

# Introduction to Approximation Algorithms, part I

08-1 2025, Srikanth Srinivasan,  
DIKU (Slides: Mikkel Abrahamsen)

## APPROX-VERTEX-COVER( $G$ )

$$C := \emptyset$$

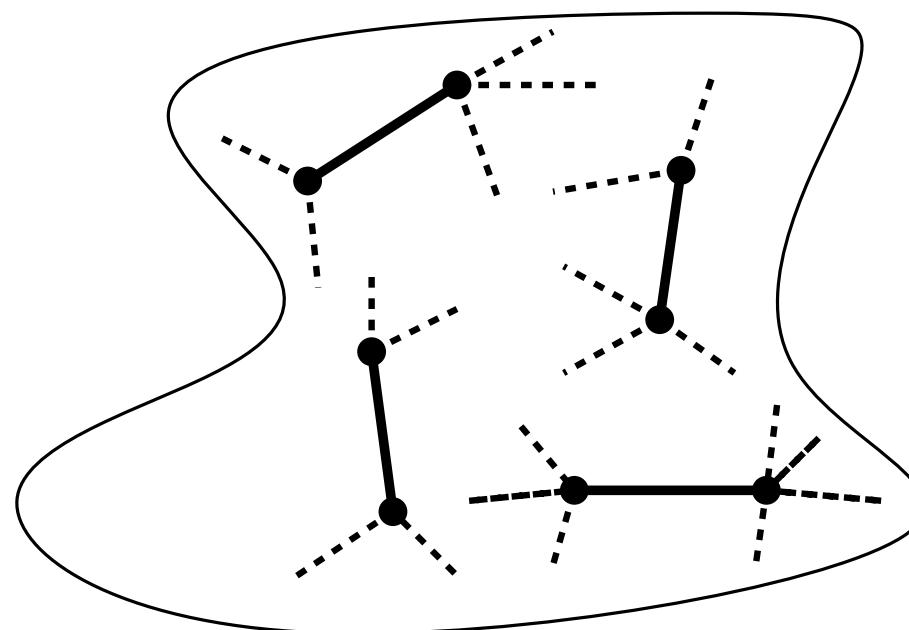
**while**  $E(G) \neq \emptyset$

choose  $uv \in E(G)$

$$C := C \cup \{u, v\}$$

remove all edges incident on  $u$  or  $v$  from  $E(G)$

return  $C$



# The big picture

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**Today:** Approximation algorithms (good when suboptimal solutions are acceptable).

## Definition

**Def.:** An algorithm for an optimization problem has *approximation ratio*  $\rho(n)$  if for every input of size  $n$ ,

$$\max \left\{ \frac{C}{C^*}, \frac{C^*}{C} \right\} \leq \rho(n).$$

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minimization problem

$C := \text{cost(produced sol.)}$

maximization problem



## Vertex Cover

**Def.:** Let  $G = (V, E)$  be a graph. A set  $V' \subseteq V$  of vertices is a *vertex cover* if for all  $uv \in E$ , we have  $u \in V'$  or  $v \in V'$ .

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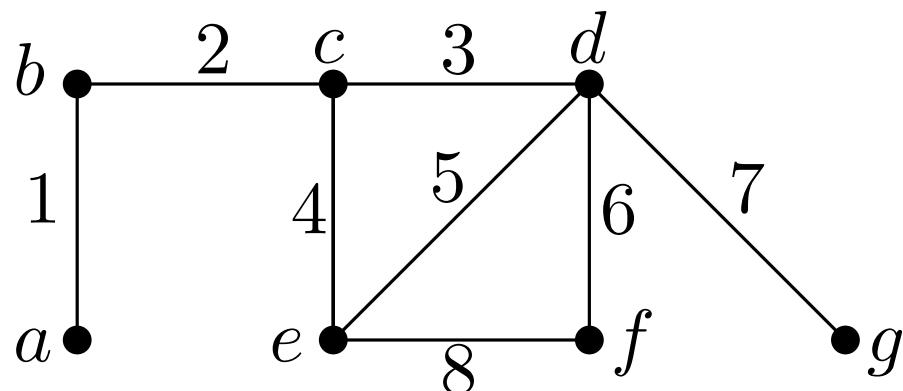
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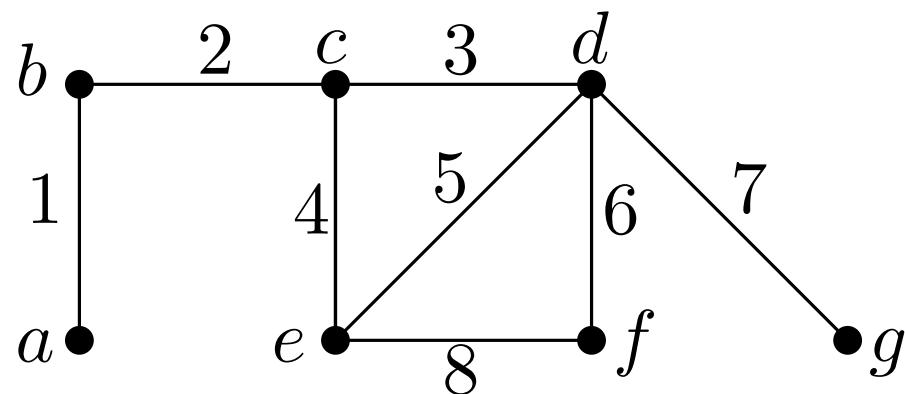
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**Exercise:**



# Implementation



Adjacency lists:

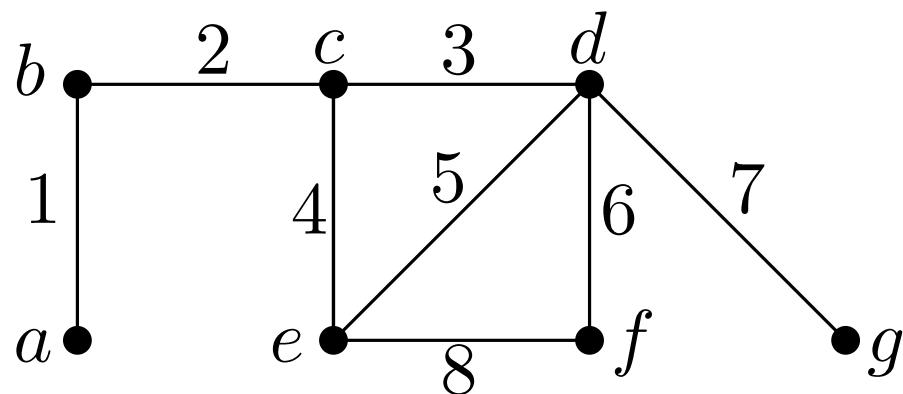
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⋮

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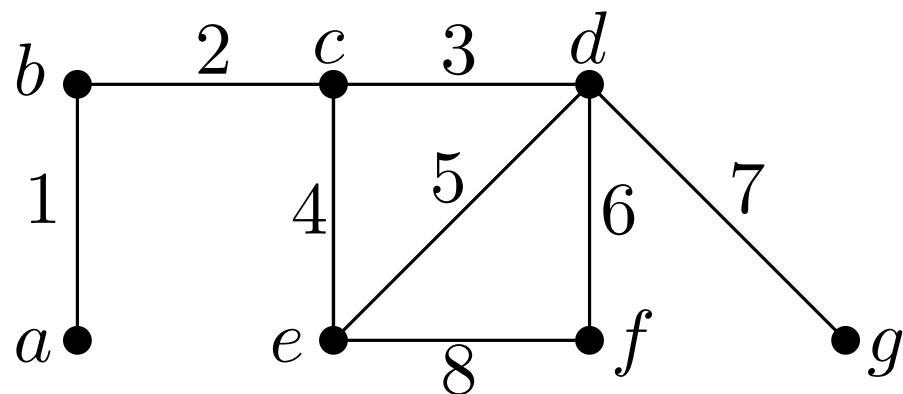
Array of edges

```
[(a, b, 1), (b, c, 1), (c, d, 1), (c, e, 1), (d, e, 1), ...]
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$\vdots$	$\vdots$

# Implementation



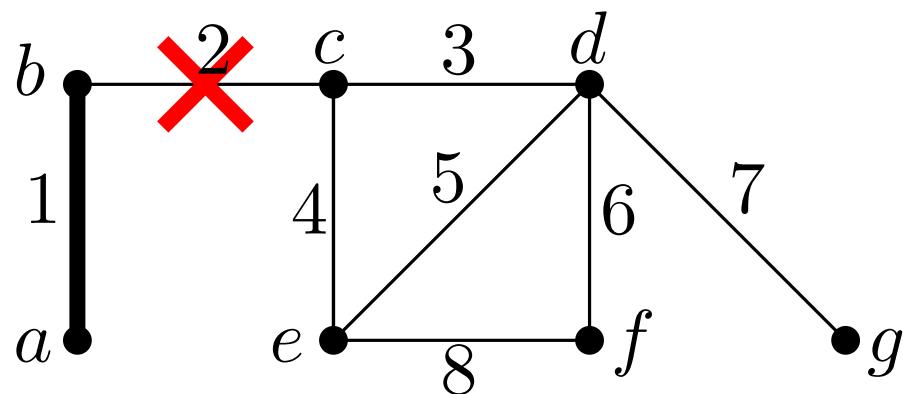
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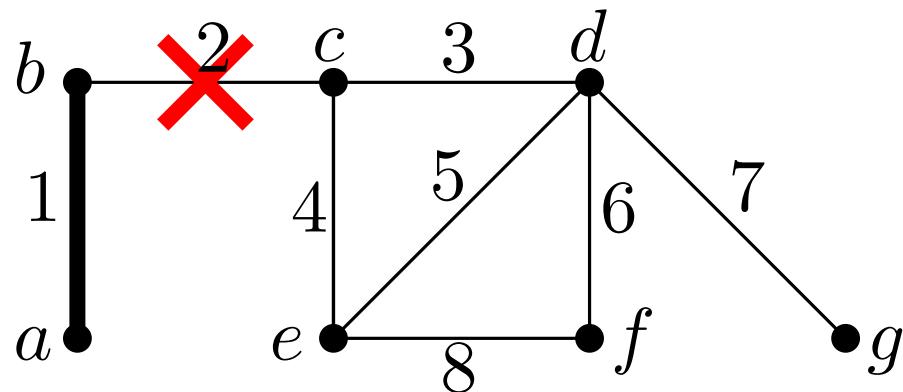
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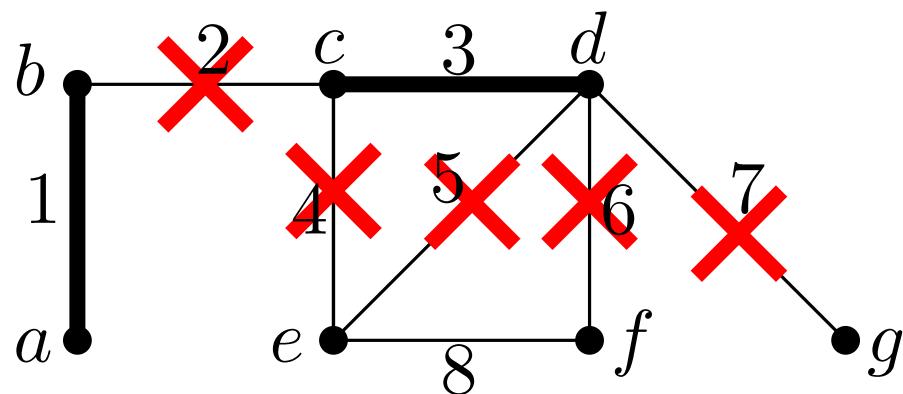
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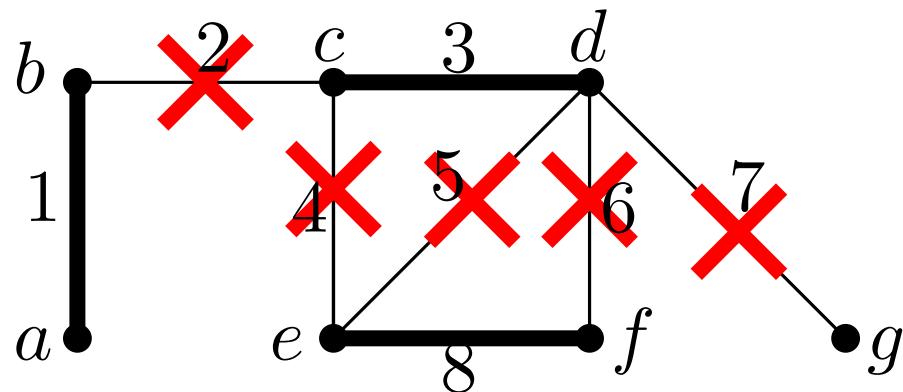
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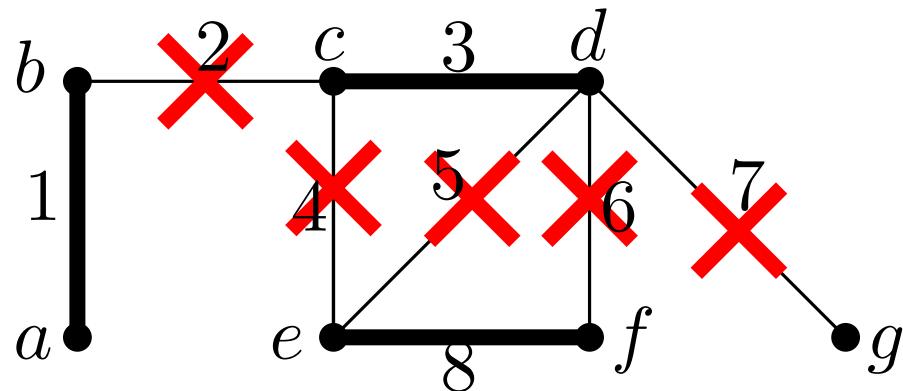
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Running time:  $O(|V| + |E|)$

## Theorem

**Thm.:** APPROX-VERTEX-COVER is a 2-approximation algorithm.

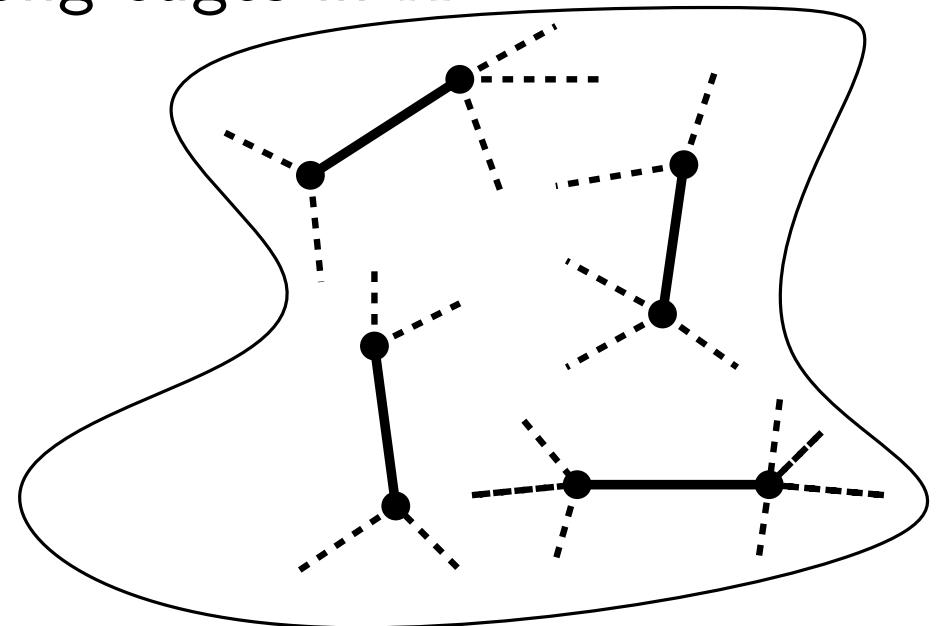
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Let  $A \subset E$  be the edges chosen by the algorithm.

Obs: No shared endpoints among edges in  $A$



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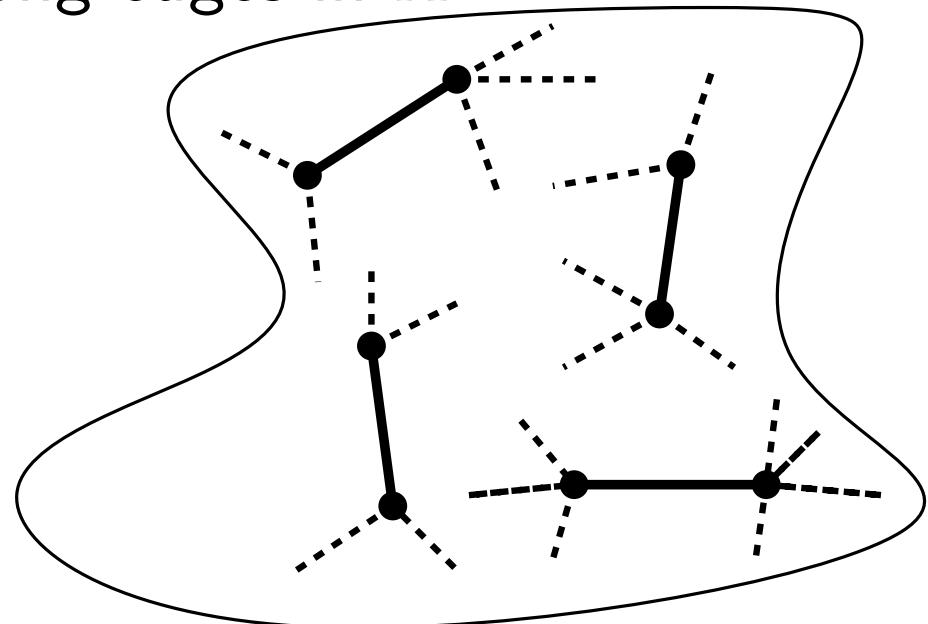
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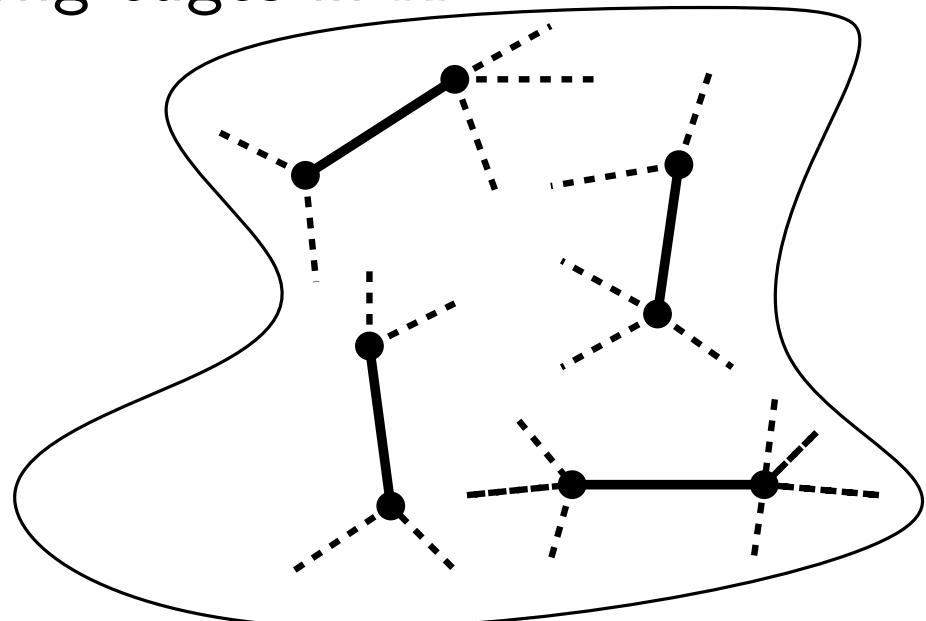
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Hence,

$$|C^*| \geq |A| = |C|/2 \implies \frac{|C|}{|C^*|} \leq 2.$$



## Reflection and methodology

How can we prove  $C/C^* \leq 2$  when we don't know  $C^*$ ?

Answer: By proving  $C \leq 2|A|$  and  $|A| \leq C^*$ .

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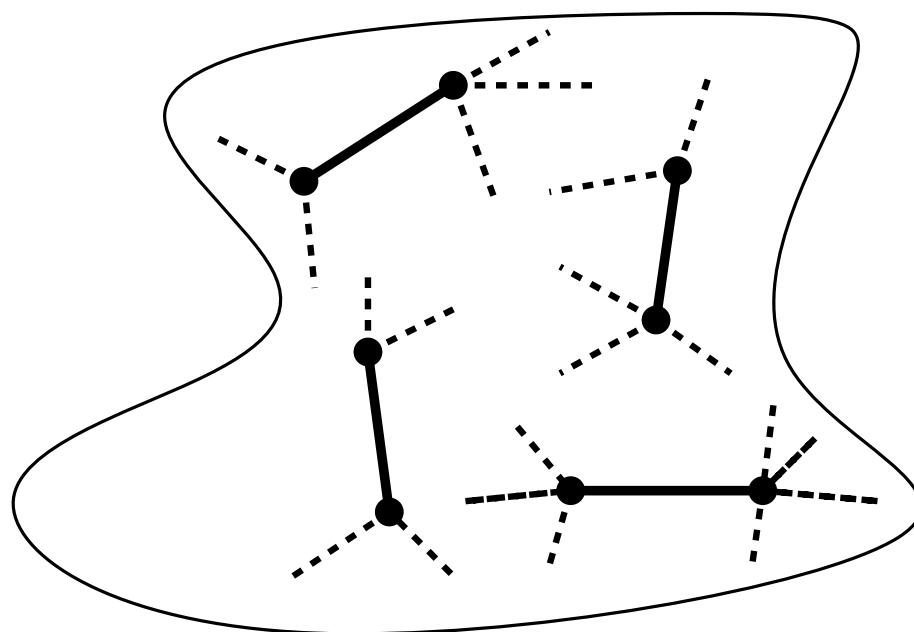
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General technique: Find a parameter  $\square$  such that  $C \leq \rho \cdot \square$  and  $\square \leq C^*$ .

For vertex cover:  $\square = |A|$  and  $\rho = 2$ .

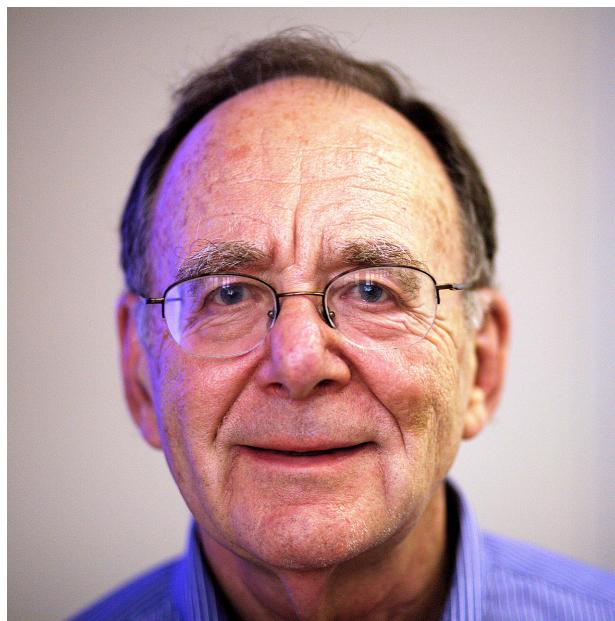
# Question

Try to guess: Is there an approximation algorithm with a better approximation ratio?



# History

1972: Karp's 21 NP-complete problems  
(including vertex cover, set cover, Hamiltonian cycle and subset sum)



Karp

Turing Award

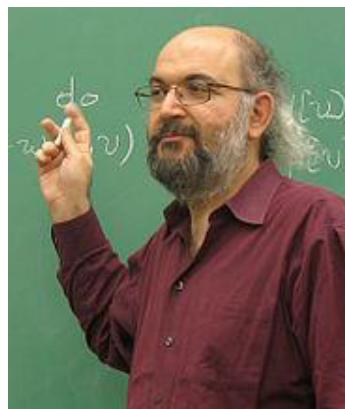


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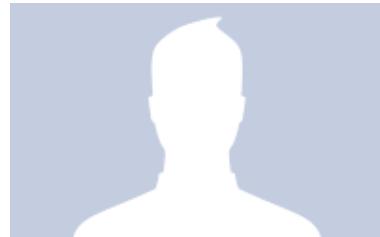
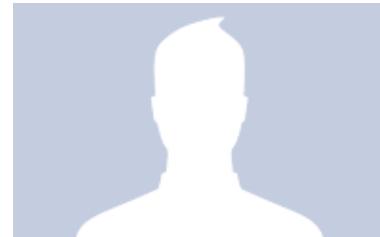
19xx: Many  $\leq 2 - o(1)$ .



Gavril



Yannakakis



...

# History

Assuming  $P \neq NP$ :

1999: Håstad,  $\geq 7/6$

2005: Dinur & Safra,  $\geq 1.38$

2018: Khot, Minzer, Safra,  $\geq 1.41$



Håstad



Dinur



Safra



Khot



Minzer

# History

2008: Khot & Regev,  $\geq 2 - \varepsilon$  assuming the  
Unique Game Conjecture.

Some, but not all people believe it.



Khot



Nevanlinna prize 2016



Regev

# Traveling Salesperson

Given a complete undirected graph  $G = (V, E)$ .

For all  $u, v \in V$ , we are given  $c(uv) \in \{0, 1, \dots\}$ .

**Goal:** Find minimum weight cycle through all vertices.

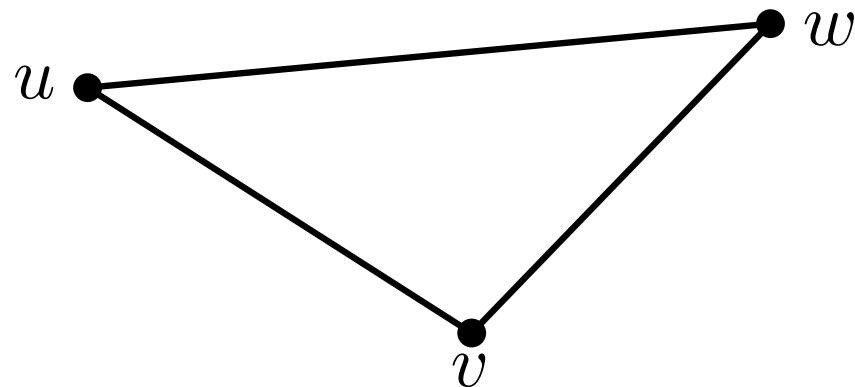
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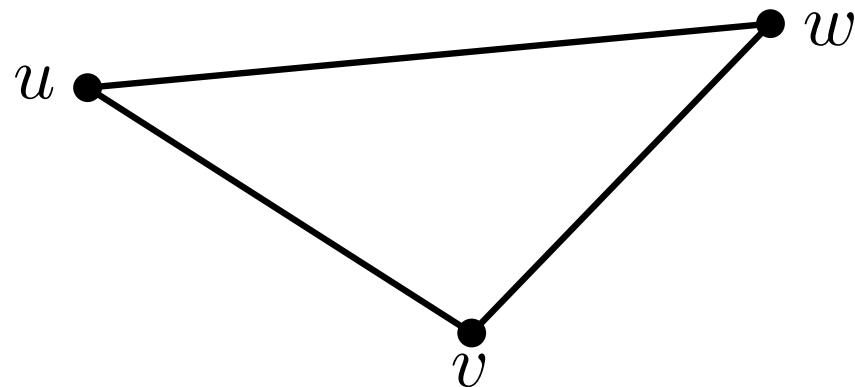
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Still NP-hard!



# Algorithm

APPROX-TSP( $G, c$ )

Find MST  $T$

Make Euler tour  $W$  using each edge of  $T$  twice

Shortcut  $W$  to  $H$  by skipping duplicates

Return  $H$

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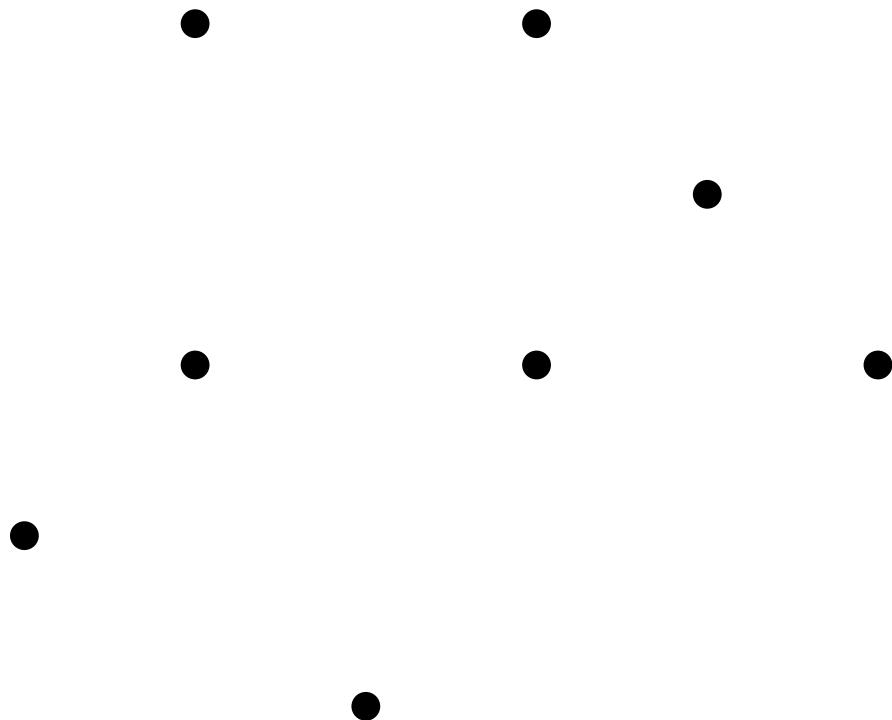
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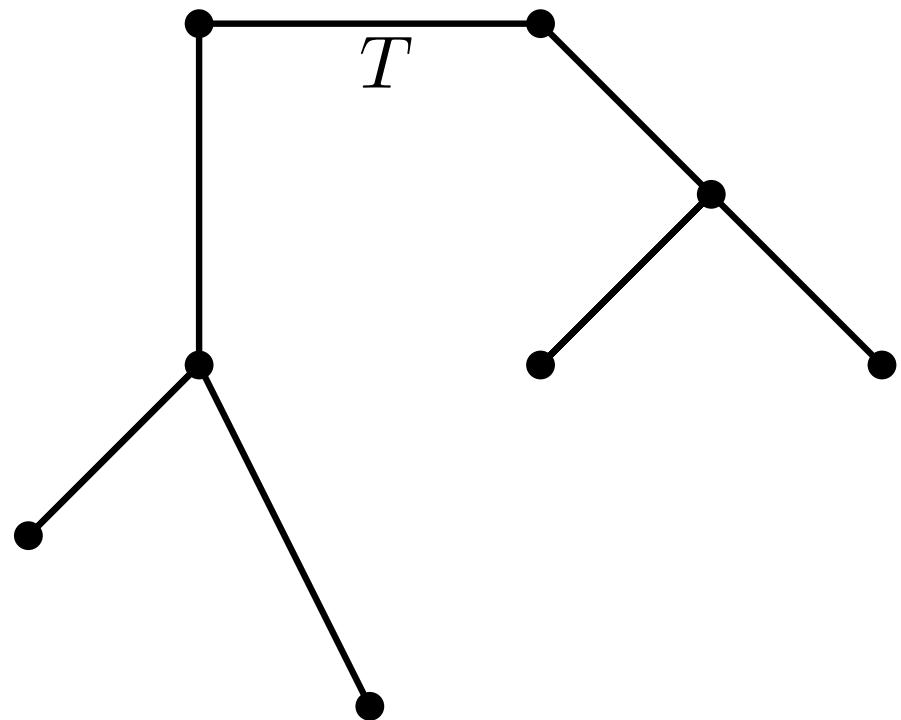
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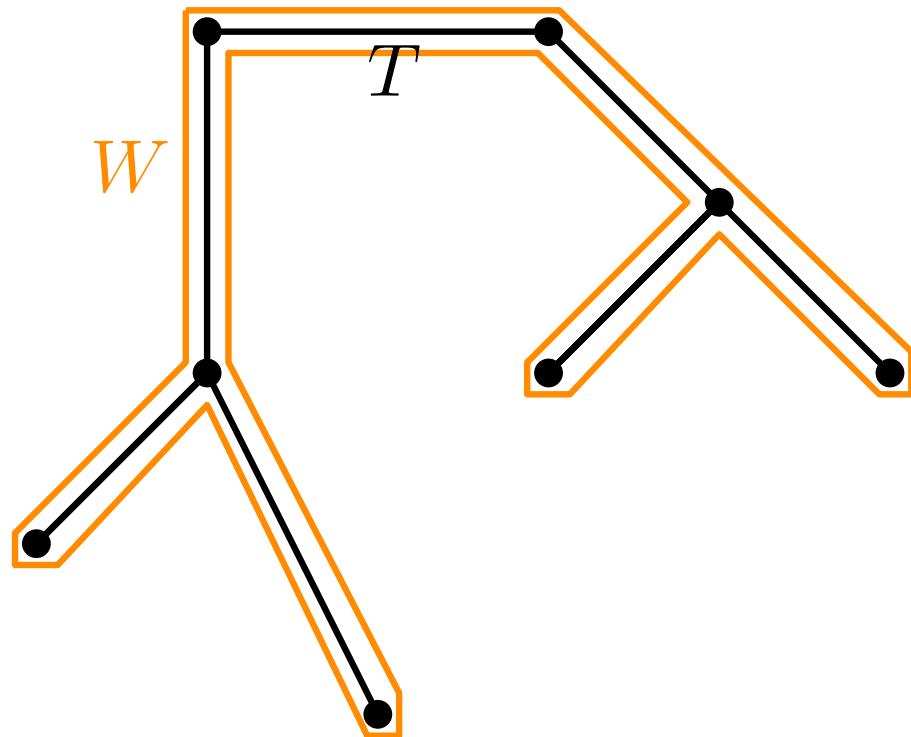
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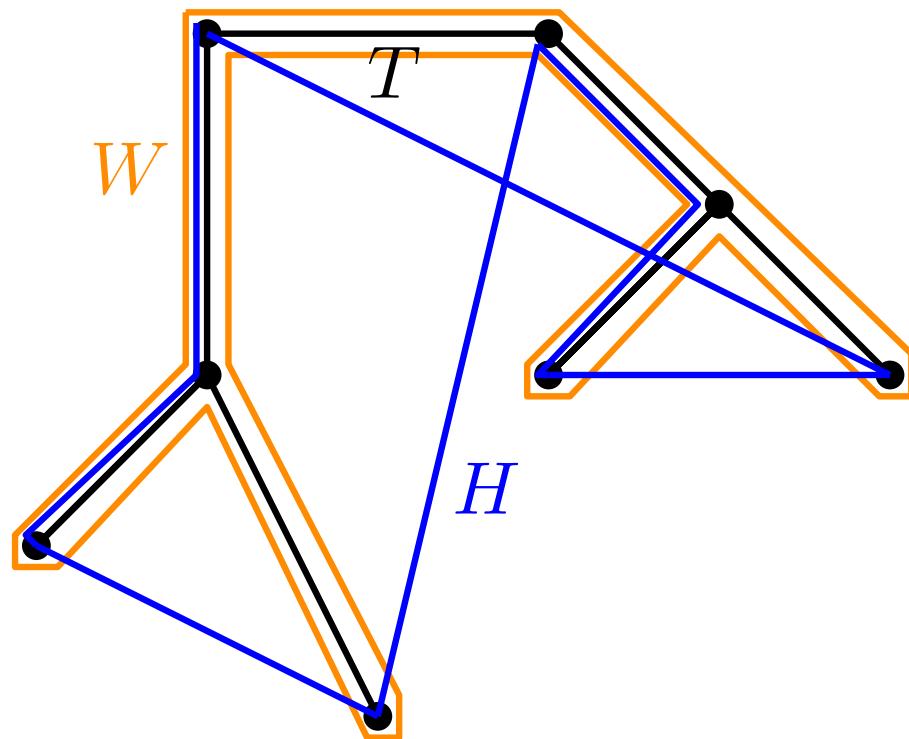
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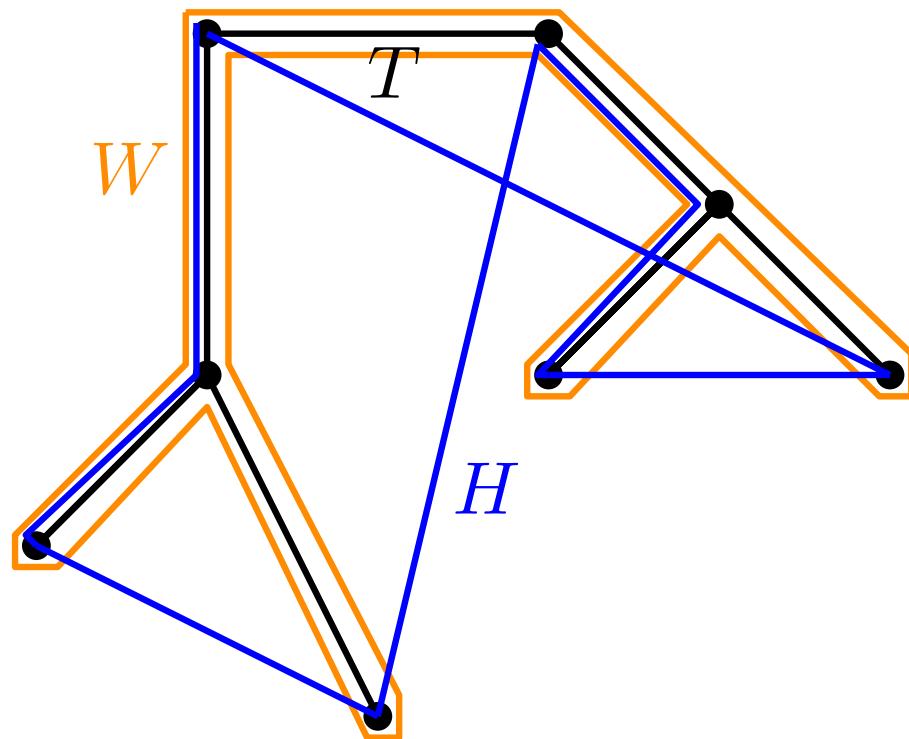
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Exercise: Run the algorithm on this instance.

$a$                      $b$   
                           
 $c$                      $d$

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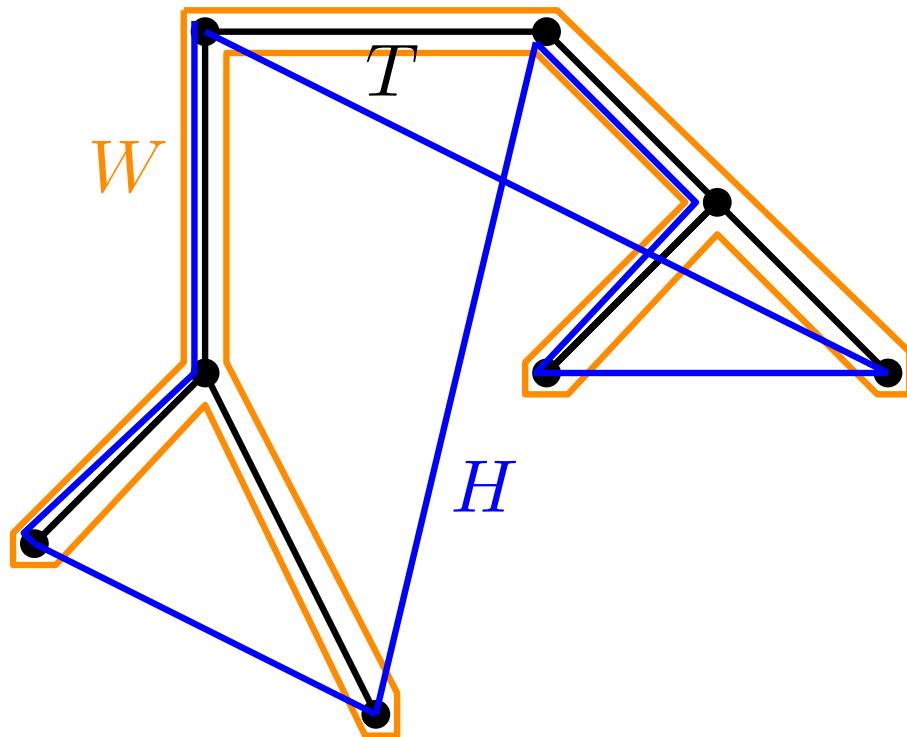
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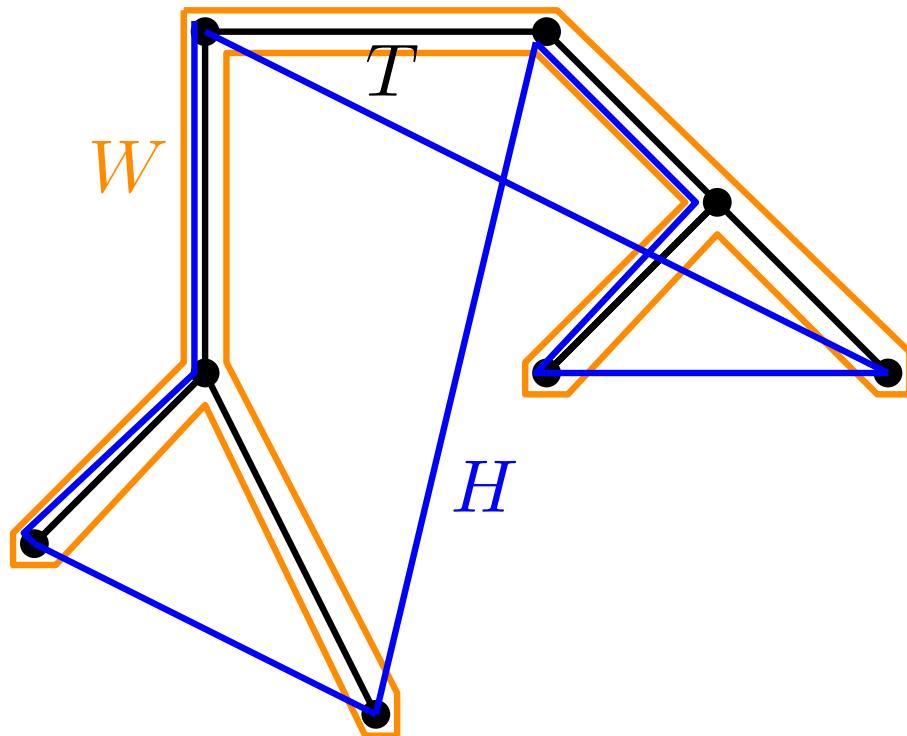
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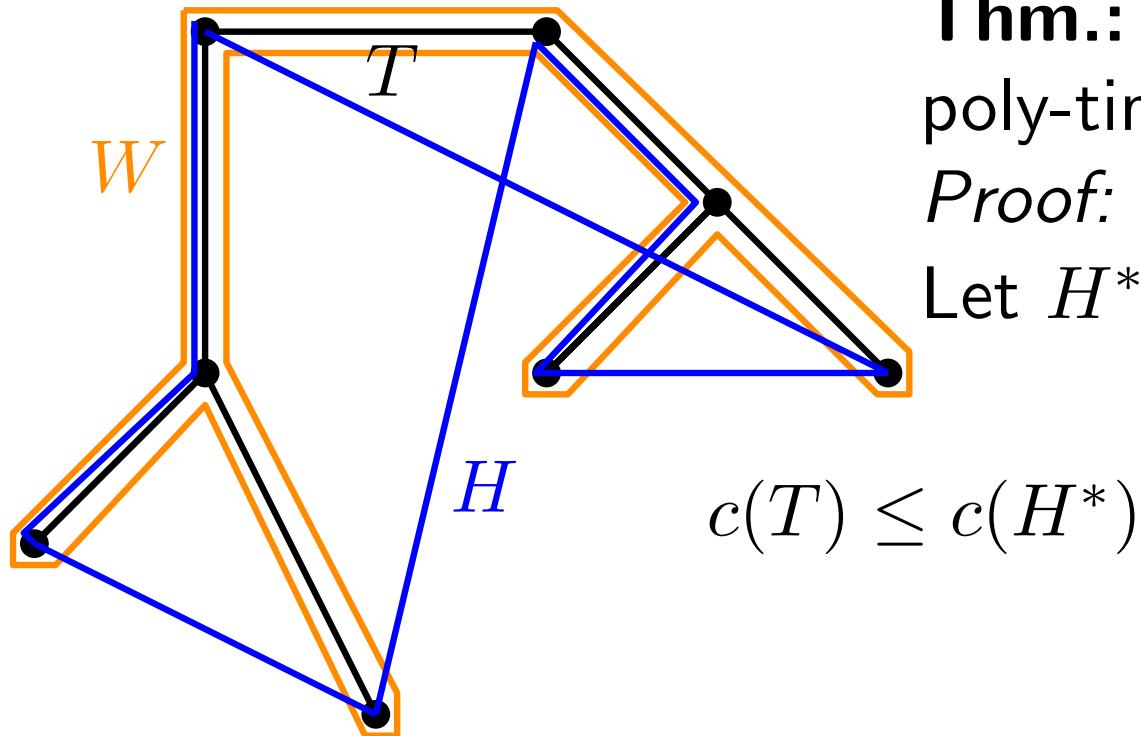
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$$c(T) \leq c(H^*)$$

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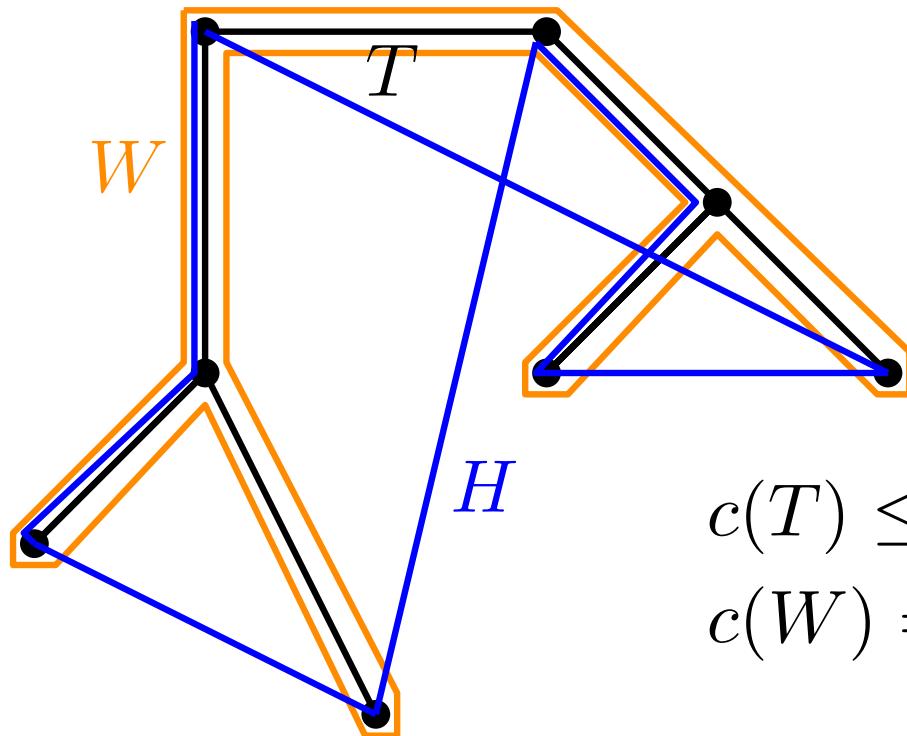
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# Theorem

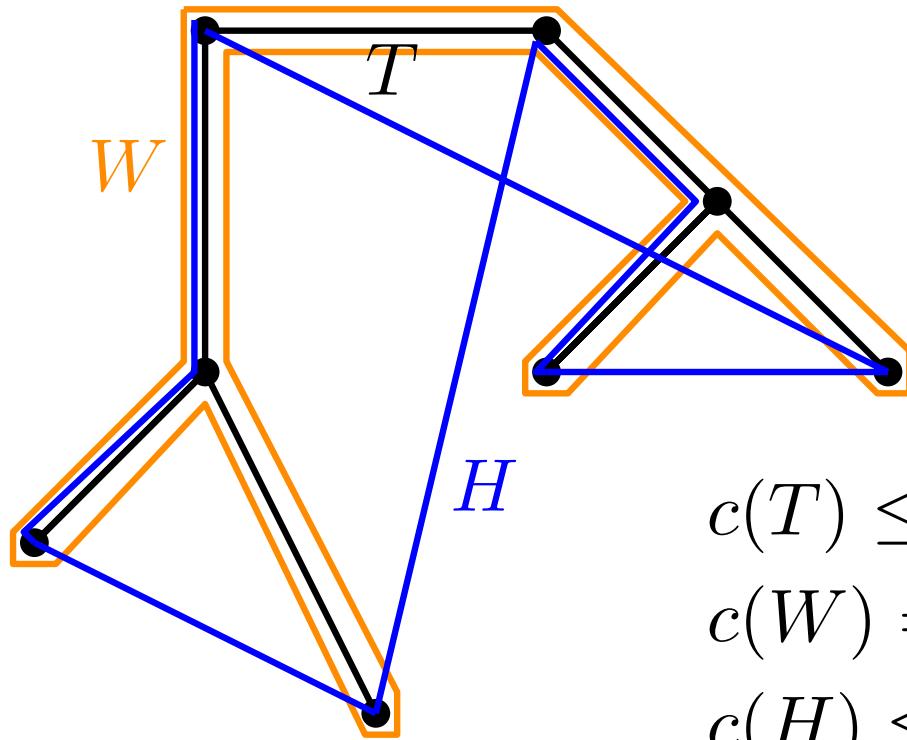
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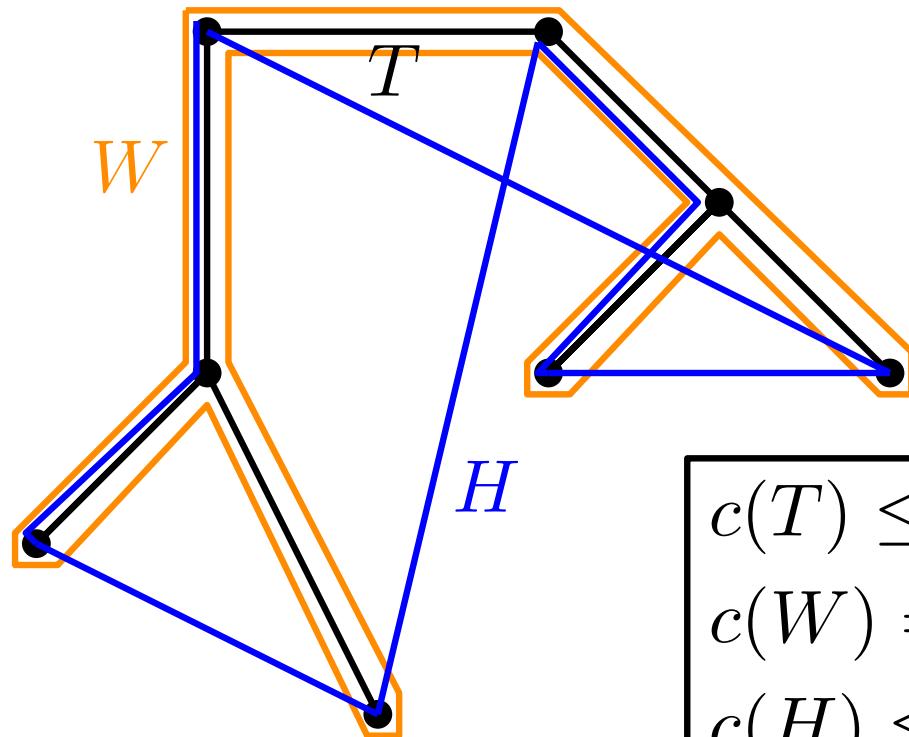
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$$\begin{aligned}c(T) &\leq c(H^*) \\c(W) &= 2c(T) \\c(H) &\leq c(W)\end{aligned}$$

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

## Reflection and methodology

How can we prove  $c(H)/c(H^*) \leq 2$  when we don't know  $H^*$ ?

Answer: By proving  $c(H) \leq 2c(T)$  and  $c(T) \leq c(H^*)$ .

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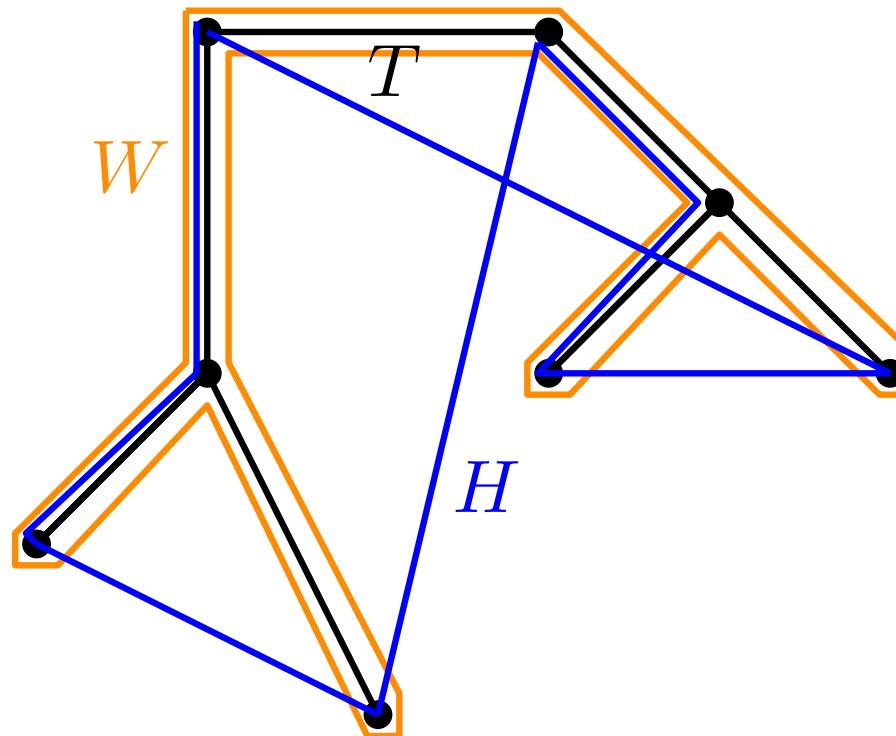
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General technique: Find a parameter  $\square$  such that  $C \leq \rho \cdot \square$  and  $\square \leq C^*$ .

For TSP:  $\square = c(T)$  and  $\rho = 2$ .

# Question

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# History

1976: Christofides, Serdyukov, 1.5-apx algorithm

It's simple! See, e.g., Wikipedia. No improvement for decades

2021: Karlin, Klein, Gharan,  $(1.5 - \varepsilon)$ -apx algorithm for some  $\varepsilon > 10^{-36}$



## Computer Scientists Break Traveling Salesperson Record

24 | 0

After 44 years, there's finally a better way to find approximate solutions to the notoriously difficult traveling salesperson problem.



# Set Cover

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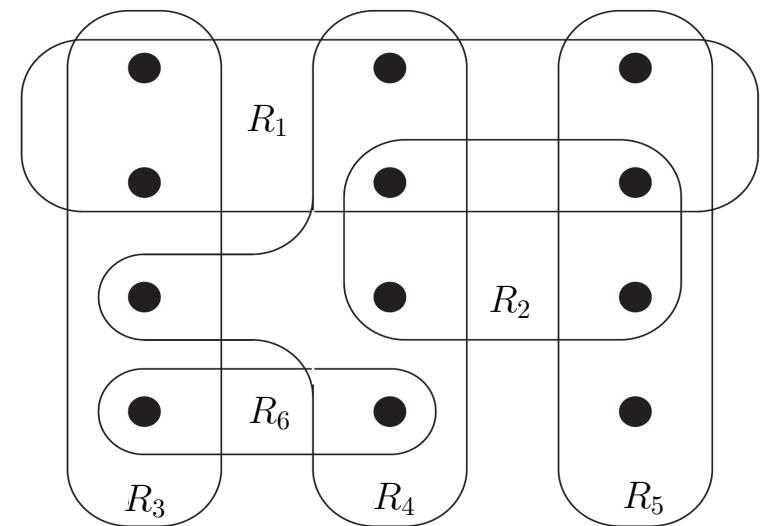
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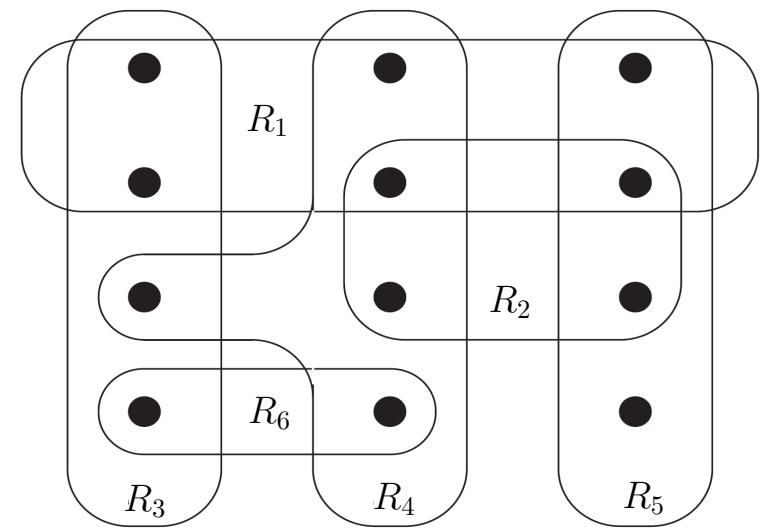
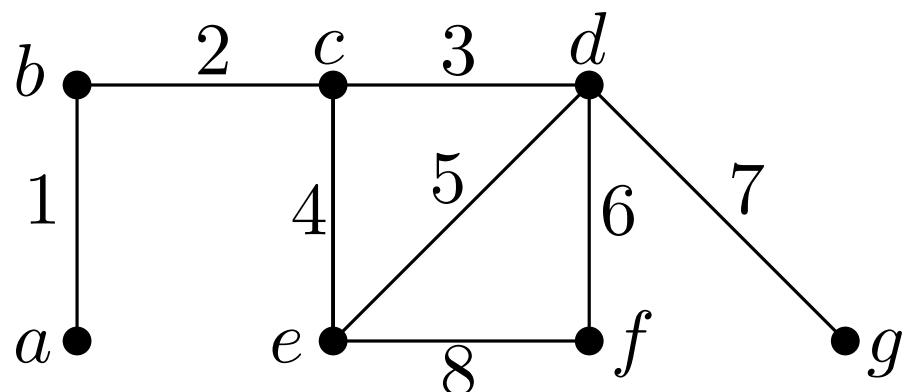


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**Exercise:** Show that vertex cover is a special case.



# Set Cover

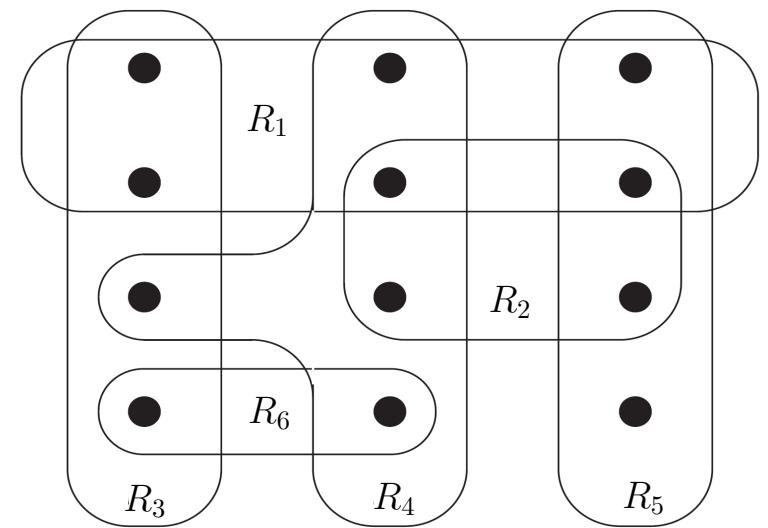
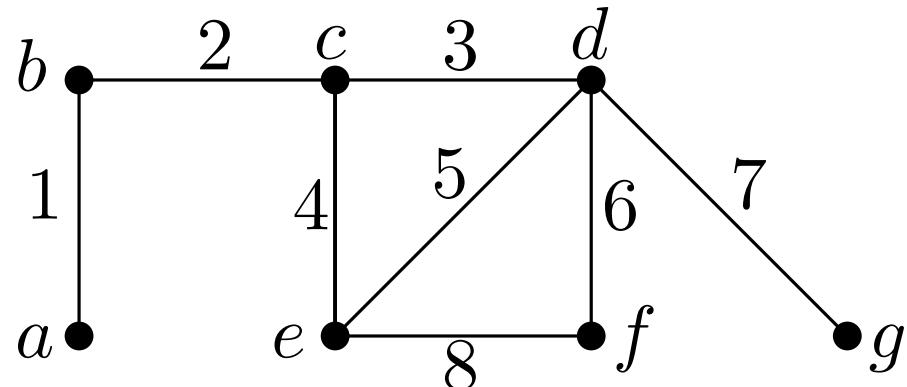
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$$X := \{1, 2, \dots, 8\}$$

$$\mathcal{F} := \{\{1\}, \{1, 2\}, \{2, 3, 4\}, \{3, 5, 6\}, \{4, 5, 8\}, \{6, 8\}, \{7\}\}$$



# Set Cover

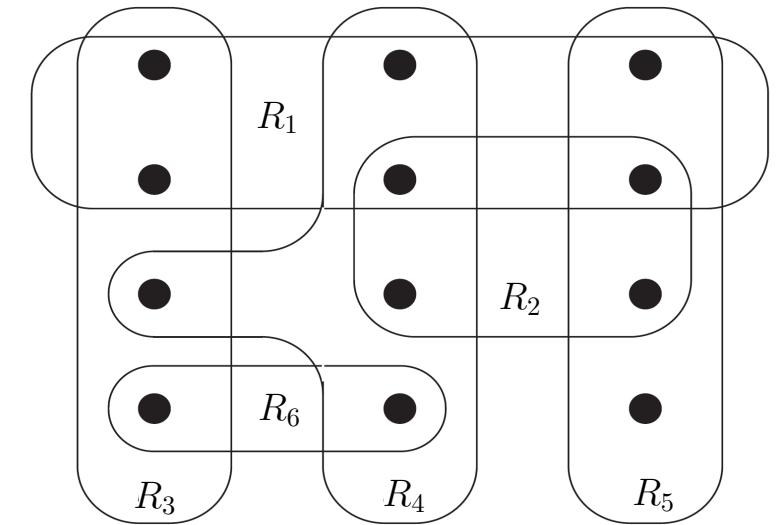
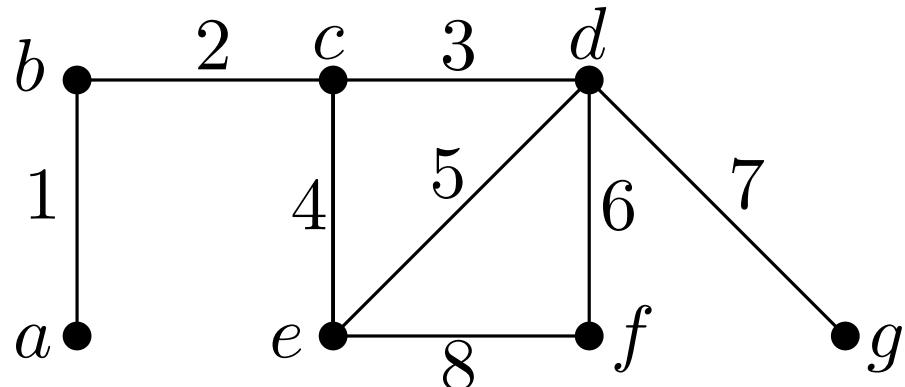
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$$X := E$$

$$\mathcal{F} := \{E(v) \mid v \in V\}$$

$$E(v) := \{uv \in E \mid u \in V\}$$

# Greedy Algorithm

GREEDY-SET-COVER( $X, \mathcal{F}$ )

$i := 0$

while  $X \setminus S_{<i+1} \neq \emptyset$

$i := i + 1$

Pick  $S_i \in \mathcal{F}$  with  $\max |S_i \setminus S_{<i}|$

Return  $\mathcal{C} := \{S_1, \dots, S_i\}$

Here,  $S_{<i} := \bigcup_{j=1}^{i-1} S_j$ .

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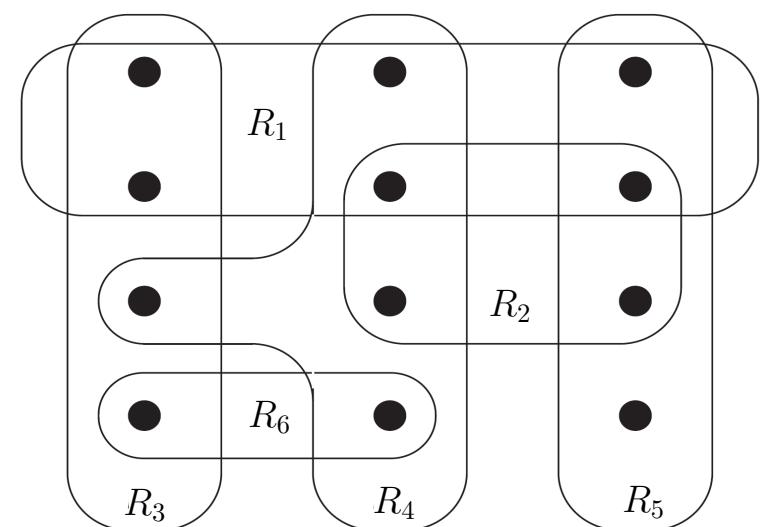
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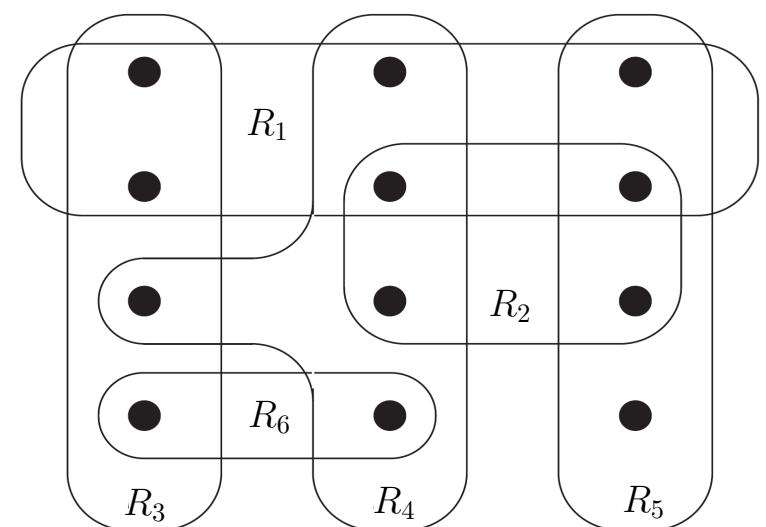
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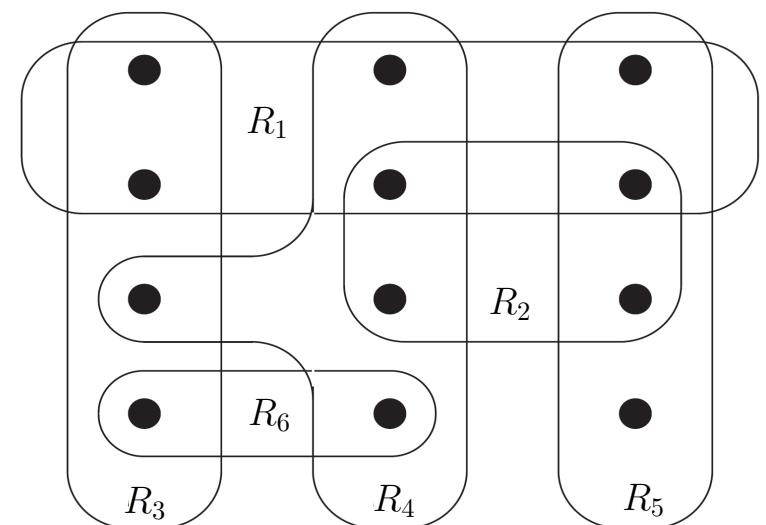
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Here,  $S_{<i} := \bigcup_{j=1}^{i-1} S_j$ .

**Exercise:** Run the algorithm on this instance.

$S_1 := R_1$

$S_2 := R_4$



# Greedy Algorithm

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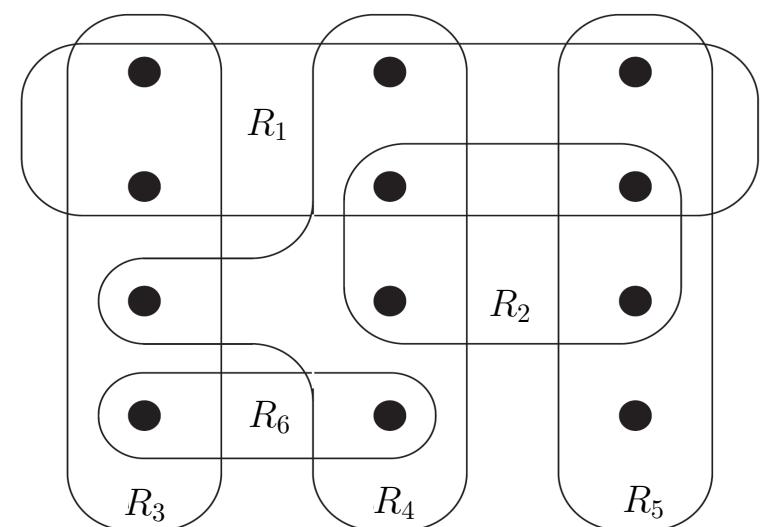
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$S_1 := R_1$

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$S_3 := R_5$



# Greedy Algorithm

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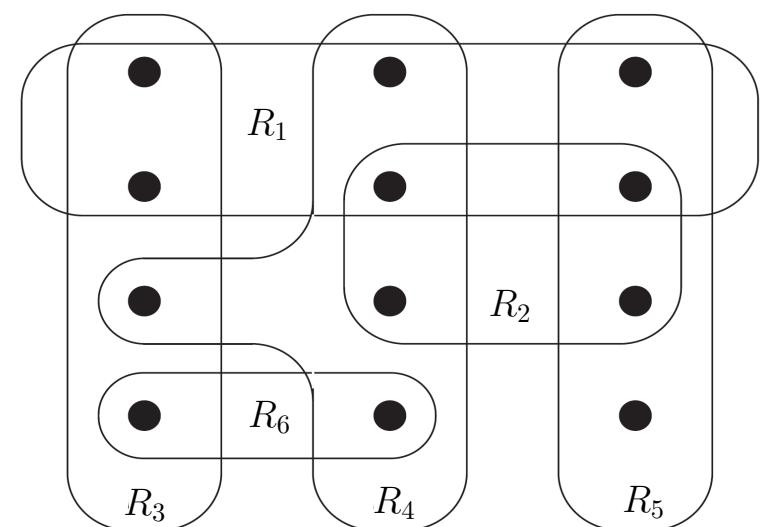
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$S_1 := R_1$

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$S_3 := R_5$

$S_4 := R_3$  or  $S_4 := R_6$



# Theorem

**Thm.:** For opt. sol.  $\mathcal{C}^*$ , we have

$$|\mathcal{C}| \leq H_{|X|} \cdot |\mathcal{C}^*|,$$

where

$$H_n := \sum_{i=1}^n 1/i \leq \ln n + 1.$$

Hence, GREEDY-SET-COVER is a  $O(\log n)$ -approx. alg.

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For  $x \in S_i \setminus S_{<i}$ , define  $c_x := \frac{1}{|S_i \setminus S_{<i}|}$ .

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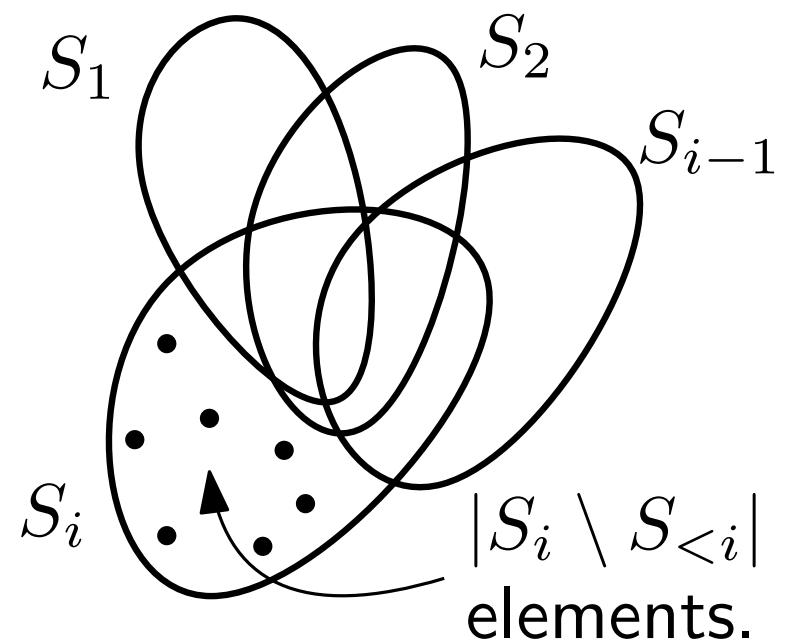
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$$c(X) = \sum_{i=1}^{|\mathcal{C}|} \sum_{x \in S_i \setminus S_{<i}} c_x = \sum_{i=1}^{|\mathcal{C}|} 1 = |\mathcal{C}|.$$

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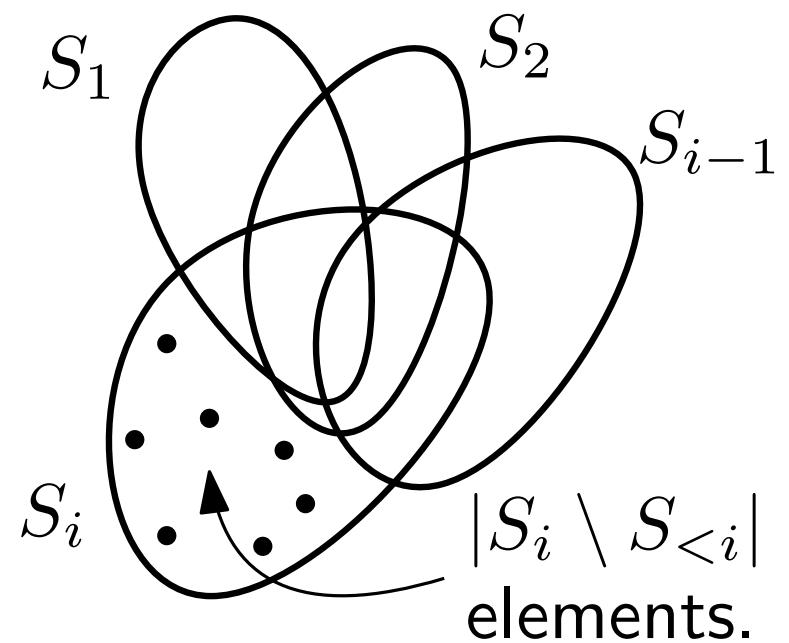
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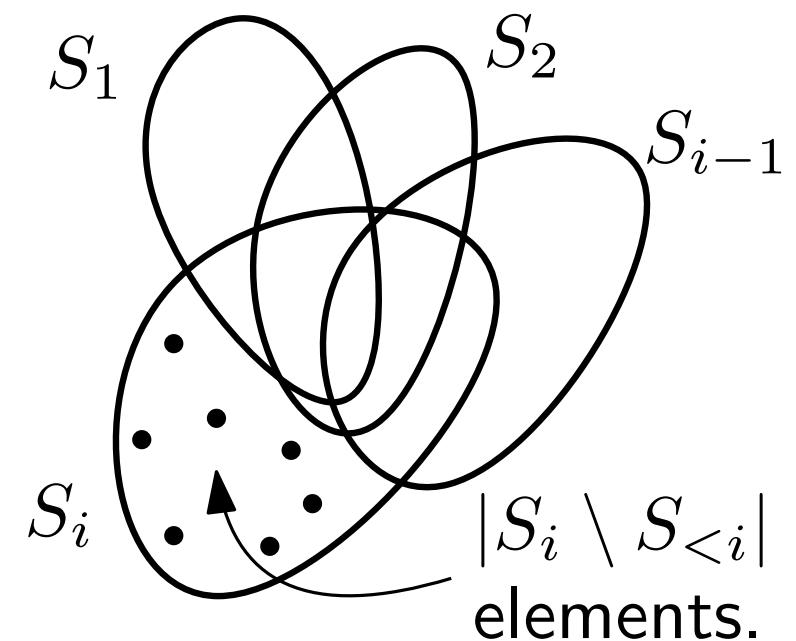
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*Proof of Thm.:*

$$|\mathcal{C}| = c(X) \leq \sum_{S \in \mathcal{C}^*} c(S) \leq \sum_{S \in \mathcal{C}^*} H_{|S|} \leq \sum_{S \in \mathcal{C}^*} H_{|X|} = |\mathcal{C}^*| \cdot H_{|X|}.$$

GREEDY-SET-COVER( $X, \mathcal{F}$ )

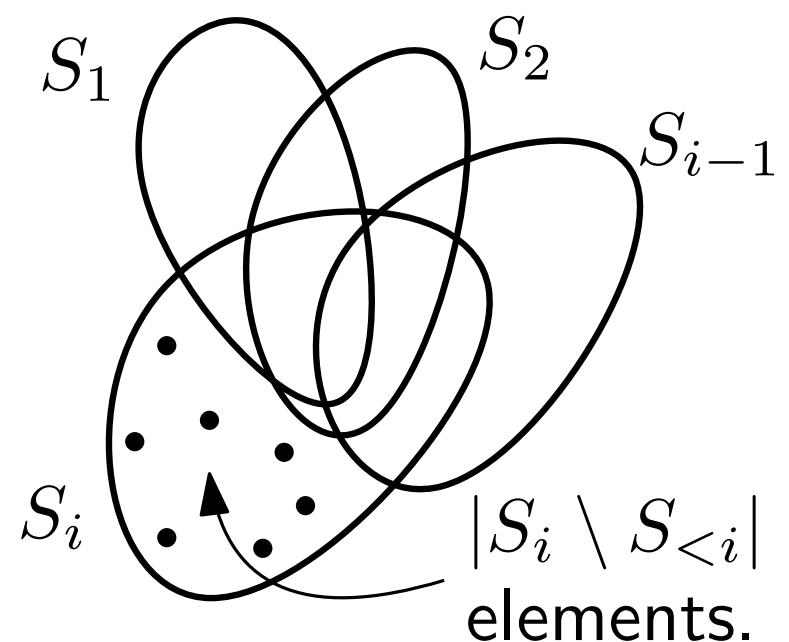
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# Lemma: Idea and Example

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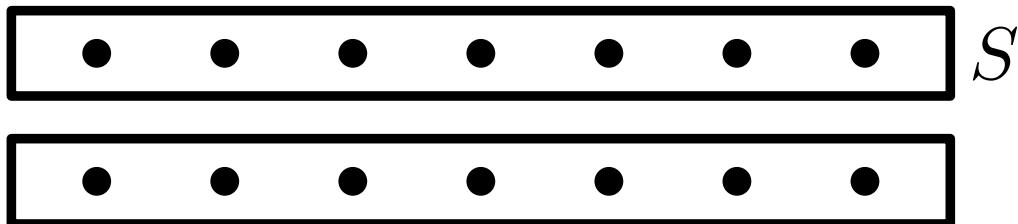
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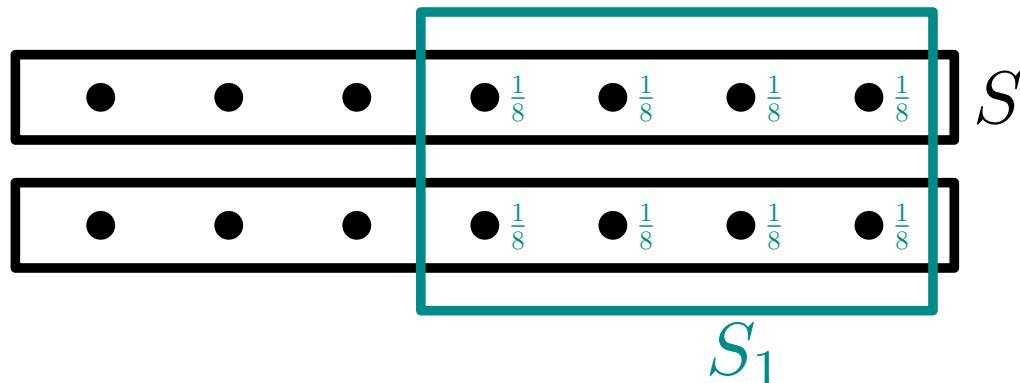
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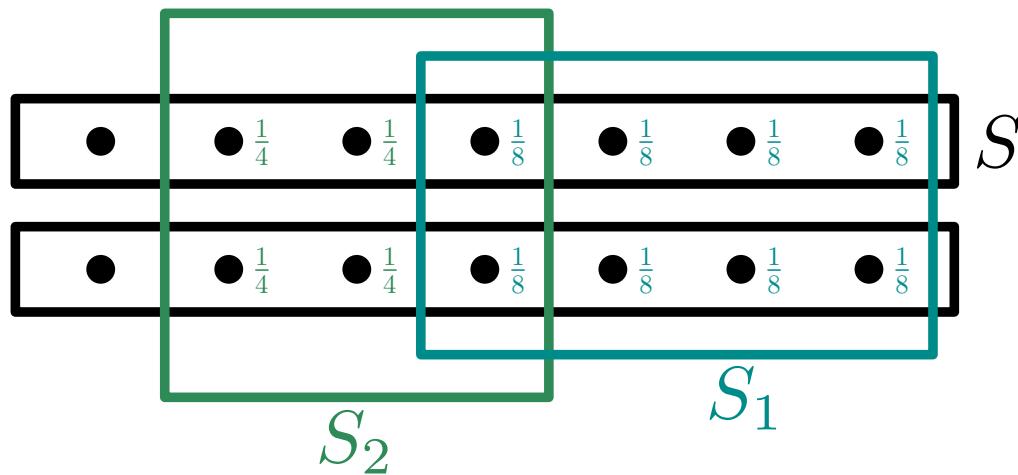
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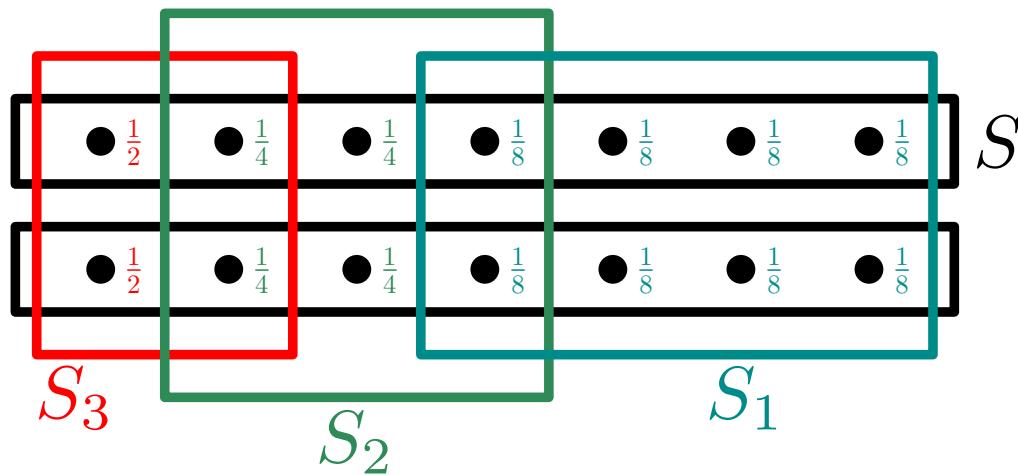
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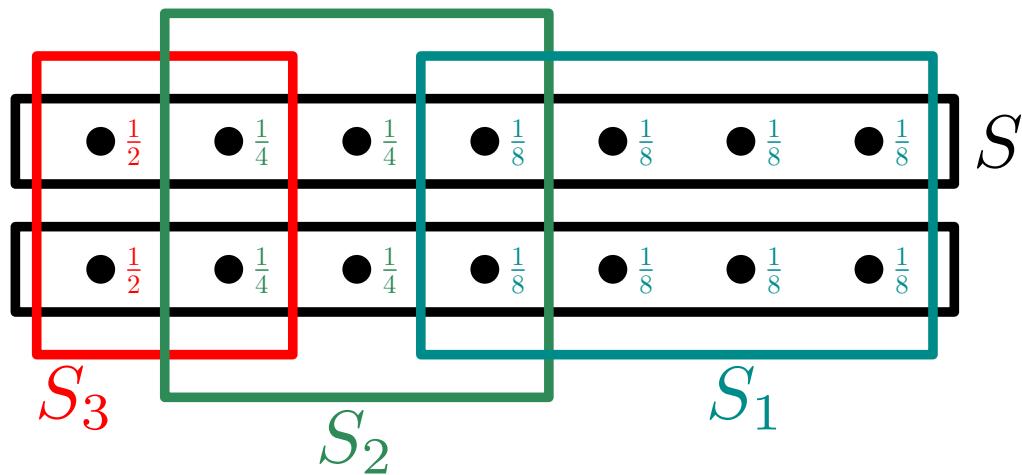
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$$\begin{aligned} c(S) &= \frac{1}{2} + \frac{1}{4} + \frac{1}{4} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} \\ &\leq 1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \frac{1}{5} + \frac{1}{6} + \frac{1}{7} \\ &= H_{|S|}. \end{aligned}$$

# Proof of Lemma

**Lemma:** For all  $S \in \mathcal{F}$ :

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(since  $S \setminus S_{<i}$  contains  $x_j, x_{j-1}, \dots, x_1$ )

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$$|S_i \setminus S_{<i}| \geq |S \setminus S_{<i}| \geq j \implies c_{x_j} = \frac{1}{|S_i \setminus S_{<i}|} \leq \frac{1}{j}.$$

by greedy choice of  $S_i$

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Pick  $S_i \in \mathcal{F}$  with  $\max |S_i \setminus S_{<i}|$

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For  $x \in S_i \setminus S_{<i}$ , define  $c_x := \frac{1}{|S_i \setminus S_{<i}|}$ .

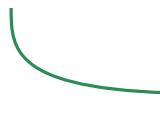
For  $Y \subset X$ , define  $c(Y) := \sum_{x \in Y} c_x$ .

**Proof:** Let  $S = \{x_k, x_{k-1}, \dots, x_1\}$ , where  $x_k$  covered first, then  $x_{k-1}$ , etc. (break ties arbitrarily).

$x_j$  covered first by  $S_i \implies |S \setminus S_{<i}| \geq j$

(since  $S \setminus S_{<i}$  contains  $x_j, x_{j-1}, \dots, x_1$ )

$$|S_i \setminus S_{<i}| \geq |S \setminus S_{<i}| \geq j \implies c_{x_j} = \frac{1}{|S_i \setminus S_{<i}|} \leq \frac{1}{j}.$$



by greedy choice of  $S_i$

$$c(S) = c_{x_1} + c_{x_2} + \dots + c_{x_k} \leq 1 + \frac{1}{2} + \dots + \frac{1}{k} = H_{|S|}$$

# Using greedy algorithm for vertex cover

```
GREEDY-VERTEX-COVER( $G$ )
```

```
     $C := \emptyset$ 
```

```
    while  $E \neq \emptyset$ 
```

```
        Choose  $v \in V$  of maximum degree
```

```
         $C := C \cup \{v\}$ 
```

```
        Remove edges incident to  $v$  from  $E$ 
```

```
    return  $C$ 
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**Exercise:** Find graph  $G$  where GREEDY-VERTEX-COVER does not produce optimal solution.

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**Exercise:** Find graph  $G$  where GREEDY-VERTEX-COVER does not produce optimal solution.

The algorithm only gives a  $\Theta(\log |E|)$ -approximation.