

Computational Revision of Ecological Process Models

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Data Mining vs. Scientific Discovery

There exist two computational paradigms for discovering explicit knowledge from data.

The *data mining* movement develops computational methods that:

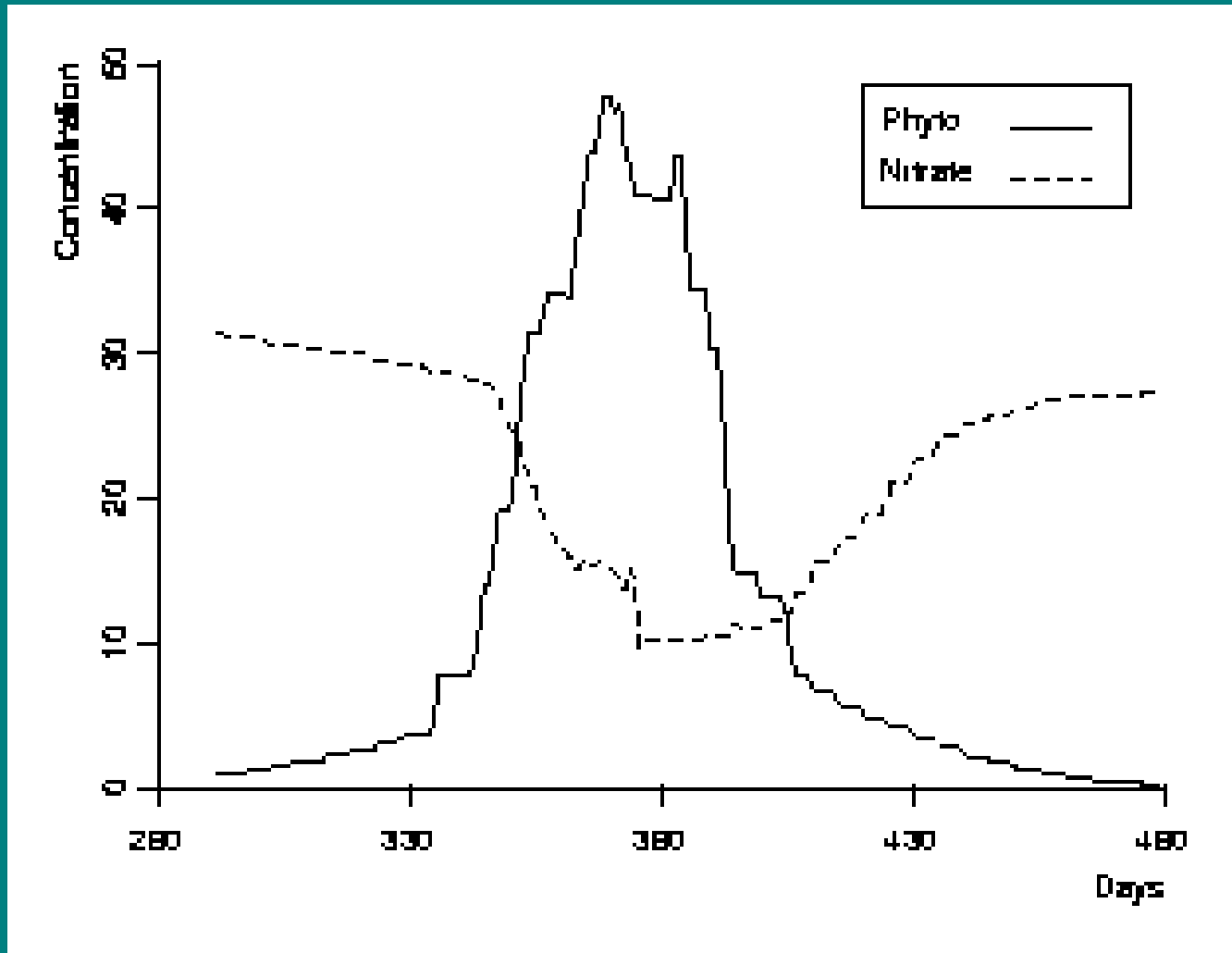
- induce predictive models from large (often business) data sets;
- represent models in notations invented by AI researchers.

In contrast, *computational scientific discovery* focuses on:

- constructing models from (often small) scientific data sets;
- stated in formalisms invented by scientists themselves.

This talk focuses on applications of the second framework to environmental and ecosystem modeling.

Observations from the Ross Sea



A Model of the Ross Sea Ecosystem

model RossSeaEcosystem

variables: phyto, zoo, nitro, residue

observables: phyto, nitro

$$d[\text{phyto}, t, 1] = -0.307 \times \text{phyto} - 0.495 \times \text{zoo} + 0.411 \times \text{phyto}$$

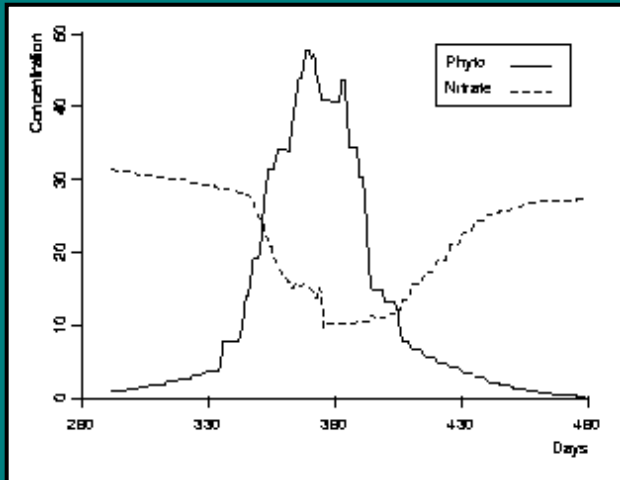
$$d[\text{zoo}, t, 1] = -0.251 \times \text{zoo} + 0.615 \times 0.495 \times \text{zoo}$$

$$d[\text{residue}, t, 1] = 0.307 \times \text{phyto} + 0.251 \times \text{zoo} + 0.385 \times 0.495 \times \text{zoo} - 0.005 \times \text{residue}$$

$$d[\text{nitro}, t, 1] = -0.098 \times 0.411 \times \text{phyto} + 0.005 \times \text{residue}$$

Inductive Revision of Ecosystem Models

observations



Revision

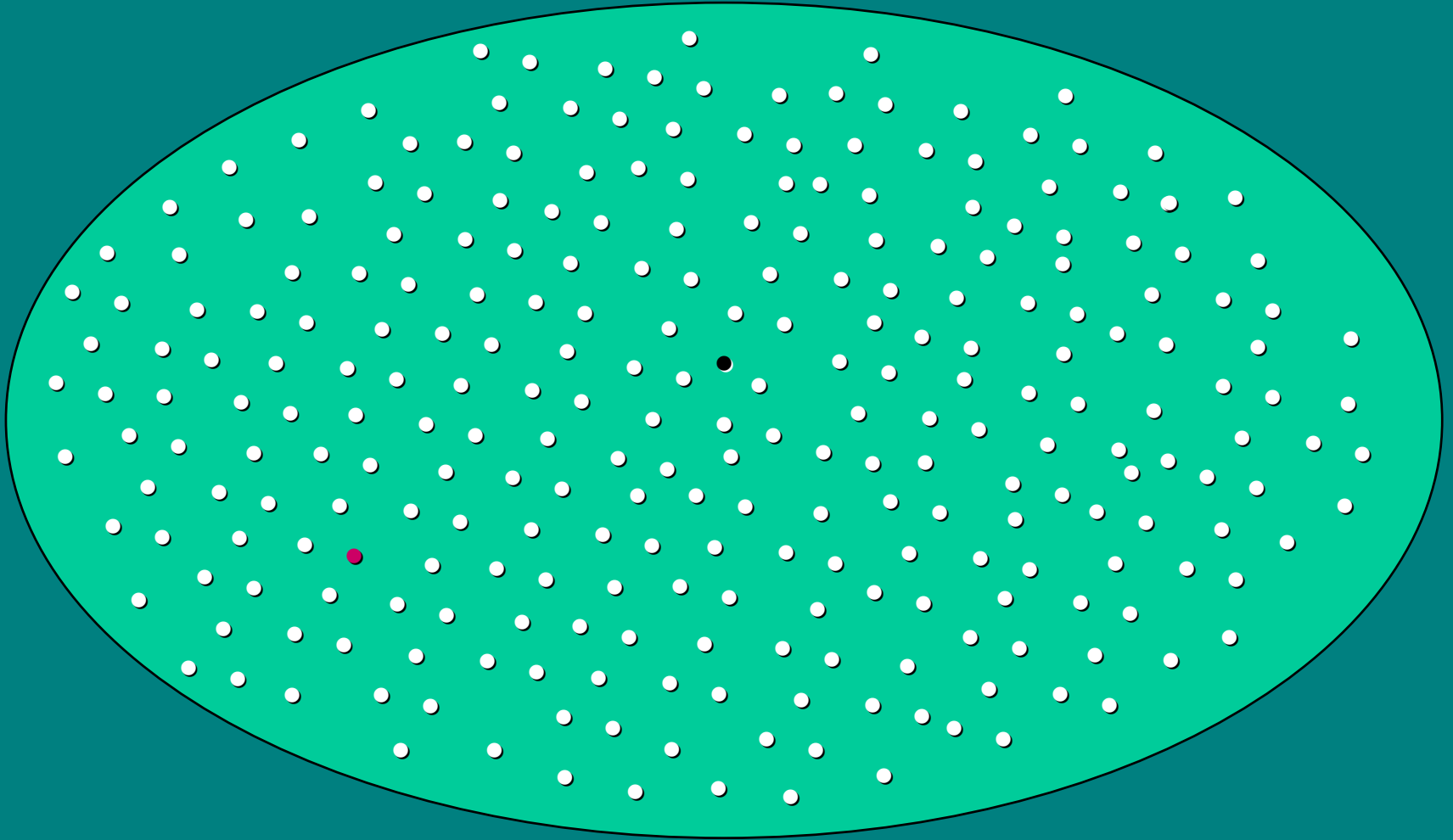
revised model

```
model RossSeaEcosystem
variables: phyto, zoo, nitro, residue
observables: phyto, nitro
d[phyto,t,1] = - 0.307 × phyto - 0.495 × zoo
               + 0.411 × phyto
d[zoo,t,1] = - 0.251 × zoo + 0.615 × 0.495 × zoo
d[residue,t,1] = 0.307 × phyto + 0.251 × zoo
                + 0.385 × 0.495 × zoo - 0.005 × residue
d[nitro,t,1] = - 0.098 × 0.411 × phyto + 0.005 × residue
```

```
model RossSeaEcosystem
variables: phyto, zoo, nitro, residue
observables: phyto, nitro
d[phyto,t,1] = - 0.307 × phyto - 0.495 × zoo
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d[residue,t,1] = 0.307 × phyto + 0.251 × zoo
                + 0.385 × 0.495 × zoo - 0.005 × residue
d[nitro,t,1] = - 0.098 × 0.411 × phyto + 0.005 × residue
```

initial model

A Space of Ecosystem Models



Model revision requires ways to constrain search through this space.

Phytoplankton Loss in the Ross Sea

model RossSeaEcosystem

variables: phyto, zoo, nitro, residue

observables: phyto, nitro

$$d[\text{phyto}, t, 1] = -0.307 \times \text{phyto} - 0.495 \times \text{zoo} + 0.411 \times \text{phyto}$$

$$d[\text{zoo}, t, 1] = -0.251 \times \text{zoo} + 0.615 \times 0.495 \times \text{zoo}$$

$$d[\text{residue}, t, 1] = 0.307 \times \text{phyto} + 0.251 \times \text{zoo} + 0.385 \times 0.495 \times \text{zoo} - 0.005 \times \text{residue}$$

$$d[\text{nitro}, t, 1] = -0.098 \times 0.411 \times \text{phyto} + 0.005 \times \text{residue}$$

Phytoplankton loss is a *process* that affects two variables; no model should include one influence without the other.

Grazing in the Ross Sea Ecosystem

model RossSeaEcosystem

variables: phyto, zoo, nitro, residue

observables: phyto, nitro

$$d[\text{phyto}, t, 1] = -0.307 \times \text{phyto} - 0.495 \times \text{zoo} + 0.411 \times \text{phyto}$$

$$d[\text{zoo}, t, 1] = -0.251 \times \text{zoo} + 0.615 \times 0.495 \times \text{zoo}$$

$$d[\text{residue}, t, 1] = 0.307 \times \text{phyto} + 0.251 \times \text{zoo} + 0.385 \times 0.495 \times \text{zoo} - 0.005 \times \text{residue}$$

$$d[\text{nitro}, t, 1] = -0.098 \times 0.411 \times \text{phyto} + 0.005 \times \text{residue}$$

We can view an ecosystem model as a set of processes that provide an alternative way to encode its assumptions.

Process Model of Ross Sea Ecosystem

model RossSeaEcosystem

variables: phyto, zoo, nitro, residue

observables: phyto, nitro

process phyto_loss

$$\begin{aligned}\text{equations: } d[\text{phyto}, t, 1] &= -0.307 \times \text{phyto} \\ d[\text{residue}, t, 1] &= 0.307 \times \text{phyto}\end{aligned}$$

process zoo_loss

$$\begin{aligned}\text{equations: } d[\text{zoo}, t, 1] &= -0.251 \times \text{zoo} \\ d[\text{residue}, t, 1] &= 0.251 \times \text{zoo}\end{aligned}$$

process zoo_phyto_grazing

$$\begin{aligned}\text{equations: } d[\text{zoo}, t, 1] &= 0.615 \times 0.495 \times \text{zoo} \\ d[\text{residue}, t, 1] &= 0.385 \times 0.495 \times \text{zoo} \\ d[\text{phyto}, t, 1] &= -0.495 \times \text{zoo}\end{aligned}$$

process nitro_uptake

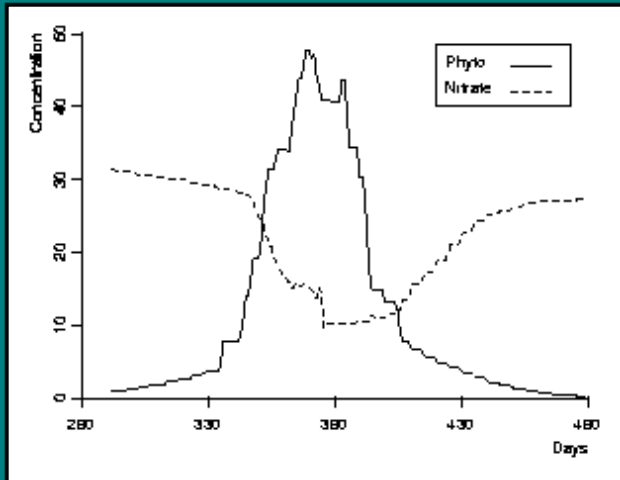
$$\begin{aligned}\text{equations: } d[\text{phyto}, t, 1] &= 0.411 \times \text{phyto} \\ d[\text{nitro}, t, 1] &= -0.098 \times 0.411 \times \text{phyto}\end{aligned}$$

process nitro_remineralization;

$$\begin{aligned}\text{equations: } d[\text{nitro}, t, 1] &= 0.005 \times \text{residue} \\ d[\text{residue}, t, 1] &= -0.005 \times \text{residue}\end{aligned}$$

Inductive Revision of Process Models

observations



revised model

```

model RossSeaEcosystem
variables: phyto, zoo, nitro, residue
observables: phyto, nitro
d[phyto,t,1] = - 0.307 × phyto - 0.495 × zoo
               + 0.411 × phyto
d[zoo,t,1] = - 0.251 × zoo + 0.615 × 0.495 × zoo
d[residue,t,1] = 0.307 × phyto + 0.251 × zoo
                + 0.385 × 0.495 × zoo - 0.005 × residue
d[nitro,t,1] = - 0.098 × 0.411 × phyto + 0.005 × residue
    
```

Revision

```

model RossSeaEcosystem
variables: phyto, zoo, nitro, residue
observables: phyto, nitro
d[phyto,t,1] = - 0.307 × phyto - 0.495 × zoo
               + 0.411 × phyto
d[zoo,t,1] = - 0.251 × zoo + 0.615 × 0.495 × zoo
d[residue,t,1] = 0.307 × phyto + 0.251 × zoo
                + 0.385 × 0.495 × zoo - 0.005 × residue
d[nitro,t,1] = - 0.098 × 0.411 × phyto + 0.005 × residue
    
```

initial model

```

process exponential_growth
variables: P {population}
equations: d[P,t] = [0, 1, ∞] × P

process logistic_growth
variables: P {population}
equations: d[P,t] = [0, 1, ∞] × P × (1 - P / [0, 1, ∞])

process constant_inflow
variables: I {inorganic_nutrient}
equations: d[I,t] = [0, 1, ∞]

process consumption
variables: P1 {population}, P2 {population}, nutrient_P2
equations: d[P1,t] = [0, 1, ∞] × P1 × nutrient_P2,
            d[P2,t] = - [0, 1, ∞] × P1 × nutrient_P2

process no_saturation
variables: P {number}, nutrient_P {number}
equations: nutrient_P = P

process saturation
variables: P {number}, nutrient_P {number}
equations: nutrient_P = P / (P + [0, 1, ∞])
    
```

generic processes

Generic Processes for Aquatic Ecosystems

generic process exponential_loss

variables: S{species}, D{detritus}

parameters: α [0, 1]

equations: $d[S, t, 1] = -1 \times \alpha \times S$
 $d[D, t, 1] = \alpha \times S$

generic process remineralization

variables: N{nutrient}, D{detritus}

parameters: π [0, 1]

equations: $d[N, t, 1] = \pi \times D$
 $d[D, t, 1] = -1 \times \pi \times D$

generic process grazing

variables: S1{species}, S2{species}, D{detritus}

parameters: ρ [0, 1], γ [0, 1]

equations: $d[S1, t, 1] = \gamma \times \rho \times S1$
 $d[D, t, 1] = (1 - \gamma) \times \rho \times S1$
 $d[S2, t, 1] = -1 \times \rho \times S1$

generic process constant_inflow

variables: N{nutrient}

parameters: v [0, 1]

equations: $d[N, t, 1] = v$

generic process nutrient_uptake

variables: S{species}, N{nutrient}

parameters: τ [0, ∞], β [0, 1], μ [0, 1]

conditions: $N > \tau$

equations: $d[S, t, 1] = \mu \times S$
 $d[N, t, 1] = -1 \times \beta \times \mu \times S$

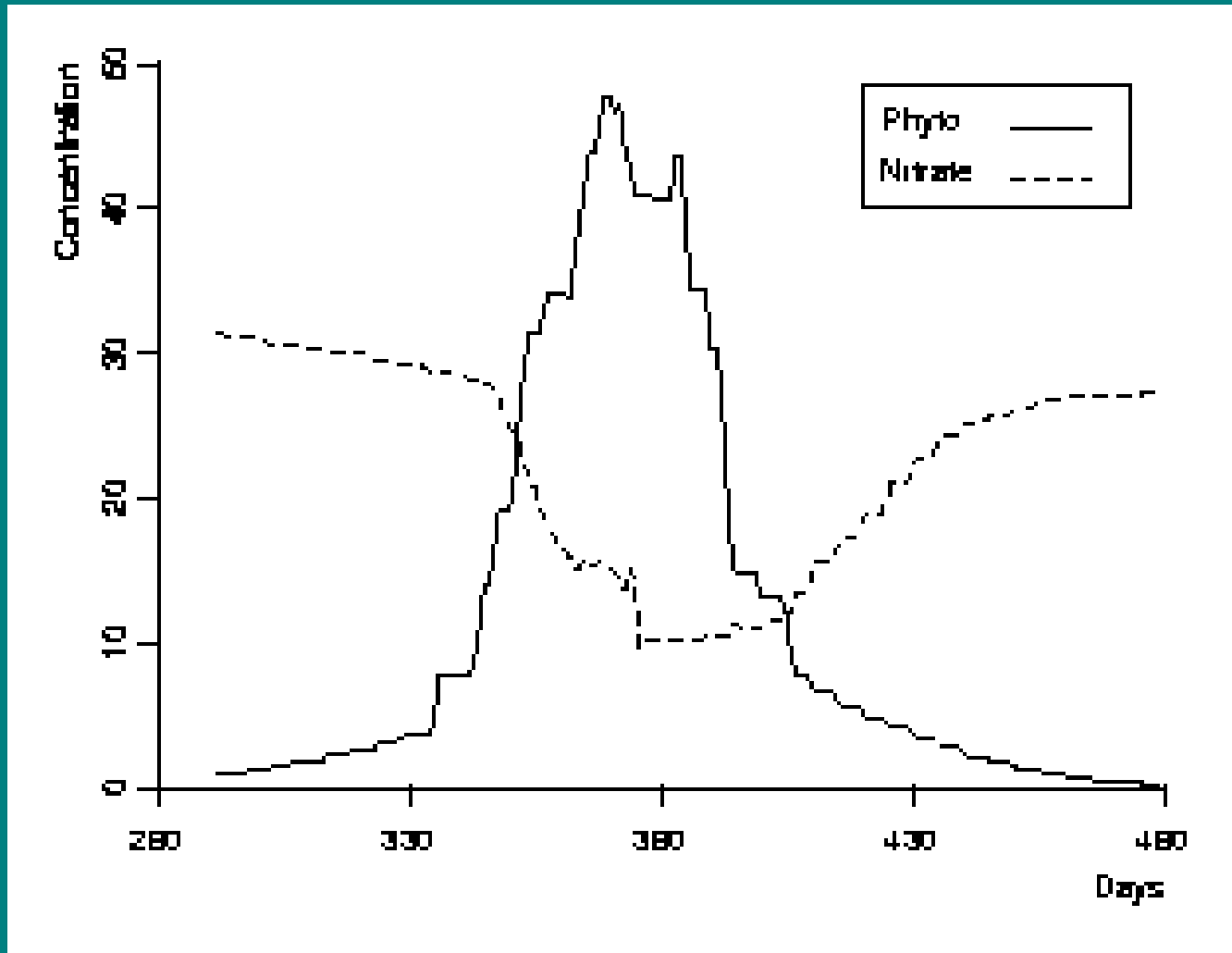
A Method for Process Model Revision

We have implemented RPM, an algorithm that revises an initial process model in four main stages:

1. Find all ways to instantiate available generic processes with specific variables, subject to type constraints;
2. Generate candidate model structures by deleting the current processes and adding new ones, subject to complexity limits;
3. For each generic model, carry out search through parameter space to find good coefficients [difficult];
4. Return a list of revised models ordered by their overall scores.

The evaluation metric can be squared error or description length based on error and distance from the initial model.

Observations from the Ross Sea



Revised Model of Ross Sea Ecosystem

model RossSeaEcosystem

variables: phyto, zoo, nitro, residue, light, G, growth_rate, nitro_rate, light_rate

observables: phyto, nitro, light

$$d[\text{phyto}, t, 1] = -0.307 \times \text{phyto} - G \times \text{zoo} + \text{growth_rate} \times \text{phyto}$$

$$d[\text{zoo}, t, 1] = 0.615 \times G \times \text{zoo}$$

$$d[\text{residue}, t, 1] = 0.307 \times \text{phyto} + 0.385 \times G \times \text{zoo} - 0.083 \times \text{residue}$$

$$d[\text{nitro}, t, 1] = -1 \times n_to_c \times \text{growth_rate} \times \text{phyto} + 0.083 \times n_to_c \times \text{residue}$$

$$G = 0.415 \times (1 - \exp(-1 \times 0.27 \times \text{phyto}))$$

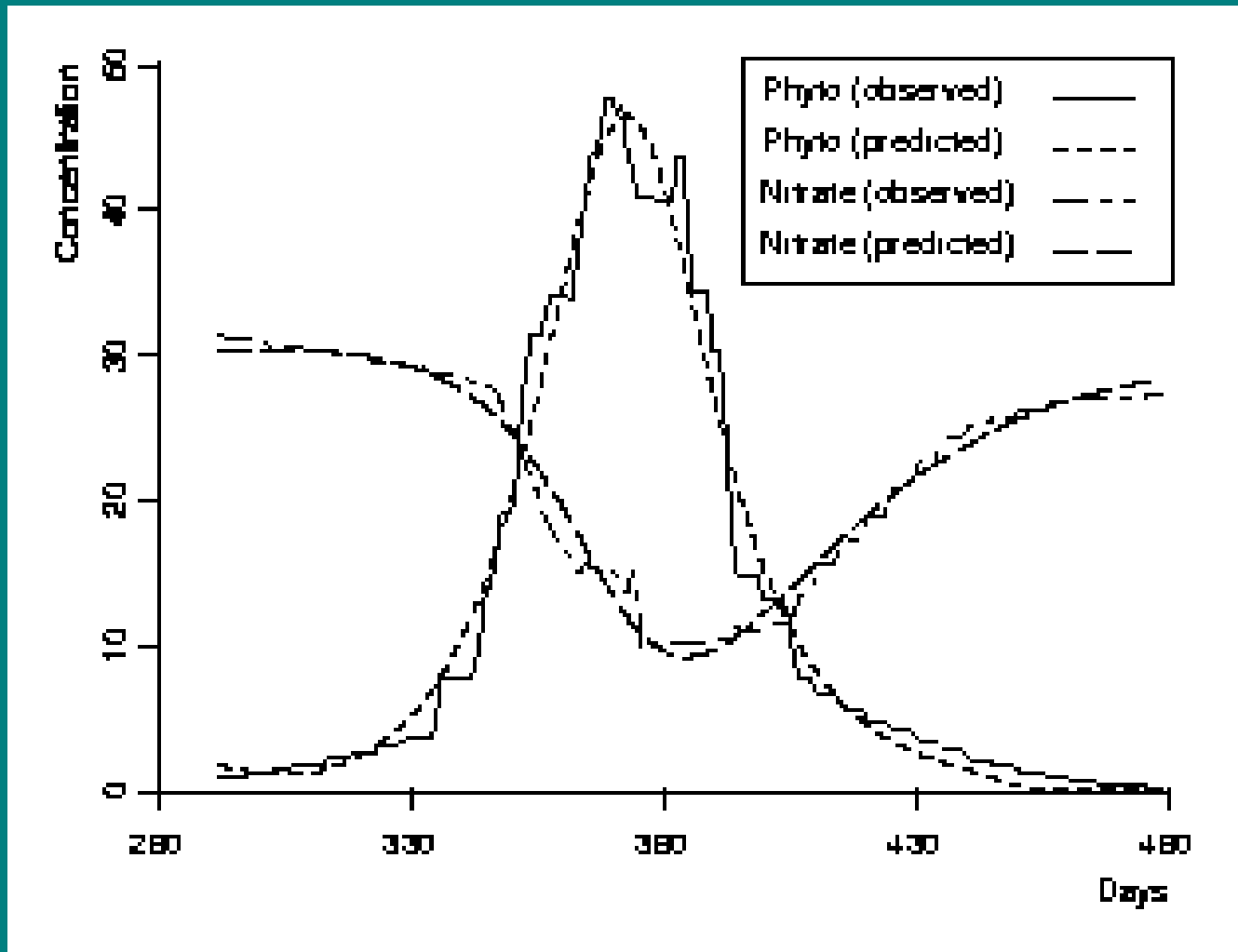
$$\text{growth_rate} = r_max \times \min(\text{nitro_rate}, \text{light_rate})$$

$$\text{nitro_rate} = \text{nitro} / (\text{nitro} + 4.33)$$

$$\text{light_rate} = \text{light} / (\text{light} + 11.67)$$

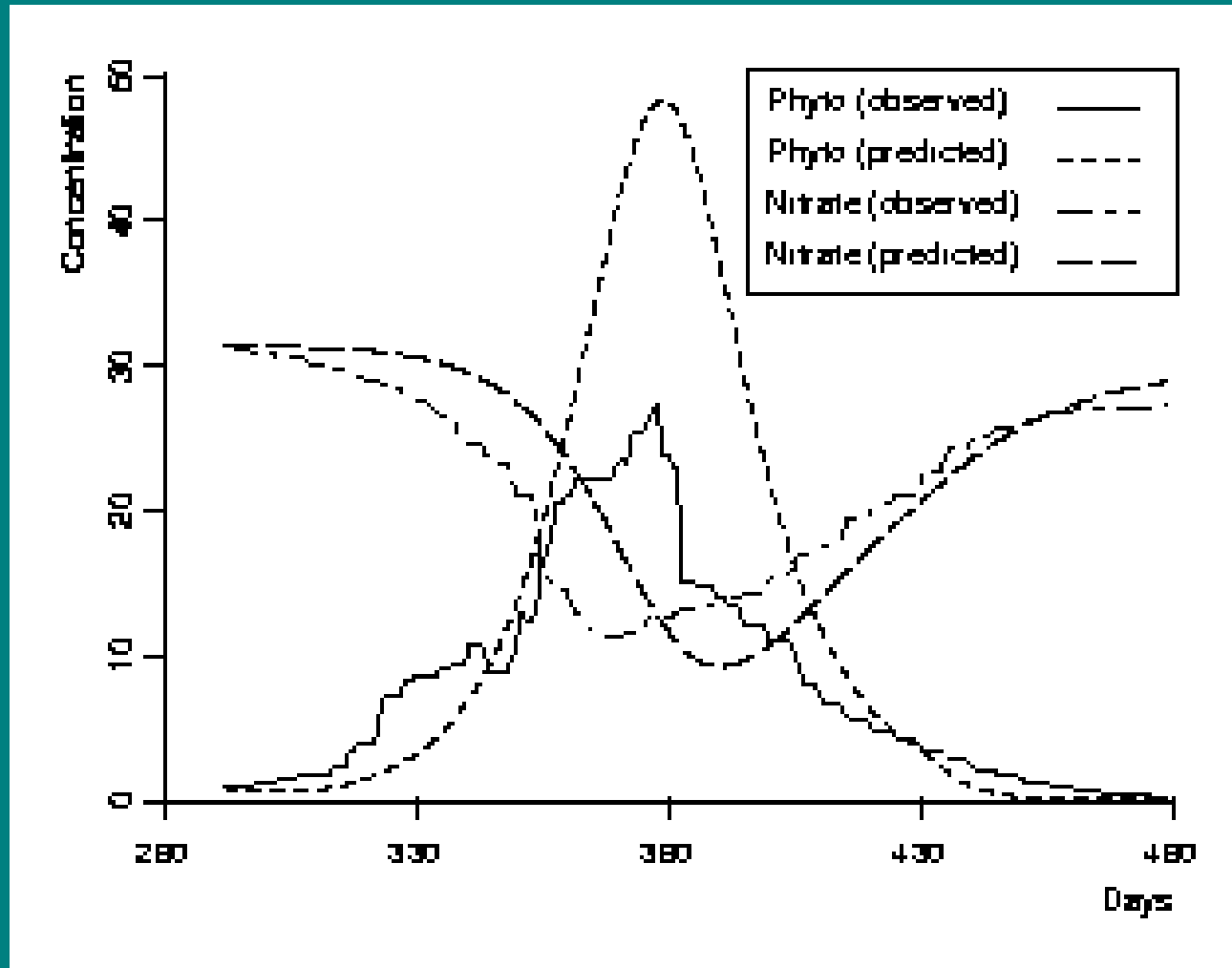
$$n_to_c = 0.251, r_max = 0.194, \text{remin_rate} = 0.0676$$

Initial Results on Ross Sea Training Data



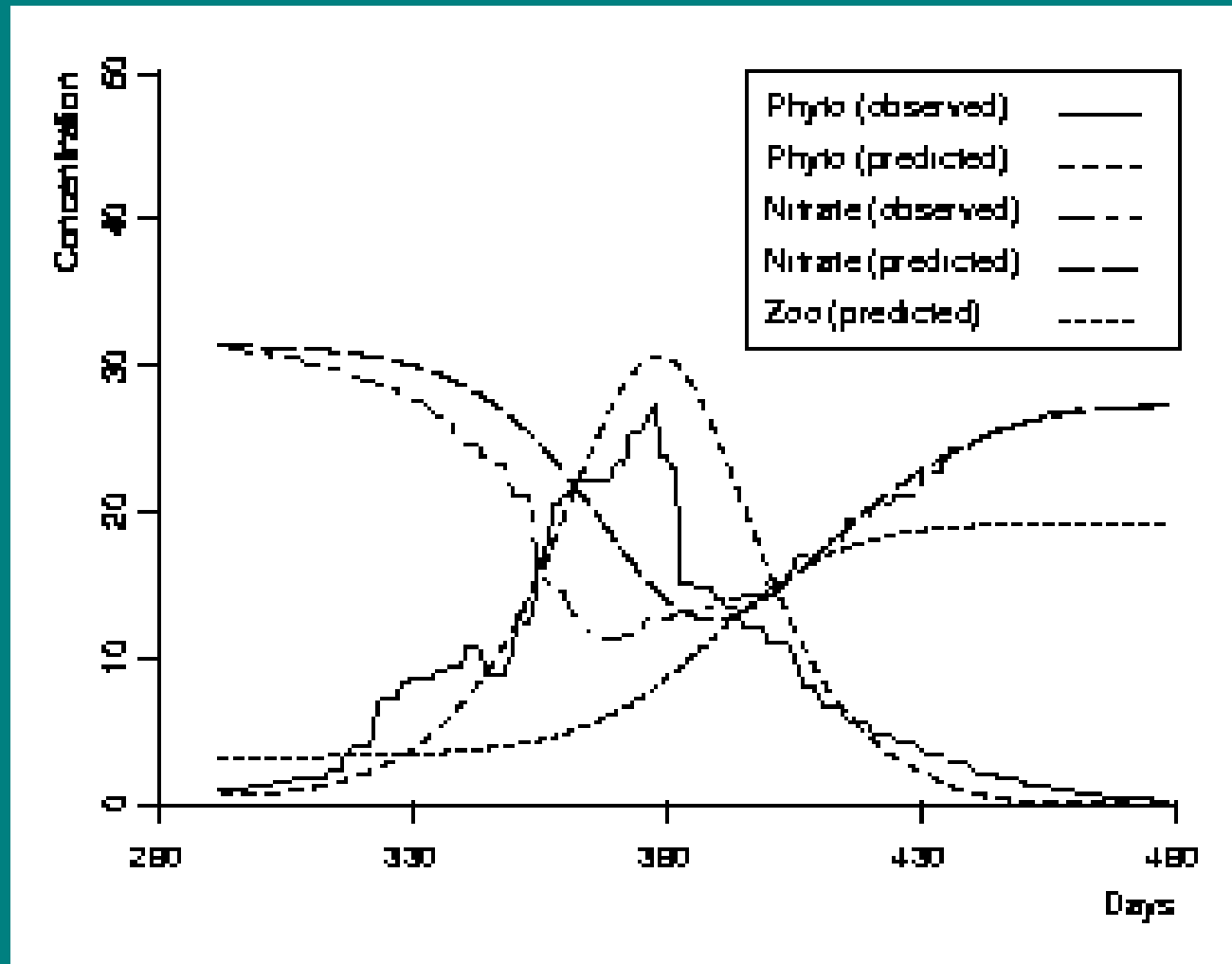
The best revised model reproduces the observations quite well.

Initial Results on Ross Sea Test Data



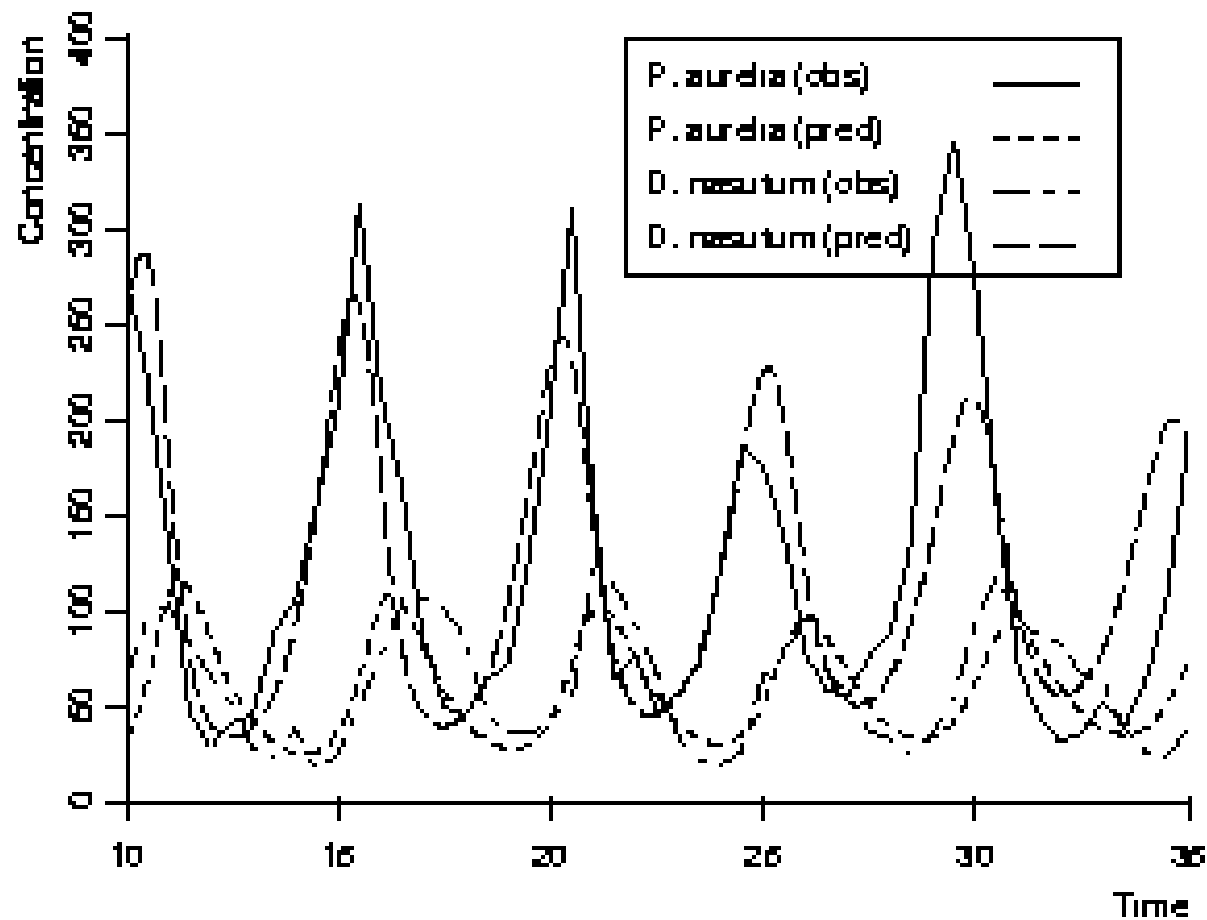
But the model predicts nearly the same behavior for both years.

Revised Results on Ross Sea Test Data

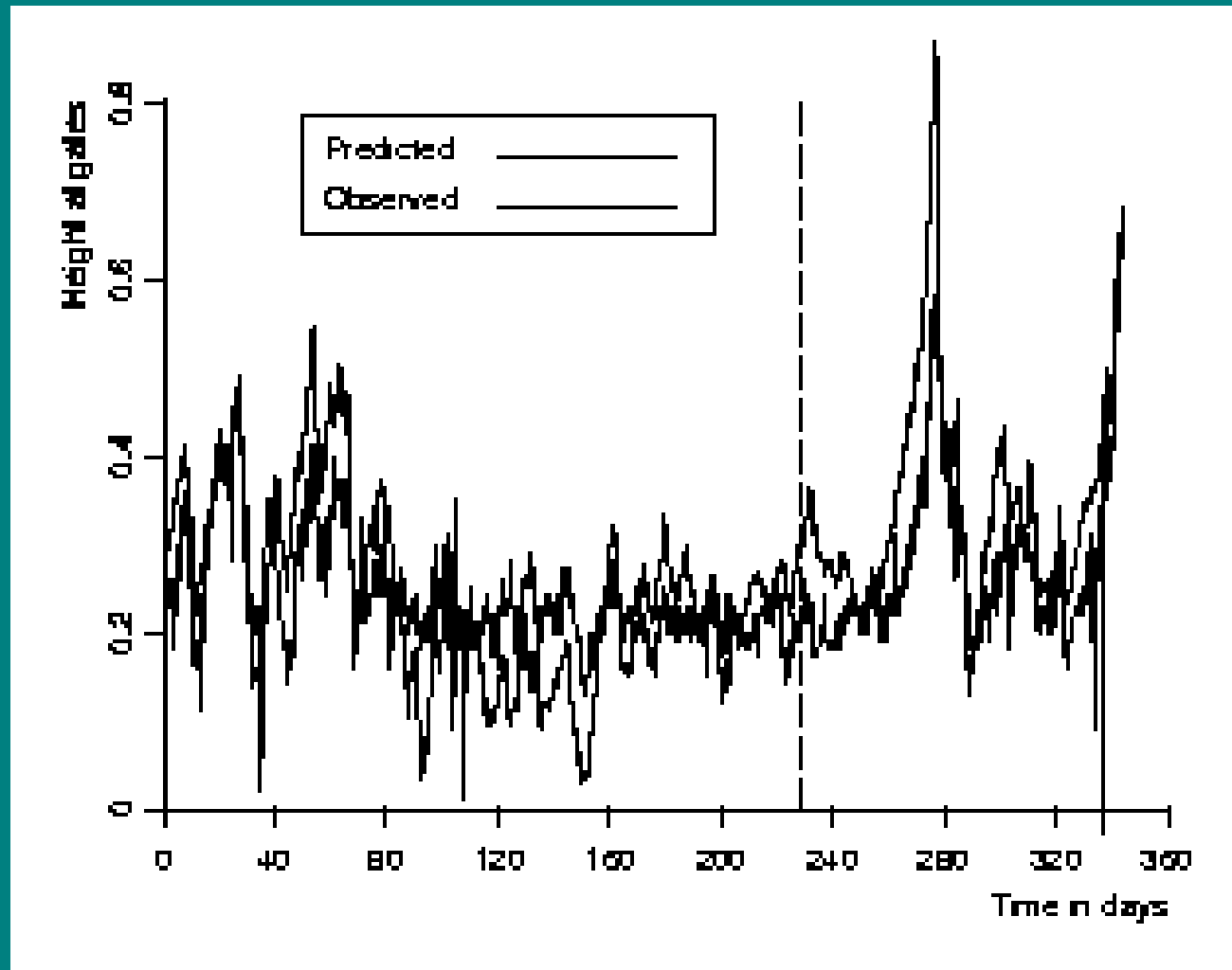


Refitting initial values for zooplankton gives better generalization.

Results on Data from Protist Study



Results on Data from Rinkobing Fjord



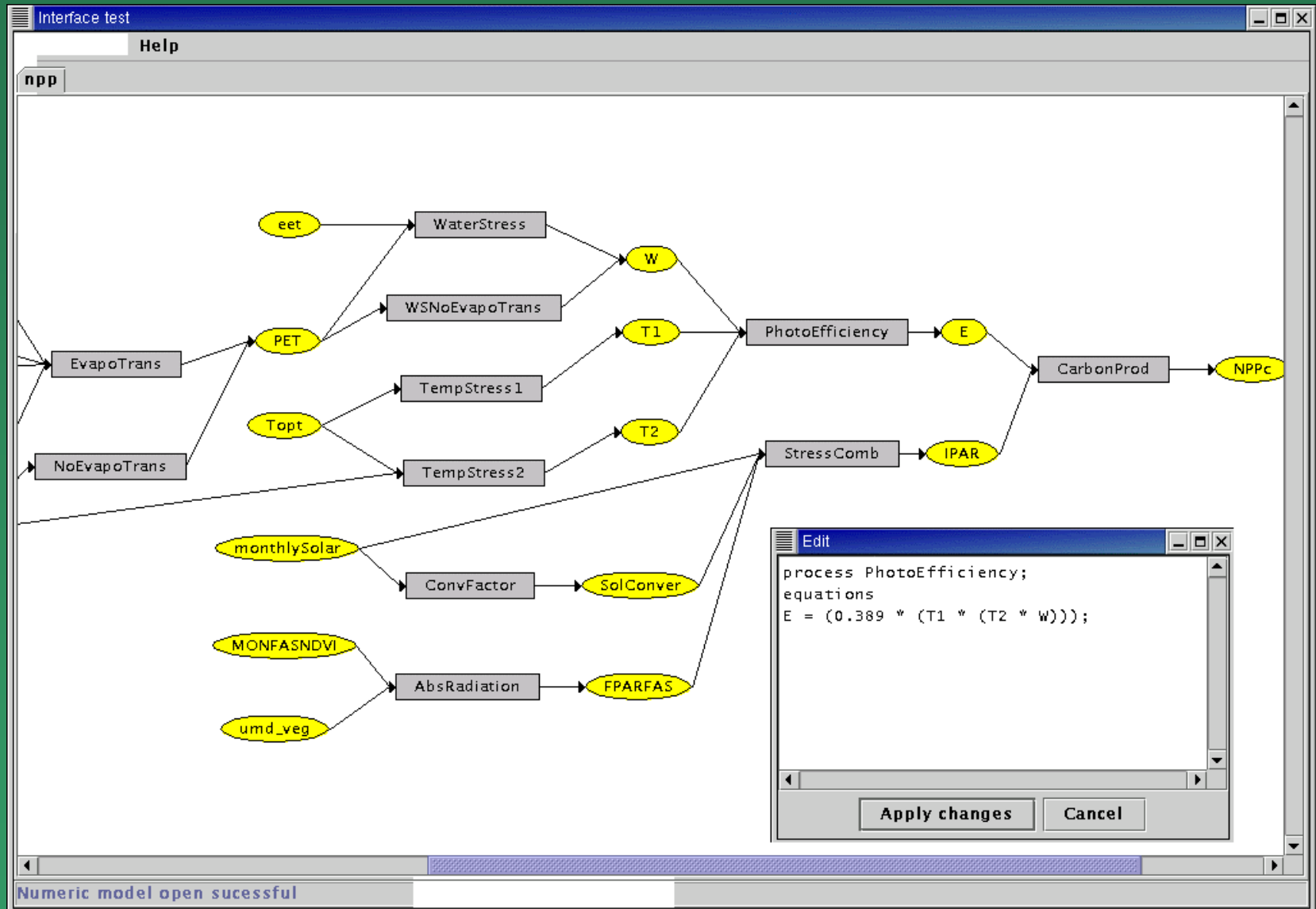
Interfacing with Scientists

Because few scientists want to be replaced, we are developing PROMETHEUS, an interactive environment that lets users:

- specify a quantitative process model of the target system;
- display and edit the model's structure and details graphically;
- simulate the model's behavior over time and situations;
- compare the model's predicted behavior to observations;
- invoke a revision module in response to detected anomalies.

The environment offers computational assistance in forming and evaluating models but lets the user retain control.

Viewing and Editing a Process Model



Intellectual Influences

Our approach to computational discovery incorporates ideas from many traditions:

- computational scientific discovery (e.g., Langley et al., 1983);
- theory revision in machine learning (e.g., Towell, 1991);
- qualitative physics and simulation (e.g., Forbus, 1984);
- languages for scientific simulation (e.g., STELLA, MATLAB);
- interactive tools for data analysis (e.g., Schneiderman, 2001).

Our work combines ideas from machine learning, AI, programming languages, and human-computer interaction.

Directions for Future Research

Despite our progress to date, we need further work in order to:

- produce additional results on other ecosystem modeling tasks
- develop improved methods for fitting model parameters
- implement heuristic methods for searching the structure space
- utilize knowledge of subsystems to further constrain search
- augment the modeling environment to make it more usable

Process modeling has great potential to aid model development in environmental science.

Contributions of the Research

In summary, our work on computational discovery has produced:

- a new formalism for representing scientific process models;
- an encoding for background knowledge as generic processes;
- a method for revising process models using time-series data;
- an interactive environment for model construction/utilization.

We have demonstrated this approach to model revision on both ecosystem modeling and an environmental domain.

The PROMETHEUS modeling/revision environment is available at:

<http://www.isle.org/process.html>

End of Presentation