

### Problem 1: (Weighted MINCUT and MAXCUT)

Let  $G(V, E)$  be an undirected *weighted* graph, with  $w_{ij} > 0$  the weight associated with every edge  $(i, j) \in E$ . The weight of a cut  $(C, \bar{C})$  is now the sum of the weights of edges across the cut, i.e.,  $\delta(C, \bar{C}) = \sum_{i,j \in E(C, \bar{C})} w_{ij}$ . We now try and extend our MAXCUT and MINCUT algorithms to this setting.

#### Part (a)

Let  $W = \sum_{(i,j) \in E} e_{ij}$  be the total weight of all edges in then graph. Modify the MAXCUT algorithm presented in class to return a cut  $(C, \bar{C})$  with expected weight satisfying:  $\mathbf{E}[\delta(C, \bar{C})] \geq \frac{W}{2}$

#### Part (b)

Next suppose we modify the CONTRACT algorithm to pick edges proportional to their weights. Show that any minimum weight cut  $(C, \bar{C})$  is returned by CONTRACT with probability  $\geq \frac{2}{n(n-1)}$ .

### Problem 2: (Recursive Randomized Selection)

Given a unsorted array  $S = \{x_1, x_2, \dots, x_n\}$ , with corresponding sorted array  $\{y_1, y_2, \dots, y_n\}$ , a *selection algorithm* is one that finds the median element  $y_{\frac{n}{2}}$  (or more generally, the  $k^{\text{th}}$ -largest element  $y_k$  for any  $k \in \{1, 2, \dots, n\}$ ). One way to do so is by first sorting the array, and then returning  $y_k$  for any  $k$  – this takes time  $O(n \log n)$ . However, consider the following simple randomized algorithm to find  $y_k$  for a given  $k$ :

QUICKSELECT( $S, k$ )

- Given array  $S$  of  $n$  elements, we want to output the  $k^{\text{th}}$  largest element  $y_k$ .
- Choose a random pivot  $\sigma$ , and partition  $S$  into two parts:

$$S_\ell = \{y_i \in S | y_i < \sigma\} \quad , \quad S_h = \{y_i \in S | y_i > \sigma\}$$

- If  $|S_\ell| = k - 1$ , return  $\sigma$
- If  $|S_\ell| > k$ , then run QUICKSELECT( $S_\ell, k$ ); else run QUICKSELECT( $S_h, k - |S_\ell| - 1$ )

It is easy to see that this will find  $y_k$  – we now want to show that QUICKSELECT has a running time of  $O(n)$ .

#### Part (a)

To build some intuition as to why this works, assume in given an array  $S$  of size  $n$ , the two arrays  $S_\ell, S_h$  were guaranteed to be of size at most  $\alpha n$ , for some  $\alpha \in [1/2, 1)$ . Argue that the runtime of QUICKSELECT would then obey:  $T(n) = T(\alpha n) + O(n)$ . Solve this to show  $T(n) = O(n)$ .

#### Part (b)

Given any array of size at most  $n$ , argue that after splitting about the pivot, the sets  $S_\ell$  and  $S_h$  both have size less than  $3n/4$  with probability at least  $1/2$ . Using this, find an upper bound on

the expected number of times an array of size  $n$  needs to be split about random pivots before the sub-array containing  $y_k$  is of size  $\leq 3n/4$ .

*Hint: Consider an alternate algorithm, where you pick a pivot, check to make sure that both  $S_\ell$  and  $S_h$  are less than  $3|S|/4$ , and then split – if not, you keep the array  $S$  as before and again pick a random pivot. Prove the above result for this modified algorithm. Convince yourself that QUICKSELECT can only be faster.*

### Part (c)

Let's define the algorithm to run in *phases*, where in phase  $i$ , the size of the sub-array containing  $y_k$  is between  $(3/4)^{j-1}n$  and  $(3/4)^jn$ . Also let  $X_j$  denote the number of splits required in phase  $j$  (so for example,  $X_1$  is the expected number of splits required to go from the original array  $S$  to one of size  $3n/4$ ).

Argue that  $T(n) \leq \sum_{\text{phase } j} c(3/4)^{j-1}n \cdot X_j$  for some constant  $c$ . Finally, via linearity of expectation, prove that  $\mathbf{E}[T(n)] = O(n)$ .

### Problem 3: (Multi-stage MINCUT Algorithm)

In class we saw the CONTRACT Algorithm for finding the MINCUT of a multigraph  $G$  – we were given that each run of CONTRACT took time  $O(n^2)$ , and argued that if  $G$  had a unique minimum cut  $(C, \bar{C})$ , then CONTRACT finds it with probability  $\Omega(1/n^2)$ .

### Part (a)

Suppose CONTRACT returned  $(C, \bar{C})$  with probability at least  $1/n^2$  – show that  $n^2 \ln 2$  independent runs of CONTRACT are sufficient to find cut  $(C, \bar{C})$  with probability at least  $1/2$ .

More generally, convince yourself that if an algorithm is successful with probability at least  $p$ , then  $\ln 2/p$  independent runs are sufficient to guarantee success with probability at least  $1/2$ .

*Hint: Use  $(1 - x) \leq e^{-x}$ .*

### Part (b)

The above problem shows that the overall runtime of CONTRACT is  $O(n^4)$  – on the other hand, we learnt in class that the best deterministic MINCUT algorithm had a runtime of  $O(n^3)$ . We also saw that if we ran CONTRACT until the number of vertices in the multigraph is  $t$ , then it takes time  $O(n^2)$  (as long as  $t = o(n)$ ) and preserves the minimum cut  $(C, \bar{C})$  with probability  $O(t^2/n^2)$ .

Now consider running CONTRACT until the number of vertices in the multigraph is  $t$ , followed by a deterministic MINCUT algorithm for the  $t$ -node graph – as before, we can do this multiple times to improve the probability. Show that the best possible choice of  $t$  results in a running time of  $O(n^{8/3})$  for finding  $(C, \bar{C})$  with probability at least  $1/2$ .

### Problem 4: (The FASTCUT Algorithm and the Branching Process)

Recall that in class, we briefly saw the FASTCUT algorithm, where given a graph, we first ran two independent executions of CONTRACT, stopping them when the resulting subgraph retained the minimum cut with probability  $\geq 1/2$ , and then proceeded recursively. We now try and understand why this algorithm works.

#### Part (a)

Assume we can choose  $\alpha$  such that contracting the graph to  $t = \alpha n$  nodes ensures that a minimum cut is preserved with probability *exactly*  $1/2$  – let us call this the  $\alpha$ -CONTRACT step. Also assume the original graph  $G$  had a unique minimum cut  $(C, \bar{C})$ .

Now suppose in the first recursive step, we do 2 independent runs of  $\alpha$ -CONTRACT on the original graph  $G$ , and at each recursive step, we do 2 independent runs of  $\alpha$ -CONTRACT for each input sub-graph. After  $k$  recursions (where  $k \in \{1, 2, \dots, \log_{1/\alpha} n\}$ ), what is the expected number of sub-graphs which retain the minimum cut  $(C, \bar{C})$ ?

#### Part (b)

Suppose instead of doing 2 independent runs of  $\alpha$ -CONTRACT on each subgraph, we instead ran it once, and just duplicated the resulting subgraph. Now what is the expected number of sub-graphs which retain the minimum cut  $(C, \bar{C})$  after  $k$  recursions? Why do you think this is different from part (a)?

#### Part (c)

Let  $p(k)$  be the probability that the minimum cut  $(C, \bar{C})$  survives in at least one subgraph if we stop after doing  $k$  recursions (thus  $p(0) = 1$ ).

Argue that in the procedure in part (b) – where we do one run of  $\alpha$ -CONTRACT for each subgraph and duplicate the output – the function  $p(k)$  obeys  $p(k+1) = \frac{p(k)}{2}$ , and thus  $p(k) = 1/2^k$ .

On the other hand, argue that the procedure in part (a) – where we do two independent runs of  $\alpha$ -CONTRACT for each subgraph – the function  $p(k)$  obeys  $p(k+1) = 1 - \left(1 - \frac{p(k)}{2}\right)^2$ .

#### Part (d)

**(OPTIONAL)** Try to show that the solution to the recursive equation  $p(k+1) = 1 - \left(1 - \frac{p(k)}{2}\right)^2$  obeys  $p(k) = \Theta(1/k)$ .

*Hint: Note that  $p(k) = \Theta(1/k)$  is same as saying  $c_1/k \leq p(k) \leq c_2/k$  – now substitute this in the above recursive equation, and prove it holds by induction.*