

Lecture 9: The Simplex method

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An algebraic derivation of the pricing step

- $z^* = \text{minimum } c^T x = \text{minimum } c_B^T x_B + c_N^T x_N$
 subject to $Ax = b$, subject to $Bx_B + Nx_N = b$,
 $x \geq 0^n$ $x_B \geq 0_m$; $x_N \geq 0_{n-m}$
- $= c_B^T B^{-1}b + \text{minimum } [c_N^T - c_B^T B^{-1}N]x_N$
 subject to $B^{-1}b - B^{-1}Nx_N \geq 0_m$,
 $x_N \geq 0_{n-m}$
- $x_N = 0_{n-m}$ feasible. Let $\bar{c}_N := c_N^T - c_B^T B^{-1}N$.
- If reduced cost $\bar{c}_N \geq 0_{n-m}$ then $x_N = 0_{n-m}$ is optimal.

- If $\bar{c}_N \not\leq \mathbf{0}_{n-m}$ then $\exists j \in N$ with $\bar{c}_j > 0$. Then the current point $\mathbf{x}_N = \mathbf{0}_{n-m}$ may be non-optimal.
- Generate a feasible descent direction!
- Choose one that leads to a neighboring extreme point!
- Swap one variable in B for one in N !
- Increase one variable in N from zero!
- Choose j^* to be among $\arg \min_{j \in N} \bar{c}_j$.

The basis change

- In x_N -space: $p_N = e_{j^*}$ (unit vector)
- In x_B -space: $x_B = B^{-1}b - B^{-1}N x_N \iff p_B = -B^{-1}N p_N = -B^{-1}N_{j^*}$
- So, search direction in \mathbb{R}^n :

$$p = \begin{pmatrix} p_B \\ p_N \end{pmatrix} = \begin{pmatrix} -B^{-1}N_{j^*} \\ e_{j^*} \end{pmatrix}$$

- Descent? Yes, because $e^T p = \bar{c}_{j^*} > 0!$
- Feasible? Must check that $\forall p = \mathbf{0}_m$ and that $p_i \geq 0$ if $x_i = 0, i \in B$.

- (a) $Ap = Bp_B + Np_N = -BB^{-1}N_{j^*} + Ne_{j^*} = \mathbf{0}_m$

- (b) Suppose that $x_B > \mathbf{0}_m$. Then, at least a small step in p keeps $x_B \geq \mathbf{0}_m$. But if there is an i^* with $(x_B)_{i^*} = 0$ and $(p_B)_{i^*} > 0$ then it is not a feasible direction.

- Must then perform a basis change without moving! A *degenerate basis change*: swap x_{j^*} for x_{i^*} in the basis.

- Otherwise (and normally), we utilize the unit direction.
- Line search? Linear objective; move the maximum step!
- Maximum step: If $p_B \geq \mathbf{0}_m$ there is no maximum step!

We have found an extreme direction p along which $c^T x$ tends to $-\infty$! *Unbounded solution.*

- Otherwise: Some basic variables will reach zero eventually. Choose as the outgoing variable a variable $i \in B$ with minimum in

$$\left\{ \min_{i \in B} \left\{ \frac{(B^{-1}b)_i}{(B^{-1}N_{j^*})_i} \mid (B^{-1}N_{j^*})_i > 0 \right\} \right.$$
- Done. In the basis, replace i^* by j^* ! Go back to the pricing step.

Computational notes—how do we do all of this?

- Given basis matrix B , solve

$$Bx_B = b.$$

- Gives us BFS: $x_B = B^{-1}b$.

- Pricing step: (a) Solve

$$B^T y = c_B.$$

- (b) Calculate $\bar{c}_N^T = c_N^T - y^T N$, the reduced cost vector.

- Choose the incoming variable, x_{j^*} .

- Outgoing variable: Solve

$$Bp_B = -N_{j^*}.$$

- The quotient rule for $(B^{-1}b)_i / (-p_B)_i$ gives the outgoing variable, x_{i^*} .

- Note: Three similar linear systems in B ! LU

factorization + three triangular substitutions!

- Factorizations can be updated after basis change rather than done from scratch.

- LP solvers like Cplex and XPRESS-MP have excellent numerical solvers for linear systems.

- Linear systems the bulk of the work in solving an LP.

Convergence

- Theorem 10.10: *If all of the basic feasible solutions are non-degenerate, then the Simplex algorithm terminates after a finite number of iterations.*
- Rough argument: Non-degeneracy implies that the step length is > 0 ; hence, we cannot return to an old BFS once we have left it. There are finitely many BFSs. \square

- Degereracy: Can actually lead to cycling—the same sequence of BFSs is returned to indefinitely!
- Remedy: Change the incoming/outgoing criterial!

Bland's rule: Sort variables according to some index ordering. Take the first possible index in the list.

Incoming variable first in the list with the right sign of the reduced cost; outgoing variable the first in the list among the minima in the quotient rule.

Initial BFS: Phase I of the Simplex method

- If a starting BFS cannot be found, do the following.
- Suppose $\mathbf{b} \geq \mathbf{0}_m$. Introduce artificial variables a_i in every row (or rows without a unit column).

- Solve the following Phase-I problem:

$$\text{minimize } w = (\mathbf{1}_m)^T \mathbf{a}$$

subject to

$$\mathbf{A}\mathbf{x} + \mathbf{I}_m \mathbf{a} = \mathbf{b},$$

$$\mathbf{x} \geq \mathbf{0}_n,$$

$$\mathbf{a} \geq \mathbf{0}_m.$$

- Possible cases: (a) $w_* = 0$, meaning that $\mathbf{a}_* = \mathbf{0}_m$ must hold. There is then a BFS in the original problem.

- Start Phase-II, to solve the original problem, starting from this BFS.
- (b) $w_* > 0$. The optimal basis then has some $a_i^* > 0$; due to the objective function construction, there exists no BFS in the original problem. The problem is then infeasible!
- What to do then? Modelling errors? Can be detected from the optimal solution. In fact some LP problems are pure feasibility problems.