

JAN 11 NOTES FROM CSE 599: ONLINE ALGORITHMS

Previously, we saw that the competitive ratio for algorithms on the fractional online bipartite matching problem is bounded above by $1 - \frac{1}{e}$. Below we will see an algorithm that achieves this bound.

1. WATER LEVEL ALGORITHM

First, we describe the algorithm. Let x_{ij} denote the value placed on edge (i, j) (the amount of water flowing from j to i). Let $y_i = \sum_{j \sim i} x_{ij}$ be the sum of the values of all the edges incident to i (the level of water in bin i).

Now, when a new vertex j arrives, we do the following: While $\sum_{i \sim j} x_{ij} < 1$ and $\min_{i' \sim j} y_{i'} < 1$, increment x_{ij} by dx for each i in $\arg\min_{i' \sim j} y_{i'}$.

Proposition 1. *The algorithm described above has competitive ratio at least $1 - \frac{1}{e}$.*

Proof. Let α_i denote the total money placed on a left vertex i by the algorithm (ALG) and let β_j denote the total money placed on a right vertex j by ALG. Define the function $g(x) = e^{x-1}$. Let y_i^f denote the value of y_i at the termination of the algorithm (similarly for α_i^f, \dots).

Fix an edge (i, j) . If the algorithm assigns value dx to the edge (i, j) , then we increment α_i by $g(y_i)dx$ and β_j by $(1 - g(y_i))dx$.

Now if we show that $\alpha_i^f + \beta_i^f \geq 1 - \frac{1}{e}$ for edge (i, j) , the proof will be complete. Note that

$$\alpha_i^f = \int_0^{y_i^f} g(y_i) dy_i = g(y_i^f) - \frac{1}{e}$$

By the monotonicity of g , we have that $g(y_i) \leq g(y_i^f)$ throughout the process. Hence $\beta_j^f \geq 1 - g(y_i^f)$. Thus we have that $\alpha_i^f + \beta_i^f \geq 1 - \frac{1}{e}$. \square

Remark 2. The g in the above proof can be derived from the properties we desire it to have. If we let $G(y) = \int_0^y g(x) dx$ and γ be the target competitive ratio, then we have the desired properties $G(1) - G(0) = \gamma$ and $G(y_i^f) - G(0) + 1 - g(y_i^f) = \gamma$. From these equations, we can get a differential equation (by differentiating the second equation) and boundary conditions. Solving these, we obtain the function g .

We can also look at this from the viewpoint of linear programming (LP). Written as an LP, the fractional bipartite matching problem has the form:

$$\begin{aligned} & \max \sum_{(i,j) \in E} x_{ij} \text{ such that} \\ & \bullet \sum_{j \sim i} x_{ij} \leq 1 \text{ for every } i \in L \\ & \bullet \sum_{i \sim j} x_{ij} \leq 1 \text{ for every } j \in R \\ & \bullet x_{ij} \geq 0 \end{aligned}$$

This LP has the following dual:

$$\begin{aligned} & \min \sum_i \alpha_i + \sum_j \beta_j \text{ such that} \\ & \bullet \alpha_i + \beta_j \geq 1 \text{ for all } (i, j) \in E \\ & \bullet \alpha_i, \beta_j \geq 0 \end{aligned}$$

Above, we showed that $\alpha_i + \beta_j \geq \gamma = 1 - \frac{1}{e}$. Thus, if we scale the constraints in the dual problem by γ (i.e. we require $\geq \gamma$ instead of ≥ 1), then we obtain a feasible dual problem. This scaling corresponds to multiplying the objective function in the primal problem by γ . Thus, using weak duality we have the following inequality:

$$\gamma \text{PRIMAL} \leq \gamma \text{OPT} \leq \text{FEASIBLE DUAL}$$

Now by noting that our algorithm determines an instance of FEASIBLE DUAL and by dividing by OPT, we see a proof that the competitive ratio is bounded below by γ .

2. BUDGET ALLOCATION (ADWORDS) PROBLEM

For this problem, we introduce the following notation. Let L (the left vertices) be a set of advertisers. Each $i \in L$ has a budget B_i . Let R (the right vertices) be a set of “queries.” For every $j \in R$ and $i \in L$, we have a bid b_{ij} . Now each query can be matched to at most one advertiser. Then the quantity that we want to maximize is:

$$\sum_i \min \left\{ \sum_{j:j \rightarrow i} b_{ij}, B_i \right\}$$

For the adwords case of this problem, we make the additional assumption that $\frac{b_{ij}}{B_i} \ll 1$.

The fractional version of this problem is easier to work with and, under the adwords assumption, produces similar results. In the fractional setting we define x_{ij} to be the weight that we place on an edge b_{ij} . Then the problem becomes the following LP:

$$\begin{aligned} & \max \sum_{(i,j)} b_{ij} x_{ij} \text{ such that} \\ & \bullet \sum_i x_{ij} \leq 1 \text{ for every } j \in R \\ & \bullet \sum_j b_{ij} x_{ij} \leq B_i \text{ for every } i \in L \\ & \bullet x_{ij} \geq 0 \end{aligned}$$

This problem has the dual:

$$\begin{aligned} & \min \sum_i \alpha_i B_i + \sum_j \beta_j \text{ such that} \\ & \bullet \alpha_i b_{ij} + \beta_j \geq b_{ij} \text{ for all } (i, j) \\ & \bullet \alpha_i, \beta_j \geq 0 \end{aligned}$$