

Optimization and Analytics

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G u r o b i
Optimization

The Early History

- ▶ 1947 – George Dantzig invents simplex for LP
 - Gives a global view of solutions
- ▶ 1951 – First computer code for solving LPs
- ▶ 1960 – LP commercially viable
- ▶ 1970 – MIP commercially viable

The Decade of the 70' s

▶ Interest in optimization flowered

- Numerous new applications identified
- Many companies created OR Departments

▶ Significant difficulties emerged

- Building application was very expensive and very risky
 - 3-4 year development cycles
 - Technology just wasn't ready: LPs were hard and MIP was a disaster
- **Result:** *Disillusionment and much of that disillusionment persists to this day.*

The Decade of the 80's

- ▶ Several key developments
 - IBM PC introduced in 1981
 - Relational databases developed
 - ERP systems introduced
 - First algebraic modeling language: GAMS
 - Karmarkar's 1984 paper on interior-point methods

The Decade of the 90' s

- ▶ LP performance takes off
- ▶ Data became plentiful and accessible
 - ERP systems became commonplace
- ▶ Optimization shown to be feasible on selected, difficult, real problems
 - Business: Airlines, Supply-Chain
 - Academic: Traveling Salesman Problem

Mixed Integer Programming

A Definition

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

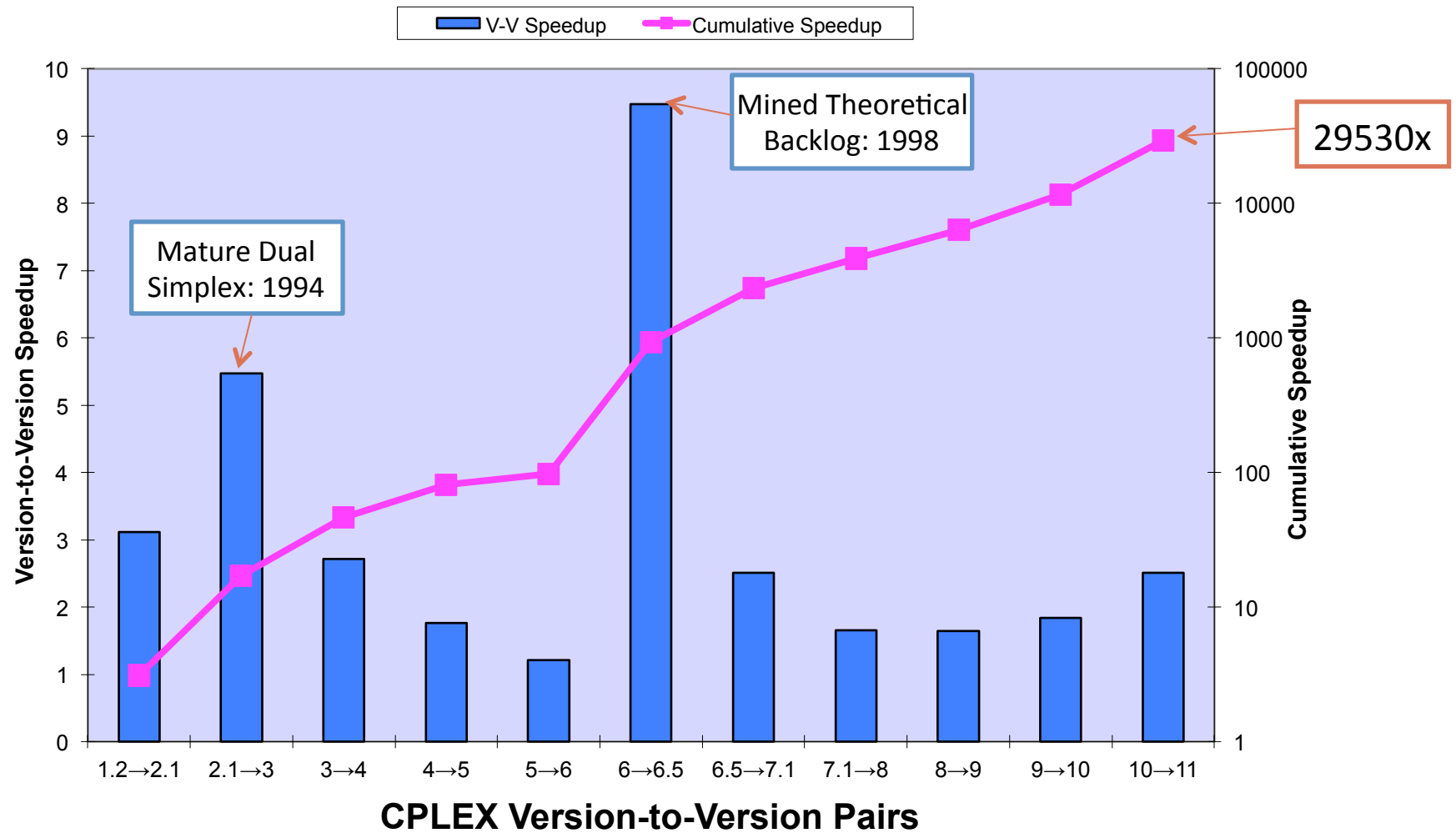
$$\begin{array}{ll}\textit{Minimize} & c^T x \\ \textit{Subject to} & Ax = b \\ & l \leq x \leq u \\ & \text{some or all } x_j \text{ integer}\end{array}$$

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:
$$\sum x_j \leq (\sum u_j) y, \quad y = 0/1$$
$$\rightarrow x_j \leq u_j y \text{ (for all } j)$$
- Cutting planes
 - Gomory, mixed-integer rounding (MIR), knapsack covers, flow covers, cliques, GUB covers, implied bounds, zero-half cuts, path cuts

MIP Performance Improvements

Speedups 1991-2008



Progress: 2009 – Present

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)
- ▶ Machine-independent IMPROVEMENT since 1991
 - Over 800,000X -- 1.8X/year

Big Data: Million-Sized Models

- ▶ **One view:** Optimization is a technique for analyzing/understanding large data sets and producing actionable conclusions.
- ▶ **Customer regularly send us models**
 - Over 300 customers have sent models
 - More than 10,000 models in total
- ▶ **Trend:** Models are getting larger and larger
 - One of our customers has recently “reported” a model with 2 billion nonzeros.

MIP “million sized” models

- ▶ Customer models with > 1 million variables or constraints
 - 2011: 47 models
 - 2012: 68 models
 - 2013: 156 models
 - 2014: 286 models
 - From energy, transportation, finance, mining, forestry, social media, ...
- ▶ Largest
 - Trading – stock market
 - 39M rows, 39M columns, 100M nonzeros
 - Solution time: 1040 seconds
- ▶ Testing (2014 – 156 models)
 - Ran defaults
 - 7200 second time limit

Summary – 156 Models in Total

▶ Solvability – defaults (tuned): 144

- Solved to optimality 81 (88)
- Feasible < 10% gap 27 (33)
- Feasible > 10% gap 12 (23)

▶ Others – no feasible solution: 12

- Could not solve root LP: 2

▶ Summary

- Defaults: 70% acceptable solutions
- Tuned: 78% acceptable solutions
- No feasible solution found: 8% found
- Mean solve time for solvable: 496 sec.



The Curse of the Free Solver

Mittelmann MIPLIB2010 tests (87 models):

	Factor Slower than Gurobi	% Solved
GLPK	27X	1%
LPSOLVE	24X	6%

The Curse: These solvers *very popular* (in particular in statistics) and lead to *completely wrong conclusions* about what is doable!

Statistics and Prescriptive Analytics

- ▶ **Dan Gusfield**
 - Statistical applications in computational biology
 - web.cs.ucdavis.edu/~gusfield
- ▶ **Dimitris Bertsimis**
 - Application of MIP to regression methods
 - “Best subset selection via a modern optimization Lens”, with A. King and R. Mazumder
- ▶ **Stan Uryasev – Finance**
 - Statistical applications in finance
 - www.ise.ufl.edu/uryasev/research/testproblems/advanced-statistics
- ▶ **José R. Zubizarreta**
 - MISMATCH: optimal matching in observational studies using MIP
 - www.columbia.edu/~jz2313

The Analytics Wave

The Analytics Taxonomy

(Thomas Davenport)

- ▶ *Descriptive Analytics*

- Often what people mean when they say “Big Data”
- Understanding, representing your data

- ▶ *Predictive Analytics*

- Forecasting
- Making predictions based on your data
- Statistics plays a central role

- ▶ *Prescriptive Analytics*

- Using your data to make decisions
- Optimization plays a central role
 - Mixed-Integer Programming (MIP) is THE primary optimization technique used in business.

Remarks I: The Future of Analytics and Optimization

- ▶ We should ride the wave
- ▶ Analytics is much bigger than optimization
 - Analytics has real chance of becoming a “standard” business function
 - Concretely: Companies of the future may well have *Chief Analytics Officers*
- ▶ Our objective should be: Profit from this movement
 - Make sure we are part of the analytics education process
 - View corporate analytics groups as providing a natural home for the OR / Management Science function

Remarks II: The Importance of Optimization

- ▶ Making the right DECISIONS based on your data is the first priority in managing a modern enterprise
 - Optimization is about making decisions!
- ▶ Many people say prescriptive analytics follow predictive analytics.
 - In my view this puts the cart before the horse: There is no better tool than building a model to force you to understand your data and determine what data you need

Thomas H. Davenport “Analytics 3.0”, HBR, December 2013

“Although Analytics 3.0 includes all three types [descriptive, predictive, prescriptive], it emphasizes the last. Prescriptive models involve large-scale testing and optimization and are a means of embedding analytics into key processes and employee behaviors.”