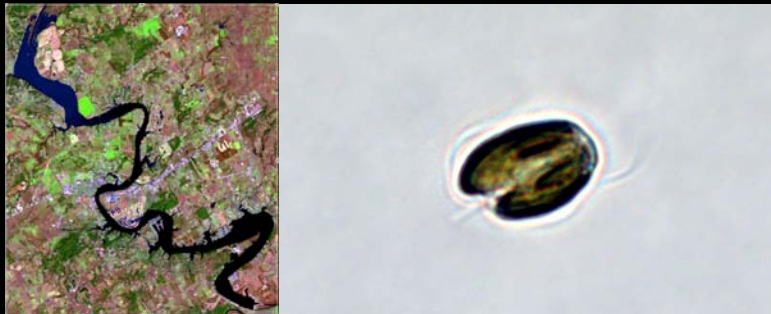


Water quality research for Lake Granbury, TX

Support from
2006/07 congressional earmark

Championed by
Rep. Chet Edwards



Leading Institutions

Texas A&M University
University of Texas at Arlington
Baylor University

Collaborators

Texas Water Research Institute
Brazos River Authority
Texas Parks and Wildlife

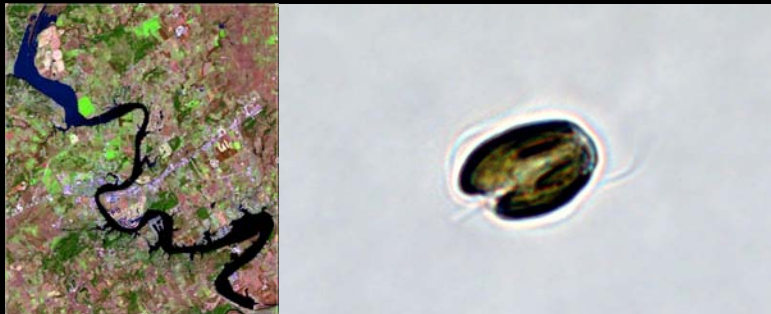
Water quality research for Lake Granbury, TX

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Rep. Chet Edwards

Researchers

Daniel Roelke
James Grover
Bryan Brooks
Steve Davis
George Gable
Anna-Marie Gable
Jason Baker
Jacob Stanley
Fabiola Ureña-Boeck
Mieke Lahousse



Prymnesium parvum blooms (golden algae)

TAMU, UTA, BU



Prymnesium parvum blooms (golden algae)

TAMU, UTA, BU



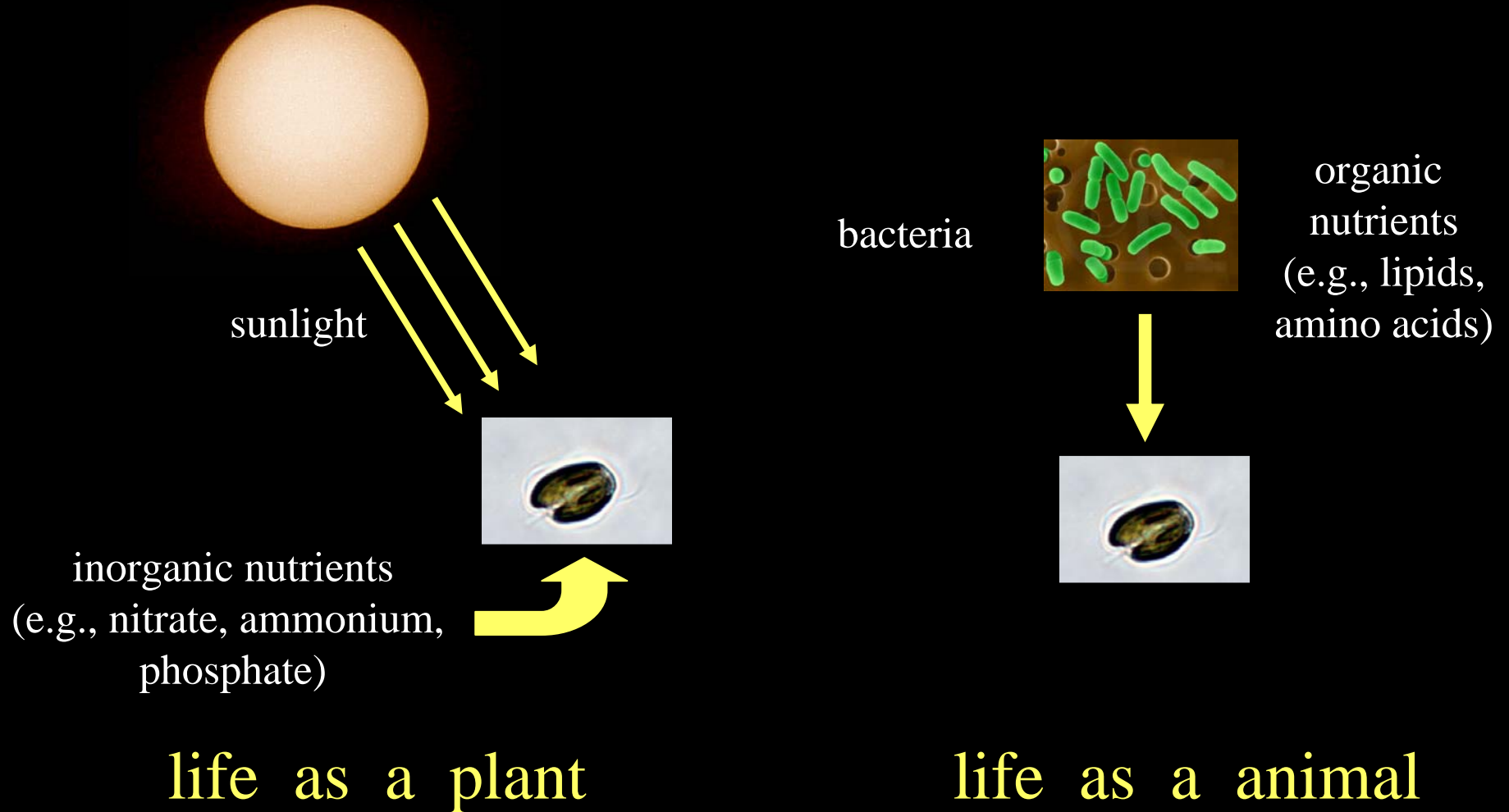
Fish are dying in
Lake Granbury - now

Pat and Dan Loomis



Prymnesium parvum: life as a mixotroph

TAMU, UTA, BU



Lake Granbury, TX: Water quality questions

TAMU, UTA, BU



- What causes golden algae blooms?
- Are golden algae and *E. coli* problems linked?
- Might “leaky” septic systems play a role?
- What can we do about it?

Research approach

TAMU, UTA, BU

1. In-lake monitoring

fixed-station sampling

high-resolution spatial mapping

2. Direct measurements of toxicity

bioassays using a fish

bioassays using a crustacean

3. Predictive modeling

laboratory studies

mathematical equations

validation and scenario testing

1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

Ten locations

#1 Head of reservoir

#2, 3

#4, 5

#6, 7

#8, 9

} Paired-stations
(shallow and deep)

#10 Dam



1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

Parameters sampled

P. parvum

E. coli

Fecal coliform

Dissolved org.-carbon

Toxicity

Chlorophyll *a*

Phytoplankton composition

Zooplankton composition

Total bacteria



1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

Parameters sampled

Nutrients

- Nitrate/nitrite
- Ammonium
- Phosphate
- Total nitrogen
- Total phosphorus

Light

- Transmission
- Secchi depth



1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

Parameters sampled

Temperature

Salinity

Dissolved oxygen

pH

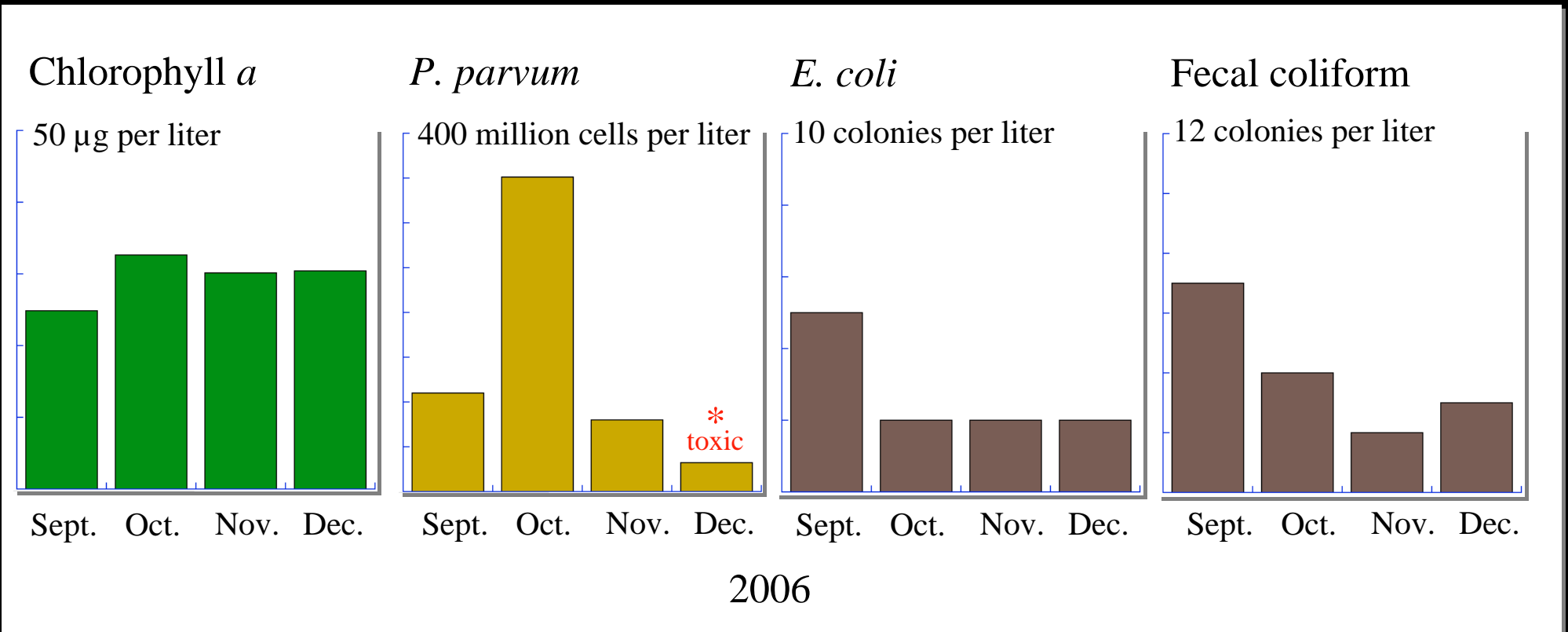
Total suspended solids

Oxidation-Reduction Potential



1. In-lake monitoring: Fixed-stations

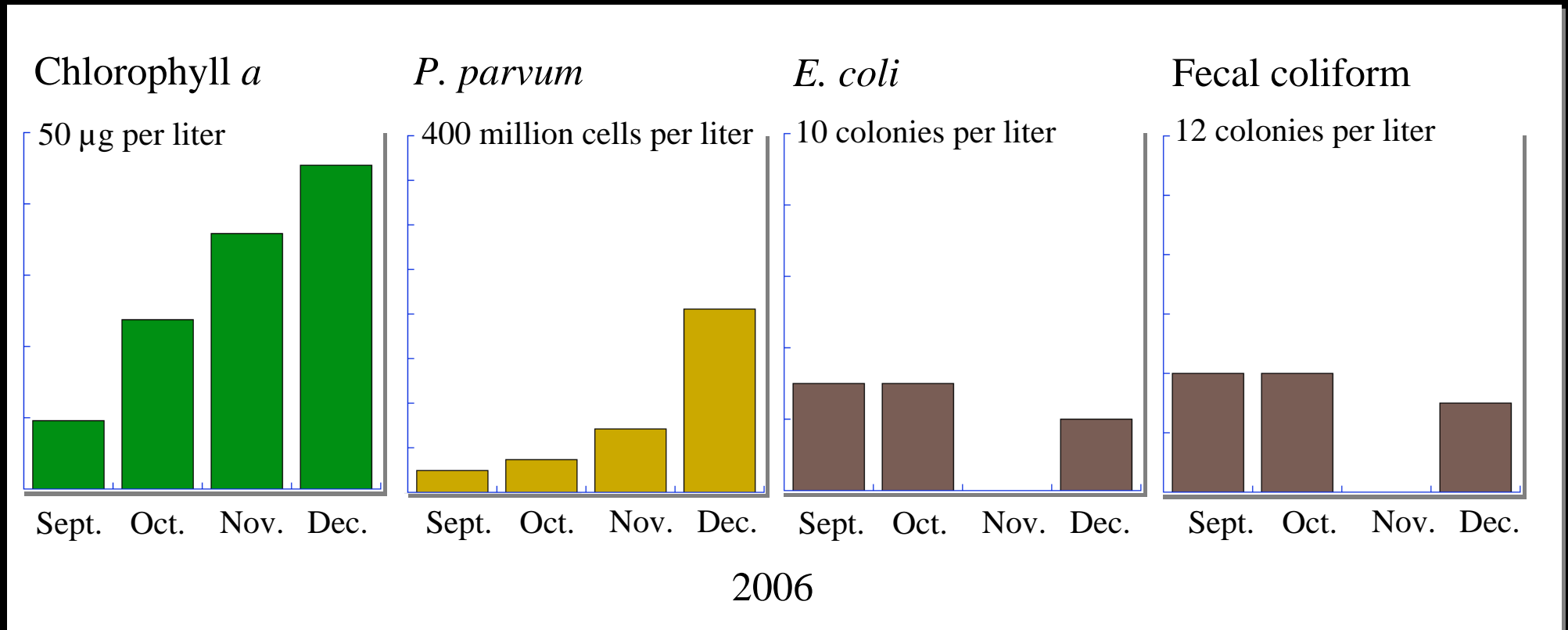
TAMU, UTA, BU



“Upper” Lake Granbury - Representative trends

1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

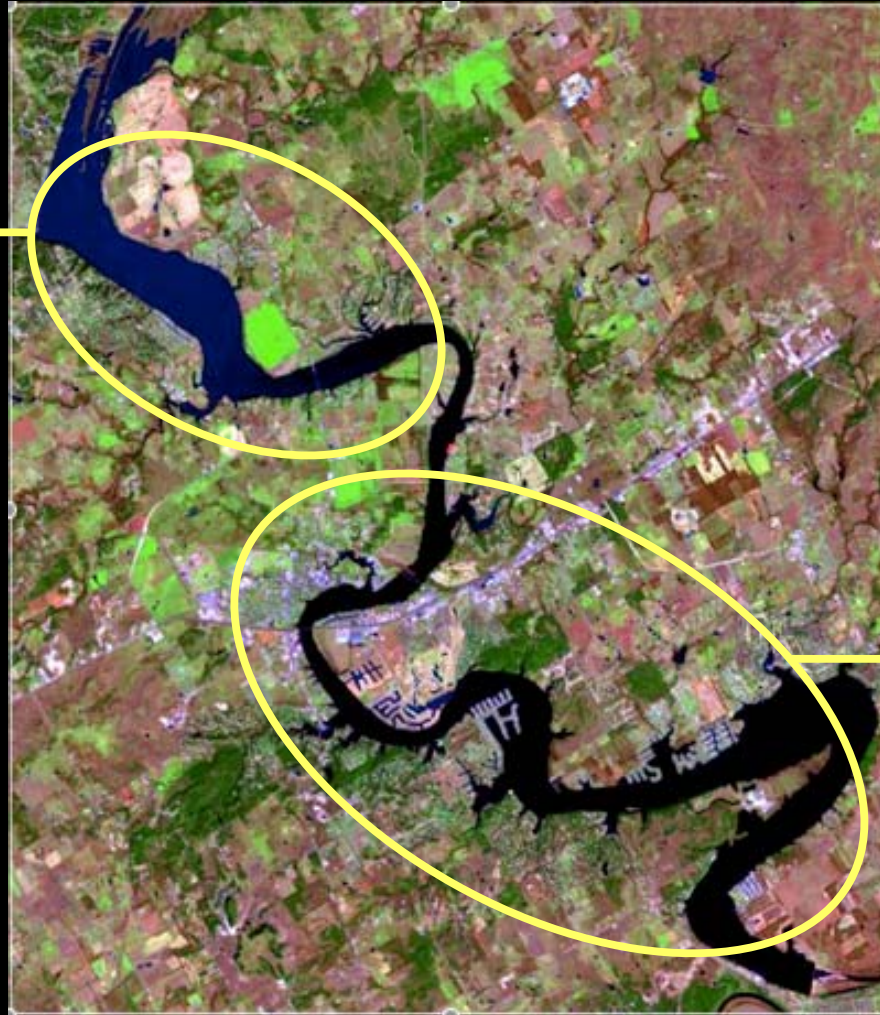


“Lower” Lake Granbury - Representative trends

1. In-lake monitoring: Fixed-stations

TAMU, UTA, BU

Persistent
phytoplankton
biomass, and
variable
golden algae



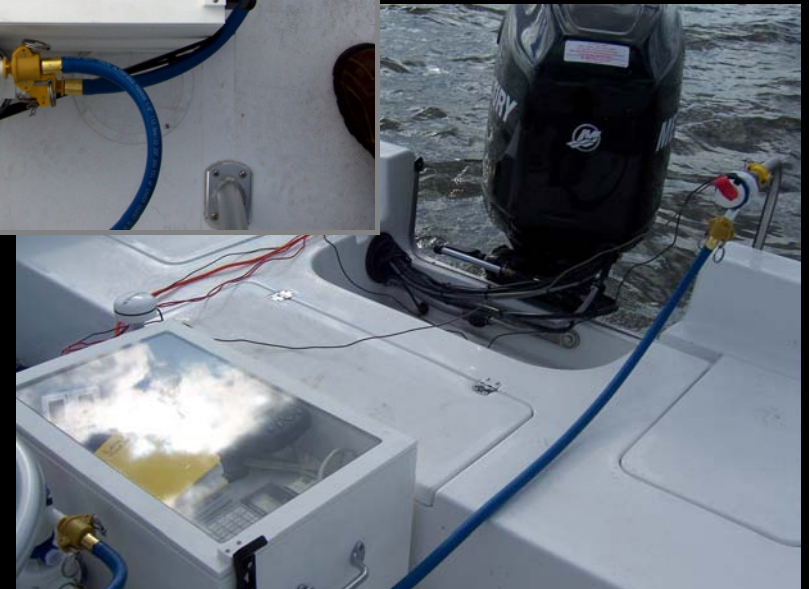
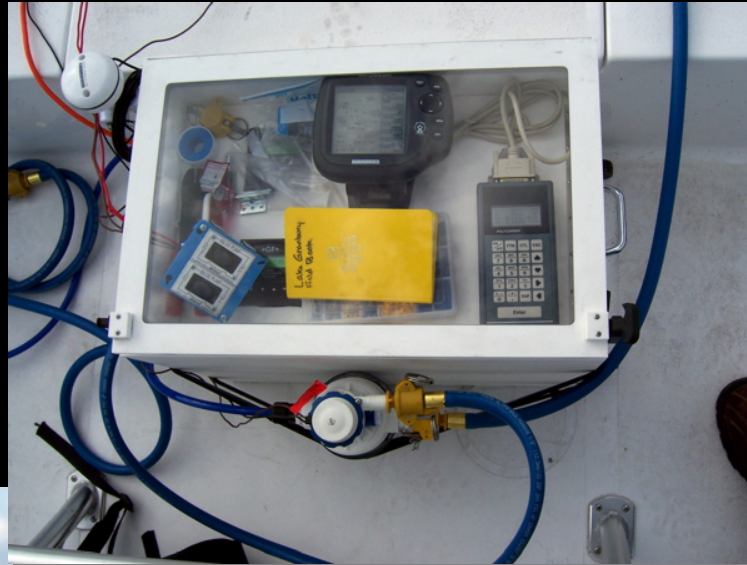
Growing
phytoplankton
biomass and
golden algae

1. In-lake monitoring: Mapping

TAMU, UTA, BU

Dataflow

On-board, flow through system with geo-referenced data collection



1. In-lake monitoring: Mapping

TAMU, UTA, BU

Parameters sampled

Chlorophyll *a*

Dissolved org. carbon

Salinity

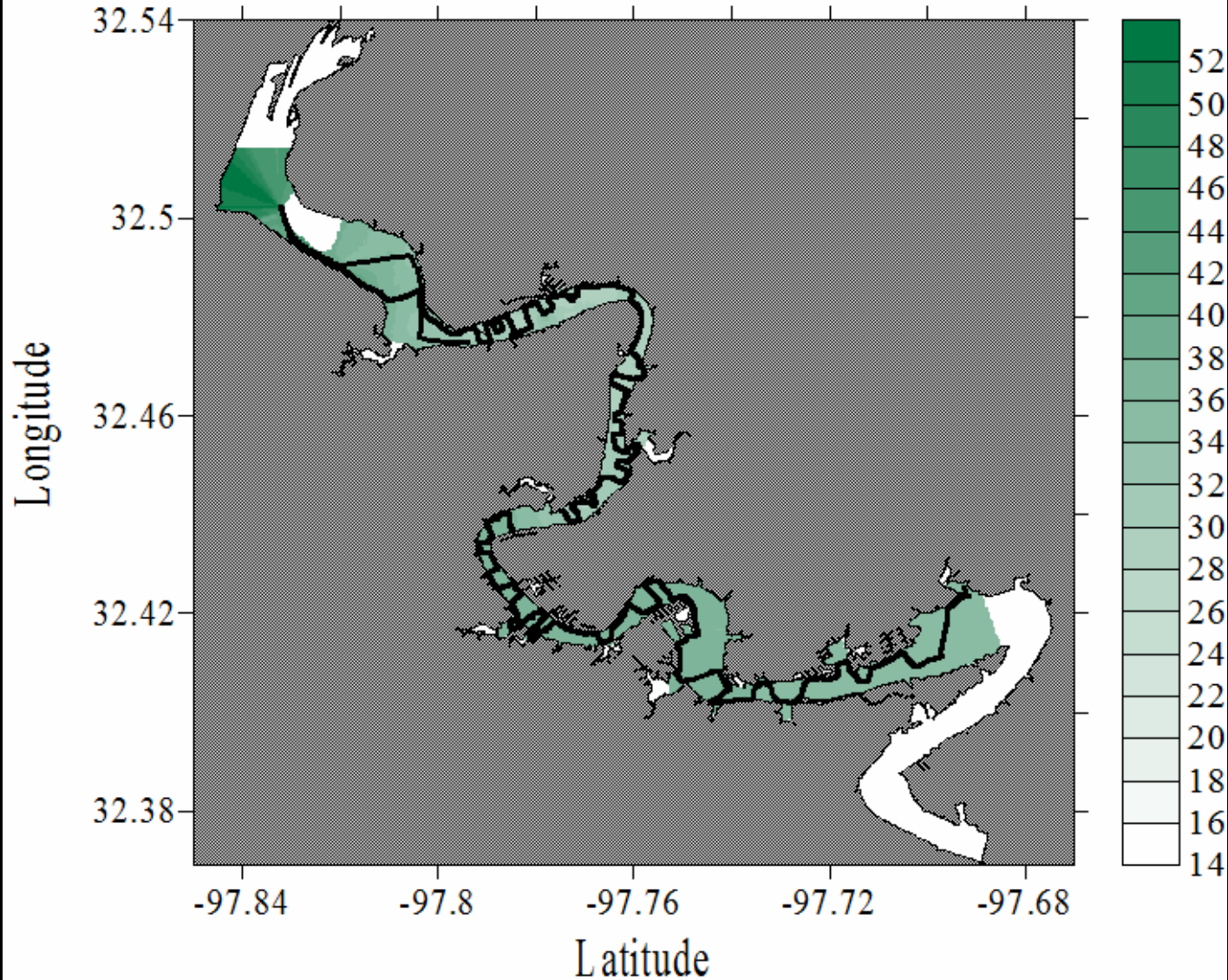
Temperature

Transparency



Lake Granbury, Texas
November 11, 2006

Chlorophyll - a ($\mu\text{g L}^{-1}$)



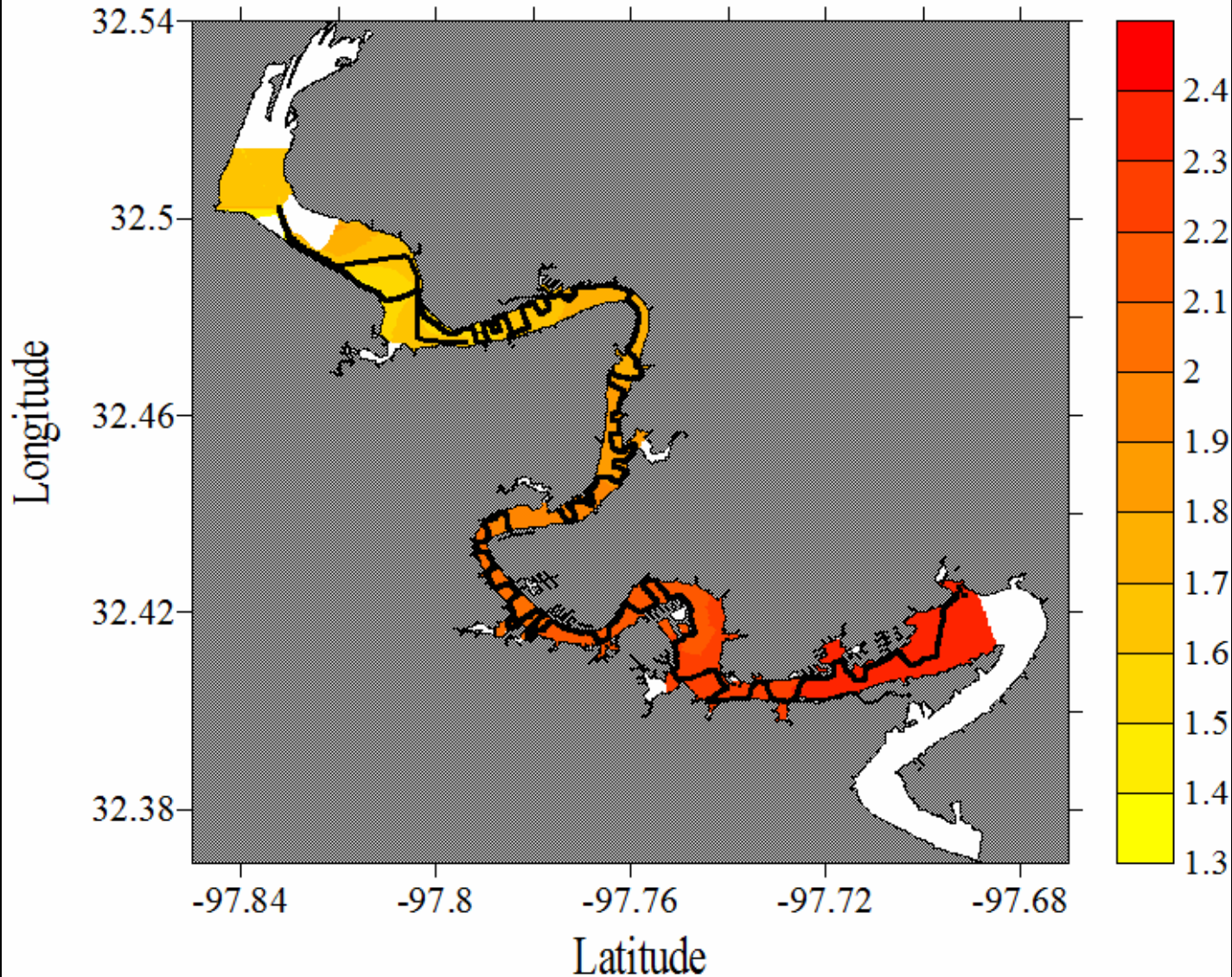
1. Mapping

TAMU, UTA, BU

Well-mixed
conditions

Lake Granbury, Texas
November 11, 2006

Salinity



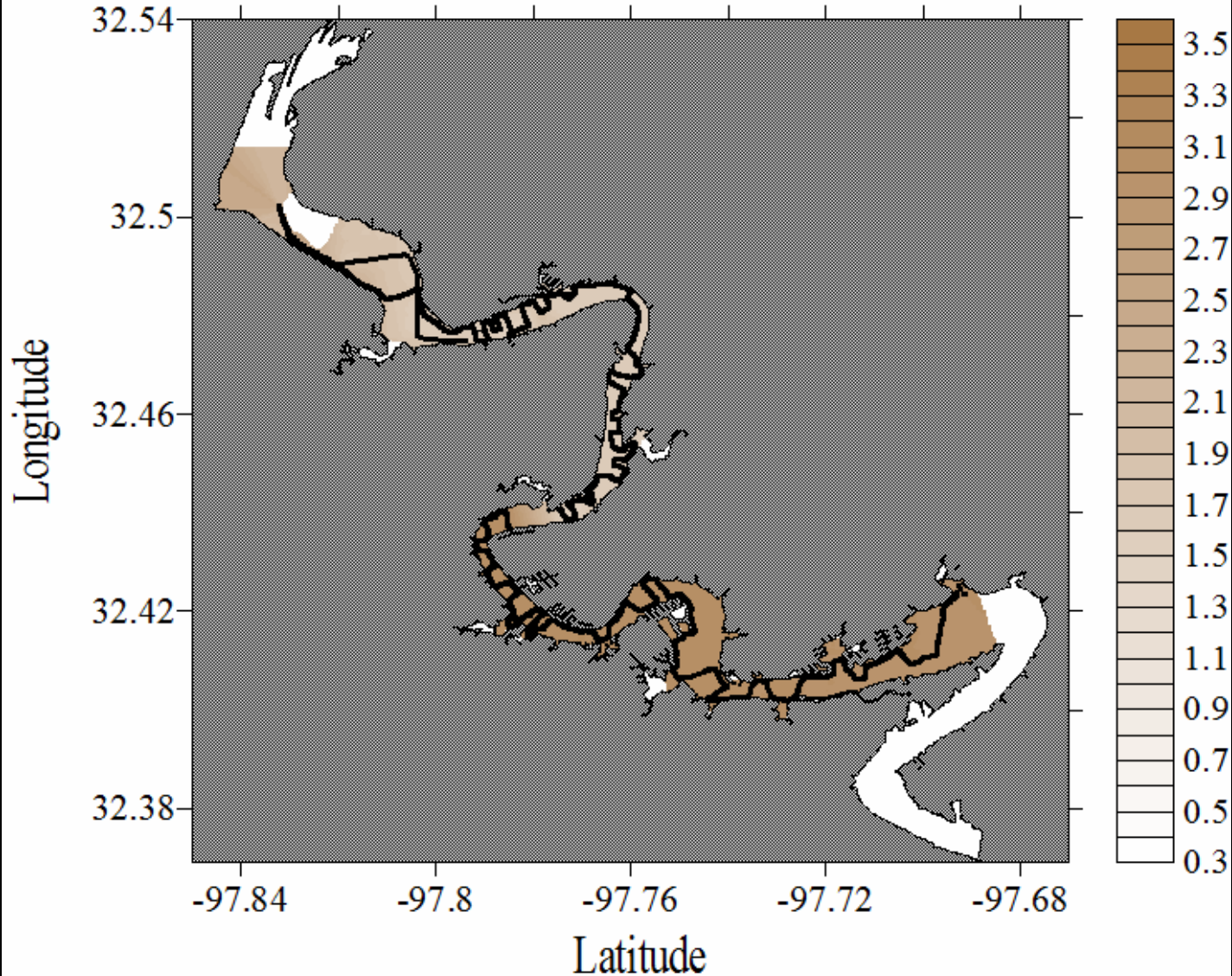
1. Mapping

TAMU, UTA, BU

Well-mixed
conditions

Lake Granbury, Texas
November 11, 2006

FDOM (voltage)



1. Mapping

TAMU, UTA, BU

Well-mixed
conditions

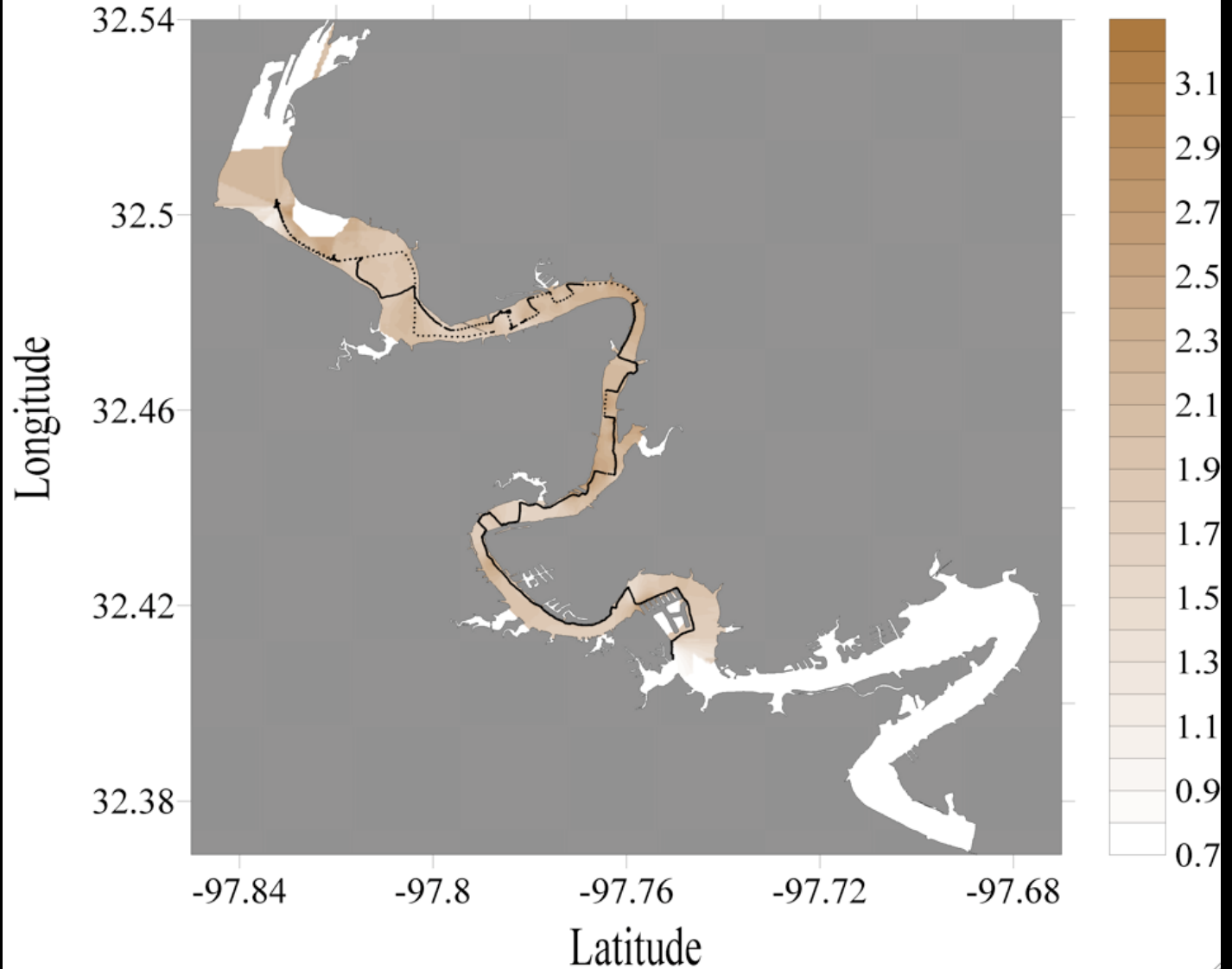
1. Mapping

TAMU, UTA, BU

Spatial
heterogeneity

Lake Granbury, Texas
September 13, 2006

FDOM (voltage)



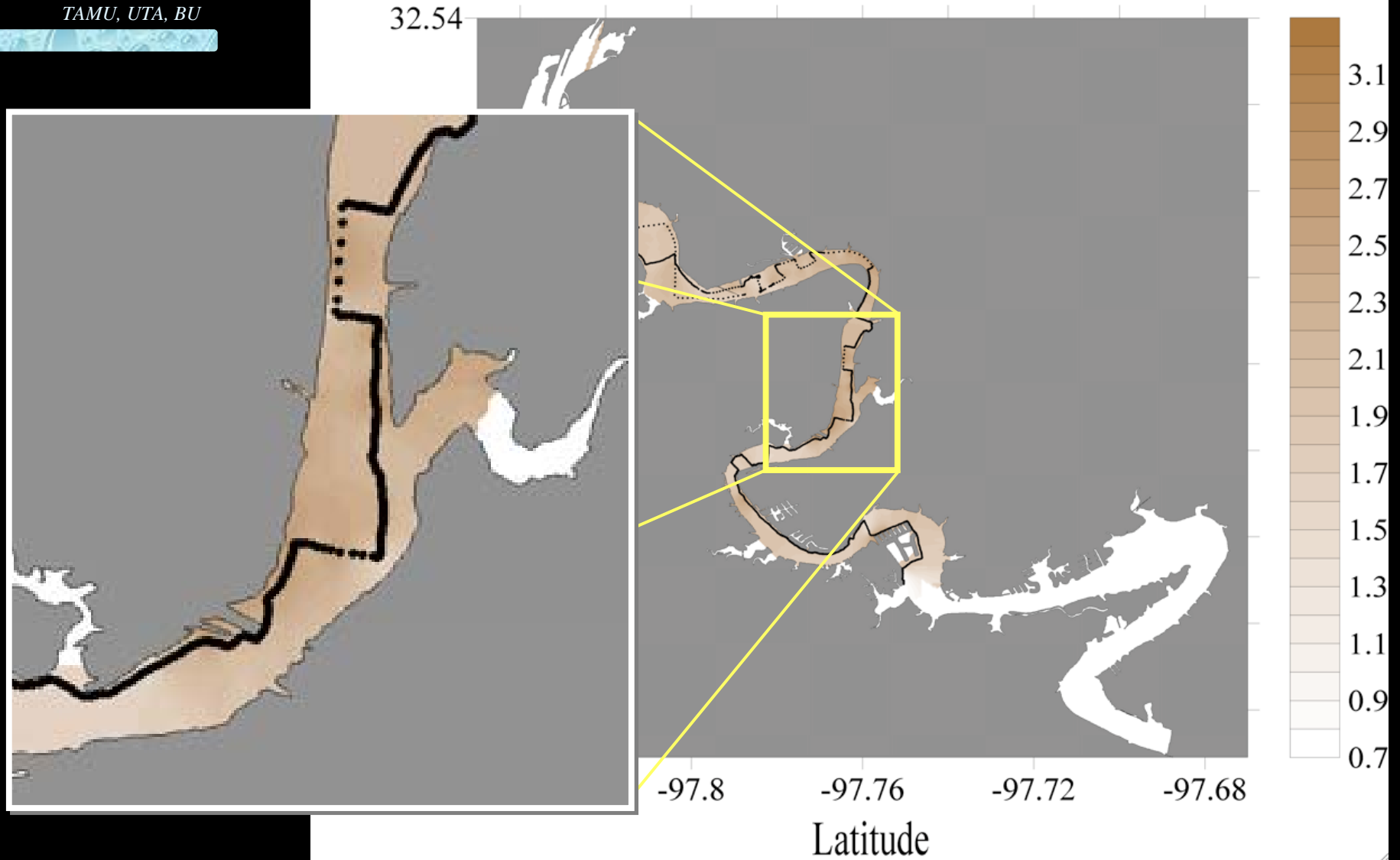
1. Mapping

TAMU, UTA, BU



Lake Granbury, Texas
September 13, 2006

FDOM (voltage)

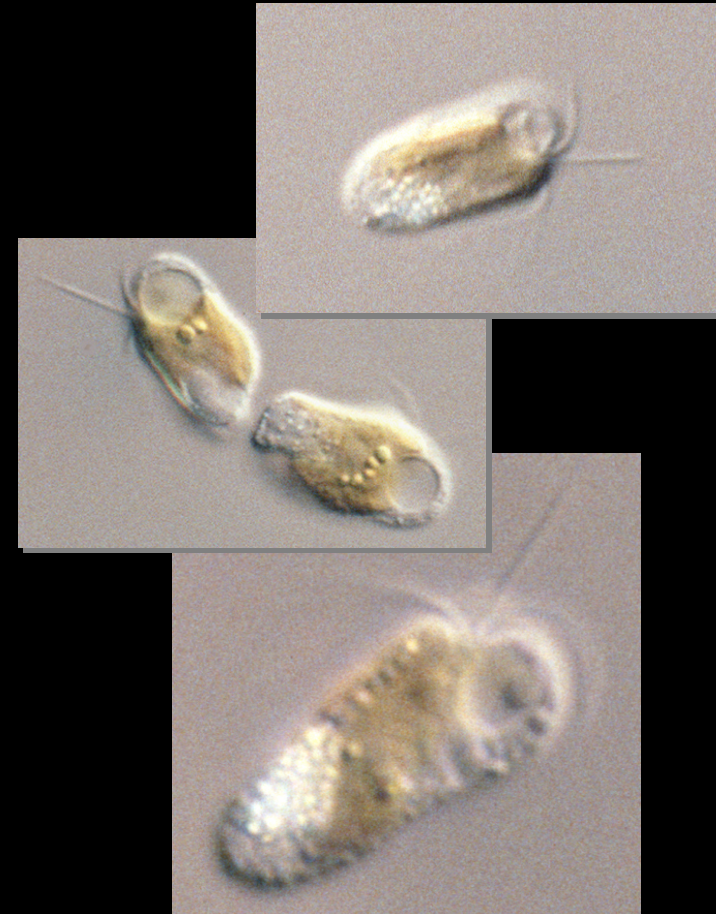
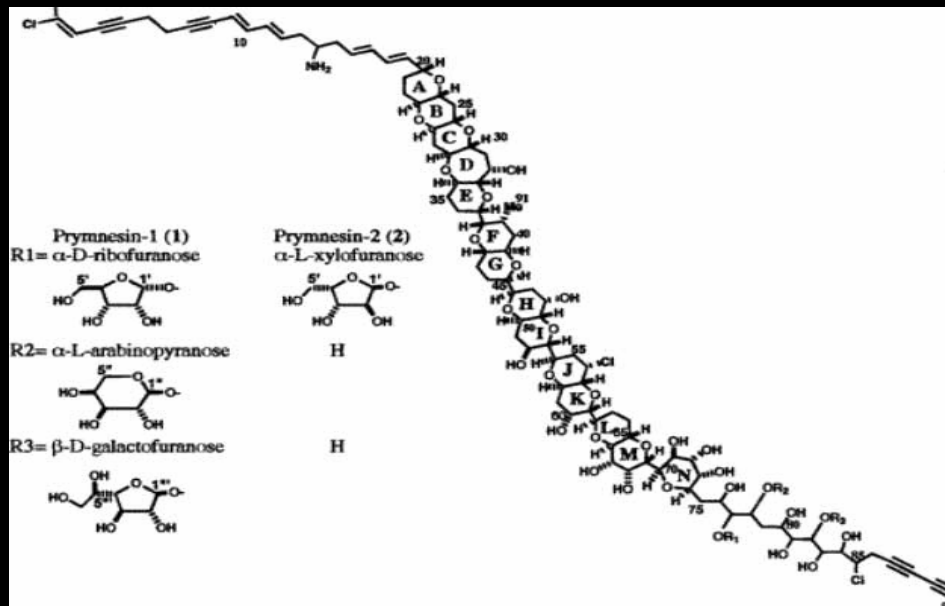


2. Direct measurements of toxicity

TAMU, UTA, BU

Prymnesium parvum

Fish kills believed to result from exposure to prymnesium-1 ($C_{107}H_{154}Cl_3NO_{44}$) and/or prymnesium-2 ($C_{96}H_{136}Cl_3NO_{35}$)



J La Claire, UT Austin

2. Direct measurements of toxicity

TAMU, UTA, BU

Coal miners used canaries to signal if there was a problem in the mine shaft



Similarly, we use sensitive organisms to signal if toxic *P. parvum* blooms occur



Daphnia magna
- A "water flea"



Pimephales promelas
- A common minnow

2. Direct measurements of toxicity

TAMU, UTA, BU

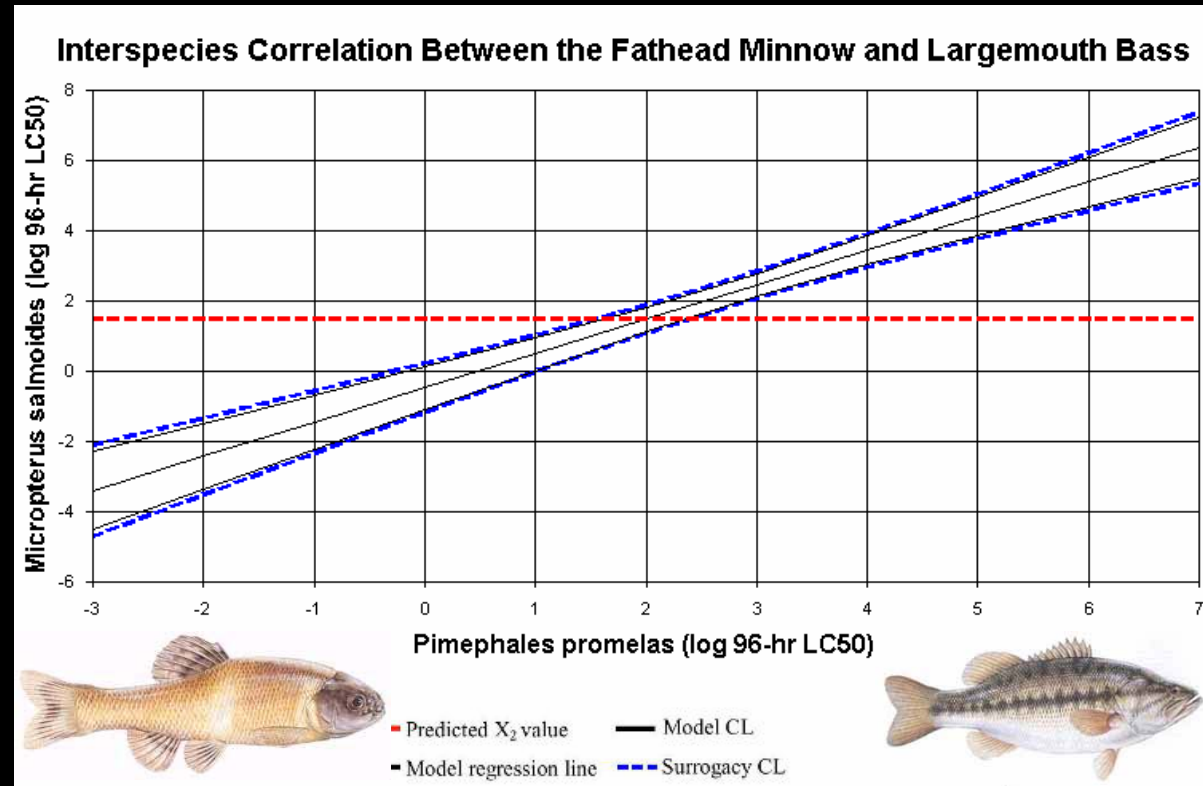
Even though a minnow...



is not a largemouth...



the minnow can be protective of other fish.



2. Direct measurements of toxicity

TAMU, UTA, BU

Prymnesium parvum Toxicity

Why Use Aquatic Biosensors?

1. Sensitivity - excellent “sentinels”
2. Ecological Relevance - representative of other cladocerans and fish
3. Availability - species widely used for monitoring water quality
4. Precision - reproducibility of responses within and between labs



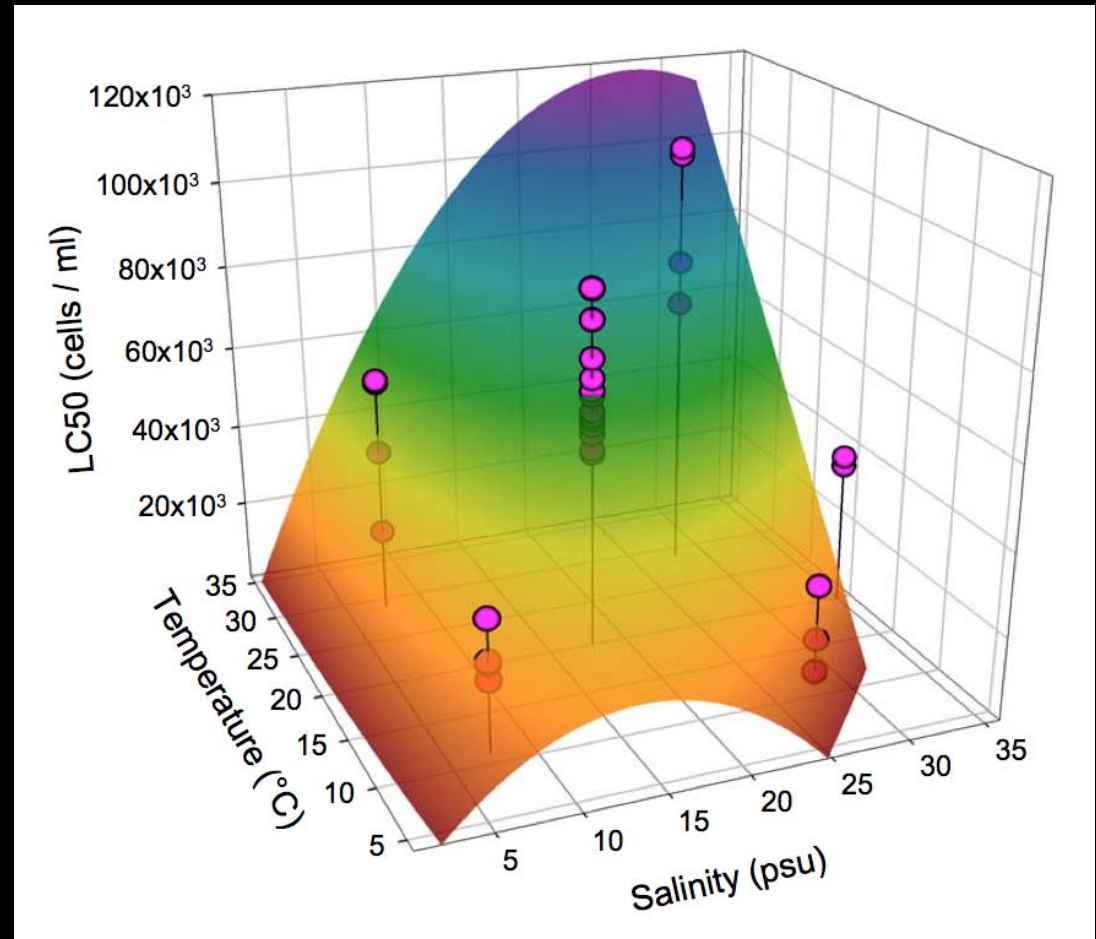
2. Direct measurements of toxicity

TAMU, UTA, BU

P. parvum toxicity to fish affected by temperature and salinity

Most pronounced at lower salinity and temperatures similar to those experienced during winter blooms in Texas reservoirs (e.g., Granbury, Possum Kingdom, Whitney)

Perhaps a recipe for fish kills?



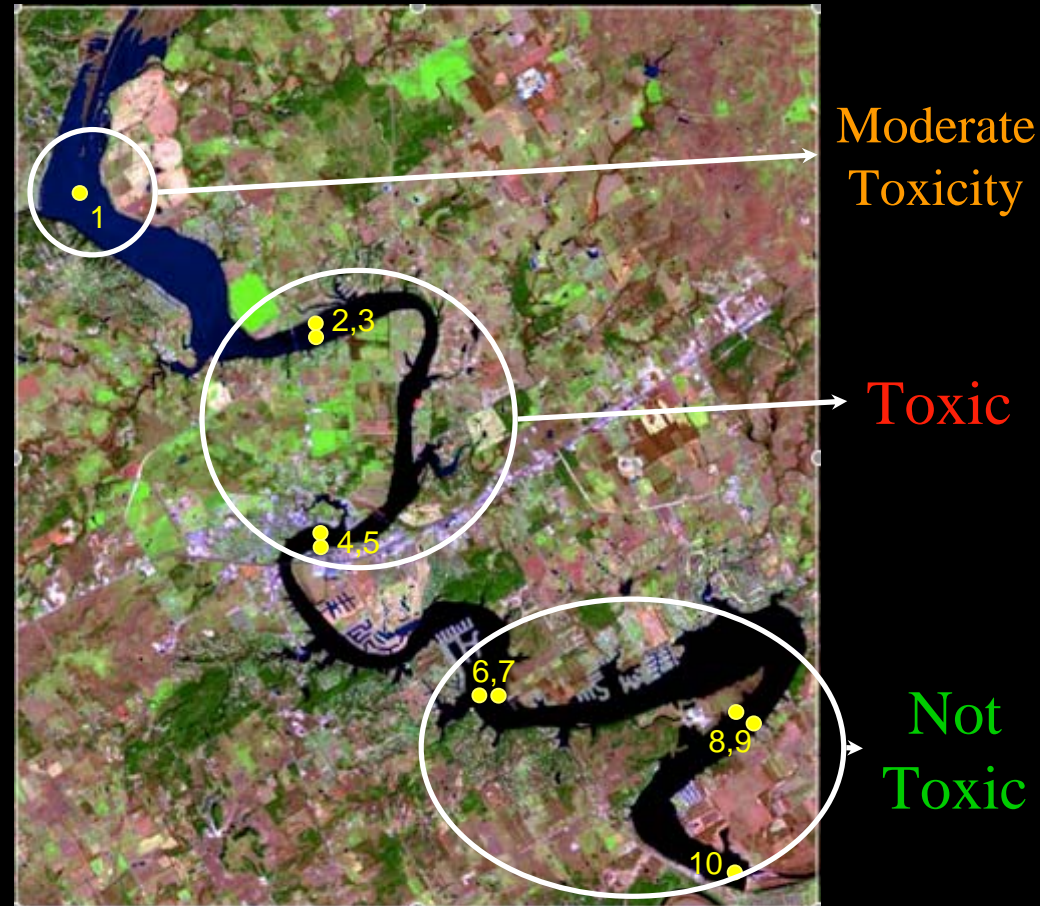
2. Direct measurements of toxicity

TAMU, UTA, BU

Fish Biosensor Responses to the January 2007 Toxic Bloom

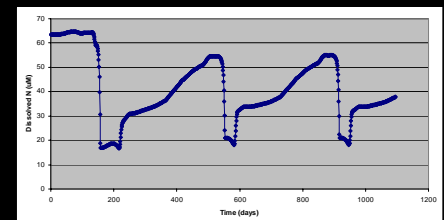
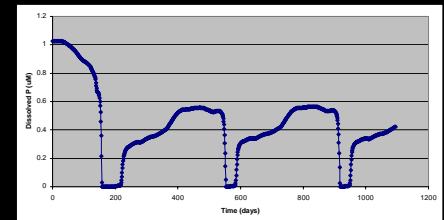
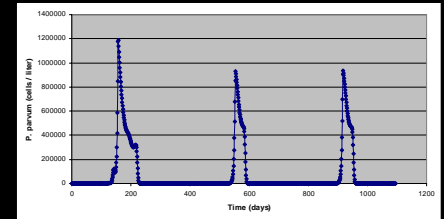
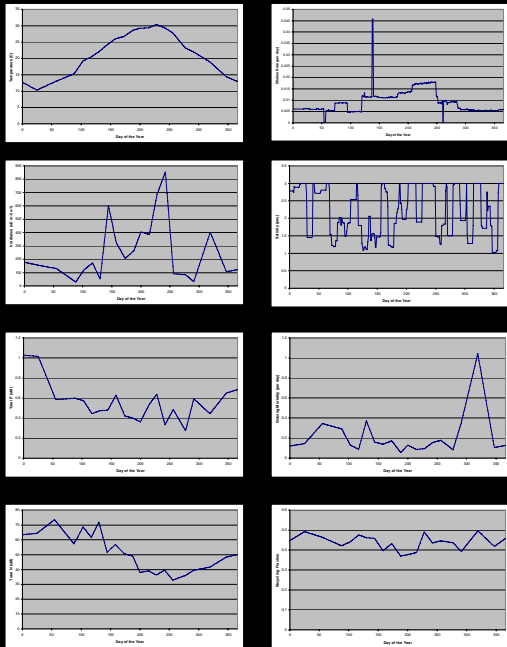
Site Number Percent Mortality

| | |
|----|-----|
| 1 | 27% |
| 2 | 93% |
| 3 | 87% |
| 4 | 93% |
| 5 | 73% |
| 6 | 0 |
| 7 | 0 |
| 8 | 0 |
| 9 | 0 |
| 10 | 0 |



3. Predictive modeling - overview

TAMU, UTA, BU



Input Data

Equations & Calculations

Output Data

Based on lake characteristics

Based on knowledge and guesswork

Predicted *P. parvum* density, dissolved nitrate and phosphate

3. Predictive modeling – Knowledge & guesswork

TAMU, UTA, BU

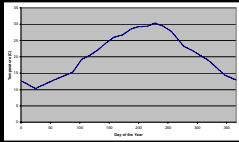


Equations &
Calculations

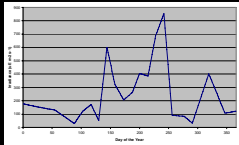
Population Change = Growth (Reproduction) - Mortality

3. Predictive modeling – Growth

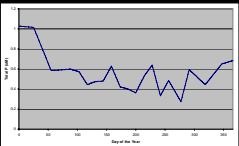
TAMU, UTA, BU



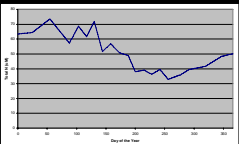
Water Temperature



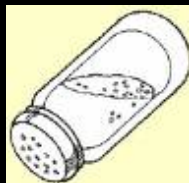
Underwater Light



Nutrient Supply
(Nitrate & Phosphate)



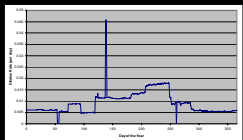
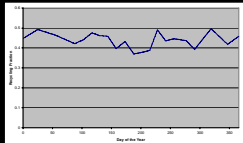
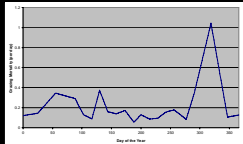
Salinity



Growth of *P. parvum*

3. Predictive modeling – Mortality

TAMU, UTA, BU



Zooplankton
“Grazing”



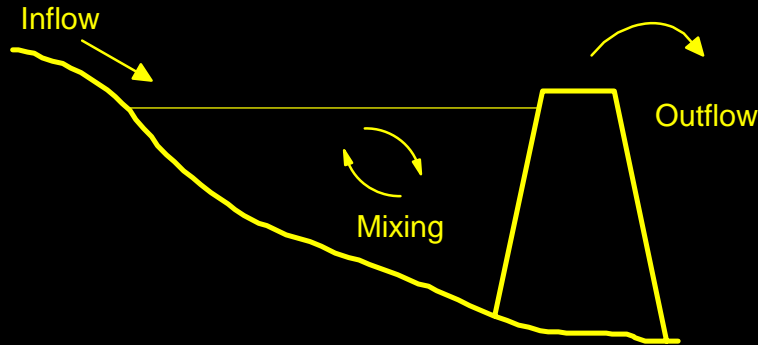
Hydraulic
Flushing



Mortality of *P. parvum*

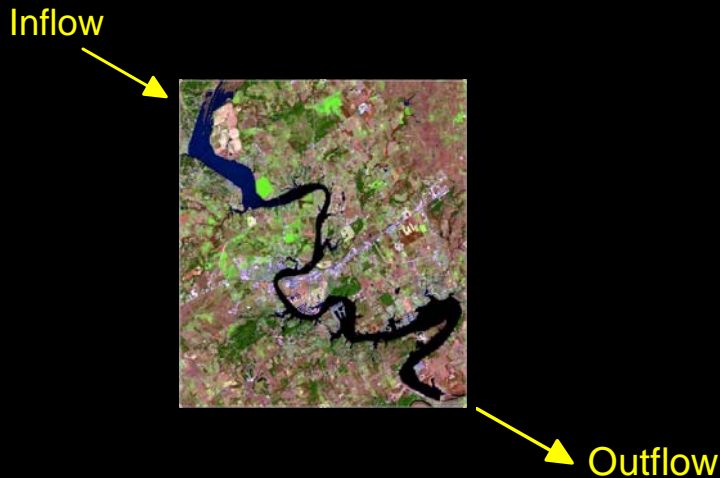
3. Predictive modeling – Physical settings

TAMU, UTA, BU



Well – mixed conditions

Unrealistic, but easy



Long, narrow reservoir

Realistic, but challenging

3. Predictive modeling – More complications

TAMU, UTA, BU

Other algae live in the lake and compete with *P. parvum*, what is their effect?

Toxicity of *P. parvum* is not in the current version.

“Life as an animal” is not in the current version.

All processes in the model are highly simplified and could be more realistic.

3. Predictive modeling – Uses

TAMU, UTA, BU

Summarizes knowledge systematically, identifies gaps.

“What if” questions and management scenarios can be explored.

Forecasting and prediction...

Timeline and What's next?

TAMU, UTA, BU

Monthly sampling, and increased scope

- on-going lake sampling (as described)
- correlations between golden algae, bacteria and DOM
- add anthropogenic tracers (e.g., nicotine, caffeine, etc.)
- expand to regional studies (multiple lakes, historical analysis)

Predictive modeling

- develop model of golden algae with competitors
- extend model to long, narrow reservoir setting
- compare model to field data
- extend realism of model (toxicity, life as an animal)