

① a) Given: $C_u = \alpha n_u + \beta \sum_v (\text{dist}(u,v))^2$, $\alpha, \beta \geq 0$

let us assume three cases: $\alpha = \beta$, $\alpha < \beta$, $\alpha > \beta$

for $\alpha = \beta = k$

$$C_u = k(n_u + \sum_v \text{dist}^2(u,v))$$

as α & β grow, the cost grows linearly with respect to k .

if the graph is a star, then $\text{dist}(u,v)$ is at max 2

but if G is complete then $\text{dist}(u,v)$ is max 1 and $n_v = \frac{n(n-1)}{2}$
any other graph would have $\text{dist}(u,v) \geq 1$ with $n_v \geq n-1$ nodes,
meaning the OPT is a star graph for $\alpha = \beta$

we can use a ratio of sums for the other 2 cases:

★ $\frac{C_{\text{star}}}{C_{\text{complete}}} = \frac{4\beta + \alpha(n-1)}{\beta + \alpha(\frac{n(n-1)}{2})}$ [note: $\lim_{n \rightarrow \infty} \frac{C_s}{C_c} = \frac{1}{2}$ for $\beta = \alpha$]

if $\beta < \alpha$: then a star is OPT since adding a node decreases cost by $4\beta - \alpha$ ^{per edge} and removing an edge from a complete graph decreases cost by α
so if $\alpha \geq 4\beta$ then adding a node is not incentivized and only removing nodes is.

if $\alpha \geq \beta$ then a complete graph is OPT since
reducing an ~~edge~~ ^{edge} lowers cost by 1 and adding an edge
can reduce a cost by $\alpha - \beta$ for $n-1$ paths where
all nodes are incentivized to add a node.

b) Similarly ★ shows us a ★ graph is NE since ^{paths} ~~all paths~~
want to minimize their max cost so from a star graph, adding
a node doesn't decrease their max cost unless it creates a complete
graph, and from a complete graph, the cost decreases
By $4\beta - \alpha(\frac{n(n-1)}{2})$ so a star graph is ideal since
removing nodes to get ^{cost} $4\beta + \alpha(n-1)$ is ideal.

(2) Given $s_i = k \forall m$, social objective $O_G = \sum_i L(A_i) = \sum_i \sum_j w_j$
 $POA = \frac{\max_{NE}(O)}{OPT}$

Since all machines have the same speed, average cost is $C_i = \frac{1}{m} \sum_i L(A_i)$

$C_i = \frac{1}{m} \sum_i \sum_j w_j$ which is social optimum since no machine is incentivized to have a higher cost than average since they are all the same speed, $\therefore OPT = \min \{C_i \in C\} = \frac{1}{m} \sum_i L(A_i)$. In case one machine requires a higher load, then you can assign jobs in order of weight for $\max_{NE}(O)$:

Let us assume not all jobs are equal such that

$$w_1 \leq w_2 \leq \dots \leq w_n \quad \forall w \in W$$

$$\frac{w_k}{s_i} \leq \frac{w_k + (a \neq k)}{s_i} \text{ where 'a' is a subset of any size in } W$$

this implies $\max_{NE}(O) = \max \left\{ \frac{w_k}{s_i} \right\} \forall i \in m \text{ \& } \forall k \in W$
 $\max_{NE}(O) = \frac{\max\{w_k\}}{s_i}$

where the \max_{NE} is determined by the 'heaviest' job.

hence $POA = \frac{\max\{w_k\}/s_i}{\frac{1}{m} \cdot \sum_i L(A_i)} = \frac{m}{s_i} \cdot \frac{\max\{w_k\}}{\sum_i L(A_i)} = \frac{m}{s_i} \cdot \frac{\max\{w_k\}}{O_G}$

let $\frac{m}{s_i} = \alpha$ since s_i & m are constants so:

$$POA = \alpha \cdot \frac{\max\{w_k\}}{O_G}$$

③ Given: $\text{Cost}_e = c(e) \geq 0$, $\text{delay}_e = d(e) \geq 0$, $N = K_n$ with $\binom{n}{2}$ edges

Find: path $p \in N$ where $\sum_p d(e) < D$ where $\sum_p c_e$ is minimized

Note: $c(e)$ is a bid by ISP & $d(e)$ is not

Answer:

We want to find a VCG mechanism s.t. $f(v_1, \dots) = \arg \max_{p \in N} \sum_i v_i(p)$
 in terms of our problem, we find the shortest path p where
 $d_p = \sum_p d(e) < D$. That is, we find a path p minimizing delay
 where $d(\hat{e}) = \sum_{e \in p} d(e) - \sum_{e \in p - \{\hat{e}\}} d(e)$.

Essentially, we bid for a min cost path from the set
 of min delay paths with delay less than D .

so we find $\arg \min_p \left\{ \arg \min_{p \in N} \left\{ \sum_p d(e) \right\} \right\}$.

so given a complete graph N , we can assume there is
 always a path containing e and not containing e .

hence: total delay $d = \sum_{e \in p^*} d(\hat{e})$ where p^* is the ^{winner} path

$$\text{so } d = \sum_{e \in p^*} \sum_{e \in p} d(e) - \sum_{e \in p^*} \sum_{e \in p - \{e\}} d(e)$$

where there can be any path $\{p^*\}$ with delay $\{d_{p^*}\} < D$.

since we only care if $d_{p^*} < D$ and we want to minimize
 cost, Then the problem is incentive compatible
 where we procure any path with delay $d(p^*) < D$ and
 minimize the cost such that we get:

$$p^* = \arg \min_{\sum_p c(e) \in N} \left\{ \sum_{e \in p} \left(\sum_{e \in p} d(e) - \sum_{e \in p - \{e\}} d(e) \right) < D \right\}$$

4) a) Traders maximize utility when all endowment is spent.

so: when goods j & j' are bought at prices p_j & $p_{j'}$, then

$$k \cdot \frac{p_j}{p_{j'}} = \frac{u_j}{u_{j'}} \quad \text{where the ratio of prices are proportional to the utility gained of the goods.}$$

$$u_{ij} = v_{ij} x_{ij} \quad \text{where } v_{ij} = r_{ij} s_i \quad \text{so:}$$

$$k \cdot \frac{p_j}{p_{j'}} = \frac{x_{ij}}{x_{ij'}} \cdot \frac{r_{ij}}{r_{ij'}} \cdot \frac{s_i}{s_i} \quad \text{so to prove ratios equal}$$

we construct an endowment m_i where $k = \frac{x_{ij'}}{x_{ij}} = 1$

We know for linear Fisher: $m_i = \sum_j p_j x_{ij} \rightarrow p_j = \frac{m_{ij}}{x_{ij}} \rightarrow p_i = \frac{m_i}{\sum_j x_{ij}}$

\rightarrow and if a product has a $p_j > 0$ then $\sum_i x_{ij} = \sum_i x_{ij'} = 1$

which shows us $x_{ij} = x_{ij'}$ since the endowment spent per good equals $x_{ij} = x_{ij'}$

$$\text{hence: } \frac{p_j}{p_{j'}} = \frac{r_{ij}}{r_{ij'}}$$

b) Market equilibrium occurs when profits are maximized which happens when a_j of j is sold and m_i for i is spent.

from part A, we can get $p_{ij} r_{ij'} = p_{ij'} r_{ij}$ and we know

$m_i = \sum_j p_{ij} x_{ij}$ where x_{ij} converges to 1 so $m_i \leq p_i$, meaning

we want to sell products to break even or make a profit

so we set the price $p_{ij}^* \geq p_{ij} \geq m_{ij}$ or the price at least what we bought but no greater than our competitor. This incentivizes i to buy more j while i' does not since they would not gain anything.

hence allocation $a_{ij} = \begin{cases} 1 & \text{if } (*) \text{ satisfied} \\ 0 & \text{if } (*) \text{ not satisfied} \\ [0, a_j - a_{ij}] & \text{if competitor sells at our price} \end{cases}$