

Advanced linear programming

11.1, 11.2: Solving integer linear programming problems:

Totally Unimodular Matrices

Cutting planes

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This lecture: More on solving ILPs

- Totally Unimodular Matrices: LP-relaxation is integral ☺

If LP-relaxation is fractional (usually)

- Cutting planes
- Branch-and cut



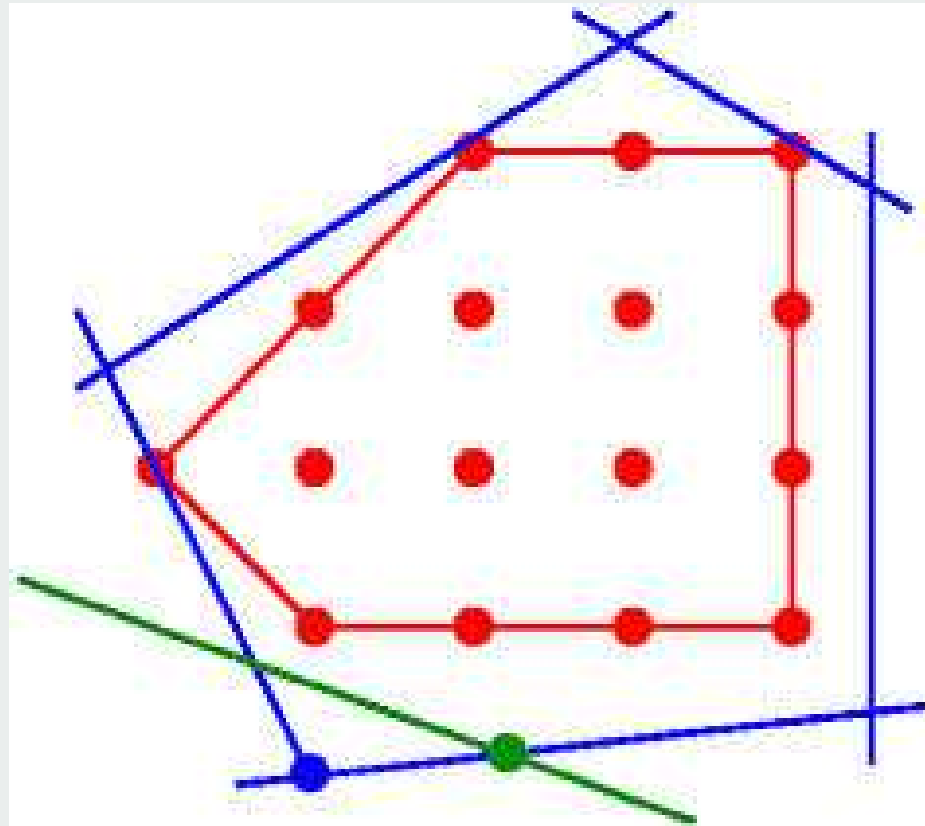
Cutting plane algorithm

1. Solve the LP-relaxation. Let x^* be an optimal solution.
2. If x^* is integral, stop x^* optimal solution to the integer linear programming problem.
3. If not, add a *valid inequality* that is not satisfied by x^* and go to Step 1. *Solve separation problem.*

Definition: A *valid inequality* is a linear constraint that is satisfied by all integral solutions.
If it cuts off fractional solutions it acts as *cutting plane*.



Cutting plane algorithm



Gomory cuts

- Take row from end tableau corresponding to fractional basic variable

$$x_i + \sum_{j \in N} \bar{a}_{ij} x_j = \bar{a}_{io} \text{ with } \bar{a}_{io} \text{ fractional}$$

- Gomory cut

$$x_i + \sum_{j \in N} \lfloor \bar{a}_{ij} \rfloor x_j \leq \lfloor \bar{a}_{io} \rfloor$$

- You can find the optimum by adding Gomory cuts, no practical method since you may need exponentially many steps



Problem-specific valid inequalities

- Usually classes of inequalities
- Cover inequalities for knapsack problems
- Weighted independent set
- Single-machine scheduling



“Cutting plane”



Branch-and-cut

- Combination of cutting planes and ILP solving by branch-and-bound

- Applied in well-known ILP-solvers:
 - CPLEX,
 - GUROBI,
 - EXPRES-MP
 - COIN-OR



Solving ILP by branch-and-bound or branch-and-cut

Let x^* be the best known feasible solution

Search strategy

1. Select an active sub problem F_i (unevaluated node)
2. If F_i is infeasible: delete node
3. Compute upper bound $Z_{LP}(F_i)$ by solving LP-relaxation and feasible solution x_f (by rounding)

If $Z_{LP}(F_i) \leq$ value x^* delete node (bounding)

x_f is better than x^* : update x^*

If solution x_{LP} to LP-relaxation is integral,

then If x_{LP} is better than x^* : update x^* and node finished, otherwise split node into two new subproblems (branching)

4. Go to step 1

Branching strategy

How many cuts? Which classes?

Primal heuristic

Optional

This if for maximization problem, the book uses a minimization problem.



Branch-and-cut is a framework algorithm!!

Last century, tailoring to your own problem was necessary in most cases and a huge amount of research has been undertaken in doing this:

- Pre-processing
- Classes of valid inequalities
- How many cuts in each node?
- Search strategy
- Branching strategy
- Primal heuristics
- Reduced cost fixing

Now solvers like CPLEX and Gurobi successfully (and secretly) apply a lot of the above techniques.

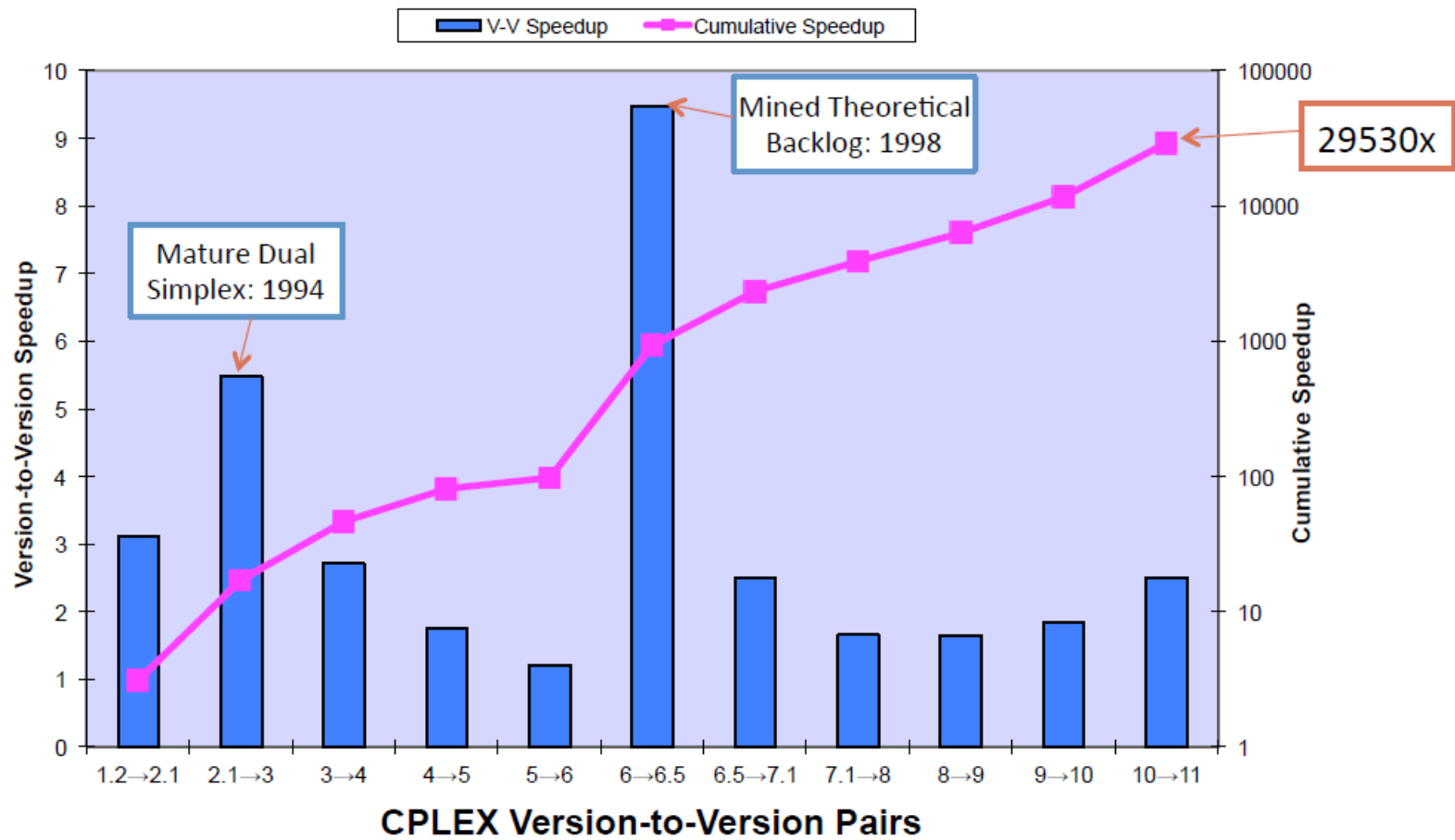
- Tailoring of branch-and-cut sometimes successful, especially valid inequalities in case of a weak LP-relaxations



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

Speedups 1991-2008



World record exact TSP solving

- Branch-and-cut



CONCORDE:

<http://www.math.uwaterloo.ca/tsp/concorde/index.html>



Still we are talking about combinatorial optimization

- We have many, many, many variables
- Still large computation times
- Decomposition approaches often help
- **We now go to Chapter 6: Large-scale optimization**
 - *Decompositions for large LP's*
 - *Since these principles are usually used for solving ILP's, Ch 10 was done first.*

