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# Algorithm Theory Sample Solution Exercise Sheet 7

# **Exercise 1: Load Balancing**

# (2+6+5+5+4 Points)

Recall the load balancing problem from the lecture: Given m machines, n jobs and for each job i a processing time  $t_i$ , we want to assign each job to a machine such that the makespan (largest total processing time of any machine) is minimized. We have seen that the modified greedy algorithm, in which we go through the jobs by decreasing length and assign each job to the machine that currently has the smallest load, has an approximation ratio of 3/2.

In this exercise, we want to prove that the algorithm has an even better approximation ratio.

Assume we have n jobs with lengths  $t_1 \ge t_2 \ge \cdots \ge t_n$ . Let T be the makespan of the greedy solution and let i be a machine with load T. Further, let  $\hat{n}$  be the last job that is scheduled on machine i.

- (a) Shortly argue why it is sufficient to ignore jobs  $\hat{n} + 1, \ldots, n$  and instead prove the desired ratio between greedy and optimal for jobs  $1, \ldots, \hat{n}$ .
- (b) Show that if an optimal solution for jobs  $1, \ldots, \hat{n}$  assigns at most two jobs to each machine, the algorithm computes an optimal solution.

*Hint:* Think of a "canonical" way to assign at most two jobs to each machine and show that  $T \leq T_{canonical} \leq T_{opt}$ .

- (c) Show that therefore, either  $t_{\hat{n}} \leq T_{opt}/3$  or the greedy algorithm computes an optimal solution.
- (d) Conclude that the algorithm has an approximation ratio of at most 4/3.
- (e) Show that the 4/3 bound is tight, i.e., there is a sequence of instances for which the ratio between Greedy and OPT converges to 4/3. *Hint: Consider 2m + 1 jobs for m machines, three jobs with processing time m and two jobs with processing times m + 1, m + 2,..., 2m − 1 each.*

## Sample Solution

- (a) If we run the greedy algorithm on jobs  $1, \ldots, \hat{n}$ , we obtain the same makespan as if we run it on jobs  $1, \ldots, n$ . An optimal solution for  $1, \ldots, \hat{n}$  can only be smaller than one for  $1, \ldots, n$ . It follows that the ratio between greedy and optimal for  $1, \ldots, n$  is at least as good as for  $1, \ldots, \hat{n}$ .
- (b) If OPT assigns at most two jobs to each machine, there are at most 2m jobs. For convenience, if  $\hat{n} < 2m$ , we add  $2m \hat{n}$  empty jobs (with processing time 0) such that we have exactly 2m jobs. Assume we have  $t_1 \ge t_2 \ge \cdots \ge t_{2m}$ . The canonical algorithm pairs job j with job 2m j + 1 and assigns it to machine j. We show that we can transform OPT to the canonical solution without increasing the makespan.

Assume OPT differs from the canonical solution and let j be the smallest integer for which job j is not paired with job 2m - j + 1 in OPT. It follows that in OPT, job j is paired with some job k < 2m - j + 1 and job 2m - j + 1 is paired with some job j' > j. We have  $t_{j'} \leq t_j$  and

 $t_{2m-j+1} \leq t_k$  and thus  $t_j + t_{2m-j+1} \leq t_j + t_k$  and  $t_{j'} + t_k \leq t_j + t_k$ . Therefore, building the pairs  $(t_j, t_{2m-j+1})$  and  $(t_{j'}, t_k)$  instead of  $(t_j, t_k)$  and  $(t_{j'}, t_{2m-j+1})$  does not increase the makespan. By continuing this procedure we can transform OPT to the canonical solution without increasing the makespan. It follows  $T_{canonical} \leq T_{opt}$ .

Next we prove  $T \leq T_{canonical}$ . The first m jobs are assigned to the same machines in both the greedy and the canonical solution. We first observe that if at some point of the Greedy execution some machine k has only one job, then also all machines j < k have only one job (because otherwise Greedy would have assigned a job to a machine though there was another machine with less load). It follows that before assigning job 2m + 1 - k for some  $1 \leq k \leq m$ , machine k has only one job (that is job k with length  $t_k$ ). If Greedy does not assign job 2m + 1 - k to machine k but to some machine j > k, machine j must have load  $\leq t_k$  (before the assignment) and hence afterwards load  $\leq t_k + t_{2m+1-k} \leq T_{canonical}$ . So in each step of Greedy, the load of the machine that Greedy latest assigned a job is at most  $T_{canonical}$ .

- (c) If OPT assigns at most two jobs to each machine, then by (b) Greedy is optimal. Otherwise, there is at least one machine that OPT assigns three jobs and as  $t_{\hat{n}}$  is the minimum job length, the load on this machine (and hence  $T_{opt}$ ) is at least  $3t_{\hat{n}}$ .
- (d) Recall that T is the makespan of the greedy solution, i a machine with load T and  $\hat{n}$  the last job that is scheduled on machine i. It follows that before assigning job  $\hat{n}$ , all machines must have a load of at least  $T t_{\hat{n}}$ . Therefore  $T_{opt} \geq T t_{\hat{n}}$  (the optimal makespan is at least the average load which is at least  $T t_{\hat{n}}$ ). By (c) we know that either Greedy is optimal or  $t_{\hat{n}} \leq T_{opt}/3$ , so we obtain  $T_{opt} \geq T T_{opt}/3$  and hence  $\frac{4}{3}T_{opt} \geq T$ .
- (e) Given m machines and three jobs with processing time m and two jobs with processing times  $m+1, m+2, \ldots, 2m-1$  each, Greedy assigns the first 2m jobs to the machines such that each machine has load 3m-1 and puts the last job to some arbitrary machine, so the makespan is 4m-1. An optimal solution assigns the three jobs with length m to one machine and the remaining 2m-2 jobs to the other machines such that each machine has load 3m. Thus the approximation ratio is  $\frac{4m-1}{3m}$  which converges to 4/3 for  $m \to \infty$ .

## Exercise 2: Two Knapsacks

### (2+6 Points)

Consider the following variation of the knapsack problem: Given items  $1, \ldots, n$  where each item *i* has a positive integer weight  $w_i \in \mathbb{N}$  and a positive value  $v_i > 0$  and **two** knapsacks of capacities  $W_1$  and  $W_2$ , we want to pack the items into the knapsacks such that

- for  $j \in \{1, 2\}$ , the total weight of the items in knapsack j is at most  $W_j$ .
- The total value of the items that are packed in either knapsack is maximized.
- (a) Prove that this problem is not equivalent to the standard knapsack problem with one knapsack of capacity  $W_1 + W_2$  by showing that the total value that can be packed into one knapsack of capacity  $W_1 + W_2$  can be *arbitrarily* larger than the total value that can be packed into two knapsacks of capacities  $W_1$  and  $W_2$ .
- (b) Assume that  $W_1 \ge W_2$ . A simple strategy would be to first compute an optimal solution for a knapsack of capacity  $W_1$  and afterwards, with the remaining elements, an optimal solution for a knapsack of capacity  $W_2$ . Show that this algorithm always computes at least a 2-approximation for the problem.

### Sample Solution

(a) Consider an instance with  $W_1 = W_2 = 1$  and one item with weight 2 and arbitrarily large value.

(b) Let  $v_{opt}(W_1)$  ( $v_{opt}(W_2)$ , resp.) be the total value of items packed into the knapsack of capacity  $W_1$  ( $W_2$ , resp.) by an optimal algorithm and  $v_{alg}(W_1)$  the total value of items packed into the knapsack of capacity  $W_1$  by the described algorithm.

 $v_{opt}(W_1)$  is a feasible (but not necessarily optimal) solution for the problem with **one** knapsack of capacity  $W_1$ . As  $v_{alg}(W_1)$  is an optimal solution for this, we have  $v_{alg}(W_1) \ge v_{opt}(W_1)$ . Similarly,  $v_{opt}(W_2)$  is a feasible solution for the problem with one knapsack of capacity  $W_2$  and hence we have  $v_{alg}(W_1) \ge v_{opt}(W_2)$  because  $W_1 \ge W_2$  (making the knapsack larger can only increase the optimal solution). So we have

$$v_{alg}(W_1) + v_{alg}(W_2) \ge v_{alg}(W_1) \ge \max\{v_{opt}(W_1), v_{opt}(W_2)\} \ge \frac{v_{opt}(W_1) + v_{opt}(W_2)}{2} .$$

# Exercise 3: Vertex Cover Approximation (4 Points)

Show that taking all nodes is a 2-approximation algorithm for the vertex cover problem in regular graphs (graphs where all nodes have the same degree)

### Sample Solution

In an r-regular graph, each vertex can cover at most r edges. As the graph has rn/2 edges, at least n/2 vertices are needed to cover all edges.