

## Problem Set Solution 1

1. Consider  $P = \{x : Ax \leq b, x \geq 0\}$ , where  $A$  is  $m \times n$ . Show that if  $x$  is a vertex of  $P$  then we can find sets  $I$  and  $J$  with the following properties.

- (a)  $I \subseteq \{1, \dots, m\}$ ,  $J \subseteq \{1, \dots, n\}$  and  $|I| = |J|$ .
- (b)  $A_J^I$  is invertible where  $A_J^I$  is the submatrix of  $A$  corresponding to the rows in  $I$  and the columns in  $J$ .
- (c)  $x_j = 0$  for  $j \notin J$  and  $x_J = (A_J^I)^{-1}b^I$  where  $b^I$  denotes the restriction of  $b$  to the indices in  $I$ .

(Hint: Consider  $Q = \{(x, s) : Ax + Is = b, x \geq 0, s \geq 0\}$ .)

Using the hint we turn our attention to  $Q = \{(x, s) : Ax + Is = b, x \geq 0, s \geq 0\}$ . If we let  $(x, s)$  be a pair such that  $x \in P$  and  $s$  is the unique vector of slack variables associated with  $x$  ( $s = b - Ax$ ), it is not hard to show that if  $x$  is a vertex of  $P$  then  $(x, s)$  is a vertex of  $Q$ . Assume that  $x$  is a vertex of  $P$  but there is a  $(y, t)$  such that  $(x, s) \pm (y, t) \in Q$  and  $(y, t) \neq 0$ . Then we have  $A(x \pm y) + (s \pm t) = b, x \pm y \geq 0, s \pm t \geq 0$ . This implies that  $A(x + y) \leq b$  and  $A(x - y) \leq b$ . Since  $x$  is a vertex, this implies  $y = 0$ . Then solving for  $t$  in  $Ax + (s + t) = b$  we find that it must be zero as well which contradicts  $(y, t) \neq 0$ .

We can now take advantage of the fact that  $Q$  is in the special form ( $Ax = b, x \geq 0$ ). If  $(x, s)$  is a vertex of  $Q$  then there is a subset  $B \subseteq \{1, \dots, n + m\}$  such that  $|B| = m$  and

- (a)  $(x, s)_N = 0$  for  $N = \{1, \dots, n + m\} \setminus B$
- (b)  $(A|I)_B$  is non singular
- (c)  $(x, s)_B = (A|I)_B^{-1}b \geq 0$ .

Let  $J \subseteq B$  be the set of columns involving  $A$  of  $(A|I)$ . Notice that if  $|J| = k$ ,  $k$  of the  $x$  variables are basic and  $m - k$  of the  $s$  variables are basic. So  $x_j = 0$  for  $j \notin J$  and  $k$  of the  $s$  variables are zero. We take the rows corresponding to the zero components of  $s$  as the set  $I$ . Then

$$A^I x = A_J^I x_J = b^I$$

and  $A_J^I$  is invertible, so

$$x_J = (A_J^I)^{-1}b^I.$$

2. In his paper in FOCS 92, Tomasz Radzik needs a result of the following form (Page 662 of the Proceedings):

**Lemma 1** *Let  $c \in \mathbb{R}^n$  and  $y_k \in \{0, 1\}^n$  for  $k = 1, \dots, q$  such that  $2|y_{k+1}c| \leq |y_k c|$  for  $k = 1, \dots, q - 1$ . Assume that  $|y_q c| = 1$ . Then  $q \leq f(n)$ .*

In other words, given any set of  $n$  (possibly negative) numbers, one cannot find more than  $f(n)$  subsums of these numbers which decrease in absolute value by a factor of at least 2.

Radzik proves the result for  $f(n) = O(n^2 \log n)$  and conjectures that  $f(n) = O^*(n)$  where  $O^*$  denotes the omission of logarithmic terms. Using linear programming, you are asked to improve his result to  $f(n) = O(n \log n)$ .

- (a) Given a vector  $c$  and a set of  $q$  subsums satisfying the hypothesis of the Lemma, write a set of inequalities in the variables  $x_i \geq 0, i = 1 \dots n$ , such that  $x_i = |c_i|$  is a feasible vector, and for any feasible vector  $x'$  there is a corresponding vector  $c'$  satisfying the hypothesis of the Lemma for the same set of subsums.

We have a set of inequalities of the form  $2|y_{i+1}c| \leq |y_i c|$  for  $1 \leq i \leq q-1$  and  $|y_1 c| = 1$ , and we have a vector  $c$  which satisfies them. To obtain a linear system, we need to remove the absolute value signs. Let  $y'_i = y_i \text{sgn}(y_i c)$ . Then  $|y_i c| = y'_i c$ . The system becomes  $2y'_{i+1}c \leq y'_i c$  for  $1 \leq i \leq q - 1$  and  $y'_1 c = 1$  and the original  $c$  is still feasible. To limit the solution space to vectors of the form  $x \geq 0$  a similar trick is used. We replace elements  $c_j$  that are negative by  $-x_j$  and non-negative elements by  $x_j$ . The linear system that remains has a solution  $x \geq 0$ , namely  $x_j = |c_j|$ . A solution  $c'$  to the original inequalities can be obtained from any feasible  $x$  in the newly constructed set of inequalities by negating the value of the  $i^{\text{th}}$  element of  $x$  if the  $i^{\text{th}}$  element of  $c$  was negative. This results in the same number of inequalities as we had originally, namely  $q$ .

- (b) Prove that there must exist a vector  $c'$  satisfying the hypothesis of the Lemma, with  $c'$  of the form  $(d_1/d, d_2/d, \dots, d_n/d)$  for some integers  $|d|, |d_1|, \dots, |d_n| = 2^{O(n \log n)}$ .

(Hint: see Problem 1.)

The polytope defined by the inequalities above is nonempty. This means that the polytope has a vertex. Our system looks like  $Ax \geq b$  and  $x \geq 0$ , where  $A$  is a  $q$  by  $n$  matrix containing entries between  $-3$  and  $3$  (since every entry is the difference of two integers, one being  $\pm 2$  or  $0$ , the other being  $\pm 1$  or  $0$ ) and  $b$  has one nonzero element which is  $\pm 1$ . From the first problem, we know that the nonzero components, say  $y$ , of a vertex satisfy

$A'y = b'$  where  $A'$  is an invertible submatrix of  $A$  and  $b'$  is a subvector of  $b$ . Notice that  $|\det(A')| \leq n!3^n = 2^{O(n \log n)}$ . As in class (by Cramer's rule), we know that we can set  $d$  to be  $|\det(A')|$  and the (nonzero)  $d_i$ 's to be determinants of submatrices of  $A'$ . By the same argument, these determinants are also upper bounded by  $2^{O(n \log n)}$ , proving the result.

(c) **Deduce from the above that  $f(n) = O(n \log n)$ .**

Multiplying the vector  $c'$  by  $d$  yields an integer solution to  $2y'_{i+1}x \leq y'_i x$  for  $1 \leq i \leq q - 1$  with elements of value  $2^{O(n \log n)}$ . Thus the largest sum that can be obtained by a subset is  $2^{O(n \log n)}$ . As the first subset sums to at least one (since the  $d_i$ 's are integers), the number of times the sum can double is at most  $O(n \log n)$ .

3. **The maximum flow problem on the directed graph  $G = (V, E)$  with capacity function  $u$  (and lower bounds 0) can be formulated by the following linear program:**

$$\max w$$

**subject to**

$$\sum_j x_{ij} - \sum_j x_{ji} = \begin{cases} w & i = s \\ 0 & i \neq s, t \\ -w & i = t \end{cases}$$

$$x_{ij} \leq u_{ij}$$

$$0 \leq x_{ij}.$$

( $x_{ij}$  represents the flow on edge  $(i, j)$ ; the flow has to be less or equal to the capacity on any edge and flow conservation must be satisfied at every vertex except the source  $s$ , where we try to maximize the flow, and the sink  $t$ .)

(a) **Show that its dual is equivalent to:**

$$\min \sum_{(i,j) \in E} u_{ij} y_{ij}$$

**subject to**

$$z_i - z_j + y_{ij} \geq 0 \quad (i, j) \in E$$

$$z_s = 0, z_t = 1$$

$$y_{ij} \geq 0.$$

This is an immediate consequence of the definition of the dual. If one takes the dual of the system of equations and inequalities above, then one gets

$$\begin{aligned} \min \quad & \sum_{(i,j) \in E} u_{ij} y_{ij} \\ z_i - z_j + y_{ij} & \geq 0 & \forall (i,j) \in E \\ z_t - z_s & = 1 \\ y_{ij} & \geq 0. \end{aligned}$$

Since adding a constant to all  $z_i$ 's doesn't change anything, we can require that  $z_s = 0$  and  $z_t = 1$ .

- (b) **A cut is a set of edges of the form  $\{(i,j) \in E : i \in S, j \notin S\}$  for some  $S \subset V$  and its value is**

$$W = \sum_{(i,j) \in E : i \in S, j \notin S} u_{ij}.$$

**It separates  $s$  from  $t$  if  $s \in S$  and  $t \notin S$ .**

**Show that a cut of value  $W$  separating  $s$  from  $t$  corresponds to a feasible solution  $(y, z)$  of the dual program such that**

$$W = \sum_{(i,j) \in E} u_{ij} y_{ij}.$$

For a cut defined by  $S \subset V$ , we define  $z_i = 0$  for  $i \in S$ ,  $z_i = 1$  for  $i \notin S$ ,  $y_{ij} = 1$  for  $i \in S, j \notin S, (i,j) \in E$  and  $y_{ij} = 0$  otherwise. Obviously,  $(y, z)$  is a feasible solution and its value is

$$\sum_{(i,j) \in E} u_{ij} y_{ij} = \sum_{i \in S, j \notin S} u_{ij} = W$$

- (c) **Given any (not necessarily integral) optimal solution  $y^*, z^*$  of the dual linear program and an optimal solution  $x^*$  of the primal linear program, show how to construct from  $z^*$  a cut separating  $s$  from  $t$  of value equal to the maximum flow.**

**(Hint: Consider the cut defined by  $S = \{i : z_i \leq 0\}$  and use complementary slackness conditions.)**

We divide the vertices into two sets defined as follows:

$$\begin{aligned} S &= \{i \in V \mid z_i^* \leq 0\} \\ \bar{S} &= \{i \in V \mid z_i^* > 0\} \end{aligned}$$

Every edge  $(i, j)$  with  $i \in S$  and  $j \notin S$  satisfies  $z_i^* - z_j^* + y_{ij}^* \geq 0$ . Since  $z_i^* \leq 0$  and  $z_j^* > 0$  we have that  $y_{ij}^* > 0$ , which by complementary slackness implies that  $x_{ij}^* = u_{ij}^*$ . Every edge  $(j, i)$  with  $i \in S$  and  $j \notin S$  satisfies  $z_j^* - z_i^* + y_{ji}^* > 0$ , since  $z_j^* > 0$  and  $z_i^* \leq 0$ . By complementary slackness we have that  $x_{ji}^* = 0$ . Thus, we can write

$$\sum_{i \in S, j \notin S} u_{ij} = \sum_{i \in S, j \notin S} x_{ij}^* - \sum_{i \notin S, j \in S} x_{ij}^* = \sum_{i \in S} \left( \sum_j x_{ij}^* - \sum_j x_{ji}^* \right) = w^*,$$

which is the value of the maximum flow.

- (d) **Deduce the max-flow–min-cut theorem: the value of the maximum flow from  $s$  to  $t$  is equal to the value of the minimum cut separating  $s$  from  $t$ .**

From (b) and weak duality, the value of any cut is greater or equal to the maximum flow value. By the analysis above, we can find a cut which is equal to the maximum flow. Thus, the minimum cut value must be the same as the maximum flow value.

4. Consider the following property of vector sums.

**Theorem 2** *Let  $v_1, \dots, v_n$  be  $d$ -dimensional vectors such that  $\|v_i\| \leq 1$  for  $i = 1, \dots, n$  (where  $\|\cdot\|$  denotes any norm) and*

$$\sum_{i=1}^n v_i = 0.$$

*Then there exists a permutation  $\pi$  of  $\{1, \dots, n\}$  such that*

$$\left\| \sum_{j=1}^k v_{\pi(j)} \right\| \leq d$$

*for  $k = 1, \dots, n$ .*

In this problem, you are supposed to prove this theorem by using linear programming techniques.

- (a) Suppose we have a nested sequence of sets

$$\{1, \dots, n\} = V_n \supset V_{n-1} \supset \dots \supset V_d$$

where  $|V_k| = k$  for  $k = d, d+1, \dots, n$ . Suppose further that we have numbers  $\lambda_{ki}$  satisfying:

$$\sum_{i \in V_k} \lambda_{ki} v_i = 0, \tag{1}$$

$$\sum_{i \in V_k} \lambda_{ki} = k - d, \quad (2)$$

$$0 \leq \lambda_{ki} \leq 1 \quad i \in V_k, \quad (3)$$

for  $k = d, \dots, n$ . Define a permutation  $\pi$  as follows: set  $\pi(1), \dots, \pi(d)$  to be elements of  $V_d$  in any order, and set  $\pi(k)$  to be the unique element in  $V_k \setminus V_{k-1}$  for  $k = d+1, \dots, n$ .

Show that this permutation satisfies the conditions of Theorem 2.

For  $k \leq d$ , the theorem is trivial. By the definition of  $\pi$  and  $\lambda_{ki}$ , for  $k > d$  we have

$$\left\| \sum_{j=1}^k v_{\pi(j)} \right\| = \left\| \sum_{i \in V_k} v_i \right\| = \left\| \sum_{i \in V_k} (1 - \lambda_{ki}) v_i \right\| \leq \sum_{i \in V_k} (1 - \lambda_{ki}) = d.$$

- (b) Show that there exist  $\lambda_{ni}$ ,  $i = 1 \dots n$ , satisfying (1), (2) and (3) for  $k = n$ .

We choose simply

$$\lambda_{ni} = 1 - \frac{d}{n}.$$

Then

$$\sum_{i \in V_n} \lambda_{ni} v_i = \left(1 - \frac{d}{n}\right) \sum_{i \in V_n} v_i = 0$$

and

$$\sum_{i \in V_n} \lambda_{ni} = n - d.$$

- (c) Suppose we have constructed  $V_n, \dots, V_{k+1}$  and  $\lambda_{ji}$  for  $j = k+1, \dots, n$  and  $i \in V_j$  satisfying (1), (2) and (3) for  $k+1, \dots, n$  (where  $k \geq d$ ). Prove that the following system of  $d+1$  equalities ((4) contains  $d$  equalities),  $k+1$  inequalities and  $k+1$  nonnegativity constraints has a solution with at least one  $\beta_i = 0$ :

$$\sum_{i \in V_{k+1}} \beta_i v_i = 0, \quad (4)$$

$$\sum_{i \in V_{k+1}} \beta_i = k - d, \quad (5)$$

$$0 \leq \beta_i \leq 1 \quad i \in V_{k+1}. \quad (6)$$

Deduce the existence of the nested sequence and the  $\lambda$ 's as described in (a).

By the induction hypothesis, we have

$$\beta_i = \frac{k-d}{k+1-d} \lambda_{k+1,i}$$

satisfying our inequalities, so the polytope of feasible solutions is nonempty. We want to find a solution with at least one zero coordinate. Consider a vertex of the polytope and suppose that for each  $i$ ,  $\beta_i > 0$ . There are  $k+1$  coordinates summing up to  $k-d$ , so at most  $k-d-1$  of them can be equal to 1. (If  $k-d$  coordinates are equal to 1, the rest is zero.) Therefore we have at least  $d+2$  coordinates  $\beta_i, 0 < \beta_i < 1$ . Let's denote this set of coordinates by  $J$ . The corresponding vectors  $v_j, j \in J$  cannot be affinely independent, so there exists a linear combination with  $\gamma \neq 0$  such that

$$\sum_{j \in J} \gamma_j v_j = 0,$$

$$\sum_{j \in J} \gamma_j = 0.$$

For a small enough  $\epsilon > 0$ , we obtain two feasible solutions by replacing the coordinates of  $\beta_j$  for  $j \in J$  by  $\beta_j \pm \epsilon \gamma_j$ , which contradicts the assumption that  $\beta_i$  is a vertex. Therefore a vertex has always a zero coordinate and by removing this coordinate we obtain the subset  $V_k$  and the corresponding coefficients  $\lambda_{ki} = \beta_i$  which completes the induction.

5. Consider the following optimization problem with "robust conditions":

$$\min\{c^T x : x \in \mathbb{R}^n; Ax \geq b \text{ for any } A \in F\},$$

where  $b \in \mathbb{R}^m$  and  $F$  is a set of  $m \times n$  matrices:

$$F = \{A : \forall i, j; a_{ij}^{\min} \leq a_{ij} \leq a_{ij}^{\max}\}.$$

- (a) Considering  $F$  as a polytope in  $\mathbb{R}^{m \times n}$ , what are the vertices of  $F$ ?

$F$  is an  $(m \times n)$ -dimensional product of intervals. It has  $2^{mn}$  vertices  $A^{(k)}$  where each coordinate  $a_{ij}^{(k)}$  is either  $a_{ij}^{\min}$  or  $a_{ij}^{\max}$ .

- (b) Show that instead of the conditions for all  $A \in F$ , it is enough to consider the vertices of  $F$ . Write the resulting linear program. What is its size?

Suppose that  $x$  satisfies

$$A^{(k)}x \geq b$$

for every vertex  $A^{(k)}$ . Any  $A \in F$  can be written as a convex linear combination of the vertices:

$$A = \sum_k \lambda_k A^{(k)},$$

$$\sum_k \lambda_k = 1, \lambda_k \geq 0.$$

By taking the corresponding linear combination of inequalities (with non-negative coefficients), we get

$$\sum_k \lambda_k A^{(k)}x \geq \sum_k \lambda_k b$$

which is

$$Ax \geq b.$$

Therefore  $x$  is a feasible solution if and only if it satisfies the condition for every vertex of  $F$ . We can write the optimization problem in the following form:

$$\min\{c^T x : \forall k; A^{(k)}x \geq b\}.$$

This is a linear program; however, it has an exponential number of inequalities, namely  $m2^{mn}$ , in  $n$  variables.

- (c) **Derive a more efficient description of the linear program: Write the condition on  $x$  given by one row of  $A$ , for all choices of  $A$ . Formulate this condition as a linear program. Use duality and formulate the original problem as a linear program. What is the size of this one?**

Let us consider a fixed vector  $x$ . It is feasible if the following condition is satisfied for each row  $a_i$  of the matrix  $A$ :

$$\forall a_{ij} \in [a_{ij}^{\min}, a_{ij}^{\max}]; \sum_j a_{ij}x_j \geq b_i.$$

We can regard this condition as a linear programming problem:

$$b_i \leq \min \left\{ \sum_j x_j a_{ij} : a_{ij} \geq a_{ij}^{\min}, -a_{ij} \geq -a_{ij}^{\max} \right\}.$$

Note that the variables are now  $a_{ij}$ , while  $x$  is fixed! By duality, we get an equivalent condition for a linear program with variables  $p_{ij}, q_{ij}$ :

$$b_i \leq \max \left\{ \sum_j p_{ij} a_{ij}^{\min} - \sum_j q_{ij} a_{ij}^{\max} : p_{ij} - q_{ij} = x_j, p_{ij} \geq 0, q_{ij} \geq 0 \right\}.$$



This means that  $x$  is feasible if there exist  $p_{ij} \geq 0, q_{ij} \geq 0$  such that

$$p_{ij} - q_{ij} = x_j$$

and

$$\sum_j p_{ij} a_{ij}^{\min} - \sum_j q_{ij} a_{ij}^{\max} \geq b_i.$$

All together, we can write our optimization problem as the following:

$$\min\{c^T x : \begin{array}{ll} p_{ij} - q_{ij} - x_j = 0 & \forall i, j \\ \sum_j p_{ij} a_{ij}^{\min} - \sum_j q_{ij} a_{ij}^{\max} \geq b_i & \forall i \\ p_{ij}, q_{ij} \geq 0 & \forall i, j \end{array}\}$$

which is a linear program in variables  $x_j, p_{ij}, q_{ij}$ . It has  $2mn + n$  variables,  $mn$  equalities,  $m$  inequalities and  $2mn$  nonnegativity constraints. The linear program from part (b) has size which is exponential in the size of this one.