

# Advanced Operations Research Techniques

## IE316

### Lecture 24

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## Reading for This Lecture

- Bertsimas Sections 10.2, 10.3, 11.1, 11.2

## Preprocessing

- Often, it is possible to **simplify** a model using logical arguments.
- Most commercial LP solvers do preprocessing automatically, but if you are developing a model that will be solved repeatedly, it may be worthwhile.
- The constraint  $a^1x \leq b_1$  **dominates** the constraint  $a^2x \leq b_2$  if

$$a_i^1 \geq a_i^2 \quad \forall i, \text{ and}$$

$$b_1 \leq b_2$$

- In this case, the dominated inequality could be deleted.
- We can also derive **implied bounds** for variables from each constraint  $ax \leq b$ . If  $a_0 > 0$ , then

$$x_1 \leq (b - \sum_{j:a_j>0} a_j l_j - \sum_{j:a_j<0} a_j u_j) / a_0$$

## More Preprocessing

- The constraint  $ax \leq b$  is **redundant** if

$$\sum_{j:a_j>0} a_j u_j + \sum_{j:a_j<0} a_j l_j \leq b.$$

- The LP is **infeasible** if

$$\sum_{j:a_j>0} a_j l_j + \sum_{j:a_j<0} a_j u_j > b.$$

- For an LP of the form  $\min\{c^T x \mid Ax \geq b, l \leq x \leq u\}$ ,
  - If  $a_{ij} \geq 0 \forall i \in [1..m]$  and  $c_j < 0$ , then  $x_j = u_j$  in any optimal solution.
  - If  $a_{ij} \leq 0 \forall i \in [1..m]$  and  $c_j > 0$ , then  $x_j = l_j$  in any optimal solution.

## More Preprocessing

- More sophisticated rules can also be applied.
- It is possible to take advantage of **problem structure**.
- Effect of preprocessing (example from the book, p. 540):

	Rows	Columns	Iterations	Seconds
No preproc.	13,689	17,148	39, 429	10, 094
Preproc.	5,579	9,508	18,975	2,381

## Round-off Error and Scaling

- One of the big issues faced in the practice of linear programming is *round-off error*.
- Round-off error occurs simply because computers do not perform *exact arithmetic*.
- Computers perform *floating point* arithmetic, which means that computations get rounded off.
- These round-off errors can accumulate until they become significant.
- To avoid round-off error, it is important that the matrix be *scaled* properly.
- The existence of variables and constraints that are on vastly different scales can cause *numerical problems*.
- Most LP solvers will scale the problem automatically, but it's always better to start with a *well-scaled model*.

## Error Tolerances

- Because of round-off error, LP solvers have specified *error tolerances*.
- Some have different error tolerances for different operations.
- These error tolerances are especially important in integer programming.
- Often, a variable will have a value that is very close to an integer value.
- If it is within the error tolerance, then it is considered integer.
- However, the rounded solution may not be feasible.
- If you are having numerical problems, you may have to adjust the error tolerances or consider *scaling*.
- Another possibility is to adjust the *basis refactorization frequency*.

## Commercial LP Solvers

- Not all LP solvers are created equal.
- Some issues that set LP solvers apart from each other.
  - Preprocessing
  - Numerical stability
  - Handling of degeneracy
  - Methods of pricing
  - Methods for determining the pivot element
  - Methods other than simplex (barrier methods)
- Some commercial codes (see NEOS guide for full list)
  - CPLEX (ILOG)
  - OSL (IBM)
  - XPRESS-MP (Dash)
- There are also numerous free solvers available on the Web.



## Improvements to Simplex: Case Study

- This table is for a large linear program (approximately 50K rows, 175K columns, and 400K nonzeros), taken from the *ILOG Optimization Times*.
- These tests were all run on the same machine (296 MHz Sun Ultrasparc).
- This shows the dramatic difference implementation can make.

CPLEX Version	Year	Time (seconds)
1.0	1988	57,840
3.0	1994	4,555
5.0	1996	3,835
6.5	1999	165

## How to Choose an Algorithm

- The three basic choices for algorithm are
  - Primal simplex
  - Dual simplex
  - Barrier
- The choice must be made more or less empirically.
- As a rule of thumb, dual simplex seems to outperform primal simplex, especially with degenerate problems, which often occur in practice.
- The availability of an advanced basis could determine the method.
- For large, sparse problems, barrier methods can outperform simplex.
- However, barrier methods cannot be “warm started” with an advanced basis.
- Starting with an advanced basis can be a significant advantage.

## Parameter Settings

- LP solvers have numerous parameter settings for both linear programming and integer programming.
- Examples of **LP parameters**
  - Algorithm type
  - Pricing method
  - Basis refactorization frequency
- Examples of **IP Parameters**
  - Search order
  - Method of branching
  - Methods for tightening relaxations
- For a summary of interesting parameters in **CPLEX**, see the **CPLEX/AMPL** guide.