

Assignment1

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2015CS10600

1 Question-1

Optimal has atmost l edge-disjoint paths of length greater than length l . Otherwise, the graph will have $> l^2 \geq m$ edges which is a contradiction. If the greedy algorithm gives k edge disjoint paths of length less than l , then each of the paths in optimum of length less than l should intersect edges in one of these k paths, otherwise greedy would have picked the path in the optimum which doesn't intersect any of these paths, and no two paths in optimum can intersect the same edge of the k -paths given by the greedy algorithm. As the total number of edges in the optimum is less than kl , we can only have atmost kl disjoint paths of length less than l in the optimum. So, the no. of disjoint paths in optimum is $\leq l + kl$. But k is greater than 1 as $l > \text{diameter}(G)$. Hence, greedy can choose the shortest path between any of the pairs(The length will be less than l) which gives $k \geq 1$. So, $OPT \leq l + kl \leq kl + kl \leq 2kl \Rightarrow l \geq (1/2k)OPT$. Hence, the algorithm is a $\Omega(1/k)$ -optimal algorithm.

2 Question-2

We are given m machines and n jobs each with $p_j \geq OPT/3$. Let the jobs be ordered in a way such that $p_1 \geq p_2 \geq \dots \geq p_n$. And k be the largest integer such that $p_k > 2OPT/3$. So, in the optimum solution, we have that machines which are allotted jobs 1 through k are given only one job and all other machines are allotted atmost 2 jobs. So, the total number of jobs is atmost $2n - k$. Now, the greedy algorithm, allots first k jobs to k machines as do the optimal solution. We claim that the greedy algorithm allots the remaining jobs to machines in the following way:

- if jobs i and j are allotted to l th machine then, $i + j = 2n + 1$.

Proof. The greedy algorithm allots the next $n - k$ jobs to the remaining $n - k$ machines one after another. Now the machine with the lightest load is n th machine. So, greedy algorithm allots the $n + 1$ th job to n th machine and load on n th machine is $\geq OPT/3 + OPT/3 = 2OPT/3$. But the load on the machines $k + 1 \dots n - 1 \leq 2OPT/3$ as the jobs that are allotted to the machines are of size $\leq 2OPT/3$. Hence, $n + 2$ th job is allotted to $n - 1$ th machine. Now the load on $n - 1$ th machine is $\geq 2OPT/3$. Hence, the next job is assigned to $n - 2$ th machine and so on. We have atmost $2n - k$ jobs and hence in the worst case, the last job is assigned to $n - k + 1$ th machine.

Claim: There is an optimum solution which assigns the jobs as the greedy algorithm.

Let O be an optimal solution. If it is same as greedy, we are done. Otherwise there exists two machines to which jobs (i, j) and (i', j') are assigned such that $i < i'$ and $j < j'$ (We can assume WLOG that $i < j$ and $i' < j'$). As O is optimal, $p_i + p_j < OPT$ and $p_{i'} + p_{j'} < OPT$. But $j < j' \Rightarrow p_i + p_{j'} \leq p_i + p_j \leq OPT$ and $i < i' \Rightarrow p_j + p_{i'} \leq p_j + p_i \leq OPT$. So, the schedule O' which assigns the jobs (i, j') and (i', j) to the machines will have makespan atmost makespan(O). But O is optimal. So, makespan(O') = OPT . By continuing this we arrive at the solution given by the greedy algorithm. \square

3 Question-3

Let V be the set of vertices and S_1 and S_2 be the partition of V . Pick an arbitrary vertex and put it into S_1 . Now iterate over the remaining vertices and one after another put them into S_1 or S_2 depending on which of the configurations gives greater cut size based on the vertices added to S_1 and S_2 till that point. This is a $1/2$ -approximation. Notation: v_i be the vertex picked in i th iteration, $S_1(i), S_2(i)$ be the sets S_1 and S_2 at the end of the i th iteration, $V(i) = S_1(i) \cup S_2(i)$ and $G(i)$ be the graph induced by the vertices $V(i)$.

Claim: Partition $(S_1(i), S_2(i))$ is a $1/2$ -approximation for $G(i)$.

Proof. Proof by induction on i .

True for $i = 1$ as the optimal partition has weight 0.

Assume that the claim is true for all $k < i$.

Without loss of generality, assume that $v_i \in S_1$. Let $(A, V(i) - A)$ be the optimal partition for $G(i)$ and $v(i) \in A$. We have $OPT(G(i)) = wt((A, V(i) - A)) = wt((A - v(i), V(i) - A)) + \text{contribution of } v_i$. But $wt((A - v(i), V(i) - A)) \leq OPT(G(i - 1))$ and contribution of $v_i \leq \text{weight of edges incident on } v_i \text{ in } G(i)$. So,

$$OPT(G(i)) \leq OPT(G(i - 1)) + \text{weight of edges incident on } v_i \quad (1)$$

By the induction hypothesis we have $wt(S_1(i) - v_i, S_2(i)) = wt(S_1(i - 1), S_2(i - 1)) \geq OPT(G(i - 1))/2 \geq wt((A - v(i), V(i) - A))/2$. We also have $wt(v_i, S_2(i)) \geq (\text{weight of edges incident on } v_i)/2$ as v_i is added to S_1 by the greedy algorithm. So, $wt(S_1(i), S_2(i)) = wt(S_1(i) - v_i, S_2(i)) + \text{contr. of } v_i \geq OPT(G(i - 1))/2 + (\text{wt. of edges incident on } v_i)/2 \geq OPT(G(i))/2$. \square

Thus the partition $(S_1(n), S_2(n))$ is a $1/2$ -approximation for the graph $G(n) = G$.

The generalized algorithm for k partitions is as follows. Start with $S_1, S_2, \dots, S_k = \phi$. Iterate over the vertices v_i . Put the vertex v_i into the set S_j such that

$$j = \text{argmax}_j wt(S_1, \dots, S_{j-1}, S_j \cup v_i, S_{j+1}, \dots, S_k) \quad (2)$$

So,

$$\text{contribution of } v_i \text{ at the end of } i\text{th iteration} = \max_j wt(S_1, S_2, \dots, S_{j-1}, v_i, S_{j+1}, \dots, S_k) \quad (3)$$

$$\geq \frac{1}{k} \sum_j wt(S_1, S_2, \dots, S_{j-1}, v_i, S_{j+1}, \dots, S_k) \quad (4)$$

$$\geq \frac{1}{k} (k - 1) \text{wt. of edges incident on } v_i \quad (5)$$

So, given that $wt(S_1(i-1), S_2(i-1), \dots, S_k(i-1)) \geq (1 - 1/k)OPT(G(i-1))$ we have $wt(S_1(i), \dots, S_j(i), \dots, S_k(i)) = wt(S_1(i-1), S_2(i-1), \dots, S_j(i-1) \cup v_i, \dots, S_k(i-1)) = wt(S_1, S_2, \dots, S_j, \dots, S_k) + \text{contr. of } v_i \geq (1 - 1/k)OPT(G(i-1)) + (1 - 1/k)\text{wt. of edges incident on } v_i \geq (1 - 1/k)OPT(G(i))$. Hence $(S_1(n), S_2(n), \dots, S_k(n))$ is $(1 - 1/k)$ approximate partition of $G(n) = G$.

4 Question-4

Let $S = \{1, 2, 3, \dots, k\}$. And each of them have a coverage requirement of $\alpha_i \in \mathbb{Z}^+$. Let $\sum_i \alpha_i = n$. The greedy algorithm is as follows:

1. $\beta_i = \alpha_i$ for all $i \in S$ and $r_i = 0 \forall S_i$.

2. if $\exists i \beta_i > 0$, continue else exit
3. find j such that $\frac{w(S_j)}{|S_j \cap \{i | \beta_i > 0\}|}$ is minimum.
4. $r_j := r_j + 1$ and $\beta_i := \beta_i - 1 \forall i \in S_j$
5. Go to step 2

The above algorithm is a $\ln n$ -approximate algorithm. Let $n_i = \sum_j \beta_j$ before i th iteration. So, $n_1 = n$ and $n_i - n_{i+1}$ is the number of elements covered in i th iteration. Let S_k be the set chosen in i th iteration. Optimum solution can cover n_i elements with cost OPT . So, there should be a set in OPT which can cover some elements with average cost less than OPT/n_i . But average cost of S_k should be even less. Hence,

$$w(S_k) \leq OPT * (n_i - n_{i+1})/n_i \quad (6)$$

Summing the above inequality over all the iterations, we have

$$\sum_j r_j wt(S_j) \leq OPT * H_n \quad (7)$$

So, this algorithm is a $\ln n$ -approximation.

5 Question-5

Consider three sets of vertices $\{r\}$, $\{v_1, v_2, \dots, v_m\}$ one for each of the sets S_i and $\{u_1, u_2, \dots, u_n\}$ one for each of the elements of set S . Construct a graph G with vertices as the union of $\{r\}$, $\{v_1, v_2, \dots, v_m\}$ and $\{u_1, u_2, \dots, u_n\}$ with the edges as follows, $r \rightarrow v_i$ for all $i = 1..n$ with cost of edge $r \rightarrow v_i = wt(S_i)$ and the edges $v_i \rightarrow u_j \forall i, j$ such that $j \in S_i$ of cost 0. Set the vertices $r, \{u_1, u_2, \dots, u_n\}$ as required and r as the special vertex and the vertices $\{v_1, v_2, \dots, v_m\}$ as steiner vertices. Any tree with root r and containing the vertex u_j should have an edge to v_k such that there is an edge.

Let $T = (V', E')$ is a steiner tree. Using this we shall construct a set-cover. Consider the set-cover $\mathcal{S} = \{S_j | r \rightarrow v_j \in E'\}$.

Claim: \mathcal{S} is a vertex cover and $cost(\mathcal{S}) = wt(T)$.

Proof. Consider an element $e_i \in S$. There is a corresponding vertex u_i in the vertex set of G and u_i is a required vertex. So, there exists a vertex v_k such that $r \rightarrow v_k$ and $v_k \rightarrow u_i$ which implies $S_k \in \mathcal{S}$ and $e_i \in S_k$. So, $e_i \in \cup_{S_j \in \mathcal{S}} S_j$. As e_i is an arbitrary element of S , we have \mathcal{S} is a set cover. $cost(\mathcal{S}) = \sum_{S_j \in \mathcal{S}} wt(S_j) = \sum_{v_j: r \rightarrow v_j \in E'} wt(r \rightarrow v_j) = wt(T)$. So, cost of the set-cover \mathcal{S} is $wt(T)$. \square

Let \mathcal{S} be a vertex cover. We shall construct a steiner tree T . Consider the graph $T = (V, E)$ with V as union of $\{r\}, \{v_j | S_j \in \mathcal{S}\}$ and $\{u_1, u_2, \dots, u_n\}$ and edges as union of $\{r \rightarrow v_j | v_j \in V\}$ and $\{v_j \rightarrow u_k | e_k \in S_j\}$. Remove edges if there are morethan one edges incoming to a vertex of form u_k . Given that \mathcal{S} is a setcover, we can see that the tree T is a steiner tree and cost of tree T is equal to weight of set cover \mathcal{S} .

From above,

$$OPT(SteinerTree) = OPT(SetCover) \quad (8)$$

Hence, given a $O(\log(n))$ approximation algorithm for steiner-tree, we can solve the steiner tree problem corresponding to the setcover problem and give $O(\log(n))$ approximation to set-cover problem.