

Approximation Algorithms

- 'Efficient Algorithms' which give answers 'close' to desired ones.

efficient - polynomial running time

- Why? (1) NP-Hard : VERY hard to find P-time algorithm
NP-Hard optimization

- (2) Sometimes even 'easy' problems do not have 'good enough' ~~solution~~ algorithm.

Eg: EDIT DISTANCE $O(mn)$

Till now no subquadratic ($< n^2$) algo.

So compromise on the algo for fast solution \leftarrow Approx algo.

Why taught by sir?

1. Explain interesting design + analysis techniques beyond AOA.
2. Research :- Fun Math

Evaluation

Drills : 10% submit (email) by next day

HW : 40% (only one which is coding)

\rightarrow groups of 2 or individual

Exams : 50% long exam - half day / entire day
(25% + 25%)

Resources - posted on GC.

Lecture Notes

Optimization Problem

Π is an optimization problem

Π : Set of instances : \mathcal{I}

\mathcal{S} : Set of "feasible" solutions

~~all algorithms~~

$c : \mathcal{S} \rightarrow \mathbb{R}$ [cost function]

$A : \mathcal{I} \rightarrow \mathcal{S}$ which minimizes/ maximizes $c(S)$, $S \in \mathcal{S}$
algo
solution

$\text{opt}(\Pi, \mathcal{I})$: min / max feasible solution for \mathcal{I} on Π

Approximation Algorithms

For a minimization problem Π ,
 A is an α -approximation if

$$\max_{\mathcal{I}} \frac{c(A(\Pi, \mathcal{I}))}{c(\text{opt}(\Pi, \mathcal{I}))} = \alpha \quad [\alpha \geq 1]$$

$\alpha = 1 \rightarrow$ exact algo

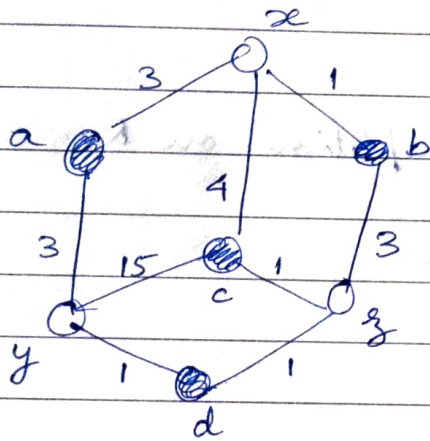
$\alpha > 1 \rightarrow$ strictly approximation algo

For a maximization problem Π ,
 A is an α -approximation if,

$$\min_I \frac{c(A(\Pi, I))}{c(\text{opt}(\Pi, I))} = \alpha \quad [\alpha \leq 1]$$

Example

Steiner Tree



$G = (V, E)$;

$l: E \rightarrow \mathbb{R}_{\geq 0}$

$R \subseteq V$ [Terminals]

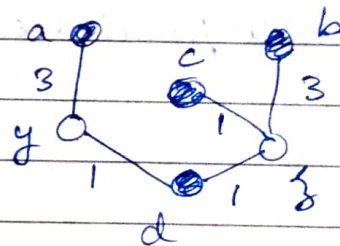
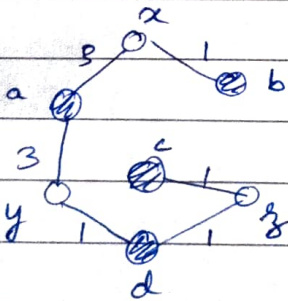
shaded.

Non-mandatory nodes \leftarrow Steiner vertices.

Minimum tree that spans
 all the terminals.

[NP-Hard]

Feasible
 solution



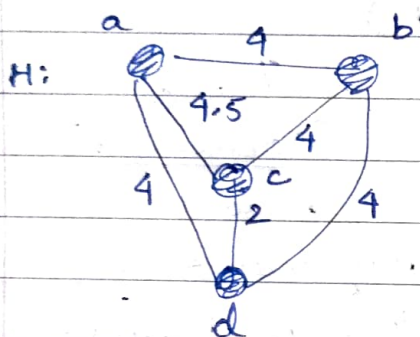
Ex 1: Find an example graph such that for any constant α
 cost of MST+Prune $> \alpha \cdot c(\text{opt}(G))$.

[i.e. MST+Prune is not a constant factor approximation]

MST+Prune \Rightarrow

$\begin{matrix} \text{terminals} \\ \downarrow \\ \text{new cost function} \end{matrix}$
 Given G , create H (R, ω) is a complete graph.

$\omega(u, v) = \text{Shortest path length between } u, v \text{ in } G.$



Metric Completion on R .

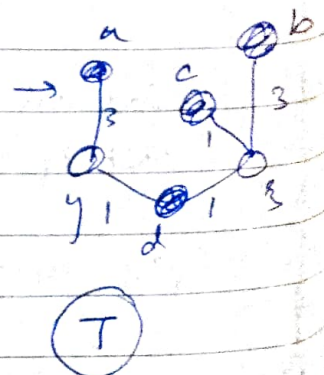
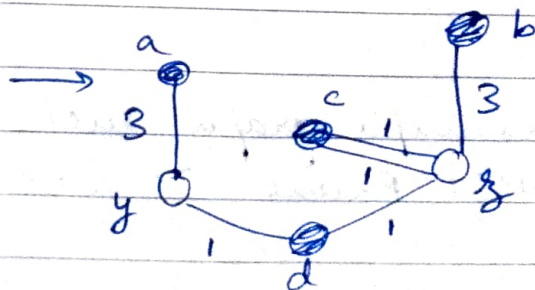
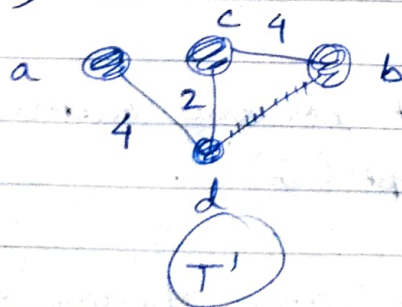
Alg: Create H (All pair shortest path)

Find MST on H

unfurl the terminal edges

delete parallel edges

(MST)



Thm: MST-Heuristic is a 2-approx for Steiner Tree problem

Proof

Let T : Tree returned by the algorithm

T' : intermediate tree

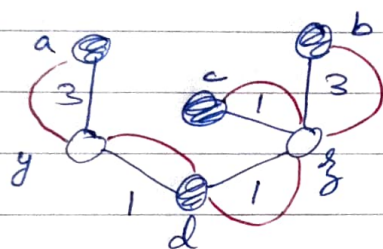
$$l(T) \leq w(T')$$

[the cost can only go down by removing redundant edges]

Enough to prove:

$$w(T') \leq 2 l(\text{opt}(G))$$

Analysis :-



Thought process

- duplicate all edges of optimal solution.
- all vertices will have even degree
- ↓
- will have Eulerian walks
- ↓
- [start and stop at same vertex]
- [each edge traversed once.]

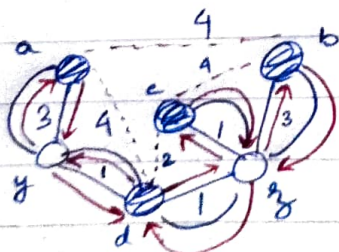
- start a walk from one of the terminals.

skip Steiner nodes.

skip used nodes

end at start node.

MST \subset subgraph.



$$\begin{aligned} \text{Cost of dotted graph } H' \\ \leq 2 \cdot l(\text{opt}(G)) \end{aligned}$$

Also, H' spans all of R and is a subgraph of H .

~~Also~~

~~MST~~ H' spans all of R (MST $\subset T' \subset$ dotted graph H')

$$\begin{aligned} H' \subset H \quad \therefore \quad \underset{\text{subgraph}}{w(T')} &\leq \text{cost of dotted graph } H' \\ \text{MST} \end{aligned}$$

Ex 2:- There exists a graph s.t.

$$\text{MST-heuristic} \approx 2 \cdot c(\text{opt}(G)).$$