

MATH 170 – PROBLEM SET 3 (DUE TUESDAY FEBRUARY 7)
SOLUTIONS BY FREDERICK LAW

1. (B&T 2.18) Consider a polyhedron $P = \{\mathbf{x} : \mathbf{Ax} \geq \mathbf{b}\}$. Given any $\varepsilon > 0$, show that there exists some $\bar{\mathbf{b}}$ with the following two properties: (a) The absolute value of every component $\mathbf{b} - \bar{\mathbf{b}}$ is bounded by ε . (b) Every basic feasible solution in the polyhedron $P = \{\mathbf{x} : \mathbf{Ax} \geq \bar{\mathbf{b}}\}$ is nondegenerate.

Solution: Intuitively, we shall find such a $\bar{\mathbf{b}}$ by first identifying excessive constraints and perturbing the necessary constraints by some ε . That is, if $P = H_1 \cap \dots \cap H_m$, where H_j are the half-spaces whose intersection gives us P , then we identify those H_j which are unnecessary in defining P . If H_j is excessive, then P is well defined apart from H_j , which means that $\bigcap_{k \neq j} H_k = P$. Let M be the set of all indices of excessive half-spaces. Then $P = \bigcap_{k \notin M} H_k$, that is, we can remove all the half-spaces of M and still get the same polyhedron. This is true just because of our construction, since all the half-spaces removed are excessive. If $H_j = \{\mathbf{x} \in \mathbb{R}^n : a'_j \mathbf{x} \geq b_j\}$, then let $H'_k = \{\mathbf{x} \in \mathbb{R}^n : a'_k \mathbf{x} \geq b_j - \varepsilon\}$. Let $\bar{\mathbf{b}}$ be defined by:

$$\bar{\mathbf{b}} = \begin{cases} b_j & j \notin M \\ b_j - \varepsilon & j \in M \end{cases}$$

Then our new polyhedron is $P' = \bigcap_{k \in M} H_k \cap \bigcap_{j \in M} H'_j$. Note that in P , the degenerate basic feasible solutions can be interpreted as having too many hyperplanes, boundary of half-spaces, touching at that point. For example, on a cube in three dimensions, all the points are nondegenerate as they all have 3 hyperplanes touching them, the three facets that connect at a corner. Moreover, every facet of a polyhedron is not excessive, since it serves as a boundary of the polyhedron. Thus, we can interpret the excessive constraints as half-spaces whose hyperplane boundary either touches the polyhedron only at some degenerate solution or at not degenerate solution. Therefore, by perturbing these outward by ε , we remove all the degeneracy from our basic feasible solutions. Thus all the basic feasible solutions in P' are nondegenerate. Also, since we have only perturbed the excessive half-spaces by ε , it follows that by construction $|\bar{b}_i - b_i| \leq \varepsilon$ for all i .

2. (B&T 2.21) Suppose that Fourier-Motzkin elimination is used in the manner described at the end of Section 2.8 to find the optimal cost in a linear programming problem. Show how this approach can be augmented to obtain an optimal solution as well.

Solution: To get an optimal solution using the Fourier-Motzkin elimination, we first find the optimal solution. This is done by extending our LP by one variable x_0 , so we get a new polyhedron in \mathbb{R}^{n+1} defined by $\{(x_0, \mathbf{x}) : \mathbf{x} \in P, \mathbf{c}'\mathbf{x} = x_0\}$. Then we use the Fourier Motzkin elimination to project onto the first variable, which gives us $\{x_0 \in \mathbb{R} : \exists \mathbf{x} \in P \text{ s.t. } \mathbf{c}'\mathbf{x} = x_0\}$. Then we minimize over this subset of \mathbb{R} to find our optimal cost, call this c^* . To get an optimal solution, we really just project back upwards on n dimensions, by inverting the Fourier Motzkin algorithm. Moreover, to save time in our algorithm, we really only need to project upwards starting at c^* . That is, if we imagine our entire process as a mapping $\Phi : P \rightarrow \mathbb{R}$ where $\mathbf{x} \in P$ gets sent to $\mathbf{c}'\mathbf{x}$, then we take the preimage over c^* which is just the level set $\Phi^{-1}(c^*)$ and see what points lie on the level set. This is also the same as taking the plane $\{\mathbf{x} \in \mathbb{R}^n : \mathbf{c}'\mathbf{x} = c^*\}$ and finding where this hyperplane intersects P .

3. (B&T 2.22) Let P and Q be polyhedra in \mathbb{R}^n . Let $P + Q = \{\mathbf{x} + \mathbf{y} : \mathbf{x} \in P, \mathbf{y} \in Q\}$.

(a) Show that $P + Q$ is a polyhedron.

Solution: Let us define M as $M = \{(\mathbf{z}, \mathbf{x}, \mathbf{y}) : \mathbf{x} \in P, \mathbf{y} \in Q, \mathbf{z} = \mathbf{x} + \mathbf{y}\}$. If P is constructed with n_1 linear constraints and Q is constructed with n_2 linear constraints, then M is constructed with $n_1 + n_2 + n$ linear constraints, where n of them come from

$\mathbf{z} = \mathbf{x} + \mathbf{y}$ component wise. Therefore since M is constructed using linear constraints, M is a polyhedron in \mathbb{R}^{3n} . Then we use Fourier-Motzkin elimination to reduce to the first n coordinates: $\Pi_n(M) = \{\mathbf{z} \in \mathbb{R}^n : \exists \mathbf{x} \in P, \mathbf{y} \in Q \text{ s.t. } \mathbf{x} + \mathbf{y} = \mathbf{z}\}$. This can be rewritten as $\Pi_n(M) = \{\mathbf{x} + \mathbf{y} \in \mathbb{R}^n : \mathbf{x} \in P, \mathbf{y} \in Q\} = P + Q$. By the Fourier-Motzkin elimination algorithm, we know that $\Pi_n(M)$ is a polyhedron, and thus $P + Q$ is a polyhedron. \square

- (b) Show that every extreme point of $P + Q$ is the sum of an extreme point of P and an extreme point of Q .

Solution: Suppose not. Then there exists $\mathbf{x} + \mathbf{y}$ which is extreme in $P + Q$ but either \mathbf{x} is not extreme in P or \mathbf{y} is not extreme in Q or both. WLOG, suppose that \mathbf{x} is not an extreme in P , \mathbf{y} may or may not be extreme in Q . Since \mathbf{x} is not extreme in P then that means there exists $\mathbf{z}, \mathbf{z}' \in P, \lambda \in [0, 1]$ such that $\mathbf{x} \neq \mathbf{z}$ and $\mathbf{x} \neq \mathbf{z}'$ and $\mathbf{x} = \lambda \mathbf{z} + (1 - \lambda)\mathbf{z}'$. Then we have

$$\mathbf{x} + \mathbf{y} = \lambda \mathbf{z} + (1 - \lambda)\mathbf{z}' + \mathbf{y} = \lambda(\mathbf{z} + \mathbf{y}) + (1 - \lambda)(\mathbf{z}' + \mathbf{y})$$

But now we have written $\mathbf{x} + \mathbf{y}$ as a convex combination of $\mathbf{z} + \mathbf{y}$ and $\mathbf{z}' + \mathbf{y}$, where $\mathbf{x} + \mathbf{y} \neq \mathbf{z} + \mathbf{y}$ and $\mathbf{z}' + \mathbf{y}$, since $\mathbf{x} \neq \mathbf{z}$ and $\mathbf{x} \neq \mathbf{z}'$. Therefore $\mathbf{x} + \mathbf{y}$ is not an extreme point. This is a contradiction, so we are done. \square

4. (B&T 3.2) (**Optimality conditions**) Consider the problem of minimizing $\mathbf{c}'\mathbf{x}$ over a polyhedron P . Prove the following:

- (a) A feasible solution \mathbf{x} is optimal if and only if $\mathbf{c}'\mathbf{d} \geq 0$ for every feasible direction \mathbf{d} at \mathbf{x} .

Solution: First we prove the forward direction. Suppose that \mathbf{x} , a feasible solution, is optimal. Then it follows that $\forall \mathbf{y} \in P, \mathbf{c}'\mathbf{x} \leq \mathbf{c}'\mathbf{y}$. Suppose \mathbf{d} is an arbitrary feasible direction at \mathbf{x} . Then there exists $\theta > 0$ such that $\mathbf{x} + \theta\mathbf{d} \in P$. Then $\mathbf{c}'\mathbf{x} \leq \mathbf{c}'(\mathbf{x} + \theta\mathbf{d}) = \mathbf{c}'\mathbf{x} + \theta\mathbf{c}'\mathbf{d}$. Thus $\theta\mathbf{c}'\mathbf{d} \geq 0$, and since $\theta > 0$, we divide by θ and get $\mathbf{c}'\mathbf{d} \geq 0$.

Now we prove the backward direction. Suppose that $\mathbf{c}'\mathbf{d} \geq 0$ for every feasible direction \mathbf{d} at \mathbf{x} . Let $\mathbf{y} \in P$ be arbitrary. Then let $\mathbf{d} = \mathbf{y} - \mathbf{x}$. Then if we let $\theta = 1$, then $\mathbf{x} + \theta\mathbf{d} = \mathbf{y} \in P$, so \mathbf{d} is a feasible direction at \mathbf{x} and so $\mathbf{c}'\mathbf{d} \geq 0$. Thus it follows that $\mathbf{c}'(\mathbf{y} - \mathbf{x}) \geq 0$, and so $\mathbf{c}'\mathbf{y} - \mathbf{c}'\mathbf{x} \geq 0$ and $\mathbf{c}'\mathbf{x} \leq \mathbf{c}'\mathbf{y}$. Since \mathbf{y} was an arbitrary point in P , it follows that \mathbf{x} is an optimal solution. This proves the equality. \square

- (b) A feasible solution \mathbf{x} is the unique optimal solution if and only if $\mathbf{c}'\mathbf{d} > 0$ for every nonzero feasible direction \mathbf{d} at \mathbf{x} .

Solution: Our argument will be similar to that in part (a). First we prove the forward direction. Suppose that \mathbf{x} , a feasible solution, is unique optimal. Then this means that for any $\mathbf{y} \in P$ such that $\mathbf{y} \neq \mathbf{x}$, then $\mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{y}$. Let \mathbf{d} be any nonzero feasible direction at \mathbf{x} . Then there exists $\theta > 0$ such that $\mathbf{x} + \theta\mathbf{d} \in P$. Since $\theta > 0$ and $\mathbf{d} \neq 0$, then it follows that $\mathbf{x} \neq \mathbf{x} + \theta\mathbf{d}$. Therefore $\mathbf{c}'\mathbf{x} < \mathbf{c}'(\mathbf{x} + \theta\mathbf{d}) = \mathbf{c}'\mathbf{x} + \theta\mathbf{c}'\mathbf{d}$. Subtracting $\mathbf{c}'\mathbf{x}$ from both sides and dividing by $\theta > 0$, it follows that $\mathbf{c}'\mathbf{d} > 0$.

Now we prove the backward direction. Suppose that $\mathbf{c}'\mathbf{d} > 0$ for any nonzero feasible direction at \mathbf{x} . Let $\mathbf{y} \in P$ be an arbitrary point such that $\mathbf{y} \neq \mathbf{x}$. Since $\mathbf{y} \neq \mathbf{x}$, then $\mathbf{d} = \mathbf{y} - \mathbf{x} \neq 0$. Using $\theta = 1$, then $\mathbf{x} + \theta\mathbf{d} = \mathbf{y} \in P$, so \mathbf{d} is a feasible, nonzero, direction at \mathbf{x} . Therefore it follows that $\mathbf{c}'\mathbf{d} > 0$. Thus $\mathbf{c}'(\mathbf{y} - \mathbf{x}) = \mathbf{c}'\mathbf{y} - \mathbf{c}'\mathbf{x} > 0$. And thus it follows that $\mathbf{c}'\mathbf{y} > \mathbf{c}'\mathbf{x}$. Since \mathbf{y} was any point in P not equal to \mathbf{x} , it follows that \mathbf{x} is the unique optimal solution. This proves the equality. \square

5. (B&T 3.3) Let \mathbf{x} be an element of the standard form polyhedron $P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{A}\mathbf{x} = \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}$. Prove that a vector $\mathbf{d} \in \mathbb{R}^n$ is a feasible direction at \mathbf{x} if and only if $\mathbf{A}\mathbf{d} = \mathbf{0}$ and $d_i \geq 0$ for every i such that $x_i = 0$.

Solution: We first prove the forward direction. Suppose that \mathbf{d} is feasible at \mathbf{x} . Then there exists $\theta > 0$ such that $\mathbf{x} + \theta\mathbf{d} \in P$. This means $\mathbf{A}(\mathbf{x} + \theta\mathbf{d}) = \mathbf{A}\mathbf{x} + \theta\mathbf{A}\mathbf{d} = \mathbf{b}$. Since $\mathbf{x} \in P$, then

$\mathbf{Ax} = \mathbf{b}$, thus subtracting \mathbf{b} from both sides, and dividing by $\theta > 0$, it follows that $\mathbf{Ad} = \mathbf{0}$. If i is an index such that $x_i = 0$, and since $\mathbf{x} + \theta\mathbf{d} \geq \mathbf{0}$, then $x_i + \theta d_i = \theta d_i \geq 0$. Dividing by $\theta > 0$, it follows that $d_i \geq 0$ for any index i such that $x_i \geq 0$.

Now we prove the backward direction. Suppose $\mathbf{d} \in \mathbb{R}^n$ such that $\mathbf{Ad} = \mathbf{0}$ and $d_i \geq 0$ for any index i such that $x_i = 0$. Let us choose θ^* satisfying

$$0 < \theta^* < \inf \left\{ -\frac{x_j}{d_j} : x_j > 0, d_j < 0 \right\}$$

Since $\mathbf{x} \in P$, then $x_i \geq 0$ for all i . If $x_i = 0$, then $x_i + \theta d_i = \theta d_i \geq 0$, since $d_i \geq 0$ and $\theta > 0$. If $x_i > 0$, there are two cases. If $d_i \geq 0$, then $x_i + \theta d_i \geq 0$. If $d_i < 0$, then $\theta^* < -\frac{x_i}{d_i}$, and multiplying through by $d_i < 0$, we get $\theta^* d_i > -x_i$ and thus $x_i + \theta^* d_i \geq 0$. Thus $\mathbf{x} + \theta^* \mathbf{d} \geq \mathbf{0}$. Also, $\mathbf{A}(\mathbf{x} + \theta^* \mathbf{d}) = \mathbf{Ax} + \theta^* \mathbf{Ad} = \mathbf{Ax} = \mathbf{b}$ since $\mathbf{x} \in P$. Thus it follows that $\mathbf{x} + \theta^* \mathbf{d} \in P$. This proves the equality, and we are done. \square

6. (B&T 3.5) Let $P = \{\mathbf{x} \in \mathbb{R}^3 : x_1 + x_2 + x_3 = 1, \mathbf{x} \geq \mathbf{0}\}$ and consider the vector $\mathbf{x} = (0, 0, 1)$. Find the set of feasible directions at \mathbf{x} .

Solution: We find the set of feasible directions by working with the definition of feasible directions. We say that \mathbf{d} is feasible at \mathbf{x} if there exists some $\theta > 0$ such that $\mathbf{x} + \theta\mathbf{d} \in P$. Let $\mathbf{d} = (d_1, d_2, d_3)$. Then $\mathbf{x} + \theta\mathbf{d} = (\theta d_1, \theta d_2, 1 + \theta d_3)$. To require this to be in P , we need that $\theta(d_1 + d_2 + d_3) + 1 = 1$, which means $d_1 + d_2 + d_3 = 0$, and thus $d_3 = -d_1 - d_2$. We also require d_1, d_2 to be non-negative, since then for any $\theta > 0$, $\theta d_1, \theta d_2 \geq 0$. Lastly, we shall require $1 + \theta d_3 \geq 0$. To do this, we shall choose θ so that $1 + \theta d_3 = 0$. Then $1 = \theta(d_1 + d_2)$ and $\theta = \frac{1}{d_1 + d_2}$. Note this θ will always be positive, unless $d_1 = d_2 = 0$. But in that case, $d_3 = 0$, and $\mathbf{d} = \mathbf{0}$, which is always a feasible direction. Thus our set of feasible directions is

$$F = \{(d_1, d_2, -d_1 - d_2) : d_1 \geq 0, d_2 \geq 0\}$$

As a safe mental check, if $d_1 = d_2 = 0$, then $\mathbf{d} = \mathbf{0}$, and for $\theta = 1$, $\mathbf{x} + \mathbf{d} = \mathbf{x} \in P$. If either d_1 or $d_2 > 0$, then let $\theta = \frac{1}{d_1 + d_2} > 0$, and then $\mathbf{x} + \theta\mathbf{d} = \left(\frac{d_1}{d_1 + d_2}, \frac{d_2}{d_1 + d_2}, 0\right) \in P$, since $\frac{d_1}{d_1 + d_2} + \frac{d_2}{d_1 + d_2} = \frac{d_1 + d_2}{d_1 + d_2} = 1$, and $\frac{d_i}{d_1 + d_2} \geq 0$ for $i = 1, 2$. \square

7. (B&T 3.6) (**Conditions for a unique optimum**) Let \mathbf{x} be a basic feasible solution associated with some basis matrix \mathbf{B} . Prove the following:

- (a) If the reduced cost of every nonbasic variable is positive, then \mathbf{x} is the unique optimal solution.

Solution: Suppose that $\bar{c}_j > 0$ for every nonbasic variable j . Let $\mathbf{y} \in P$ be arbitrary such that $\mathbf{y} \neq \mathbf{x}$. Then $\mathbf{Ax} = \mathbf{Ay} = \mathbf{b}$. Thus it follows that $\mathbf{A}(\mathbf{y} - \mathbf{x}) = \mathbf{0}$. Let $\mathbf{d} = \mathbf{y} - \mathbf{x}$. Then using $\theta = 1$, $\mathbf{x} + \mathbf{d} = \mathbf{y} \in P$, so \mathbf{d} is a feasible direction at \mathbf{x} and $\mathbf{Ad} = \mathbf{0}$. We can rewrite this as $\mathbf{B}\mathbf{d}_B + \sum_{j \in I} A_j d_j = \mathbf{0}$ where I is the set of nonbasic indices. Since \mathbf{B} is invertible, then $\mathbf{d}_B = -\sum_{j \in I} \mathbf{B}^{-1} A_j d_j$. Then

$$\mathbf{c}'\mathbf{d} = \mathbf{c}'_B \mathbf{d}_B + \sum_{j \in I} c_j d_j = \sum_{j \in I} (c_j - \mathbf{c}'_B \mathbf{B}^{-1} A_j) d_j = \sum_{j \in I} \bar{c}_j d_j$$

Since $\mathbf{y} \neq \mathbf{x}$, there must exist some $j \in I$ such that $y_j \neq x_j = 0$, where $x_j = 0$ for all $j \in I$ since \mathbf{x} is a basic feasible solution. If $\mathbf{y} \in P$ such that $y_j = 0$ for all $j \in I$, then $\mathbf{Ay} = \mathbf{B}\mathbf{y}_B = \mathbf{b}$. Then $\mathbf{y}_B = -\mathbf{B}^{-1}\mathbf{b} = \mathbf{x}_B$, and so $\mathbf{y} = \mathbf{x}$. Therefore there must exist some nonbasic index on which \mathbf{y} and \mathbf{x} disagree. Let $k \in I$ be such an index. Since $\mathbf{y} \in P$, then $y_k \geq 0$, and since $y_k \neq x_k = 0$, then $y_k > 0$. Since $d_j = y_j - x_j$ and since $x_j = 0$ for all $j \in I$:

$$\mathbf{c}'\mathbf{d} = \sum_{j \in I} \bar{c}_j d_j = \sum_{j \in I} \bar{c}_j y_j \geq \bar{c}_k y_k > 0$$

where we use the fact that $\bar{c}_j \geq 0$ for all $j \in I$ and $y_j \geq 0$ for all $j \in I$ since $\mathbf{y} \in P$. Thus we have $\mathbf{c}'(\mathbf{y} - \mathbf{x}) > 0$ and thus $\mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{y}$. Since this holds for all $\mathbf{y} \neq \mathbf{x}$, it follows that \mathbf{x} is the unique optimal solution. \square

- (b) If \mathbf{x} is the unique optimal solution and is nondegenerate, then the reduced cost of every nonbasic variable is positive.

Solution: Suppose not. Then there exists an index $m \in I$, borrowing notation from part (a), such that $\bar{c}_m \leq 0$. Let \mathbf{d} be the m th basic direction. We know from our results in class and in Bertsimas and Tsitsiklis that \mathbf{d} is always a feasible direction at \mathbf{x} since \mathbf{x} is nondegenerate. Then there exists $\theta > 0$ such that $\mathbf{x} + \theta\mathbf{d} \in P$. Then

$$\mathbf{c}'\mathbf{d} = \sum_{j \in I} \bar{c}_j d_j = \bar{c}_k \leq 0$$

since \mathbf{d} is the k th basic direction, so $d_k = 1$ and $d_j = 0$ for $j \in I, j \neq k$. Therefore $\mathbf{c}'\mathbf{d} \leq 0$. Choosing $\theta > 0$ sufficiently small so that $\mathbf{x} + \theta\mathbf{d} \in P$, we get $\theta\mathbf{c}'\mathbf{d} \leq 0$ which implies that $\mathbf{c}'(\mathbf{x} + \theta\mathbf{d}) = \mathbf{c}'\mathbf{x} + \theta\mathbf{c}'\mathbf{d} \leq \mathbf{c}'\mathbf{x}$. But $\mathbf{x} + \theta\mathbf{d} \in P$, which means that \mathbf{x} is not the unique optimal solution. This is a contradiction, which completes the proof. \square

8. (B&T 3.10) Show that if $n - m = 2$, then the simplex method will not cycle, no matter which pivoting rule is used.

Solution: Note that there is a lot of possible choices in the simplex method, thus if this fact is indeed true it, then the fact that $n - m = 2$ must do something eliminate our possible choices in directions to move when executing the simplex method. One possible explanation could be that when $n - m = 2$, then either the polyhedron is unbounded, in which case simplex terminates at optimal cost being $-\infty$ or the polyhedron is bounded and it happens that in the polytope all the basic feasible solutions are nondegenerate, and then the simplex method on this polytope will have to terminate.