

# Modeling Languages and Global Optimization



#### **Languages and Conceptual Design Features**

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## The COCONUT project



- European Union research and development project
- Partners from six European universities:
   Nantes, Lausanne, Vienna
   Louvain-la-Neuve, Coimbra, Darmstadt
   and an industrial partner:
   II OG
- Aimed at the integration of the existing approaches to continuous global optimization and constraint satisfaction
- December 2000 November 2003
- Uses AMPL as modeling language, also a GAMS interface
- http://www.mat.univie.ac.at/coconut







### **Modeling**



- From the Latin word "modellus", typical human way of coping with the reality
- Mechanical models
  - Astronomy
  - Architecture
- Mathematical models used since Hellenic Age
  - Geometry (~600 BC (Thales) ~300 BC (Euclid))
  - Circumference / Diameter of Earth (~250 BC (Erathostenes))
  - Arithmetic (~240 AD (Diophantus) ~800 AD (Al-Khwarizmi))
  - Astronomy (~150 AD (Ptolemy))
  - Physics
  - Economics
- Visual models
  - Anatomy



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## Modeling (ctd.)



- A model is a simplified version of something that is real.
- Represents a real-world object by other, simpler objects.
- Structures the knowledge about the real world.
- Reduction to those phenomena and aspects which are considered important.
- Usefulness is restricted to its scope of application.
- Serve many different needs ≈ the modeling goal
  - explain phenomena, make predictions,
  - control environment, decision making,
  - maintenance, normative components,
  - communication



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## **Mathematical Modeling**



- Represents a real-world problem by mathematical objects in a formalized mathematical language.
- Can be analyzed by mathematical theory and algorithms.
- Represents the knowledge about the real-world by concepts, variables, relations, data.
- Makes models accessible to computers.
  - First computers were human (usually underpaid women).
  - Since ENIAC (1945) larger and more complicated mathematical models have become accessible.
- Models became attractive to military and industry.

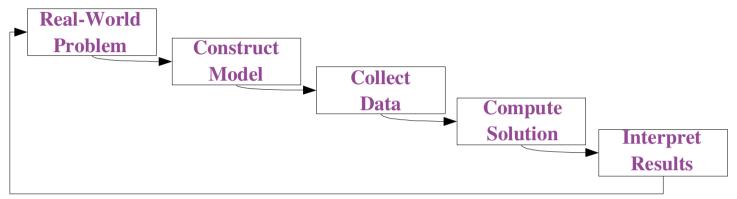






## Modeling – Scheme





- Modeling is done in "cycles".
- Some stages need the help of the end-user.
- For complex models, a computer is needed in the Compute Solution step.
- The Collect Data step is nowadays computer based, as well.

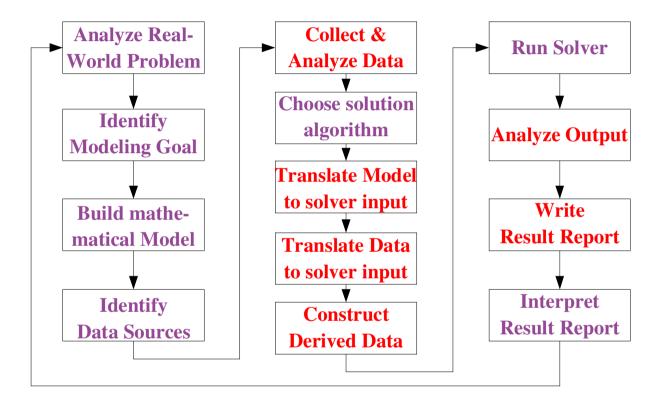


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# **Modeling – Detailed Scheme**











## **History of Modeling Languages**



- Mathematical Modeling is used in philosophy and scientific communities since antiquity.
- Useful algorithms (e.g. Runge-Kutta, FFT) made model analysis "tractable".
- The development of computers made bigger models accessible.
- Problems were input as programs.
- Data was read in from files (which were hand-made)
- Derived information (e.g. derivatives) were programmed.
- This had many disadvantages: error-prone, unstructured, not maintainable, no re-usable, not flexible, badly scaling







# History of Modeling Languages (ctd.)



- In 1947 George Dantzig developed the simplex algorithm for solving linear programming problems for the US air force.
- LP was used by non-scientific users to optimize various tasks.
- Modeling sometimes was more expensive than the profit.
- For LP mainly matrices were needed.
- The MPS input format was invented (for MPSX by IBM).
- For translating data to MPS or another input format, computers were used — matrix generators. (PDS, MAGEN, OMNI)
- Then MGG (matrix generator generator) was developed.







# History of Modeling Languages (ctd.)



- Even then the Modeling process was still expensive.
- Modeling was difficult and needed expert knowledge, modelers came from very different areas, and had very different background knowledge.
- Many self-devised modeling systems were (and still are) written.
- Optimization was most interesting for the industry.
- Late 1970s (papers in 1980) GAMS was the first algebraic modeling language.
- Late 1980s SIF as input language to Lancelot (paper 1990),
- Around 1985 AMPL was developed (book 1992),
- Xpress-MP Dash Optimization, 1987







# History of Modeling Languages (ctd.)



- Around the 1990s the number of modeling systems increased.
- LINGO LINDO Systems; Chicago; 1988
- MPL Maximal Software; 1989
- AIMMS Paragon; 1993
- ECLiPSe IC-Parc; London; 1995
- LPL Hürlimann; Zürich; 1995
- NOP, NOP-2 Neumaier, Dallwig, Schichl; Vienna; 1997-1999
- Numerica Van Hentenryck, ILOG; Paris; 1997
- OPL ILOG; Paris; 1998
- MINOPT Schweiger, Floudas; Princeton; 1999
- Mosel Dash Optimization; 2002







## **Modeling Languages – why?**



- Make modeling convenient and reduce the probability for errors,
- Reduce modeling costs,
- Increase the flexibility of the models,
- Independence from solvers,
  - Automatically translate models to internal solver formats.
  - Different solvers can be tried on hard problems
- Independence from data sources.
  - Automatically retrieve data and convert them to solver formats.
- Automatize the error-prone generation of derived information like derivatives
  - Automatic differentiation



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## **Algebraic Modeling Languages**



- Declarative Languages neither imperative, functional nor logic programming
- Models are expressed in an index-based formulation.
- Typical: Sets, Indices, Parameters, Variables
- Conceptual similar entities are grouped in Sets.
- These entities are later referenced by indices to the elements of these sets.
- Groups of entities (variables, constraints,...) can be represented in a compact by one algebraic expression.
  - $\sum_{i \in X} x_i$  is represented in AMPL by the expression Sum {i in X} x[i];



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# Algebraic Modeling Languages (ctd.)



- The algebraic modeling language is responsible for creating a problem instance to work on (set-indexing ability).
- Capability of dealing with non-linear models.
- The notation is close to the mathematical formulation.
- Problem formulation can be data-independent.
- Model is scalable, so switch from toy-problem to full problem can be done by a simple change of the parameters.
- Most statements are declarative, very few procedural statements, except for conditionals and loops.
- Many provide a database interface for data retrieval.







# Algebraic Modeling Languages (ctd.)



min 
$$x^T Q x + c^T x$$
 mathematical model s.t.  $Ax \le b$  
$$||x||_2 \ge 1$$
 AMPL if  $x \in [l, u]$ 

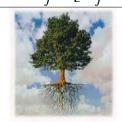
"flat" model

$$\min \sum_{i=1}^{N} \sum_{j=1}^{N} Q_{ij} x_{i} x_{j} + \sum_{k=1}^{N} c_{k} x_{k}$$
s.t. 
$$\sum_{j=1}^{N} A_{ij} x_{j} \le b_{i} \quad \forall i = 1, ..., M$$

$$\sum_{j=1}^{N} x_{j}^{2} \ge 1$$

$$x_{j} \in [l_{j}, u_{j}] \quad \forall j = 1, ..., N$$

### PARAMETERS ### param N>0 integer; param M>0 integer: param c {1..N}; param b {1..M}; param 0 {1..N,1..N}; AMPL model param A {1..M,1..N}; param 1 {1..N}; param u {1..N}; ### VARIABLES ### var x  $\{1..N\}$ ; ### OBJECTIVE ### minimize goal function:  $sum\{i in 1..N\} (sum\{j in 1..N\} Q[i,j]*x[i]*x[j]) +$  $sum{i in 1..N} c[i]*x[i];$ ### CONSTRAINTS ### subject to linear constraints {i in 1..M}: sum { j in 1..N} A[i, j] \*x[j] <=b[i]; box constraints { j in 1..N}: l[j] <= x[j] <= u[j];</pre> con add: sum { j in 1..N}  $x[j]^2 >= 1;$ data HOP2.2.dat: solve; display X;



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## Non-algebraic modeling languages



#### Object oriented modeling languages

- structured modeling (primitive entity, compound entity, attributes, tests, functions)
- gPROMS, ASCEND for chemical engineering
- EXTEND for manufacturing processes

#### Languages for constraint logic programming

- ECLiPSe
- provide a model description with very specialized constraints (all\_diff, count, ...)
- hints on the solution process, preferred branching and backtracking points can be specified



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Non-algebraic modeling languages (ctd.)

```
% ECLiPSe SAMPLE CODE
% AUTHOR:
                 Joachim Schimpf, IC-Parc
% The famous N-queens problem
:- set_flag(gc_interval, 10000000).
:- lib(lists).
:- lib(fd).
:- lib(fd search).
% The model
queens (N, Board) :-
 length (Board, N),
  Board :: 1..N,
   (fromto(Board, [Q1|Cols], Cols, []) do
    (foreach (Q2, Cols), param (Q1),
               count (Dist, 1, ) do
      noattack(Q1,Q2,Dist)
noattack (Q1,Q2,Dist) :-
        02 \# = 01,
        02 - 01 \# = Dist.
        01 - 02 \# = Dist.
```

```
% The search strategies
§_____
labeling(a, AllVars) :-
 (foreach(Var, AllVars) do
   count backtracks.
   indomain(Var)
                      % select value
) .
labeling(b, AllVars) :- % also for c,d,e
 .... % stuff deleted here ! %
first_queens(Strategy, N) :- % Find one solution
 statistics(times, [T0|]),
 queens (N, Board),
 statistics(times, [T1| ]),
 D1 is T1-T0,
 printf("Setup for %d queens done")
 init backtracks,
 labeling (Strategy, Board),
 get backtracks(B),
 statistics(times, [T2|]),
 D2 is T2-T1,
 printf("Found first solution for %d queens in %w\
          s with %d backtracks%n", [N,D2,B]).
% further stuff deleted down there...
```



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## **Other Modeling Tools**



#### Integrated Modeling Environments

- Include visualization techniques.
- Graphical User Interface (GUI) for model management and result presentation
- OPLStudio, AIMMS

#### Model Analysis Tools

- analyze mathematical problems wrt.
   constraint effectiveness checking, objective effect analysis, constraint grouping/sorting, shape analysis,...
- ANALYZE (linear), MPROBE (non-linear)

#### Spreadsheet modeling

 FRONTLINE Systems provide optimization on EXCEL spreadsheets.







# **Advantages of Modeling Languages**



- Scalability of the model
- Clear cut between modeling and numerics
- Clear cut between model structure and data
- Independence of solvers and solution algorithms
- Derived information can be generated automatically
  - notably automatic differentiation tools are widely used to provide derivatives and sparse Hessians
- The error probability in the translation of model and data to the solver's internal representation is almost zero.
- Convenience for the user.







# **Important Features of Modeling Languages**



- Close to mathematical Notation
- Little overhead in the formulation
- Convenience for the modeler
  - Names, arbitrary index sets,...
- Flexibility in the type of models
- Simple interface for solver connection
  - Modeling language to solver (model, data, derived information)
  - Solver to modeling language (progress report, solution, status info.)
- Simple and powerful data handling
  - Additional files
  - Database connectivity
- Automatic differentiation



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## Global Optimization/ Validated Computing



- Various levels of model rigor can be distinguished
   Validation ≤ Verification/Falsification ≤ Mathematical Proof
- Rigorous Computing tries to provide computer assisted mathematical proofs.
- Deterministic global optimization tries to solve

$$min f(x)$$
s.t.  $F(x) \in \mathbf{F}$ 

$$x \in \mathbf{X}$$

in a rigorous way.







## **Global Optimization (ctd.)**



- The variables typically fall into one of the following categories:
  - Continuous variables (real or complex valued), usually bounds are provided for them
  - Integer variables (also real or complex) usually bounds are specified in addition
  - Binary variables (either 0 or 1)
  - Discrete variables (can take a finite number of distinct isolated values)
    usually these values can have arbitrary symbols
  - Semi-continuous variables
     (0 or at least some threshold)
  - Partial integer variables (either  $\theta,...,N$  or continuous bound constrained in [N,M])



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## **Global Optimization (ctd.)**



- The constraints F(x) also come in different flavors
  - Continuous constraints (linear, convex, non-convex)
  - Discrete constraints
  - Logical constraints involving binary variables and logical operators
  - Global constraints
     constraints involving a lot of variables (e.g. all\_different, cardinality)
  - Special ordered sets
    - SOS-1: At most one of a set of variables is non-zero, all the others are zero
    - SOS-2: At most two 'adjacent' variables are non-zero, all the others are zero.
  - Exclusion Regions

$$x_L \notin X_L$$







# **Global Optimization (ctd.)**



- The data sometimes needs to be known exactly. Roundoff in the translation from modeling system to solver can destroy important properties or change the solution set, e.g.  $0.4 x + 0.6 y \ge 1$  becomes infeasible in IEEE floating point  $x, y \in [0,1]$  arithmetic after rounding the data.
- Sometimes the data is uncertain.
- Some constraints may be less relevant than others, or they can be slightly violated in some circumstances (soft constraints).
- Structural information about the problem should not be lost in the translation from modeling system to solver, and should be specifiable in the first place.







## **Typical Problems**



- Most modeling languages have originally been designed for non-linear local optimization.
- They pass presolved problems to the solver. Most of the problem's mathematical structure is lost (except for sparsity patterns).
- The data and the coefficients are passed as floating-point numbers (this involves roundoff and makes mathematical rigor impossible).
- Uncertain data cannot be specified.
- There is only restricted support for soft/hard constraints.







## **Typical Problems (ctd.)**



- There is no inherent support for interval arithmetic, so important derived information (such as implied bounds) cannot be generated automatically (except in Numerica).
- Usually there is no support for matrix operations.  $A^T D A x = b$  usually can be used in a model, but with difficulty, destroying the matrix structure. It is close to impossible to specify the constraint  $det(A) \ge c$ .
- Finally, many difficult global optimization problems can only be solved by using different but mathematically equivalent formulations of the same model.
   No modeling language (I know of) can be used to specify that.







# Other classes of Optimization Problems



- Multi-level Optimization
- Multi-objective Optimization
- Optimization with many objective functions
- Safety Programming
- Resource Optimization
- Stochastic Optimization
- Optimization involving differential equations
- ... and (hierarchical) combinations of all above mentioned classes







#### **Constraint Satisfaction Problems**



- No objective function, only sets of constraints
- Various goals are possible
  - find one solution (e.g. map-coloring problem)
  - find all solutions (in case of isolated solution points)
  - find outer and inner coverings of the solution set to provide a visualization
  - find N sufficiently distinct solutions
  - prove that the constraints are inconsistent
- Perform all the above approximately or with full rigor.
- It should be possible to formulate these goals in the modeling language.



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# **Multi-Objective Optimization**



- More than one objective is present
- Search for Pareto-optimal solutions

- find one
- find all
- cover the extremal set
- Multiple objectives should be allowed in a modeling language.







# Optimization with multiple objective functions



- More than one objective function is specified.
- For every objective function the optimization problem is solved, the results are recorded independently
- This should not be handled by simply passing each problem instance independent of the other to the solver.
- Support for warm starts and parallel execution.

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## **Multi-Level Optimization**



- Hierarchically structured optimization problems
- The simplest ones are minimax problems
- A two stage problem: The inner (stage 1) problem consists of g, G, y and has free variables x which appear in stage 2 together with f, F.

$$min f(x)$$

$$s.t. F(x) \in \mathbf{F}$$

$$max g(x, y) \leq c$$

$$s.t. G(x, y) \in \mathbf{G}$$

$$y \in \mathbf{Y}$$

$$x \in \mathbf{X}$$

 Recursive problem formulations would be useful (i.e. a model with parameters should be usable as a function)







# **Safety Programming**



- Verification of safety constraints for uncertain designs
- This is e.g. needed for mechanical engineering problems: Show that (or find a counterexample)

$$C(x,u) \le 0$$

$$\forall x, u : F(x,u) = 0$$

$$||x - x_0|| \le \Delta x$$

A typical problem of this type is e.g. FEM structural analysis, where F(x, u) = 0 takes the special form

$$A(x)u = b$$

and the x are bound constrained.







## **Resource Optimization**



- Optimization problems from resource optimization tend to be very large and very structured.
- All constraints can be generated from building blocks
- All data can be retrieved from databases
- There is a hierarchical structure of problems of increasing complexity
- Constraints for different resources are only similar.

\
$\boldsymbol{B}(z_1,\ldots,z_q) = \sum_{i=1}^{n} \boldsymbol{b}_i(s_i,z_1,\ldots,z_q)$
$p \in \mathbb{R}(\mathbb{Z}),  s_i \in \mathbb{R}(\mathbb{Z}) \ \forall i \in I$
$z_j \in \mathbb{R}(\mathbb{Z}) \ \forall j = 1, \dots, q$
$y_n \in \{0, 1\} \ \forall n \in E$
$s_i = f_i(p)  \forall i \in I_0 \subset I$
$S_i = S_i(p, z_1, \dots, z_q)  \forall i \in I_p$
$s_i \le 0  \forall i \in I_{\text{in}} \subset I \setminus (I_0 \cup I_p)$
$s_i \ge -M_i y_{n_i}  \forall i \in I_{\text{in}} \cap I_d$
$s_i \ge 0  \forall i \in I_{\text{out}} \subset I \setminus (I_0 \cup I_p)$
$s_i \leq M_i y_{n_i} \forall i \in I_{\text{out}} \cap I_d$
$g_{j}(p, s_{j_{1}},, s_{j_{n}}) \le 0  \forall j \in J_{u}, j_{1},, j_{n} \in I$
$h_k(p, s_{k_1},, s_{k_m}) = 0  \forall k \in J_g, k_1,, k_m \in I$
$G_j(z_1,\ldots,z_q) \le 0  \forall j \in J_u^z$
$H_k(z_1,\ldots,z_q) = 0  \forall k \in J_g^z$
$m_r$
$\sum y_{n_i} \leq e_r  \forall r \in R$
i = 1







# Optimization involving differential equations



- In design problems (e.g. aircraft design) often constraints are defined using the solution of some ordinary or partial differential equation.
- Sometimes differential algebraic systems and integral equations have to be used in the modeling process (e.g. parameter optimization for ovens used in silicon wafer production)
- In very few modeling systems (e.g. PCOMP, MINOPT) exists the possibility to specify differential equations.







## **Stochastic Optimization**



- Many real-world problems (e.g. portfolio optimization) depend on stochastic processes (e.g. the stock market)
- In addition to the before-mentioned variable classes random variables enter the mathematical problem formulation.
- Stochastic variables can be continuous or discrete.
- Usually distribution functions and/or covariance matrices are available.
- Often stochastic processes are included in the model. This leads to multistage problems.
- New classes of constraints, involving expectation values, variances, covariances, or probabilities appear.







## **Stochastic Optimization (ctd.)**



A typical stochastic optimization problem has the form

min 
$$f(x^{(k)}, \xi^{(k)})$$
  
s.t.  $F^{(k)}(\xi^{(k)}, x^{(< k)}, \xi^{(< k)}) \in \mathbf{F}$   
 $x^{(k)} \in [x_0^{(k)}]$   
 $\xi^{(k)} \sim g^{(k)}(b, x)$ 

where the k-th stage variables  $x^{(k)}$  and  $\xi^{(k)}$  depend on the lower stages in this multi-stage problem. The constraint  $F^{(k)}$  usually involves expectations or probabilities.

Today's algorithms use scenarios to "solve" this problem, they are, however, not part of the model.







## **Beyond Stochastic Optimization**



- Stochastic optimization problems are still not the most general ones.
- The modeler wants to record his knowledge about the realworld in the modeling language to construct the model.
- Typical knowledge is e.g.:
  - the opinion of expert(s)
  - within certain bounds
  - the mean value (and variance, higher moments)
  - a distribution function
  - a set of discrete values (with or without probabilities)
  - an exact value, an approximation, an estimate, a guess
- Then all objects must be set in relation, and these relations can be hard or soft or stochastic,...
- It should be possible to build all mathematically equivalent useful formulations.







#### **Data Handling**



- In addition to the declarative part, the model instance must be specified by adding data.
- Knowledge about the data should be recordable
  - Source of the data and how much it suits the model
  - Accuracy of the data
  - Is the data up-to-date?
  - Validity of the data
- Retrieval and structuring the data should be convenient
  - Database (ODBC) interface
  - Data/Model separation should be possible







# **Beyond Stochastic Optimization (ctd.)**



- In addition to the model instances, sometimes solution hints must be used to guide the solvers.
- This is neither part of the model nor part of the model instance, and includes e.g.
  - search and solution strategies
  - scenarios for stochastic problems
  - starting points for local optimization
  - bisection and backtracking hints
- Different solvers will need different hints.
- Construct a solver's view of the model instance.



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#### **Visions for the Future**



- Matrix support
- Allow several equivalent formulations at the same time
- GUI front-ends
- Object-oriented
- Recursive
- Building blocks
- Possibility to specify all knowledge







#### **GUI front-ends and back-ends**



- Like in the Mathematica 4 GUI front-end provide an even more mathematical notation.
- Provide model structuring
  - coarse grain / fine grain view
- Provide syntax-based editing capabilities with on line-help
- Visualization of the connection between model-parts
- Cross reference: variables constraints

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- Progress visualisation
- Result visualisation
- Is already done in some systems (e.g. AIMMS, Mosel, OPLstudio, MPL)



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## **Object Oriented Modeling**



- Object oriented programming is an extension (or the consequence) of imperative programming
- Modeling needs the consequence of declarative programming — derived models.
  - E.g. a screw is represented by a set of variables and constraints.
  - Steel is a material, also specified by variables and constraints.
  - Wood is another material.
  - Now it should be straight-forward to derive steel screws and wooden screws by inheritance
- Mimics the mathematical way of defining new objects.
  - E.g. a group is a monoid where inverses of all elements exist.
  - Associativity and unit element are "inherited" from the monoid.



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#### **Recursive Modeling**



- Multistage problems often arise naturally.
- E.g. first we want to prove that a designed robot is free of structural deficiencies.
- We build a model for that (a safety problem) and solve it.
- For the next design we want the cheapest robot which is free of structural deficiencies.
- This is a two-stage optimization problem (a design problem) where the first stage contains the safety problem which has been designed already. However, now the design parameters are no longer fixed but free variables).
- Models should be reusable as "functions" in bigger models.







## **Building Blocks**



- Many huge optimization problems are built from a big number of (almost) identical blocks
  - chemical engineering
  - resource optimization
  - manufacturing processes
  - optimal scheduling
- Supporting building blocks in modeling languages and enhancing it by adding GUI support will decrease the number of modeling errors for big problems and greatly reduces the modeling costs. (e.g. ProComm BoFit)



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## Specifying all knowledge



- It should be possible for the modeler to record all his knowledge about model and real-world problem in the language.
- Knowledge about objects
- Knowledge about data
- Knowledge about solvers and solution strategies
- The solver need not take all information into account if it cannot make use of certain information.







#### Open Model Exchange Format/ Model Databases



- An open format for communicating models, storing them in databases and exchanging them between modeling languages/systems would be very useful.
  - In the COCONUT project we had to write >10 converters between various modeling languages for testing the same set of benchmark problems on a number of different solvers.
- Could be XML like and would provide a convenient Webinterface for model presentation at the same time,
- Could have ASCII and binary interfaces and encryption,
- Needs a consistent interface for floating point numbers,
- Should be extendable,
- Open, public and widely supported,







#### Literature



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#### The End



# Thank you for your attention



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