

Logic-Based Optimization Methods for Engineering Design

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Idea of logic-based methods

- An alternative to integer and mixed integer/linear programming.
- Represent discrete choices with logical propositions rather than integer variables.
- Solve with branch-and-bound methods as in integer programming, but with a difference:
 - use logical inference and domain reduction methods from constraint programming as well as linear programming.
 - use new relaxations for logical constraints rather than the traditional linear programming relaxation

Advantages of logic-based approach to design

- Design typically involves a combination of discrete choices and continuous parameters.
- Logic framework provides more natural modeling of discrete choices.
- Solution approach harnesses power of logical inference and removes unnecessary integer variables from the linear relaxations.
- By distinguishing special cases logically, singularities can be avoided.

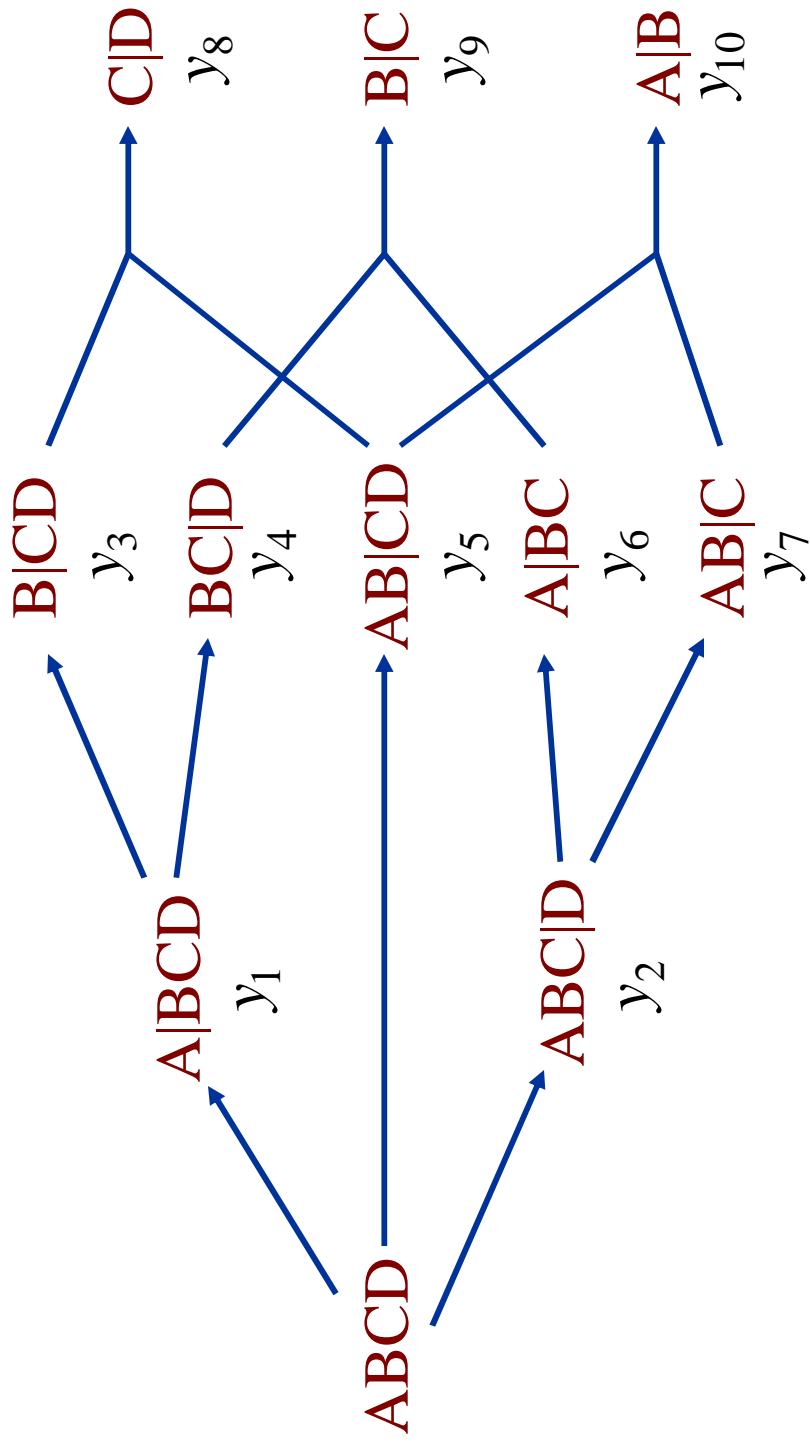
Outline

- Logic-based optimization for chemical processing network design (process synthesis).
- Logic-based optimization for truss structure design.
 - Computational results
- Integration of optimization and constraint satisfaction.

Processing network design

- Design a network of processing units, such as reactors or distillation units
- Meet demand while minimizing fixed and variable costs.
- Discrete choice is which units to install.
- Problem presented here is linear, but heat exchange and other processes give rise to nonlinear models.

4-component separation network



Model for separation problem

Unit flow cost

$$\min \sum_{ij} c_{ij} x_{ij} + \sum_i z_i$$

Fixed cost

Flow volume

$$\text{s.t. } \sum_i x_{ij} = \sum_k x_{jk}, \quad \text{all } j \quad \text{Flow balance}$$

$$x_{ij} = \alpha_{ij} \sum_k x_{ki}, \quad \text{all } i, j \quad \text{Unit output}$$

$$0 \leq x_{ij} \leq k_{ij}, \quad \text{all } i, j \quad \text{Capacity}$$

Unit is installed $\rightarrow y_i \rightarrow (z_i = f_i), \quad \text{all } i$


Unit is not installed $\rightarrow \neg y_i \rightarrow \left(\begin{array}{l} z_i = 0 \\ \sum x_{ij} = 0 \end{array} \right), \quad \text{all } i$

Integer programming model

Replace discrete constraints with

$$0 \leq \sum_j x_{ij} \leq k_i y_i, \quad \text{all } i$$

0-1 variable



A continuous (linear programming) relaxation is important for solving the problem. It can be obtained by replacing

$$y_i \in \{0,1\} \quad \text{with} \quad 0 \leq y_i \leq 1$$

Relaxation for logic constraints

$$y_i \rightarrow (z_i = f_i)$$
$$\neg y_i \rightarrow \left(\begin{array}{l} z_i = 0 \\ \sum x_{ij} = 0 \end{array} \right)$$

The logic constraints

$$\frac{z_i}{f_i} \geq \frac{1}{M_i} \sum_j x_{ij}$$

Can be relaxed:

Where M_i = upper bound on output of unit i

So integer variables add needless overhead to solution of linear programming relaxation at each node.

Additional logic constraints

Constraints such as

$$y_3 \rightarrow y_1$$

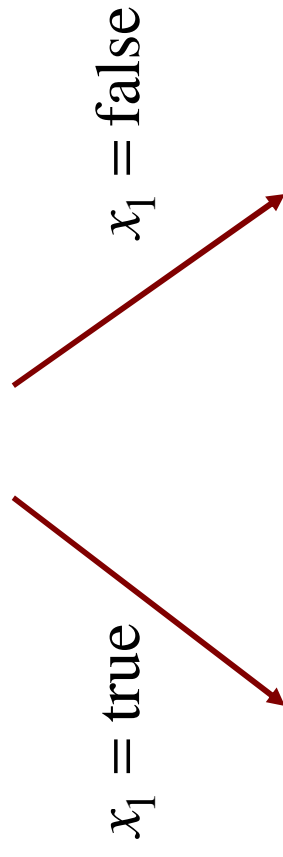
$$y_4 \rightarrow y_1$$

$$y_5 \rightarrow y_8$$

$$y_5 \rightarrow y_{10}$$

can speed processing at each node by ruling out solutions that cannot be optimal.

Logic-based branch & bound



Solve as an LP: linear part of problem, plus relaxation of conditional constraints, plus:

$$z_1 = f_1$$

$x_2 = \text{true}$

$x_2 = \text{false}$

Solve same problem except replace $z_1 = f_1$ with:

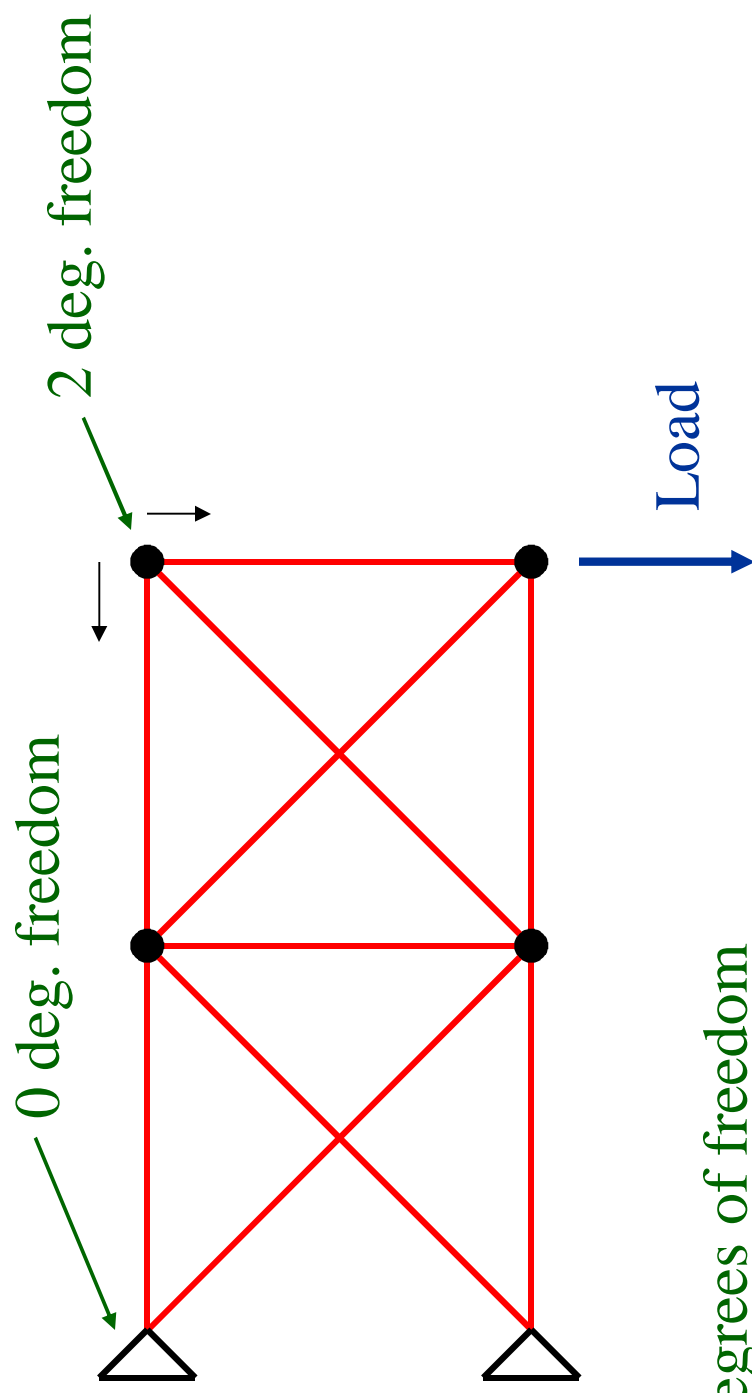
$$\left(\begin{array}{l} z_1 = 0 \\ \sum x_{1j} = 0 \end{array} \right)$$

Because $x_3 \rightarrow x_1$ and $x_4 \rightarrow x_1$, fix $x_3 = x_4 = \text{false}$ and add constraints similar to above

Truss structure design

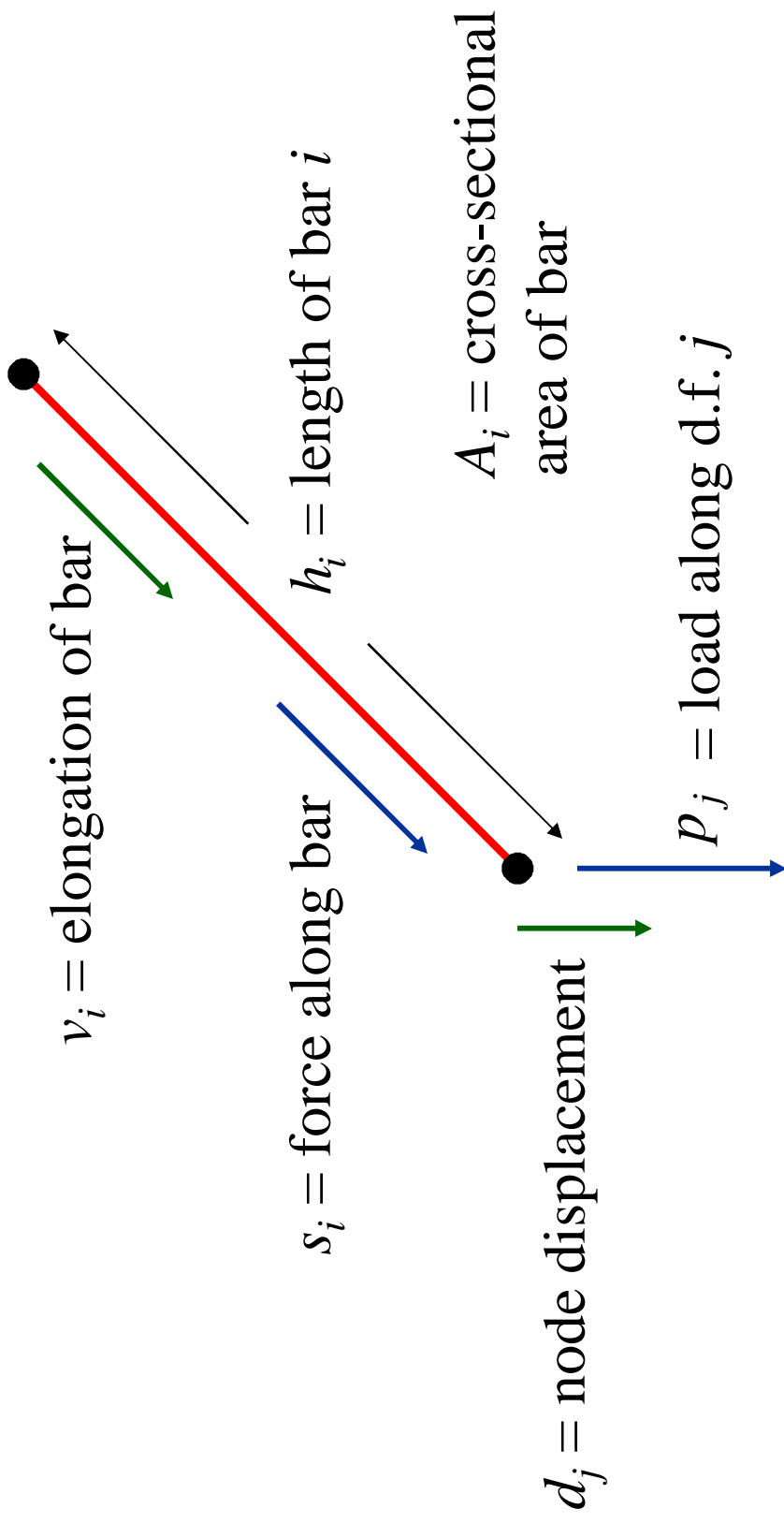
- Find truss structure that supports a given load while minimizing total weight of bars.
- Any given bar may be present or absent from the truss.
- Its cross-sectional area must be one of several discrete values.
- The model is nonlinear. Logic-based modeling avoids the singularities that occur in traditional models when the bar size goes to zero.

Planar 10-bar cantilever truss



Total 8 degrees of freedom

Notation



$$\begin{aligned}
& \min \sum_i h_i A_i \quad \} \text{Minimize total weight} \\
& \text{s.t.} \quad \sum_i \cos \theta_{ij} s_i = p_j, \text{ all } j \quad \} \text{Equilibrium} \\
& \quad \sum_i \cos \theta_{ij} d_j = v_i, \text{ all } i \quad \} \text{Compatibility} \\
& \quad \frac{E_i}{h_i} A_i v_i = s_i, \text{ all } i \quad \} \text{Hooke's law} \\
& \quad v_i^L \leq v_i \leq v_i^U, \text{ all } i \quad \} \text{Elongation bounds} \\
& \quad d_j^L \leq d_j \leq d_j^U, \text{ all } j \quad \} \text{Displacement bounds} \\
& \quad \bigvee_k (A_i = A_{ik}) \quad \} \text{Logical disjunction}
\end{aligned}$$

nonlinear \longrightarrow

Area must be one of several discrete values A_{ik}

Constraints can be imposed for multiple loading conditions

Logic-based branch & bound

- Rather than branch on variables, branch by splitting the range of areas A_i
- At each node, add the constraints

$$A_i^L \leq A_i \leq A_i^U$$

where the bounds are equal to one of the discrete areas bar i can have

Mixed integer model introduces many additional variables

$$\begin{aligned}
 \min \quad & \sum_i h_i \sum_k A_{ik} y_{ik} && \text{0-1 variables indicating size of bar } i \\
 \text{s.t.} \quad & \sum_i \cos \theta_{ij} s_i = p_j, \text{ all } j \\
 & \sum_j \cos \theta_{ij} d_j = \sum_k v_{ik}, \text{ all } i && \text{Elongation variable disaggregated by bar size} \\
 & \frac{E_i}{h_i} \sum_k A_{ik} v_{ik} = s_i, \text{ all } i && \text{Hooke's law becomes linear} \\
 & v_i^L \leq v_i \leq v_i^U, \text{ all } i \\
 & d_j^L \leq d_j \leq d_j^U, \text{ all } j \\
 & \sum_k y_{ik} = 1, \text{ all } i
 \end{aligned}$$

A useful relaxation can be obtained without the extra variables...

Linear relaxation

Use the change of variables:

$$A_i = A_i^L y_i + A_i^U (1 - y_i)$$

$$v_i = v_{i0} + v_{i1}$$

Current bounds on areas



The y_i 's are not 0-1 variables but are continuous variables that vary in the interval $[0,1]$, introduced only to form a relaxation.

The resulting “relaxation” is not a true relaxation but provides a valid bound on the optimal value...

$$\begin{aligned}
& \min \sum_i h_i [A_i^L y_i + A_i^U (1 - y_i)] \\
& \text{s.t.} \quad \sum_i \cos \theta_{ij} s_i = p_j, \text{ all } j \\
& \quad \sum_i \cos \theta_{ij} d_j = v_{i0} + v_{i1}, \text{ all } i \\
& \quad \frac{E_i}{h_i} (A_i^L v_{i0} + A_i^U v_{i1}) = s_i, \text{ all } i \\
& \quad v_i^L y_i \leq v_{i0} \leq v_i^U y_i, \text{ all } i \\
& \quad v_i^L (1 - y_i) \leq v_{i1} \leq v_i^U (1 - y_i), \text{ all } i \\
& \quad d_j^L \leq d_j \leq d_j^U, \text{ all } j \\
& \quad 0 \leq y_i \leq 1, \text{ all } i
\end{aligned}$$

Hooke's law is linearized

Elongation bounds split into 2 sets of bounds



Parallel with branch & bound

- In branch & bound, solution of the linear relaxation is often integral, which reduces branching.
- In logic-based branching, solution of “relaxation” often puts areas at endpoints of their ranges, in which case they have one of the permissible discrete values.

Computational testing

- Logic-based: Branch on upper half of interval first.
- MILP: Solve with CPLEX 4.0 with automatic SOS detection turned on.
- Problems:
 - 10-bar truss
 - 25-bar transmission tower
 - rectangular building (72, 90, 108 bars)
- Bars are “linked” when they must have the same size.

Computational results

| <i>Problem</i> | <i>Linking groups</i> | <i>Logic-based seconds</i> | <i>CPLEX MILP seconds</i> |
|------------------|-----------------------|----------------------------|---------------------------|
| 10-bar truss | 10 | 0.3 | 1.3 |
| 10-bar truss | 10 | 0.3 | 1.6 |
| 10-bar truss | 10 | 1.2 | 2.6 |
| 10-bar truss | 10 | 1.4 | 2.6 |
| 10-bar truss + L | 10 | 5.8 | 23.6 |
| 10-bar truss + D | 10 | 68 | 1089 |
| 10-bar truss + D | 10 | 1654 | 13744 |

L = 2 loading conditions

D = displacement bounds

Computational results

| <i>Problem</i> | <i>Linking groups</i> | <i>Logic-based seconds</i> | <i>CPLEX MILP seconds</i> |
|----------------------|-----------------------|----------------------------|---------------------------|
| 25-bar tower + L | 8 | 226 | 272 |
| 72-bar building + L | 16 | 208 | 12693 |
| 90-bar building + L | 20 | 169 | >72000 |
| 108-bar building + L | 24 | 329 | >72000 |
| 200-bar structure | 96 | >36000 | >72000 |

L = 2 loading conditions

An alternative model

Remove nonlinearities by writing Hooke's law directly as a disjunction

$$\bigvee_k \left(\frac{E_i}{h_i} A_{ik} v_i = s_i \right)$$

that can be relaxed with a single inequality for each i .

The disjunctive constraint $\bigvee_k (A_i = A_{ik})$

can now be dropped.

Possible enhancements

- Limit total number of bar sizes in structure.
- Enforce symmetries in topology even when bar sizes need not match.
- Enforce connectedness and other properties logically.
- Find logical characterization of stability.

Constraint programming

- Constraint programming is based primarily on logic-based methods known as *constraint satisfaction* methods.
- Discrete variables tend to be multivalued (here, they were 2-valued).
- *Global* constraints exploit problem structure and speed solution; e.g., **all-different** (Y_1, \dots, Y_n), **cumulative**.
- Inference takes the form of *domain reduction* algorithms, particularly when applied to global constraints.

Optimization + constraint programming

- Combining optimization and constraint programming is a very active area of research; e.g. OPL modeling language.
- The logic-based method described here is an example of one approach.