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Water resources system analysis: tools and challenges

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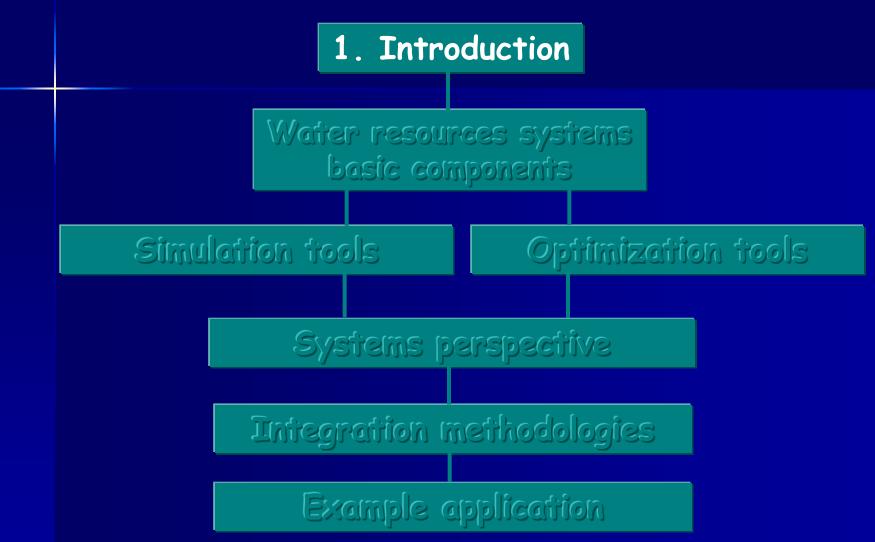
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Scientific and practical challenges in modeling water resources systems (Part 1)

 Example application: "A Hybrid Model Tree (MT) - Genetic Algorithm (GA) Scheme for Toxic Cyanobacteria Predictions in Lake Kinneret" (Part 2)

Questions





General: water resources systems analysis – operations research (OR) of "water components" (e.g., reservoirs, rivers, watersheds, groundwater, distribution systems, etc.), as standalone or integrated, for single or multiobjective problems, deterministic or stochastic

 <u>Specific:</u> water distribution systems analysis (water security), watershed management, surface water

 <u>Tools</u>: "traditional" OR (LP, NLP), data driven modeling (neural networks, model trees), evolutionary computation (Genetic Algorithms, Ant Colony, Cross Entropy) – single/multiobjective

□ The scientific and practical challenge in dealing quantitatively with water resources management problems is in taking into consideration from a <u>systems</u> <u>perspective</u>, social, economical, environmental, and technical dimensions, and integrating them into a single framework for trading - off in time and in space competing objectives

□ Inherently, such problems involve modeling of water quantity and quality for water resources systems components, such as: surface water, groundwater, water distribution systems, reservoirs, rivers, lakes, and others, as stand alone or linked elements

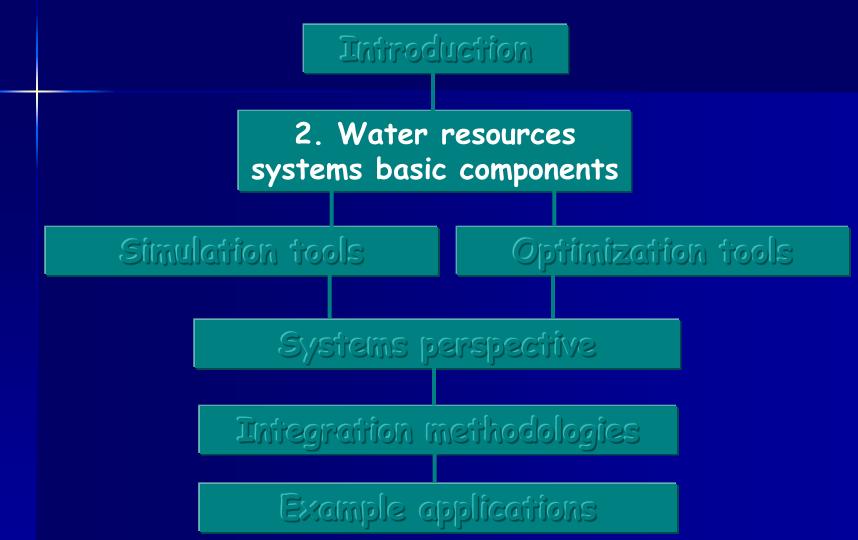
Conceptual issues which should be overcomed for constructing and solving a water resources management problem

Description of available tools for dealing with water resources management problems: available simulation programs (e.g., WMS, GMS, SWAT, AVGWLF, EPANET), optimization methodologies (e.g., simulated annealing, genetic algorithms, ant colony, cross entropy, non-linear programming, linear programming)

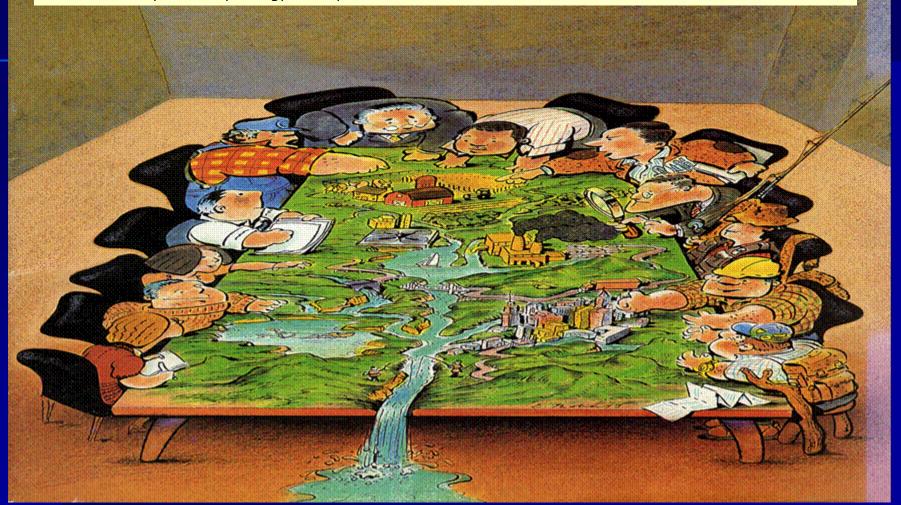
Integration approaches: "embedding" ; "linking" ; "hybridizing"

Case study for demonstrating the "hybridizing" approach

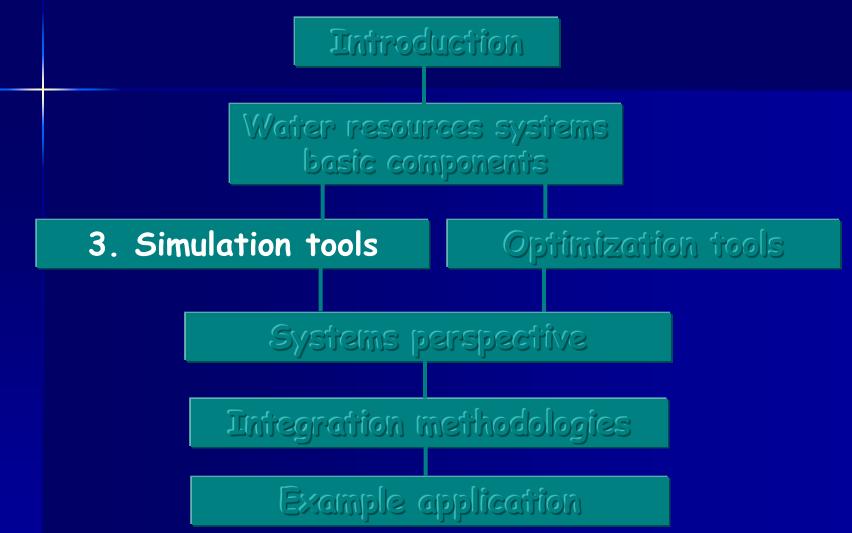




Loucks, Daniel P.; Beek, Eelco van; Stedinger, Jery R.; Dijkman, Jozef P.M.; Villars, Monique T. (2005). "Water resources systems planning and management: an introduction to methods, models and applications." Studies and reports in hydrology; 680 p., illus.



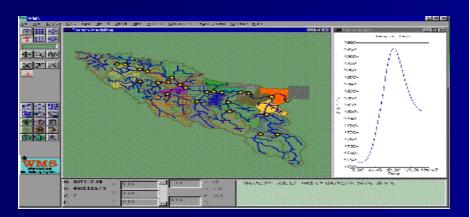
"Everything" interacts within river basins

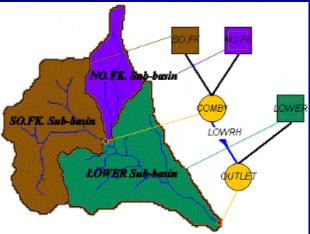


Surface water - WMS, BASINS, others <u>WMS - Watershed Modeling System</u>

Merging information obtained from terrain models and GIS with lumped parameter traditional hydrologic analysis models such as HEC-1 and TR-20, having an ability to take advantage of digital terrain for hydrologic data development

WMS uses three primary data sources for model development: 1. Geographic Information Systems (GIS) Data 2. Digital Elevation Models (DEMs) 3. Triangulated Irregular Networks (TINs)

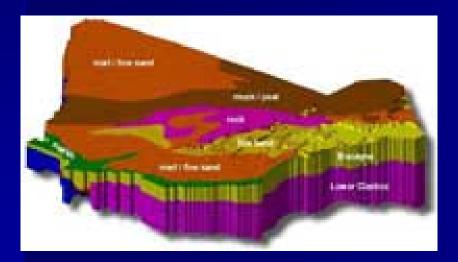


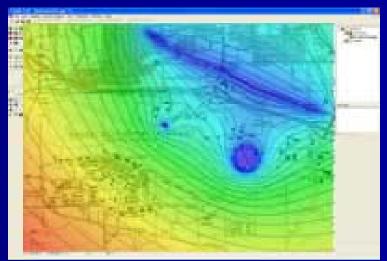


http://www.ems-i.com/

Groundwater – GMS, Visual MODFLOW, others <u>GMS – Groundwater Modeling System</u>

Providing GIS tools for groundwater simulation including site characterization, model development, calibration, postprocessing, and visualization, supporting both finitedifference and finite-element models in 2D and 3D including links to MODFLOW, MODPATH, and others

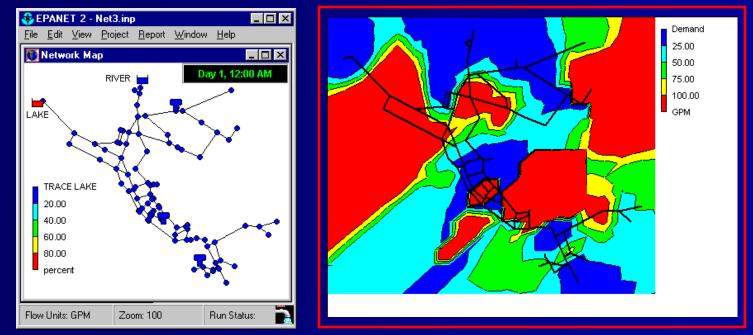




http://www.ems-i.com/

Water networks - EPANET, WaterCad, others EPANET

Performing extended period simulation of hydraulic and water-quality behavior within pressurized pipe networks consisting of pipes, consumers, pumps, valves and storage tanks

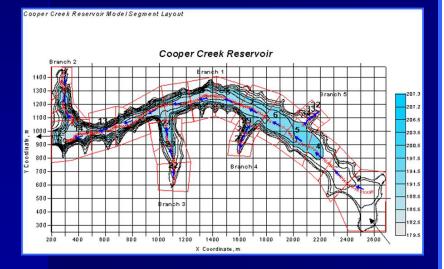


http://www.epa.gov/nrmrl/wswrd/epanet.html

□ Lakes and reservoirs - CE-QUAL-W2, ELCOM, others

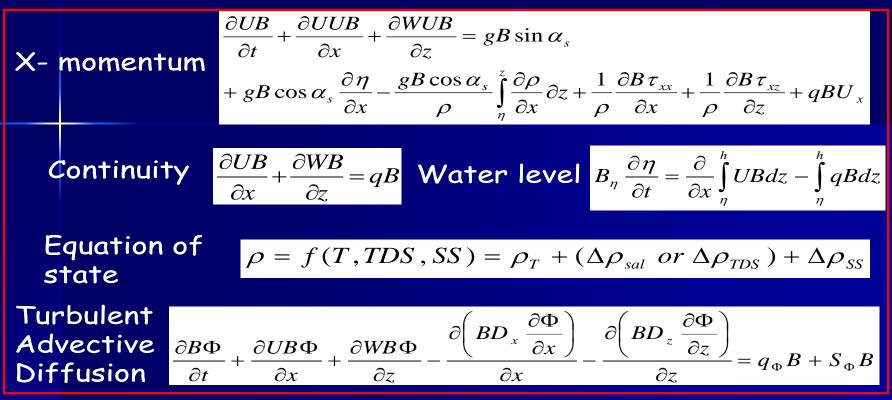
<u>CE-EQUAL-W2</u>

A two-dimensional , longitudinal/vertical, hydrodynamic and water quality model for reservoirs, lakes and estuaries, which models basic eutrophication processes such as temperature-nutrientalgae-dissolved oxygen-organic matter and sediment relationships



http://www.ce.pdx.edu/w2/

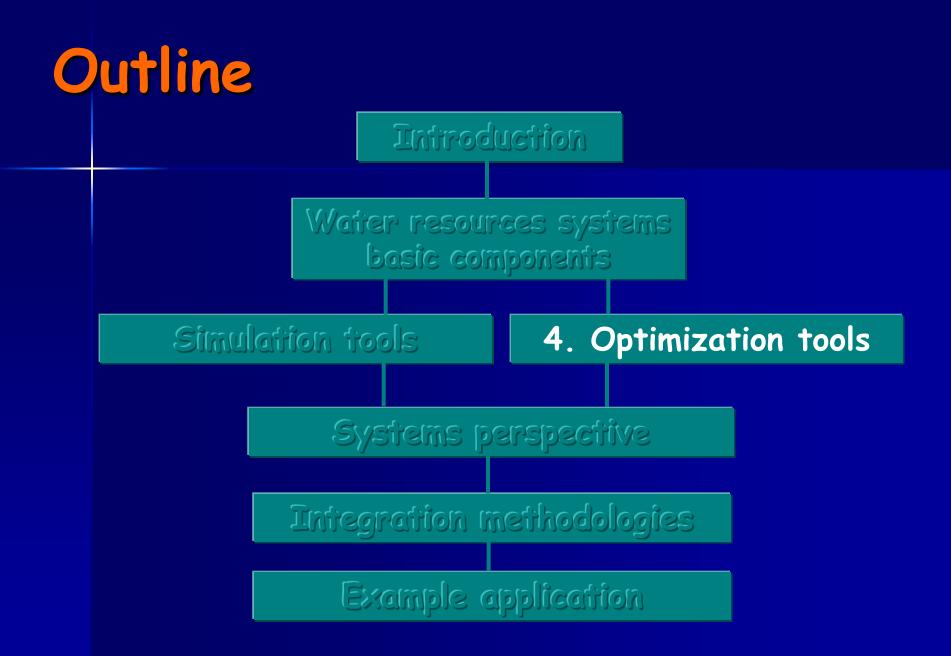
CE-QUAL-W2 Hydrodynamic equations



Two observations:

> The physical representation of water resources elements might be quite complex

This directs the possible solution methods and approaches employed for optimizing water resources systems



Optimization tools

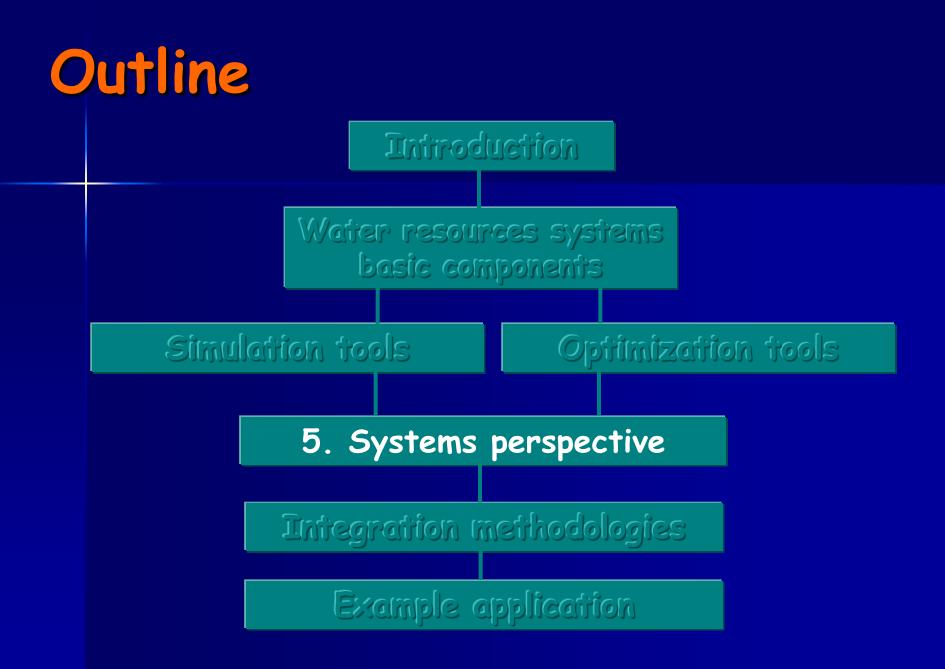
Two main categories: (1) Classical (2) Heuristic

Classical: LP (Linear Programming), NLP (Nonlinear Programming, e.g. GRG = General Reduced Gradient), Dynamic Programming, others

<u>Advantages:</u> provide analytical tools to solve optimization problems; <u>Limitations:</u> restricted (number of constraints, decision variables, model properties)

Heuristic: Simulated Annealing, Genetic Algorithms, Ant Colony, Tabu Search, Cross Entropy, others

<u>Advantages:</u> not restricted; <u>Limitations:</u> no analytical assurance of an optimal solution, highly computational intensive



Systems perspective

> The goal is to find a Pareto optimal solution set or a non-dominated solution set

Each solution in the Pareto optimal set is optimal in the sense that it is not possible to improve one objective without making at least one of the others worse

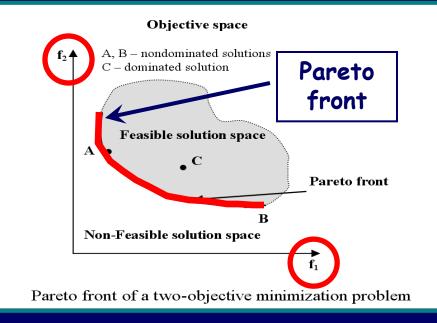
Optimize:
$$F(x) = (f_1(x), f_2(x), ..., f_M(x))^T$$

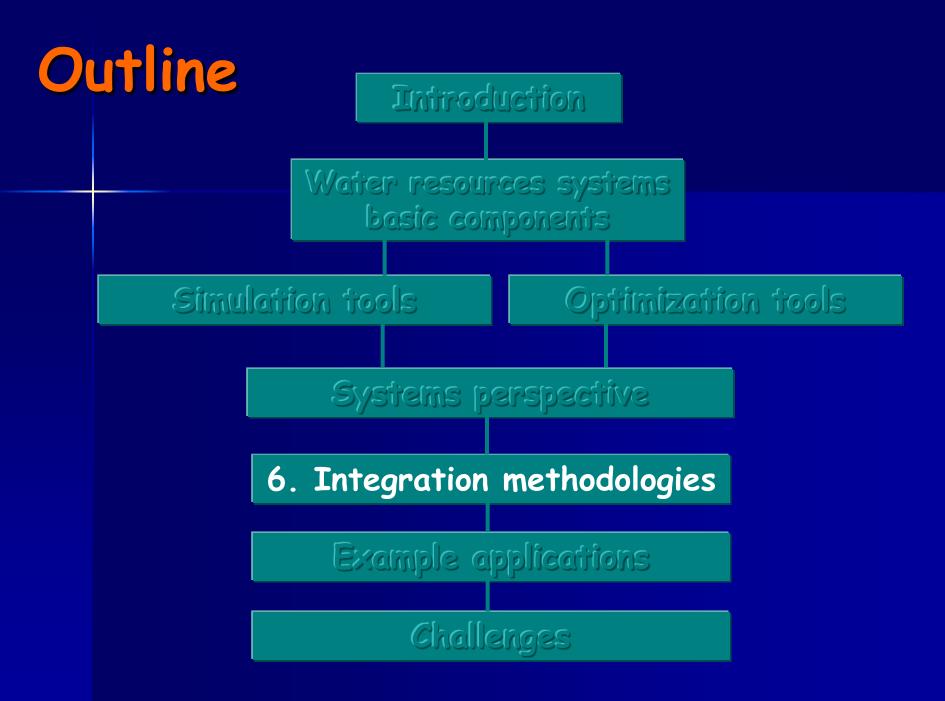
Subject to:

 $g_i(x) > 0$, i = 1, 2, ..., k k Inequality constraints

 $e_j(x) = 0, \quad j = 1, 2, ..., l \qquad l$ Equality constraints

where $x = (x_1, x_2, ..., x_n)^T$ is the vector of decision variables





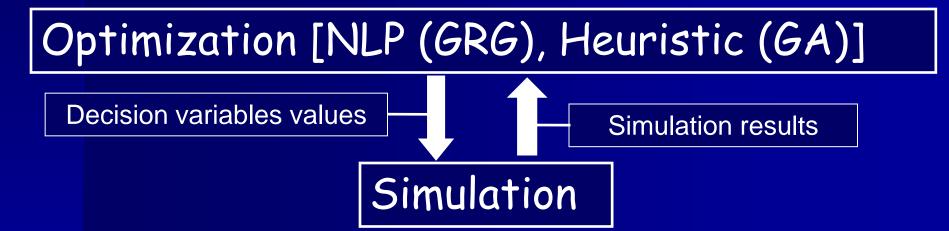
Integration methodologies

Three basic approaches:

- Embedding direct; traditional
- □ Simulation Optimization (NLP) (Late 1980's)
- Simulation Optimization (Heuristic) (Last decade)
- Hybrid modeling: Simulation Data Driven Modeling - Heuristic Optimization (Lately)

Integration methodologies

Simulation - Optimization (NLP) (Late 1980's) ; Simulation - Optimization (Heuristic) (Last decade)



If the simulation stage is time consuming then the optimization process can become endless (e.g., a one 5 minute simulation duration for 100 GA generations with 100 strings at each generation, will result a computational time effort of about 35 days)

Integration methodologies

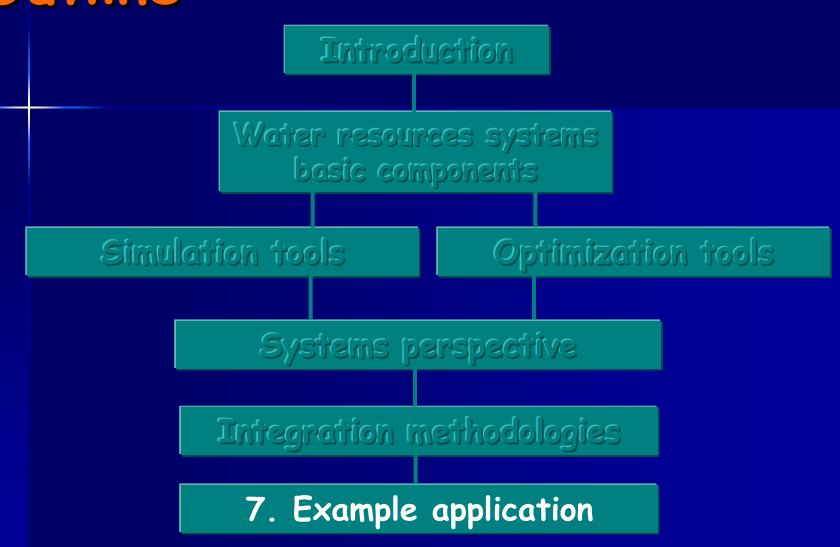
Hybrid modeling: Simulation - Data Driven modeling - Heuristic Optimization (Lately)



Decision variables values

Simulation results

Simplified Simulation + Data Driven Modeling



A Hybrid Model Tree (MT) – Genetic Algorithm (GA) Scheme for Toxic Cyanobacteria Predictions in Lake Kinneret

By Avi Ostfeld¹, Ariel Tubaltzev¹ Meir Rom², and Lea Kronaveter²

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² Lake Kinneret Watershed Unit, Mekorot, Israel National Water Company, Israel

This work presents a hybrid model tree (MT) genetic algorithm (GA) scheme for toxic Cyanobacteria (Blue Algae) predictions in Lake Kinneret (Sea of Galilee), which is the most important surface water resource in Israel

For more than 30 years, up to 1994, there have been no major problems with respect to the water quality of the lake

In 1994 toxic Cyanobacteria (blue algae) blooms appeared, which suggests that the future water quality of the lake might be at risk

□ A full physical understanding of the reasons for the toxic Cyanobacteria blooms is lacking

□ This study suggests a data driven modeling approach, relying on the vast existing data base of the lake, to explore the possible major factors causing the toxic Cyanobacteria to bloom, and to predict their possible appearance

□ The suggested model is a hybrid model-trees (MT) genetic algorithm (GA) scheme

□ A few words about Lake Kinneret, model-trees (MT), and genetic algorithms (GA's) ---

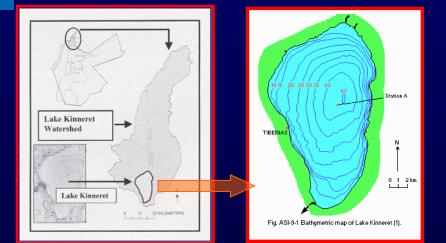
Lake Kinneret

Lake Kinneret - the Sea of Galilee 32:50N, 35:35E; -209 m above sea level.

A warm, monomictic lake, located at the northern end of the Afro. Syrian Rift Valley in Northern Israel.

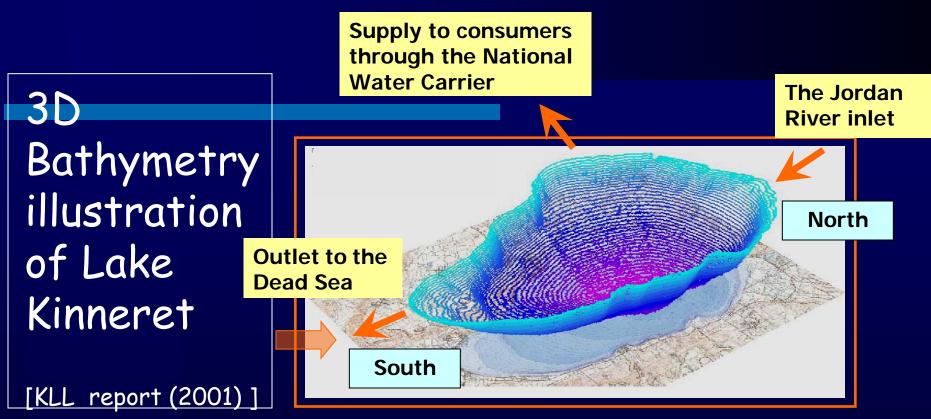
PHYSICAL DIMENSIONS

Surface area - 170 km² Volume - 4 km³ (4000 MCM) Maximum depth - 43 m Mean depth - 25.6 m Length of shoreline - 53 km Residence time - 4.8 yr Catchment area - 2,730 km²





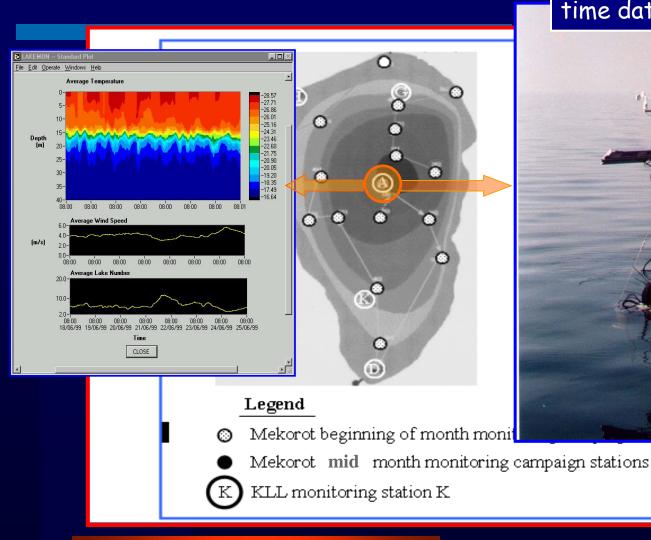
Lake Kinneret



The largest natural fresh surface water in Israel providing approximately 35% of its annual drinking water, a proportion that is constantly increasing.

Monitored Data

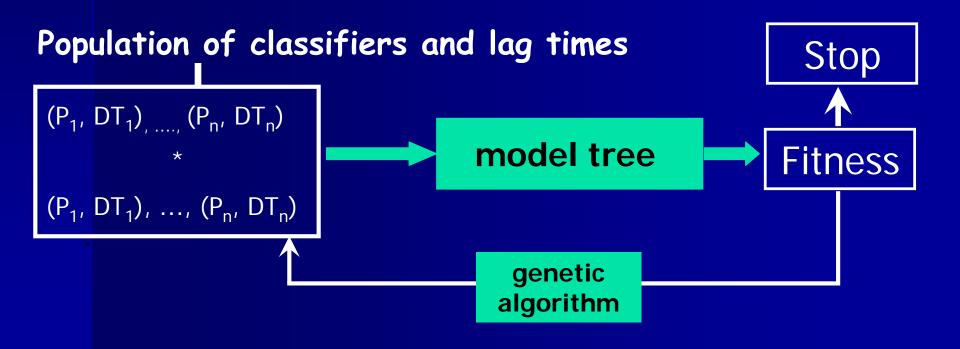
LDS- Wind recorder, Short & long-wave radiation, air temp, thermister chain & radio antenna installed at sta. A sends realtime data to KLL

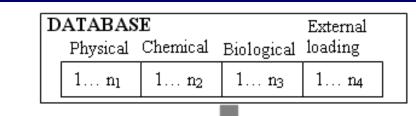


Methodo	loav			
Input (Database)		Each of the parameters is measured on a fortnight basis		
1. Physical	> what classifiers should be chosen so as to maximize the model prediction ability ?			
2. Chemical				
3. Biological	> the model trees are predicting; the GA searches the best			
4. External loading	classifiers			
	River (e.g., Nitrat Phosphorus, etc.)			

Methodology

An iterative scheme for searching the optimal set of classifiers and their corresponding lag times which maximizes the model prediction ability for a given database component





Stage A. INITIALIZATION

- * Choose randomly a predefined number of parameters from the database (e.g., one parameter from each database group OR any four parameters).
- * Choose randomly a time lag for each of the chosen parameters (e.g., three weeks).
- * <u>Repeat</u> until a set of parameters and corresponding time lags are selected (i.e., setting the first GA population).

Stage B. MODEL TREE (MT) CONSTRUCTION

- * Build a model tree for each of the selected parameters and time lags (i.e., the GA strings) using Cubist (Quinlan, 1993) for predicting a chosen parameter (e.g., Toxic Nitrogen Fixing Cyanobacteria Algae).
- * Assign each string a fitness equal to its correlation coefficient calculated by Cubist using a predefined cross validation dataset.

Stage C. GENETIC ALGORITHM (GA)

- * Perform Selection, Crossover, and Mutation on the current population using optiGA (Salomons, 2002).
- * Construct a new population of strings (i.e., parameters and time lags).
- * <u>Check</u> if STOPPING conditions are met. If stopping conditions are met define the corresponding highest correlation coefficient string as the OPTIMAL solution (i.e., parameters and time lags), <u>otherwise</u> go back to Stage B.

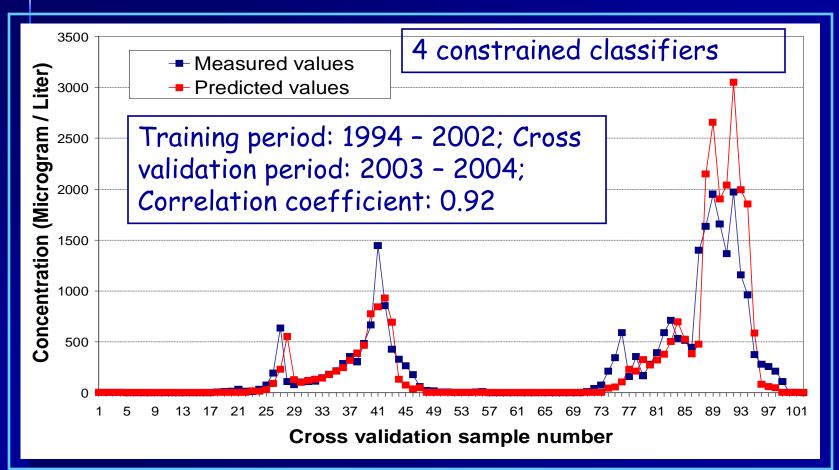
Methodology

The methodology is cast in a program entitled KAPM (Kinneret Algae Prediction Model), it's main interface menu is:

🐂 Blue Algae			
Data Settings			GA Options
Target field name:	CTXF 1m	Predicted	Population Size 100
Build the tree using data from	1994	component	Iterations Number 100
Training to	2002		
			GA
Test tree using data from Validation to	2003		
Validation to	2004		Run
Data based analysis			Make XLS file from .pred
Number of fields from Physical table Huri table	1	Classifiers	
Chemical table	1	· · · · · · · · · · · · · · · · · · ·	
Biological table		method	
🔿 General analysis			
Fields number	4		
Report			
Best Coefficient	<u></u>		
.92 Report			
Current Iteration			
100			

Application

Predicting (CTXF 1m) Toxic Nitrogen Fixing Cyanobacteria Algae blooms at station A (deepest point in the lake) at a depth of 1 m (validation results):



Application

Optimal model tree rules:

```
Rule 1:
 IF
      W T emp <= 26.55
      CTXF 5m \le 71.23
  THEN
      CTXF 1m = 3.91
                                 . . .
 Rule 13:
 IF
      W T emp > 26.81
     CTXF 5m > 725.00
     CTXF 5m <= 1328.42
  THEN
      CTXF 1m = -208.3 + 38 (WTemp) - 855 (NTOT 5-10) +
                 + 0.38 (CTXF 5m)
Rule 14:
 IF
      NTOT (Huri) <= 1.90
     NTOT 5-10 \le 0.37
      CTXF 5m > 725.00
  THEN
      CTXF 1m = 792.54 + 0.38 (CTXF 5m)
 Rule 15:
 IF
      CTXF 5m > 1328.42
  THEN
      Output = 4463.67 - 1.06 (CTXF 5m)
```

Conclusions

The methodology and application of a coupled model trees (MT) - genetic algorithm (GA) scheme for predicting blue-algae blooms in Lake Kinneret was presented and demonstrated

The model was able to satisfactorily predict bluealgae blooms using a database which incorporates physical, chemical, biological, and external loading data for the lake

One of the possible extension options of the model is to explore the tradeoff between the model prediction ability versus the optimal correlation coefficient (i.e., fitness) using multi-objective optimization, assuming that these two objectives compete Thanks! Questions?