Mixed Integer Programming The State of the Art

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A Definition

A *mixed-integer program* (MIP) is an optimization problem of the form

 $\begin{array}{ll} \text{Minimize} & c^T x \\ \text{Subject to} & Ax = b \\ & l \leq x \leq u \\ \text{some or all } x_j \text{ integer} \end{array}$



Unit-Commitment Models

Electrical Power Industry, ERPI GS-6401, June 1989: Mixed-integer programming (MIP) is a powerful modeling tool, "They are, however, theoretically complicated and computationally cumbersome"

In Other Words: MIP is an interesting modeling "toy", but it just doesn't work in practice.

This perception began to change in 1999.



From the Rutger's DIMACS Meeting 1999: California 7-Day Model

UNITCAL_7: 48939 constraints, 25755 variables (2856 binary)

Reported Results 1999 – machine unknown 2 Day model: 8 hours, no progress 7 Day model: 1 hour to solve initial LP

Desktop PC -- ran full 7-day model CPLEX 6.5 (1999): 22 minutes, optimal TODAY (2015): 15 seconds, optimal



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Computational History: 1950 -1998

- 1954 Dantzig, Fulkerson, S. Johnson: 42 city TSP
 - Solved to optimality using LP and cutting planes
- 1957 Gomory
 - Cutting plane algorithms
- 1960 Land, Doig; 1965 Dakin
 - B&B
- 1964-68 LP/90/94
 - First commercial application
- IBM 360 computer
 - 1974 MPSX/370
 - 1976 Sciconic
 - LP-based B&B
 - MIP became commercially viable

- 1975 1998 Good B&B remained the state-of-the-art in commercial codes, in spite of
 - Edmonds, polyhedral combinatorics
 - 1973 Padberg, cutting planes
 - 1973 Chvátal, revisited Gomory
 - 1974 Balas, disjunctive programming
 - 1983 Crowder, Johnson, Padberg: PIPX, pure 0/1 MIP
 - 1987 Van Roy and Wolsey: MPSARX, mixed 0/1 MIP
 - TSP, Grötschel, Padberg, ...



1998 ... A New Generation of MIP Codes

- Linear programming
 - Stable, robust dual simplex
- Variable/node selection
 - Influenced by traveling salesman problem
- Primal heuristics
 - 12 different tried at root
 - Retried based upon success
- Node presolve
 - Fast, incremental bound strengthening (very similar to Constraint Programming)

- Presolve numerous small ideas
 - Probing in constraints:
 - $\sum x_j \le (\sum u_j) y, y = 0/1$ $\Rightarrow x_j \le u_j y \text{ (for all j)}$
- Cutting planes
 - Gomory, mixed-integer rounding (MIR), knapsack covers, flow covers, cliques, GUB covers, implied bounds, zero-half cuts, path cuts



MIP Speedups



Some Test Results

- Test set: 1852 real-world MIPs
 - Full library
 - 2791 MIPs
 - Removed:
 - 559 "Easy" MIPs
 - 348 "Duplicates"
 - 22 "Hard" LPs (0.8%)
- Parameter settings
 - Pure defaults
 - 30000 second time limit
- Versions Run
 - CPLEX 1.2 (1991) -- CPLEX 11.0 (2007)



CPLEX Version Performance Improvements (1991–2008)



CPLEX Version-to-Version Pairs

Progress: 2009 – Present



Gurobi MIP Library

(3550 models)



MIP Speedup 2009–Present

Starting point

Gurobi 1.0 & CPLEX 11.0 ~equivalent on 4-core machine

Gurobi Version-to-version improvements

Gurobi 1.0 -> 2.0: 2.4X
Gurobi 2.0 -> 3.0: 2.2X (5.1X)
Gurobi 3.0 -> 4.0: 1.3X (6.6X)
Gurobi 4.0 -> 5.0: 2.0X (12.8X)
Gurobi 5.0 -> 6.0: 2.2X (27.6X)
Gurobi 6.0 -> (6.5): 1.7X (46.0X)

Machine-independent IMPROVEMENT since 1991

• Over 1.3 million X -- 1.8X/year



Suppose you were given the following choices:

- Option 1: Solve a MIP with today's solution technology on a machine from 1991
- Option 2: Solve a MIP with 1991 solution technology on a machine from today

Which option should you choose?

 Answer: Option 1 would be faster by a factor of approximately 300.



Thank you

