

Today ..→ Online Set Cover

↳ Algo + Proof

→ Online Routing

↳ Algo + Proof

 U = universe of possible elements \mathcal{S} = collection of sets = m $X \subseteq U$ is online set of elements that need to be covered, $S_A \subseteq \mathcal{S}$ is online set cover for X $|X| = n$ $e_1 \quad e_2 \quad \dots \quad e_n$ ↓
Alg has to cover (done w/o knowledge of future).Goal : minimize competitive ratio

$$CR(\text{Alg}) = \max_{\text{All possible inputs } X} \frac{\# \text{ sets alg picks for input } X}{\text{Optimal } \# \text{ sets needed for } X}.$$

Last class

↳ CR algo was proposed, but also shown why bad.

↳ Fractional algo (MW-inspired, increases confidence in sets based on past utility)

ONLINE

FRACTIONAL ALGORITHM

ONLINE ROUNDING SCHEME

Maintain fractional assignments for all sets

→ Make sure that S_i is

Maintain fractional solution for all sets

$$x_{S_1} = 0.7, x_{S_2} = 0.2, x_{S_3} = 0.3 \text{ A.C.}$$

updated monotonically over time

Make sure that S_i is included with probability $\propto x_{S_i}$

eg:- (e_1) arrived

$$x_{S_1} = 0.5, x_{S_2} = 0.3$$

$$x_{S_3} = 0.2$$

(e_2) arrived

$$x_{S_1} \rightarrow 0.8$$

$$x_{S_2} \rightarrow 0.9$$

$$x_{S_3} = 0.2$$

$$\begin{aligned} \text{Include } S_1 \text{ w.p. } &\propto C \cdot x_{S_1} \\ S_2 \text{ w.p. } &= C \cdot x_{S_2} \\ S_3 \text{ w.p. } &= C \cdot x_{S_3} \end{aligned}$$

$$\begin{aligned} \text{Include } S_1 \text{ w.p. } &\propto C \Delta x_{S_1} \\ S_2 \text{ w.p. } &\propto C \Delta x_{S_2} \\ &\vdots \end{aligned}$$

Maintaining

Invariant that at any time, every set is chosen in solⁿ w.p. = $\underbrace{C \cdot x_S}$

Thm ①

There exists online algo. for maintaining fractional solutions
st cost of fractional solⁿ $\leq O(\log m) \cdot \text{OPT}$
 $\sum x_S^{(t)} \leq O(\log m) \text{OPT}^{(t)}$

Thm ②

There is a rounding scheme to select sets (based on fractional solⁿ)

$$\text{st @ time } t, E\left[\sum Z_S^{(t)}\right] \leq O(\log n) \sum x_S^{(t)}$$

Indicator variable for whether S has been rounded or not.

Thm ① + ②

$$\Rightarrow E[\text{Cost of online Algo}] \leq O(\log m \log n) \text{OPT}^{(t)}$$

sets

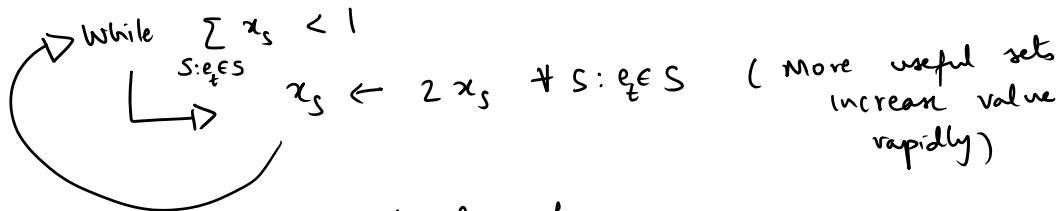
elts that show up = $|X|$

Thm ①: fractional algorithm

Initialize all sets to $x_s^{(0)} = \frac{1}{m}$

$$\sum x_s^{(0)} = 1$$

→ When element e_t arrives



Resulting $\{x_s\}$ values form fractional solution @ time t , called x_t .

<p><u>Repeatedly useful sets</u></p> <p>$\frac{1}{m} \rightarrow \frac{2}{m} \rightarrow \frac{4}{m} \rightarrow \dots$</p> <p>$\leq O(\log m)$ times</p>	}	<p>Each element participates in a doubling step step $\leq O(\log m)$ times</p>
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⌊

Thm ①

At any time t , $\sum x_s^t =$ fractional cost of online algo at time t

$\leq O(\log m) \text{OPT}^t$

↑ optimal set cover for elements arrived upto time t .

Proof :-

Min $\sum x_s$

$$\sum_{S: e \in S} x_s \geq 1 \quad \forall e_1, e_2, \dots, e_t$$

Max $\sum y_e$

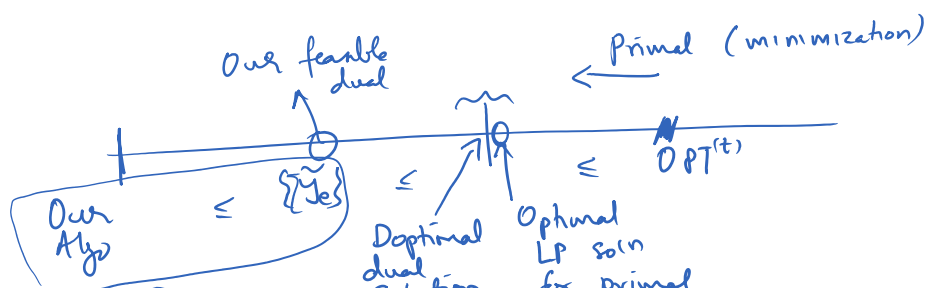
$$\sum_{e \in S} y_e \leq 1 \quad \forall S \in \mathcal{L}$$

Lemma ①

Will construct a feasible dual solution $\{\tilde{y}_e^t\}$ st

Online Frac Cost = $\sum x_s^t \leq O(\log m) \cdot \sum_e \tilde{y}_e^t \leq O(\log m) \text{OPT}^t$

↑ weak duality



Optimal dual solution

↙
twofold goal

- (a) It should be feasible
 (b) It should help me bound cost of online algorithm

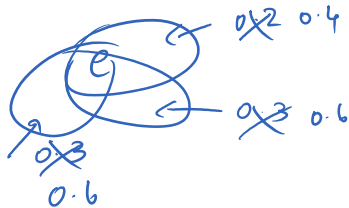
$$\sum x_s^{(0)} = 1$$

$\text{While } \sum_{S: e_t \in S} x_S < 1$

$x_S \leftarrow 2x_S \quad \forall S: e_t \in S$ (more useful sets increase value rapidly)

(

Q) How much does cost increase be? (in one iteration of ...)



$\sum x_s = 0.8$
 $s_{\text{covering}} \downarrow$ while loop is entered
 Overall new soln. from

$$\sum_{s \text{ covering } e} \pi_s = 1 - b$$

$\xleftarrow{\text{In this example}}$
 $\xrightarrow{\text{No other } x_s \text{ was increased}}$
 $\text{Overall fractional increase} = 0.8$

CI: In any iteration, fractional cost increases by $\leq \textcircled{1}$
 \hookrightarrow (lets increase $y_e \leftarrow y_e + 1$ in this iteration)

For every while loop iteration,
see which element caused that iteration,
increase $y_e \leftarrow y_e + 1$.

let $\{\tilde{y}_e^t\}$ denote final dual variables

$$\boxed{\text{Final fractional cost} = \sum_{\text{all } S} x_S^t \leq \sum_e \tilde{y}_e^t} \quad \leftarrow \text{by design}$$

Worry Dual solution may not be feasible \because
 \hookrightarrow Some y_e^s can be > 1

$$\boxed{\text{Not true that } \sum_{e \in S} \tilde{y}_e^t \leq 1 \quad \forall S} \quad \text{😞}$$

Saving grace

Constraint won't be violated by too much !!

Claim :-

$$\sum_{\substack{e \in S \\ \text{LHS}}} \tilde{y}_e^t \leq (\log m + 1) \quad \forall S$$

Proof :-

Fix a set S .

look at all the iterations which increased the LHS

\hookrightarrow Nice property :- in all such iterations, S must be covering the current element which caused the iteration.

in each such iteration, x_S value doubled.

\downarrow
can't cross 2!

$$\frac{1}{m} \rightarrow \frac{2}{m} \rightarrow \frac{4}{m} \rightarrow \dots \rightarrow \frac{m}{m} \rightarrow \frac{2m}{m}$$

$$\# \text{ iterations} = (\log m + 1)$$

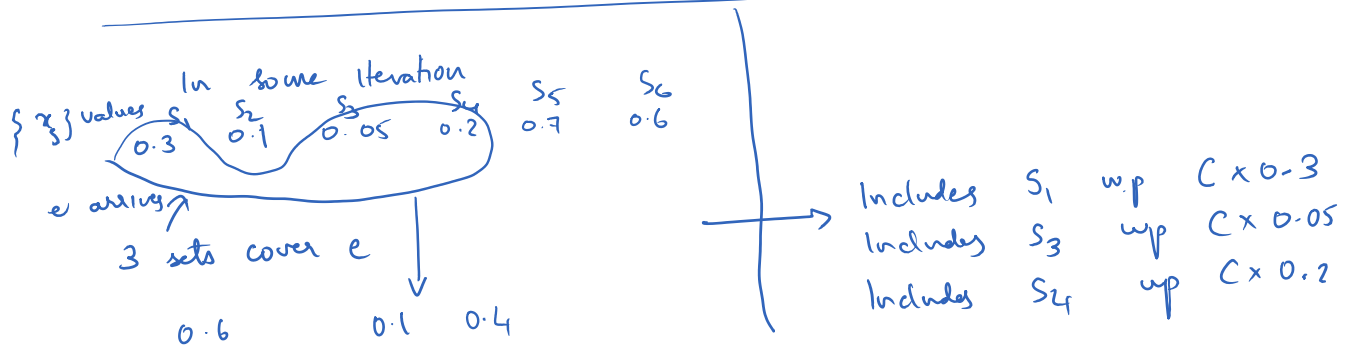
$$\text{Our fractional cost @ time } t = \sum_{\text{all sets}} x_S^t \leq \sum_e \tilde{y}_e^t \quad \textcircled{a}$$

$$\& \left\{ \frac{\tilde{y}_e^{(t)}}{\log m + 1} \right\} \text{ is feasible for dual program.} \quad \textcircled{b}$$

$\textcircled{a} \& \textcircled{b} \Rightarrow \text{lemma } \textcircled{1} \Rightarrow \text{theorem } \textcircled{1}$

Summary :

- MW inspired algo (more aggressive for more useful sets historically)
- Analyzed Algo cost vs dual solution we constructed (upto $O(\log n)$)
- duality \Rightarrow dual opt \leq Real Set Cover Optimal.



Analysis of online rounding:-

@ time t , $\Pr[\text{Set } S \text{ is included by now}] = C x_s^t$

Look at any element arrived already,

$$E[\# \text{ sets in my rounded sol}^n \text{ which cover } e] \geq C \sum_{S: e \in S} x_s^t \geq C$$

Moreover,

$$E[\text{cost of rounding solution}] = C \sum_{\text{all sets } S} x_s^t.$$

$$\Pr[\text{a fixed elt which has arrived is not covered by rounding solution}] \leq e^{-C} = \frac{1}{n^2}.$$

$$\text{Set } C = 2 \log n.$$

Then ②

Above rounding algo. maintains online set cover, st

@ time t , it covers all elements up to $1 - 1/n$

& has expected cost $\leq O(\log n) \sum x_s^t$

LOWER BOUND EXAMPLE :-

No online algorithm can do better than some α factor.

$$\alpha \cdot 1 = 1^2$$

No online algorithm can do better than some α factor.

$$U = \{e_1, e_2, \dots, e_L\}$$

$$\text{st } L = l^2$$

$$\mathcal{S} = \{ \text{all possible } l\text{-sized subsets of } U \}; |\mathcal{S}| = \binom{L}{l}$$

Bad input sequence :-

present e_1 , algo chooses say $S_1 \ni e_1$

present e_2 not in S_1 , algo has to choose new set, say S_2

present e_3 not in S_1 or S_2 , algo forced to choose S_3

present e_l not in S_1 or S_2 or \dots or S_{l-1} , algo forced to pick S_l

$$\text{Algo cost} = l$$

$$\text{Opt cost for } \{e_1, e_2, \dots, e_l\} = 1 \quad \text{gap} = l = \Omega(n)$$

\Rightarrow Competitive ratio = l . ; Ex: how does this behave with m & n ?

Huge gap b/w offline set cover & online set cover

In offline set cover, Randomized rounding $\leq O(\log n)$.

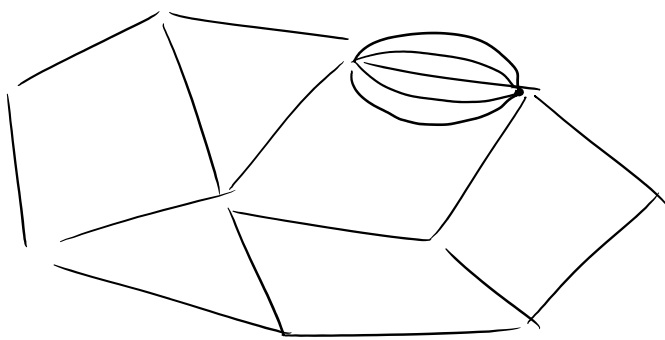
Online Routing

\rightarrow large graph G

\rightarrow Unit capacities on edges

\rightarrow Online requests arrives

source vertex,
destination vertex
unit bandwidth requirement.



Can simulate High Capacity by parallel links.

\rightarrow Goal of Algo :- $\left\{ \begin{array}{l} \text{Fix a path to route this bandwidth} \\ \rightarrow \text{can't change later} \end{array} \right. \leftarrow \text{to ensure QoS.}$

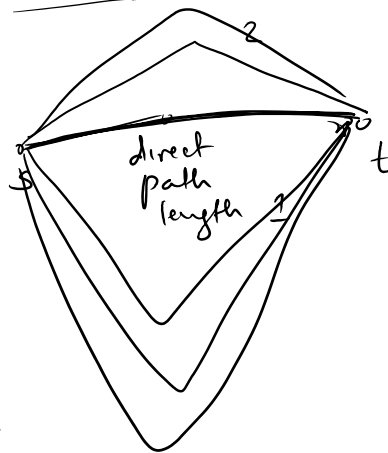
- { a) \rightarrow can't change later - QoS.
 b) Minimize the \downarrow congestion on edges \leftarrow to ensure QoS.
max

Candidate 1 :- choose shortest path. \leftarrow dumb candidate algo !!

1st request $(s, t, 1)$

2nd request $(s, t, 1)$

n th request $(s, t, 1)$



n indirect paths of length 2.

Congestion of SP = n .

Optimal congestion = 1. (using indirect paths of length 2).

Candidate 2 :- give weights to the edges? }
 (current load ??)

Candidate 3 :- Greedily pick a path which minimizes max load after that path is chosen.

\downarrow
 (eg: throw out high-congestion edges & find shortest path)

Exponential weights are the right weights !!

$$w(e) = \left(\frac{3}{2} \right)^{\text{curload}(e)}$$

: Find shortest path !!

\uparrow
 Soft form of max congestion (soft max)

\downarrow
 Nice middleground b/w max congestion & length of path