

# Algorithm Design: Homework #5

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## Problem 1

### Solution

- a) It's obviously that if we choose  $k$  times  $X = X + 1$  and  $n - k$  times  $X = X - 1$ , then  $X = k - (n - k) = 2k - n$ . So, if finally  $X = i$ , then  $k = \frac{n+i}{2}$ , and  $n - k = \frac{n-i}{2}$ . Hence,

$$\begin{aligned}\Pr[X = i] &= \binom{n}{(n+i)/2} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^{n-k} \\ &= \binom{n}{(n+i)/2} \left(\frac{1}{2}\right)^n\end{aligned}$$

Notice that the symmetry of this equation, we know that  $\binom{n}{m} = \binom{n}{n-m}$ , hence  $\Pr[X = k] = \Pr[X = -k]$ .

Thus we can calculate  $\mathbb{E}[X]$ :

$$\begin{aligned}\mathbb{E}[X] &= \sum_{i=-n/2}^n i \Pr[X = i] \\ &= \sum_{i=-n/2}^{n/2} i \binom{n}{(n+i)/2} \left(\frac{1}{2}\right)^n \\ &= 0\end{aligned}$$

- b) For even  $n$ , we know that

$$\begin{aligned}\Pr[X = 0] &= \binom{n}{n/2} \left(\frac{1}{2}\right)^n \\ &= \frac{n!}{((n/2)!)^2} \left(\frac{1}{2}\right)^n \\ &= \frac{(\frac{n}{e})^n \sqrt{2\pi n}}{\pi n (\frac{n}{2e})^n} \left(\frac{1}{2}\right)^n \\ &= \sqrt{\frac{2}{\pi n}} \frac{n^n}{e^n} \frac{2^n e^n}{n^n} \left(\frac{1}{2}\right)^n \\ &= \sqrt{\frac{2}{\pi n}}\end{aligned}$$

And for odd  $n$ 's,  $X$  can never be 0, so  $\Pr[X = 0] = 0$ .

## Problem 2

### Solution

The algorithm is, for each time, choose a vertex that is not been colored, then generate a uniform distributed random variable  $X \sim \text{Unif}(0, 3)$ . Then if  $0 < X < 1$ , assign this vertex to color 1; if  $1 < X < 2$ , assign it to color 2, else assign it to color 3. (i.e. color this vertex to 3 colors with same probability), until all vertices are colored.

Then, we can prove that, for each adjacent pair of vertices  $V_i$  and  $V_j$ , the probability of they are assigned to same color is  $P = 1 - \binom{3}{1} \left(\frac{1}{3}\right)^2 = \frac{2}{3}$ . Hence the expectation of satisfied edges found by this algorithm satisfies:  $c_{\text{found}} > \frac{2}{3} c_{\text{total}}$ , where  $c_{\text{total}}$  is the total number of edges in the graph. And it's obviously that  $c^* < c_{\text{total}}$ , so  $c_{\text{found}} \geq \frac{2}{3} c^*$ .

## Problem 3

### Solution

1. It's obviously that  $E[X_i]$  is equal to the probability of picking an 1 in the array, So

$$\mathbb{E}[X_i] = \frac{T}{n}$$

- 2.

$$\begin{aligned} \text{Var}[X_i] &= \Pr[X_i = 0](0 - E[X_i])^2 + \Pr[X_i = 1](1 - E[X_i])^2 \\ &= (1 - \frac{T}{n})(\frac{T}{n})^2 + \frac{T}{n}(1 - \frac{T}{n})^2 \\ &= (1 - \frac{T}{n})(\frac{T}{n}) \end{aligned}$$

3. By the linearity of expectation:

$$\begin{aligned} \mathbb{E}[Y] &= \mathbb{E}[\frac{n}{s} \sum_{i=1}^n X_i] \\ &= \frac{n}{s} \sum_{i=1}^n \mathbb{E}[X_i] \\ &= s \cdot \frac{n}{s} \cdot \frac{T}{n} \\ &= T \end{aligned}$$

4. By the linearity of variance of i.i.d variables:

$$\text{Var}[X_1 + X_2 \dots + X_s] = s \cdot \text{Var}[X_1]$$

and

$$\text{Var}[cX] = c^2 \text{Var}[X]$$

We have:

$$\text{Var}[Y] = \frac{n^2}{s^2} \cdot s \cdot \text{Var}[X_i] = \frac{n^2}{s} \text{Var}[X_i]$$

Hence,

$$\text{Var}[Y] = \frac{n^2}{s} (1 - \frac{T}{n})(\frac{T}{n}) = \frac{nT}{s} (1 - \frac{T}{n})$$

It takes its maximum value at  $T = \frac{n}{2}$ , so if  $T$  is at about  $\frac{n}{2}$ , the variance is large. So, the performance is not good (i.e. it's not a good estimator).

## Problem 4

### Solution

The algorithm is:

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**Algorithm 1** Maximum 3D matching
 

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Initialize  $S = \{\}$ 
while  $T$  are all non-empty do
    Randomly choose a triple  $(X_i, Y_j, Z_k)$  from  $T$ 
    Add  $(X_i, Y_j, Z_k)$  to  $S$ 
    Delete all triples related to  $(X_i, Y_j, Z_k)$  in  $T$ 
end while
  
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Notice that the matching found by this algorithm is not a subset of any other matching in  $T$ , for we found add triples until no triples can be added.

For any 2 different subset  $S_1$  and  $S_2$  found by this algorithm, we'll prove that  $|S_1| \leq 3|S_2|$ .

First notice that: each triple  $(x_i, y_j, z_k)$  in  $S_2 \setminus S_1$  can be adjacent to at most 3 edges in  $S_1 \setminus S_2$ . For  $S_1 \setminus S_2$  at most contains  $(x_i, -, -)$ ,  $(-, y_j, -)$ ,  $(-, -, z_k)$ . And each edge in  $S_1 \setminus S_2$  is adjacent to an edge in  $S_2 \setminus S_1$ . since  $S_2$  is a maximal matching. Hence

$$|S_1 \setminus S_2| \leq 3|S_2 \setminus S_1|$$

Thus we have:

$$|S_1| = |S_1 \cap S_2| + |S_1 \setminus S_2| \leq |S_1 \cap S_2| + 3|S_2 \setminus S_1| = 3|S_2|$$

So we proved  $|S_1| \leq 3|S_2|$ . By  $|S_1|, |S_2| \leq |T|$ ,  $|S_1| \leq 3|S_2|$  and the size of maximum 3D matching is also less than the size of  $T$  (i.e.  $|M| \leq |T|$ ), we can observe for any set  $S$  found by our algorithm:

$$|S| \geq \frac{1}{3}|M|$$

Hence we can find 3 dimensional matching of size at least  $1/3$  times the maximum possible size by this algorithm. For the time complexity, we choose at most  $O(n)$  times, and for each choice, at most delete  $O(n)$  triples. Hence the time complexity is:

$$T(n) = O(n^2)$$

## Problem 5

### Solution

- a) Let  $v \in T$ . If  $v \notin S$ , we can observe that there must exists a vertex  $v'$  which is the neighbor of  $v$ . And  $v'$  is selected by the greedy algorithm and  $v$  is removed. Then, we must have that:  $w(v') \geq w(v)$ , otherwise the greedy algorithm will choose  $v$  instead of  $v'$ . Thus, for each node  $v \in T$ , either  $v \in S$ , or there is a node  $v' \in S$  so that  $w(v) < w(v')$  and  $(v, v')$  is an edge of  $G$ .
- b) Let  $C = S \cap T$ , and  $S' = S \setminus C$ ,  $T' = T \setminus C$ . By (a), we know that for each node  $v \in T'$ , there is a node  $v' \in S$  so that  $w(v) < w(v')$  and  $(v, v')$  is an edge of  $G$ . We use  $v' \in S'$  to cover for  $v$  in  $T'$ , and any such  $v'$  need to cover at most 4 neighbors of  $v \in T'$  where  $w(v) \leq w(v')$ . Thus we have

$w(S') \leq 4w(T')$ . Hence,

$$\begin{aligned} w(S) &= w(S') + w(C) \\ &\geq w(C) + \frac{1}{4}w(T') \\ &\geq \frac{1}{4}(w(T') + w(C)) \\ &= \frac{1}{4}w(T) \end{aligned}$$

Hence, we've proved the "heaviest-first" algorithm returns an independent set of totalweight at least  $1/4$  times the maximum total weight of any independent set in  $G$ .

## Problem 6

### Solution

- a) If  $w_1 = 1, w_2 = 2, w_3 = 3, w_4 = 4, w_5 = 5, w_6 = 6$ , and  $K = 7$ , then the minimum possible number of trucks is 3. Because we can let first truck contain  $w_1, w_6$ , second truck contain  $w_2, w_5$ , third truck contain  $w_3, w_4$ . But by this algorithm, we need 4 trucks: first contains  $w_1, w_2, w_3$ , second contains  $w_4$ , third contains  $w_5$ , fourth contains  $w_6$ , which is not optimal.
- b) Let  $T_i$  denotes the items in truck  $i$  by our greedy approach, and  $W_i$  denotes the total weight of items in truck  $i$  (i.e.  $W_i = \sum_{a \in T_i} w_a$ ). Then we can observe that:

$$W_i + W_{i-1} > K$$

for any  $i$ . Because if  $W_i + W_{i-1} \leq K$ , our greedy algorithm will merge  $T_i$  and  $T_{i-1}$  together.

Then we discuss following 2 cases:

If  $N = 2m$  ( $N$  is even) for some integer  $m$ :

$$\sum_{j=1}^N W_j = \sum_{j=1}^m (W_{2j} + W_{2j-1}) > Km$$

and  $N = 2m + 1$  ( $N$  is odd) for some integer  $m$ :

$$\sum_{j=1}^N W_j = \sum_{j=1}^m (W_{2j} + W_{2j-1}) + W_{2m+1} > Km$$

For  $m \geq \frac{N-1}{2}$ , we can observe

$$\sum_{i=1}^n w_i = \sum_{j=1}^N W_j > Km \geq \frac{K(N-1)}{2}$$

And for the minimum number of trucks  $N^*$  we have:

$$\sum_{i=1}^n w_i \leq KN^*$$

By inequalities shown above, we know

$$\begin{aligned} \frac{K(N-1)}{2} &< KN^* \\ N &< 2N^* + 1 \end{aligned}$$

Hence,

$$N \leq 2N^*$$

## Problem 7

### Solution

- a) Let  $S = \{1, 2, 3, 101\}$ , and  $B = 102$ . then the optimal solution is choosing  $S = \{1, 101\}$ , but the solution given by this algorithm is  $S = \{1, 2, 3\}$ . then for  $6 < 102$ , this solution is less than half of some other solution.
- b) The algorithm is: That is a greedy algorithm which picks the one with maximum weight that can be

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#### Algorithm 2 Maximum subset sum

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Initialize  $S = \{\}$ ,  $i = n$ ,  $C = B$ 
Sort  $\{a_1, a_2, \dots, a_n\}$  to descending order
while  $i \neq 0$  do
  if  $a_i < C$  then
     $C = C - a_i$ 
    Add  $a_i$  to  $S$ 
  end if
   $i = i - 1$ 
end while

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added to the subset.

Proof:

If subset  $S$  found by our algorithm has total weight  $W$ , and  $W^*$  denotes the optimal solution (i.e.  $W = \sum_{a \in S} w_a$ ). We hypothesis that  $W < \frac{1}{2}W^*$ . By the hypothesis, we have  $\exists a \in (A \setminus S), w_a \geq \frac{1}{2}B$ . But the items in the set satisfies  $w < \frac{1}{2}B$ , which means we didn't pick the item with maximum weight. By this contradiction, we know that:

$$W = \sum_{a \in S} w_a \geq \frac{1}{2}B$$

And the total weight of the optimal solution  $S^*$  is:

$$W_{S^*} = \sum_{a \in S^*} w_a \leq B$$

Hence, we have:

$$W \geq \frac{1}{2}W_{S^*}$$

Thus we found an algorithm which returns a feasible set  $S \subseteq A$  whose total sum is at least half as large as the maximum total sum of any feasible set  $S' \subseteq A$ .

The time complexity is  $O(n \log n)$ , for the sort takes  $O(n \log n)$  time, and assignment takes  $O(n)$  time.