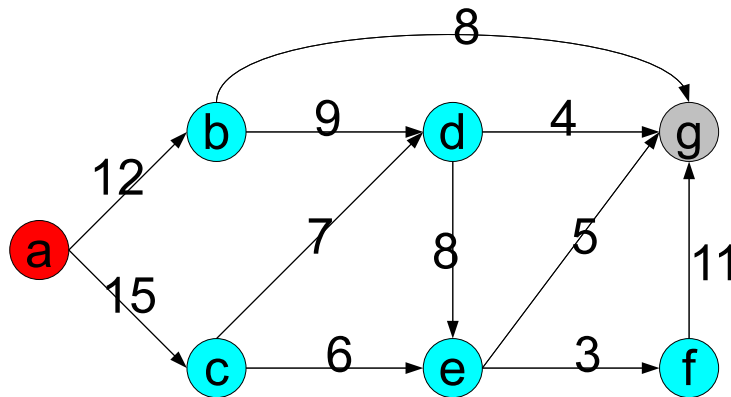


# CS5321 Numerical Optimization Homework 6

Due May 21

1. (50%) Many of famous computer science problems can be written as a linear programming problem. Considering the following weighted directed graph, which has 7 nodes. Let node *a* be the *source* and node *g* be the *sink*.



- Let the weights be capacity. Formulate the *maximum flow problem* as a linear programming problem. (Hint: capacity constraints and flow-balanced constraint.)
- Let the weight be capacity. Formulate the *minimum cut problem* as a linear programming problem.
- Show the problems in (b) and (c) are dual to each other.
- Let the weights be distances. Formulate the *shortest path problem* as a linear programming problem.
- What is the dual problem of the shortest path problem?
- (Bonus 20%) Ignore the direction of edges. Formulate the *minimum spanning tree problem* as an integer linear programming problem.

2. (50%) Implement the simplex method for linear programming. The pseudo code is in Figure 1. The calling interface will be like

`[x, case] = mysimplex(c, A, b, x0)`

which solves

$$\begin{aligned} \min_{\vec{x}} \quad & \vec{c}^T \vec{x} \\ \text{subject to} \quad & A\vec{x} = \vec{b} \\ & \vec{x} \geq 0 \end{aligned}$$

The return value `case` should be 0, 1, or 2, which means (0) solved, (1) unbounded, (2) infeasible. You can assume  $\vec{x}_0$  is a feasible point. Also, print out each  $\vec{x}_i$  during the computation.

- (1) Given a basic feasible point  $\vec{x}_0$  and the corresponding index set  $\mathcal{B}_0$  and  $\mathcal{N}_0$ .
- (2) For  $k = 0, 1, \dots$
- (3) Let  $B_k = A(:, \mathcal{B}_k)$ ,  $N_k = A(:, \mathcal{N}_k)$ ,  $\vec{x}_B = \vec{x}_k(\mathcal{B}_k)$ ,  $\vec{x}_N = \vec{x}_k(\mathcal{N}_k)$ , and  $\vec{c}_B = \vec{c}_k(\mathcal{B}_k)$ ,  $\vec{c}_N = \vec{c}_k(\mathcal{N}_k)$ .
- (4) Compute  $\vec{s}_k = \vec{c}_N - N_k^T B_k^{-1} \vec{c}_B$  (pricing)
- (5) If  $\vec{s}_k \geq 0$ , return the solution  $\vec{x}_k$ . (found optimal solution)
- (6) Select  $q_k \in \mathcal{N}_k$  such that  $\vec{s}_k(i_{q_k}) < 0$ , where  $i_{q_k}$  is the index of  $q_k$  in  $\mathcal{N}_k$
- (7) Compute  $\vec{d}_k = B_k^{-1} A_k(:, q_k)$ . (search direction)
- (8) If  $\vec{d}_k \leq 0$ , return unbounded. (unbounded case)
- (9) Compute  $[\gamma_k, i_p] = \min_{i, \vec{d}_k(i) > 0} \frac{\vec{x}_B(i)}{\vec{d}_k(i)}$  (ratio test)  
(The first return value is the minimum ratio;  
the second return value is the index of the minimum ratio.)
- (10)  $x_{k+1} \begin{pmatrix} \mathcal{B} \\ \mathcal{N} \end{pmatrix} = \begin{pmatrix} \vec{x}_B \\ \vec{x}_N \end{pmatrix} + \gamma_k \begin{pmatrix} -\vec{d}_k \\ \vec{e}_{i_{q_k}} \end{pmatrix}$   
( $\vec{e}_{i_{q_k}} = (0, \dots, 1, \dots, 0)^T$  is a unit vector with  $i_{q_k}$ th element 1.)
- (11) Let the  $i_p$ th element in  $\mathcal{B}$  be  $p_k$ . (pivoting)  
 $\mathcal{B}_{k+1} = (\mathcal{B}_k - \{p_k\}) \cup \{q_k\}$ ,  $\mathcal{N}_{k+1} = (\mathcal{N}_k - \{q_k\}) \cup \{p_k\}$

Figure 1: The simplex method for solving (minimization) linear programming