

---

## Efficient Algorithms and Data Structures II

---

*Deadline: July 8, 2019, 10:15 am in the **Efficient Algorithms** folder.*

### Homework 1

- (a) Consider the greedy algorithm for the knapsack problem: Sort the objects by decreasing ratio  $p_i/w_i$  and greedily pick items in that order. Show that the approximation ratio of this algorithm is unbounded.
- (b) Disappointed by the algorithm from part (a), we improve the algorithm as follows. Order the items as in (a) and find the smallest integer  $k$ , so that the total weight of the first  $k$  items exceeds the weight bound  $W$ . Wlog, the ordered items are named  $1, 2, \dots, n$ . Pick items  $\{1, \dots, k-1\}$  or just item  $\{k\}$ , depending on which set yields a higher profit.

Show that the improved greedy algorithm returns a 2-approximation.

### Homework 2

The *First-Fit heuristic* for solving the bin packing problem takes each object in turn and places it in the first bin that can accommodate it.

- (a) Show that this approach gives a factor 2 approximation algorithm for bin packing.
- (b) Give an example on which First-Fit does as bad as  $\frac{5}{3} \cdot \text{OPT}$ .

### Homework 3

Let  $G$  be a complete undirected graph in which all edge lengths are either 1 or 2 (clearly,  $G$  satisfies the triangle inequality). Give a  $4/3$  factor algorithm for TSP in this special class of graphs.

**Hint:** Start by finding a minimum 2-matching in  $G$ . A 2-matching is a subset  $S$  of edges so that every vertex has exactly 2 edges of  $S$  incident at it.

### Homework 4

We say that a bin packing algorithm is *monotonic* if the number of bins it uses for packing a subset of the items is at most the number of bins it uses for packing all  $n$  items. Show that whereas Next-Fit is monotonic, First-Fit is not.

The Next-Fit heuristic tries to pack the next item only in the most recently started bin. If it does not fit, it is packed in a new bin.

## Tutorial Exercise 1

In the maximum  $k$ -cut problem, we are given an undirected graph  $G = (V, E)$ , and non-negative weights  $w_{ij} \geq 0$ , for all edges  $(i, j) \in E$ . The goal is to partition the vertex set  $V$  into  $k$  parts  $V_1, \dots, V_k$  so as to maximize the weights of all edges whose endpoints are in different parts (i.e.,  $\max. \sum_{(i,j) \in E: i \in V_a, j \in V_b, a \neq b} w_{ij}$ ).

Give a randomized  $\frac{k-1}{k}$  approximation algorithm for the maximum  $k$ -cut problem.

## Tutorial Exercise 2

Consider the following variant of MAX-CUT: In addition to the graph  $G = (V, E)$ , we are given two sets of vertex pairs  $S_1, S_2 \subseteq V \times V$ . The pairs of  $S_1$  need to be separated, the pairs of  $S_2$  need to be on the same side of the cut. Under these constraints, the problem is to find the maximum weight cut.

Give a vector program relaxation for this problem. Apply an adapted form of the MAX-CUT algorithm from the lecture to round its solution to an integral solution. The approximation factor should still be at most  $\alpha \geq 0.878$ .

[Hermann Rubin] showed me the Chebyshev type of proof that gives rise to what's now called the Chernoff bound, but it is certainly Rubin's. [...] I am very unhappy about the fact that I did not properly credit Rubin at that time because I thought it was a rather trivial lemma, but many things are only trivial once you know them.

- H. Chernoff