

Lecture 9: The Simplex method

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An algebraic derivation of the pricing step

- $$z^* = \infimum c^T x = \infimum c_B^T x_B + c_N^T x_N$$
- subject to $Ax = b$, subject to $Bx_B + Nx_N = b$,
- $$x \geq 0^n \quad x_B \geq 0^m, x_N \geq 0^{n-m}$$
- $$= c_B^T B^{-1}b + \infimum [c_N^T - c_B^T B^{-1}N]x_N$$
- 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.

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- If $\bar{c}_N \not\geq 0^{n-m}$ then $\exists j \in N$ with $\bar{c}_j < 0$. Then the current point $x_N = 0^{n-m}$ 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 $\arg \min_{j \in N} \bar{c}_j$.

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The basis change

- In x_N -space: $p_N = e_{j^*}$ (unit vector)
- In x_B -space: $x_B = B^{-1}b - B^{-1}Nx_N \implies p_B = -B^{-1}Np_N = -B^{-1}N_{j^*}$
- 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 $c^T p = \bar{c}_{j^*} < 0$!
- Feasible? Must check that $Ap = 0^m$ and that $p_i \geq 0$ if $x_i = 0, i \in B$.

- (a) $A\mathbf{p} = \mathbf{B}\mathbf{p}_B + \mathbf{N}\mathbf{p}_N = -\mathbf{B}\mathbf{B}^{-1}\mathbf{N}_{j^*} + \mathbf{N}\mathbf{e}_{j^*} = \mathbf{0}^m$
- (b) Suppose that $\mathbf{x}_B > \mathbf{0}^m$. Then, at least a small step in \mathbf{p} keeps $\mathbf{x}_B \geq \mathbf{0}^m$. But if there is an i^* with $(\mathbf{x}_B)_{i^*} = 0$ and $(\mathbf{p}_B)_{i^*} < 0$ 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 $\mathbf{p}_B \geq \mathbf{0}^n$ there is no maximum step! We have found an extreme direction \mathbf{p} along which $\mathbf{c}^T\mathbf{x}$ 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

$$\text{minimum}_{i \in B} \left\{ \frac{(\mathbf{B}^{-1}\mathbf{b})_i}{(\mathbf{B}^{-1}\mathbf{N}_{j^*})_i} \mid (\mathbf{B}^{-1}\mathbf{N}_{j^*})_i > 0 \right\}$$
- 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 \mathbf{B} , solve

$$\mathbf{B}\mathbf{x}_B = \mathbf{b}.$$
- Gives us BFS: $\mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b}$.
- Pricing step: (a) Solve

$$\mathbf{B}^T\mathbf{y} = \mathbf{c}_B.$$
- (b) Calculate $\bar{\mathbf{c}}_N^T = \mathbf{c}_N^T - \mathbf{y}^T\mathbf{N}$, the reduced cost vector.
- Choose the incoming variable, x_{j^*} .

- Outgoing variable: Solve

$$\mathbf{B}\mathbf{p}_B = -\mathbf{N}_{j^*}.$$
- The quotient rule for $(\mathbf{B}^{-1}\mathbf{b})_i / (-\mathbf{p}_B)_i$ gives the outgoing variable, x_{i^*} .
- Note: Three similar linear systems in \mathbf{B} ! LU factorization + three triangular substitutions!
- Factorizations can be updated after basis change rather than done from scratch.
- LP solvers like Cplex and XPPRESS-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

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

$$\begin{array}{ll} \text{minimize} & w = (\mathbf{1}^m)^T \mathbf{a} \\ \text{subject to} & \mathbf{A}\mathbf{x} + \mathbf{I}^m \mathbf{a} = \mathbf{b}, \\ & \mathbf{x} \geq \mathbf{0}^n, \\ & \mathbf{a} \geq \mathbf{0}^m. \end{array}$$

- 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.