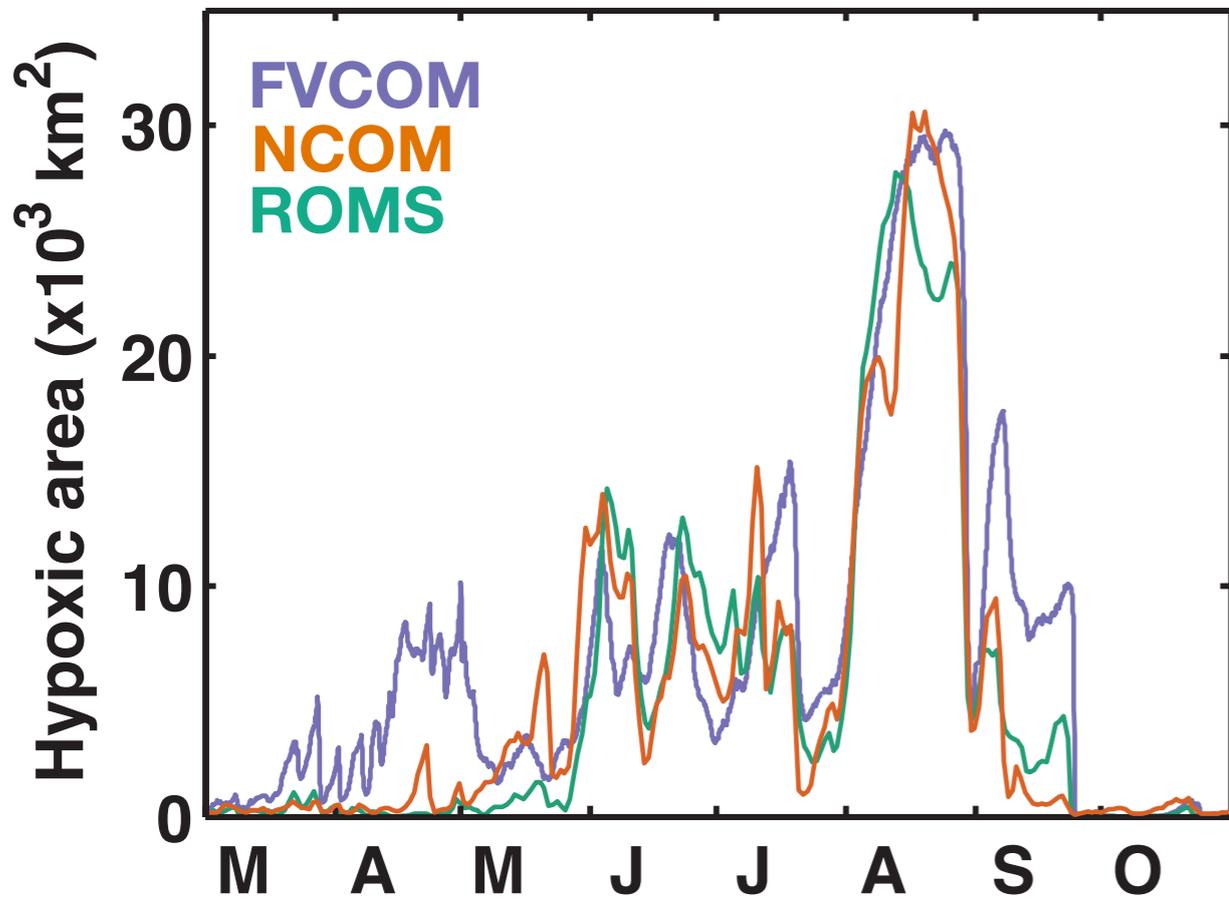
ROMS has strongest stratification; FVCOM the weakest



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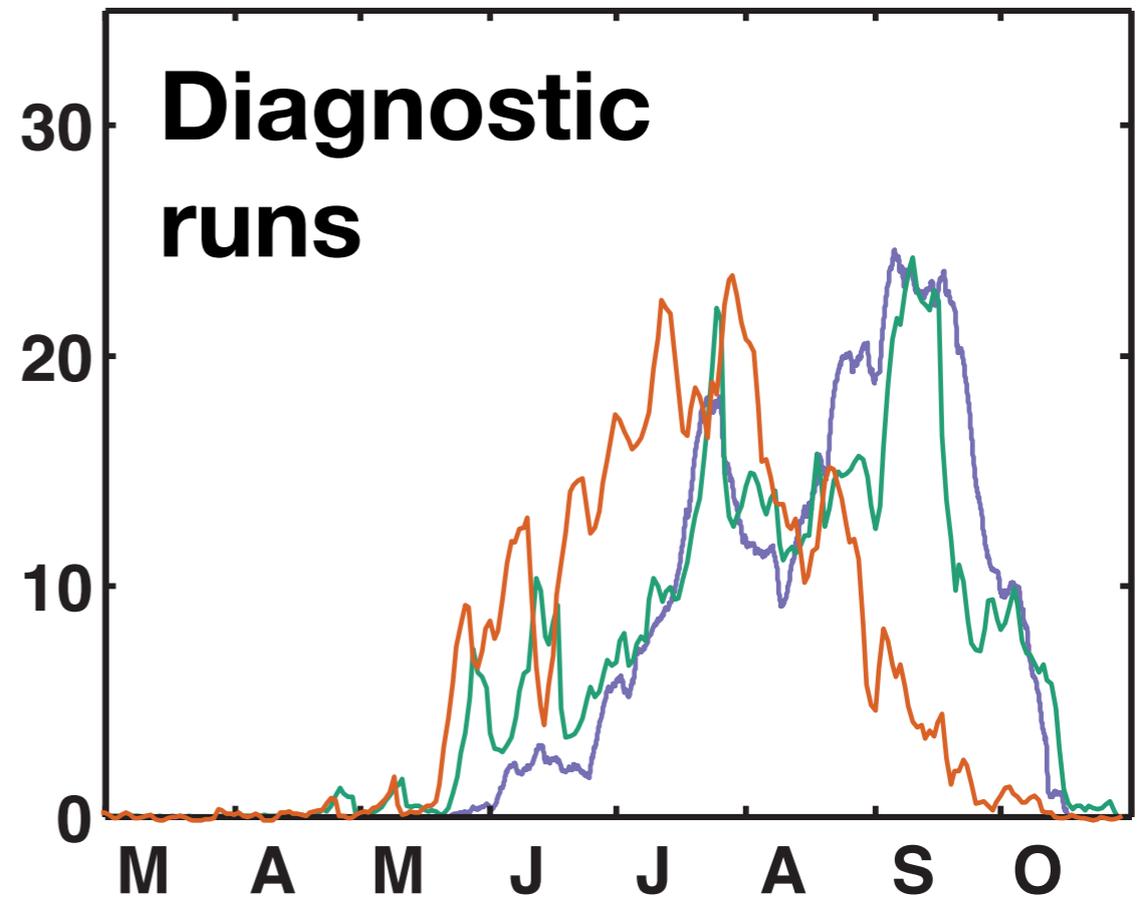
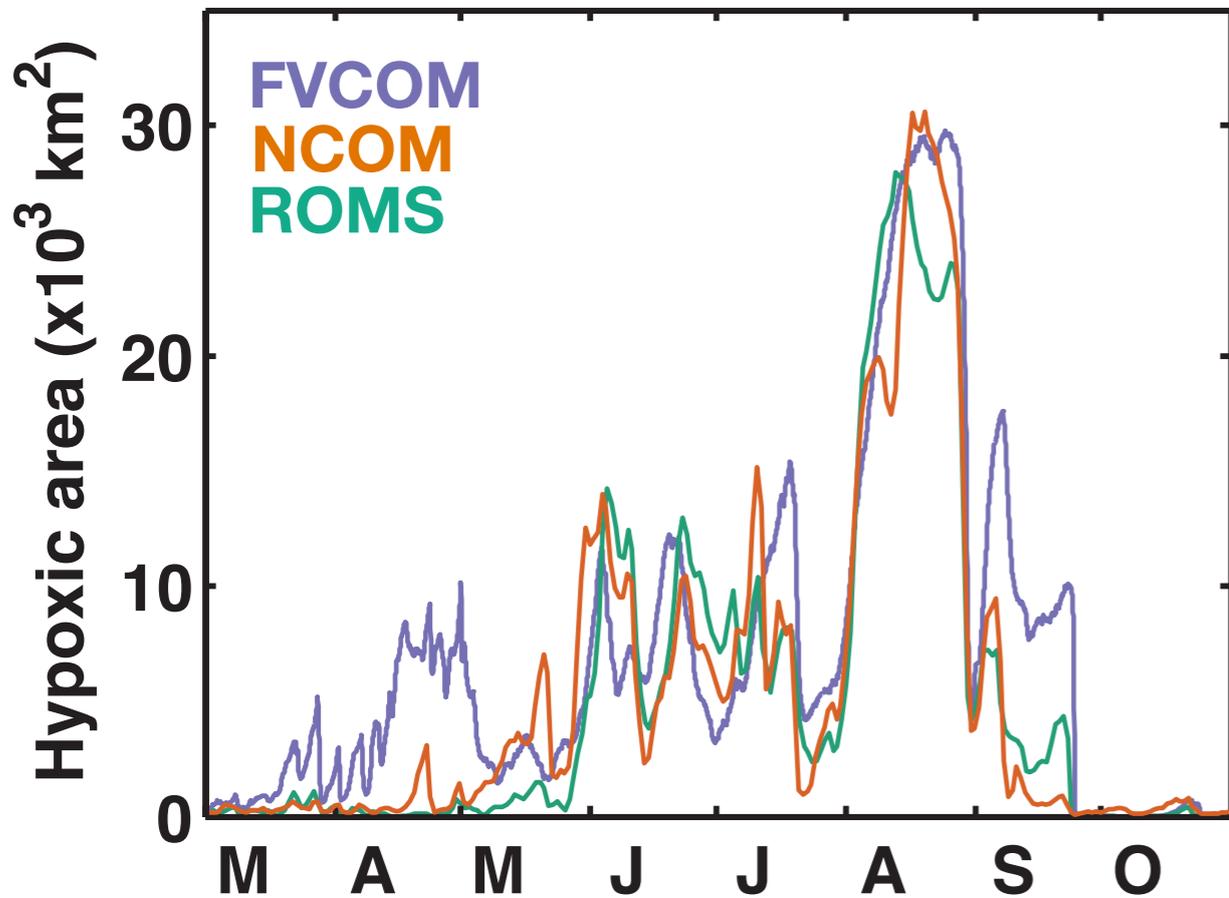
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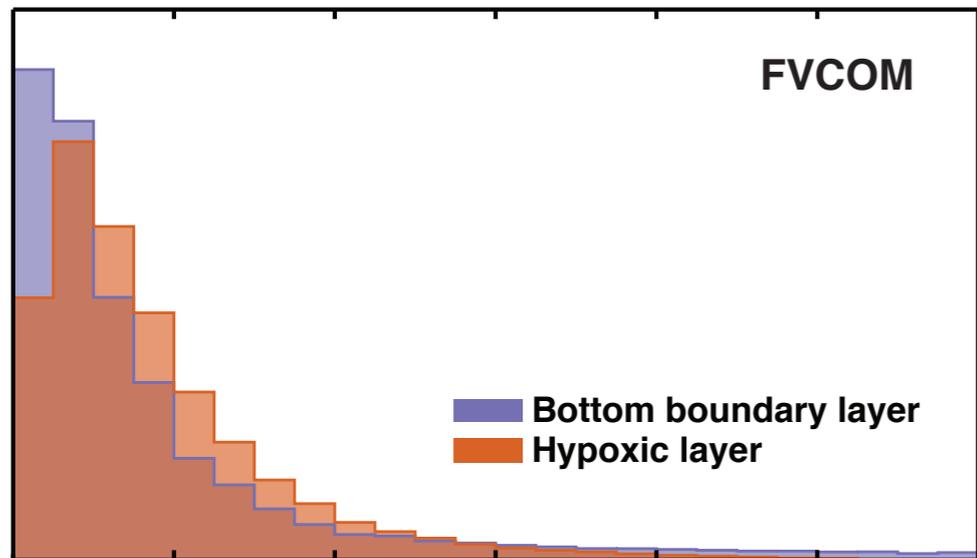


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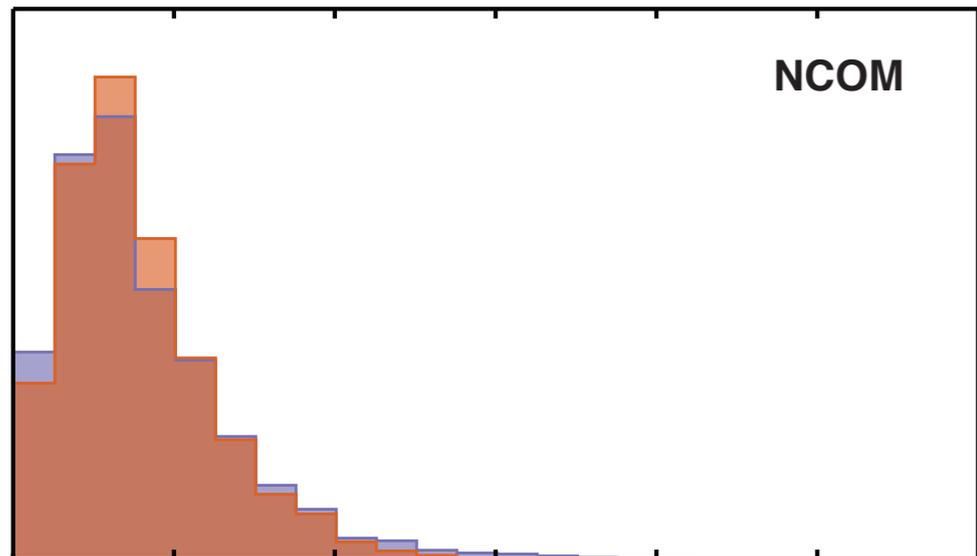
Differences in BBL (hypoxic layer = BBL).

2005

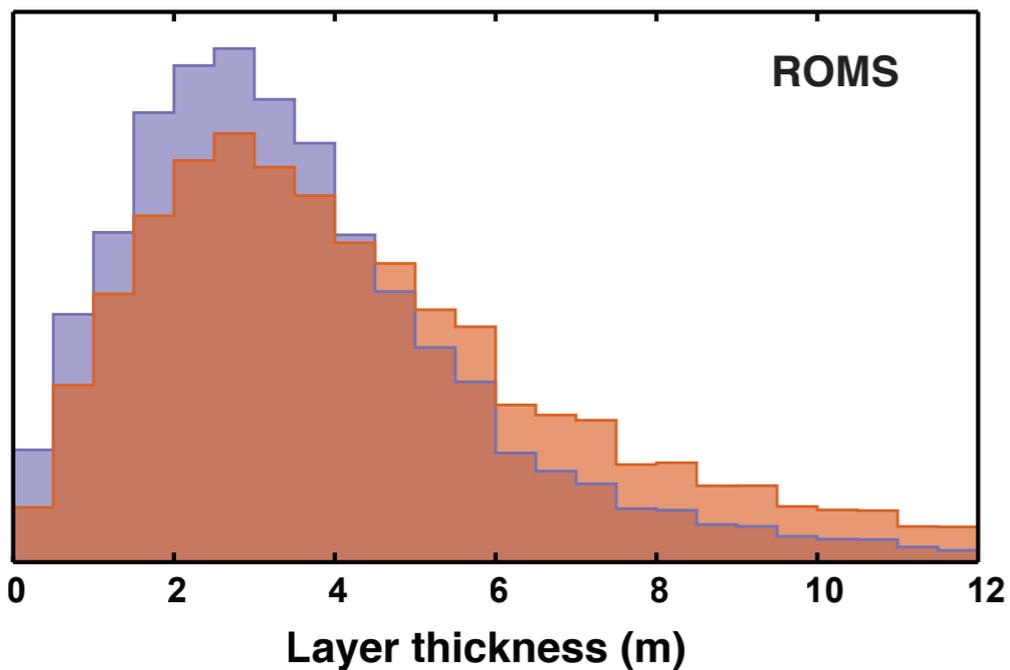


Hypoxic layer and BBL coincide in all models.

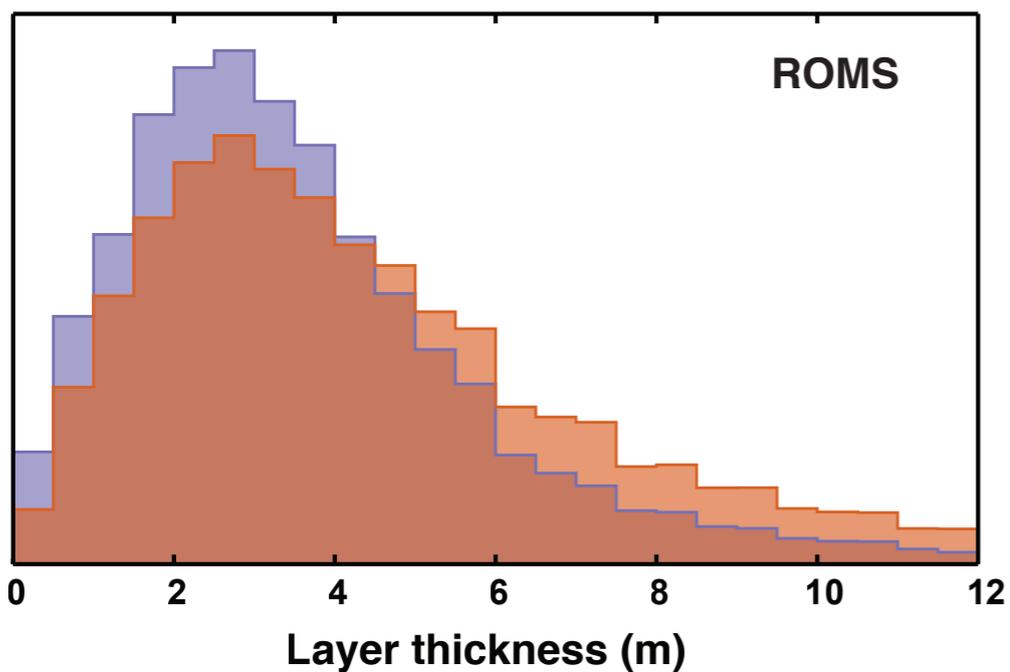
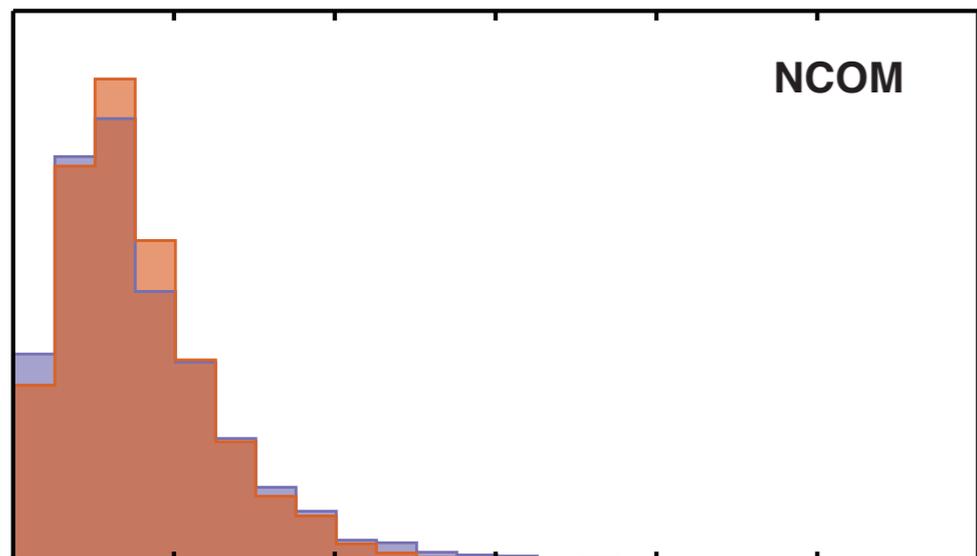
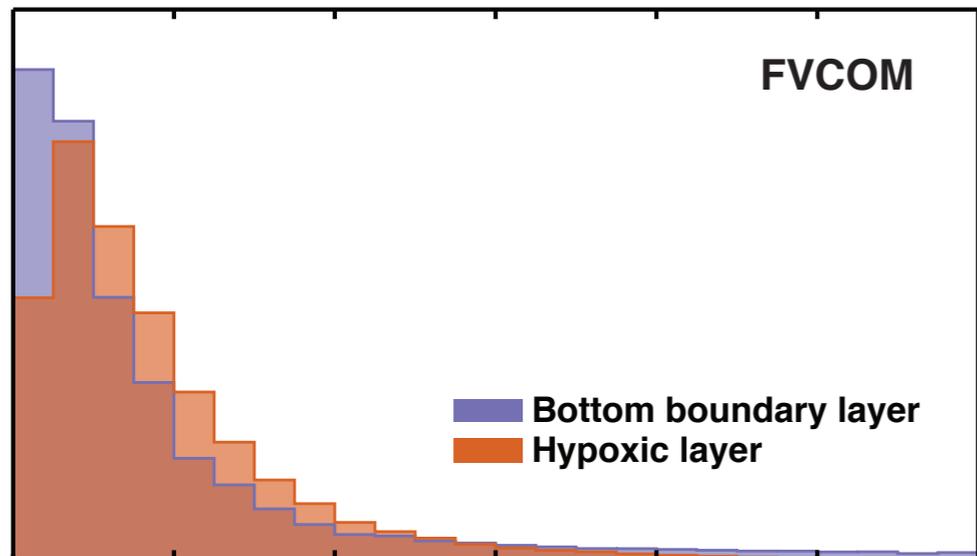
BBL in ROMS is thicker than in the other two models.



Driving a thick BBL to hypoxia requires more oxygen consumption than a thin BBL.



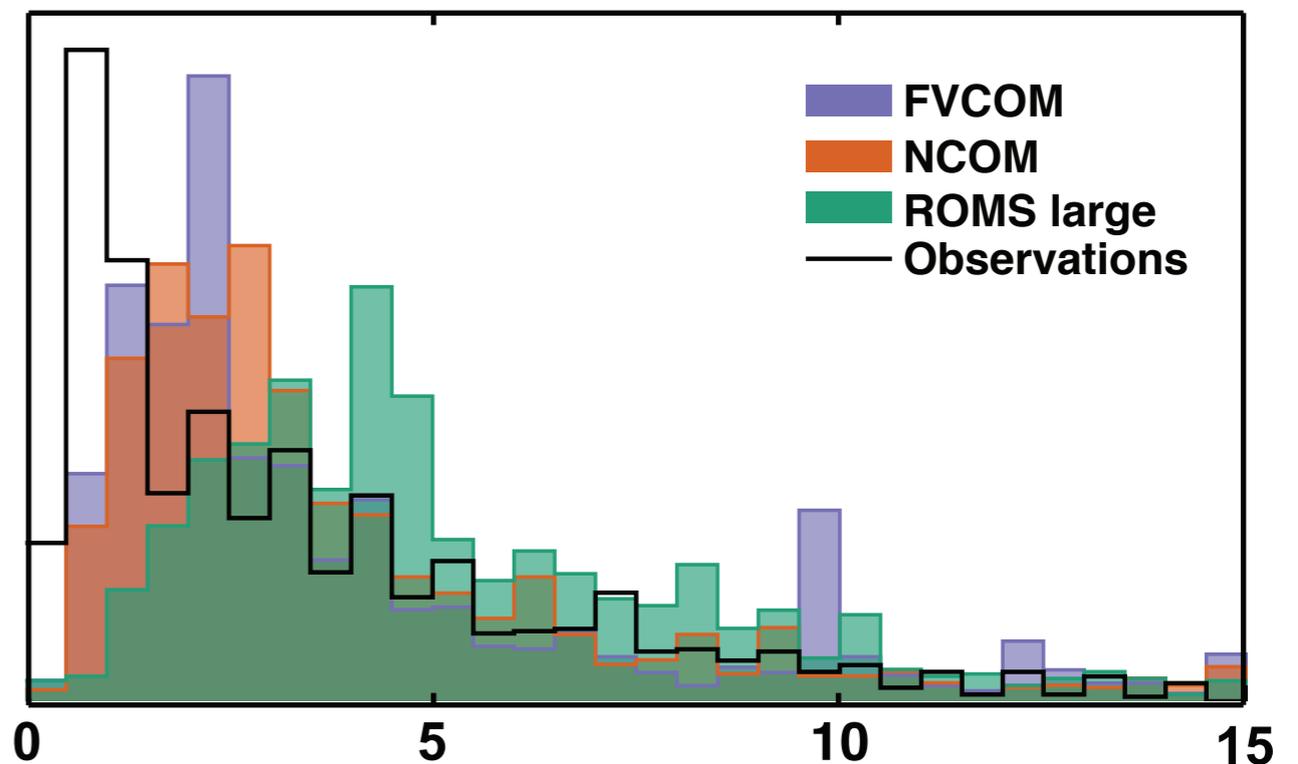
2005



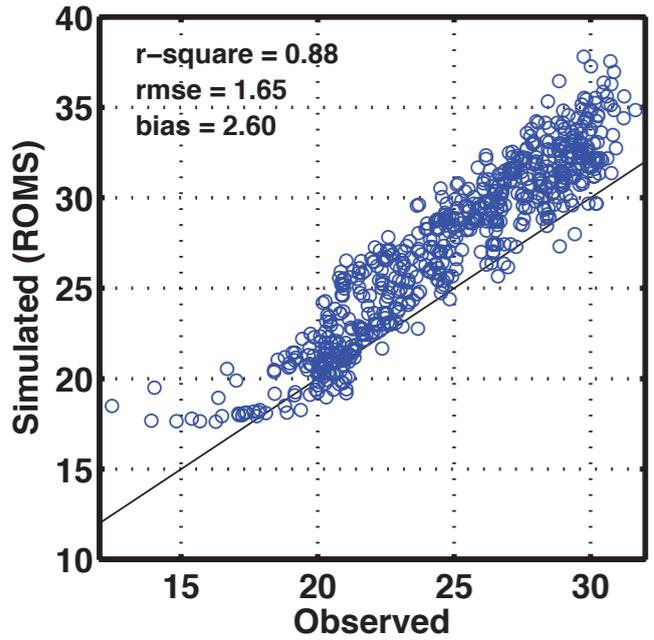
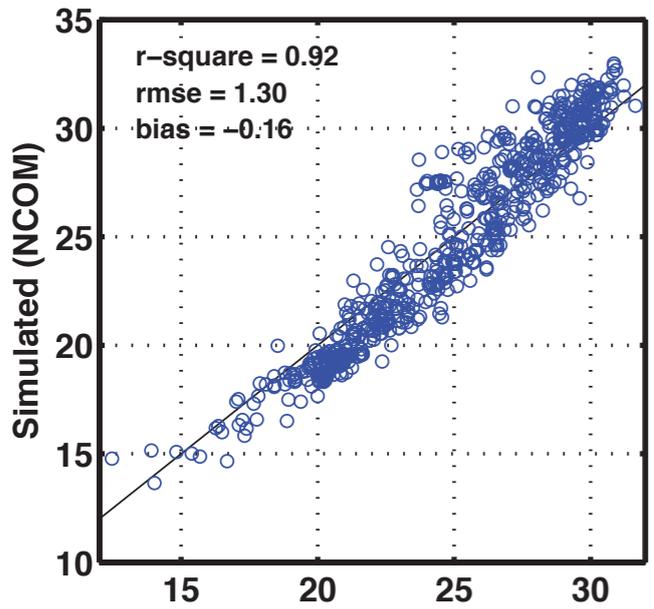
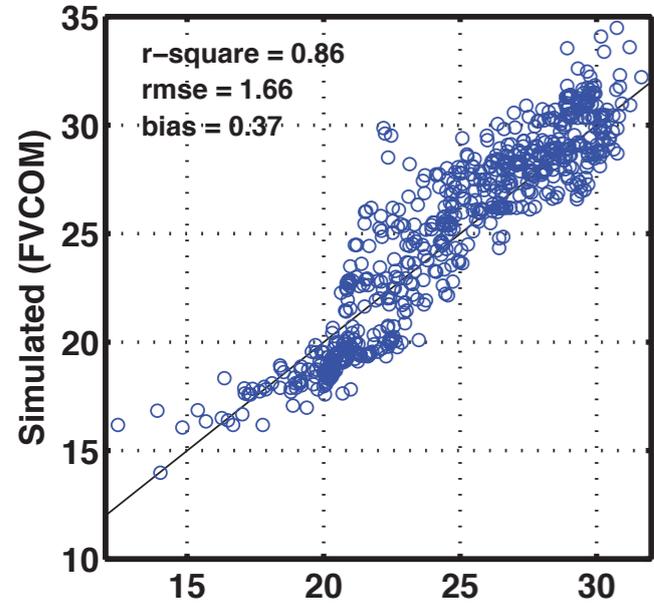
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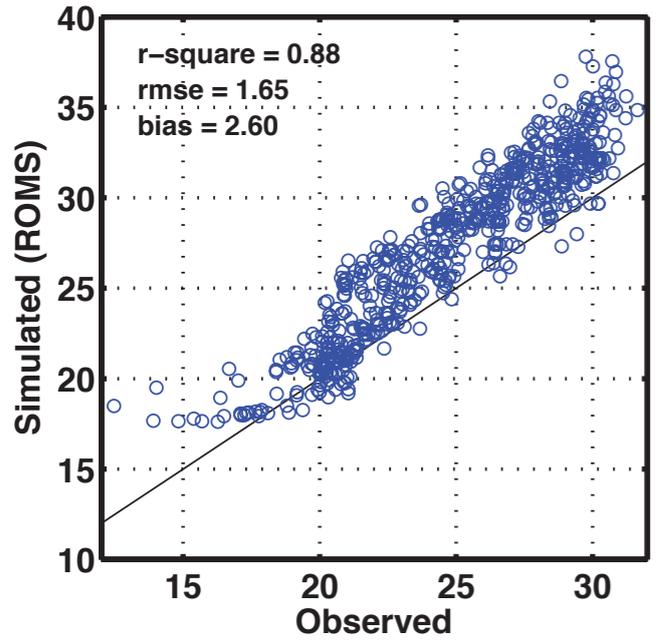
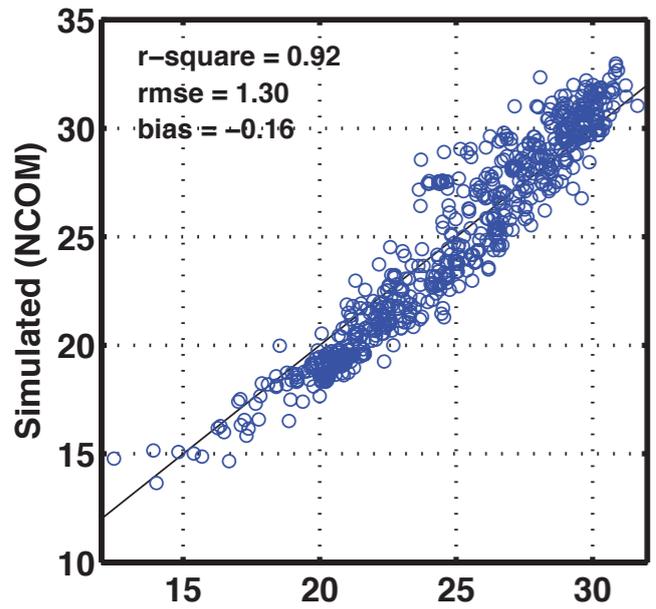
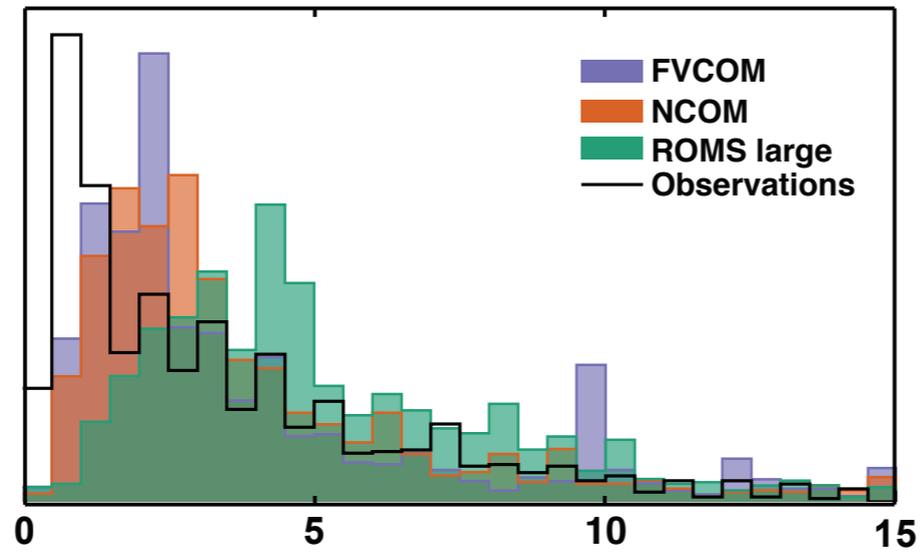
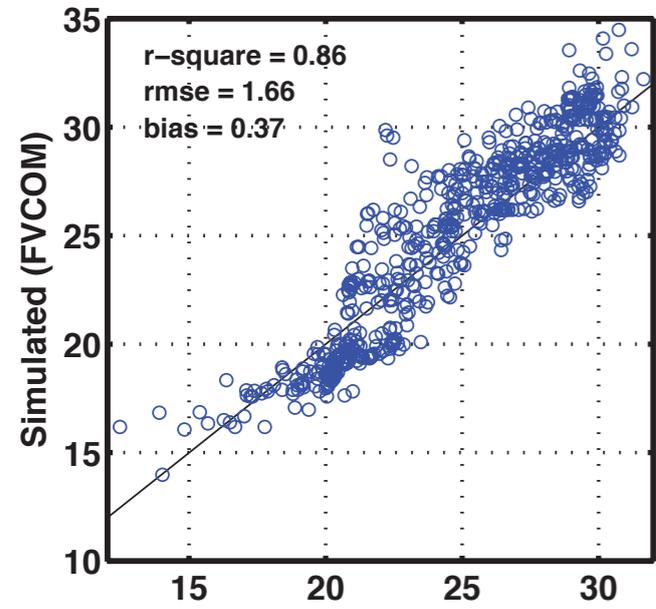
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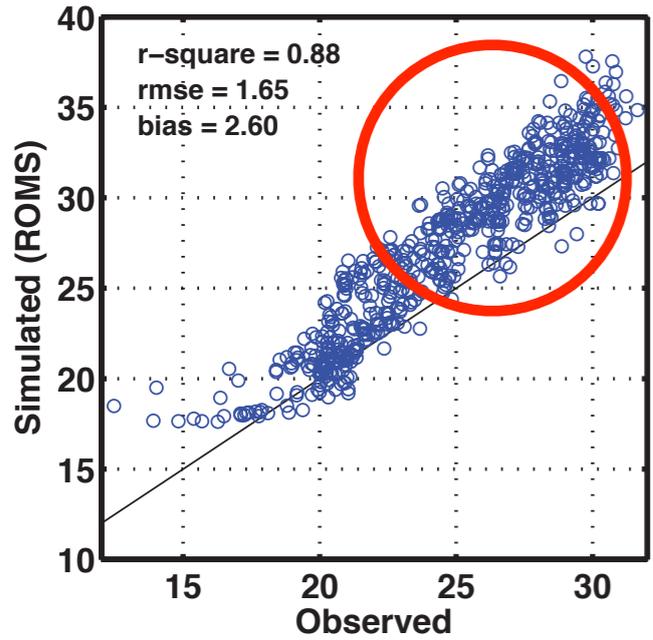
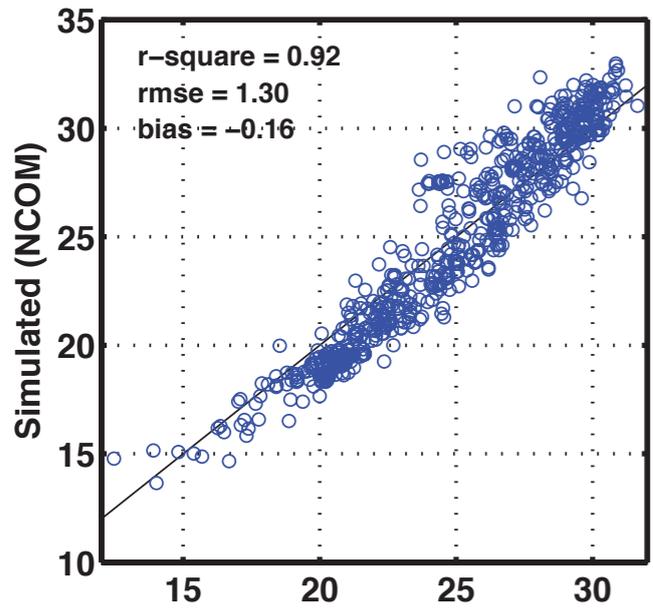
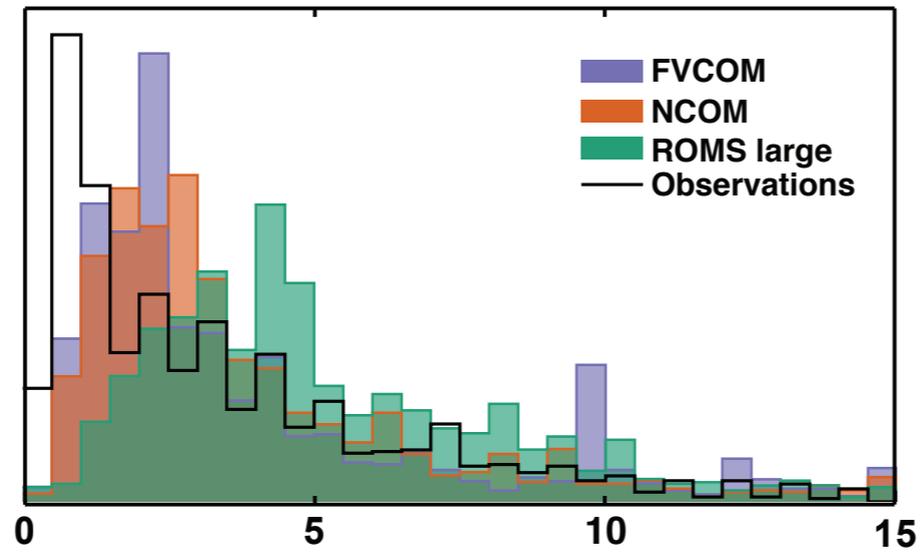
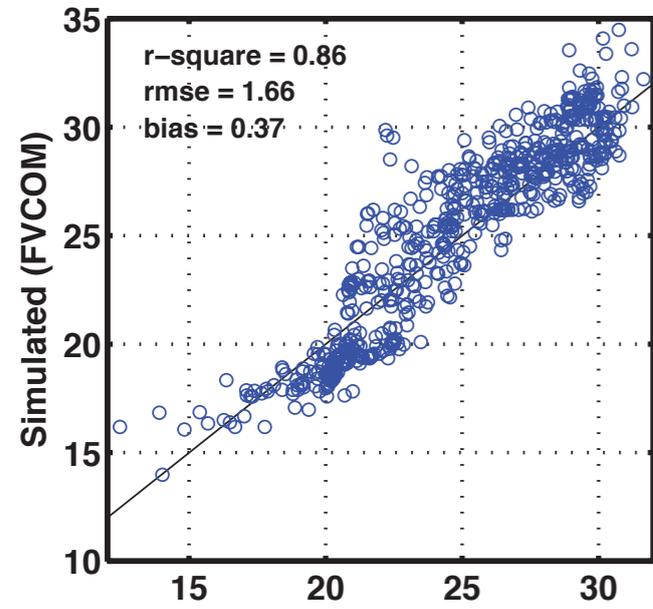
Bottom temperature (°C)



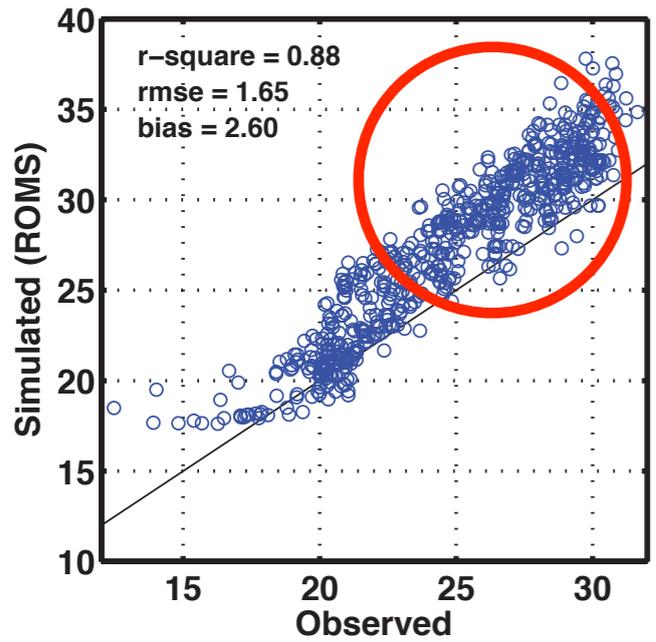
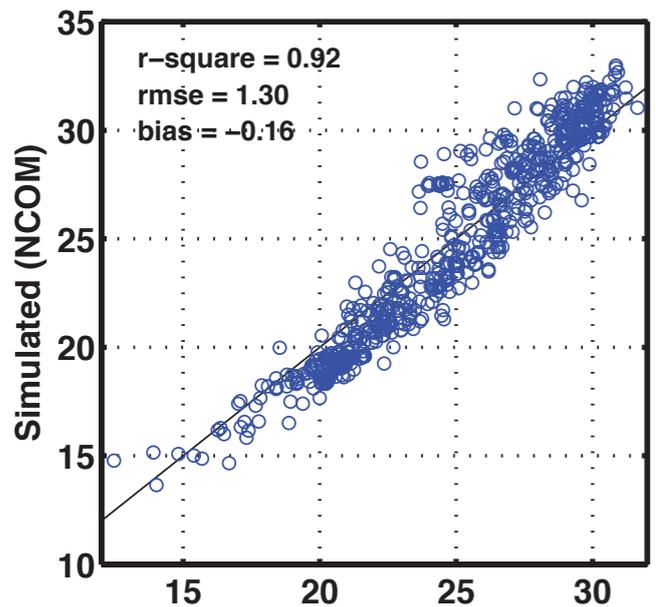
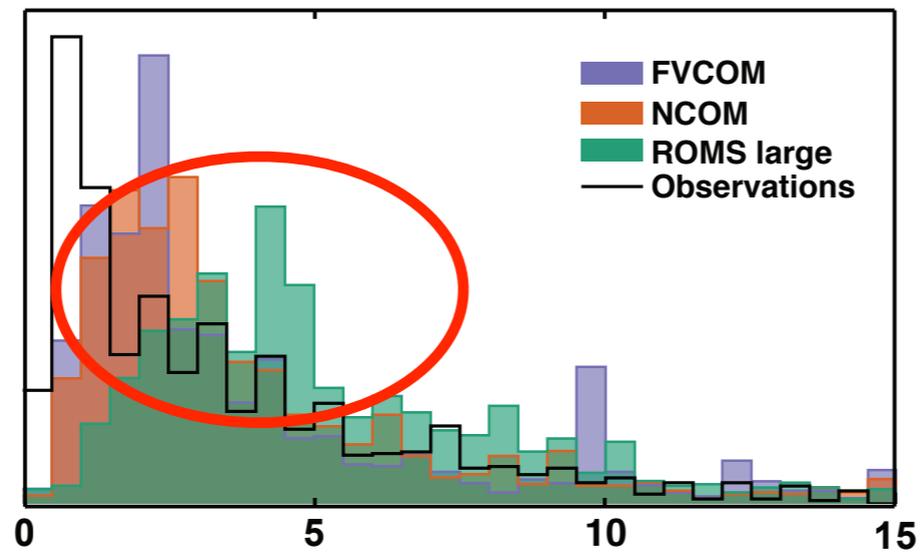
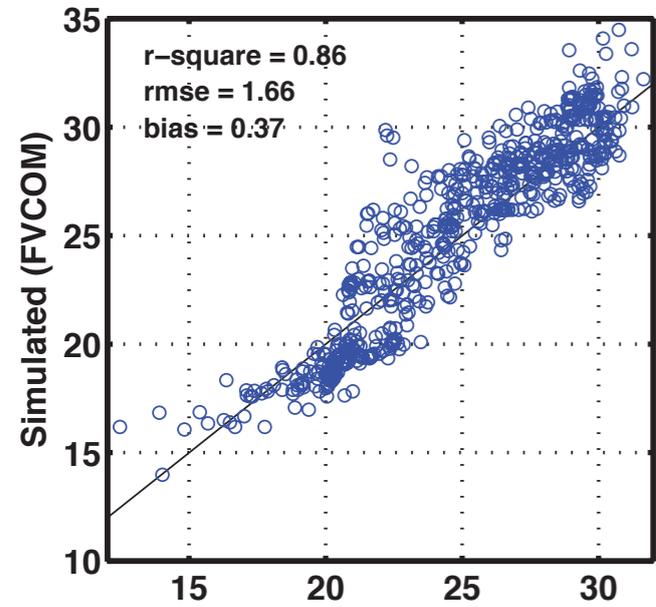
Bottom temperature (°C)



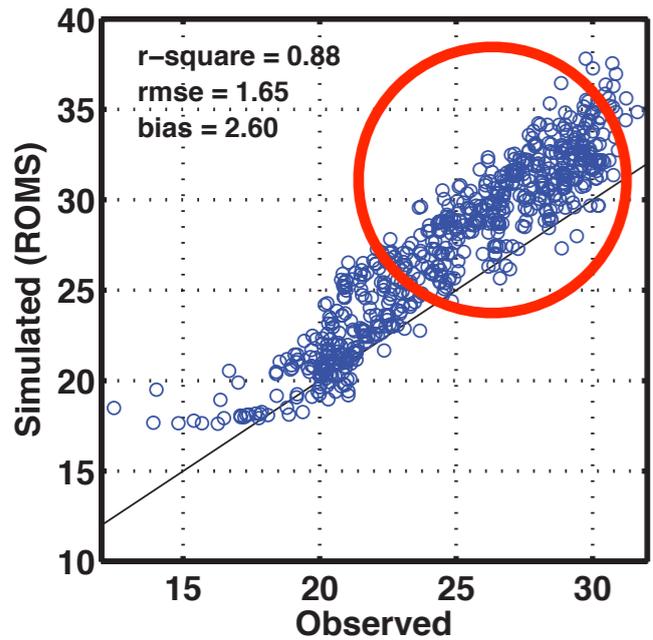
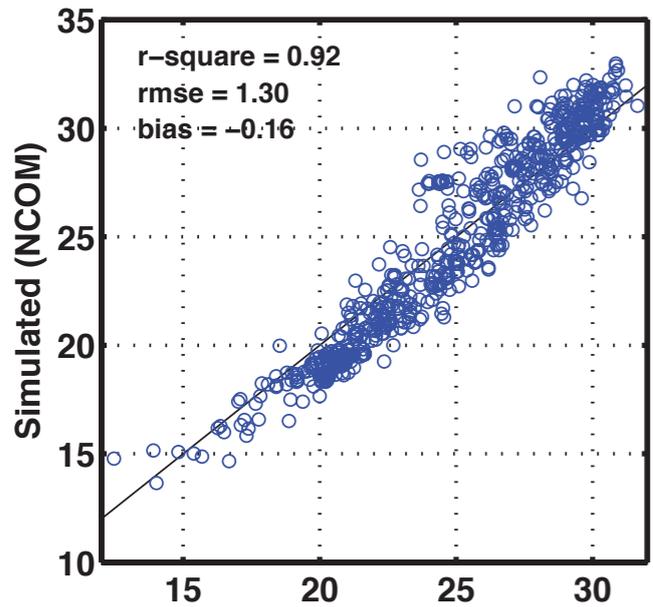
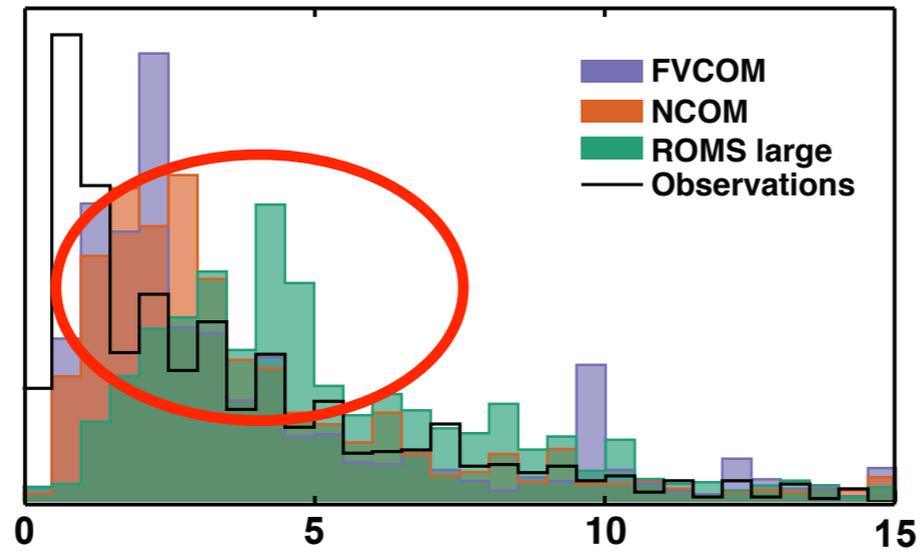
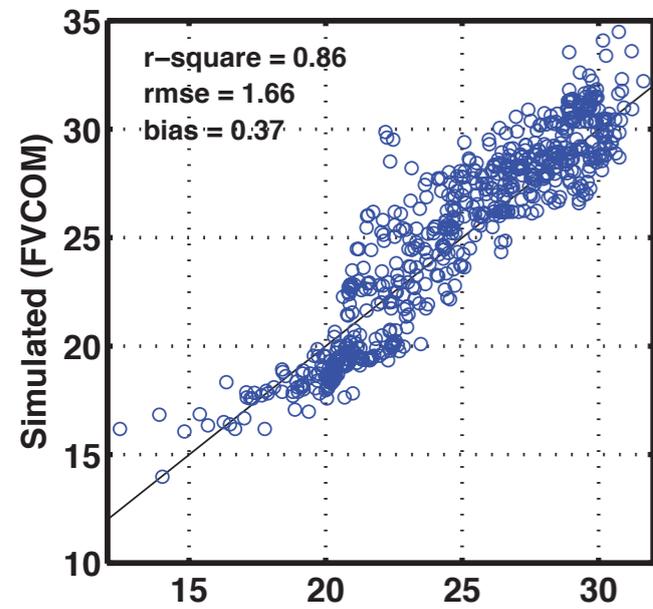
Bottom temperature (°C)



Bottom temperature (°C)

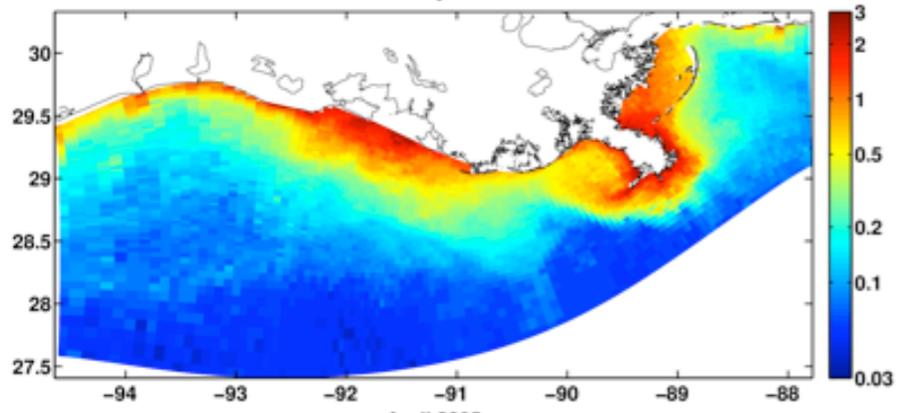


Bottom temperature (°C)

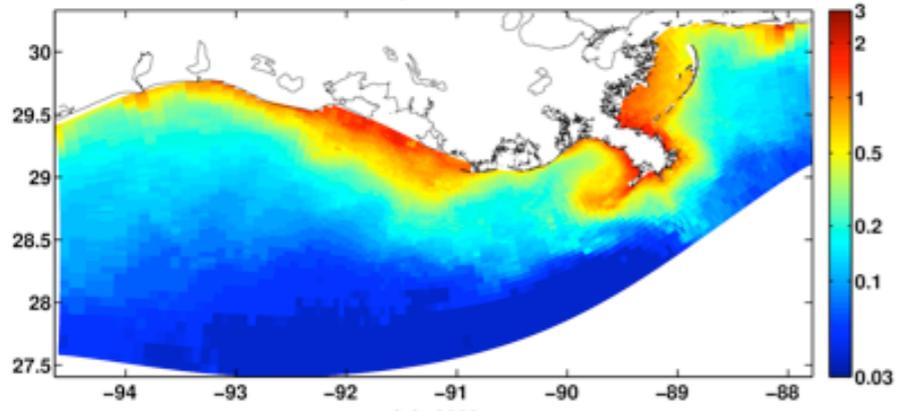


- BBL and hypoxia sensitive to bottom drag
- vertical attenuation of shortwave radiation k_D

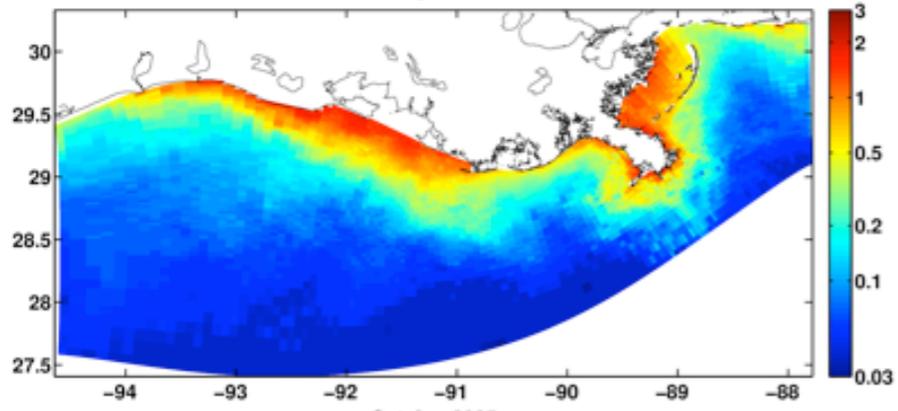
January 2005



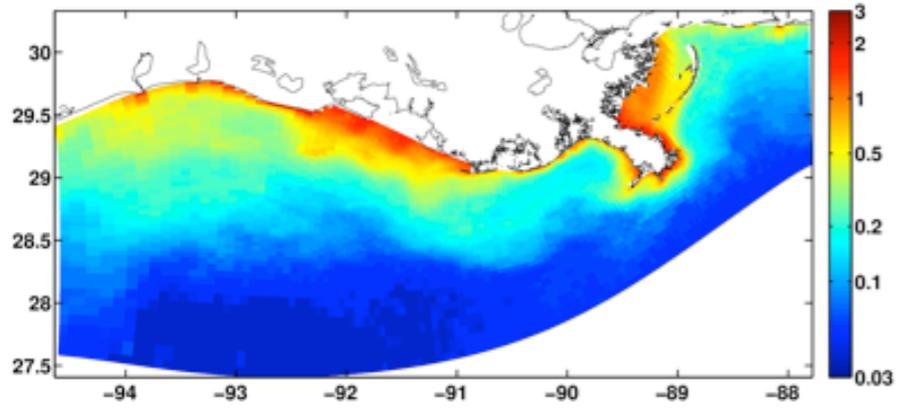
April 2005



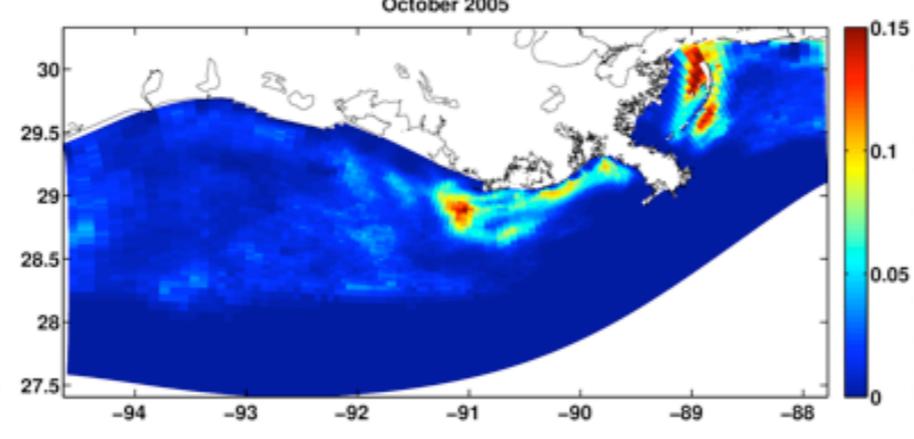
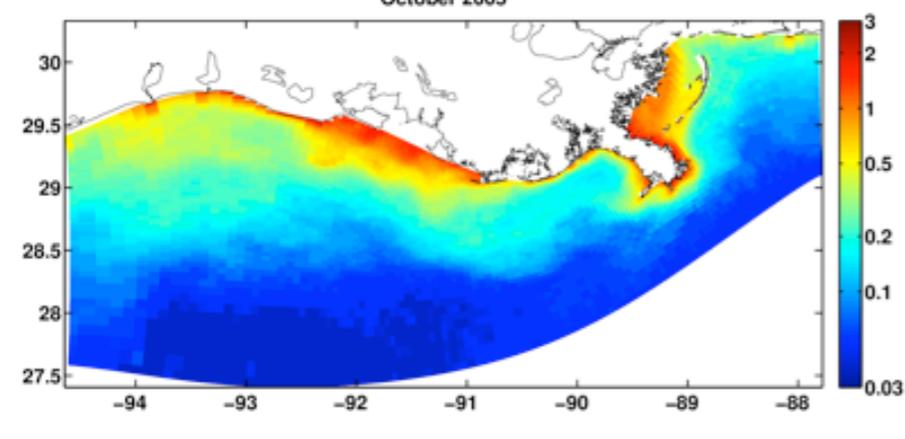
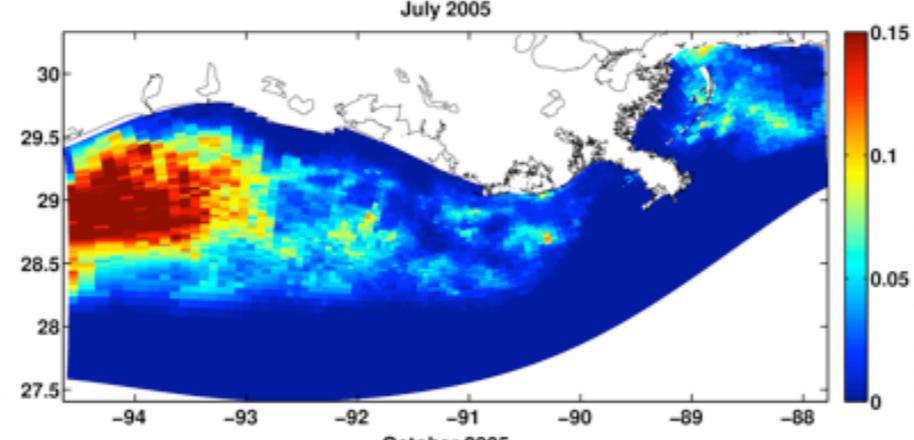
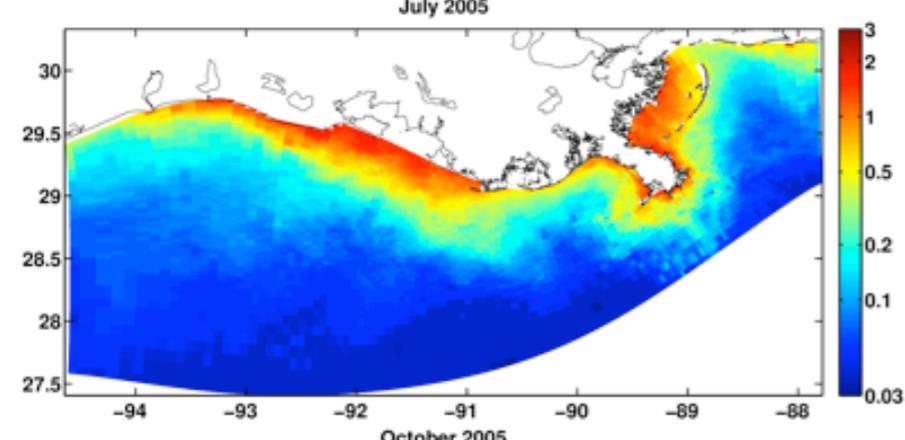
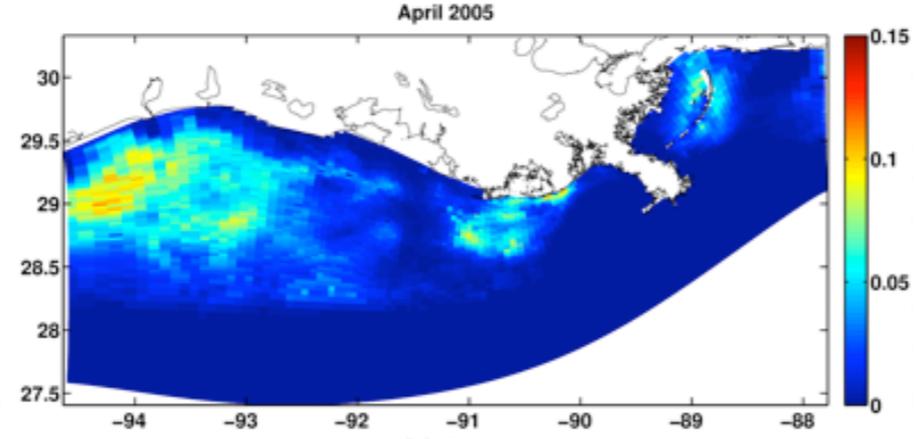
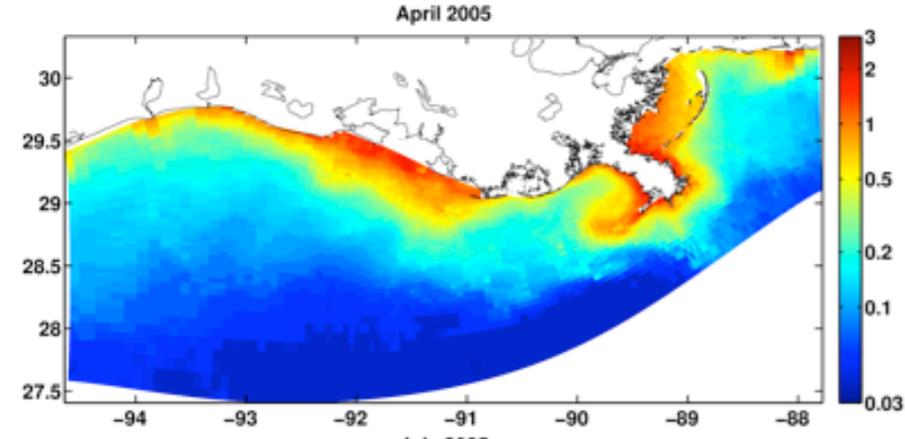
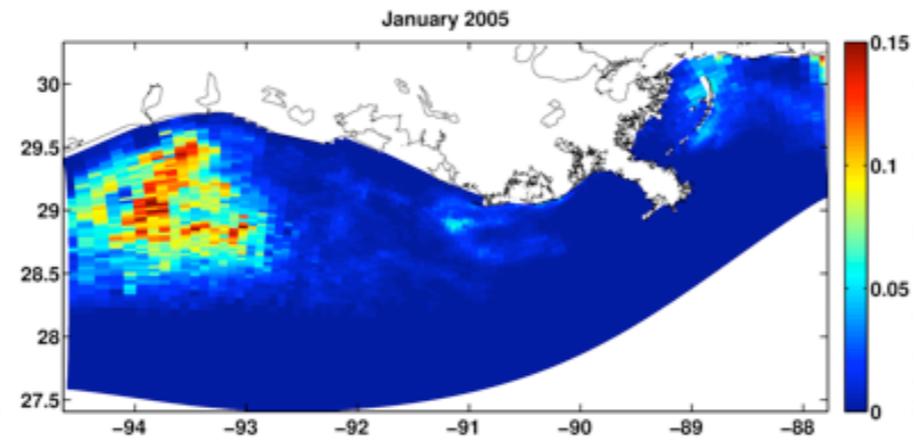
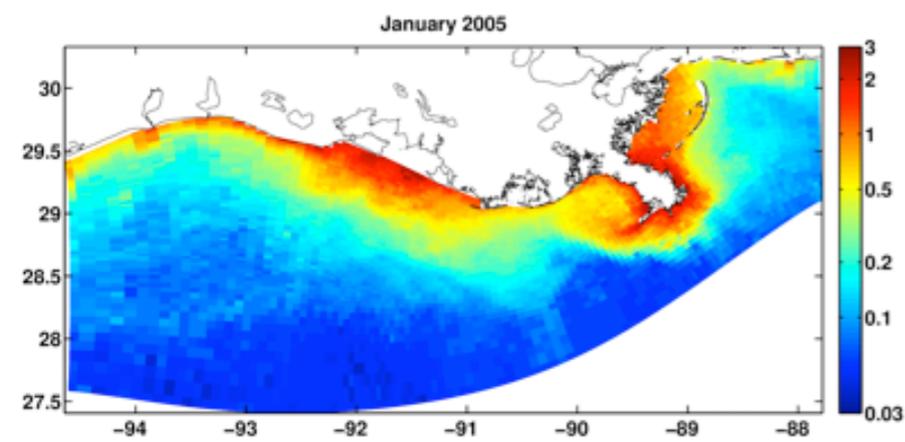
July 2005



October 2005

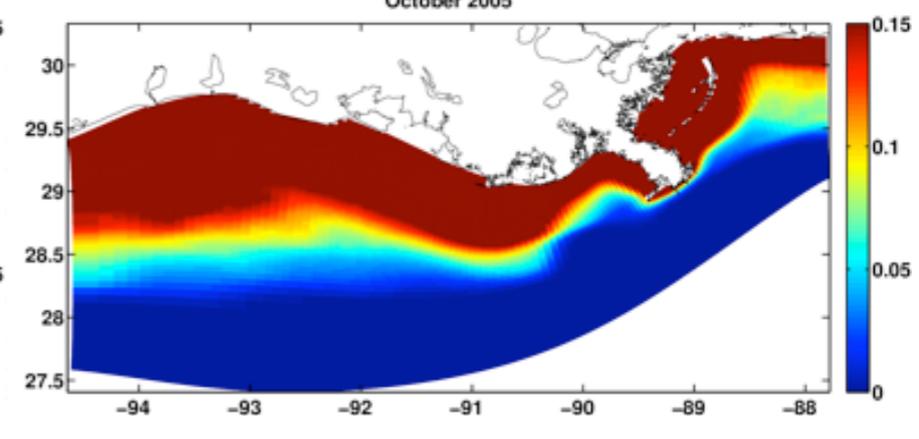
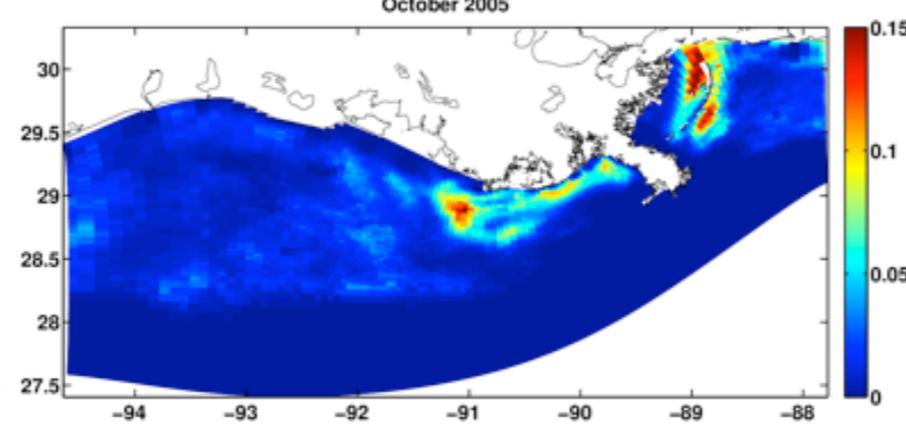
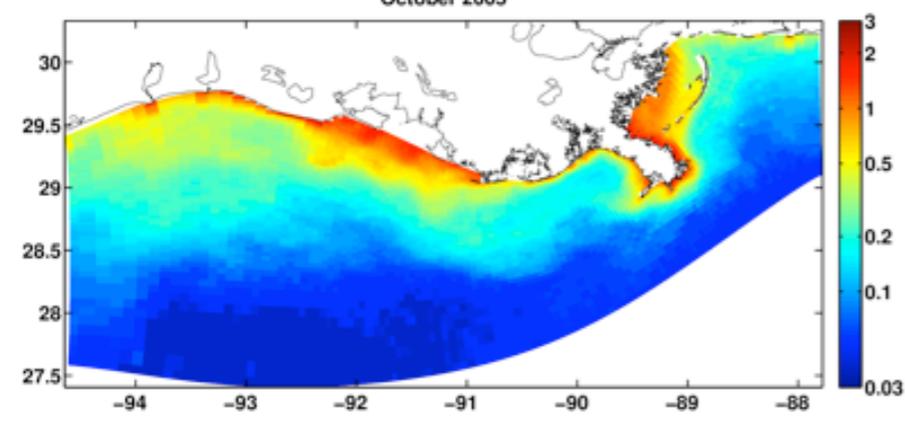
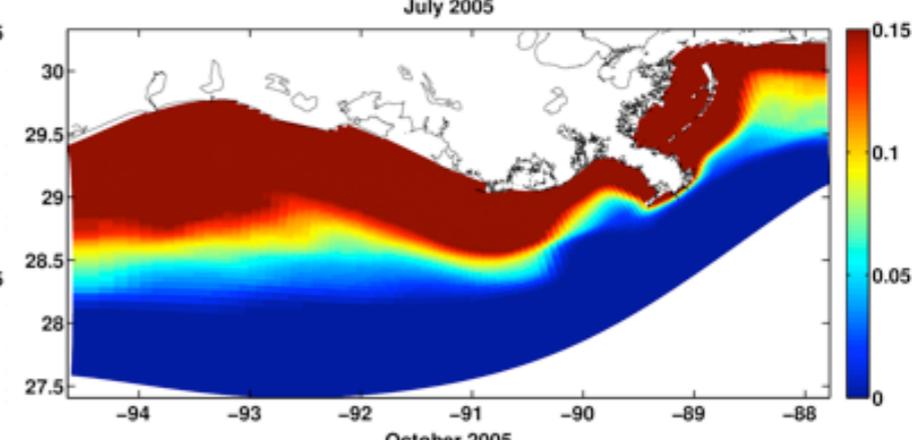
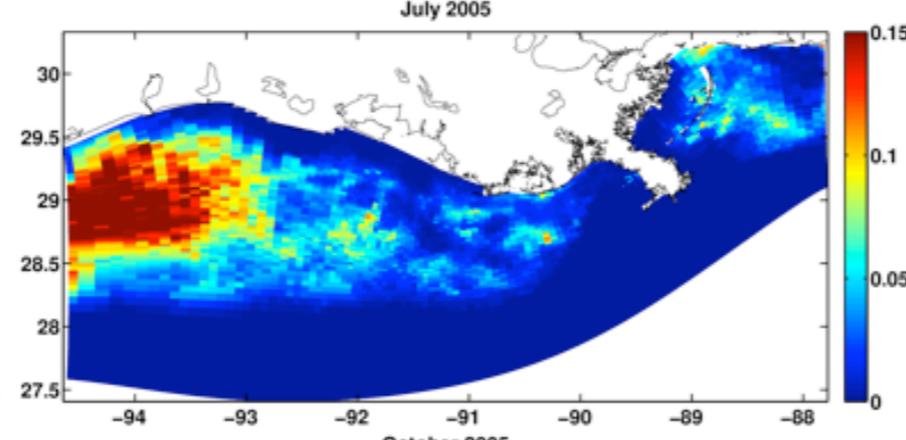
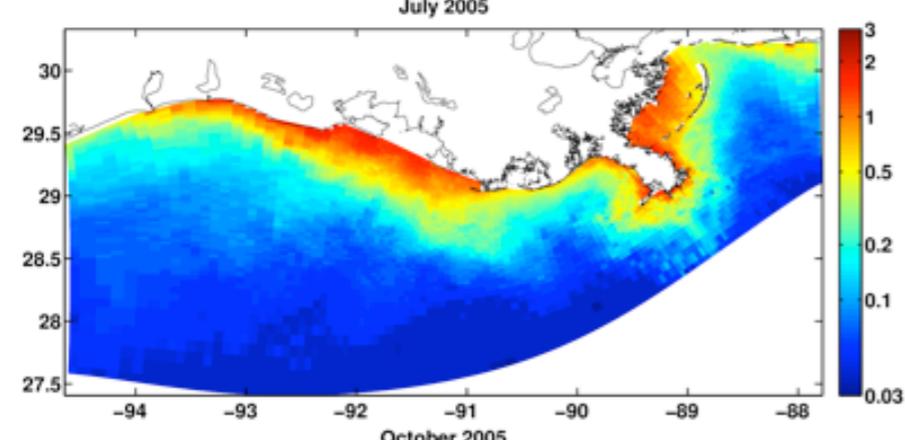
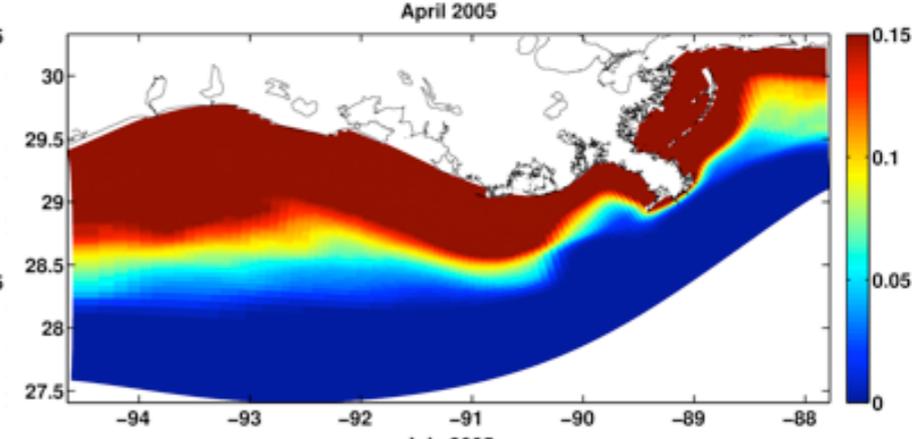
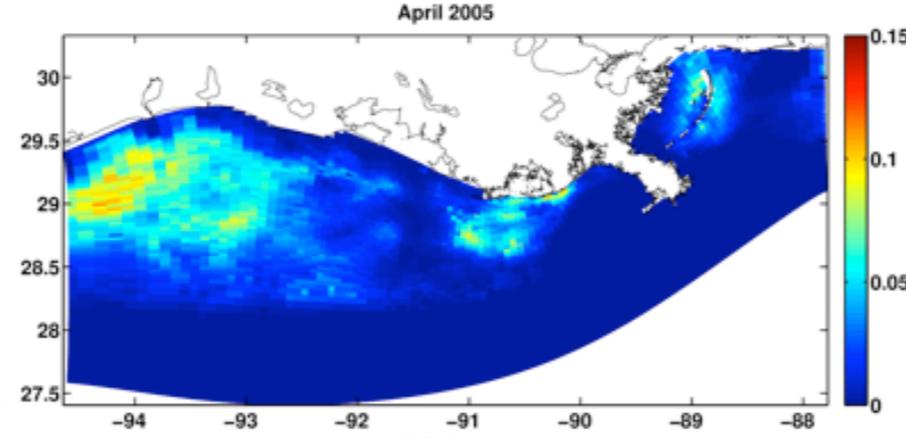
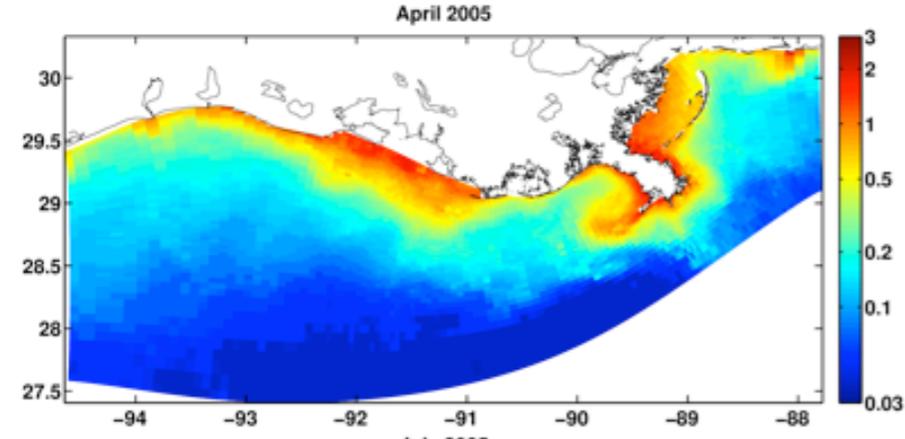
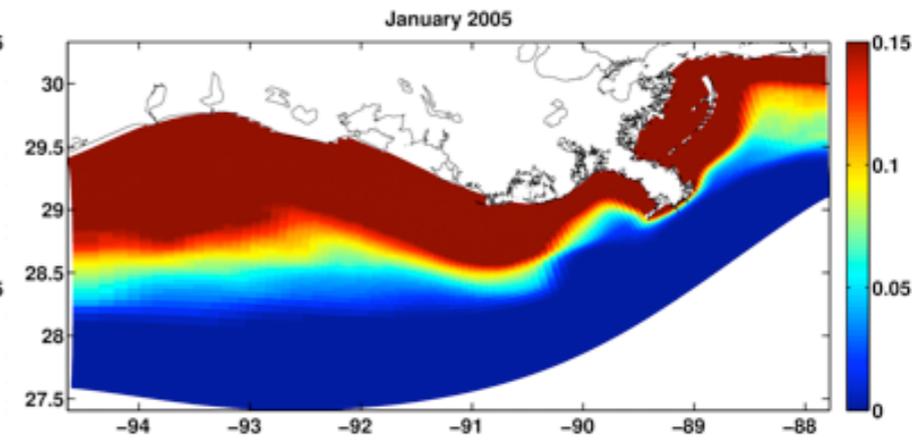
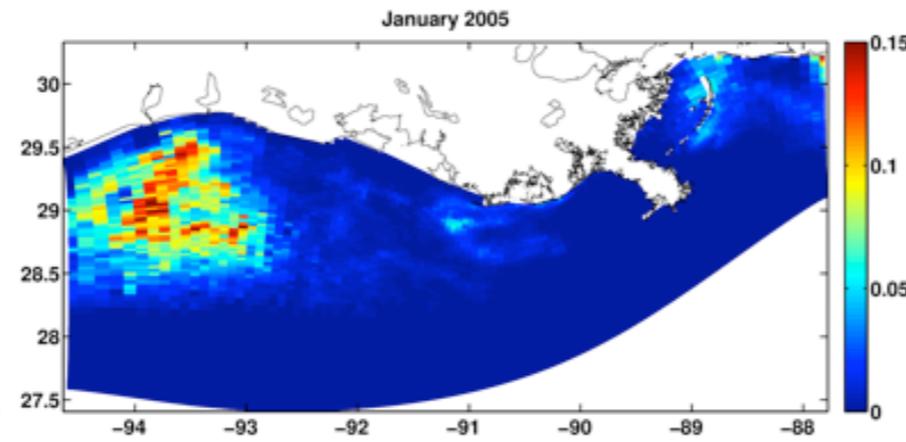
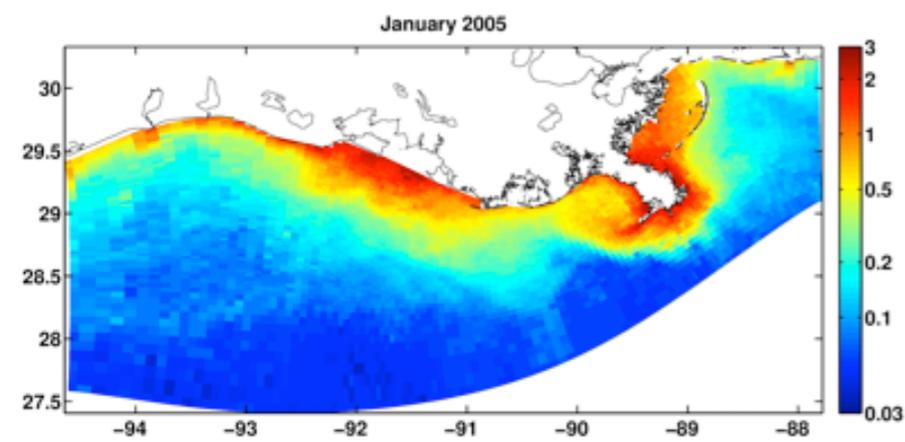


MODIS-derived values of k_D (Schaeffer et al. 2011)



MODIS-derived values of k_D (Schaeffer et al. 2011)

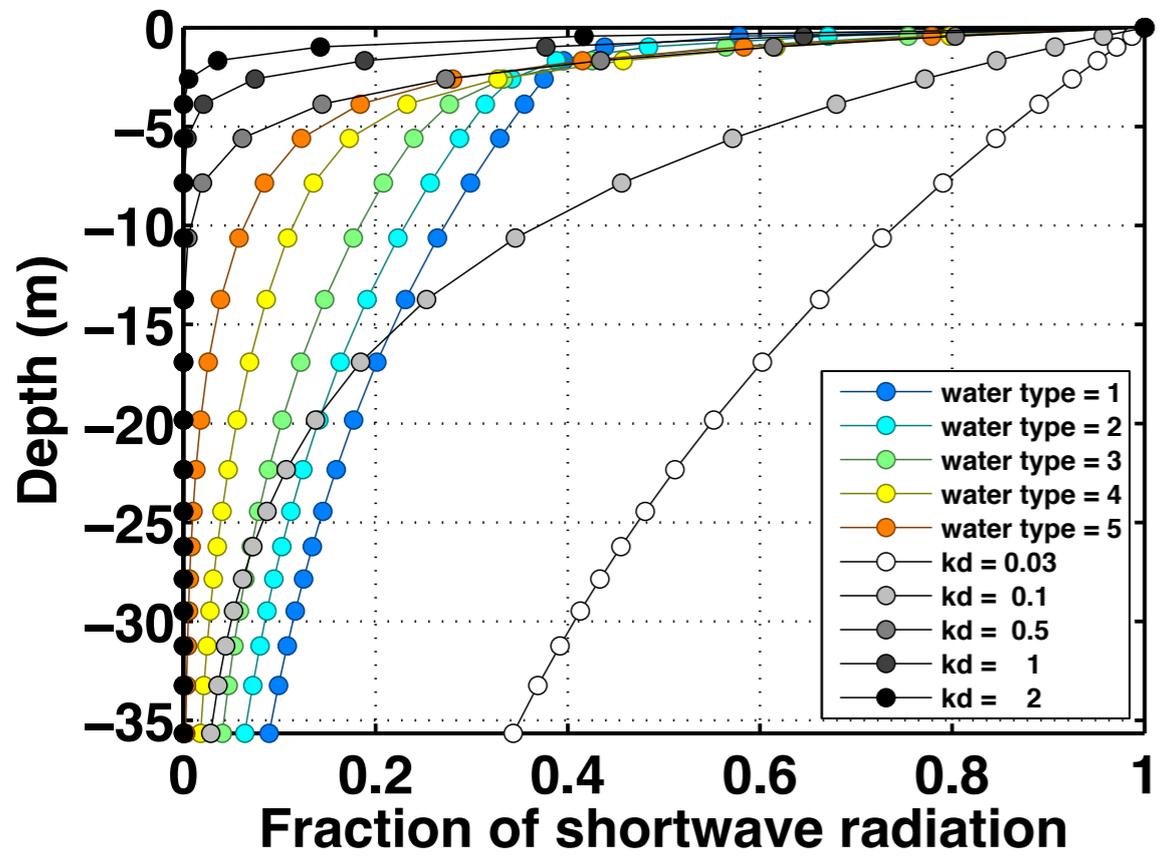
% of SW radiation reaching bottom w/ MODIS k_D

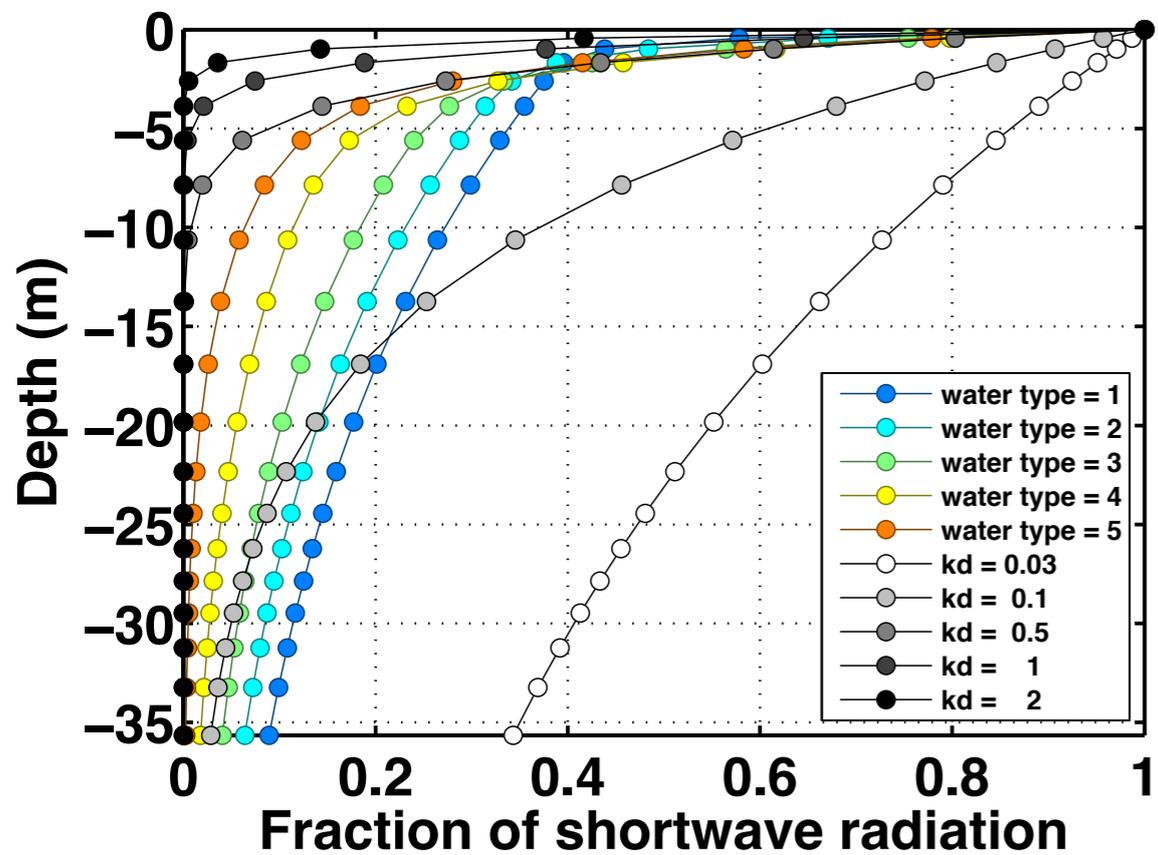


MODIS-derived values of k_D (Schaeffer et al. 2011)

% of SW radiation reaching bottom w/ MODIS k_D

% of SW radiation reaching bottom w/ wtype 1 in ROMS





$$kd = a1 + a2*H + a3*S + a4*T + a5*LON + a6*LAT$$

$$a1 = -0.421962219 \text{ (m}^{-1}\text{)}$$

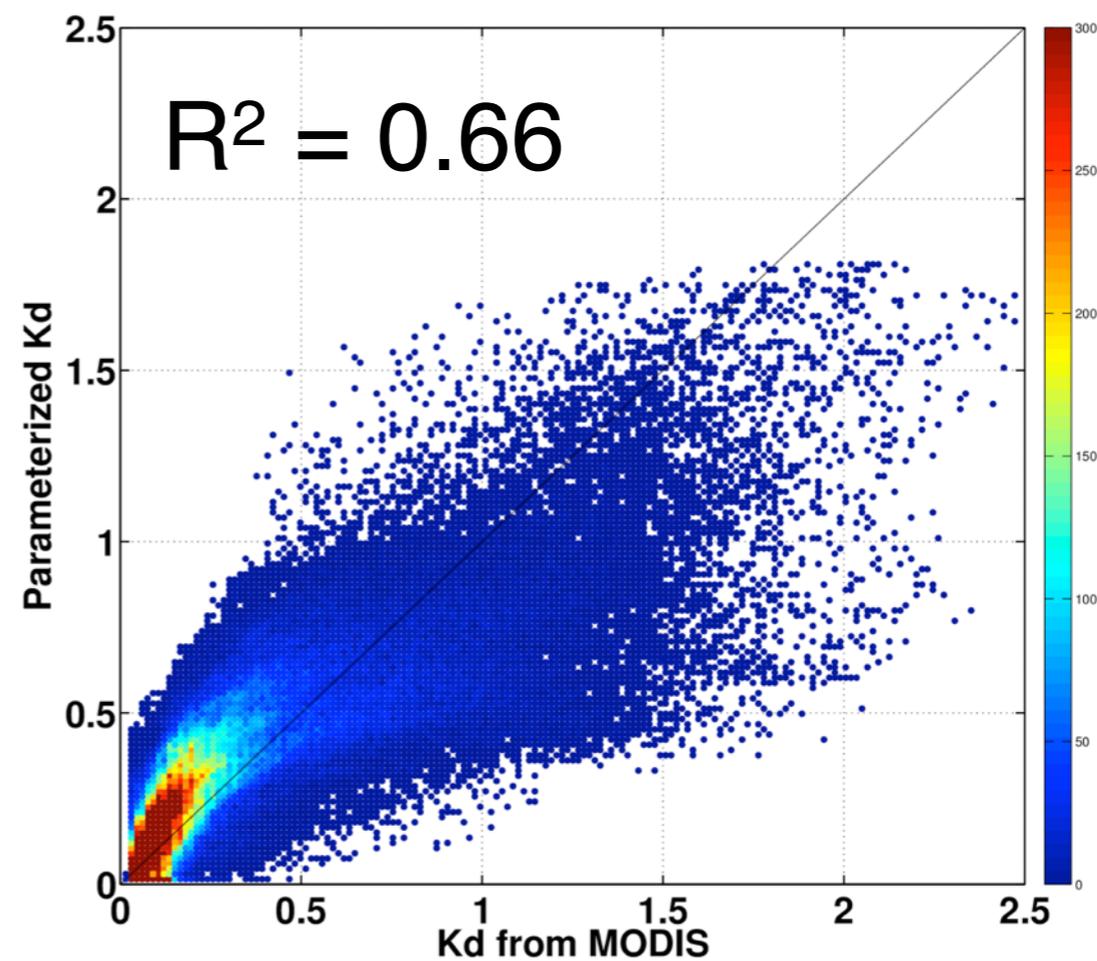
$$a2 = -0.0000338347 \text{ (m)}$$

$$a3 = -0.0478813776$$

$$a4 = -0.0145770602 \text{ (}^{\circ}\text{C)}$$

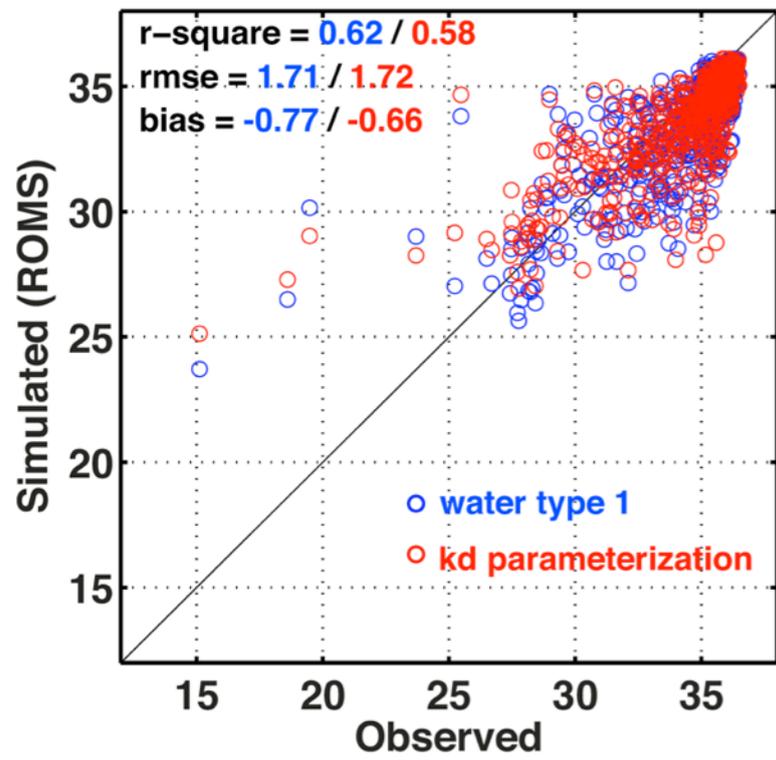
$$a5 = 0.005204748 \text{ (}^{\circ}\text{W, negative)}$$

$$a6 = 0.1058210282 \text{ (}^{\circ}\text{N)}$$

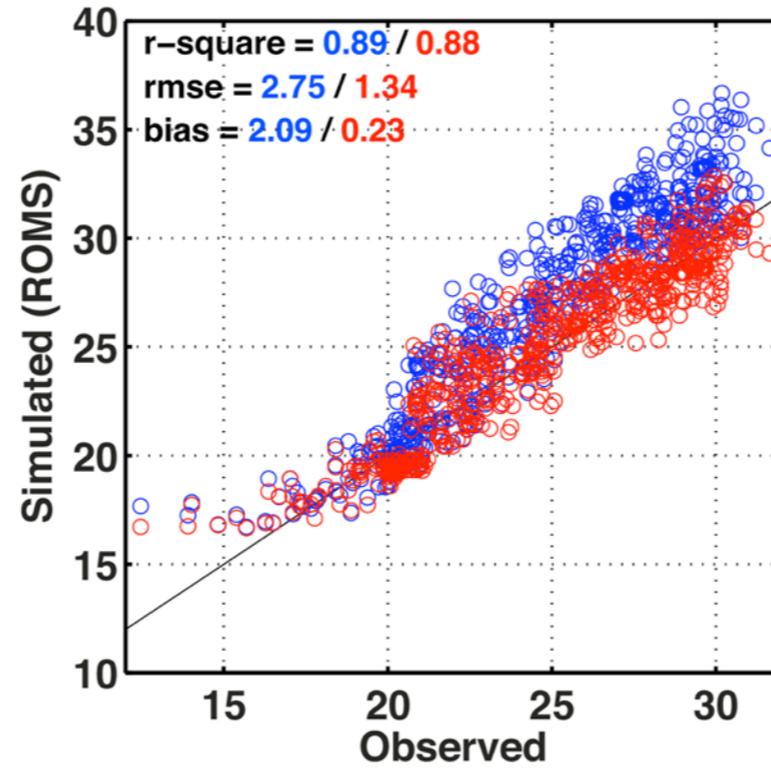


A minimum Kd is set to 0.027 m^{-1} as determined by Smith and Baker (1978) for clear ocean waters.

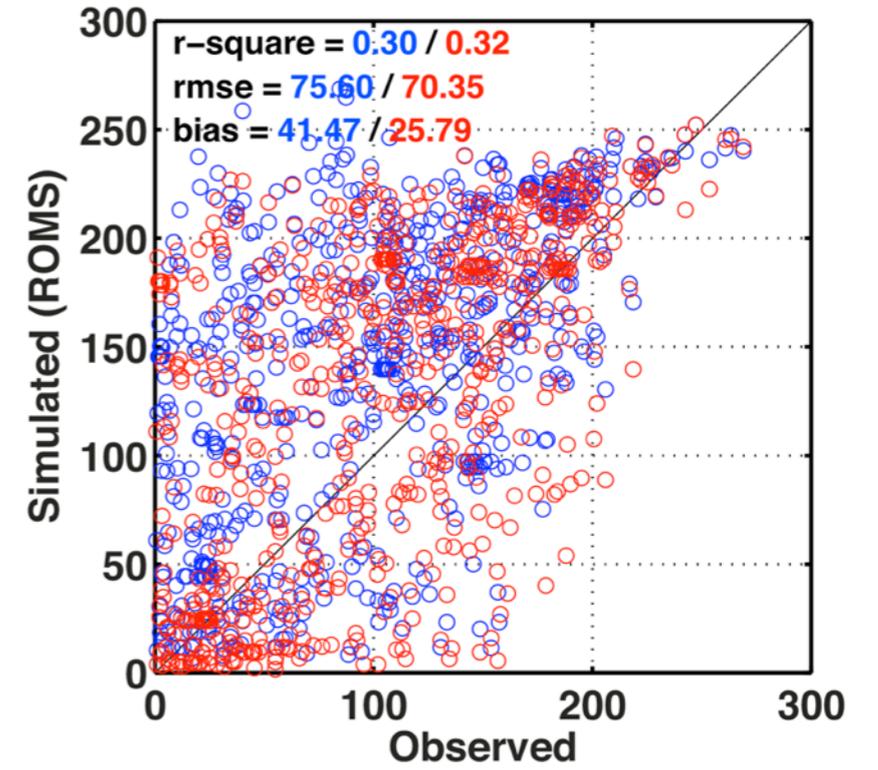
Bottom salinity



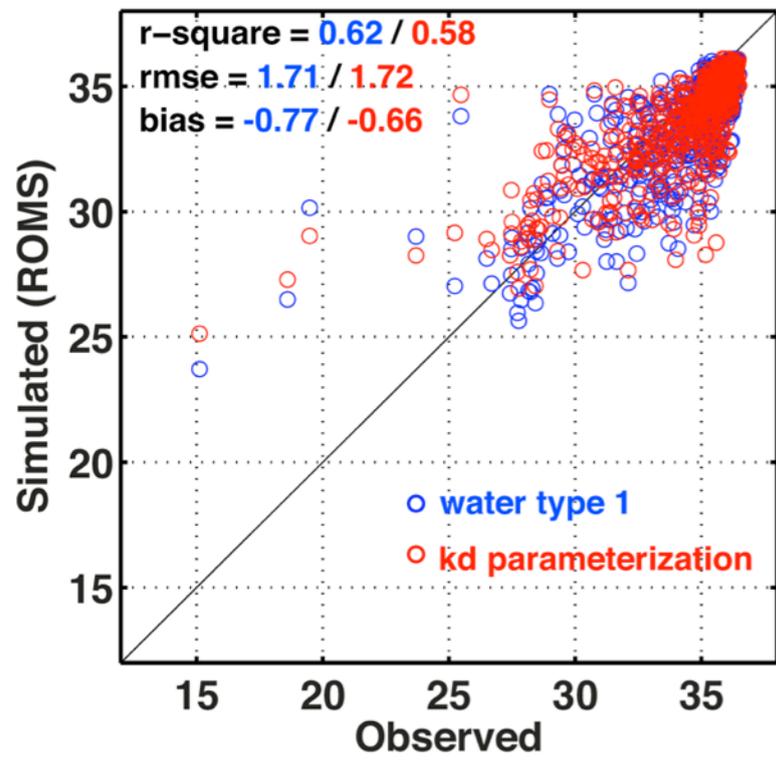
Bottom temperature (°C)



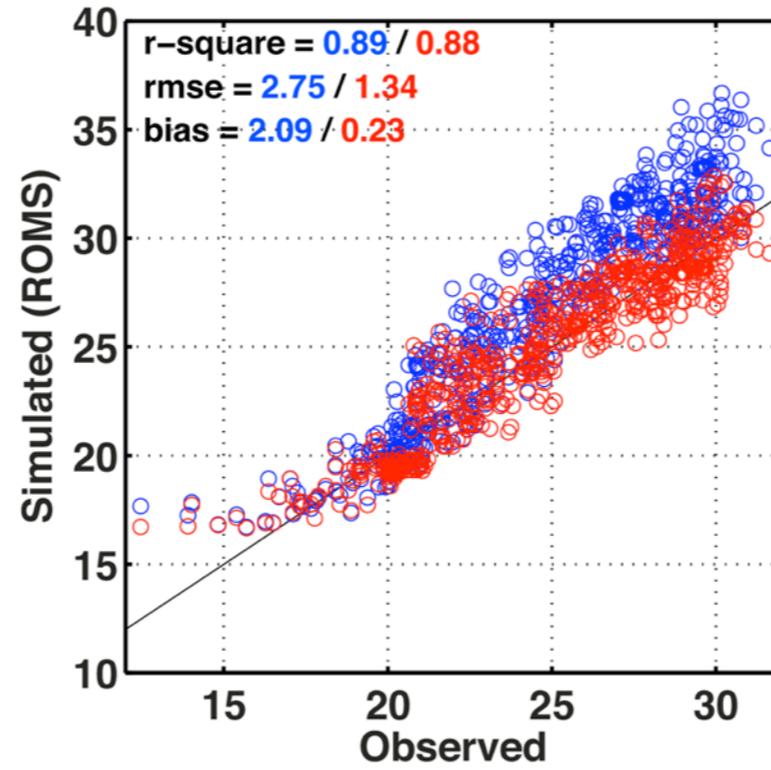
Bottom oxygen (mmol O2 m⁻³)



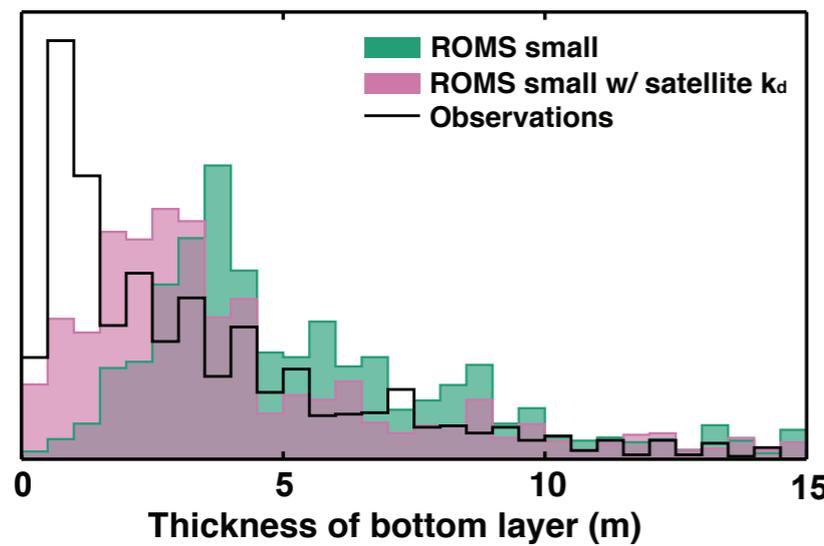
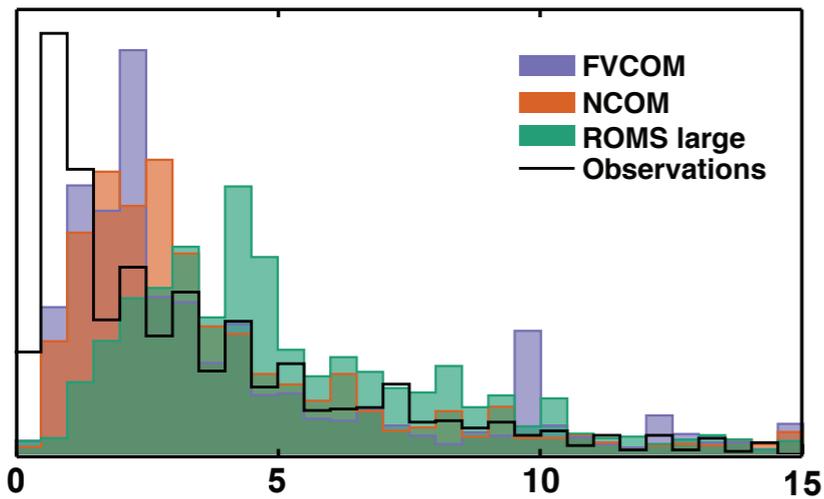
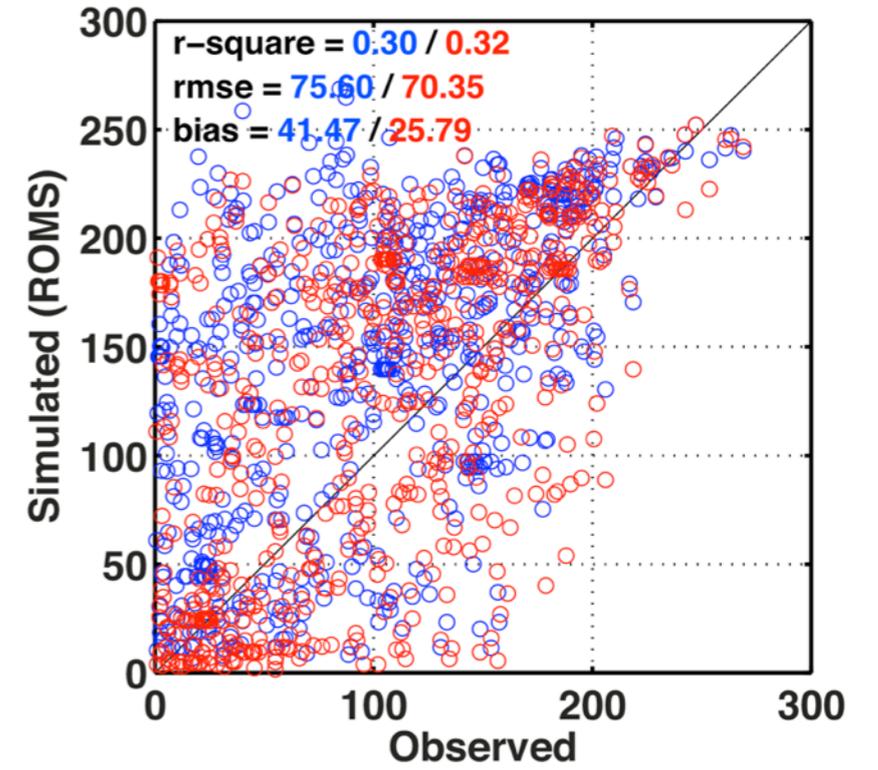
Bottom salinity

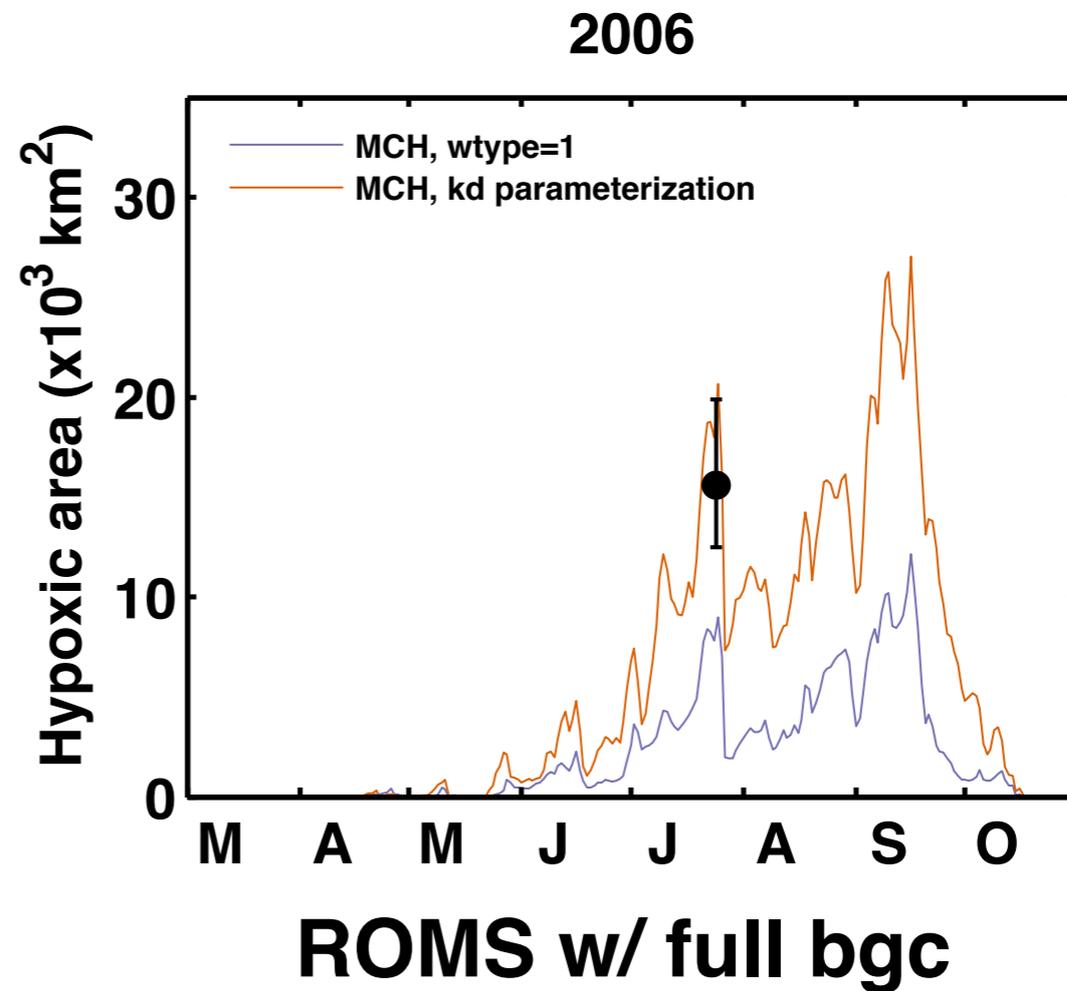
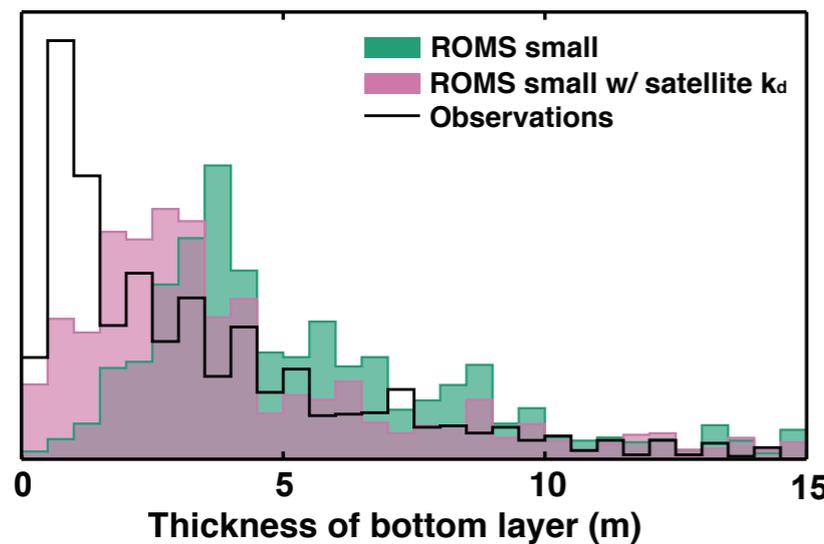
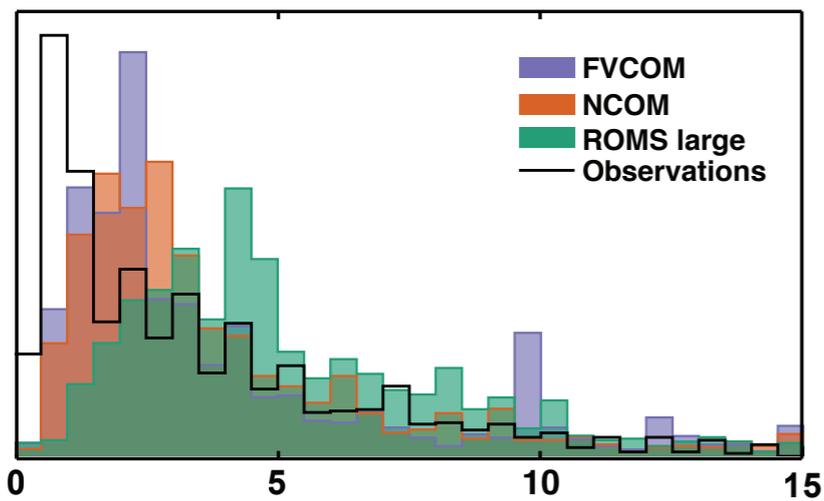
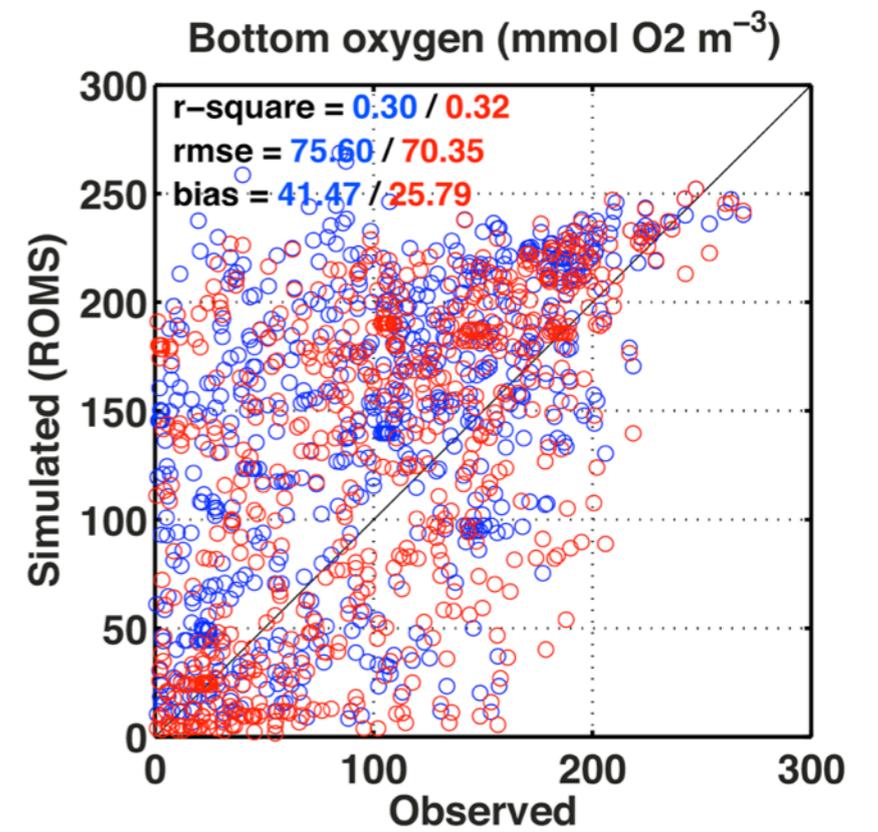
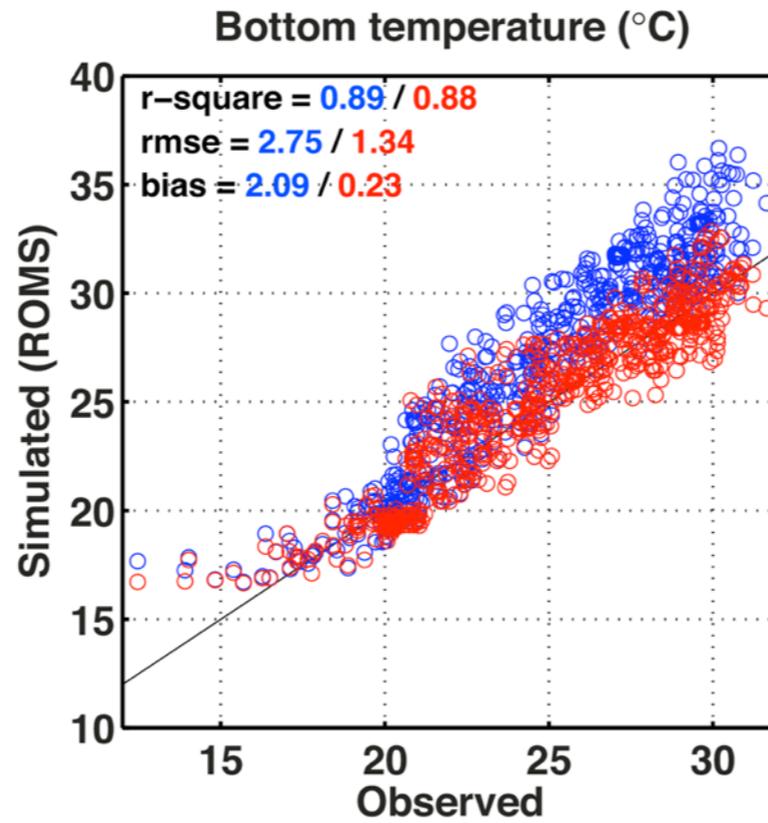
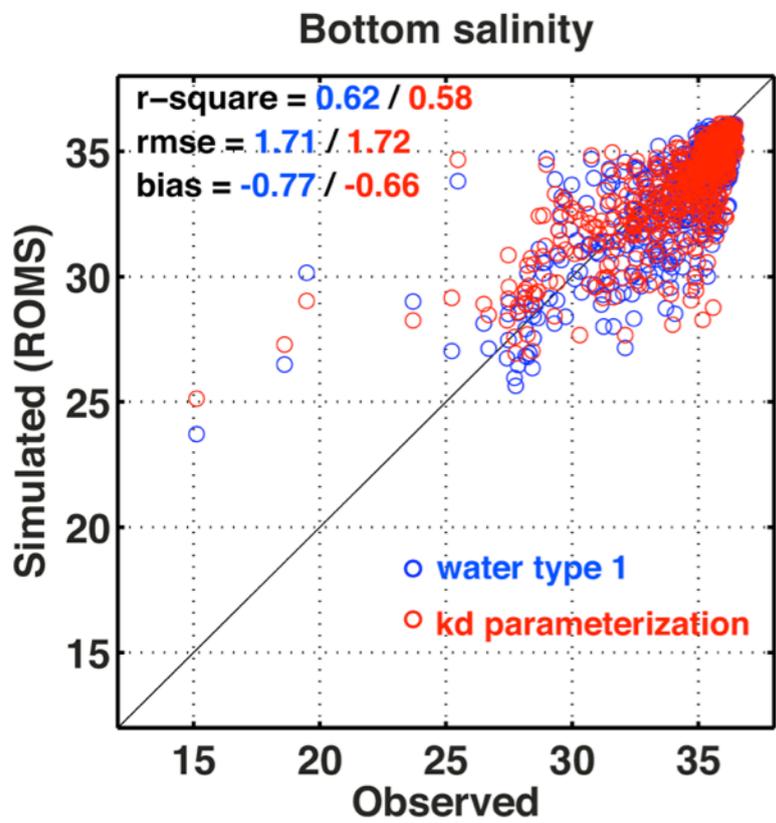


Bottom temperature (°C)



Bottom oxygen (mmol O2 m⁻³)





Find the odd one out!

oxygen
consumption
in the water
column

vertical
stratification

oxygen
consumption by
the sediment

bottom drag
parameter

vertical
attenuation
of shortwave
radiation

Find the odd one out!

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radiation

Journal of Geophysical Research: Oceans

RESEARCH ARTICLE

10.1002/2015JC011577

Key Points:

- Model intercomparison of three hypoxia models of the northern Gulf of Mexico is presented
- Bottom water temperature and bottom boundary layer thickness are important for hypoxia simulation
- Overall stratification strength does not explain model-to-model differences in hypoxic conditions

Supporting Information:

- Supporting Information S1

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Katja.Fennel@dal.ca

Citation:

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Effects of model physics on hypoxia simulations for the northern Gulf of Mexico: A model intercomparison

Katja Fennel¹, Arnaud Laurent¹, Robert Hetland², Dubravko Justić³, Dong S. Ko⁴, John Lehrter⁵, Michael Murrell⁵, Lixia Wang³, Liuqian Yu¹, and Wenxia Zhang^{1,2}

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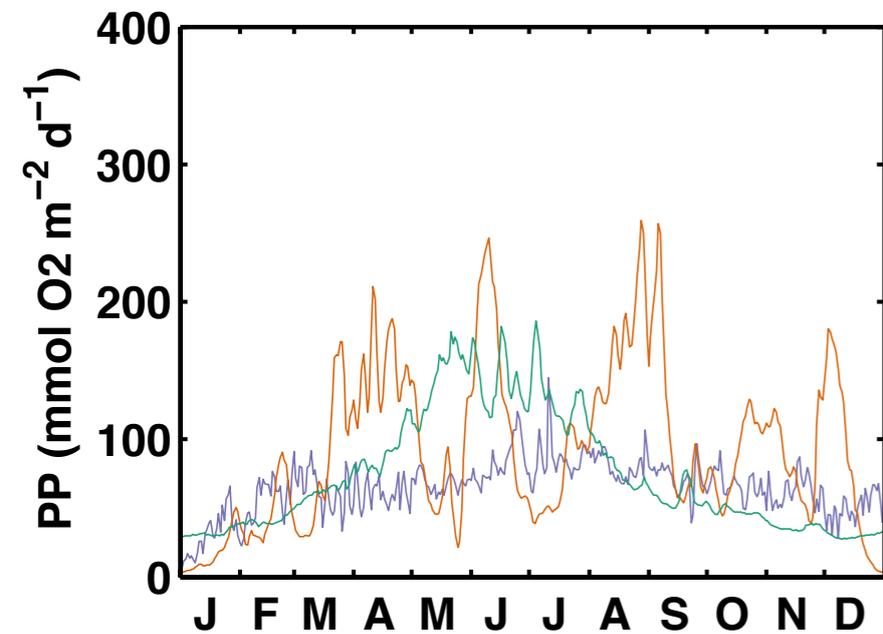
Abstract A large hypoxic zone forms every summer on the Texas-Louisiana Shelf in the northern Gulf of Mexico due to nutrient and freshwater inputs from the Mississippi/Atchafalaya River System. Efforts are underway to reduce the extent of hypoxic conditions through reductions in river nutrient inputs, but the response of hypoxia to such nutrient load reductions is difficult to predict because biological responses are confounded by variability in physical processes. The objective of this study is to identify the major physical model aspects that matter for hypoxia simulation and prediction. In order to do so, we compare three different circulation models (ROMS, FVCOM, and NCOM) implemented for the northern Gulf of Mexico, all coupled to the same simple oxygen model, with observations and against each other. By using a highly simplified oxygen model, we eliminate the potentially confounding effects of a full biogeochemical model and can isolate the effects of physical features. In a systematic assessment, we found that (1) model-to-model differences in bottom water temperatures result in differences in simulated hypoxia because temperature influences the uptake rate of oxygen by the sediments (an important oxygen sink in this system), (2) vertical stratification does not explain model-to-model differences in hypoxic conditions in a straightforward way, and (3) the thickness of the bottom boundary layer, which sets the thickness of the hypoxic layer in all three models, is key to determining the likelihood of a model to generate hypoxic conditions. These results imply that hypoxic area, the commonly used metric in the northern Gulf which ignores hypoxic layer thickness, is insufficient for assessing a model's ability to accurately simulate hypoxia, and that hypoxic volume needs to be considered as well.

Recommendation: For short-term hypoxia forecasts, a well-calibrated simple oxygen model coupled to excellent physical model is the strategy.

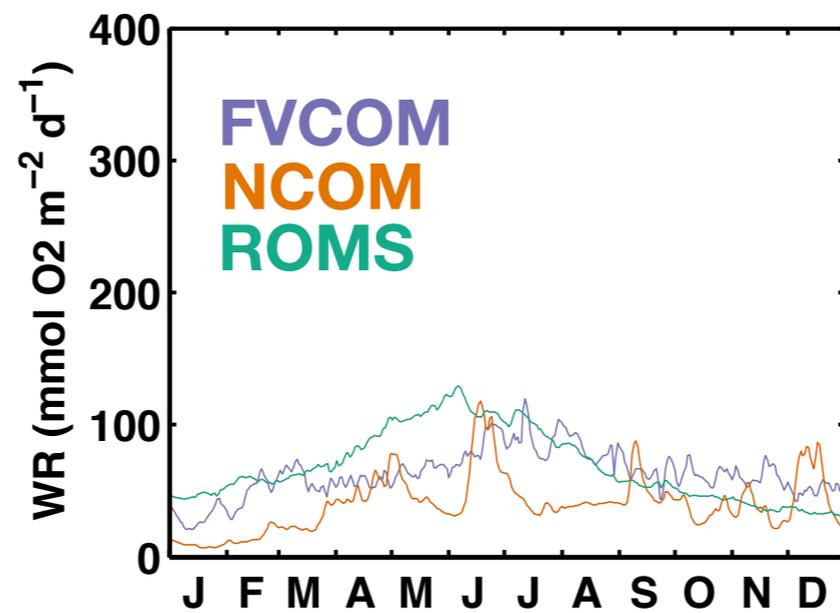
Recommendation: For short-term hypoxia forecasts, a well-calibrated simple oxygen model coupled to excellent physical model is the strategy.

But for nutrient load reduction scenarios a full biogeochemical model is needed.

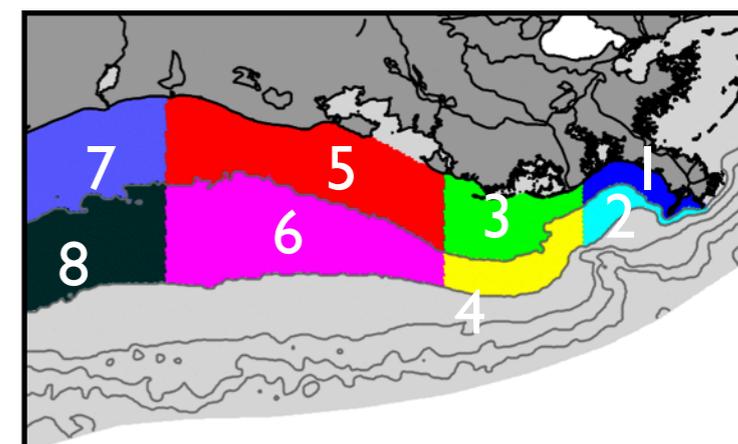
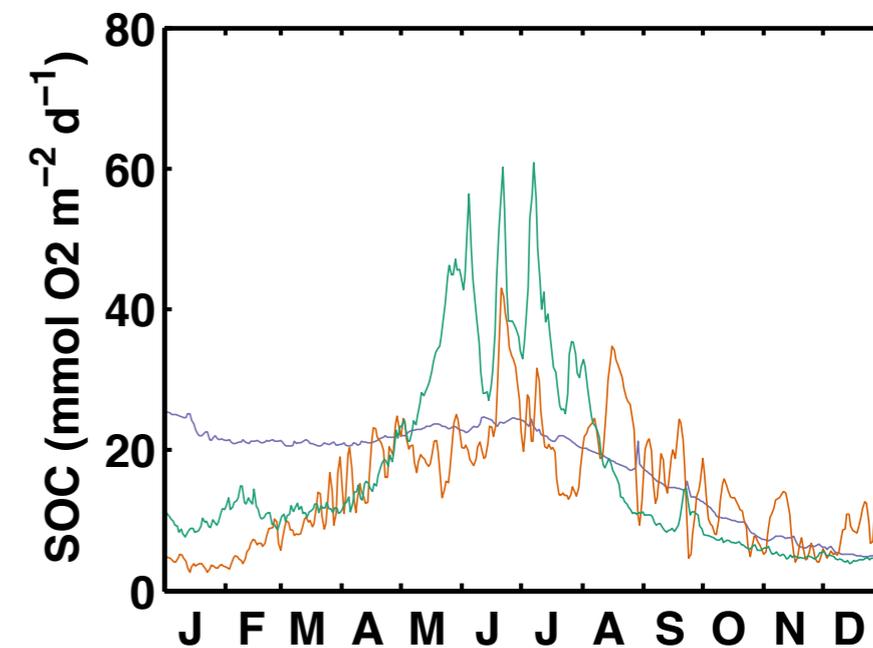
Zone 5 – 2005



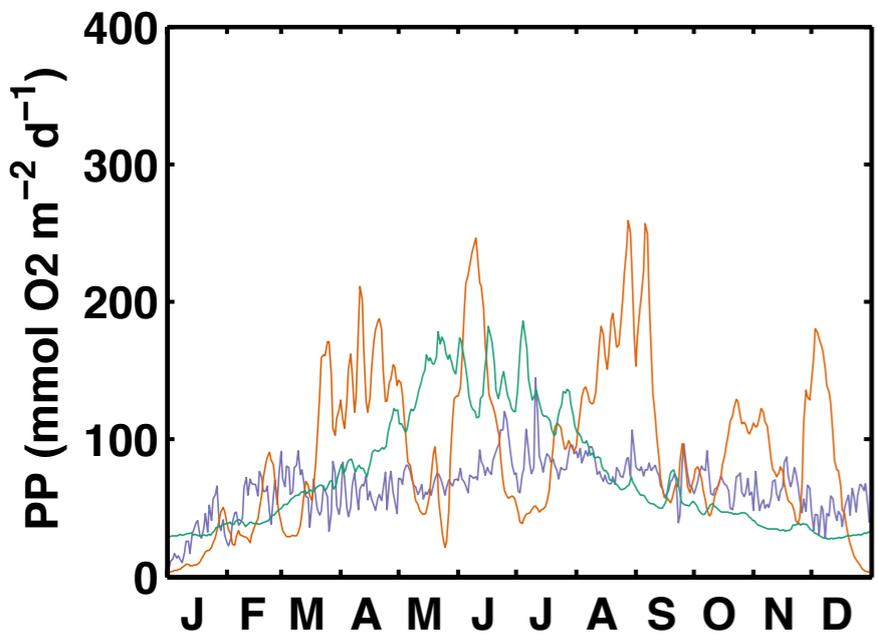
Zone 5 – 2005



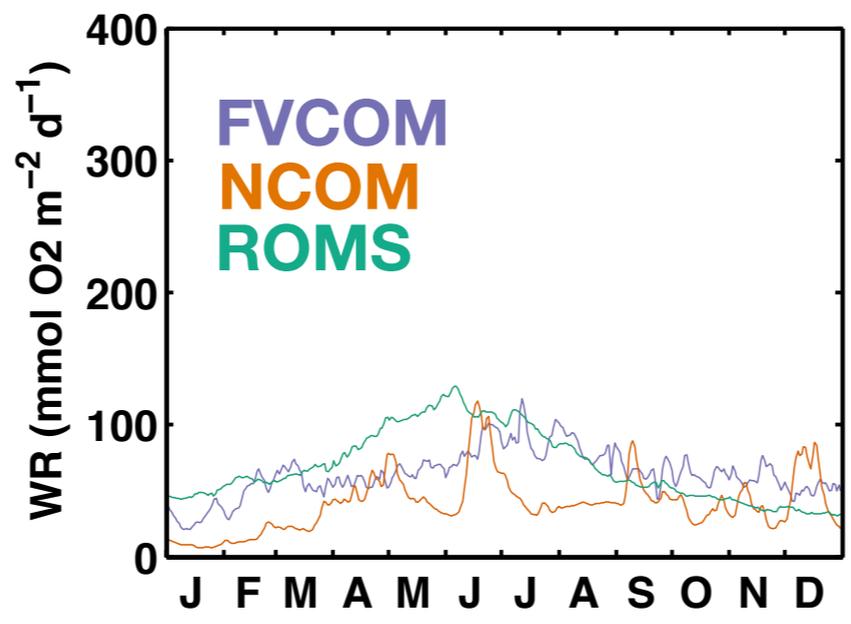
Zone 5 – 2005



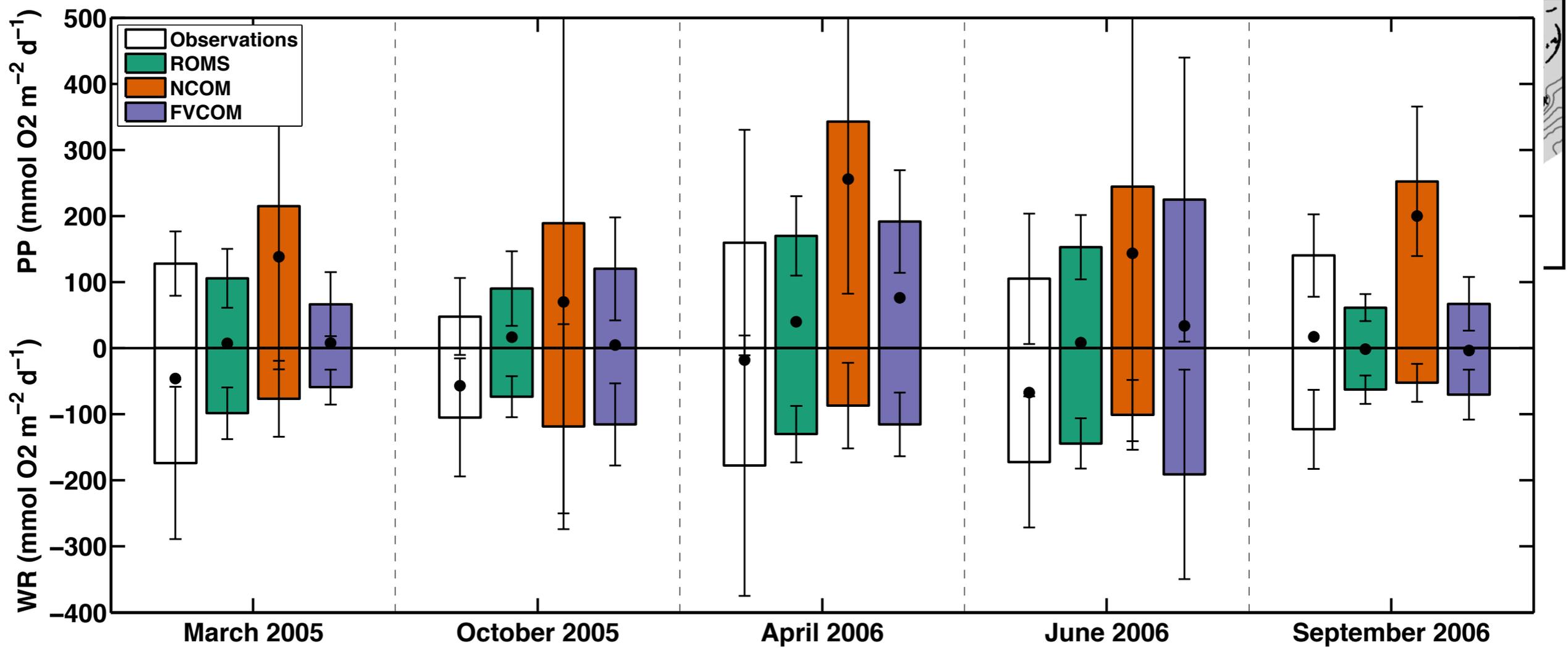
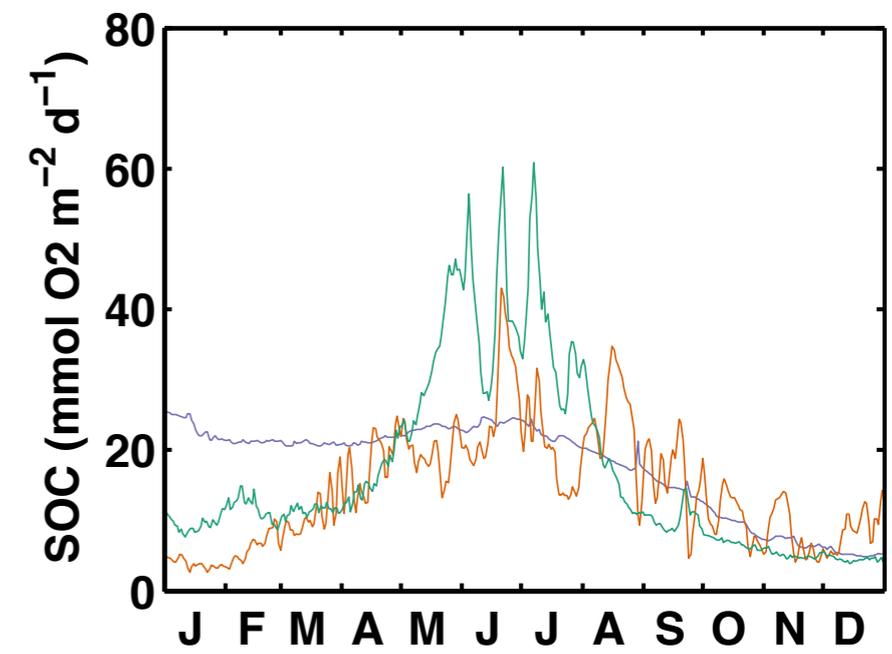
Zone 5 - 2005



Zone 5 - 2005



Zone 5 - 2005



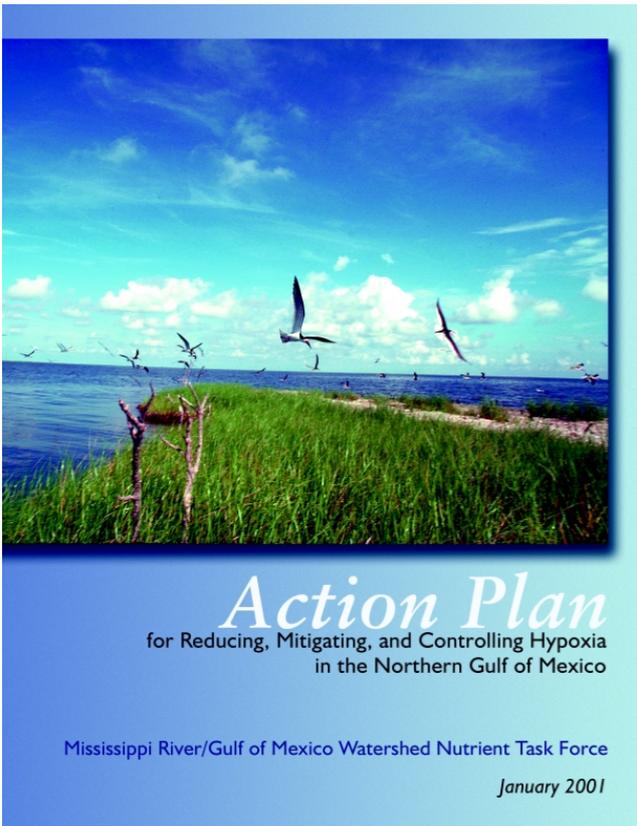
**Goal: Multi-model estimates of
necessary nutrient load reductions**

How far do N and/or P loads have to be reduced in order to affect desired reductions in hypoxia?

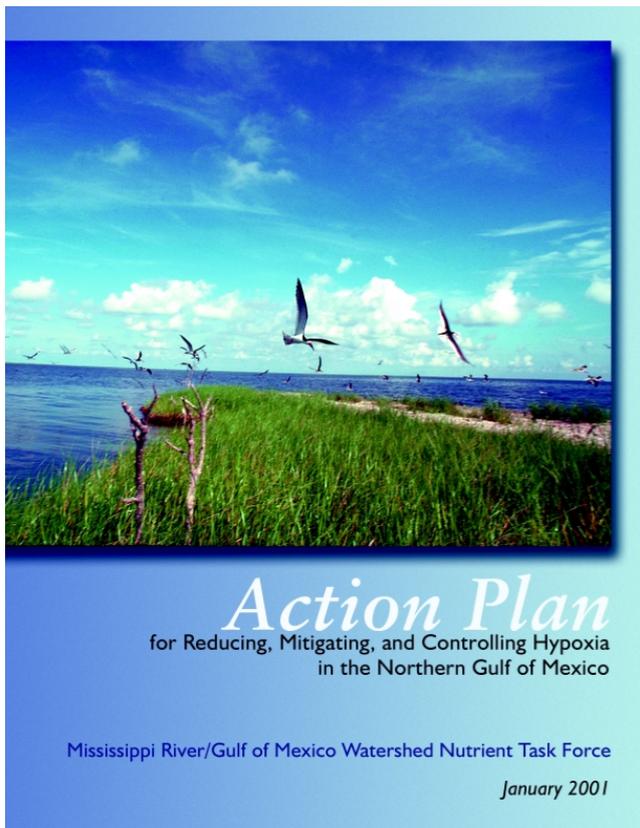
How far do N and/or P loads have to be reduced in order to affect desired reductions in hypoxia?

“By the year 2015 .. reduce the five-year running average areal extent of the Gulf of Mexico hypoxic zone to less than 5,000 square km” (p.9)

“The best current science indicates .. a 30% reduction .. in nitrogen discharges .. is consistent with the goal..” (p.21)

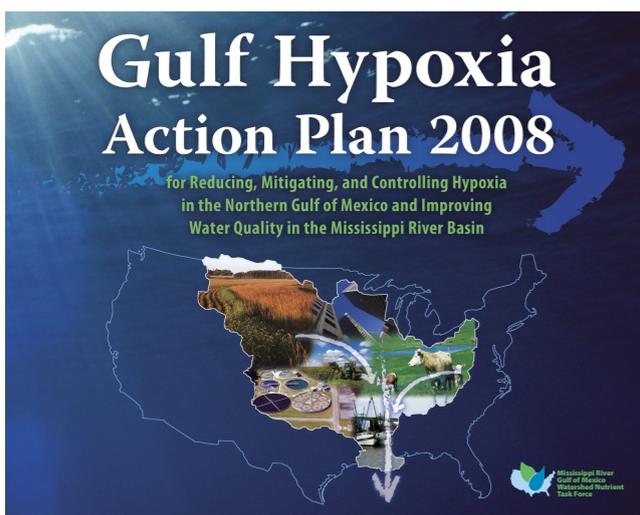


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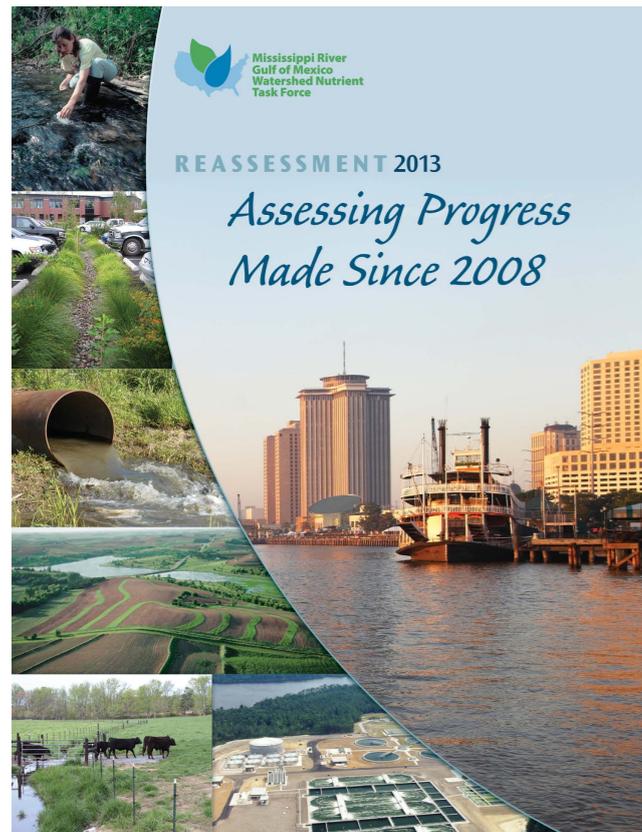
“The best current science indicates .. a 30% reduction .. in nitrogen discharges .. is consistent with the goal..” (p.21)



“.. must reduce nitrogen and phosphorus in .. the River Basin.” (p.29)

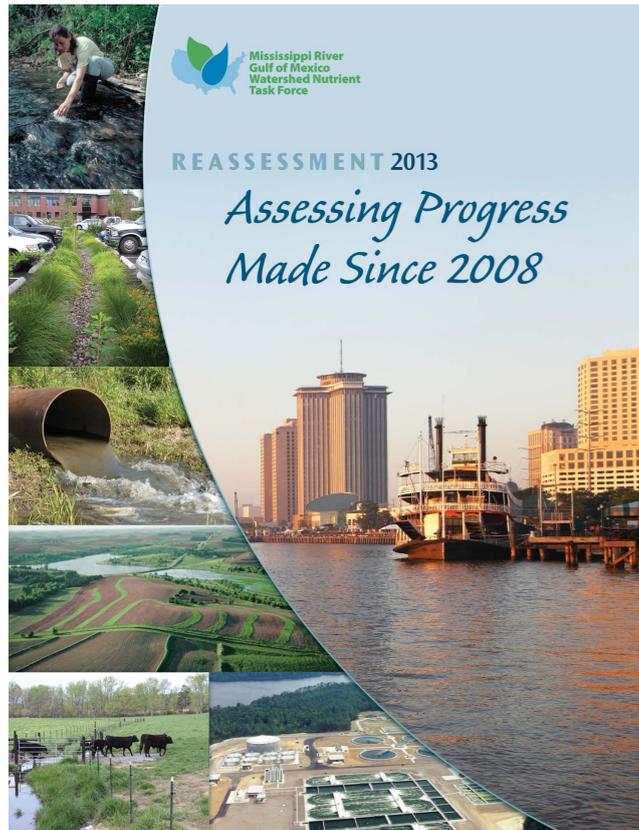
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“Decrease scientific uncertainty of nitrogen and phosphorus effects on hypoxia” (p.48)

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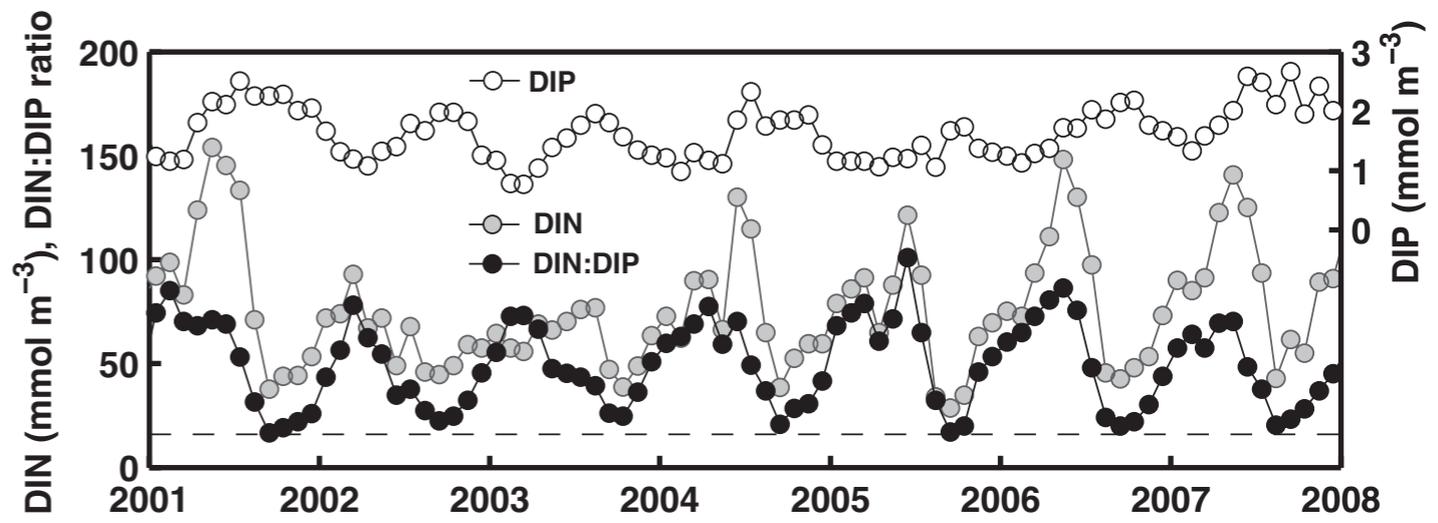


“Decrease scientific uncertainty of nitrogen and phosphorus effects on hypoxia” (p.48)

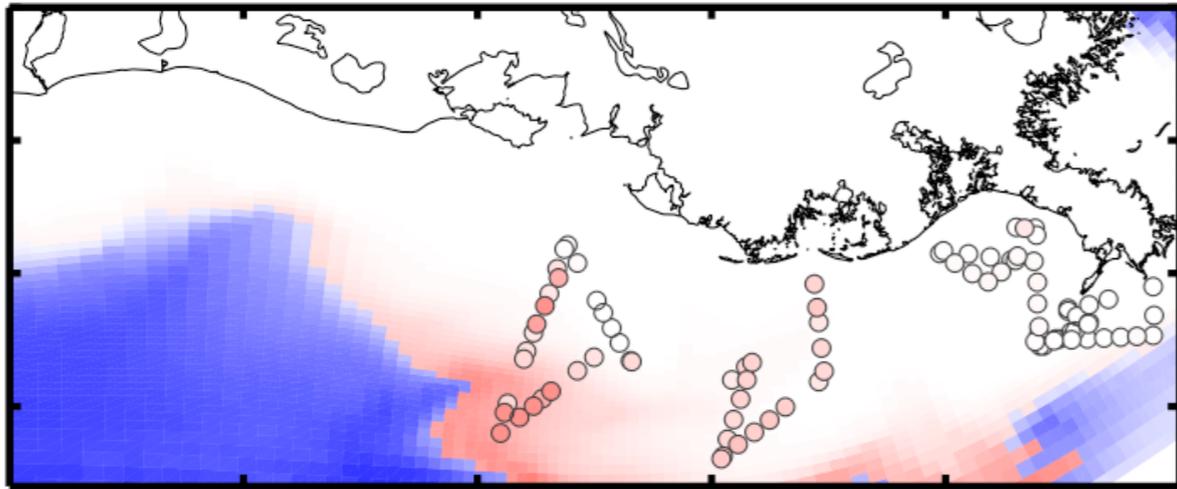


“Retain 2008 Action plan goal of 5,000 km² by year 2035.” (p.1)

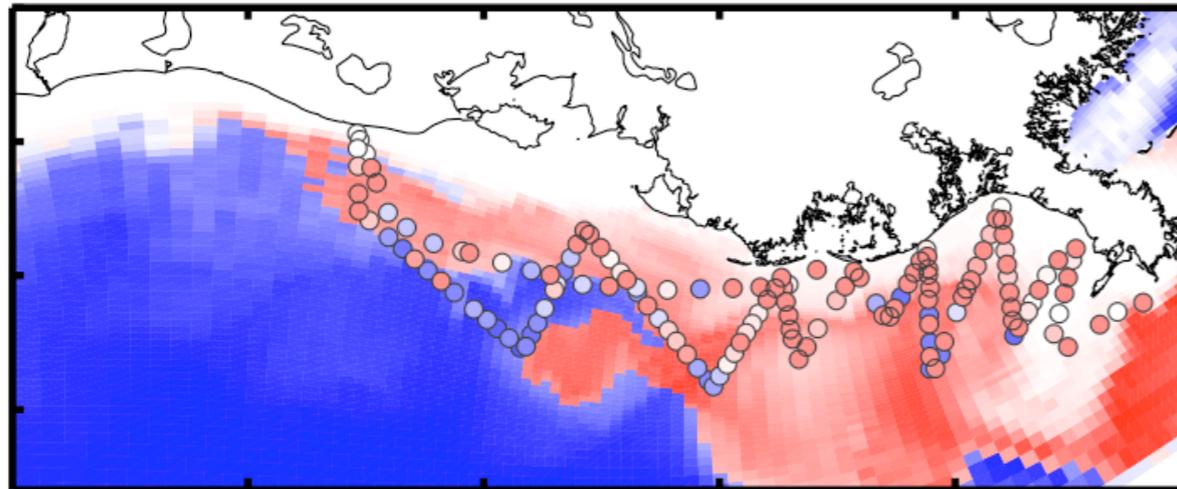
“Interim target of 20% nutrient load reduction by year 2025.” (p.1)



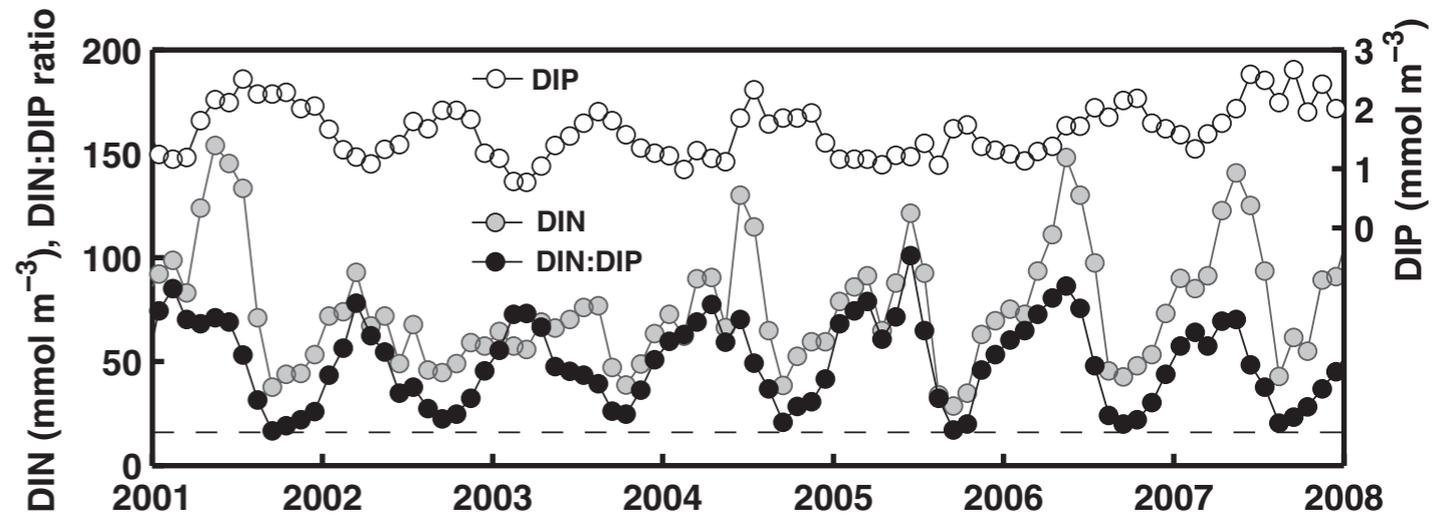
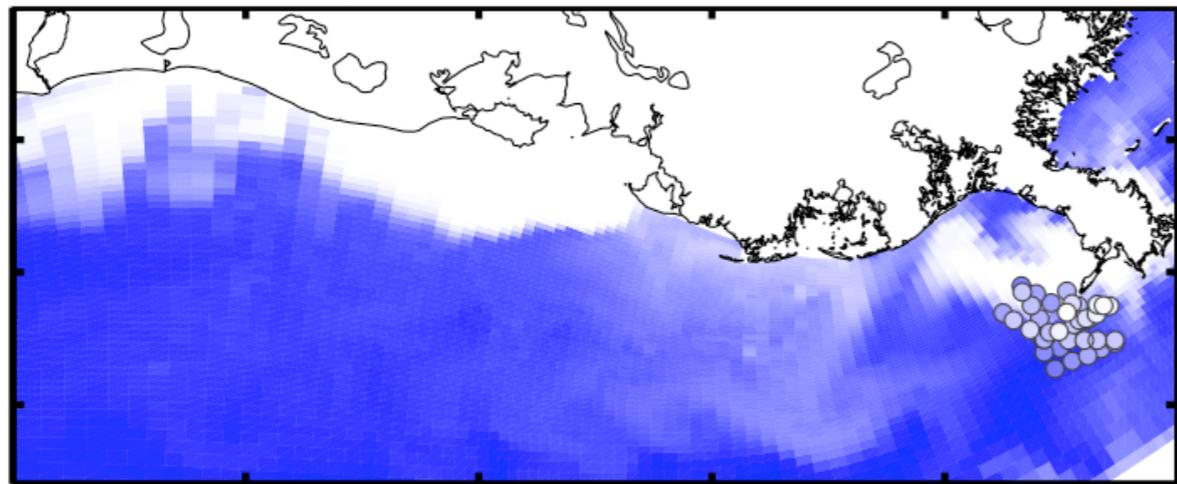
March 20-24, 2001



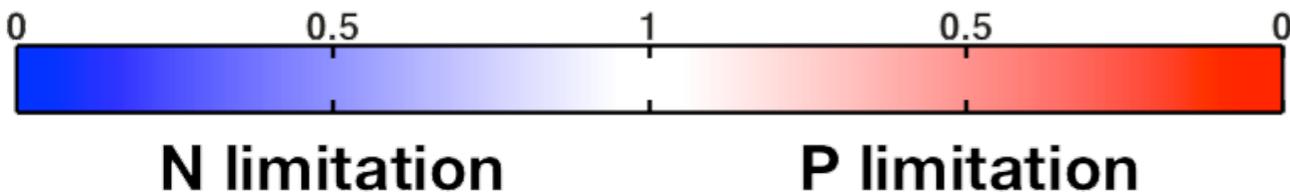
June 29 - July 3, 2002



September 21-22, 2001



Laurent et al. *Biogeosciences* (2012)
using observations from Sylvan et al.
EST (2006)



Nutrient reduction strategies in the Mississippi Basin have long focused on N assuming it is the ultimate limiting nutrient while P is only limiting in a proximate sense.

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prox•i•mate | 'präksəmit | immediate

ul•ti•mate | 'əltəmit | final

Nutrient reduction strategies in the Mississippi Basin have long focused on N assuming it is the ultimate limiting nutrient while P is only limiting in a proximate sense.

prox•i•mate | 'präksəmit | immediate

ul•ti•mate | 'əltəmit | final

In Ecology (after Tyrrell 1999):

The proximate limiting nutrient is the one that is locally (or temporarily) limiting Primary Production (PP). Its addition will immediately enhance PP.

Nutrient reduction strategies in the Mississippi Basin have long focused on N assuming it is the ultimate limiting nutrient while P is only limiting in a proximate sense.

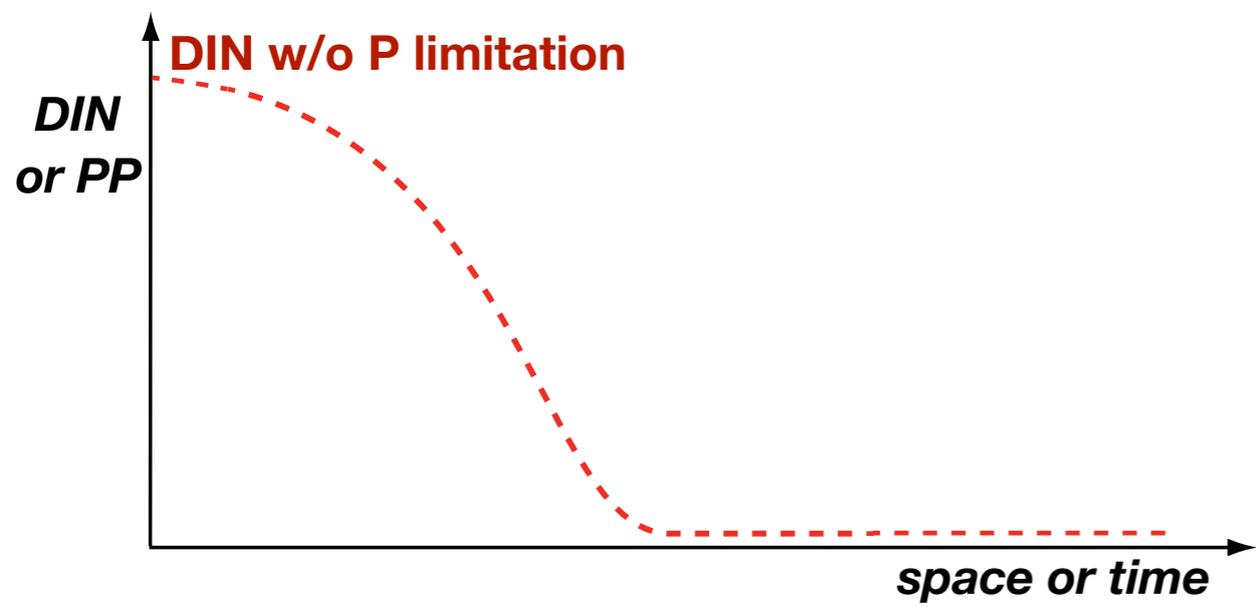
prox•i•mate | 'präksəmit | immediate

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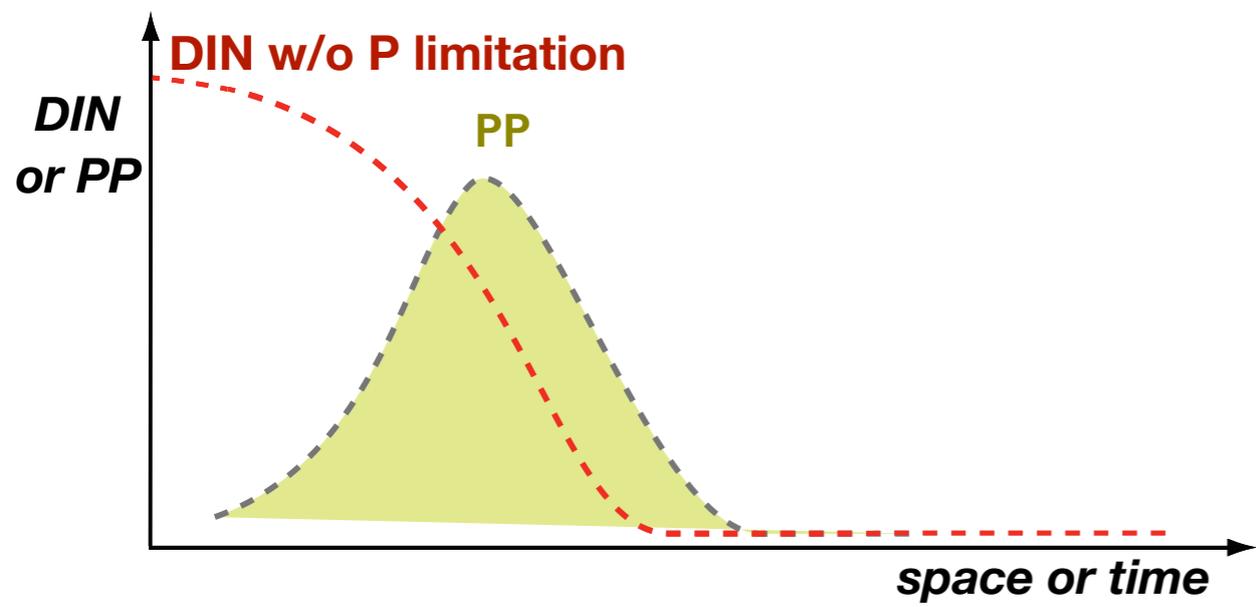
In Ecology (after Tyrrell 1999):

The proximate limiting nutrient is the one that is locally (or temporarily) limiting Primary Production (PP). Its addition will immediately enhance PP.

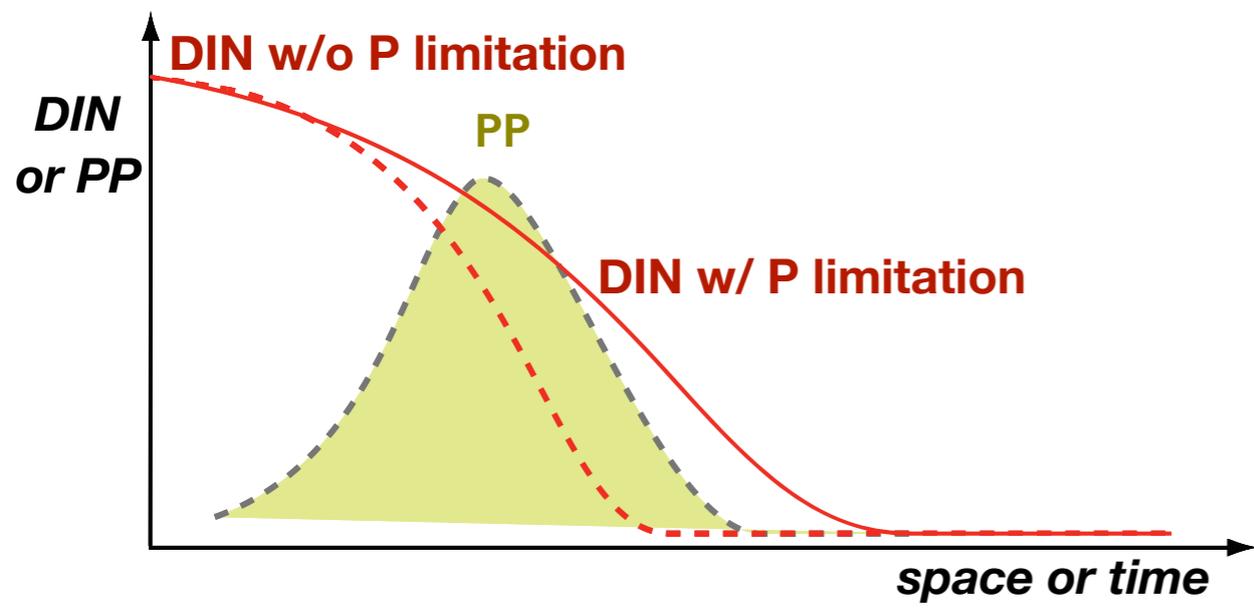
The supply of the ultimate limiting nutrient determines system productivity over long time scales.



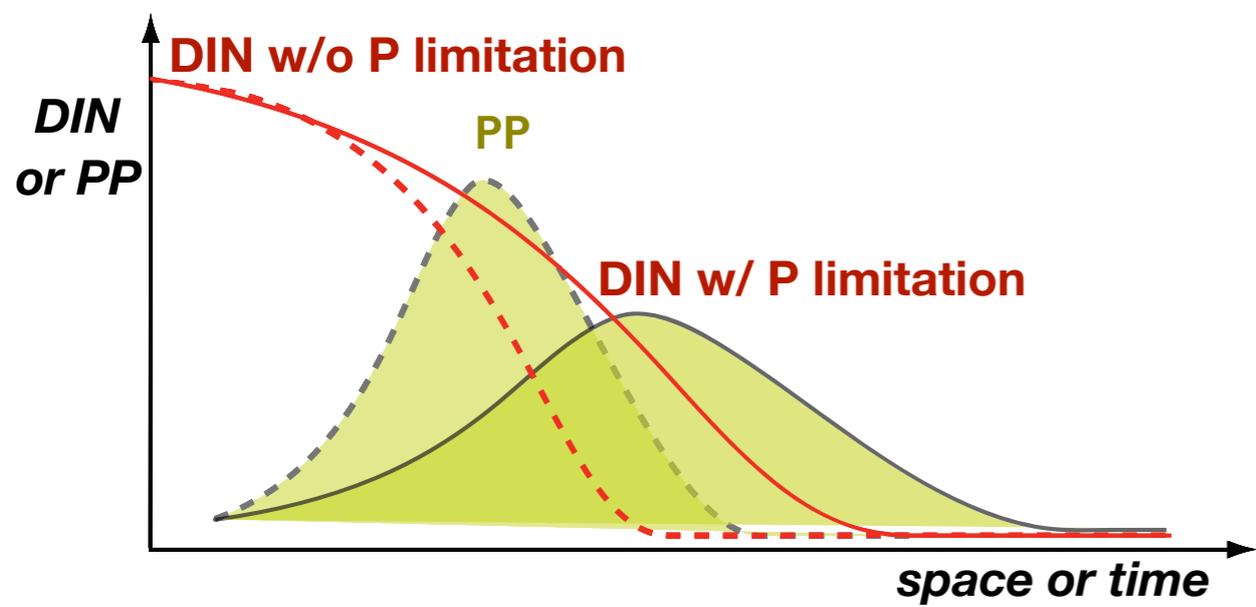
Implications of P limitation



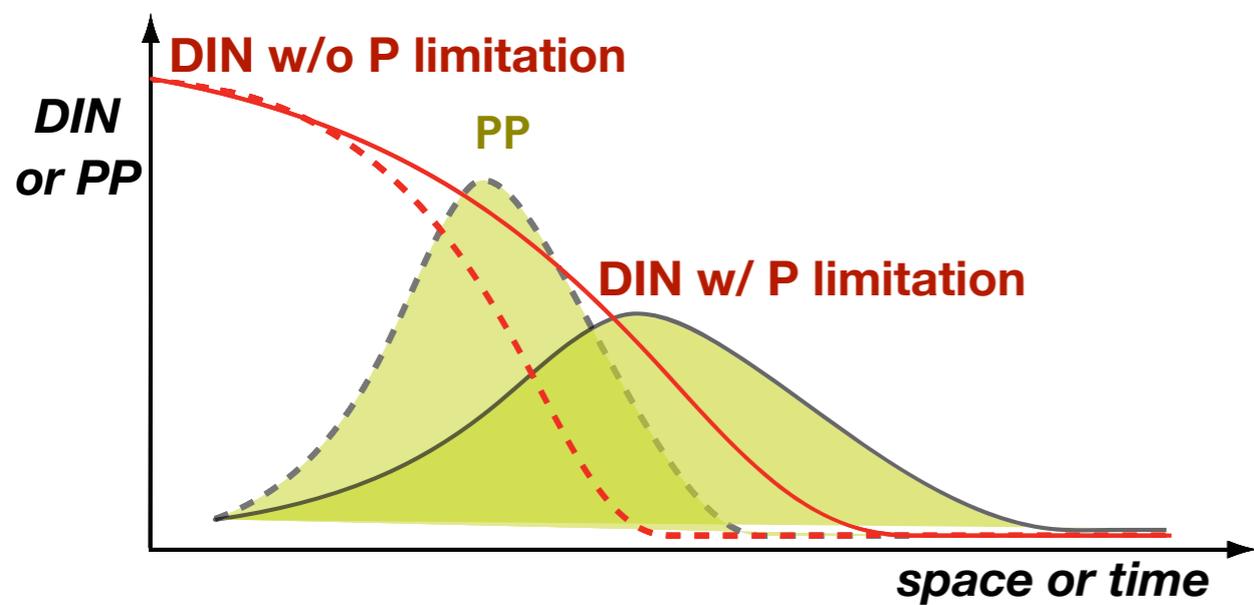
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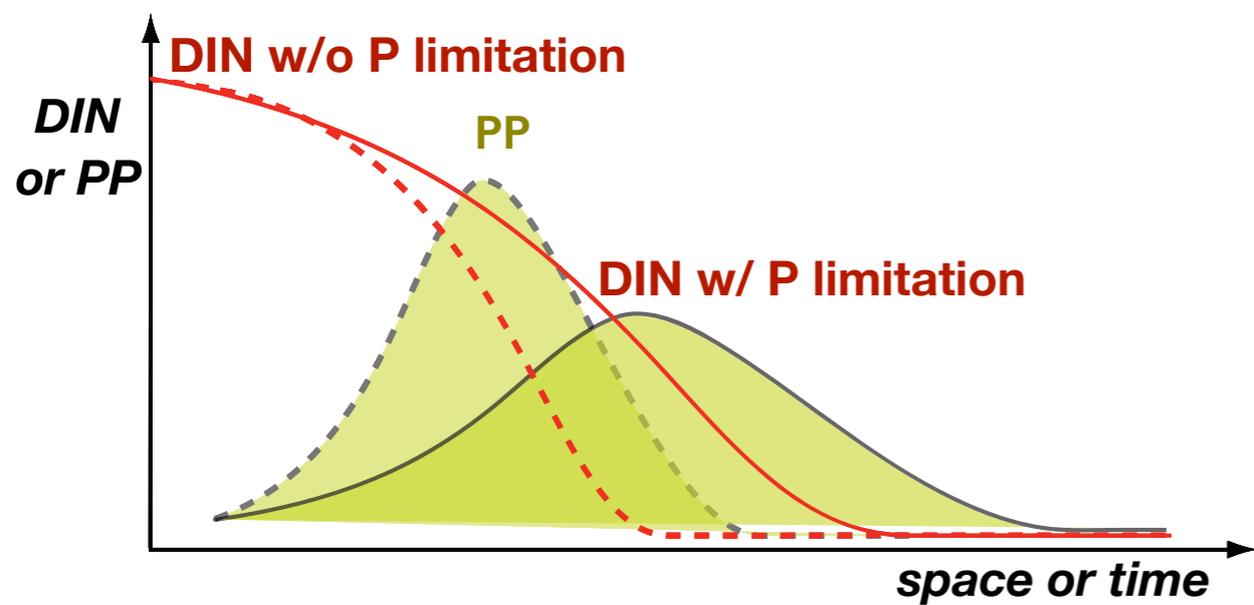


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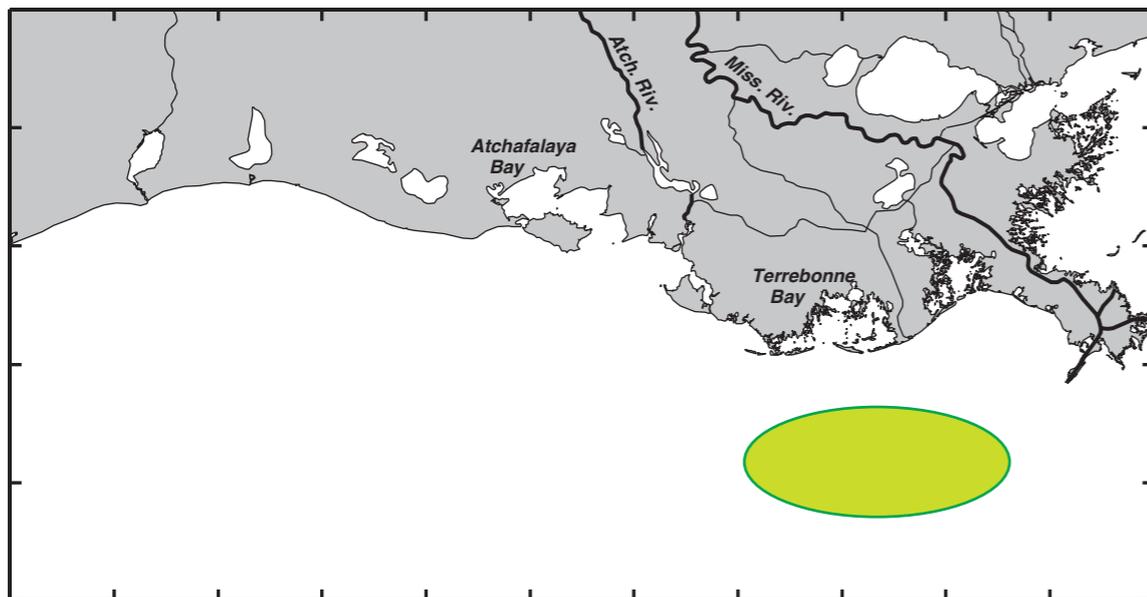
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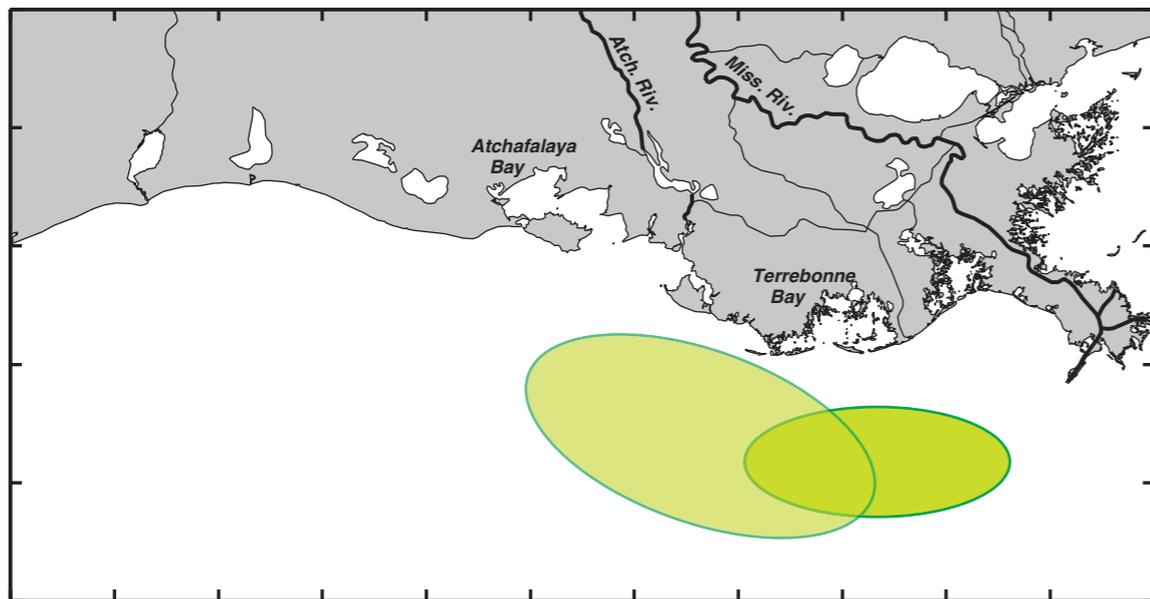
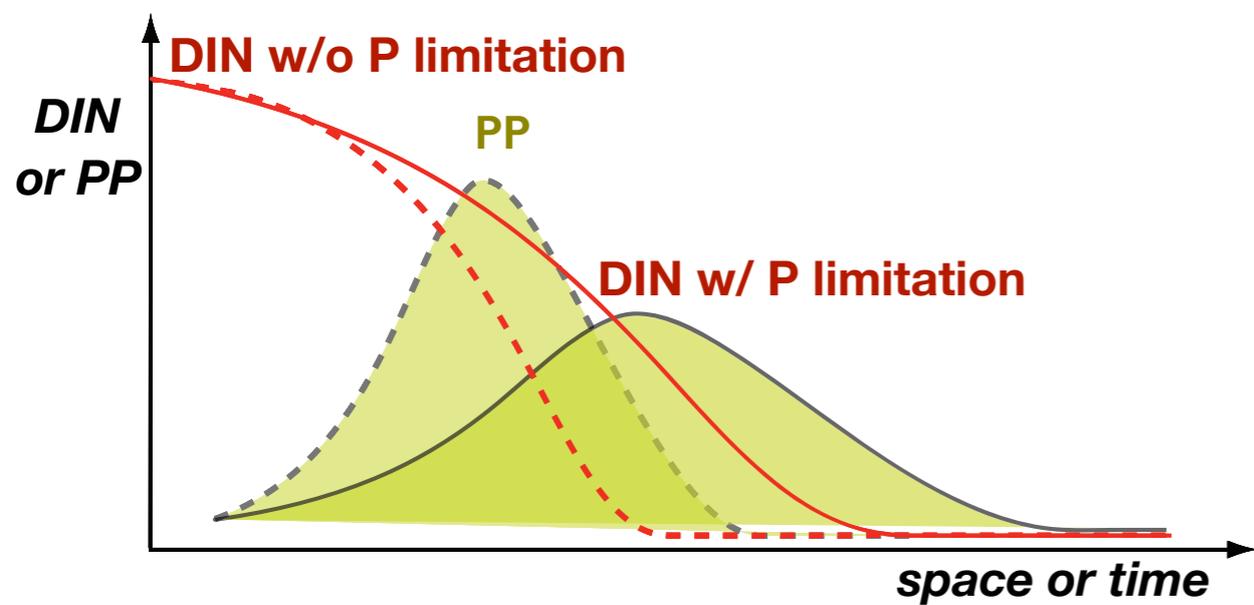
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- reduces magnitude of PP peak but results in elevated PP in larger area/over longer time period



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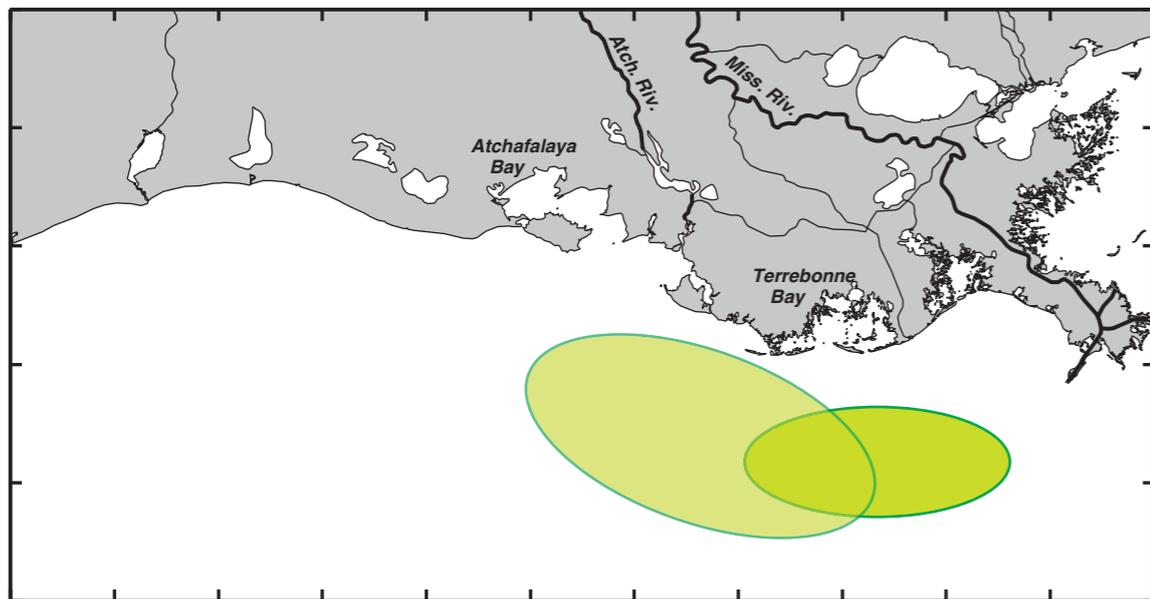
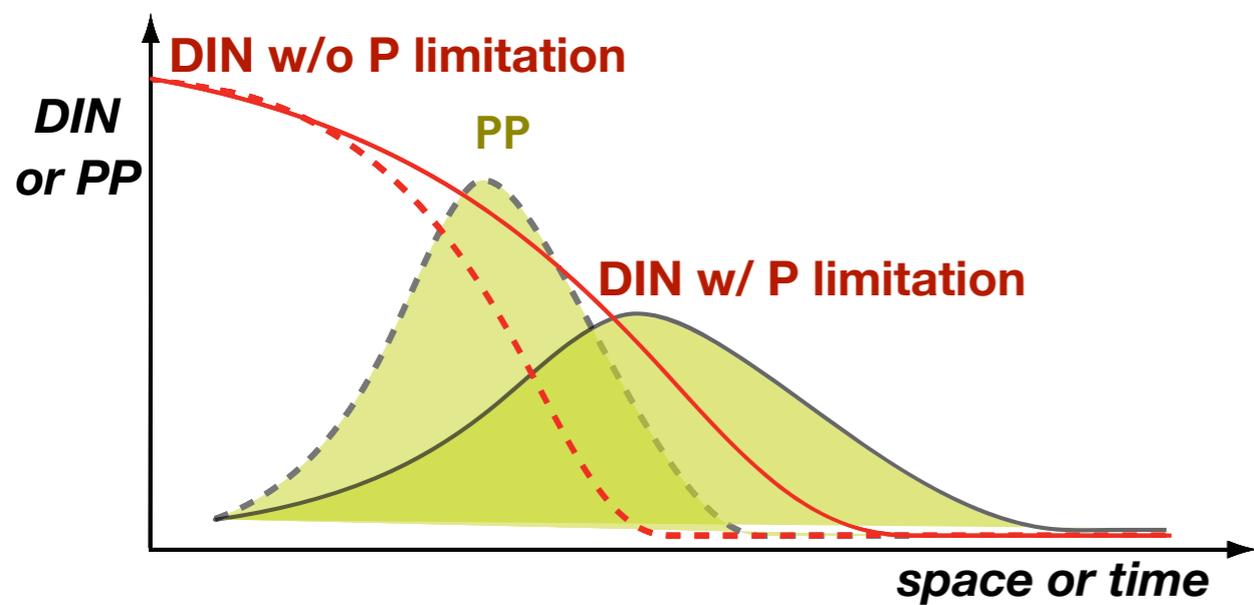
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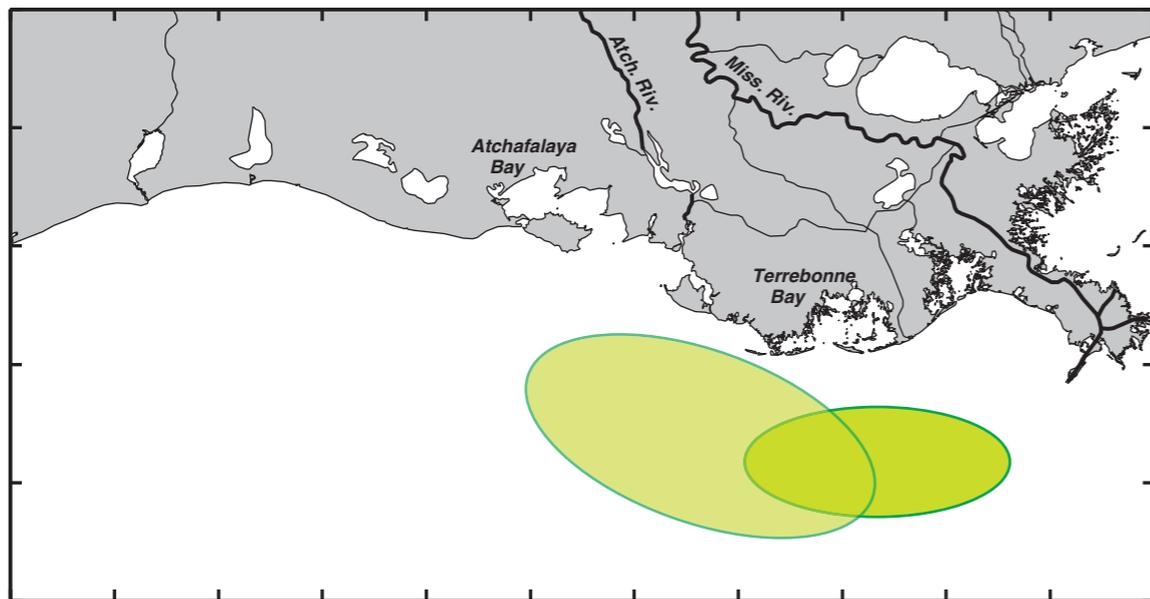
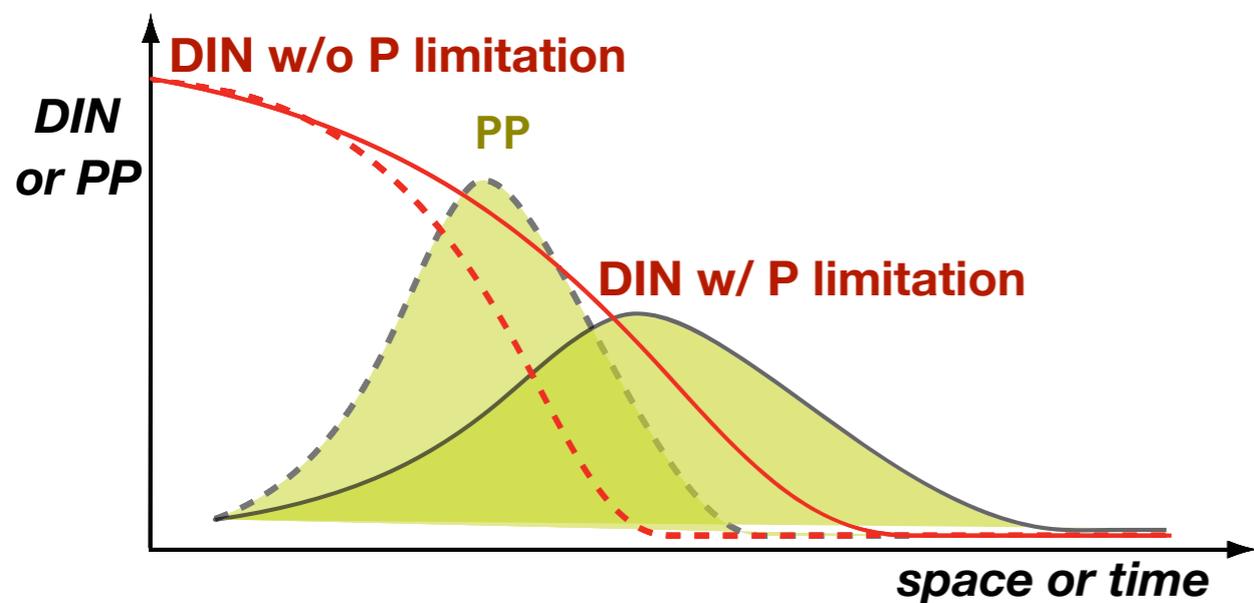
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And hypoxia?

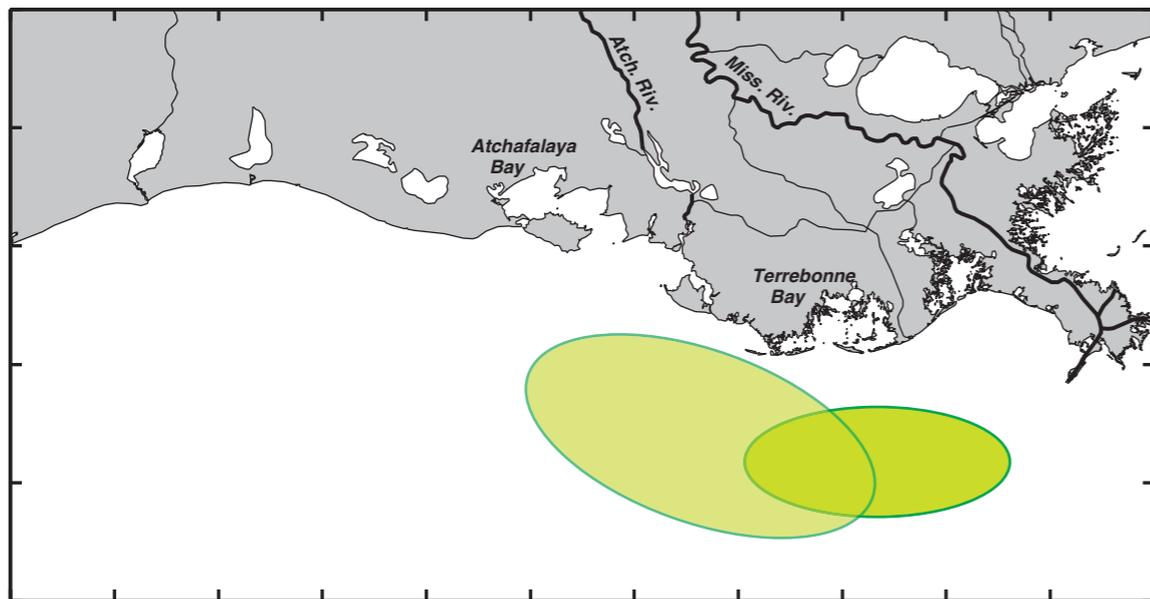
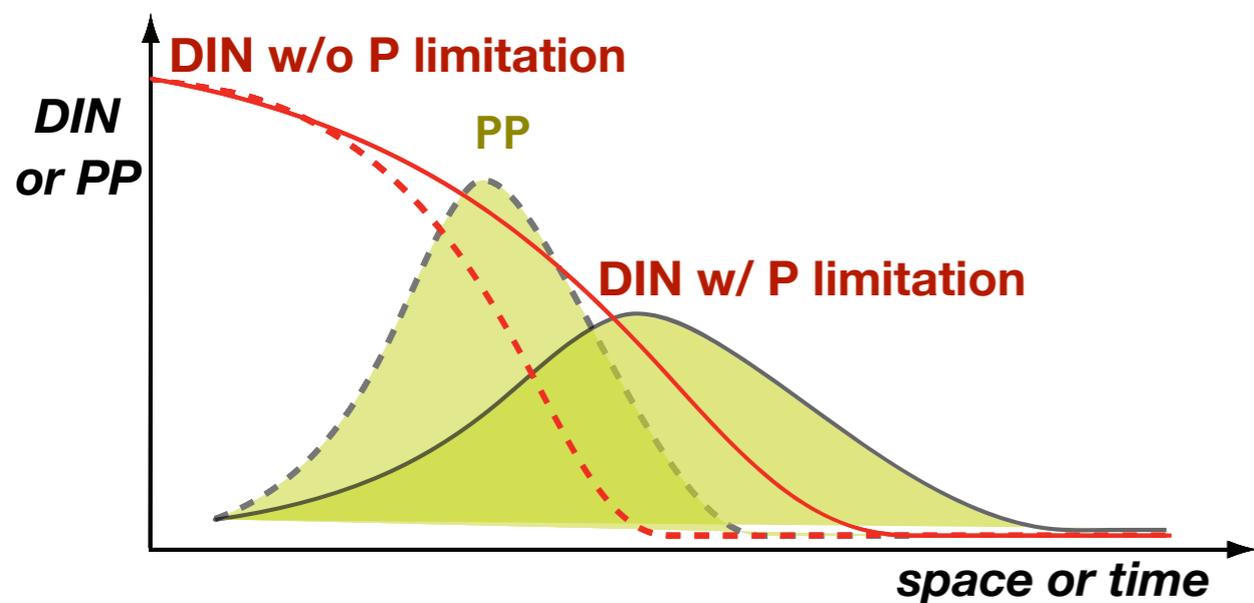


Implications of P limitation

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- could intensify (Pearl 2004: Neuse River estuary; Conley et al. 2009: Baltic Sea)

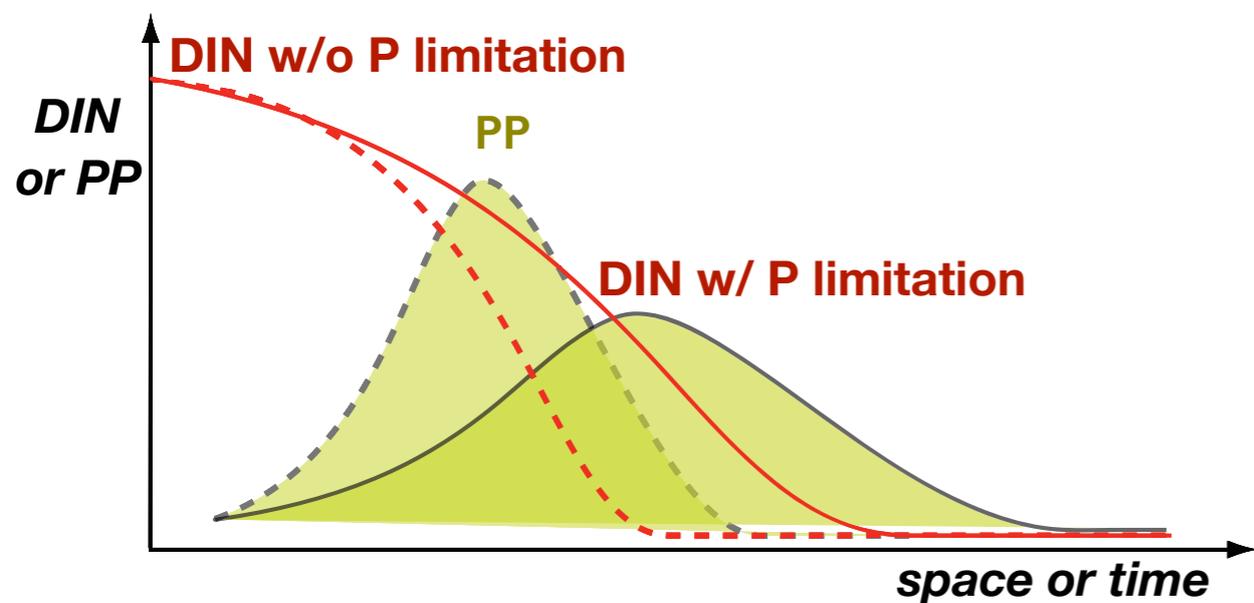


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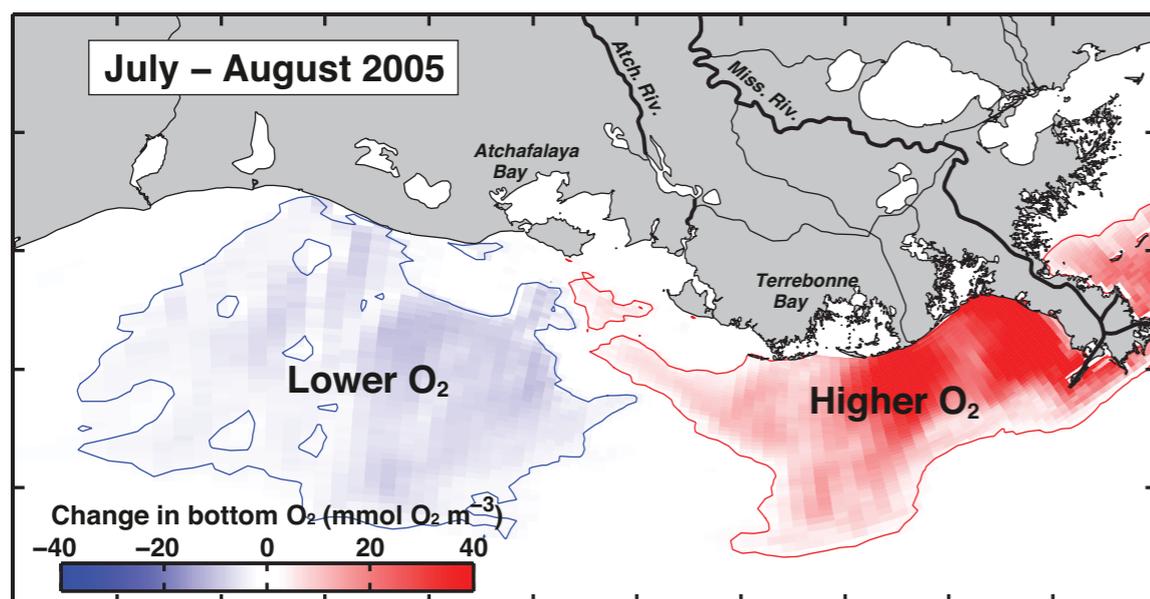
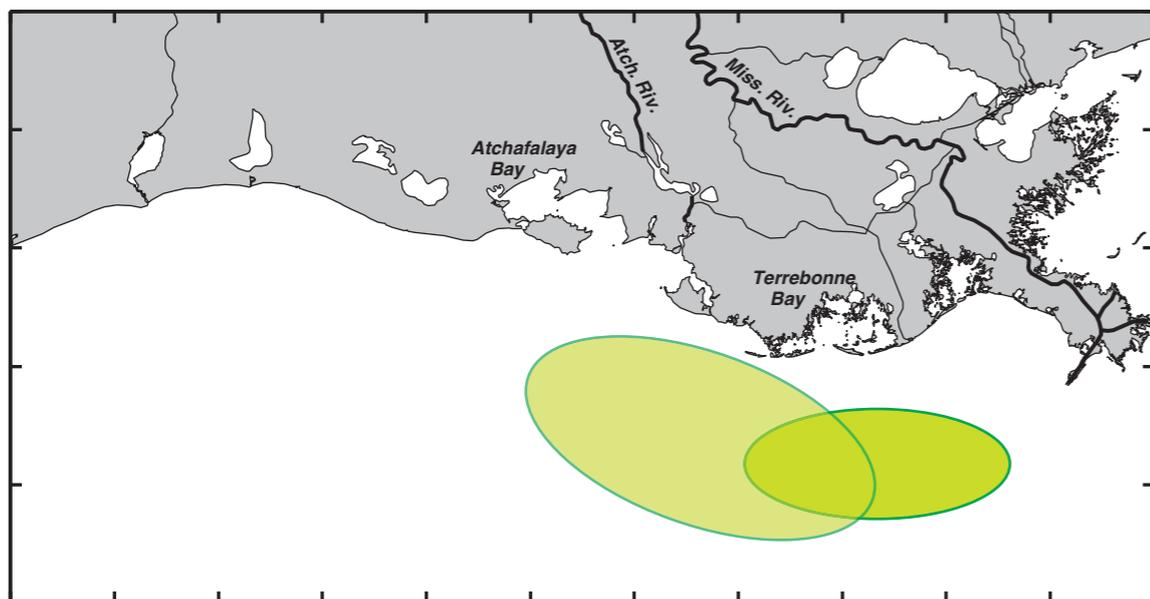


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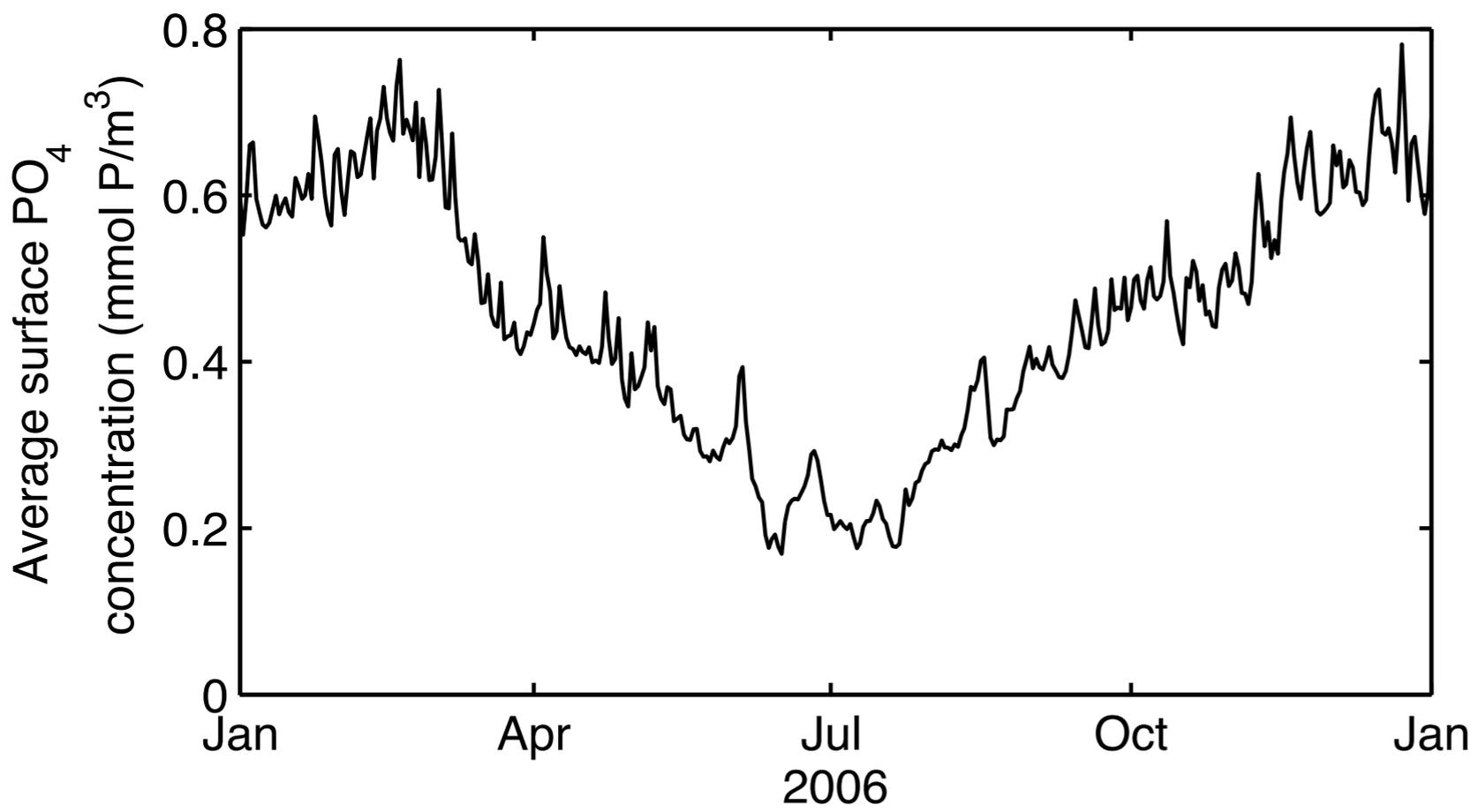
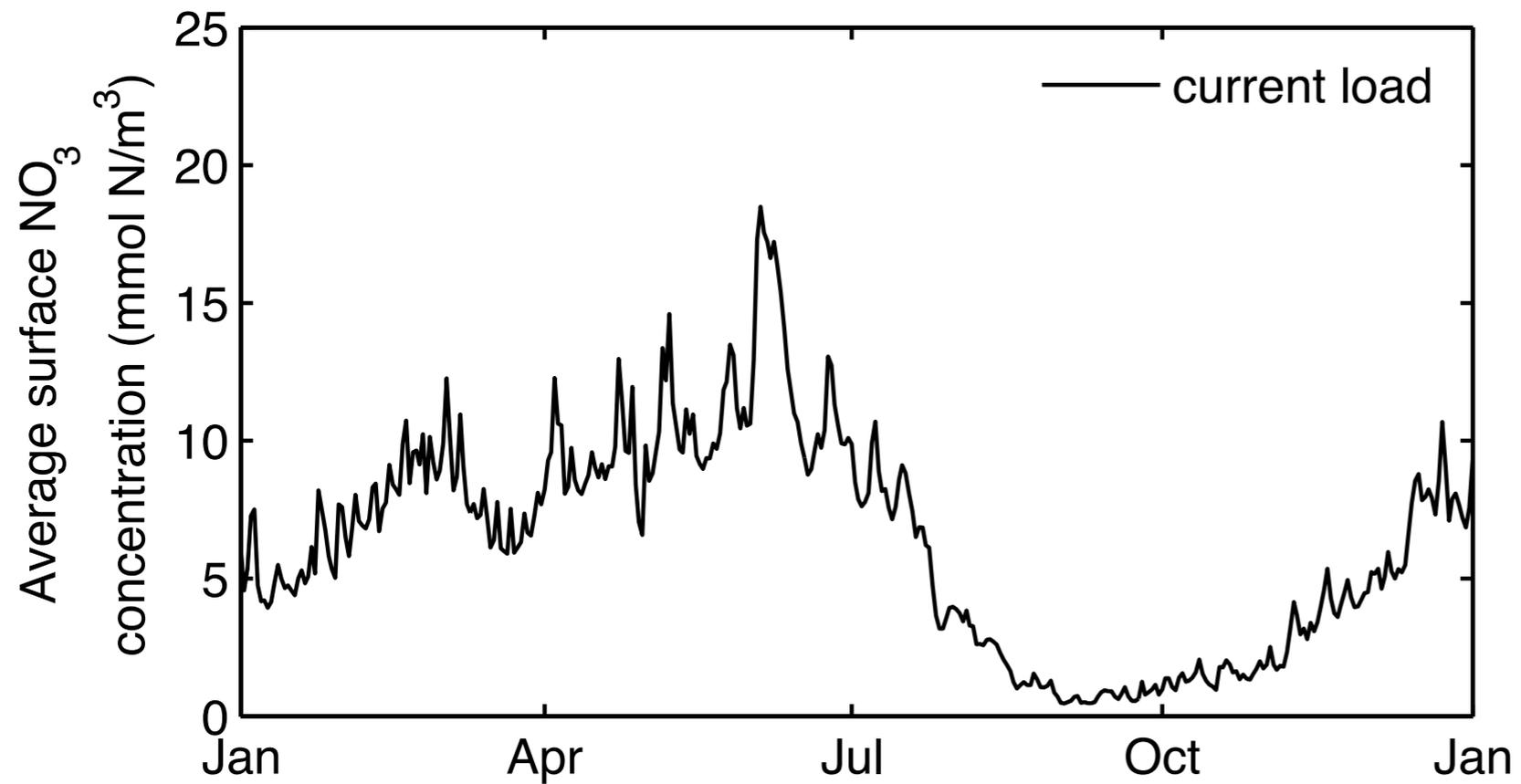
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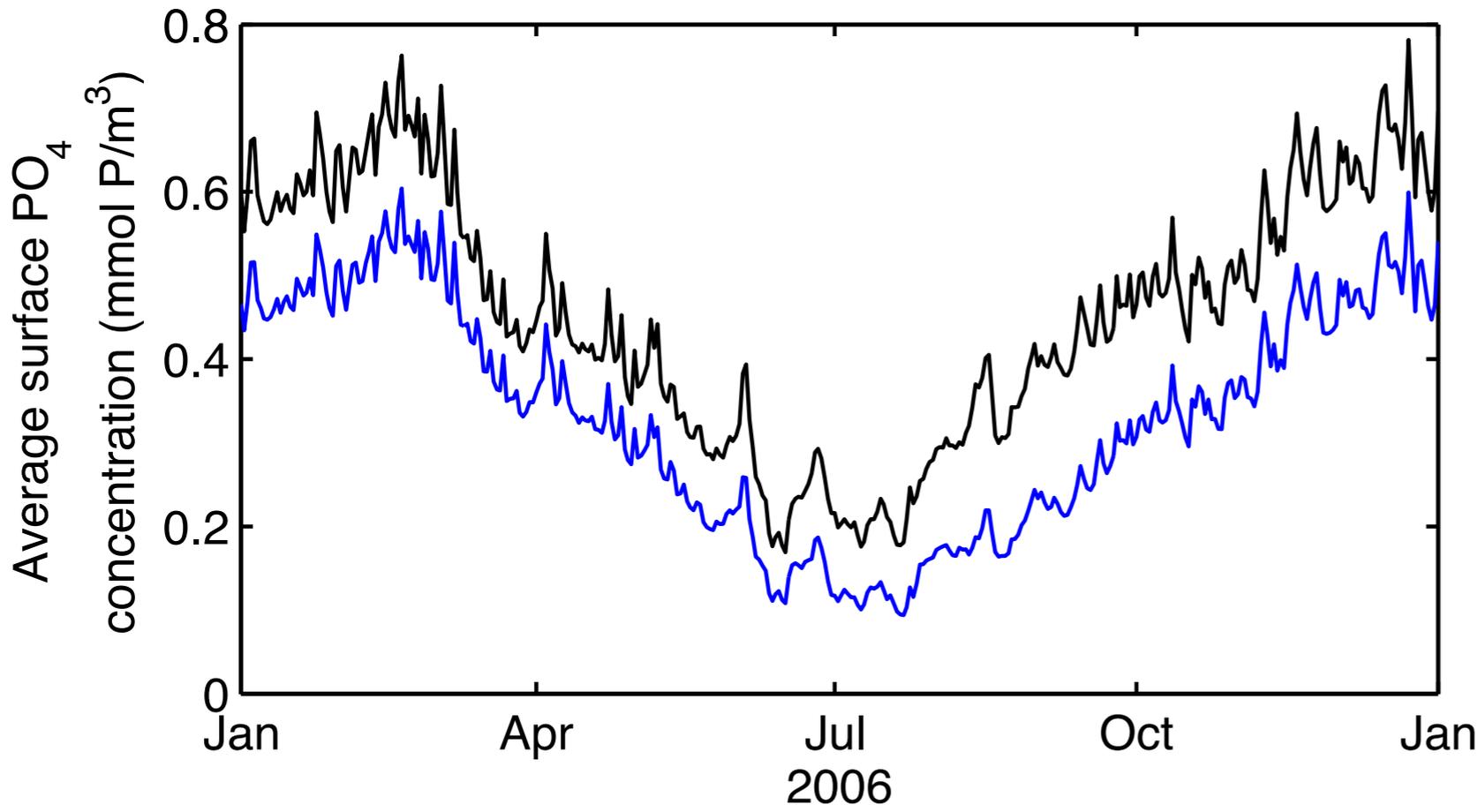
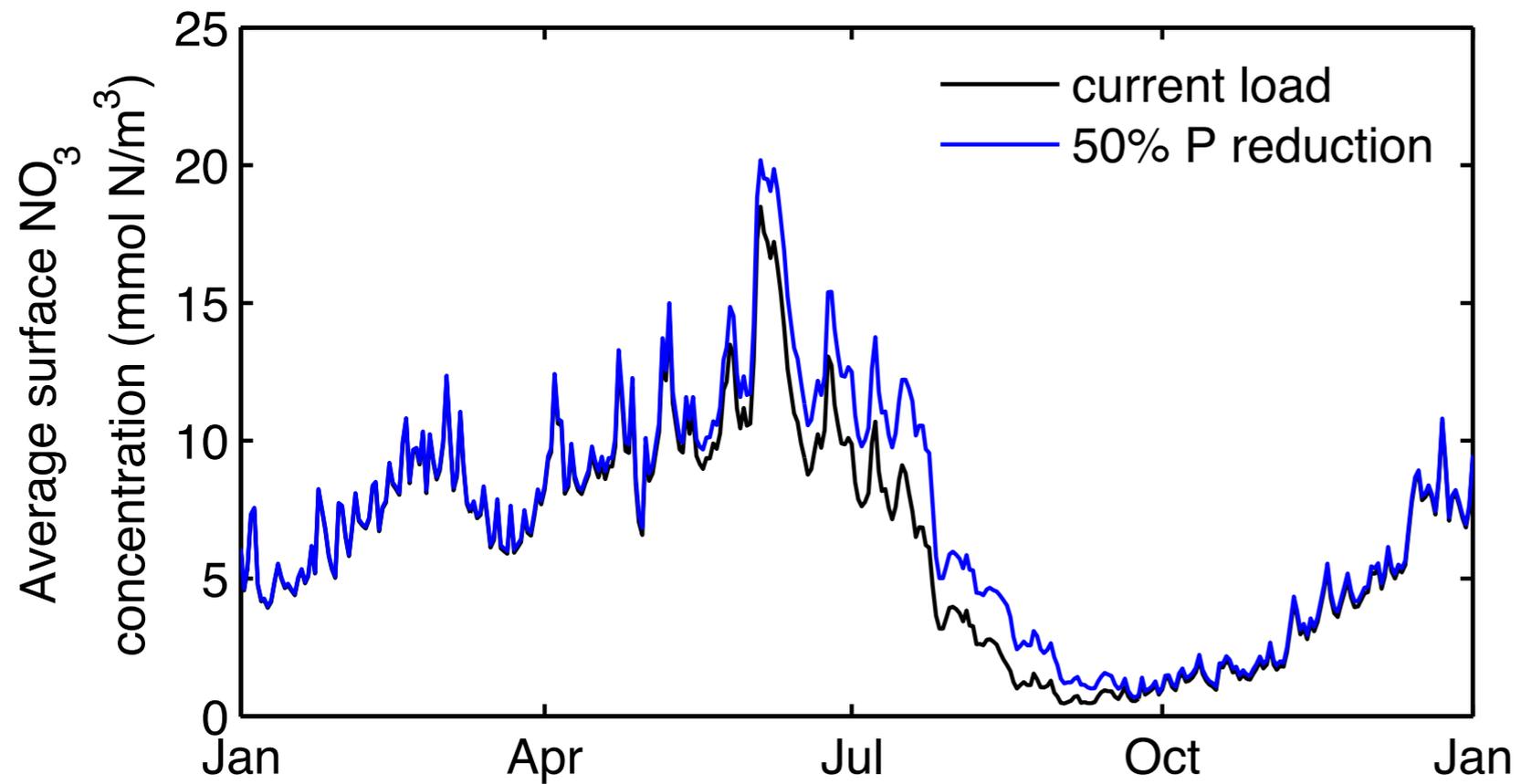
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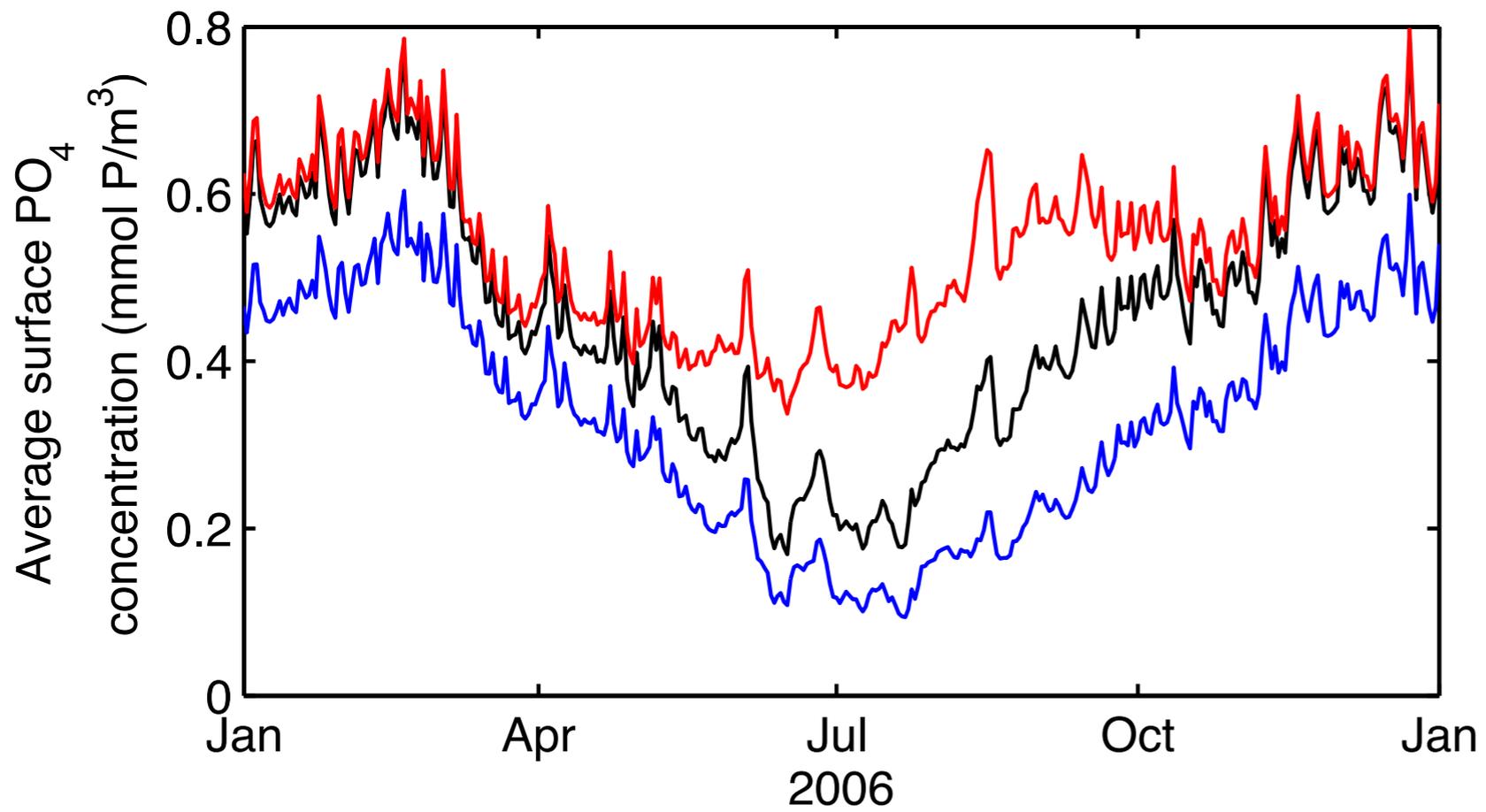
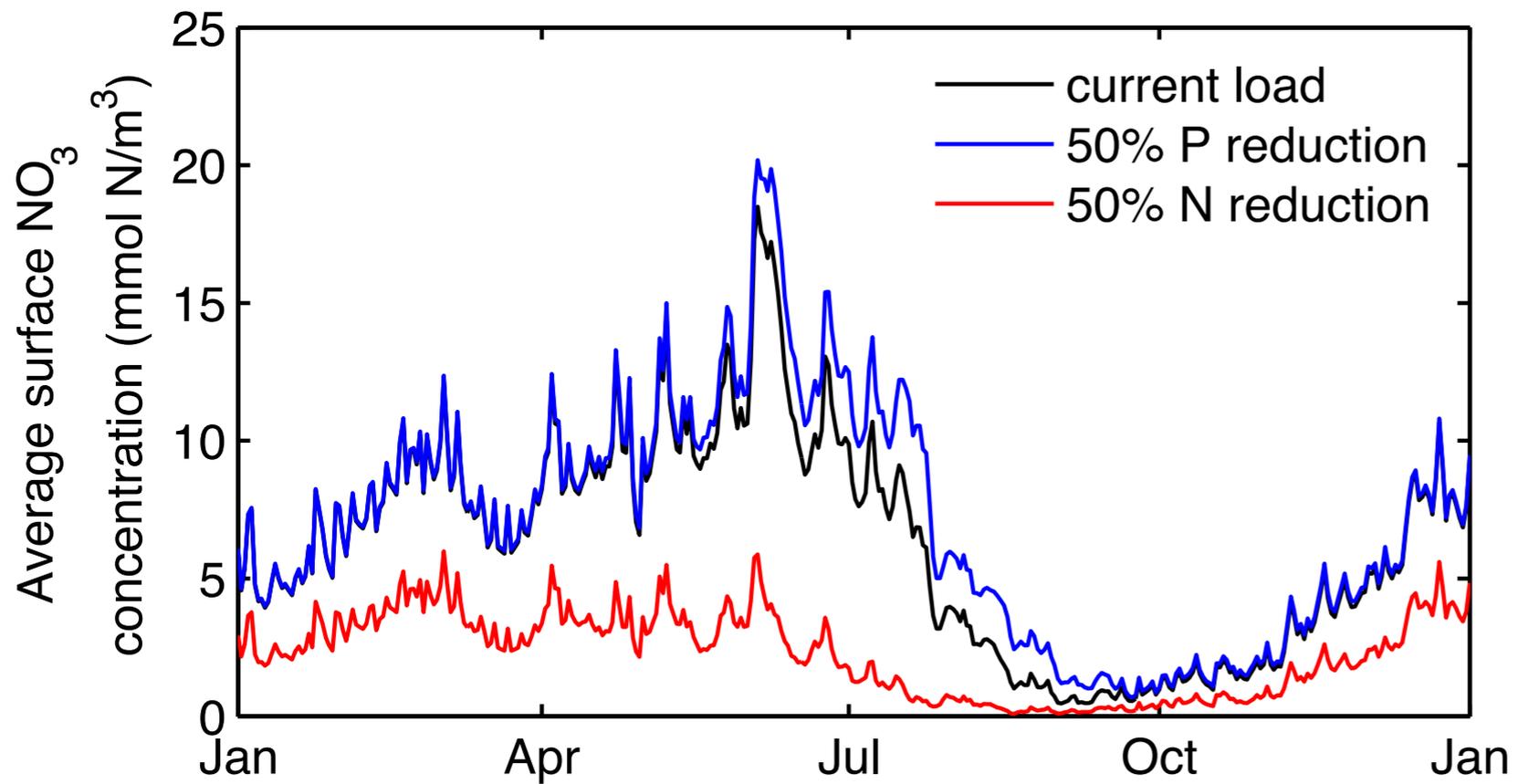
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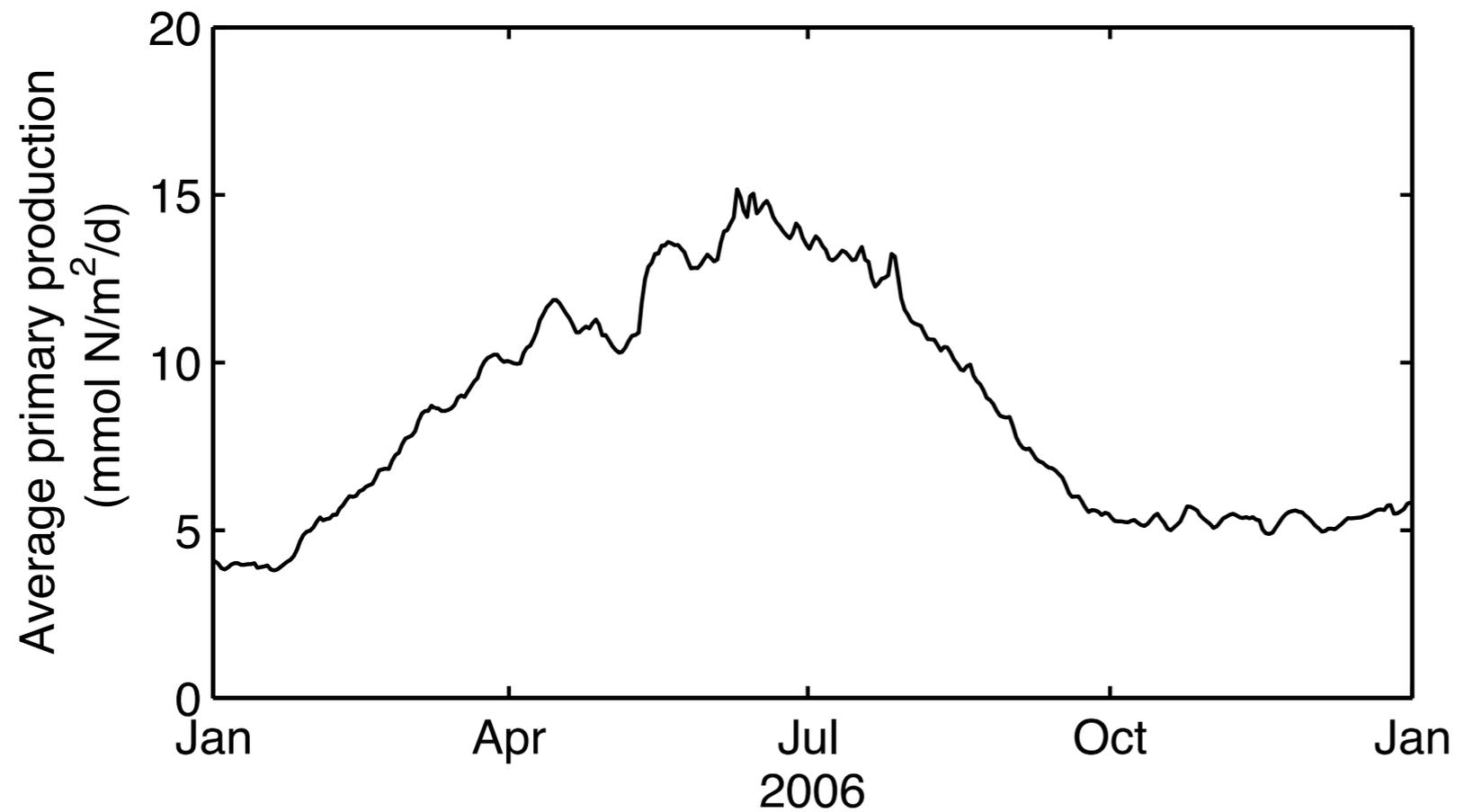
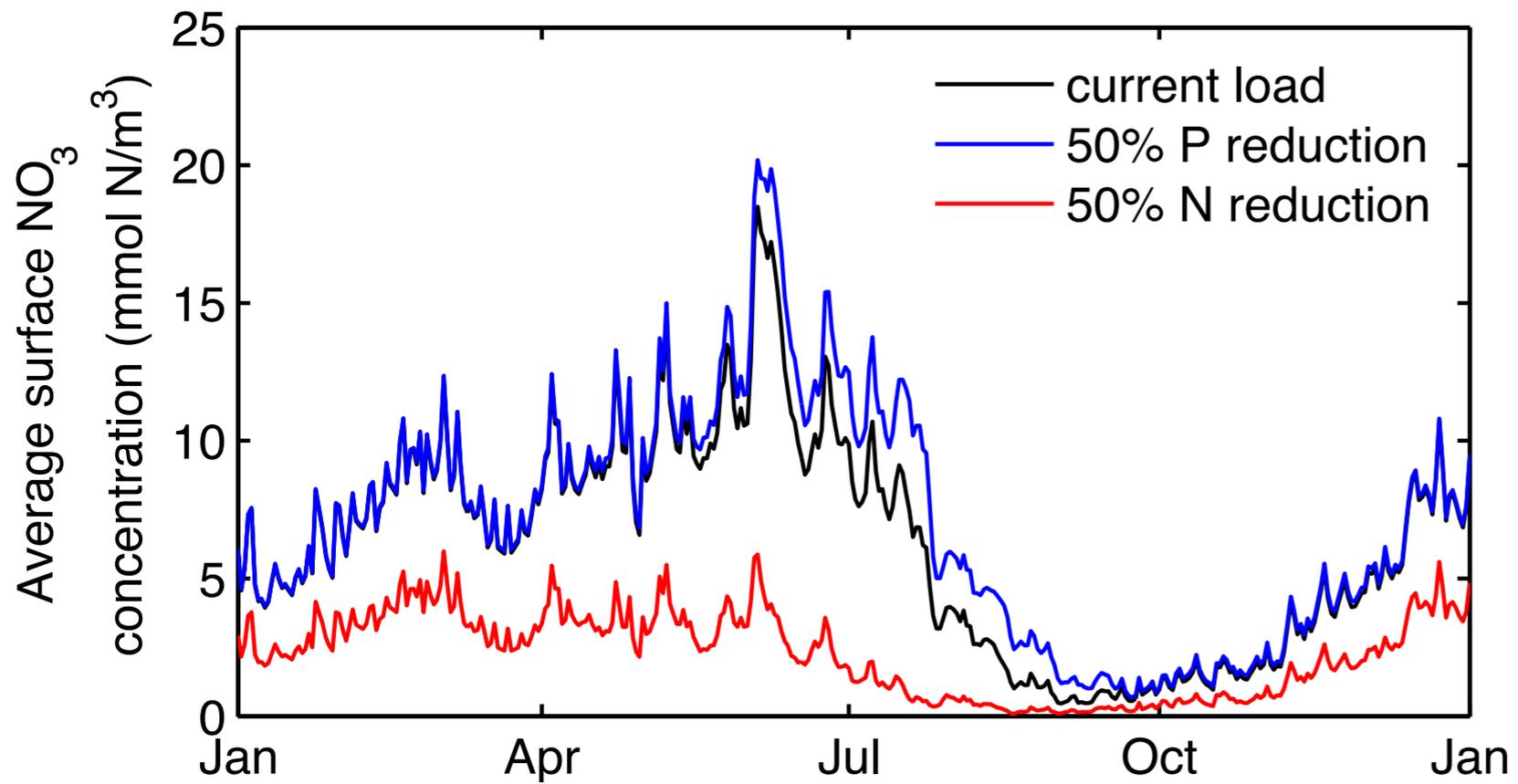


Laurent & Fennel *Elementa* (2014)



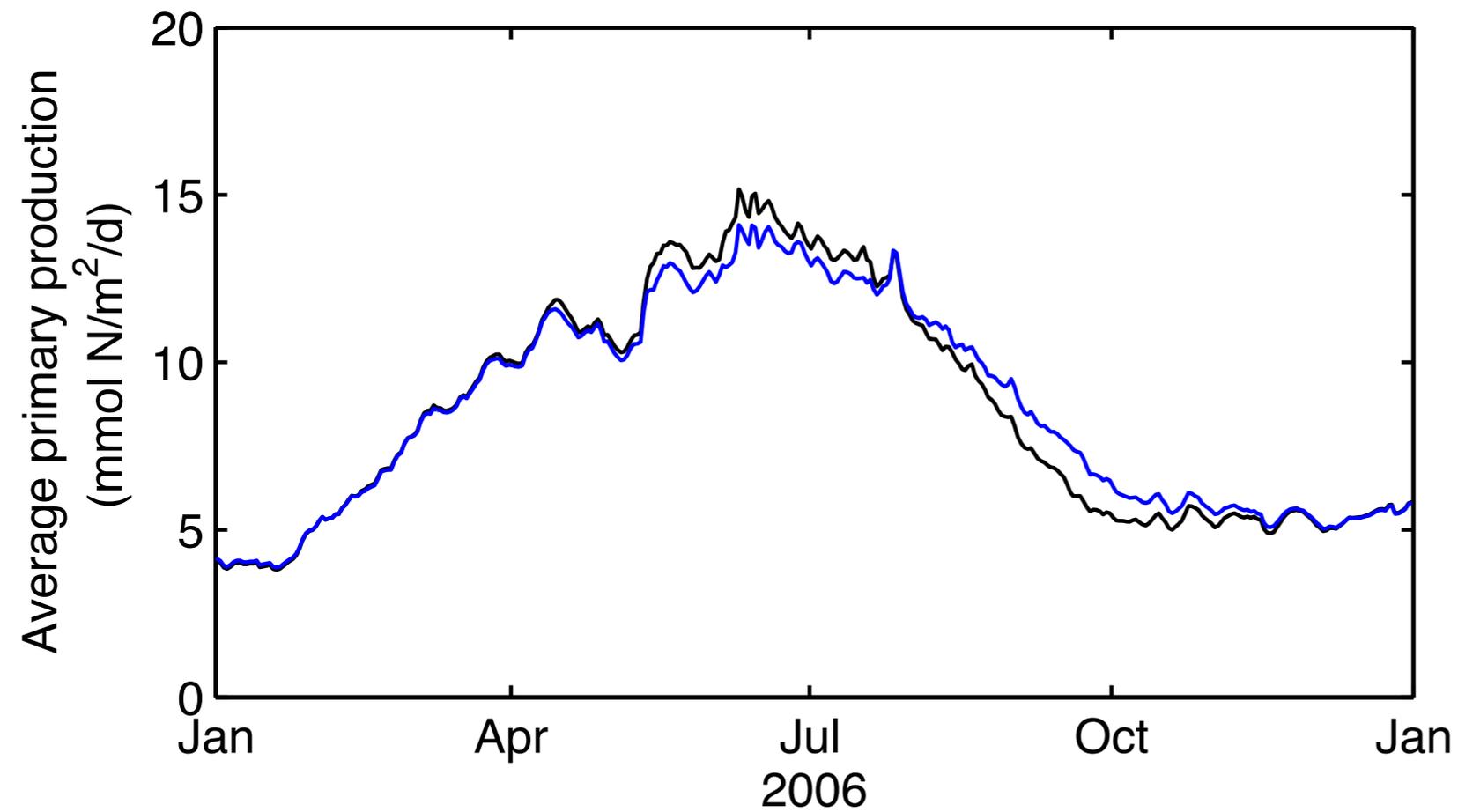
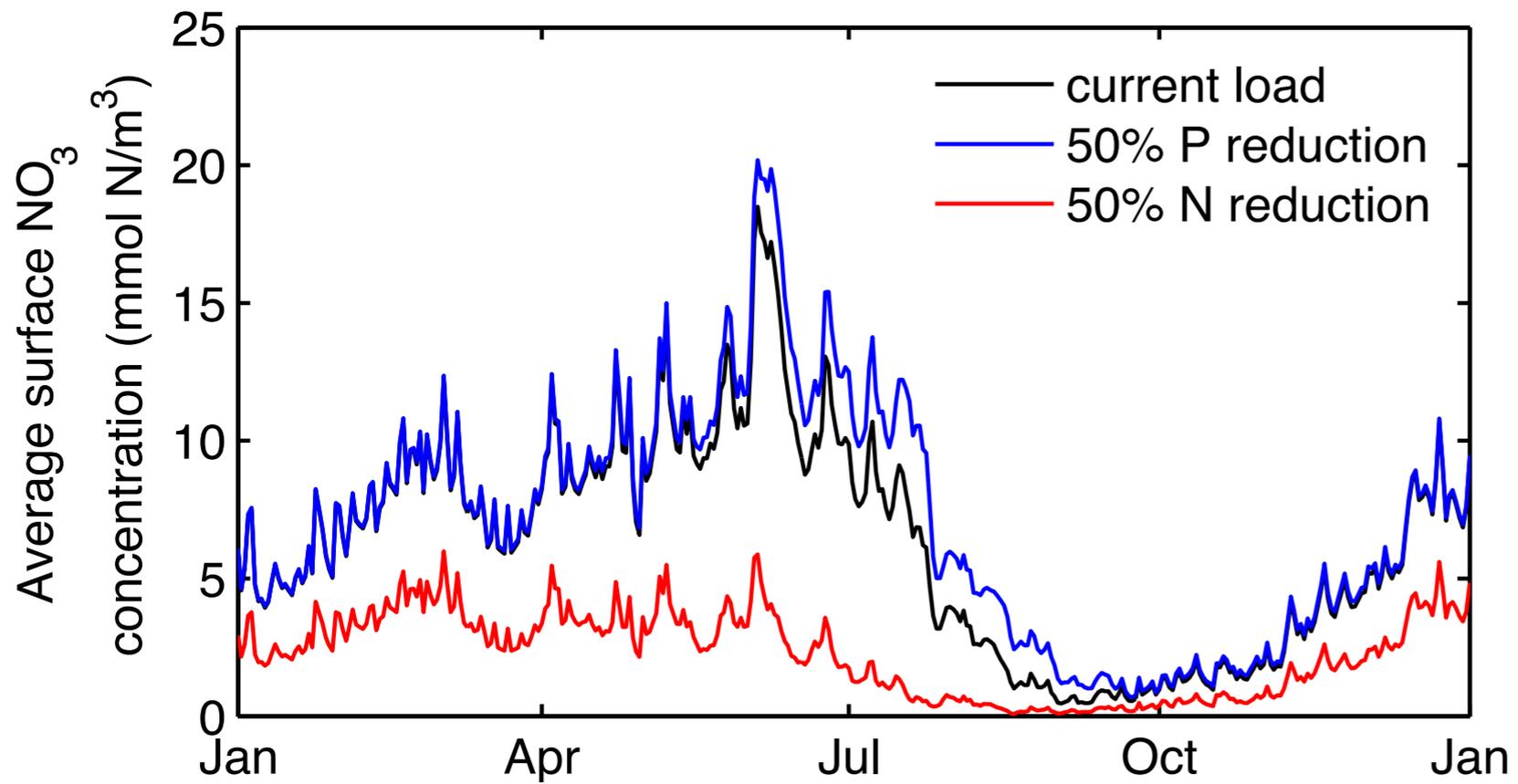






Annual PP (shelf-wide average):

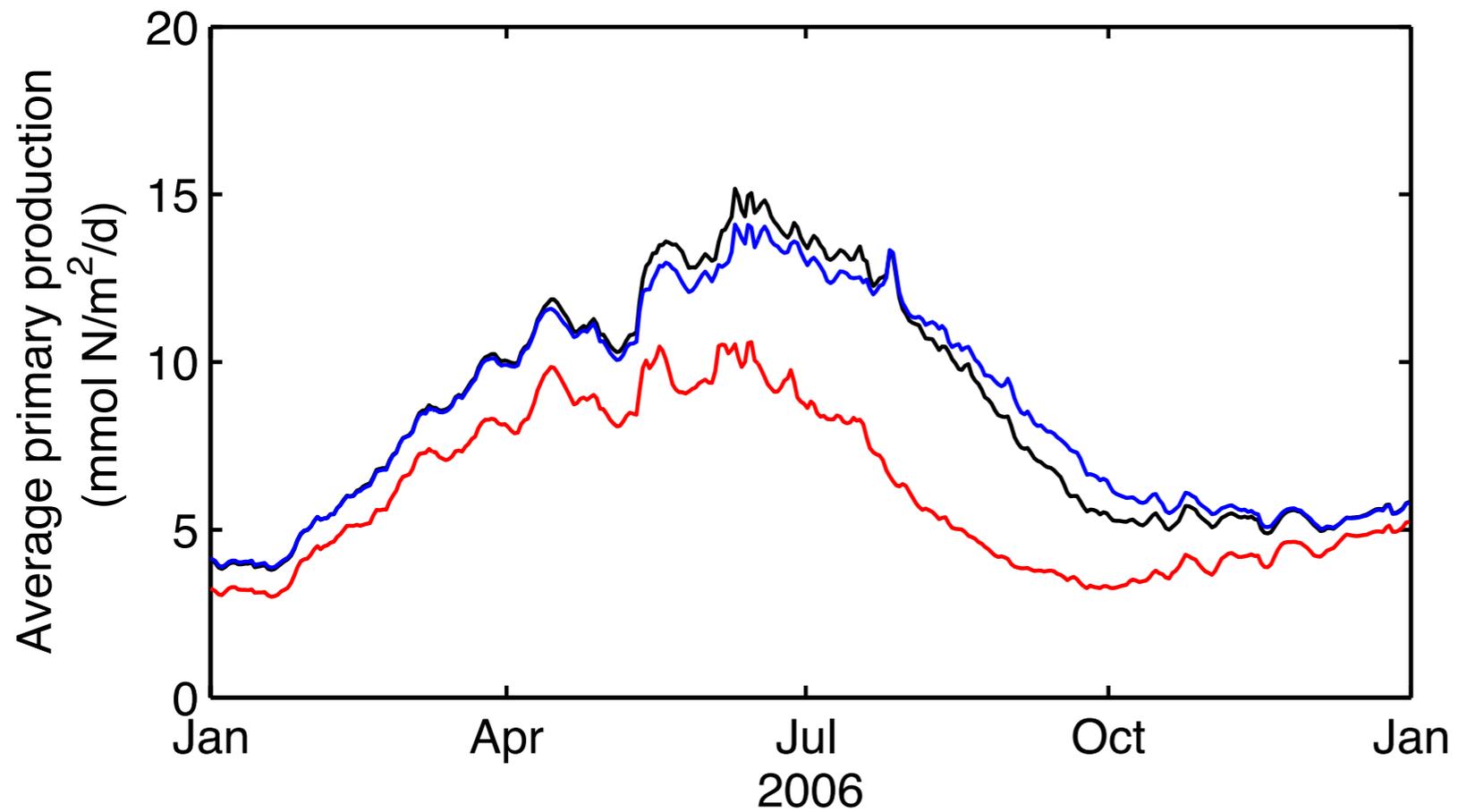
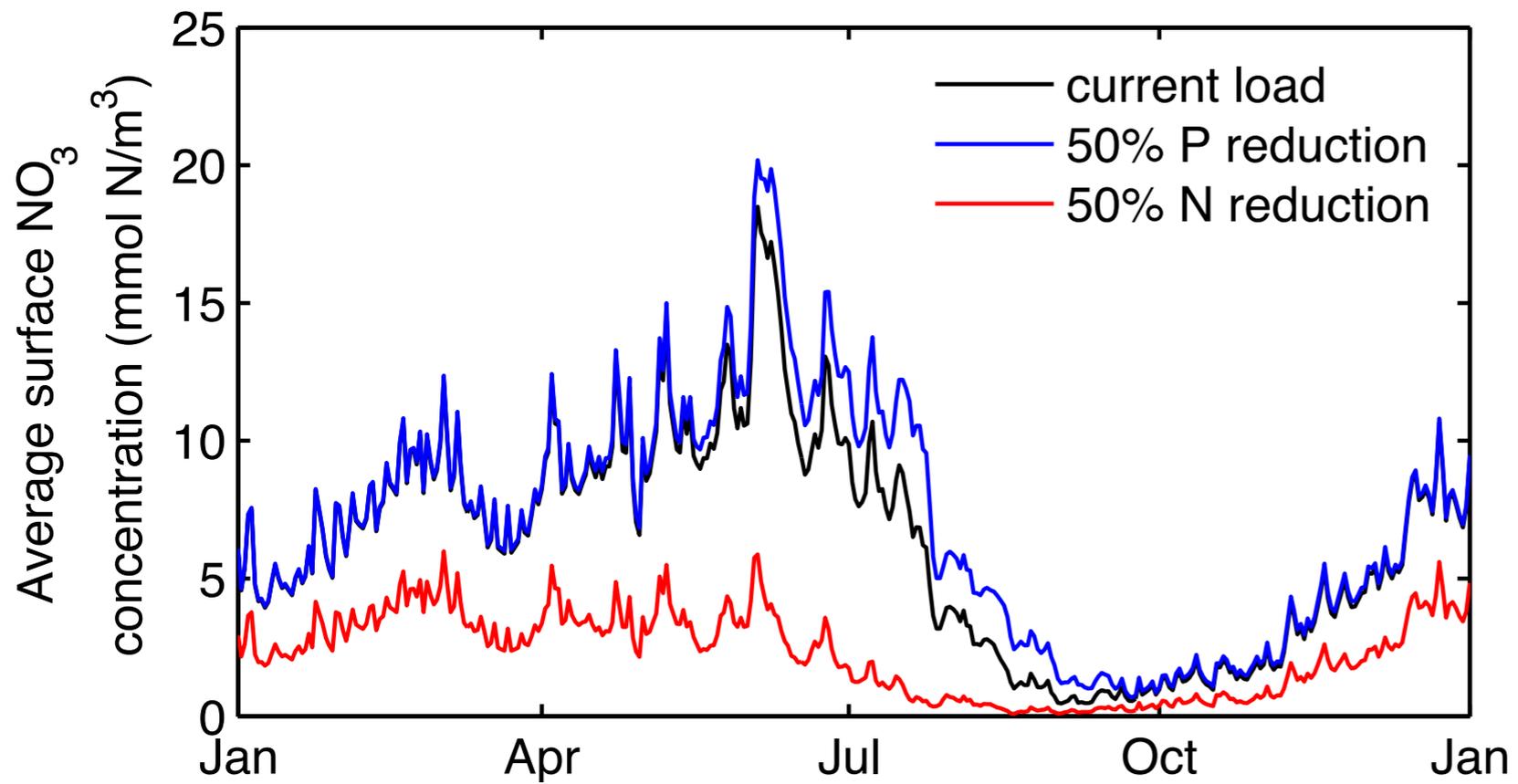
33.1 mol $\text{O}_2/\text{m}^2/\text{yr}$



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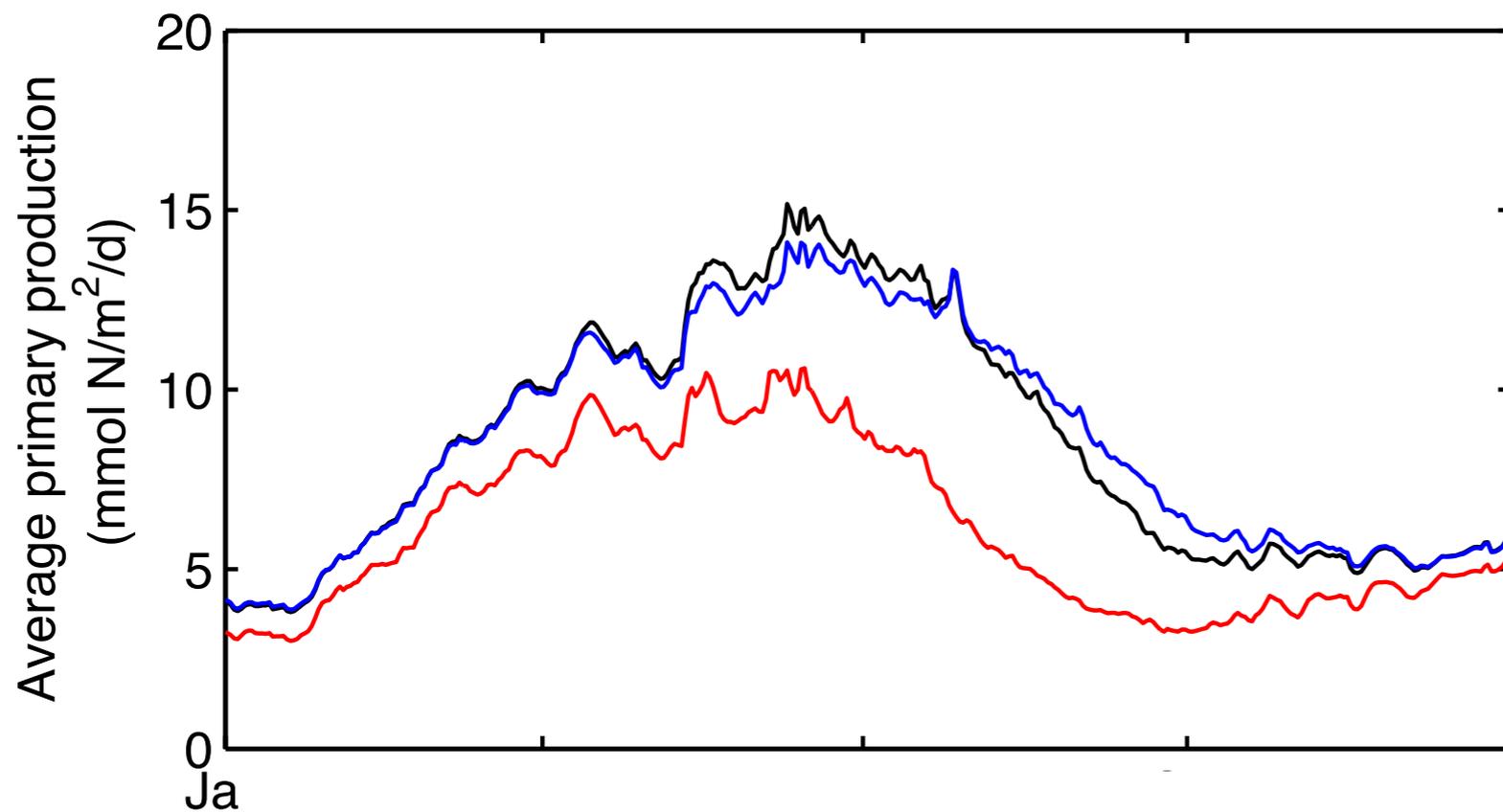
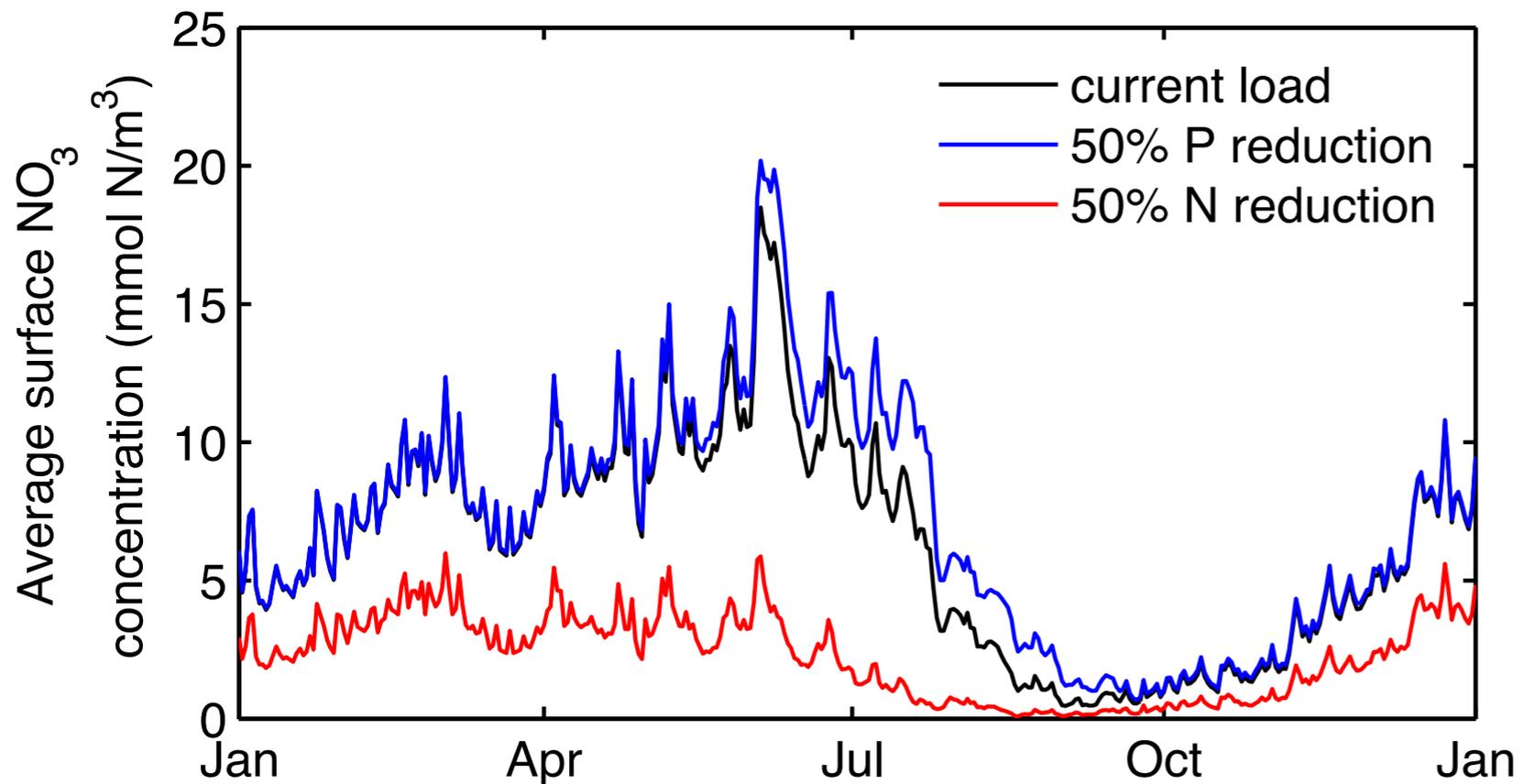


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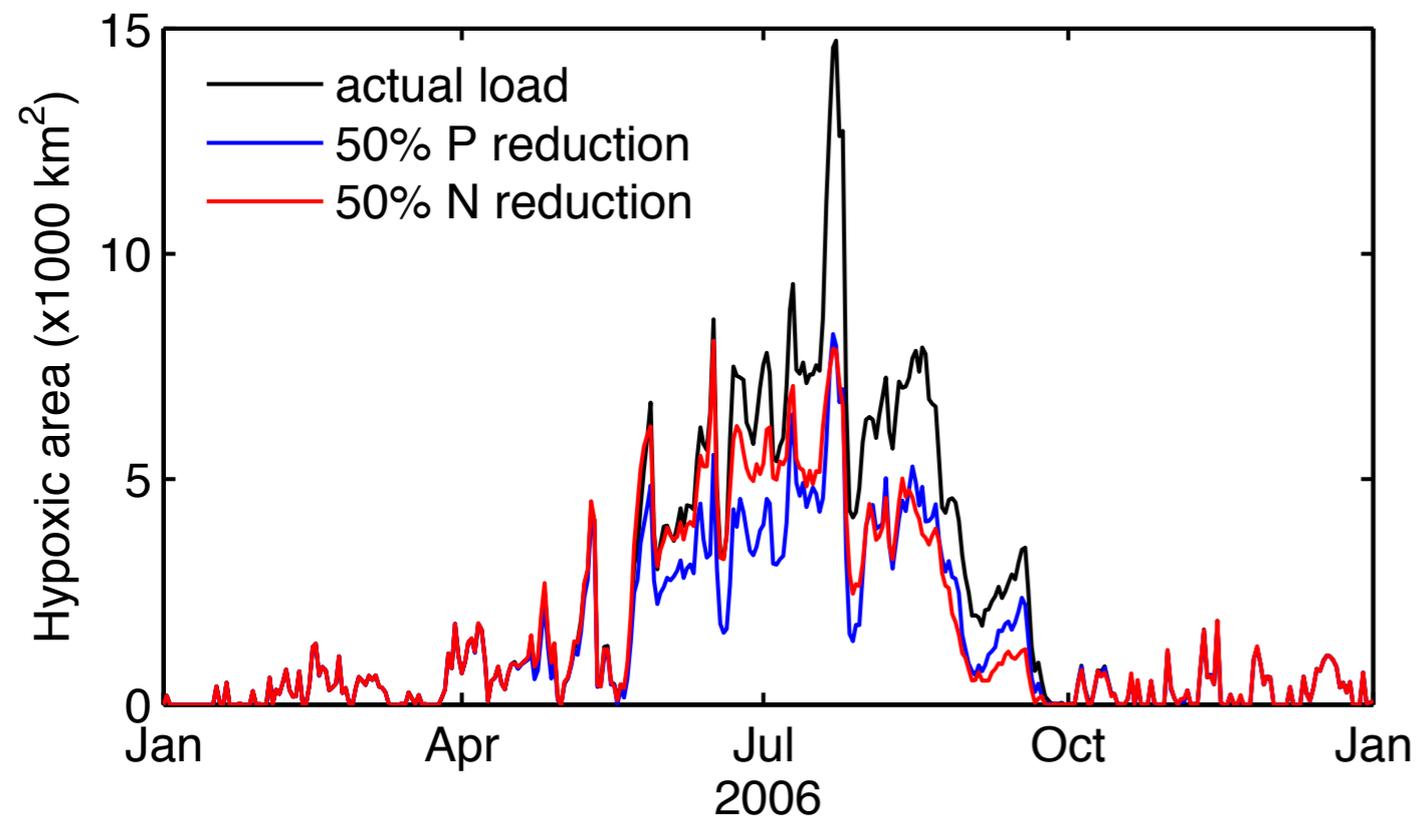
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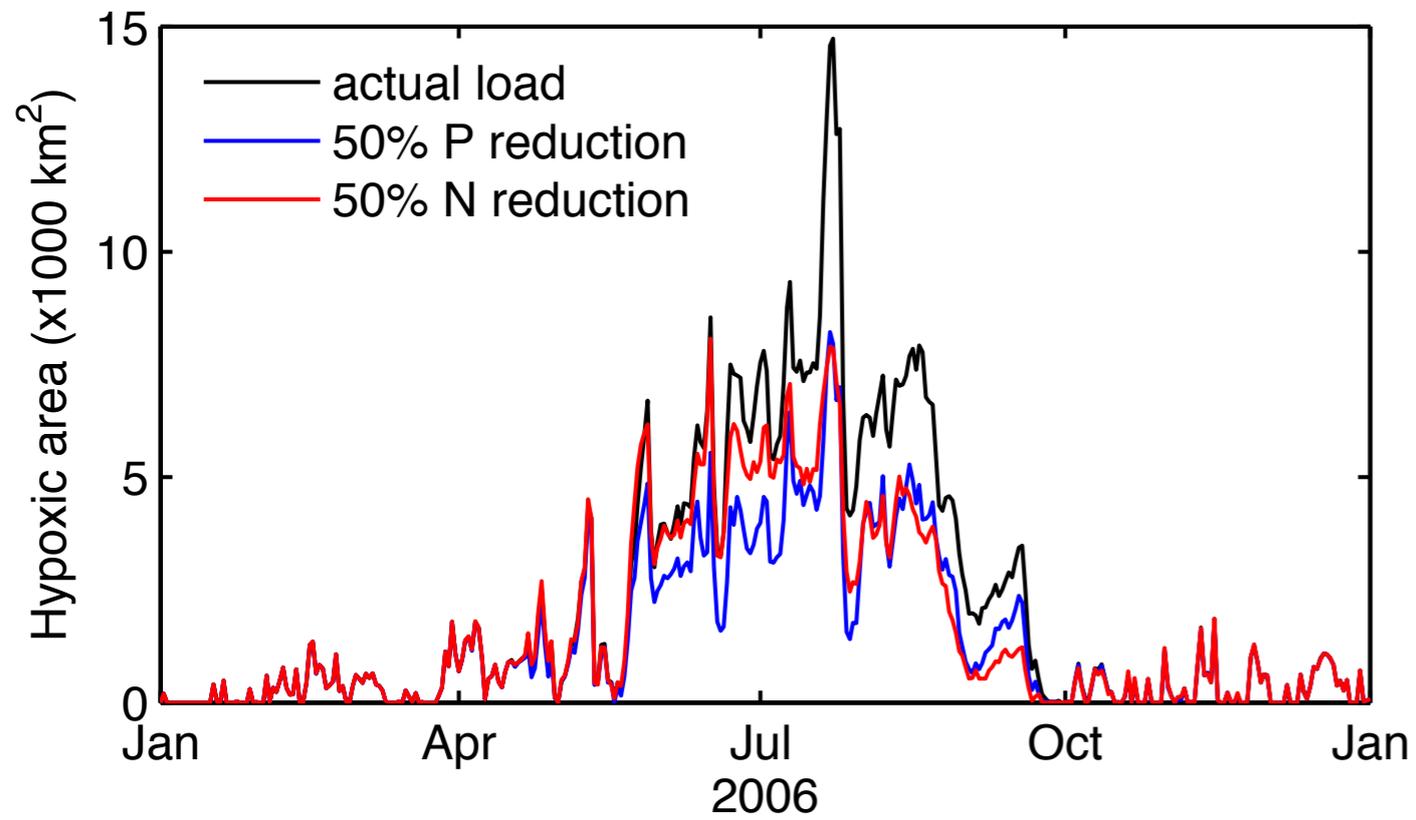
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Model results consistent with N as *ultimate* limiting nutrient and P as limiting in a *proximate* sense.





Integrated hypoxic area (H):

Current load: $2.2 \times 10^3 \text{ km}^2 \text{ yr}$

50% P reduction: $1.43 \times 10^3 \text{ km}^2 \text{ yr}$

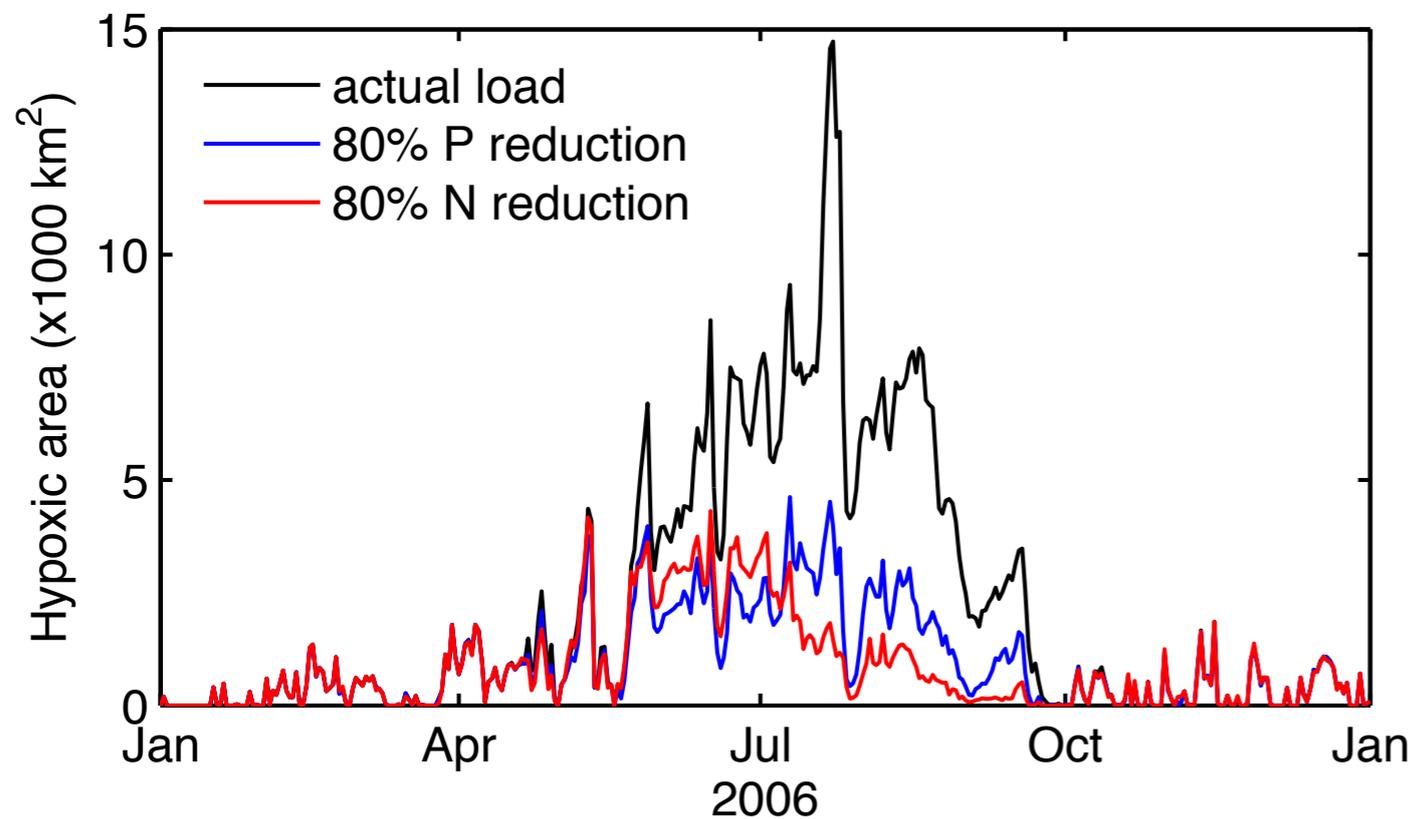
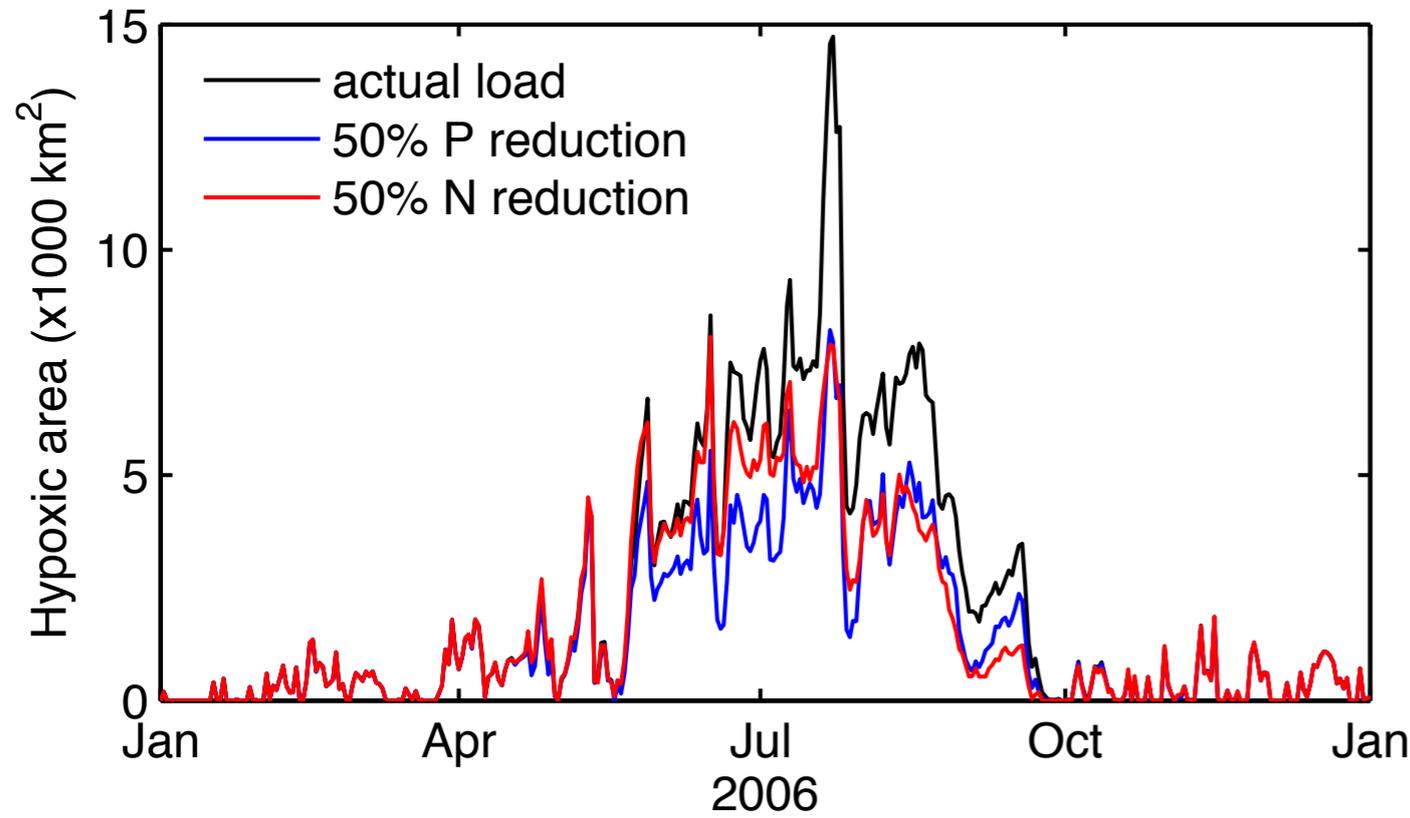
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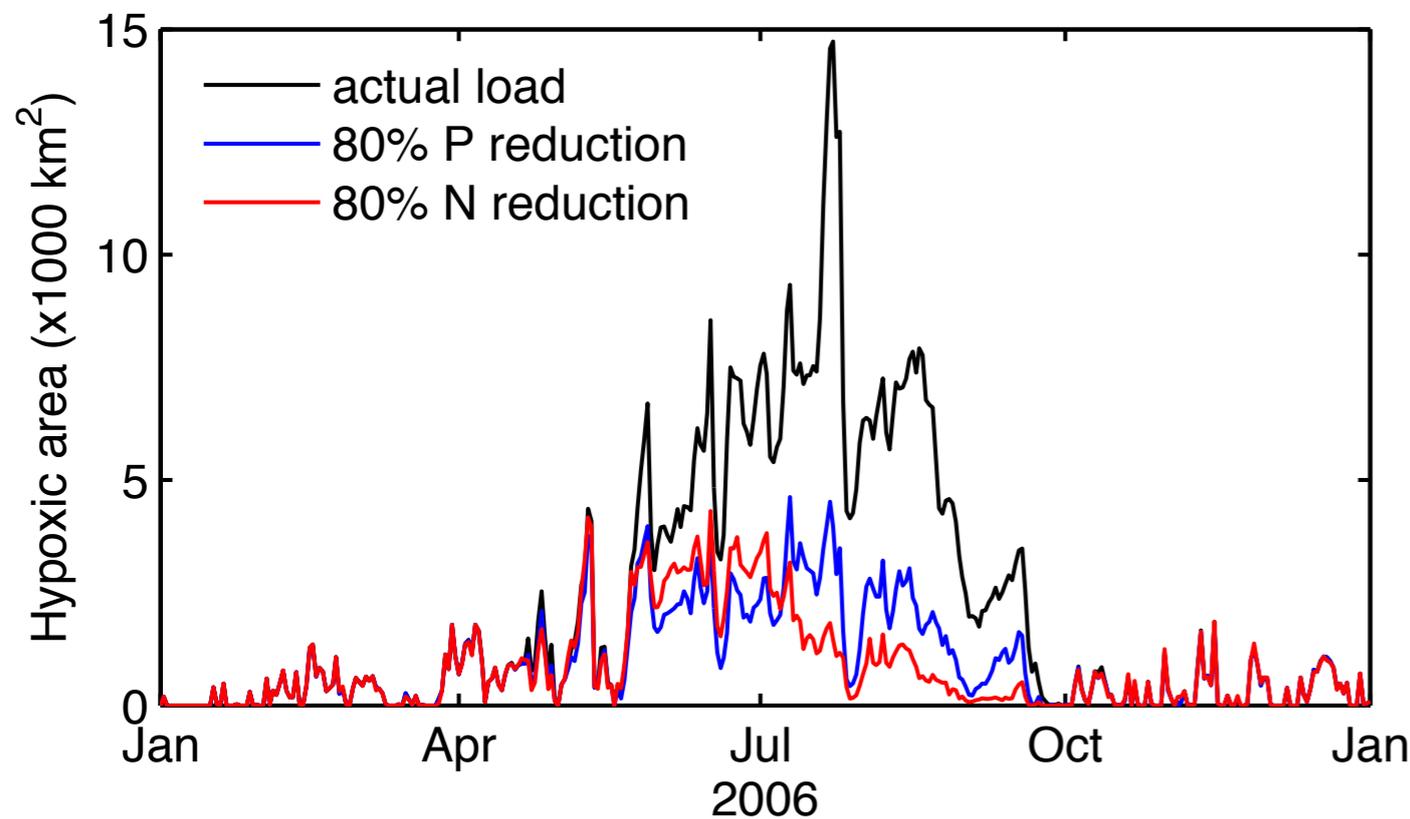
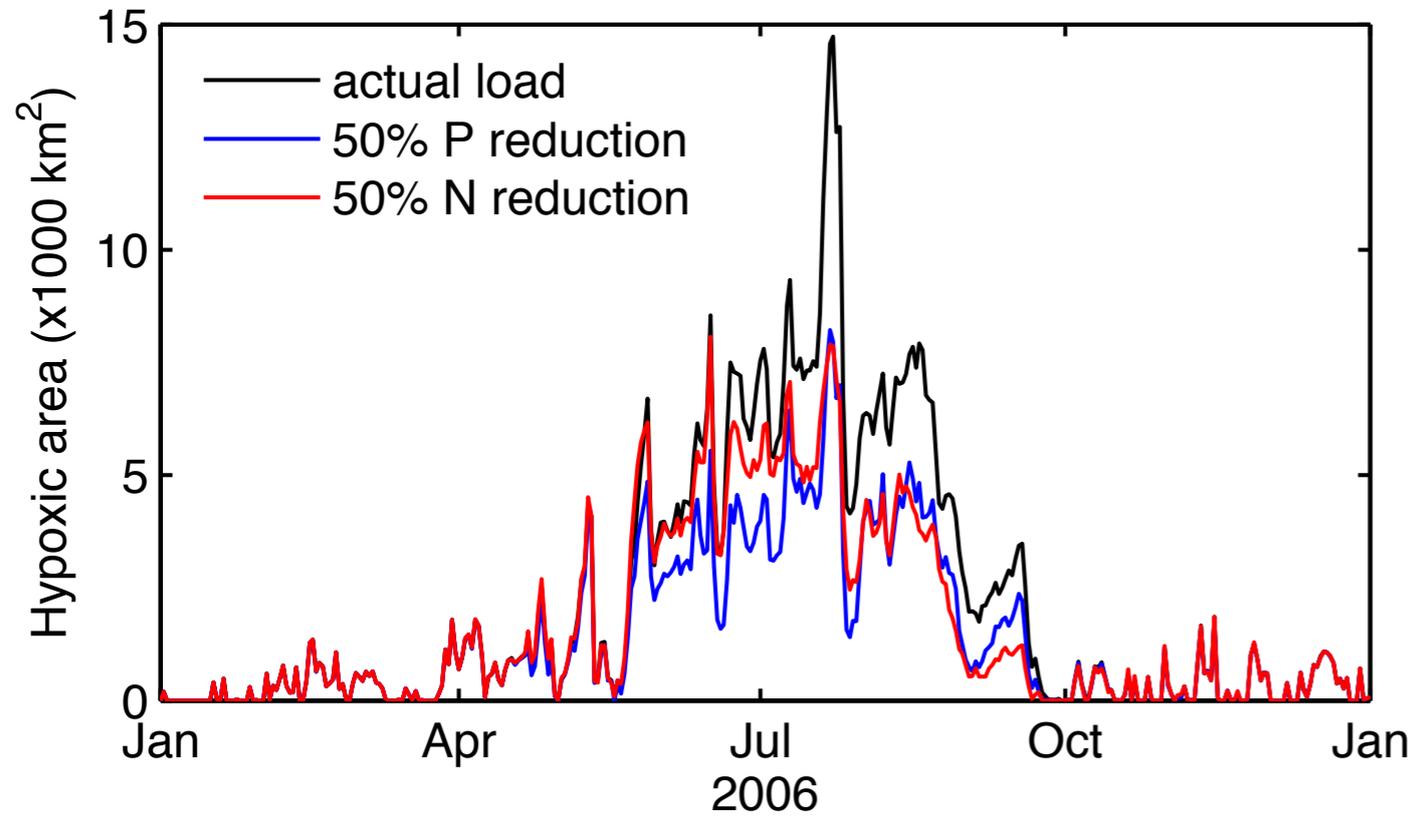


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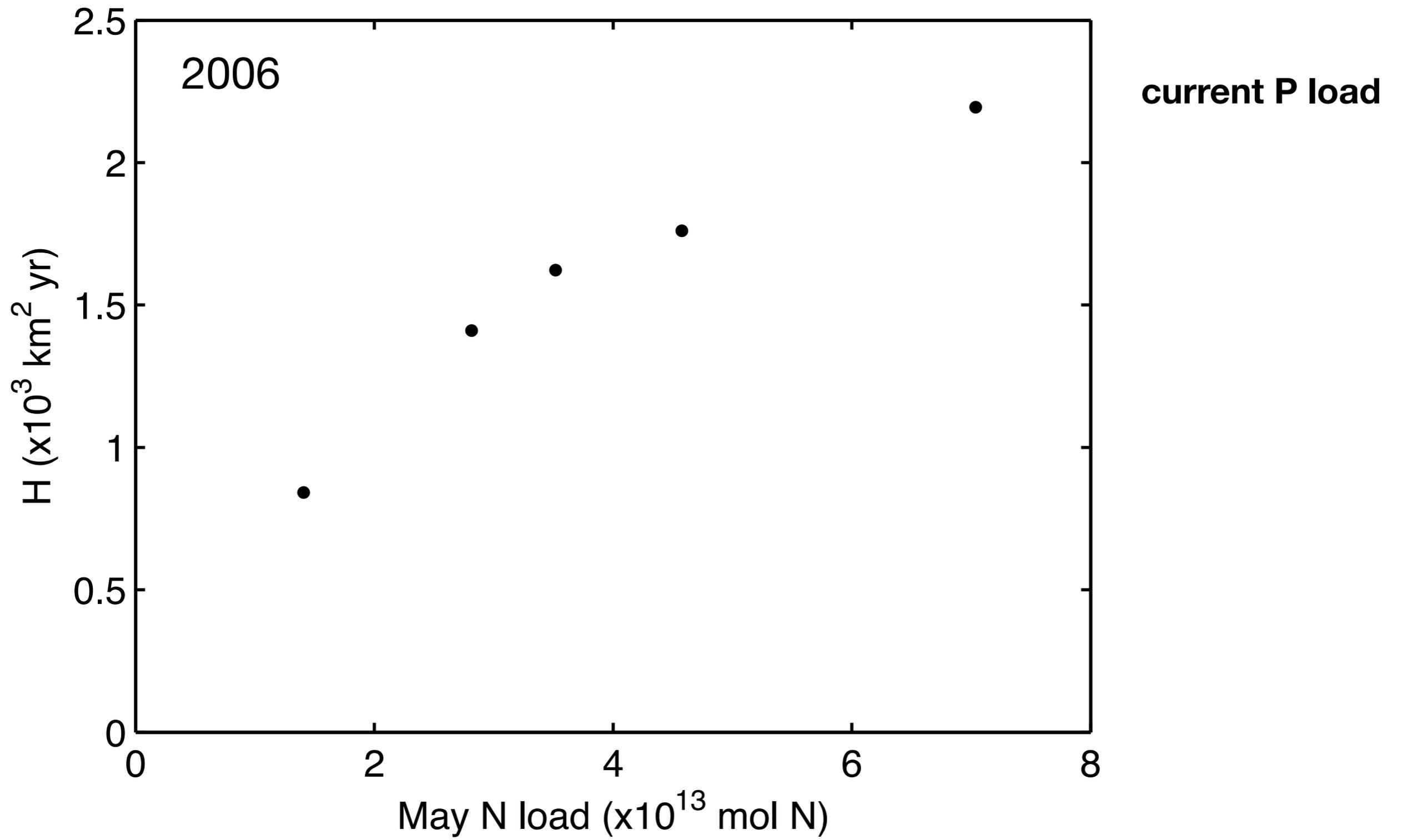
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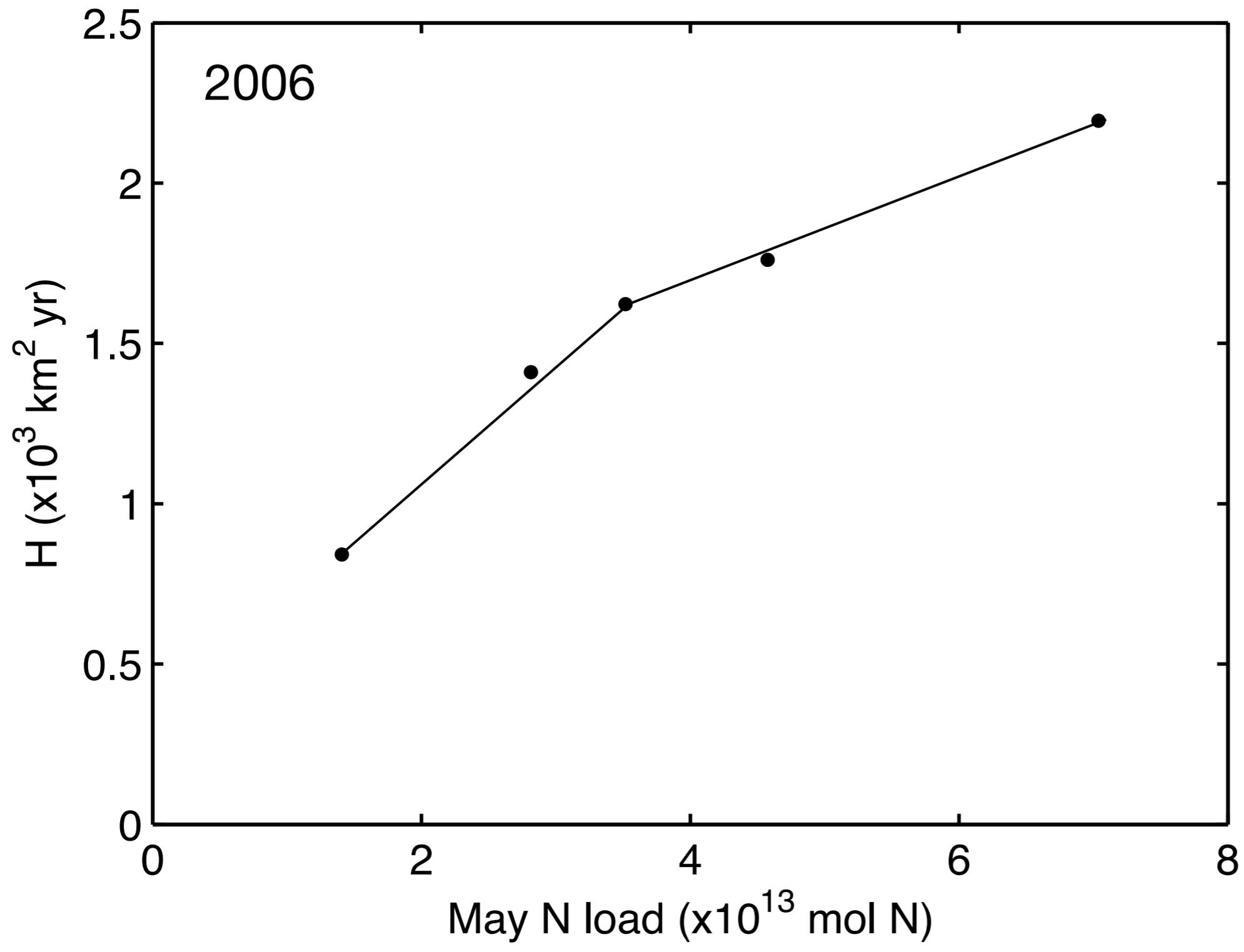
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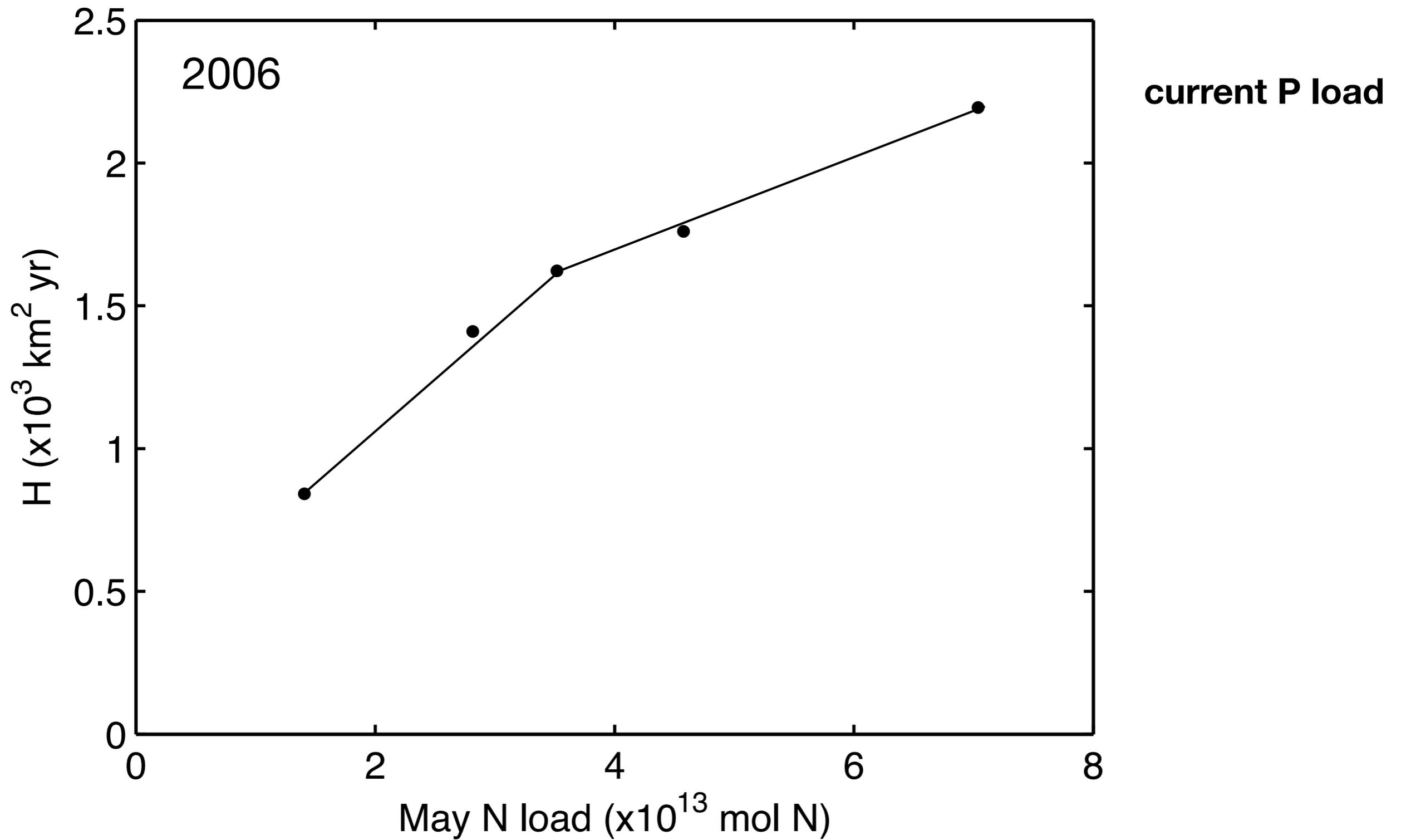


80% P reduction: $0.98 \times 10^3 \text{ km}^2 \text{ yr}$

80% N reduction: $0.84 \times 10^3 \text{ km}^2 \text{ yr}$



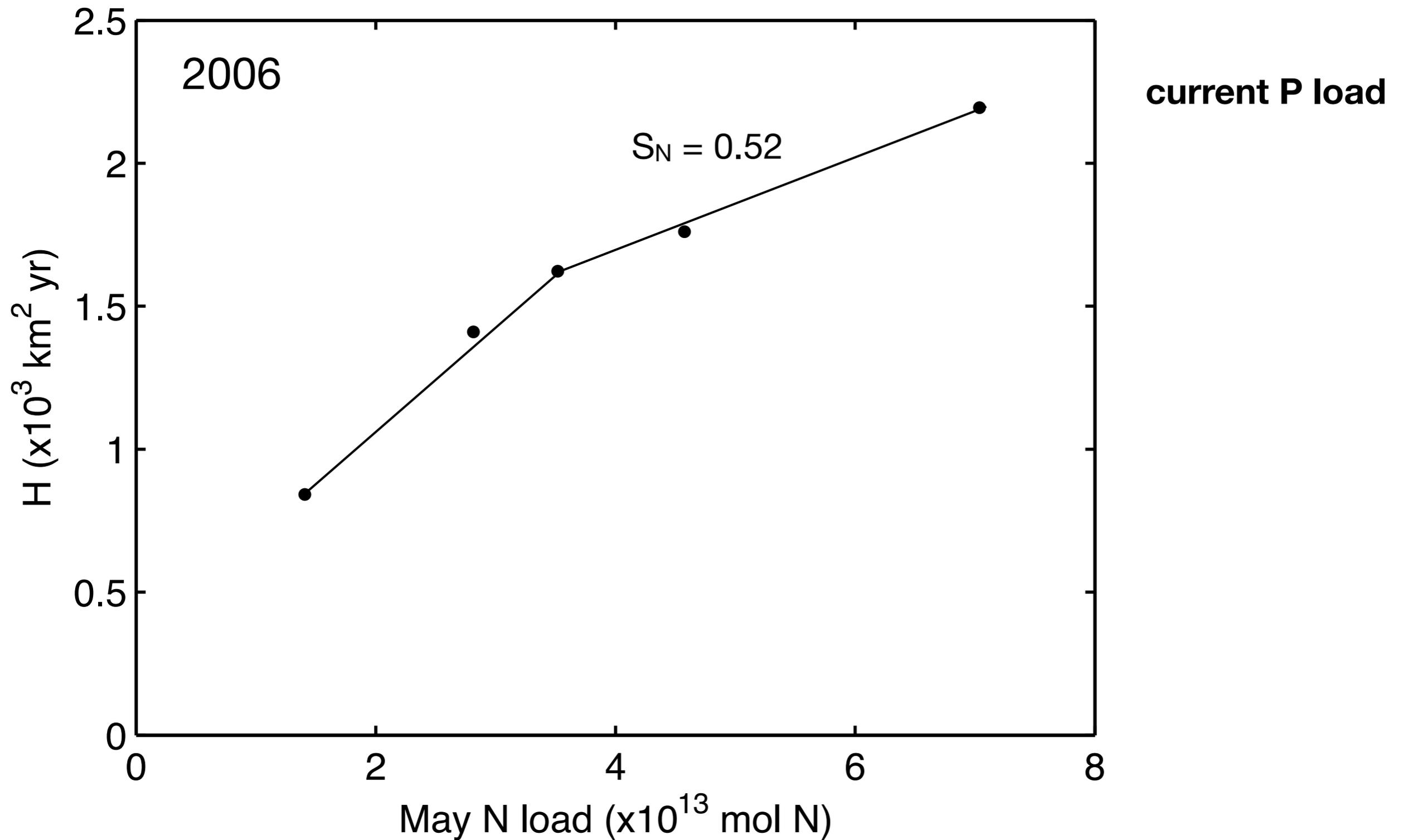




Sensitivity: $S = \Delta H(\%) / \Delta \text{load}(\%)$

For $S = 1$ a 10% reduction in nutrient load will lead to 10% reduction in H.

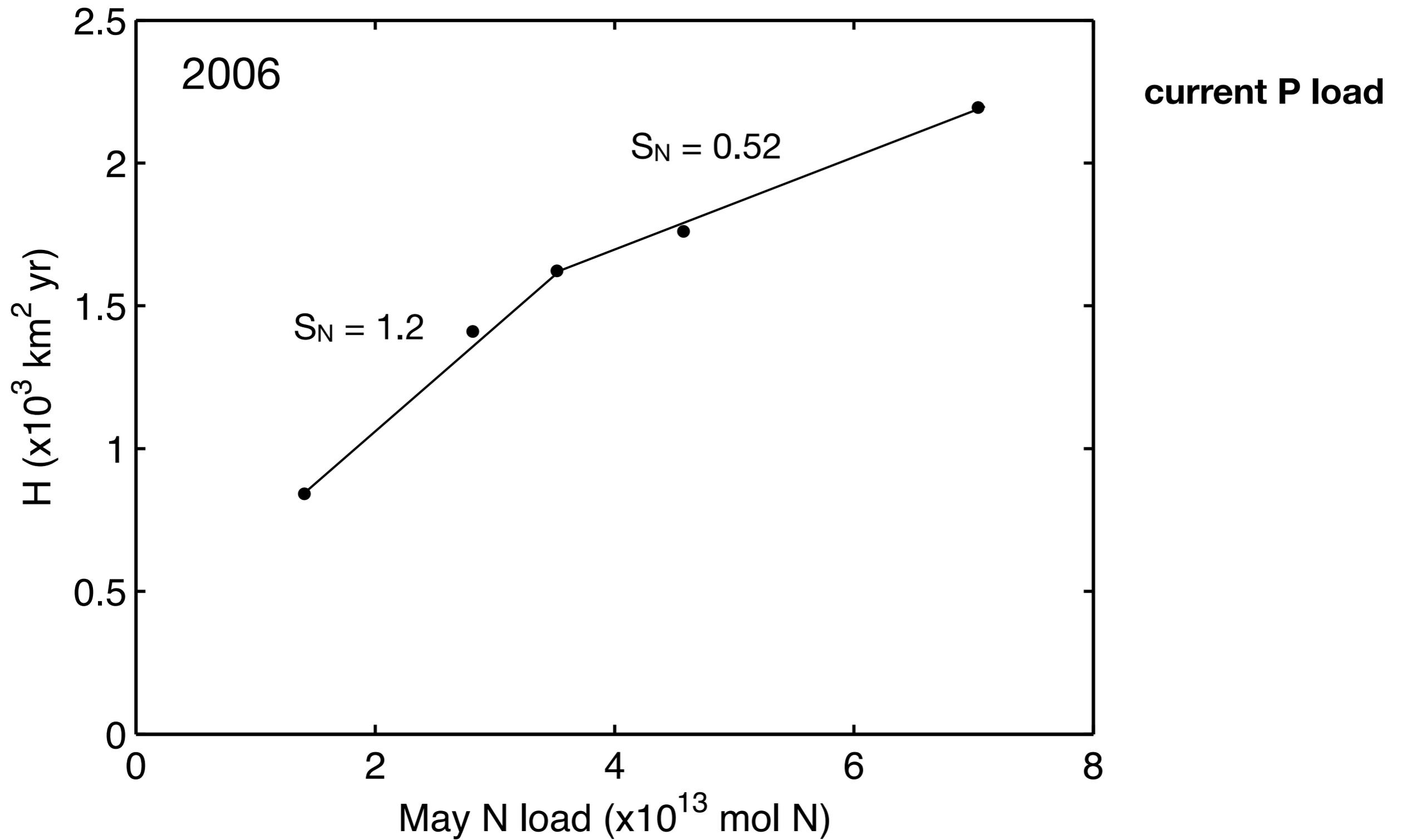
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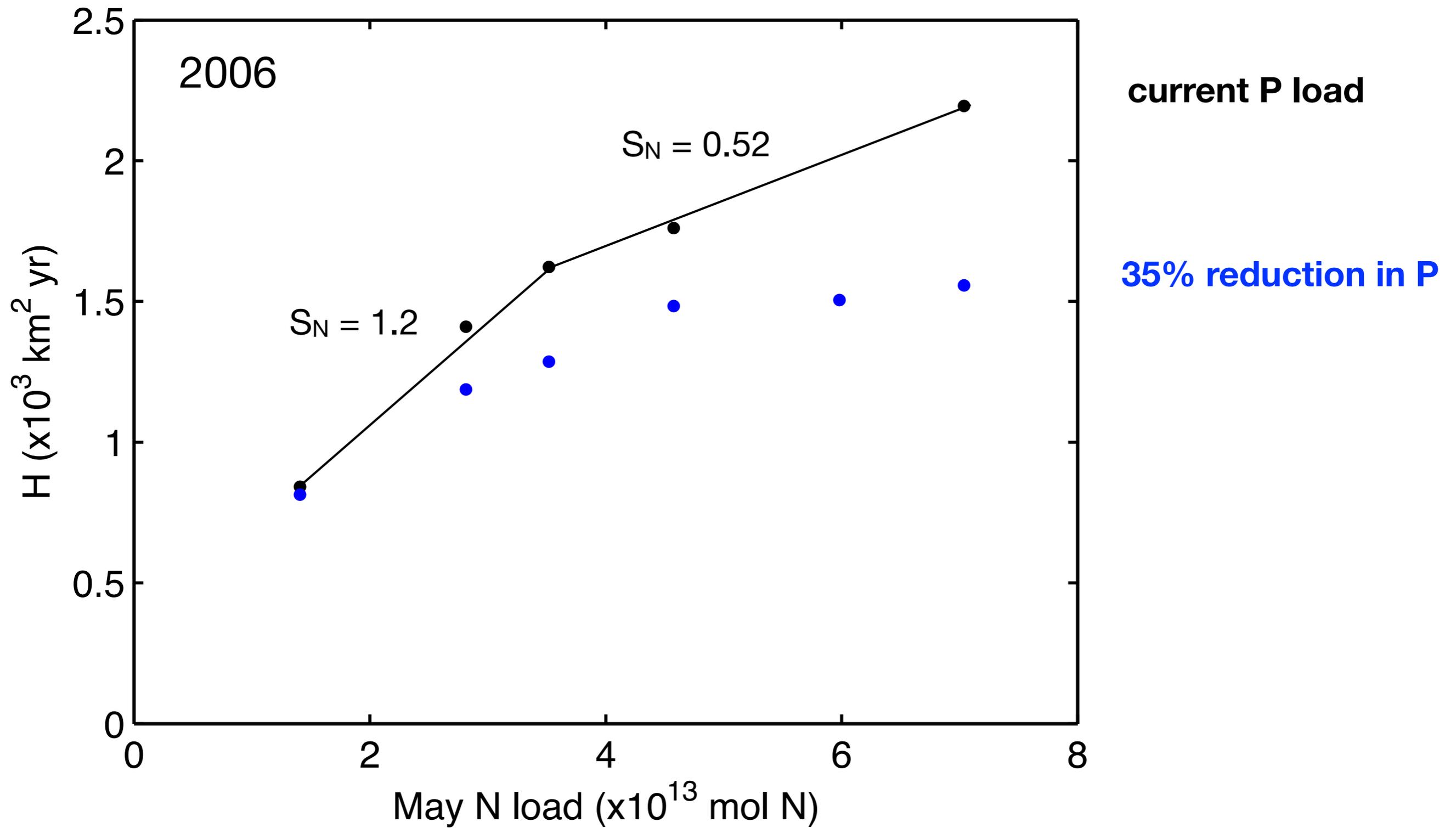
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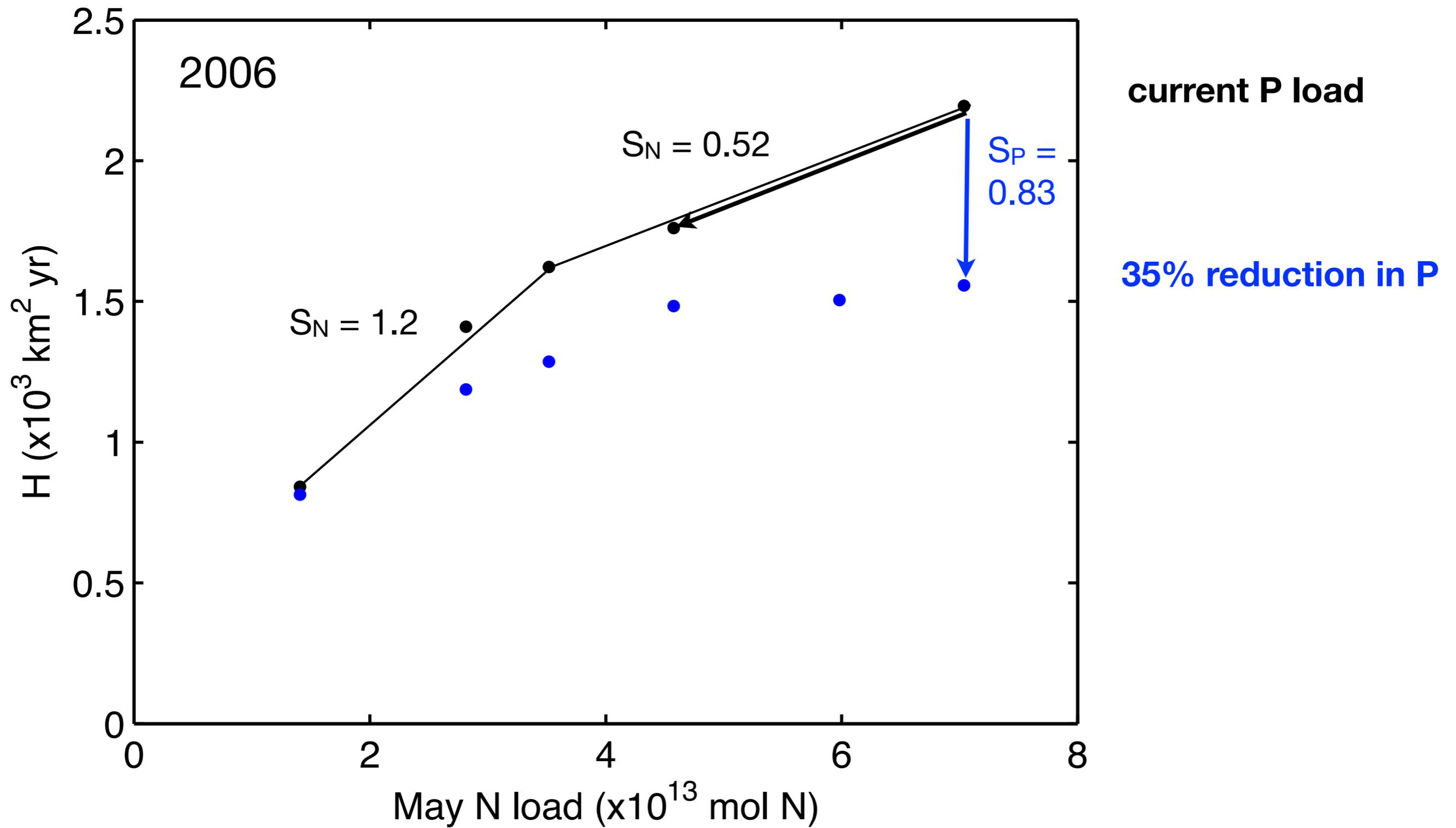
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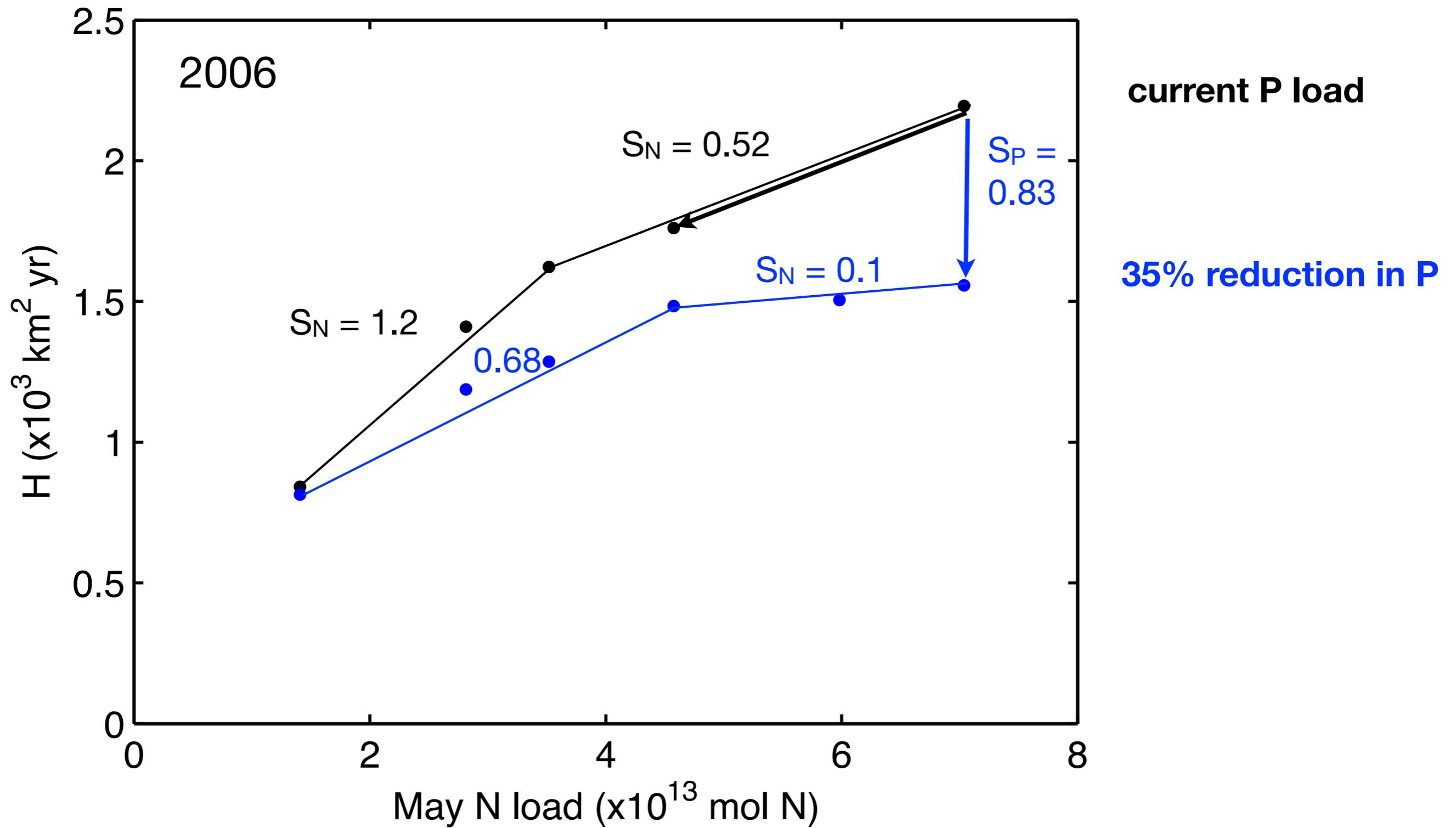
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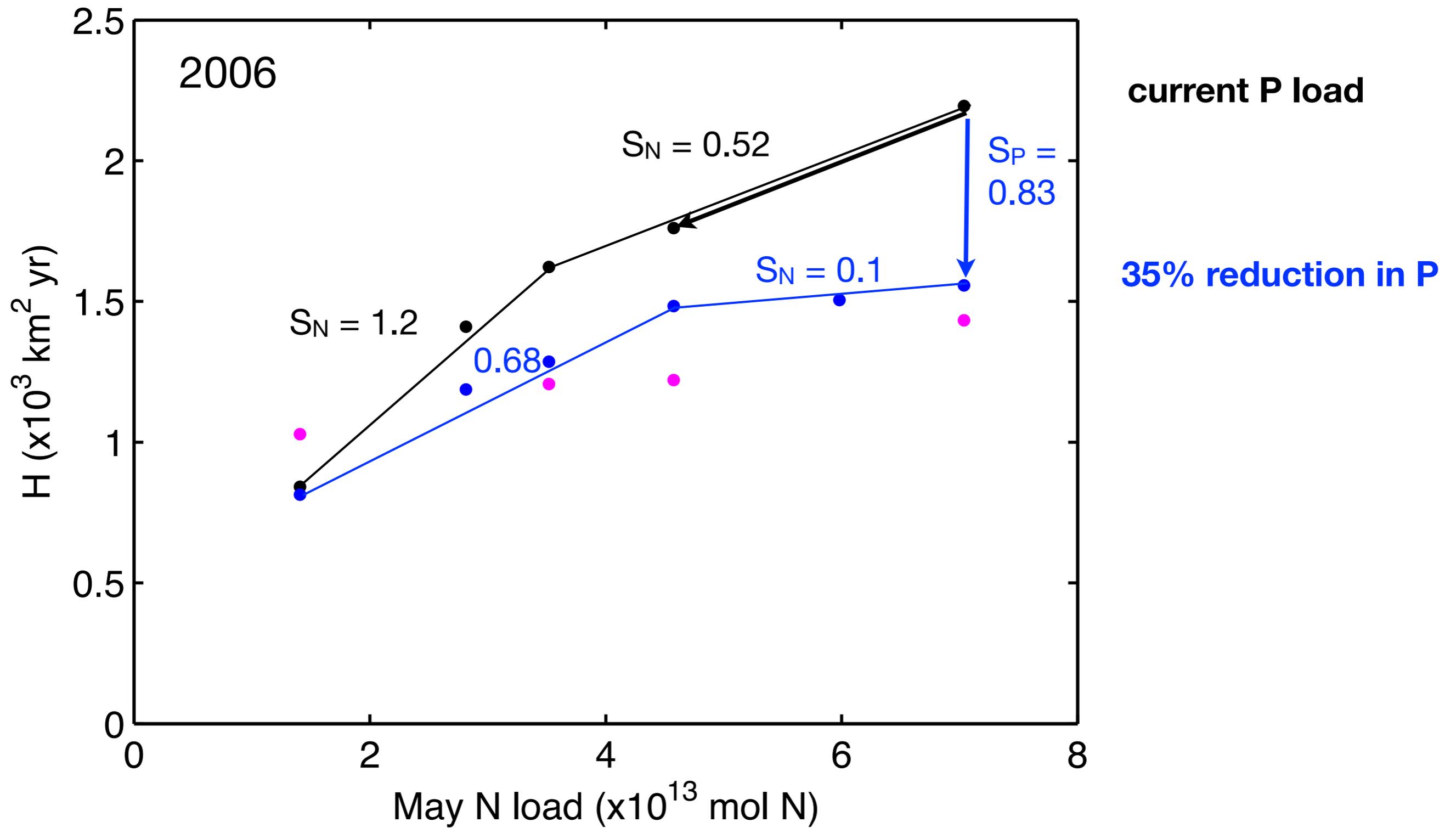
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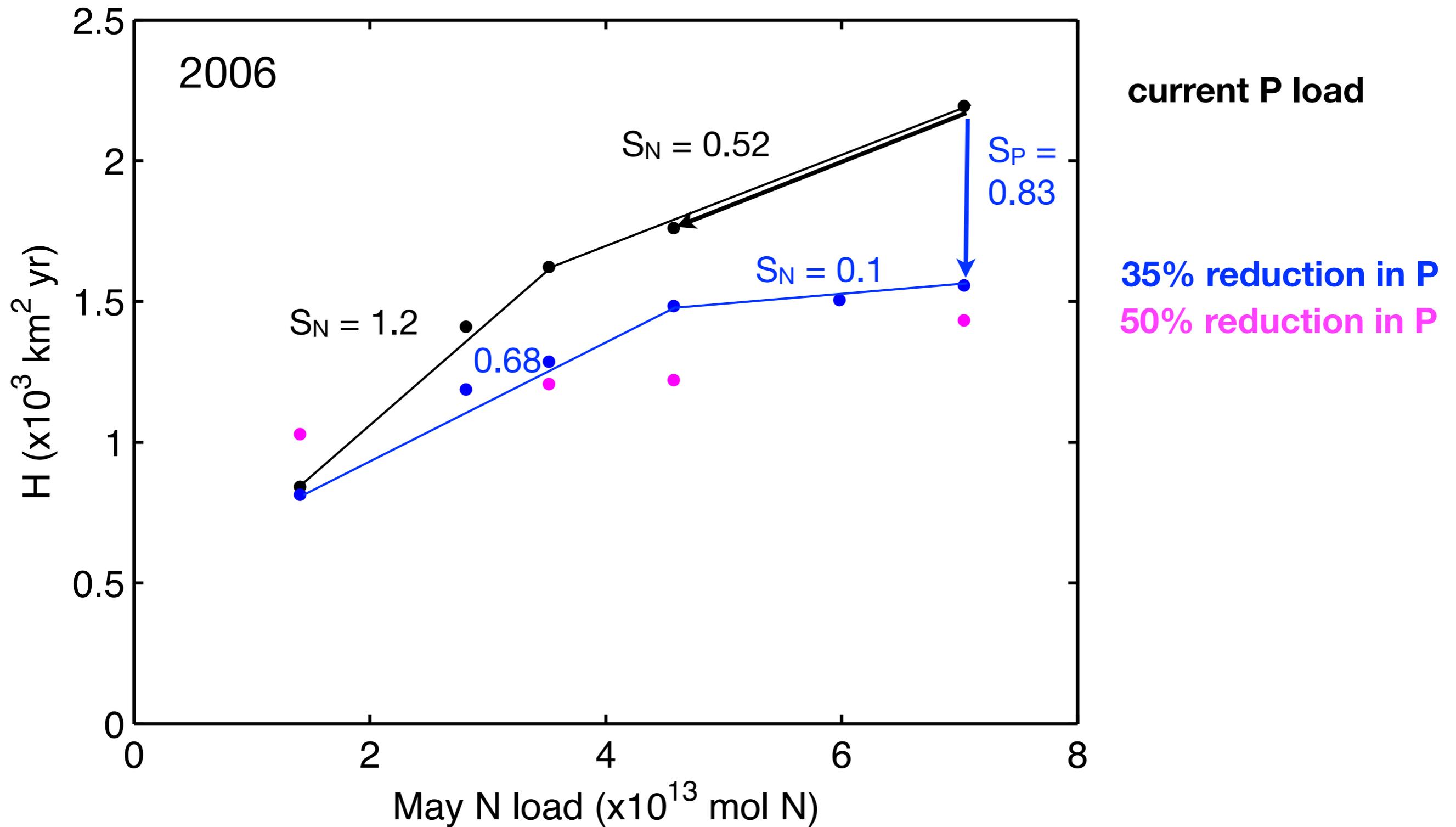
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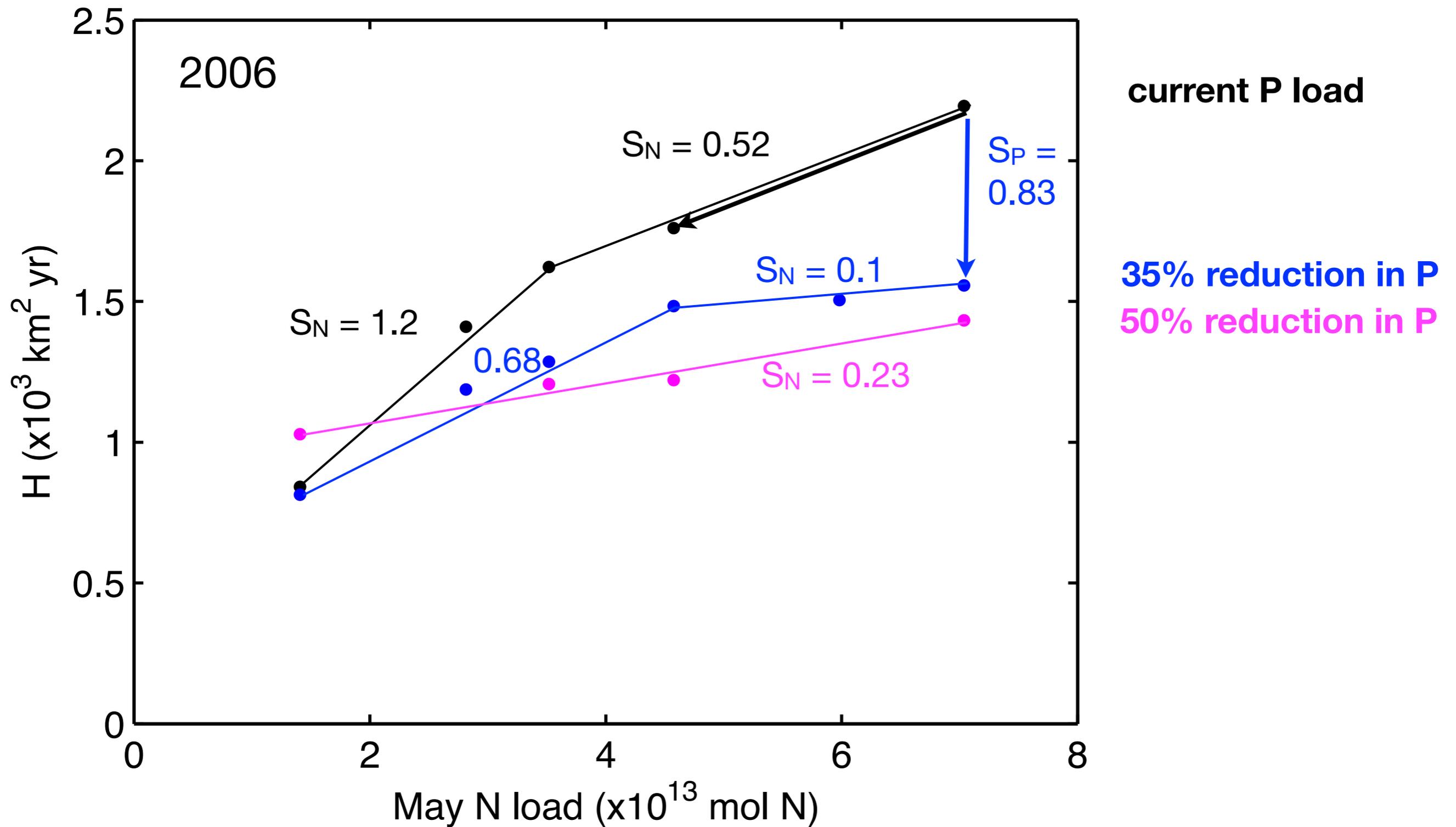
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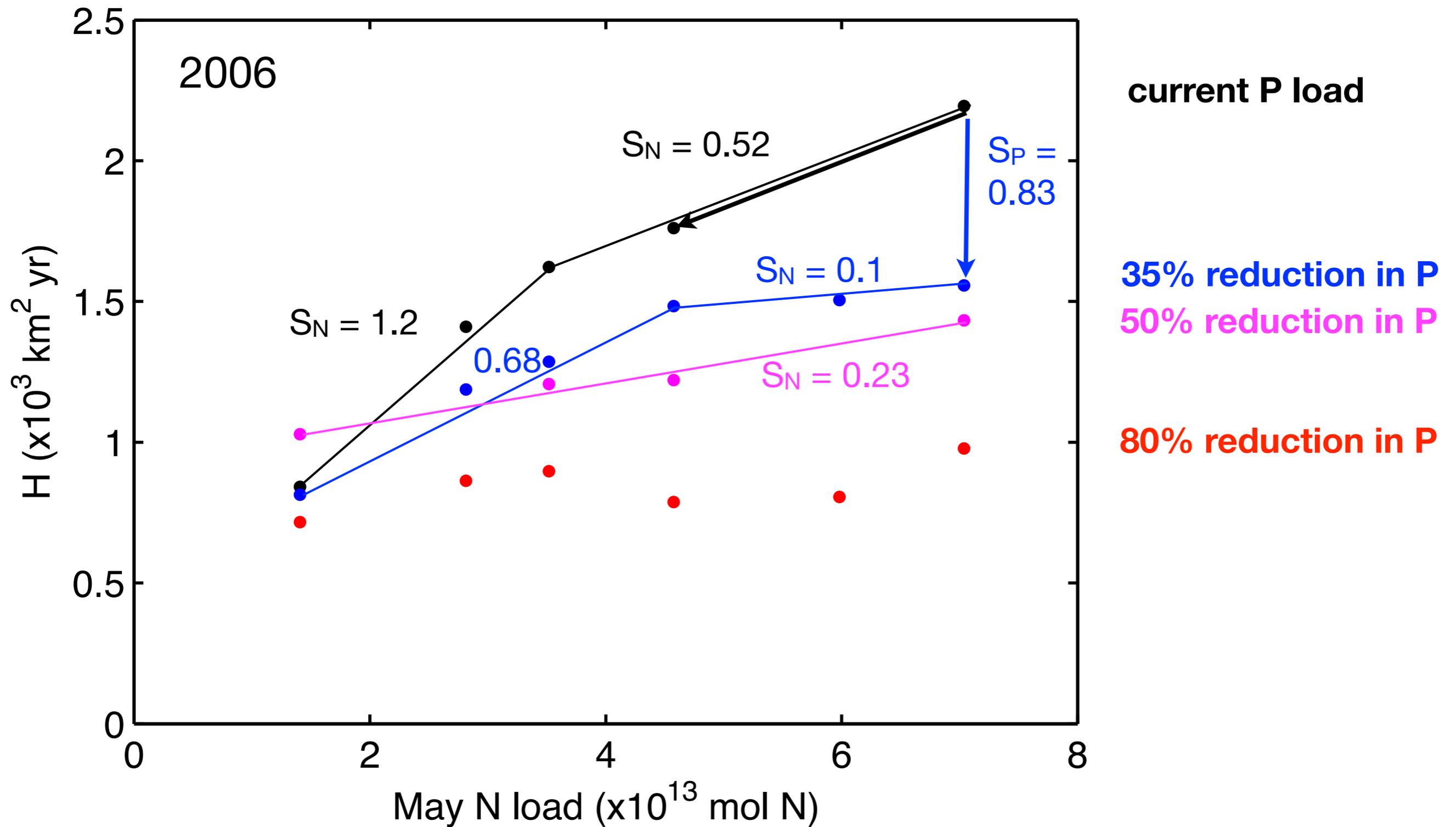
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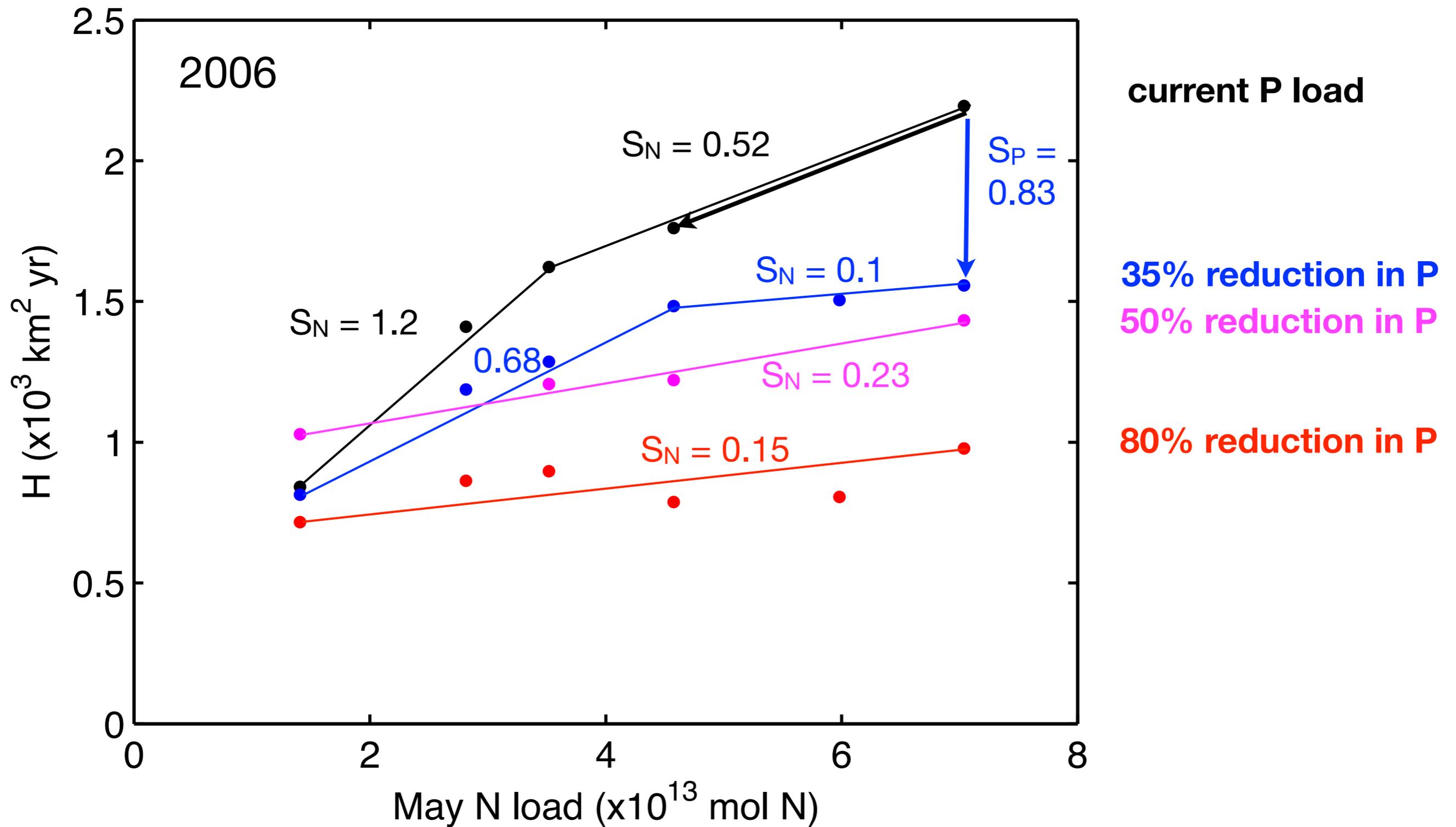
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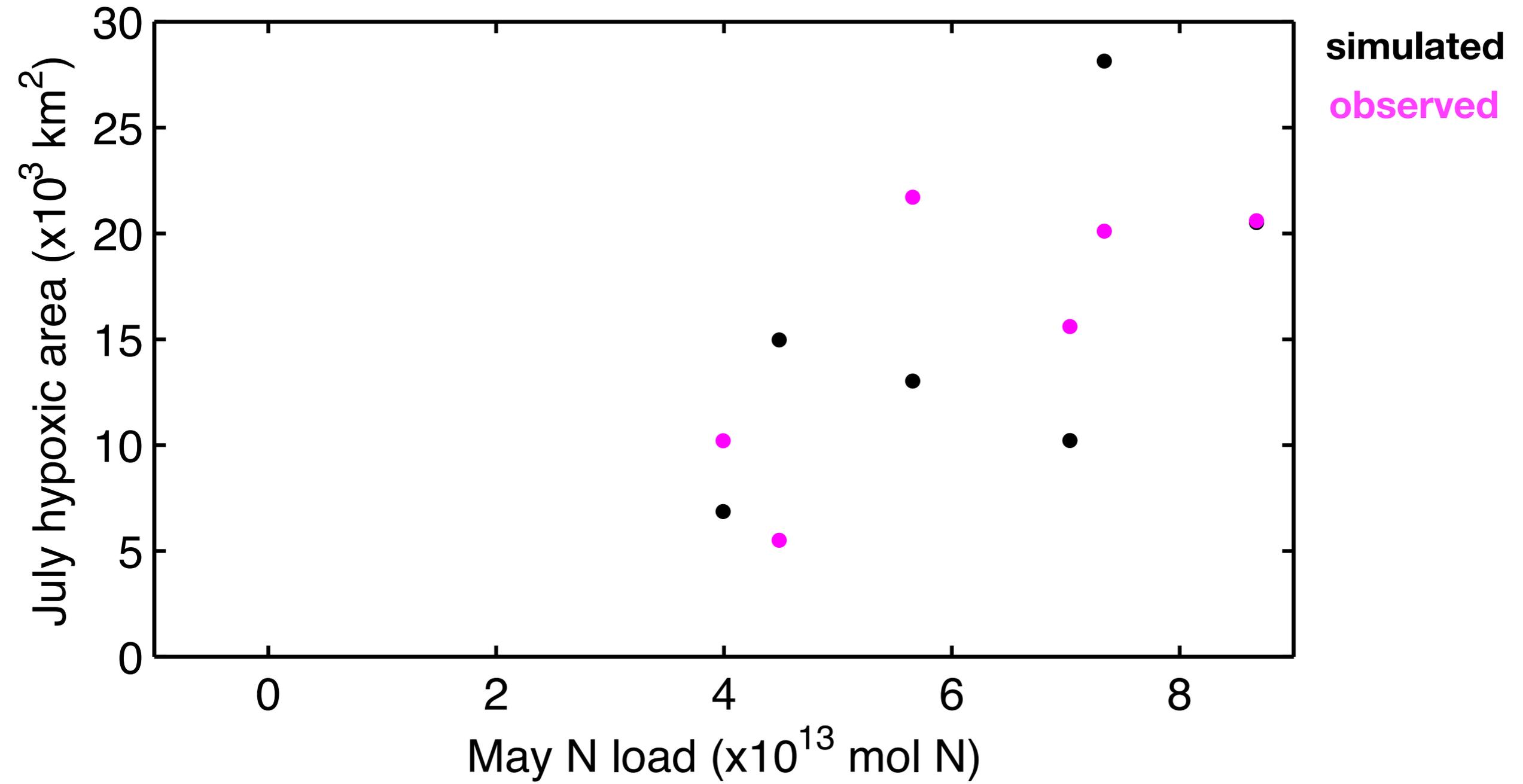


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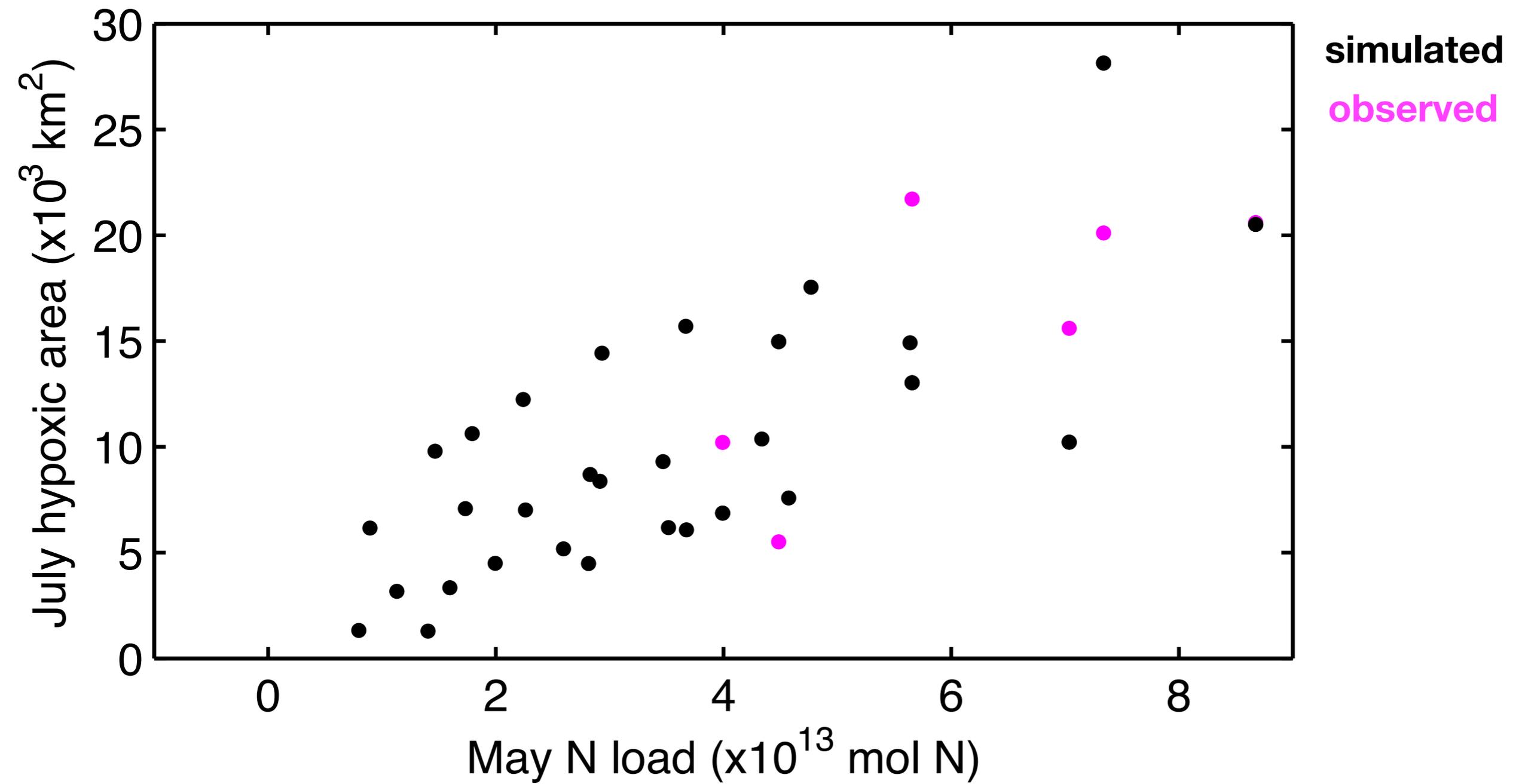
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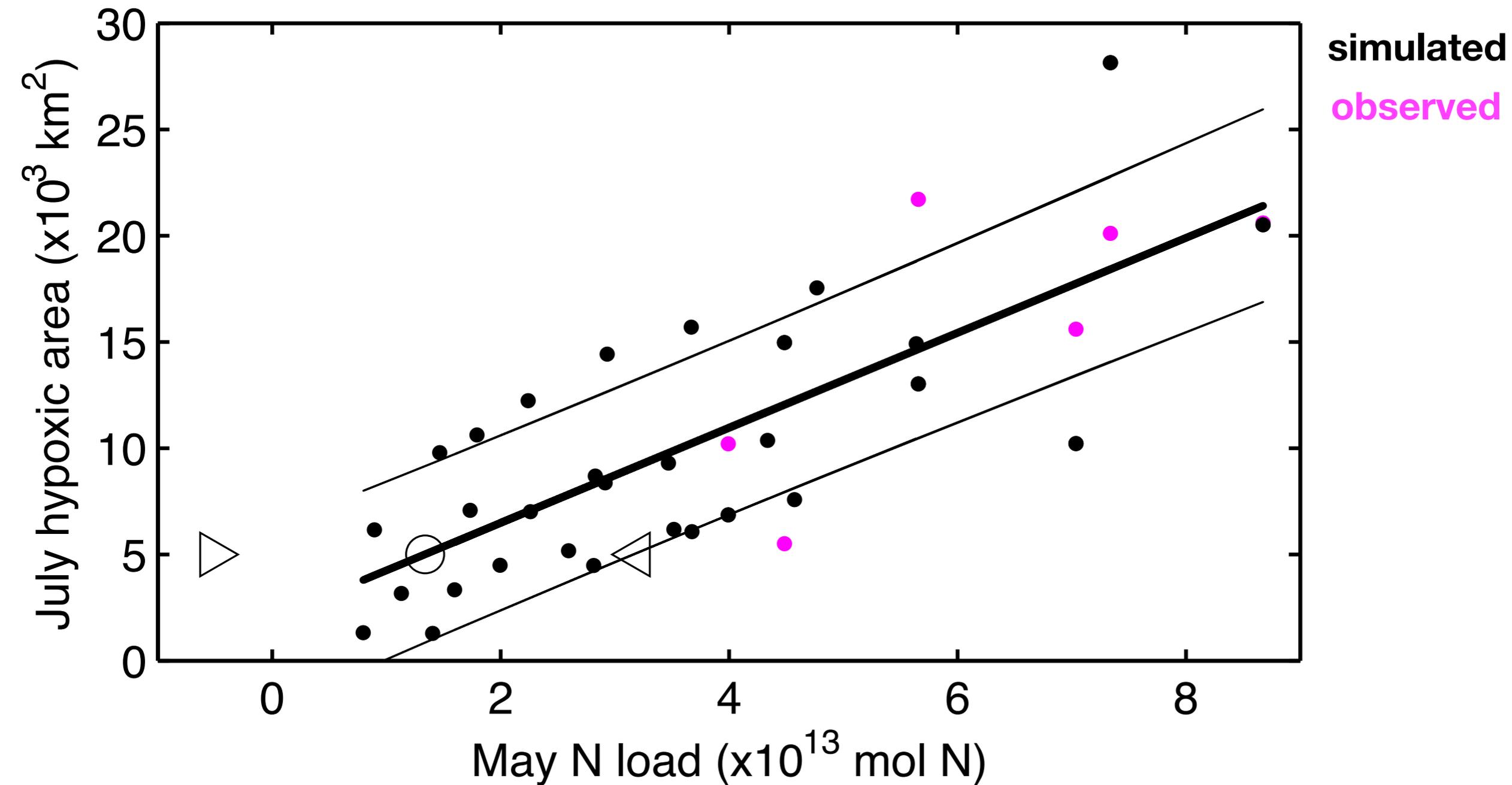
Current loads



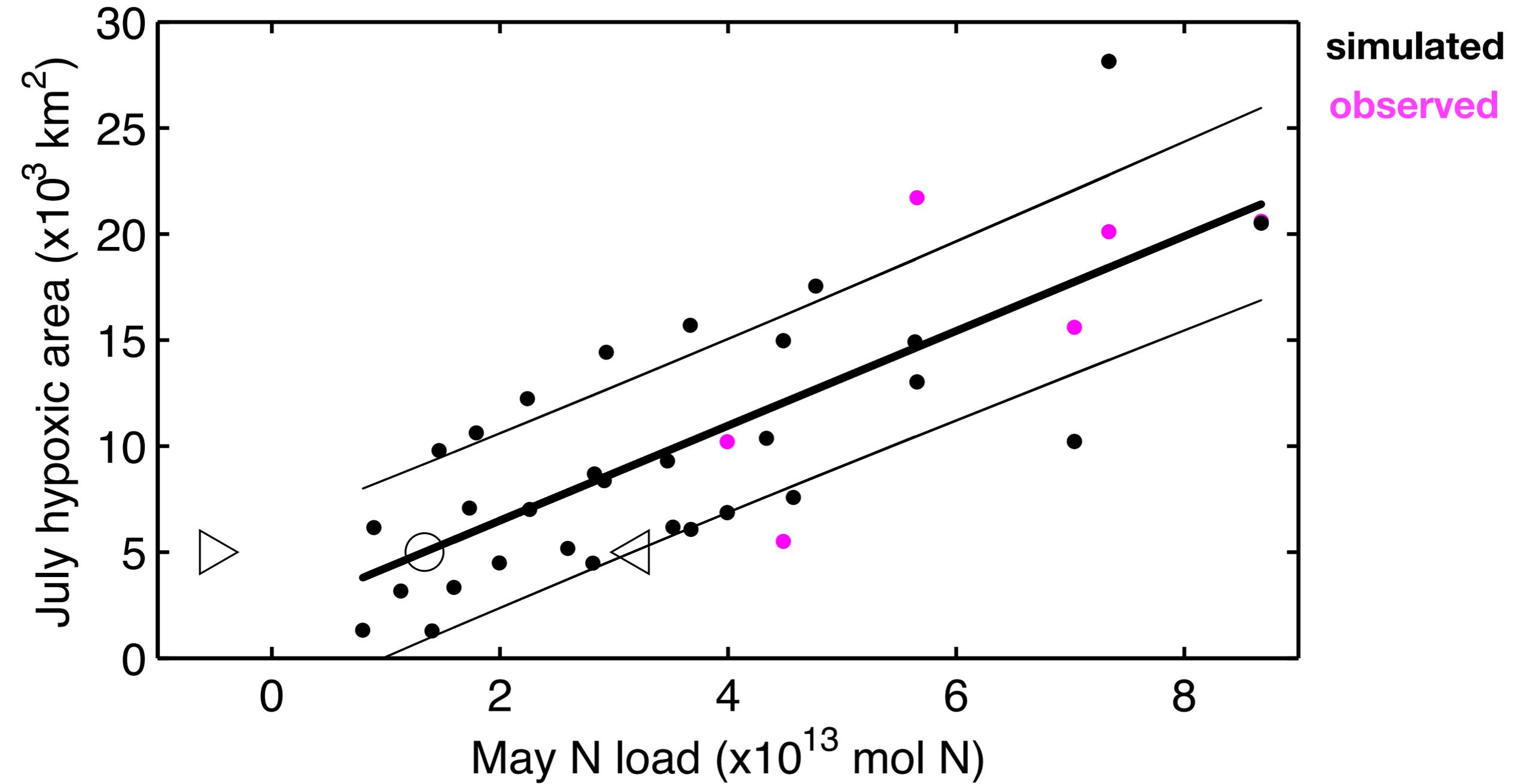
6 years with N load reductions



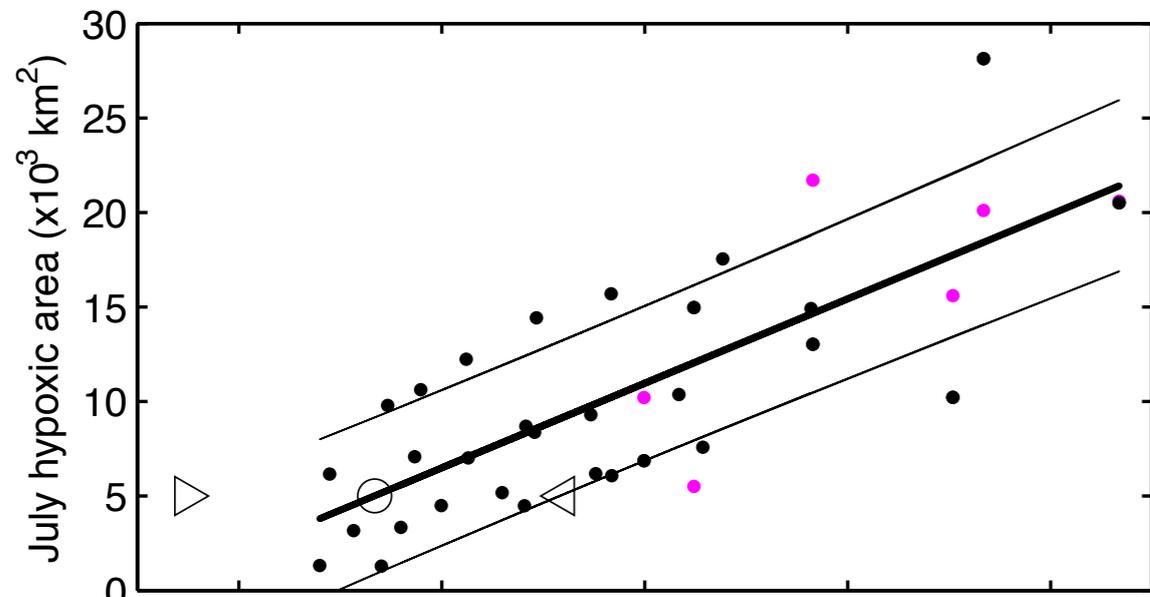
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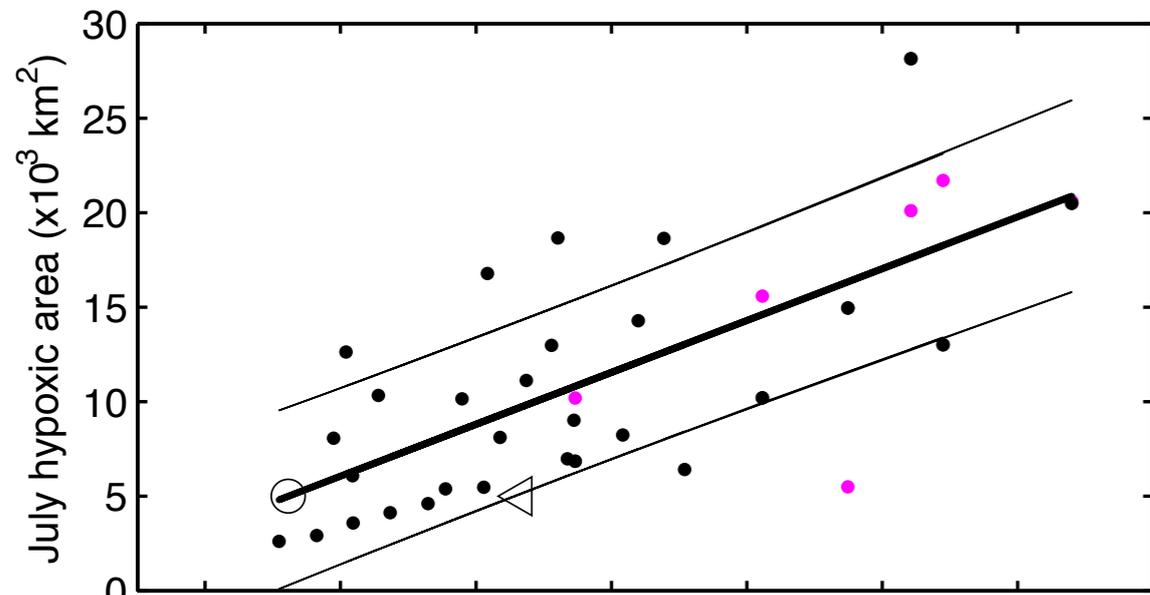
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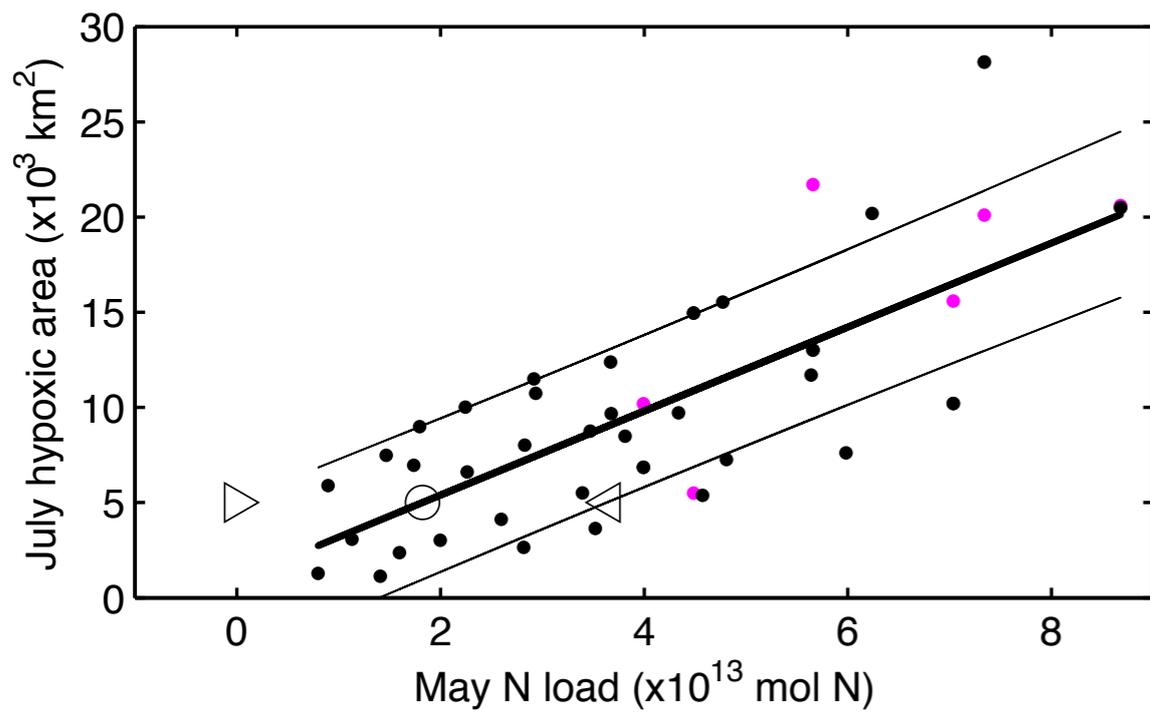
Need 78% +/- 30% N load reduction to reach 5,000 km².



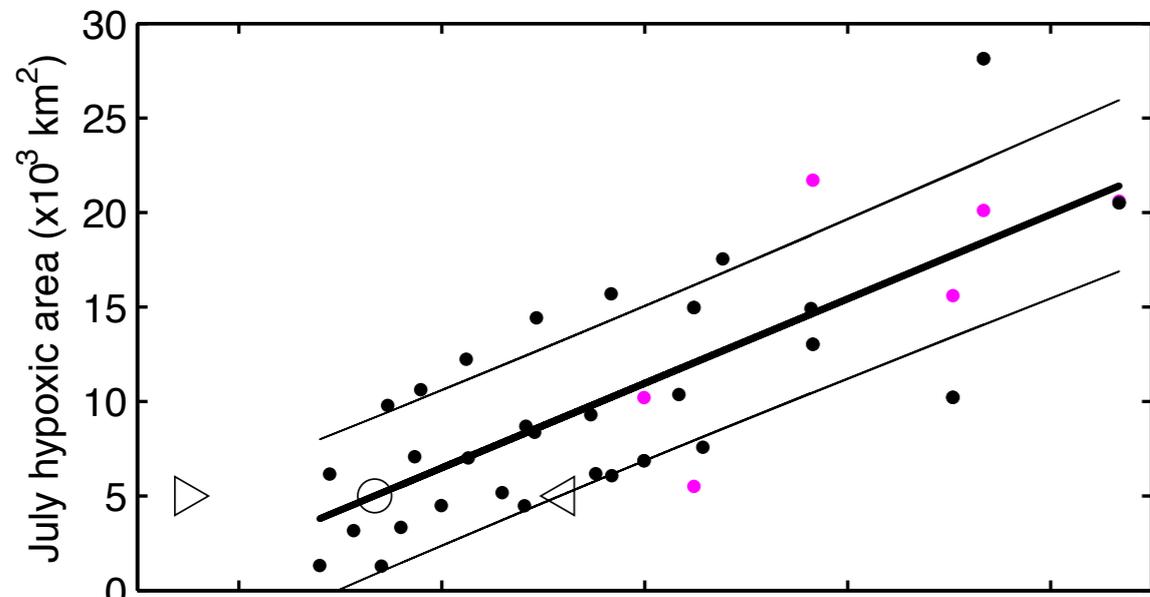
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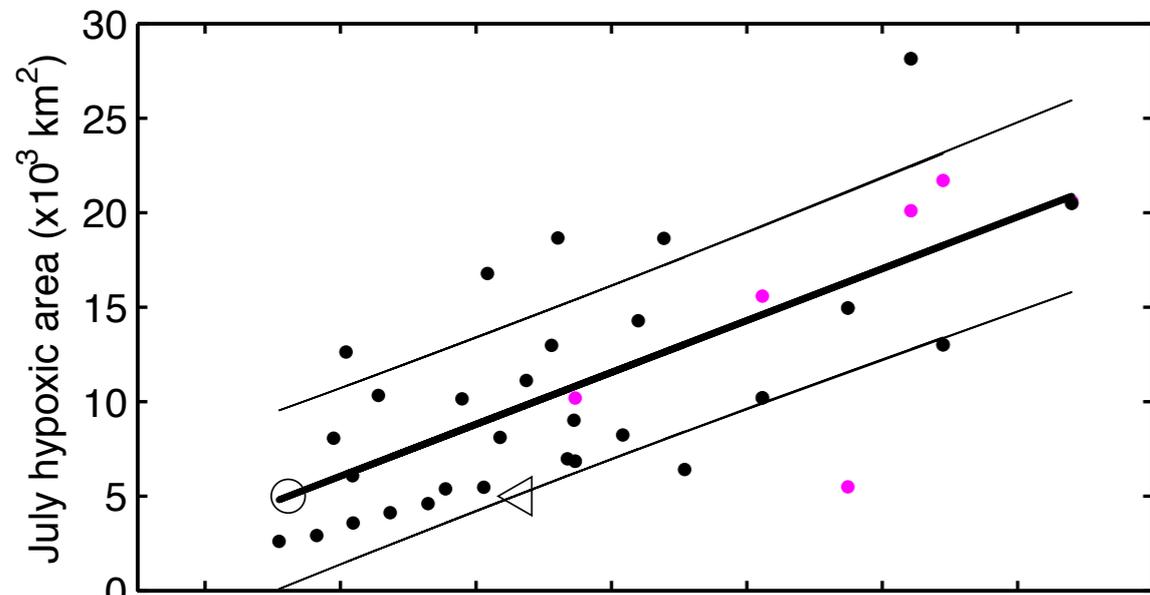
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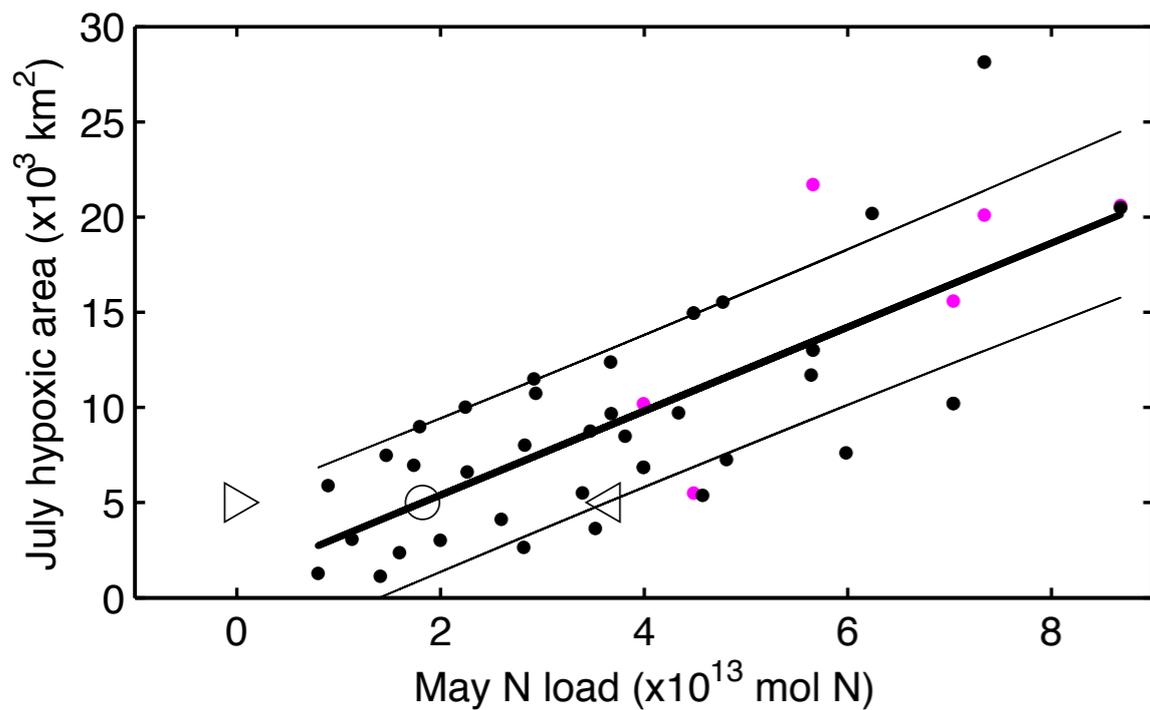
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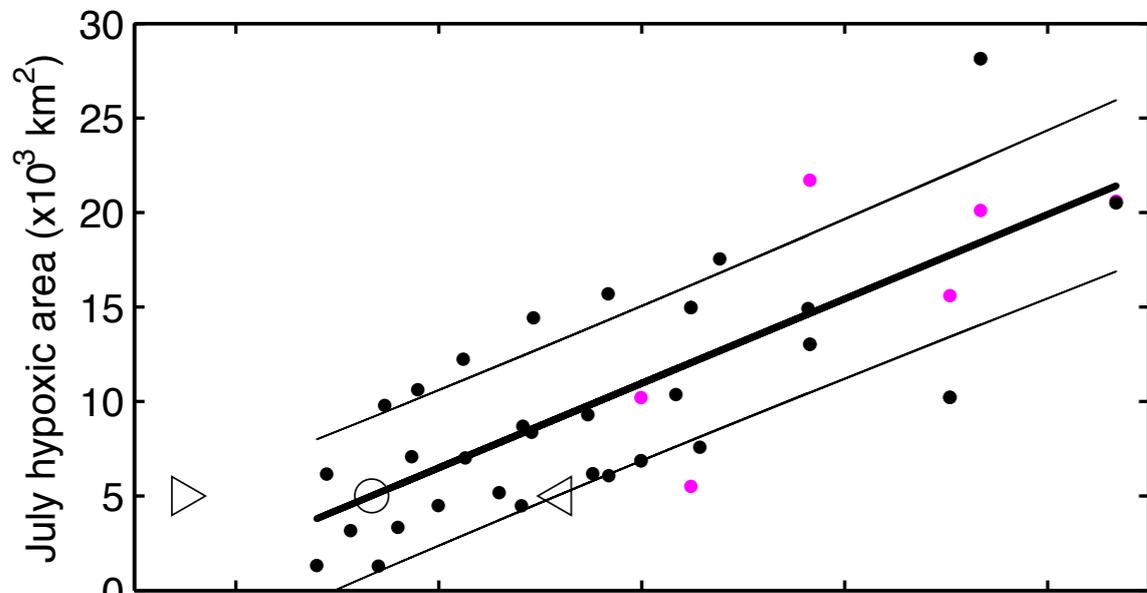


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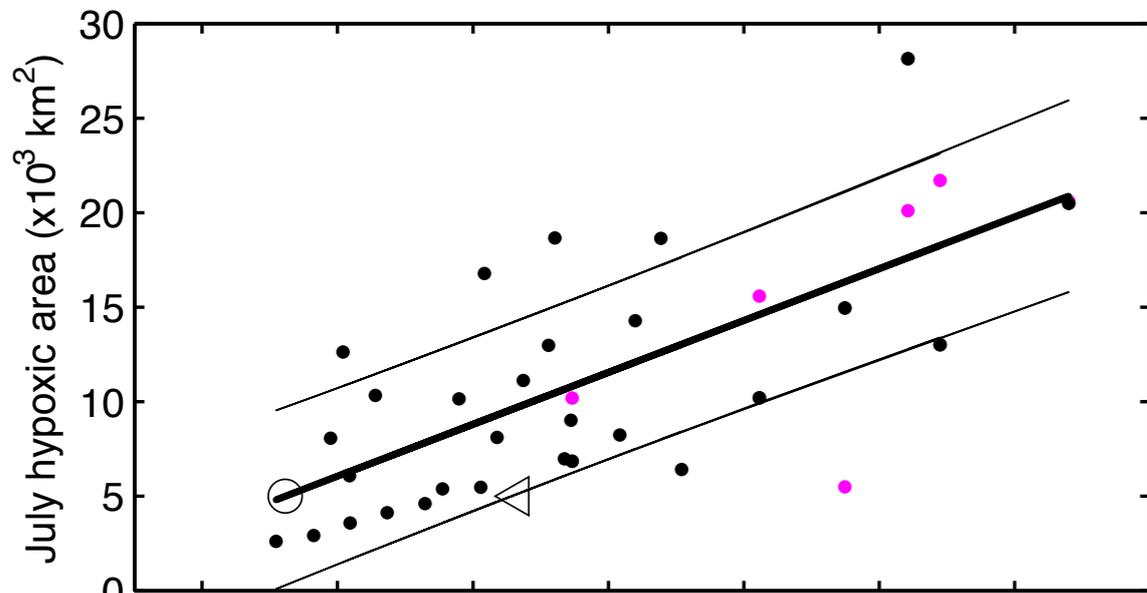
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best strategy

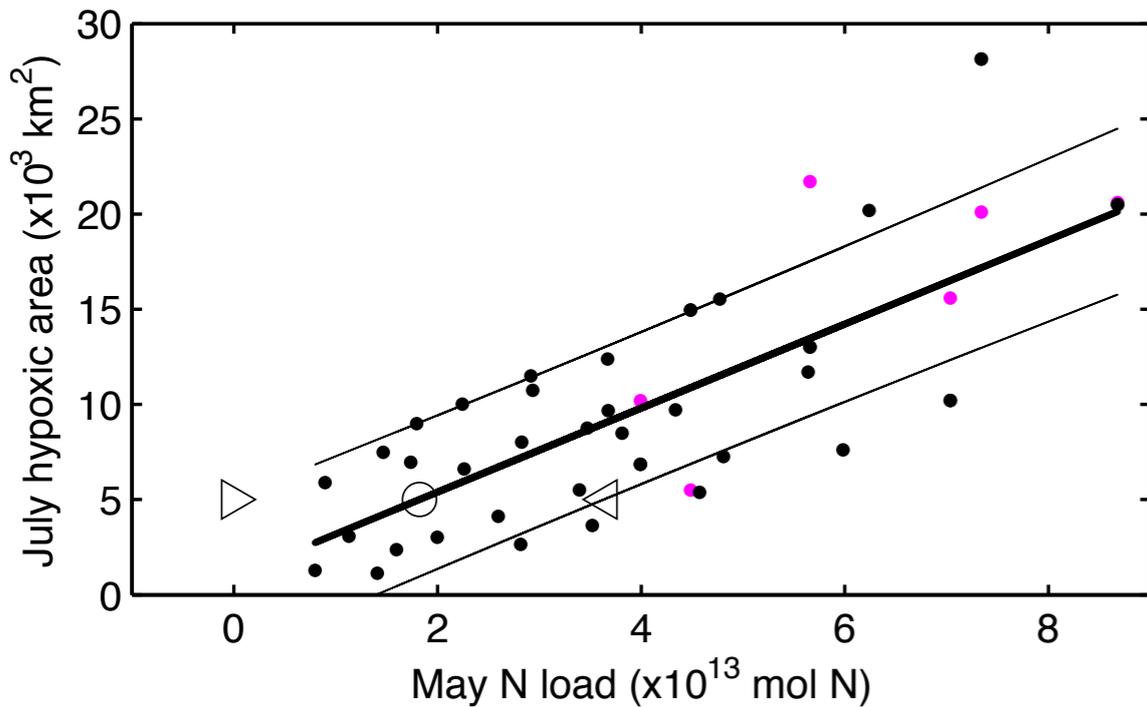


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next best



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best strategy

Previous estimates to reach 5,000 km² hypoxic area

Taskforce (2001)

30% N load reduction

Scavia et al. (2003)

40-45% N load reduction

Scavia & Donnelly (2007)

**37-45% N load reduction
or 40-50% P load reduction**

Previous estimates to reach 5,000 km² hypoxic area

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	model 12	>100% N load reduction or 42% N&P load reduction
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	UEN	>100% N load reduction
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Are these ROMS results robust? What do the other models predict?