The Simplex Algorithm

IE 496 Lecture 24
Reading for This Lecture

- Primary
  - Bazaraa, Sherali, and Sheti, Chapter 2.
  - Chvatal, Chapters 6 and 7.
Consider again the system $Ax = b$, $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$.

In this problem, there are either
- no solutions
- one solution
- infinitely many solutions (if $n > m$)

The problem of linear programming is

$$\begin{align*}
\min & \quad c^T x \\
\text{s.t.} & \quad Ax = b \\
& \quad x \geq 0
\end{align*}$$
The Simplex Algorithm

- Note that $x_B = B^{-1}b - B^{-1}Nx_N$

- Hence, $c^T x = c_B^T x_B + c_N^T x_N = c_B^T B^{-1}b + (c_N^T - c_B^T B^{-1}N)x_N$

- So if $c_N^T - c_B^T B^{-1}N \geq 0$, we have found the optimal solution (why?).

- Otherwise, suppose some component of $c_N^T - c_B^T B^{-1}N$ is negative.

- Then we raise the value of the corresponding variable as much as possible while maintaining feasibility.
More Terminology

- The matrix $B$ is called the *basis*.
- The variables corresponding to the columns of $B$ are the *basic* variables.
- All other variables are called *non-basic*.
- The fundamental step the simplex algorithm is called a *pivot*.
  - We add one basic variable and remove another.
  - We do this in such a way that feasibility is maintained and the cost deceases at each step.
Summary of the Simplex Algorithm

- **Simplex algorithm**
  - Solve $yB = c_B^T$
  - Choose a column of $a_j$ of $N$ such $ya_j < c_j$
  - Solve $Bd = a_j$
  - Find the largest $t$ such that $x_B^* - td \geq 0$
  - Set the value of $x_j$ to $t$ and the values of the basic variables to $x_B^* - td$.
  - Update the basis.

- The interesting part is implementing the last step.
Implementing the Algorithm

- Let $B_k$ be the basis after the $k^{th}$ iteration.
- Note that $B_k = B_{k-1}E_k$ where
  - $E_k$ is the identity matrix with the $p^{th}$ column replaced by $d = B_{k-1}^{-1}a_j$ (already computed).
  - $p$ is the "leaving column"
- So, we have $B_k = B_0E_1 .... E_k = LUE_1 .... E_k$
- To update at each iteration, we merely append the next eta matrix to the list.
- Often, $B_0$ is the identity matrix.
Refactorizing the Basis

- After many iterations, it can become inefficient to solve these systems.
- Periodically, throw away all the eta files and calculate a brand new LU factorization.
- How often should this be done?
- It depends on the matrix.
- Under some fairly reasonable assumptions, the "break-even" point seems to be $\approx 15$ iterations.
Another Approach

- Update the LU factorization directly
- We have $B_k = L_k U_k$.
- We also have $B_{k+1} = B_k E_{k+1}$.
- Hence, $B_{k+1} = L_k U_k E_{k+1}$.
- We can permute the rows and columns of $V = U_k E_{k+1}$ such that $V$ differs from an upper-triangular matrix in at most one row.
- It is then easy to perform an LU factorization of $V$.
- This can easily be made into an LU factorization of $B_{k+1}$.
Issues to be addressed

- Ensuring numerical accuracy
  - Conditioning
  - Stability
  - Zero tolerances

- Ensuring efficiency
  - Maintaining sparsity
  - Updating basis factorization
Dealing with Large Matrices

- Recall this step from the Simplex Algorithm:
  - Choose a column of $a_j$ of $N$ such that $ya_j < c_j$
- This step is called *pricing*.
- One approach is to choose the quantity $c_j - ya_j$ to be as large as possible.
- If the number of columns of $A$ is large, then the pricing step can be cumbersome.
- Partial pricing is the practice of only pricing out a small subset of possible columns.
Column Generation

- Notice that the problem $\max \{c_j - y_{a_j}\}$ is an optimization problem.
- Notice also that it is not necessary to have all the columns present in the matrix.
- Suppose the columns of the matrix have a special structure that allows us to generate them "automatically".
- We can solve the above optimization problem to determine the next column to be pivoted in.
- All we really need is the columns of the optimal basis.
Constraint Generation

• Consider an LP specified as follows

\[
\begin{align*}
\text{min} & \quad c^T x \\
\text{s.t.} & \quad Ax \leq b
\end{align*}
\]

• In this case, we can sometimes have \( m >> n \).

• Constraints (rows) can also be automatically generated.

• This is called separation.
Deleting Columns and Rows

- If the slack variable for a particular row is basic, then that row is "inactive".
- Inactive rows can be deleted from the problem without changing the optimal solution.
- Similarly, there are methods of proving that a particular column can never be basic in an optimal solution.
- While solving large LP's by column and constraint generation, we can simultaneously purge ineffective rows and columns and generate new ones.
- This technique can be very effective.