

APPROXIMATION ALGORITHMS

ADVANCED ALGORITHMS

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CLRS CH. 35

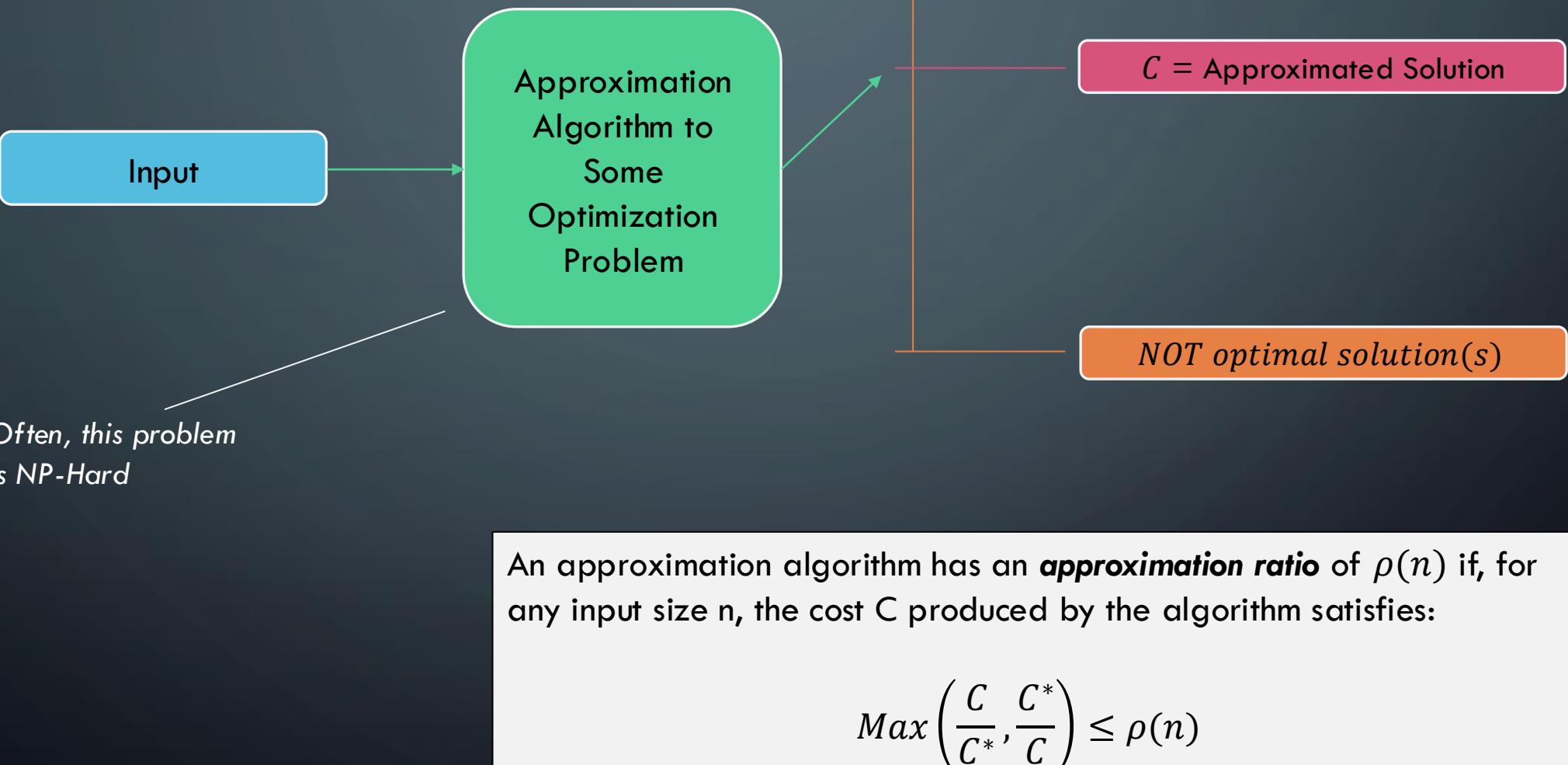
ADVANCED TREE STRUCTURES

In this deck we will look at:

- **Approximation Ratios**
- **Vertex Cover Approximation**
- **TSP Approximation**
- **Set-Covering Approximation**

INTRODUCTION: APPROXIMATION

APPROXIMATION RATIOS



SOME DEFINITIONS

$\rho(n)$ -Approximation Algorithm

An approximation algorithm that achieves an approximation of $\rho(n)$ on all inputs

Approximation Scheme

An approximation algorithm that takes as input the instance of the problem and a value $\epsilon > 0$ such that for any fixed ϵ , the scheme is a $(1 + \epsilon)$ -approximation algorithm.

Polynomial Time Approximation Scheme

A scheme is polynomial time if, for any fixed $\epsilon > 0$, the algorithm runs in polynomial time in terms of the input size n

Fully Polynomial Time Approximation Scheme

A scheme is fully polynomial time if it has a runtime that is polynomial in terms of $\frac{1}{\epsilon}$.

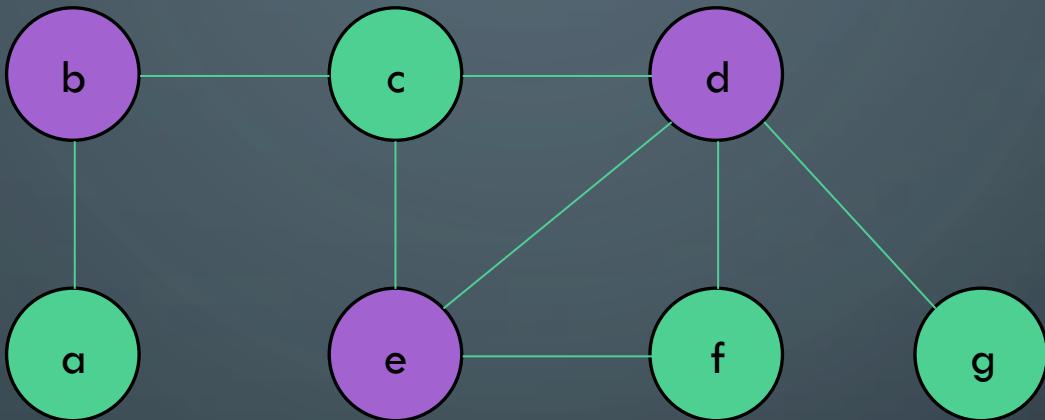
For example, a fully polynomial time approximation scheme might have runtime $\Theta\left(\left(\frac{1}{\epsilon}\right)^2 n^2\right)$

VERTEX-COVER

RECALL VERTEX-COVER

Given a Graph $G = (V, E)$, find the smallest set of nodes $V' \subset V | \forall_{e=(u,v) \in E} (u \in V' \vee v \in V')$

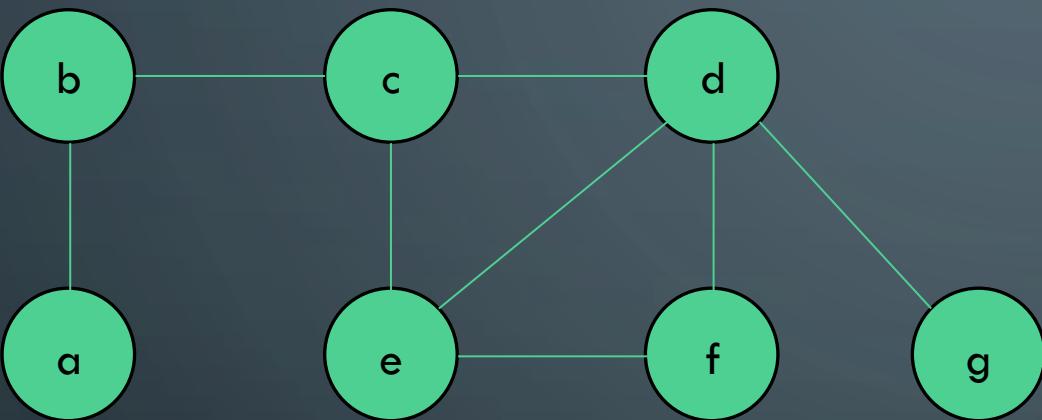
This problem is NP-Hard



In other words, find the smallest set of nodes that touches every edge at least once

The purple nodes are the smallest solution for this graph

PROPOSED ALGORITHM FOR VERTEX-COVER



Vertex-Approximation(G):

$C = \emptyset$

$E' = G.E$

while E' not empty:

let $e = (u, v)$ be arbitrary edge from E'

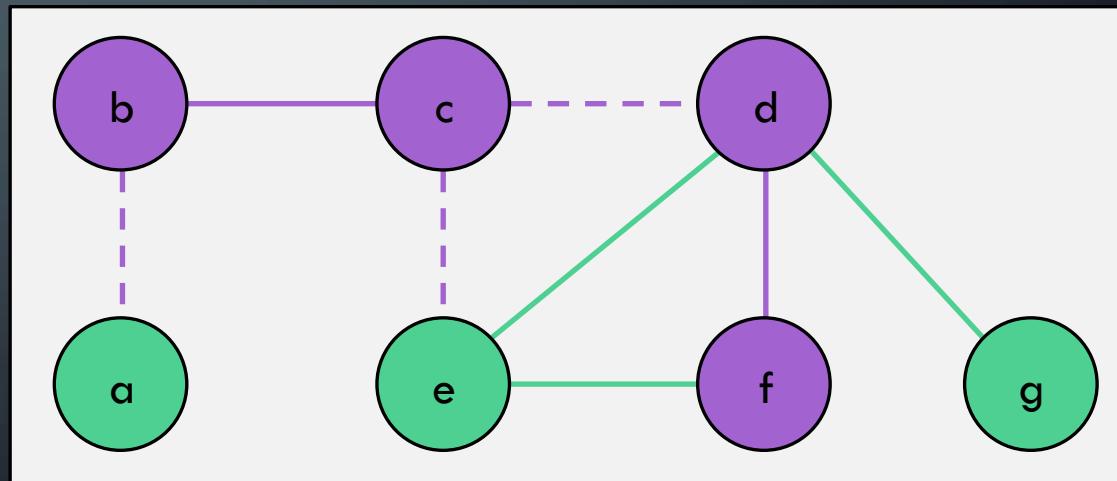
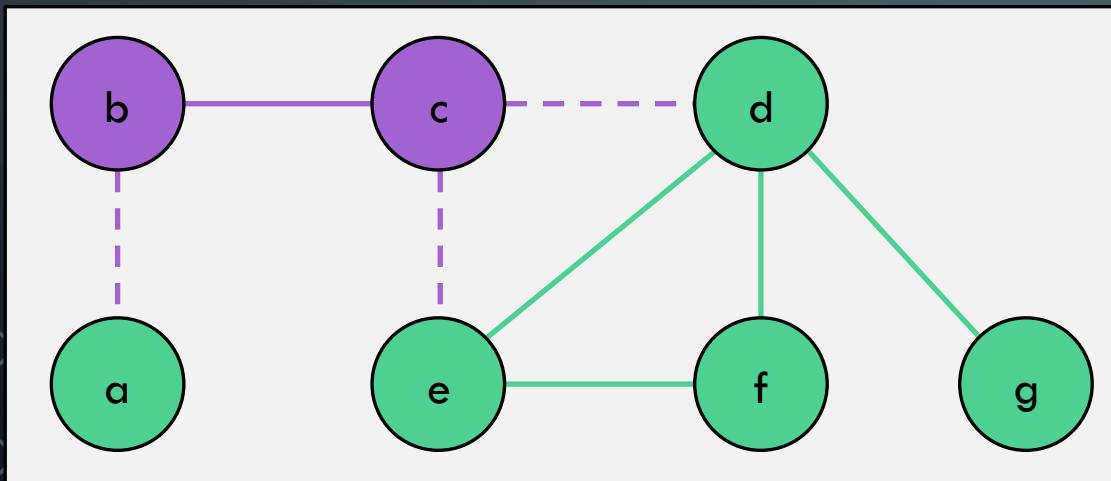
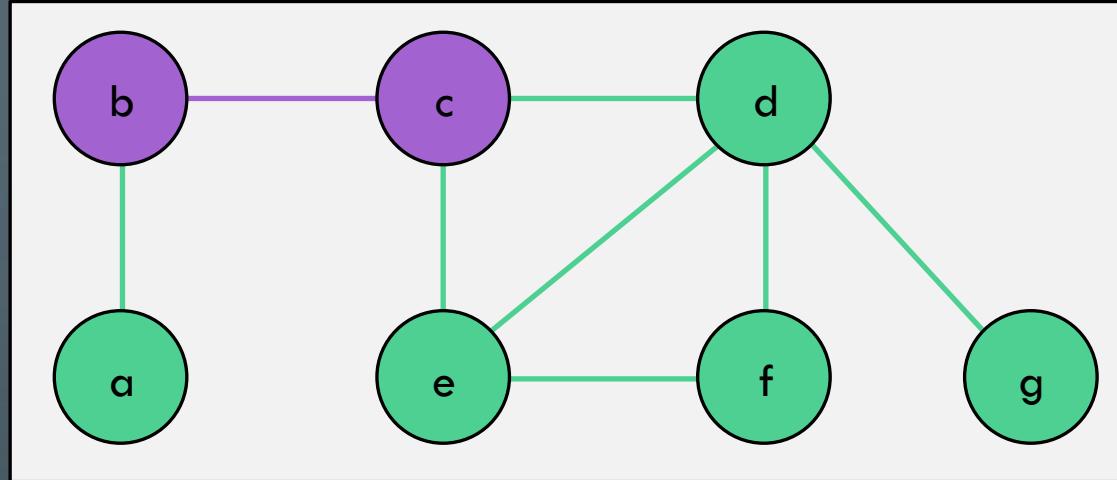
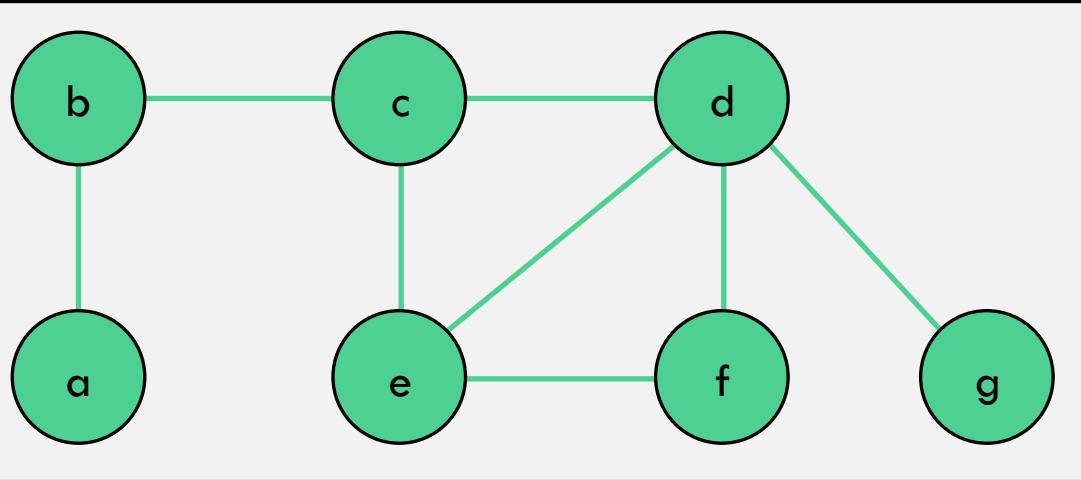
$C = C.append(\{u, v\})$

remove all edges in E' incident on u or v

return C

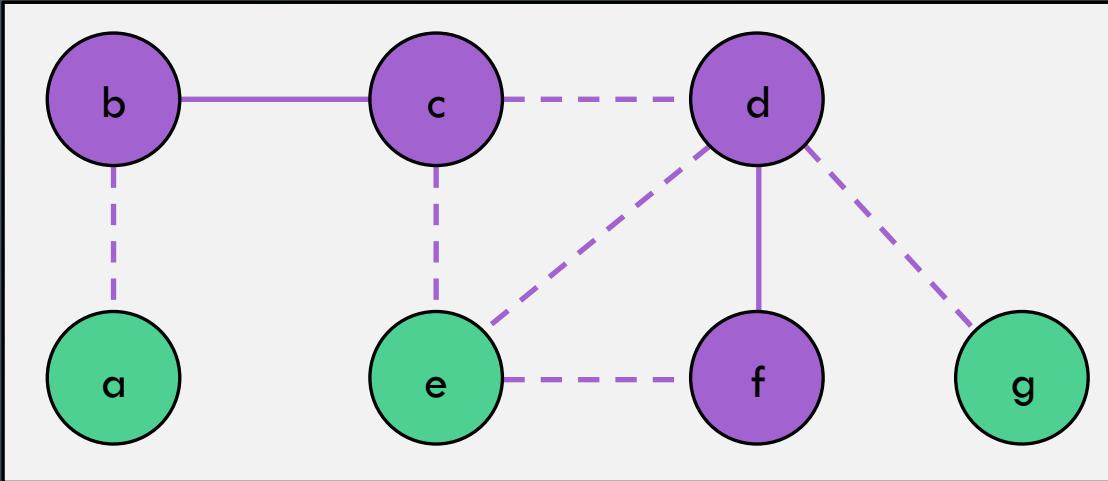
*Run this on the given graph a few times.
What is the approximation ratio? Do we
have a guaranteed ratio here?*

SAMPLE EXECUTION

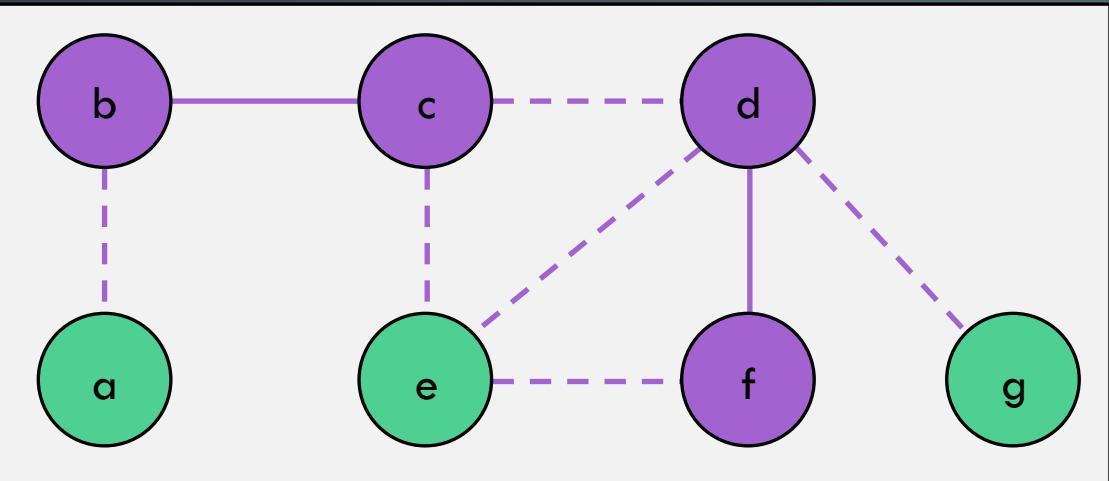


...last step is on next slide

SAMPLE EXECUTION



SAMPLE EXECUTION



Vertex-Approximation(G):

$C = \emptyset$

$E' = G.E$

while E' not empty:

let $e = (u,v)$ be arbitrary edge from E'

$C = C.append(\{u,v\})$

remove all edges in E' incident on u or v

return C

This is just one way it could have turned out. Any thoughts on the limits of "how bad" or "how good" the approximation will be?

APPROX-VERT-COVER: ANALYSIS

Runtime is polynomial? Yes, it is $O(|E|^2)$

Vertex-Approximation(G):

```
C = []
E' = G.E
while E' not empty:
    let e=(u,v) be arbitrary edge from E'
    C = C.append(e)
    remove all edges in E' incident on u or v
return C
```

For each edge e we choose on this line, the solution definitely includes u or v , we always add both

We remove as many extra edges as we can here, but of course there could be 0 each time.

APPROX-VERT-COVER: ANALYSIS

Claim: Vertex-Approximation is a 2-approximation of vertex cover

Vertex-Approximation(G):

$C = \emptyset$

$E' = G.E$

while E' not empty:

 let $e=(u,v)$ be arbitrary edge from E'

$C = C.append(\{u,v\})$

 remove all edges in E' incident on u or v

return C

Proof Part 1 (Correctness): The nodes returned by this algorithm are a valid vertex cover

This is self-evident. Every edge in E is either chosen in line 4 and both its nodes added to the solution OR removed in line 6 being incident to u or v .

Thus, every edge is incident on some node in the set C because we do not terminate until E' is empty.

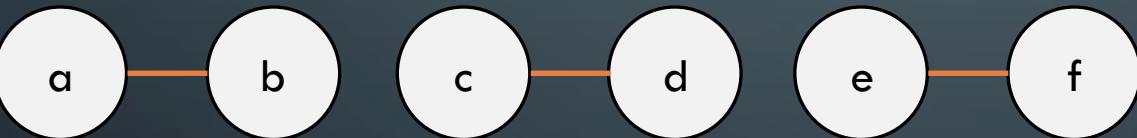
The algorithm returns a valid vertex-cover

APPROX-VERT-COVER: ANALYSIS

Claim: Vertex-Approximation is a 2-approximation of vertex cover

Proof Part 2 (Ratio): The size of C is no more than twice the size of the optimal vertex-cover

Consider the set of edges A chosen in line 4 of the algorithm. Because other incident edges are removed in line 6, the lines in A form a set of disjoint pairs



There may be edges between these pairs (not shown), but the vertex cover MUST include at least one node from each pair. Thus, we get the inequality $|C^*| \geq |A|$

Separately, note that our vertex cover returned always returns exactly two nodes for each edge in A , thus:

$$|C| = 2|A|$$

Combining these equations we get:

$$\begin{aligned}|C| &= 2|A| \\ |C| &\leq 2|C^*|\end{aligned}$$

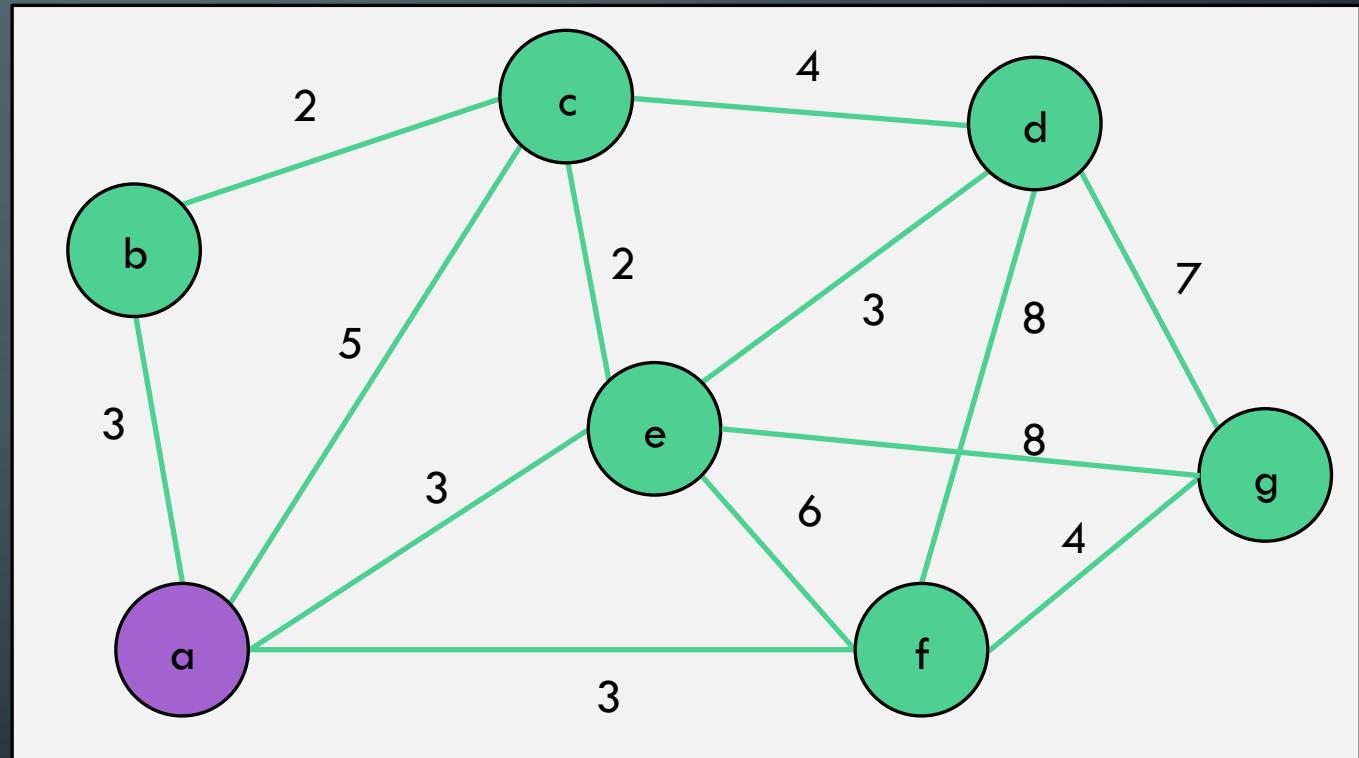
TRAVELING SALESPERSON

RECALL TRAVELING SALESPERSON PROBLEM

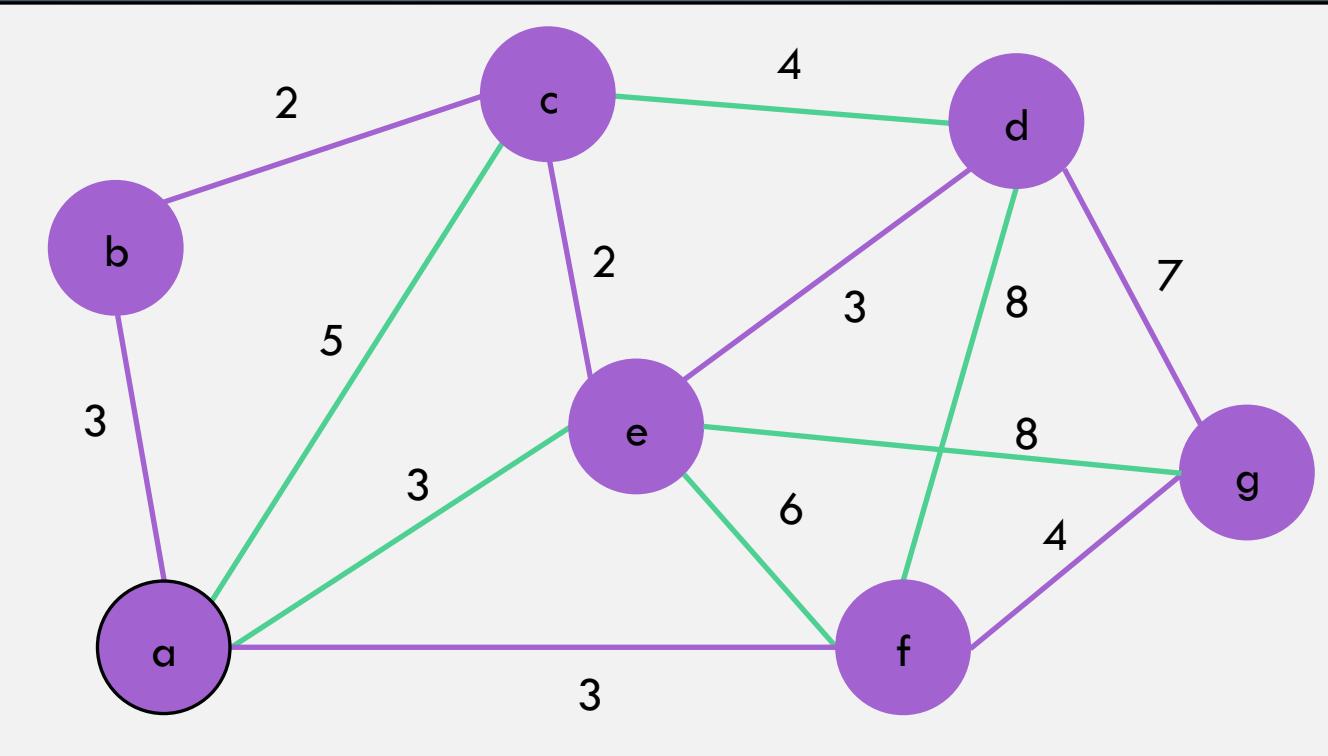
Recall the traveling salesperson problem: Given a graph G and start node, what is the minimum cost tour (visit every node exactly once then return to start)

For our approximation, we are going to assume that the triangle inequality holds.
For any three nodes u, v, w :

$$c(u, w) \leq c(u, v) + c(v, w)$$



RECALL TRAVELING SALESPERSON PROBLEM



Total Cost of this tour is: $3 + 2 + 2 + 3 + 7 + 4 + 3 = 24$

APPROXIMATION FOR TSP

```
TSP-Approx(G, s):  
    Select any vertex r in G.V  
    MST T = prims-mst(G, r)  
    prev = T.preorder().first  
    for v in T.preorder().second : ...  
        add (prev, v) to solution  
    add (prev, s) to solution  
    return solution
```

Basic Idea:

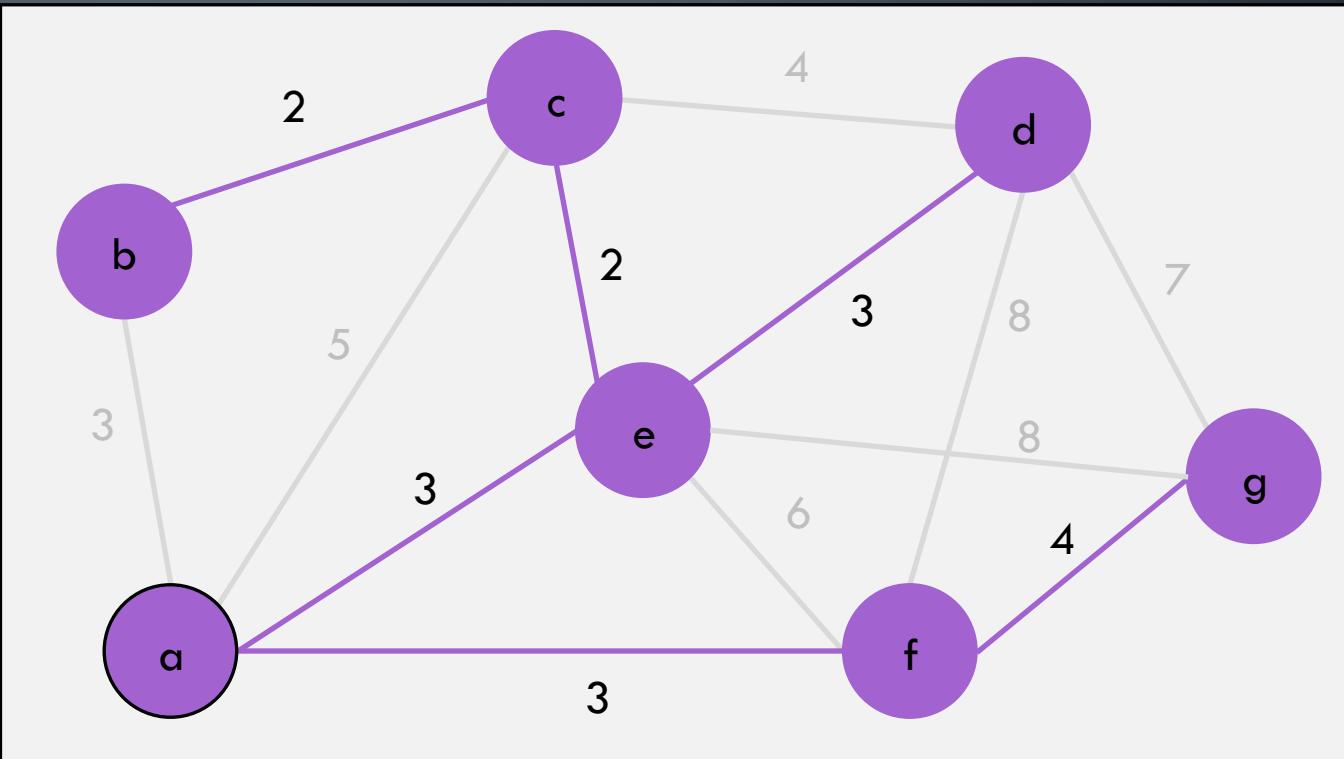
Calculate minimum spanning tree of G . Use pre-order traversal of this MST to define the tour for TSP

APPROXIMATION FOR TSP

```
TSP-Approx(G, s):  
    Select any vertex R in G.V  
    MST T = prims-mst(G, r)  
    prev = T.preorder().first  
    for v in T.preorder().second : ...  
        add (prev, v) to solution  
    add (prev, s) to solution  
    return solution
```

Step 1:

Calculate MST of G (let's root it at a)

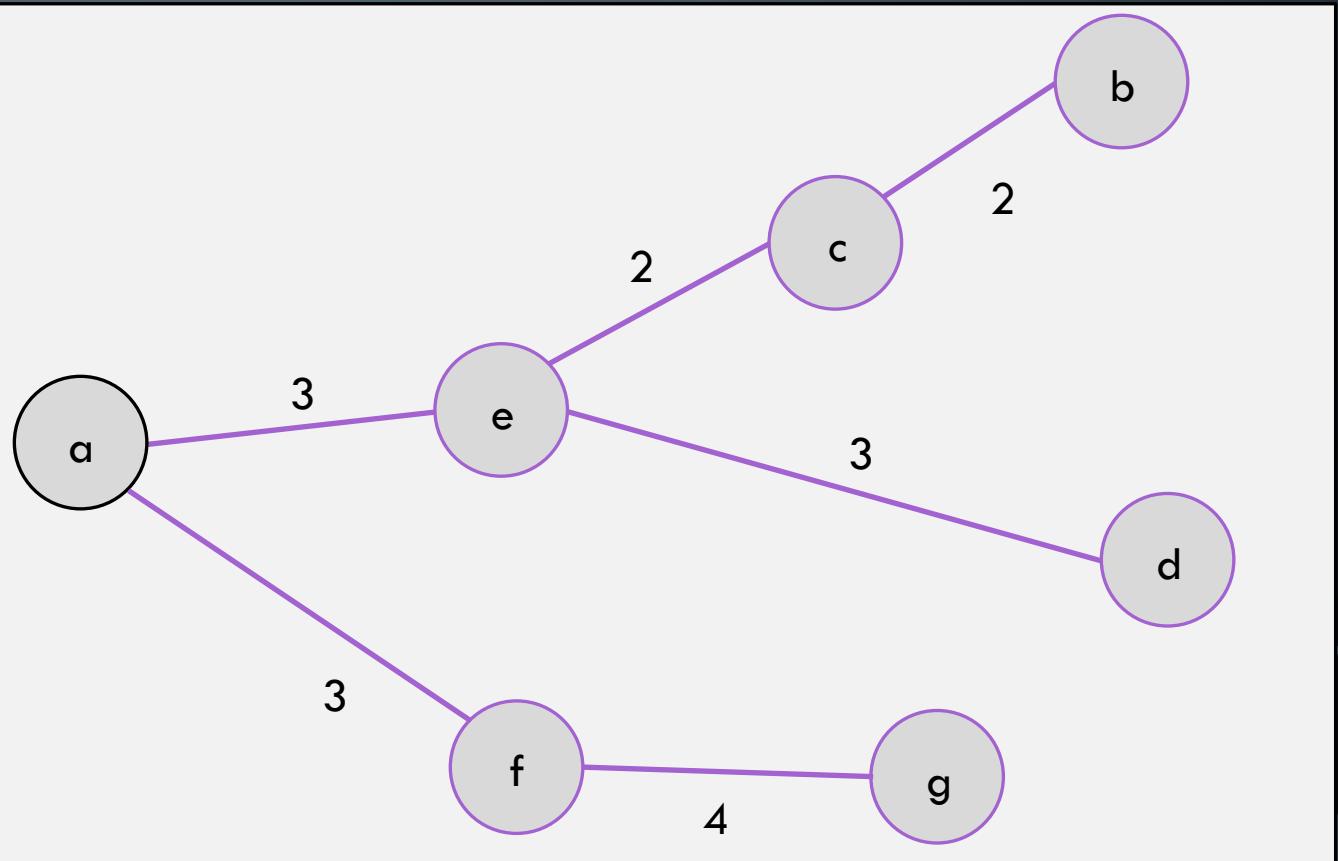


APPROXIMATION FOR TSP

```
TSP-Approx(G, s):  
    Select any vertex R in G.V  
    MST T = prims-mst(G, r)  
    prev = T.preorder().first  
    for v in T.preorder().second : ...  
        add (prev, v) to solution  
    add (prev, s) to solution  
    return solution
```

Step 1.5:

Re-arrange the MST to make it easier to view, remove unused edges

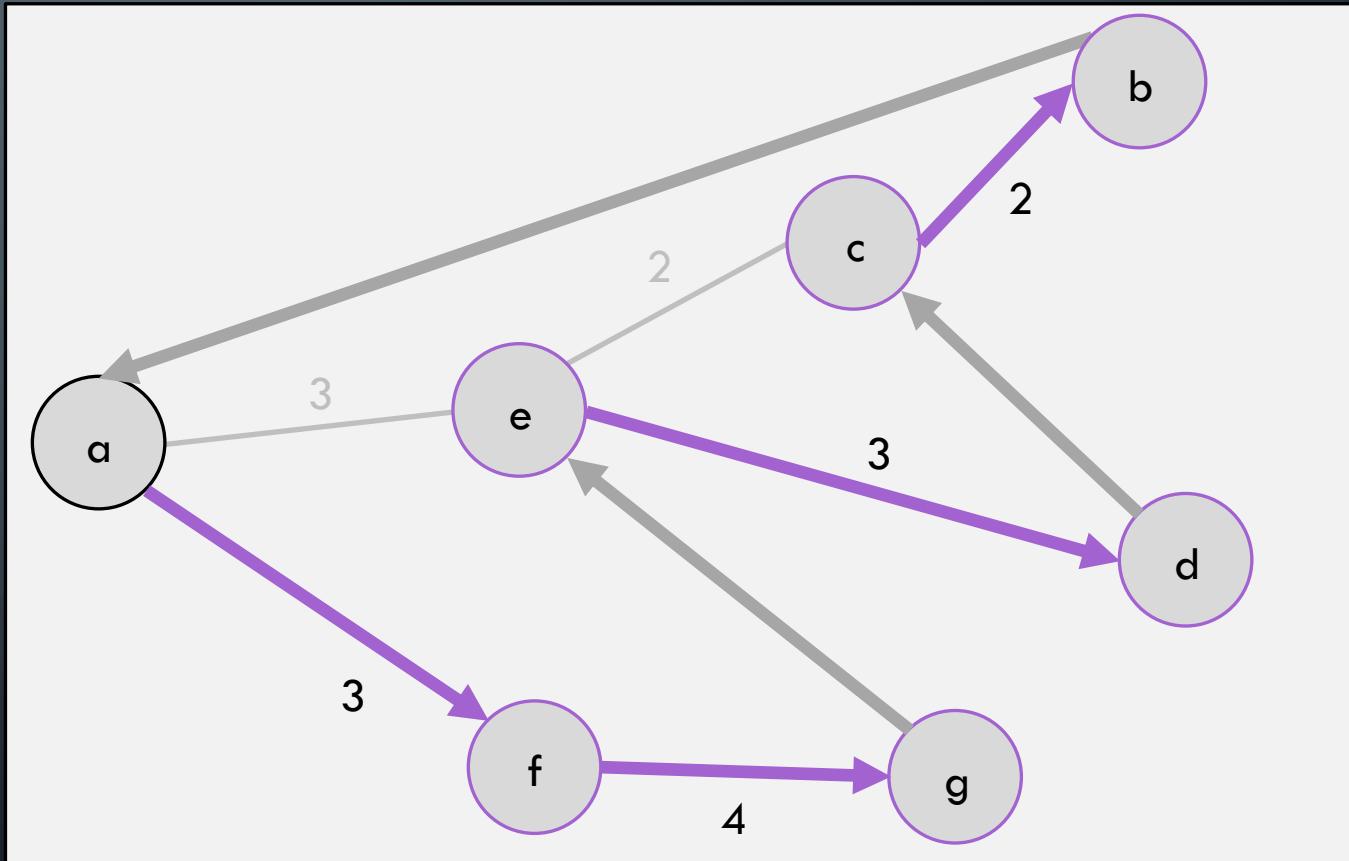


APPROXIMATION FOR TSP

```
TSP-Approx(G, s):  
    Select any vertex R in G.V  
    MST T = prims-mst(G, r)  
    prev = T.preorder().first  
    for v in T.preorder().second : ...  
        add (prev, v) to solution  
        add (prev, s) to solution  
    return solution
```

Step 2:

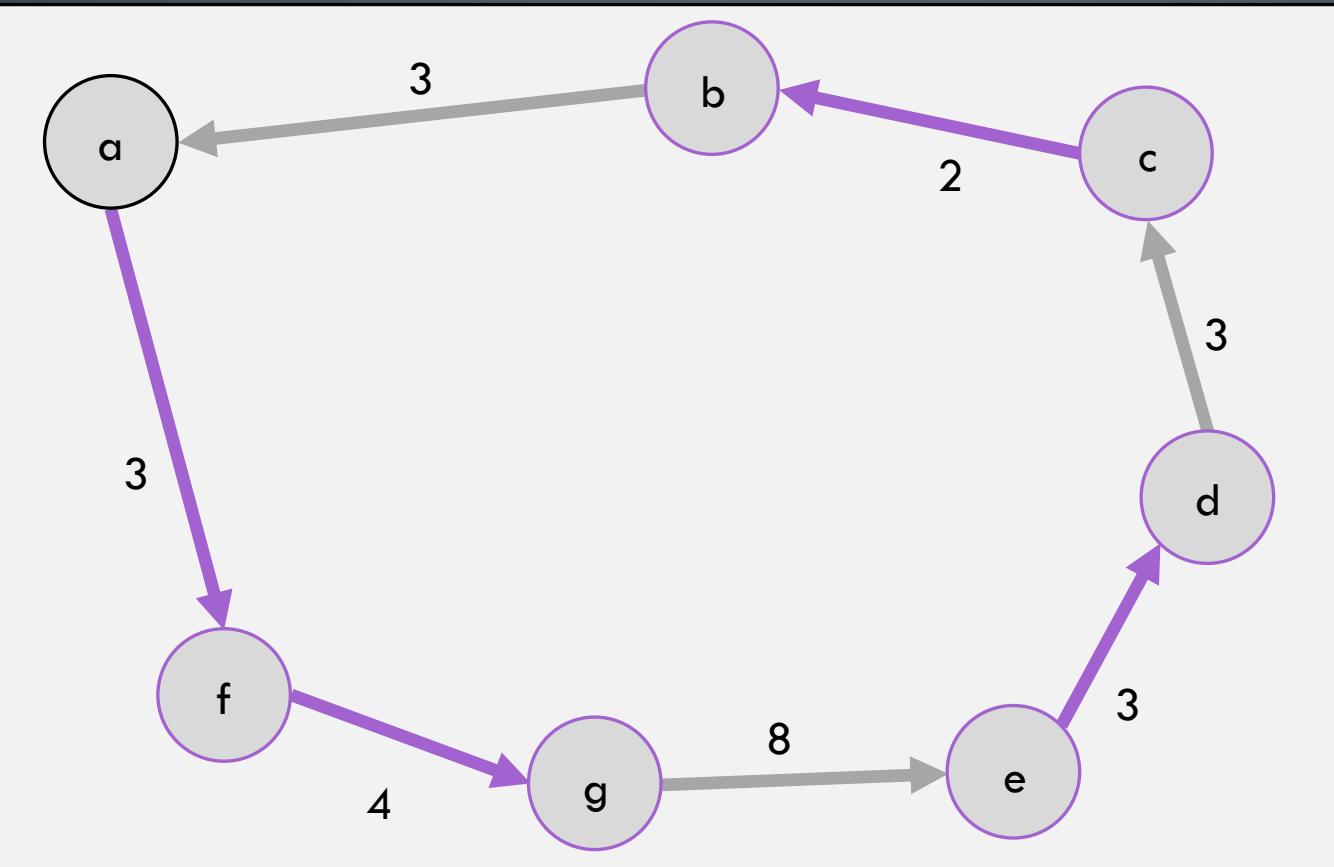
Tour is just the pre-order traversal of the MST



APPROXIMATION FOR TSP

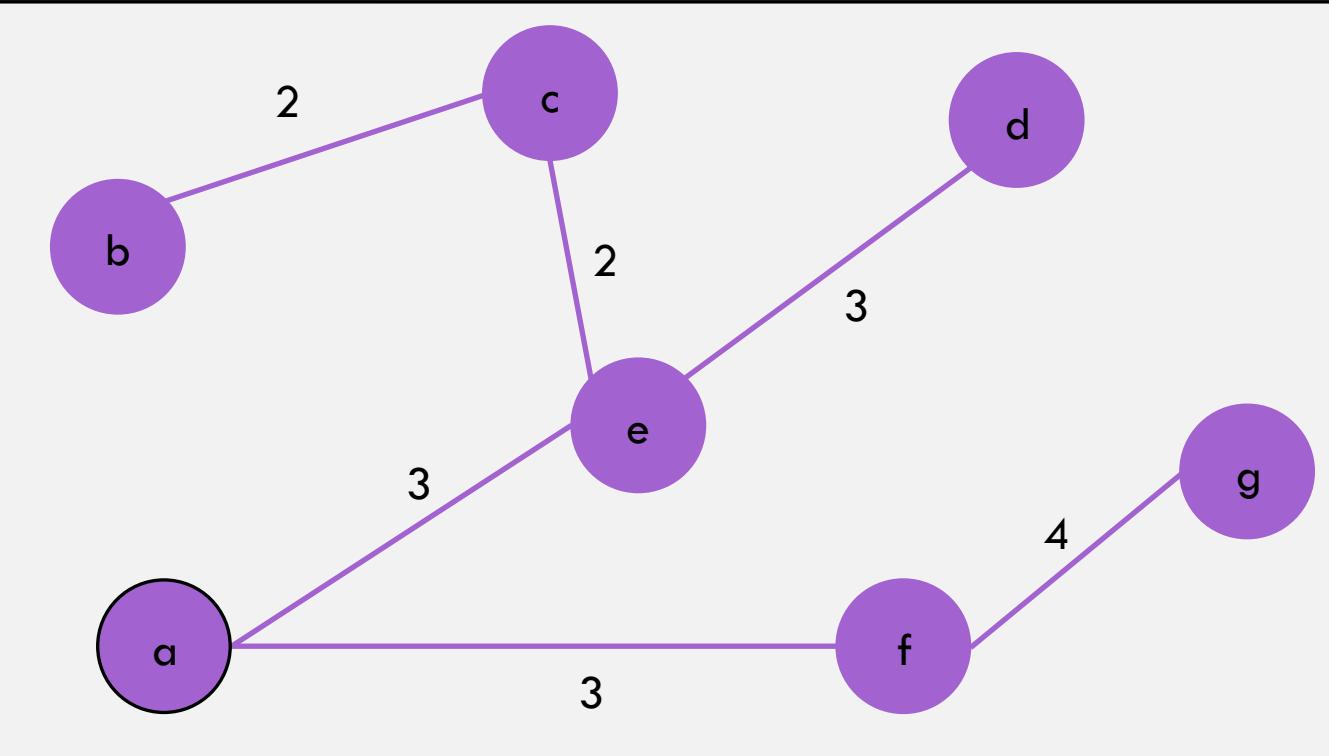
Rearrange to show tour:

Cost is $3 + 4 + 8 + 3 + 3 + 2 + 3 = 26$



How good / bad can this approximation be though?

APPROXIMATION FOR TSP: ANALYSIS



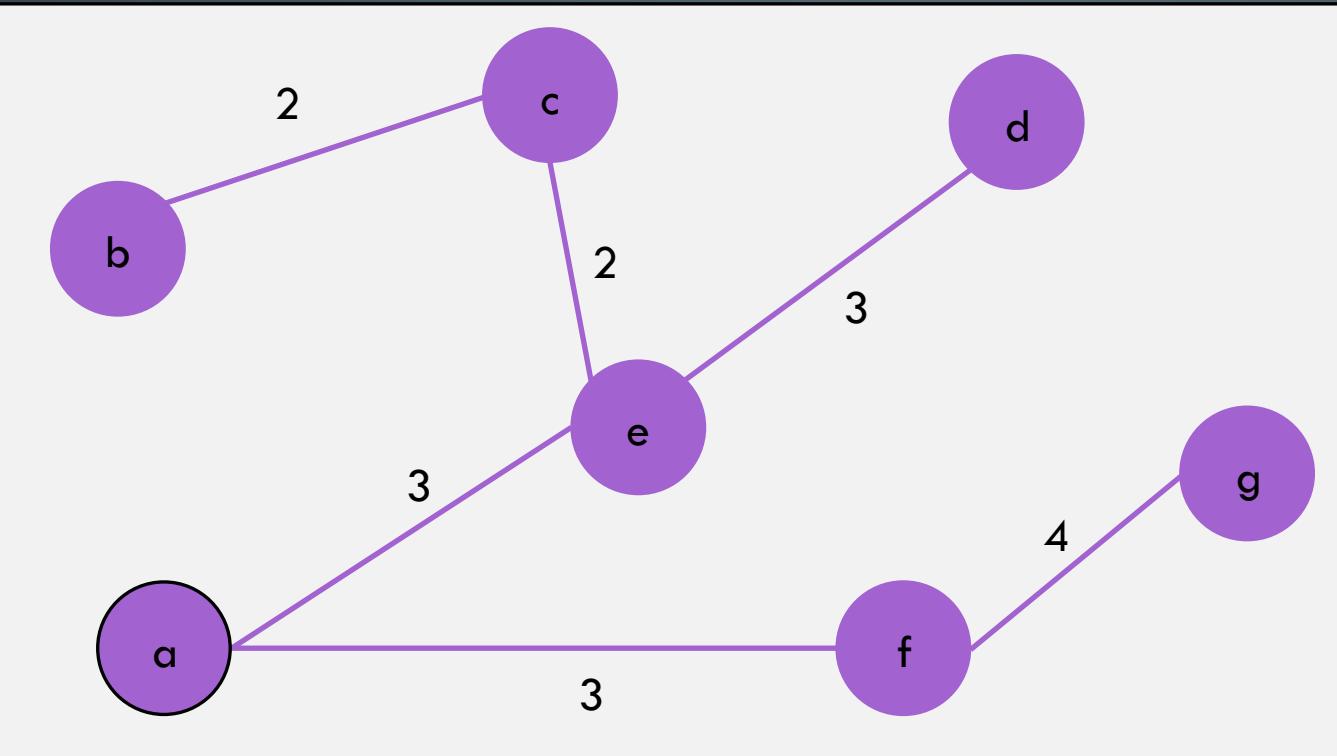
Observation 1:

The minimum spanning tree is a lower bound for the optimal tour.
Why?

So we get the inequality:

$$c(T) \leq c(H^*)$$

APPROXIMATION FOR TSP: ANALYSIS



Observation 2:

Consider the FULL WALK of T . Notice that the full walk traverses each edge exactly twice.

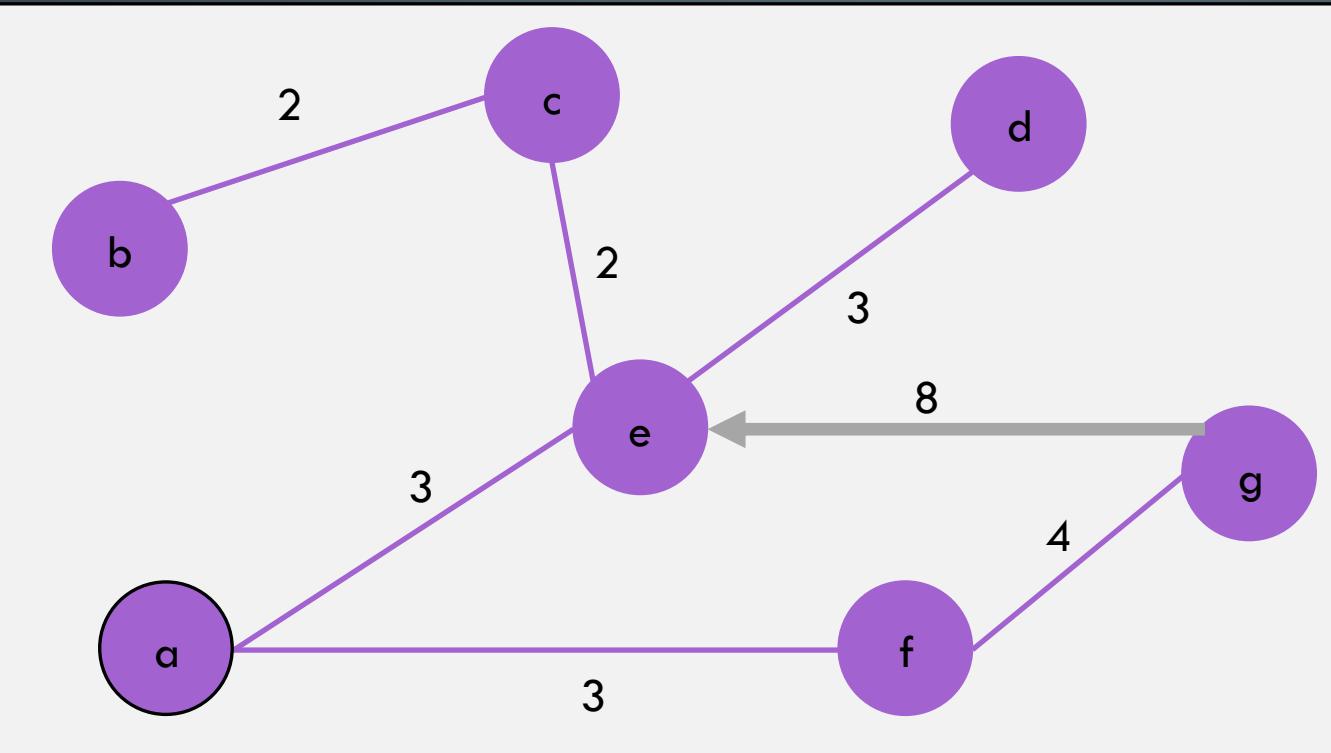
Full walk is $(a \text{ } f \text{ } g \text{ } f \text{ } a \text{ } e \text{ } d \text{ } e \text{ } c \text{ } b \text{ } c \text{ } e \text{ } a)$

So we get the inequality:

$$\begin{aligned} c(W) &= 2c(T) \\ 2c(T) &\leq 2c(H^*) \end{aligned}$$

But this walk is not a tour
(it is an mst full walk),
what about the tour that
gets returned by the
algorithm?

APPROXIMATION FOR TSP: ANALYSIS



Observation 3:

The cross edges never increase the cost of the “walk”

Consider the edge (g,e) with cost 8

The full walk goes from $(g \rightarrow f \rightarrow a \rightarrow e)$
Direct edge goes from $(g \rightarrow e)$

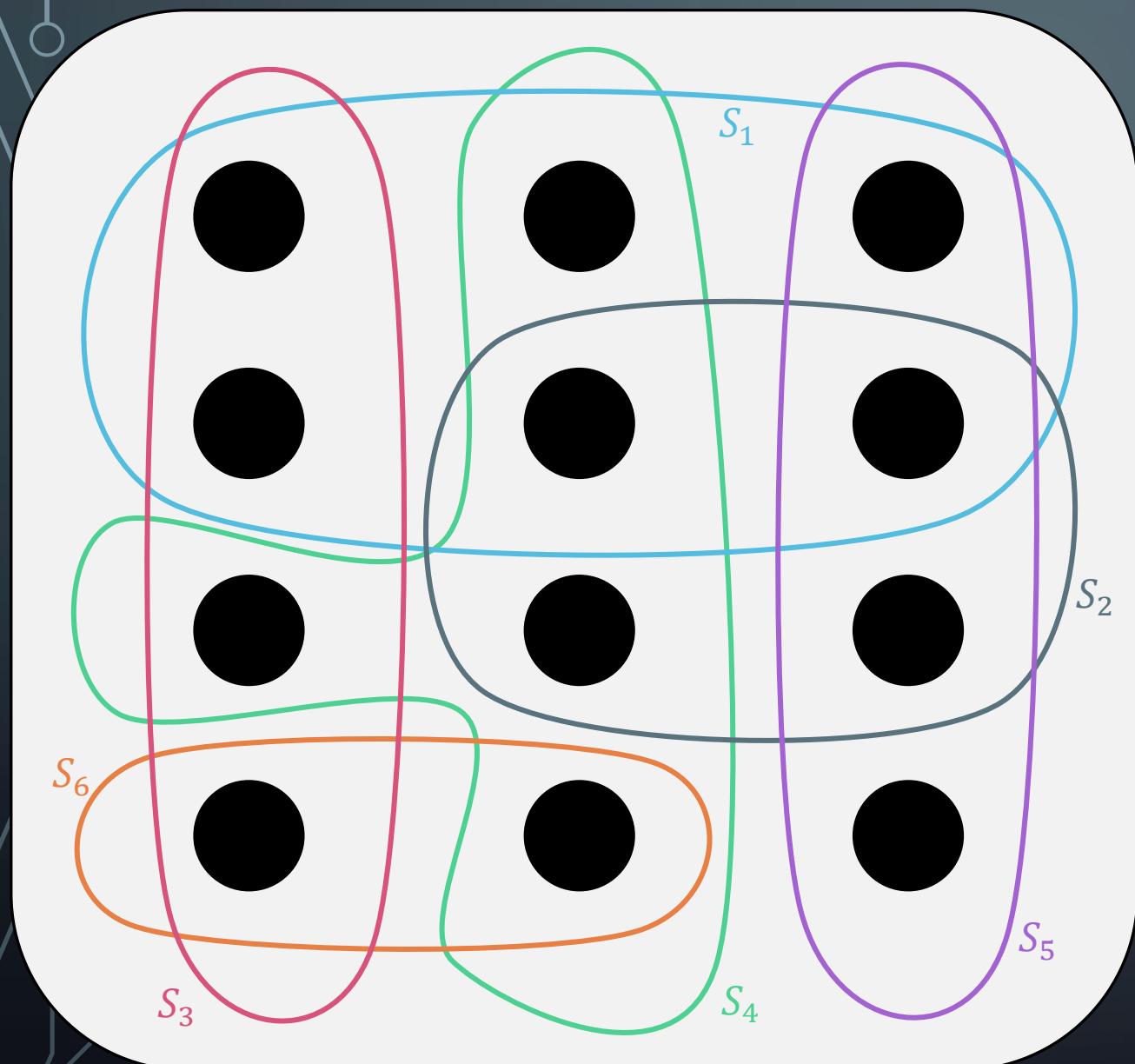
Due to triangle inequality:
 $c(g,e) \leq c(g,f) + c(f,a) + c(a,e)$
 $8 \leq 10$

Thus,

$$c(H) \leq c(W) \leq 2c(H^*)$$

SET COVER

SET COVER PROBLEM



Set Cover Problem Statement

INPUT:

A finite set X (black nodes in image)
A family F or subsets of X | $X = \bigcup_{S \in F} S$

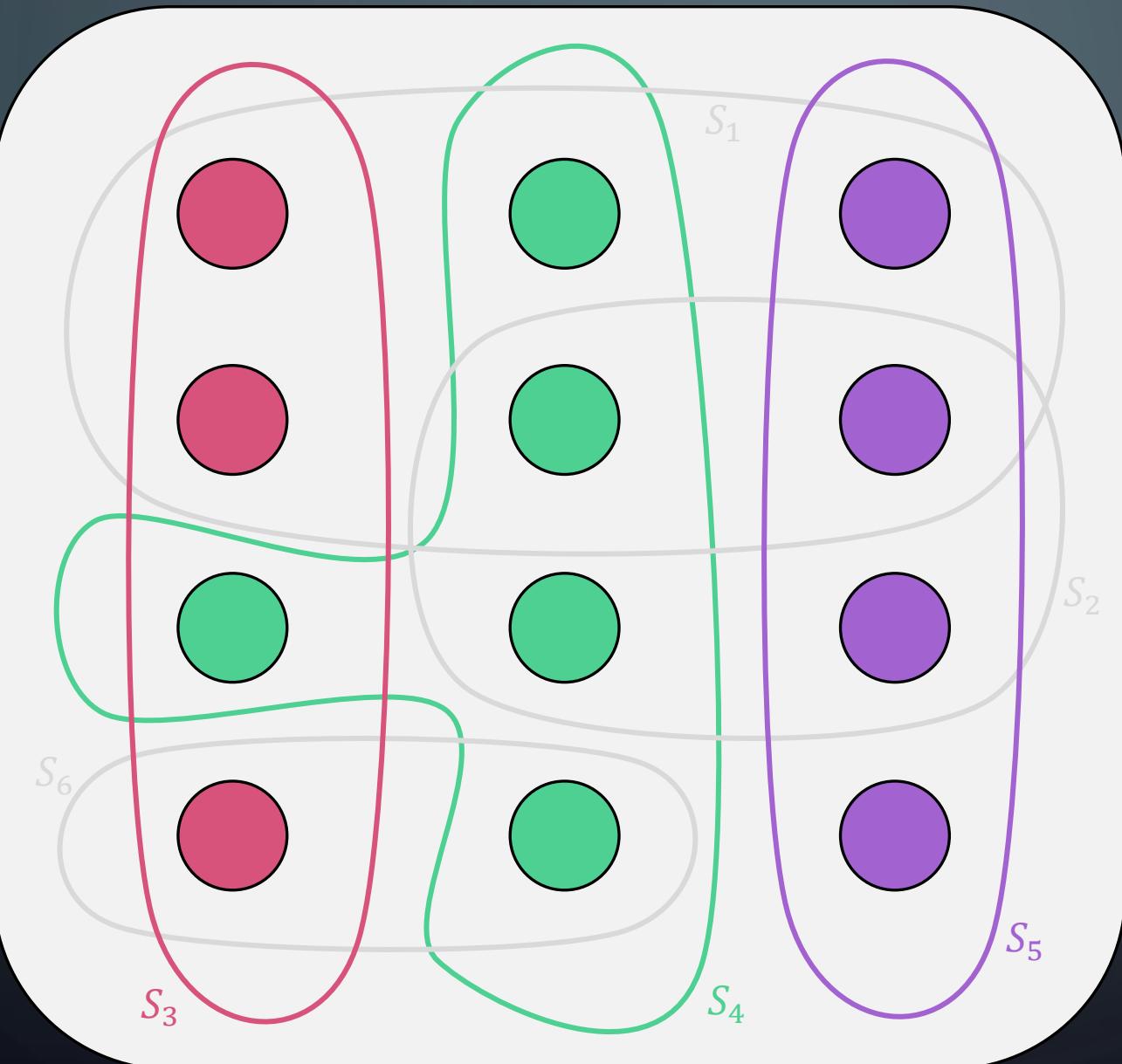
A subset S is said to cover its elements

OUTPUT:

Minimum size subset $C \subseteq F$ such that:
 $X = \bigcup_{S \in C} S$

In other words, find the
smallest group of subsets that
include all the elements.

SET COVER PROBLEM



Minimum cover is called C^* , for this case, C^* is shown and $|C^*| = 3$

Set Cover is also NP-Hard, by the way.

APPROXIMATING SET COVER

How good of a cover
will this produce? Any
“gut” reactions?

What is the runtime
of this method?

A greedy approximation for set cover

Greedy-Set-Cover(X, F):

$$U = X$$

$$C = \emptyset$$

while $U \neq \emptyset$:

 select an $S \in F$ that maximizes $|S \cap U|$

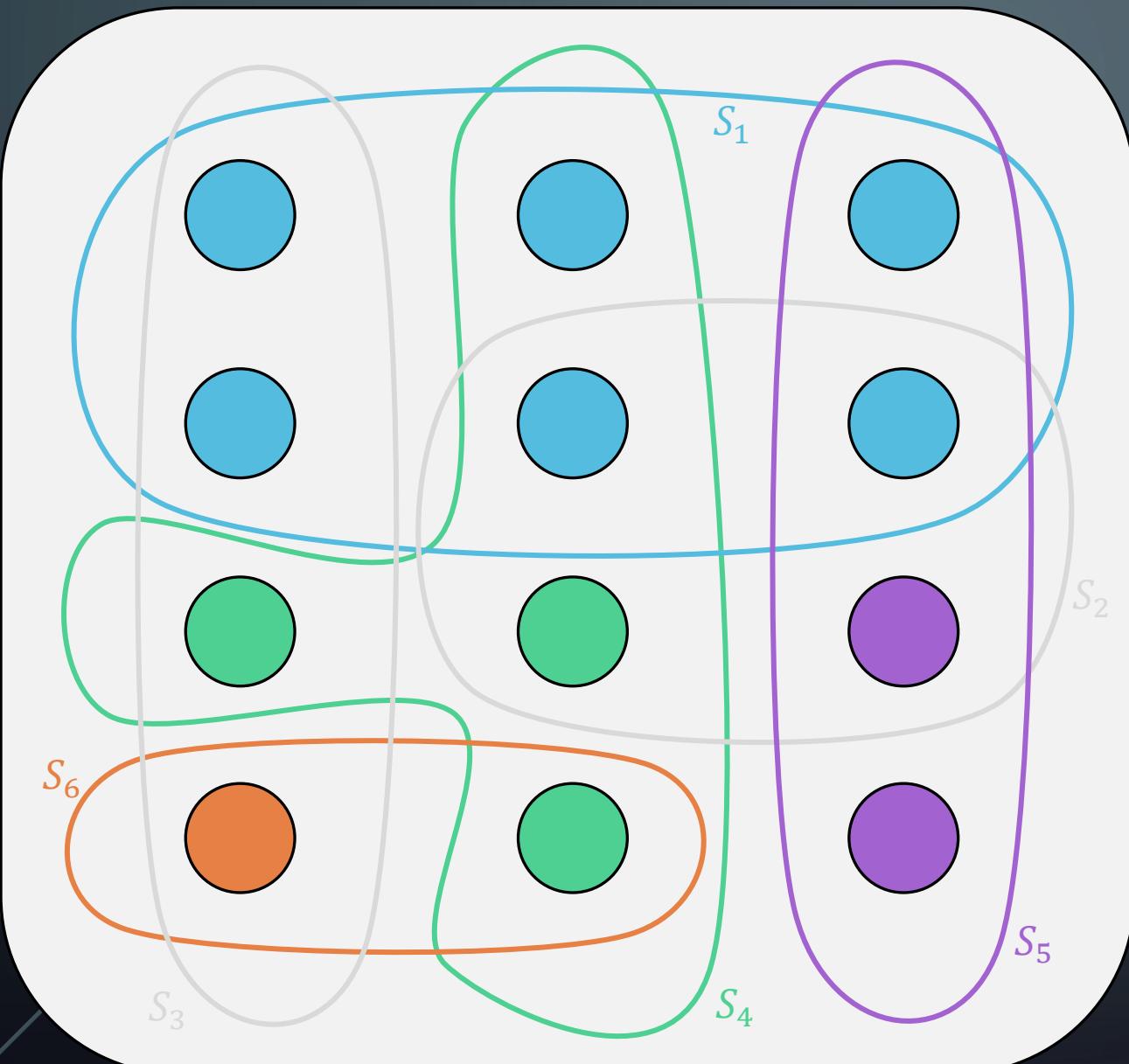
$$U = U - S$$

$$C = C \cup \{S\}$$

return C

... in other words, always select the subset that adds the most new (currently uncovered) elements to the cover

APPROXIMATING SET COVER



A greedy approximation for set cover

Greedy-Set-Cover(X, F):

$U = X$

$C = \emptyset$

while $U \neq \emptyset$:

 select an $S \in F$ that maximizes $|S \cap U|$

$U = U - S$

$C = C \cup \{S\}$

return C

Execution on test case gives:

Add S₁ first (adds six nodes to cover)

Add S₄ next (adds three nodes to cover)

Add S₅ next (adds two nodes to cover)

Add S₆ last (adds one node to cover)

$|C| = 4 > |C^*| = 3$

ANALYSIS

Theorem:

Greedy-Set-Cover(X, F) is a $\rho(n)$ -approximation of set-cover where
$$\rho(n) = H(\max(|S| : S \in F))$$

$H()$ is a harmonic number
(see next slide)

ASIDE: HARMONIC NUMBERS

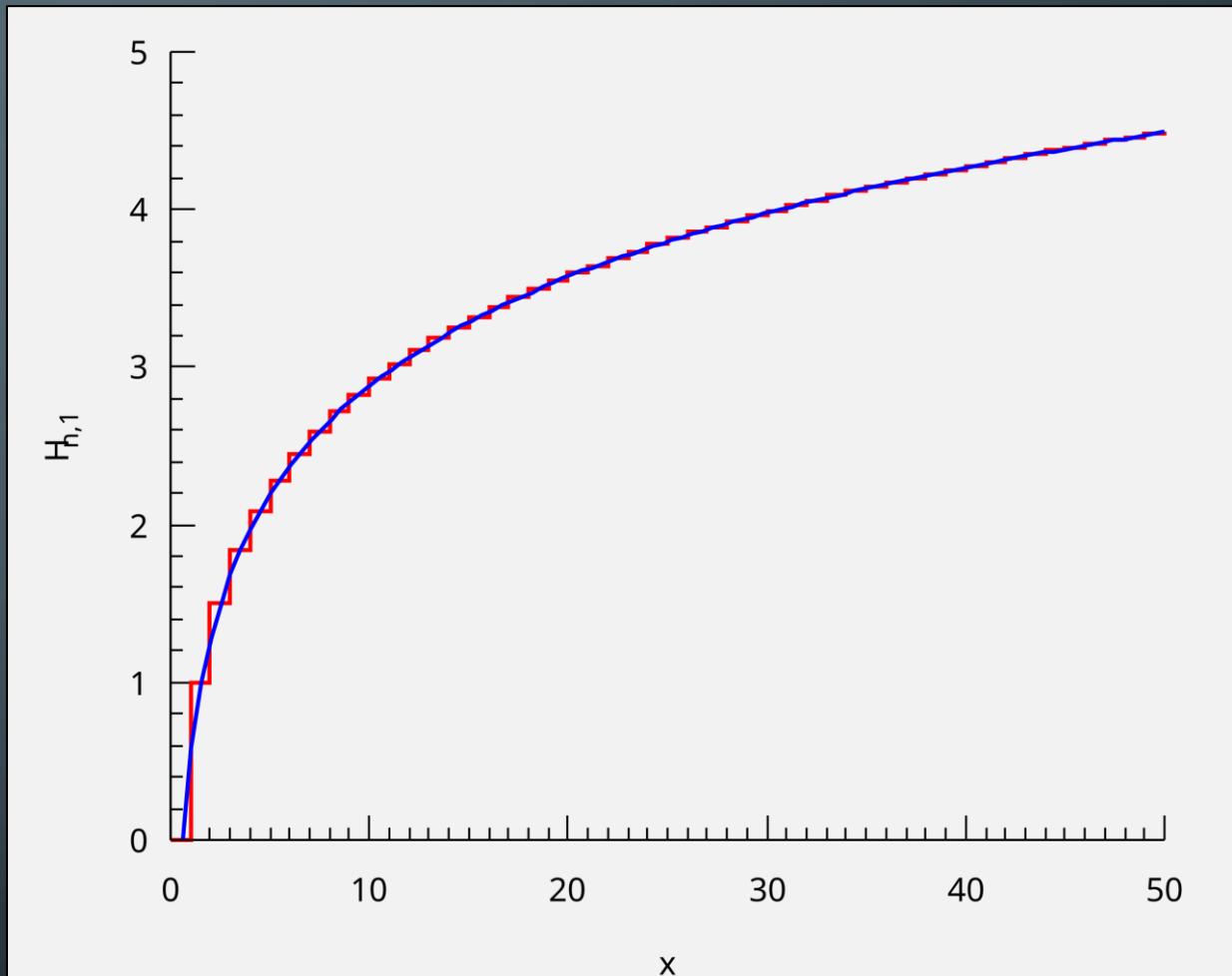
Harmonic Number:

$$H(d) = H_d = \sum_{i=1}^d \frac{1}{d} = 1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots + \frac{1}{d}$$

Why? Well set cover algorithm seems to have a similar pattern. You keep adding a percentage of nodes that are left on each iteration.

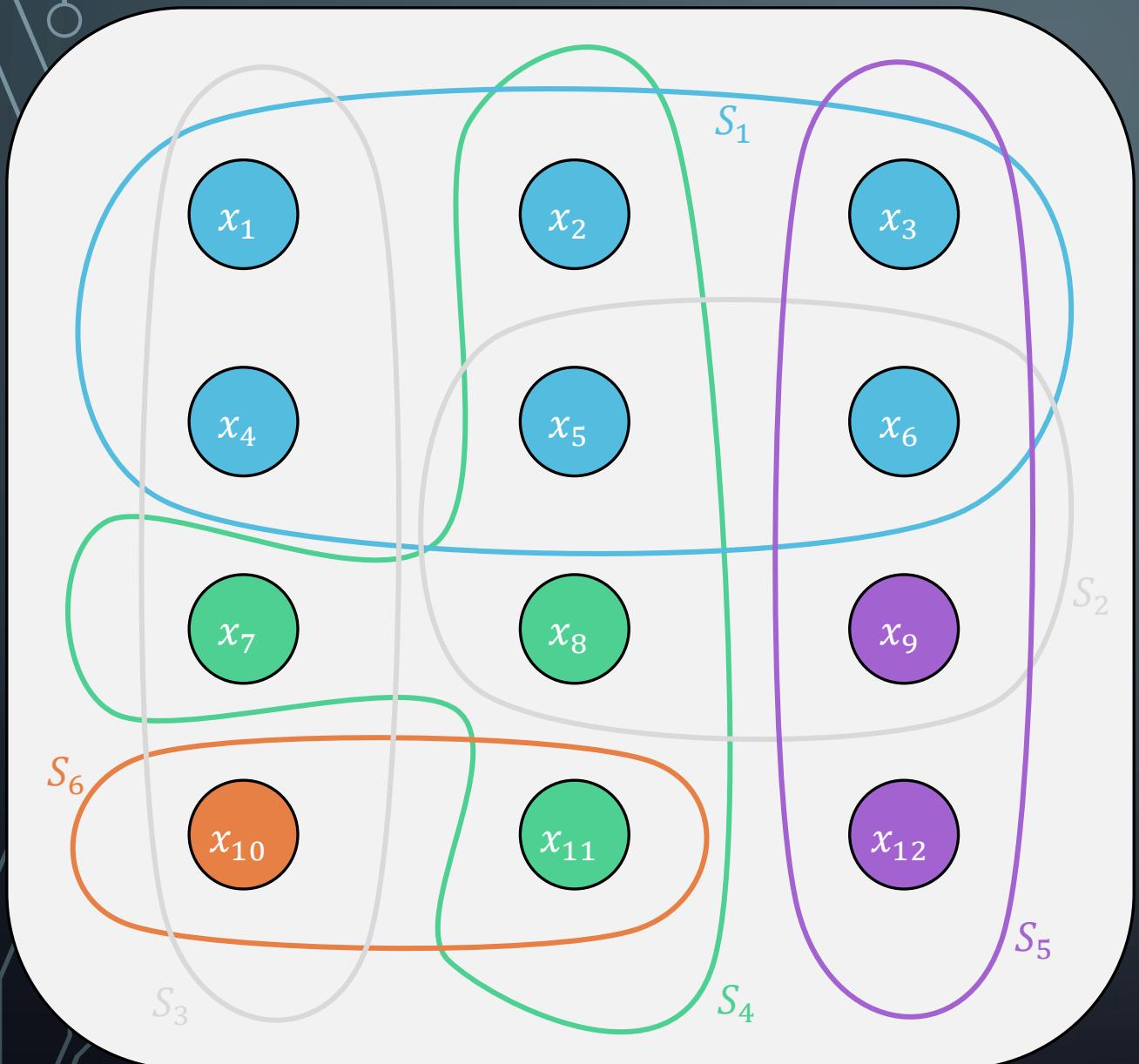
Note that it is established that:

$$H_d = \ln(d) + O(1)$$



Harmonic number (x-axis) and value of summation H_d (y-axis). **source = Wikipedia

ANALYSIS OF SET COVER GREEDY



Consider the following cost function

$$c_x = \frac{1}{|S_i - (S_1 \cup S_2 \cup S_3 \dots \cup S_{i-1})|}$$

Each set we add to the solution adds a cost of 1, but spread cost out among elements added.
Each element has exactly ONE cost.

Costs of the nodes in this example are:

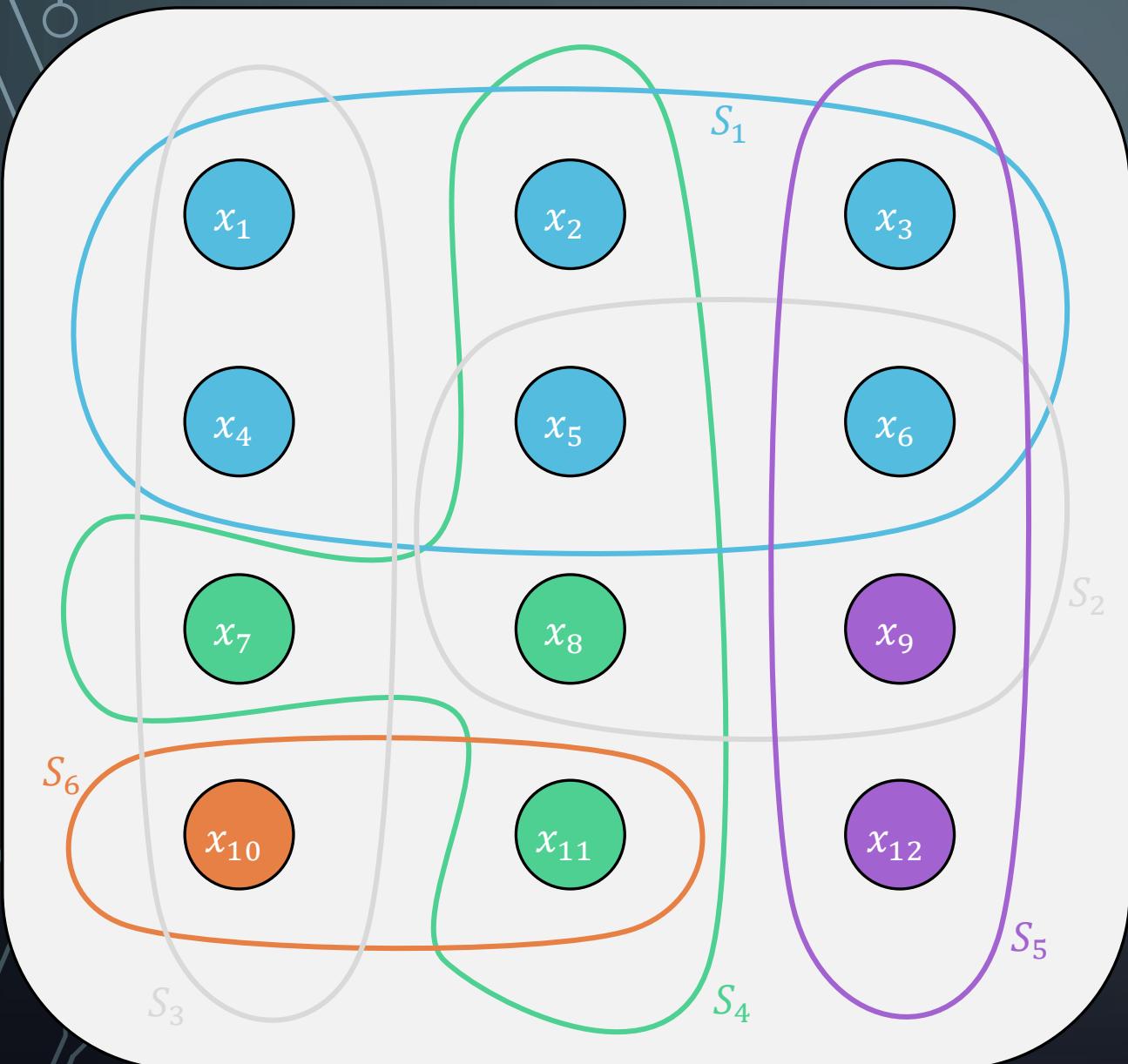
$$c_{x1} = c_{x2} = \dots = c_{x6} = \frac{1}{6}$$

$$c_{x7} = c_{x8} = c_{x11} = \frac{1}{3}$$

$$c_{x9} = c_{x12} = \frac{1}{2}$$

$$c_{x10} = 1$$

ANALYSIS OF SET COVER GREEDY



Costs of the nodes in this example are:

$$c_{x_1} = c_{x_2} = \dots = c_{x_6} = \frac{1}{6}$$

$$c_{x_7} = c_{x_8} = c_{x_{11}} = \frac{1}{3}$$

$$c_{x_9} = c_{x_{12}} = \frac{1}{2}$$

$$c_{x_{10}} = 1$$

Note the following

$$|C| = \sum_{x \in X} c_x$$

Sum of costs equals total cost. In this case 4

$$\sum_{S \in C^*} \sum_{x \in S} c_x \geq \sum_{x \in X} c_x = |C|$$

Because left side duplicates some costs that right side does not

ANALYSIS OF SET COVER GREEDY

From Previous Slide

$$|C| \leq \sum_{S \in C^*} \sum_{x \in S} c_x$$

$$\sum_{x \in S} c_x \leq H(|S|)$$

Accept as true for now, will argue why in a moment

$$|C| \leq \sum_{S \in C^*} H(|S|)$$

.....

$$|C| \leq |C^*| \times H(\max(|S| : S \in F))$$

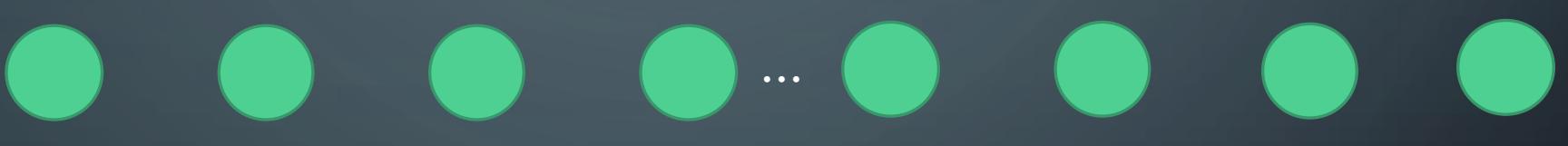
Q.E.D. if inequality assumed above is true

ANALYSIS OF SET COVER GREEDY

Analysis on previous slide relies on the following inequality. Does it actually hold?

First, suppose that there is a set S with size $|S|$ and algorithm adds all nodes from S "at once"

$$\sum_{x \in S} c_x \leq H(|S|)$$



$$\sum_{x \in S} c_x = \frac{1}{n} + \frac{1}{n} + \frac{1}{n} + \frac{1}{n} \dots \frac{1}{n} + \frac{1}{n} + \frac{1}{n} + \frac{1}{n}$$

$$< < < < < < < =$$

$$H(|S|) = 1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots \frac{1}{n-3} + \frac{1}{n-2} + \frac{1}{n-1} + \frac{1}{n}$$

ANALYSIS OF SET COVER GREEDY

Analysis on previous slide relies on the following inequality. Does it actually hold?

Now, suppose one node gets added in the previous "round" and S gets added next.

$$\sum_{x \in S} c_x \leq H(|S|)$$



$$\sum_{x \in S} c_x = \left[\frac{1}{n-1} + \frac{1}{n-1} + \frac{1}{n-1} + \frac{1}{n-1} \dots \frac{1}{n-1} + \frac{1}{n-1} + \frac{1}{n-1} + \frac{1}{n+\delta_1} \right]$$

< < < < < < = <

$$H(|S|) = \left[1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots \frac{1}{n-3} + \frac{1}{n-2} + \frac{1}{n-1} + \frac{1}{n} \right]$$

ANALYSIS OF SET COVER GREEDY

Analysis on previous slide relies on the following inequality. Does it actually hold?

Now, suppose S is added “third”,
OR after two nodes already added.

$$\sum_{x \in S} c_x \leq H(|S|)$$



$$\sum_{x \in S} c_x = \left[\frac{1}{n-2} + \frac{1}{n-2} + \frac{1}{n-2} + \frac{1}{n-2} \dots \frac{1}{n-2} + \frac{1}{n-2} + \frac{1}{n-1+\delta_2} + \frac{1}{n+\delta_1} \right]$$

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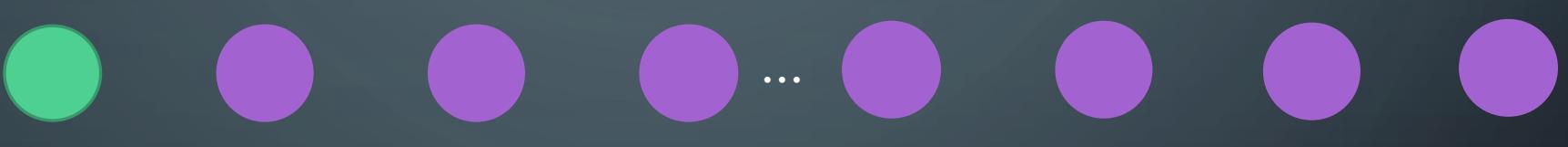
$$H(|S|) = \left[1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots \frac{1}{n-3} + \frac{1}{n-2} + \frac{1}{n-1} + \frac{1}{n} \right]$$

ANALYSIS OF SET COVER GREEDY

Analysis on previous slide relies on the following inequality. Does it actually hold?

Logic continues. What if last node
is added by itself...

$$\sum_{x \in S} c_x \leq H(|S|)$$



$$\sum_{x \in S} c_x = 1 + \frac{1}{2 + \delta_7} + \frac{1}{3 + \delta_6} + \frac{1}{4 + \delta_5} \dots \frac{1}{n - 3 + \delta_4} + \frac{1}{n - 2 + \delta_3} + \frac{1}{n - 1 + \delta_2} + \frac{1}{n + \delta_1}$$

$$= < < < < < < <$$

$$H(|S|) = 1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} \dots \frac{1}{n - 3} + \frac{1}{n - 2} + \frac{1}{n - 1} + \frac{1}{n}$$

ANALYSIS

Theorem:

Greedy-Set-Cover(X, F) is a $\rho(n)$ -approximation of set-cover where
$$\rho(n) = H(\max(|S| : S \in F))$$

Thus, it is proven!!

CONCLUSION

CONCLUSIONS

Basic Approximations:

- Are very fast (often greedy) algorithms. Algorithm / implementation not difficult.
- Analysis to figure out “how good” approximation is often much harder
- We only scratched the surface in this slide deck.