

# CS41 Lab 10: Approximation Algorithms

In typical labs this semester, you'll be working on a number of problems in groups of 3-4 students. You will not be handing in solutions; the primary purpose of these labs is to have a low-pressure space to discuss algorithm design. However, it will be common to have some overlap between lab exercises and homework sets.

1. **Traveling Salesman Problem.** In this problem, a salesman travels the country making sales pitches. The salesman must visit  $n$  cities and then return to her home city, all while doing so as cheaply as possible.

The input is a complete graph  $G = (V, E)$  along with nonnegative edge costs  $\{c_e : e \in E\}$ . A *tour* is a simple cycle  $(v_{j_1}, \dots, v_{j_n}, v_{j_1})$  that visits every vertex exactly once.<sup>1</sup> The goal is to output the minimum-cost tour.

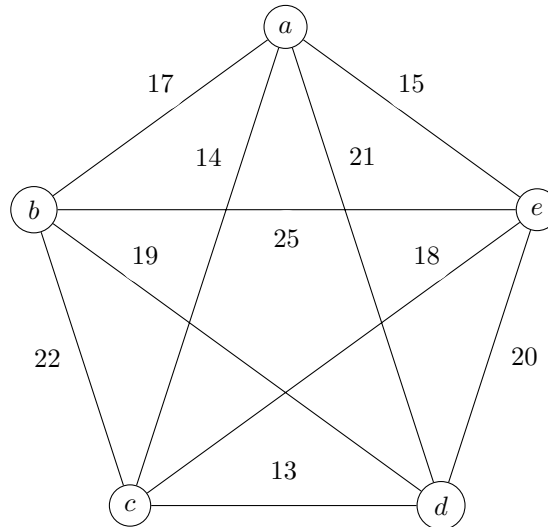
For many TSP applications (such as when the cost is proportional to the distance between two cities), it makes sense for the edges to obey the *triangle inequality*: for every  $i, j, k$ , we have

$$c_{(ik)} \leq c_{(ij)} + c_{(jk)}.$$

This version is often called METRIC-TSP.

The (decision version of the) Traveling Salesman Problem is NP-COMplete. For this problem, you will develop a 2-approximation algorithm for METRIC-TSP.

- (a) First, to gain some intuition, consider the following graph:



- (b) *On your own* try to identify a cheap tour of the graph.
- (c) Build some more intuition by computing the minimum spanning tree (MST) of the graph. Let  $T$  be your minimum spanning tree.
- (d) Let  $OPT$  be the cheapest tour. Show that its cost is bounded below by the cost of the MST:  $\text{cost}(T) \leq \text{cost}(OPT)$ .
- (e) Give an algorithm which returns a tour  $A$  which costs at most twice the cost of the MST:  $\text{cost}(A) \leq 2 \text{cost}(T)$ .
- (f) Conclude that your algorithm is a 2-approximation for METRIC-TSP.

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<sup>1</sup>except for the start vertex which we visit again to complete the cycle

2. **Toy-Storage.** William has lots of toys of all different sizes. You'd like to purchase a number of bins in which to store the toys. Approximately how many bins will you need?

Let's formalize the TOY-STORAGE problem as follows. Suppose there are  $n$  toys, with sizes  $s_1, \dots, s_n$ , with  $0 < s_i < 1$  for all  $i$ . Assume each bin has size 1 and can hold any collection of toys whose total size is less than or equal to 1.

In this problem, you'll develop a greedy approximation algorithm, which works by taking each toy in turn and placing it into the first bin that can hold it. Let  $S := \sum_{i=1}^n s_i$ .

- (a) Show that the optimal number of bins is at least  $\lceil S \rceil$ .
- (b) Show that the greedy algorithm leaves at most one bin half full.
- (c) Prove that the number of bins used by the greedy algorithm is at most  $\lceil 2S \rceil$ .
- (d) Prove that the greedy algorithm is a 2-approximation algorithm for the TOY-STORAGE problem.