

# Homework 1

## Randomized Algorithms

Due Wednesday, August 30

1. [MR 1.8]. Consider adapting the min-cut algorithm of the first class to the problem of finding an  $s$ - $t$  min-cut in an undirected graph. In this problem, we are given an undirected graph  $G$  together with two distinguished vertices  $s$  and  $t$ . An  $s$ - $t$  min-cut is a set of edges whose removal disconnects  $s$  from  $t$ ; we seek an edge set of minimum cardinality. As the algorithm proceeds, the vertex  $s$  may get amalgamated into a new vertex as the result of an edge being contracted; we call this vertex the  $s$ -vertex (initially  $s$  itself). Similarly, we have a  $t$ -vertex. As we run the contraction algorithm, we ensure that we never contract an edge between the  $s$ -vertex and the  $t$ -vertex.
  - (a) Show that there are graphs (not multi-graphs) in which the probability that this algorithm finds an  $s$ - $t$  min-cut is exponentially small.
  - (b) How large can the number of different  $s$ - $t$  min-cut solutions in an instance be?
  - (c) Can you derive a very different bound for the number of different global min-cuts, as a consequence of the algorithm presented in class?
2. Suppose we have access to a source of unbiased random bits. This problem looks at constructing biased coins or dice from this source.
  - (a) Show how to construct a biased coin, which is 1 with probability  $p$  and 0 otherwise, using  $O(1)$  random bits in expectation. [Hint: First show how to construct a biased coin using an arbitrary number of random bits. Then show that the expected number of bits examined is small.]
  - (b) Show how to sample from  $[n]$ , with probabilities  $p_1, \dots, p_n$ , using  $O(\log n)$  random bits in expectation.
  - (c) Show that the “in expectation” caveat is necessary: for example, one cannot sample uniformly over  $\{1, 2, 3\}$  using  $O(1)$  bits in the worst case.
  - (d) [Optional.] Give a *fast* algorithm to sample from  $[n]$  with probabilities  $p_1, \dots, p_n$ . That is, give an algorithm that uses in expectation  $O(\log n)$  bits and  $O(1)$  time per sample (in the word RAM model, so manipulating/indexing with  $O(\log n)$ -bit words takes  $O(1)$  time.). Your algorithm may preprocess the input, using  $O(n)$  time and space. [Hints: (a) if all the  $p_i$  came in pairs that summed to  $2/n$ , could you solve the problem? (b) can you break up any set of  $p_i$  into  $2n$  total pieces, so the pieces come in pairs that sum to  $1/n$ ?]