

Due: May 7th, 2009.

Reading: Lecture notes, Chapter 7 of DPV book.

1. **Selection.** [15 points]

QuickSelect is the following simple randomized algorithm for finding the k th smallest element in an unsorted set S .

QuickSelect(S, k):

1. Pick a pivot element p uniformly at random from S .
2. By comparing p to each element of S , split S into two pieces: $S_1 = \{x \in S : x < p\}$ and $S_2 = \{x \in S : x > p\}$.
3. If $|S_1| = k - 1$, then output p .
If $|S_1| > k - 1$ then output QuickSelect(S_1, k).
If $|S_1| < k - 1$ then output QuickSelect($S_2, k - |S_1| - 1$).

Prove that the expected number of comparisons made by QuickSelect on a set S of size n is at most $3.5n$. (Partial credit for proving it's at most $4n$.)

2. **Approximate min cuts.** [15 points]

In this problem, we revisit the Contraction algorithm for computing minimum cuts, and consider its ability to find near-minimum cuts. For an integer $\ell \geq 1$, define an ℓ -approximate cut to be a cut whose size is at most ℓ times the size of the minimum cut. (We are considering unweighted, undirected graphs in this problem).

- (a) Prove that a single trial of the contraction algorithm yields as output an ℓ -approximate cut with probability at least $\Omega(n^{-2/\ell})$, where n is the number of vertices in the graph.
- (b) For each *fixed* integer $\ell \geq 1$, give a polynomial time algorithm that outputs a list of all ℓ -approximate cuts in the graph. Prove also that, in any n -vertex graph, there are at most $n^{2\ell}$ ℓ -approximate cuts.

3. **A simplex example.** [15 points]

- (a) Solve the following linear program using the simplex method. Start with the initial solution $(0, 0, 1)$. Show your intermediate steps.

Maximize $3x_2 + x_3$ subject to

$$\begin{aligned}x_1 + 2x_2 + x_3 &\leq 2 \\2x_1 + x_2 - x_3 &\leq -1 \\3x_1 + 2x_2 + x_3 &\leq 3 \\x_1, x_2, x_3 &\geq 0.\end{aligned}$$

(b) [Removed]

4. **Integrality and matchings.** [25 points]

(a) In this problem, you will identify a very useful criterion for a linear program to have an integral optimum solution. Consider an LP in standard form:

Maximize $c \cdot x$ subject to

$$\begin{aligned} Ax &\leq b \\ x &\geq 0 \end{aligned}$$

where A is an integer matrix and b is an integer vector, but the vector c need not be integral. Suppose that the determinant of every square submatrix of A is $0, 1$, or -1 . Prove that if the linear program is bounded and has an optimal solution, then it has an optimal solution x^* whose entries are integers.

Problem: Although you will need the full strength of the preceding assertion to finish (b)-(e), you are only required to prove that there is an integral optimal solution *when A is non-degenerate*, i.e. when any subset of d rows of A are linearly independent.

Hint: Use the fact that when x is a vertex of the feasible region, we have $A_{\Gamma}x = b_{\Gamma}$ where A_{Γ} is a $d \times d$ sub-matrix of A , and b_{Γ} is a subset of d coordinates from b . (Recall that we used this fact in the proof of strong duality.)

- (b) Consider the problem of computing a maximum matching in a bipartite graph. Express this as an integer linear program using a 0/1-valued variable x_e to indicate whether an edge e is picked in the matching or not.
- (c) Consider the linear programming relaxation of the above integer linear program, obtained by relaxing the constraint $x_e \in \{0, 1\}$ to $x_e \geq 0$. Write down the dual linear program of this LP relaxation. What optimization problem is the dual a natural relaxation of?
- (d) Prove, using part (a), that the linear program above for maximum matching on bipartite graphs as well as its dual both have integral optimum solutions.
- (e) What “min-max theorem” concerning bipartite graphs can you conclude using part (d)?