

# Lecture: Approximation Algorithms

Jannik Matuschke



October 29, 2018

# Greedy Algorithms and Local Search

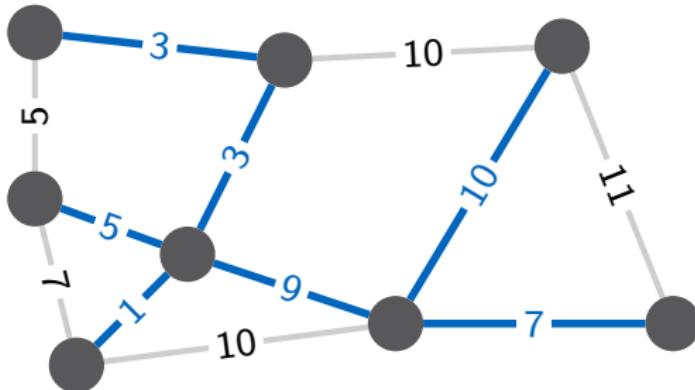
## Example I: Minimum Spanning Trees

# The Minimum Spanning Tree Problem

Input: graph  $G = (V, E)$ , distances  $d : E \rightarrow \mathbb{R}_+$

Task: find a spanning tree  $T$  in  $G$

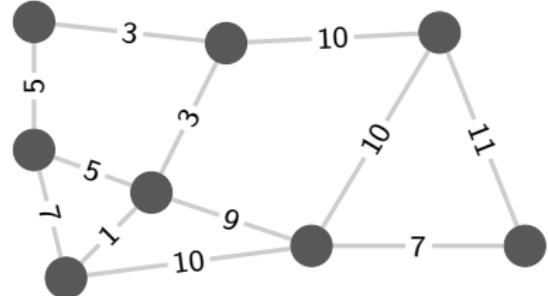
$$\text{minimizing } \sum_{e \in T} d(e)$$



# Kruskal's algorithm

## Algorithm:

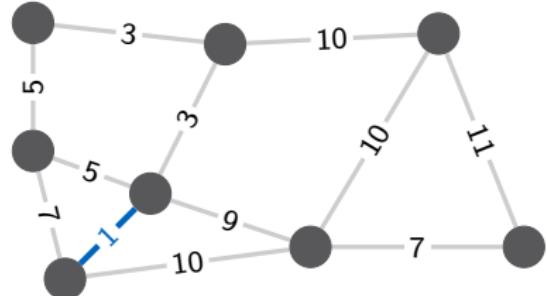
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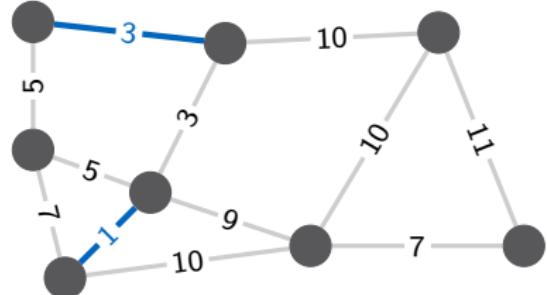
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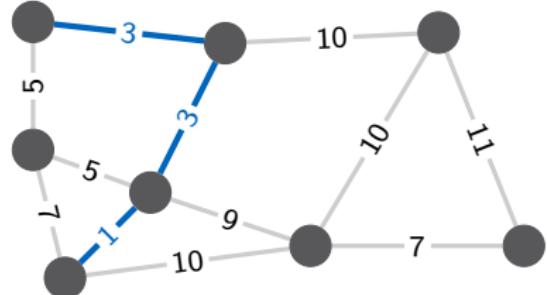
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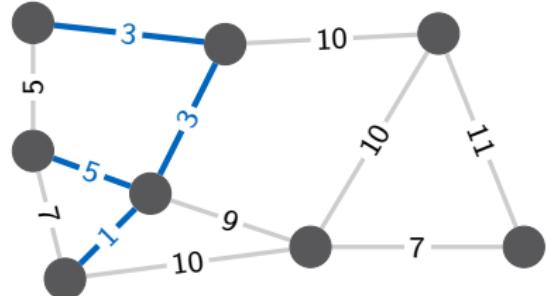
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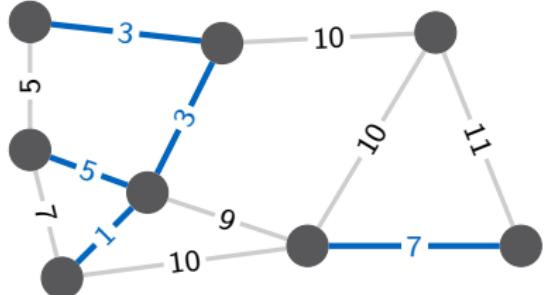
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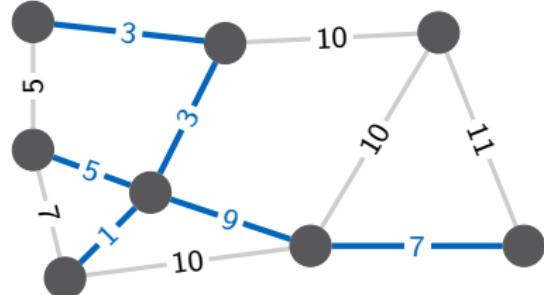
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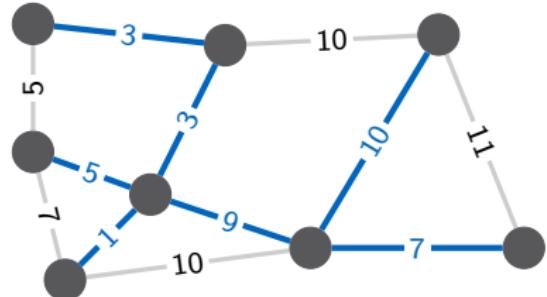
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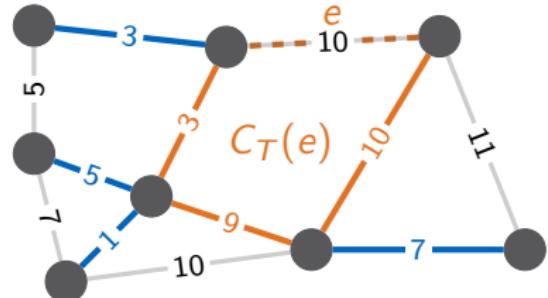
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- ▶ For  $e \in E \setminus T$ , let  $C_T(e)$  be the unique cycle in  $T \cup \{e\}$ .



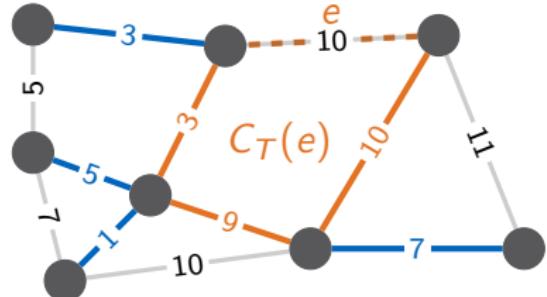
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## Definition

- ▶ For  $e \in E \setminus T$ , let  $C_T(e)$  be the unique cycle in  $T \cup \{e\}$ .
- ▶ A tree  $T$  is **swap-optimal** if  $d(e) \geq d(f)$  for all  $e \in E \setminus T, f \in C_T(e)$ .



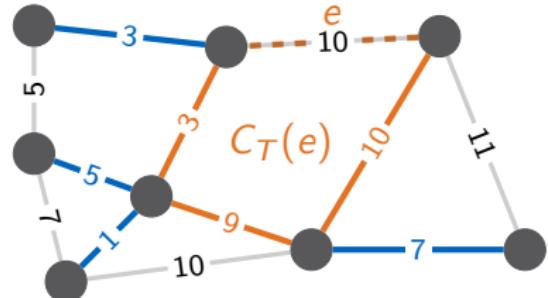
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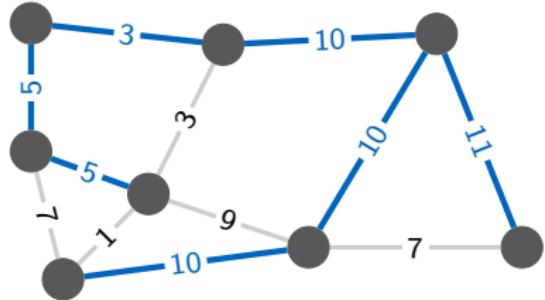
## Lemma 3.1

Kruskal's algorithm returns a swap-optimal tree.

# Local Search

## Algorithm:

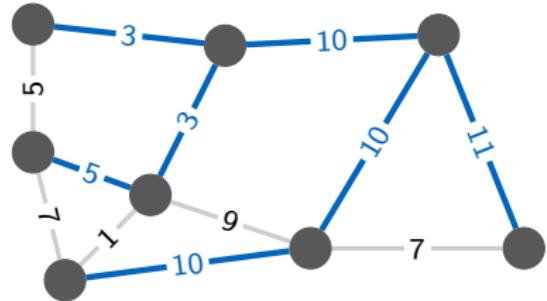
- 1  $T :=$  some spanning tree
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# Local Search

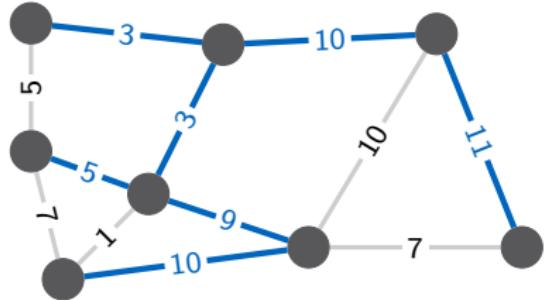
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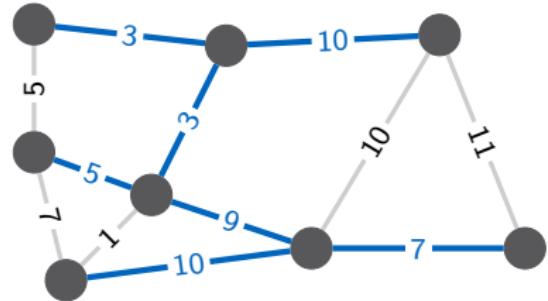
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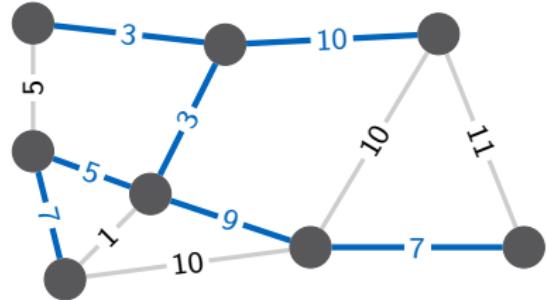
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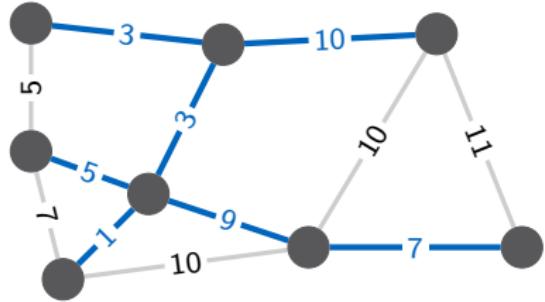
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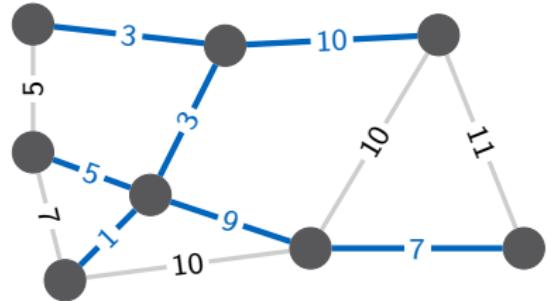


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## Lemma 3.2

The local search algorithm returns a swap-optimal tree.

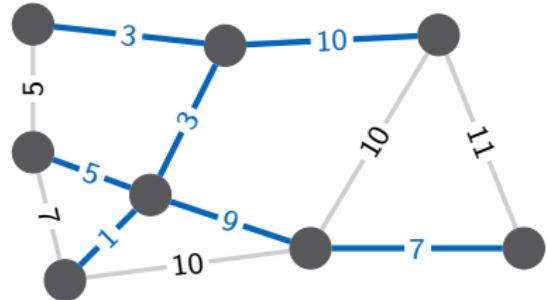


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## Theorem 3.3

A tree is a minimum spanning tree if and only if it is swap-optimal.

# Greedy Algorithms and Local Search

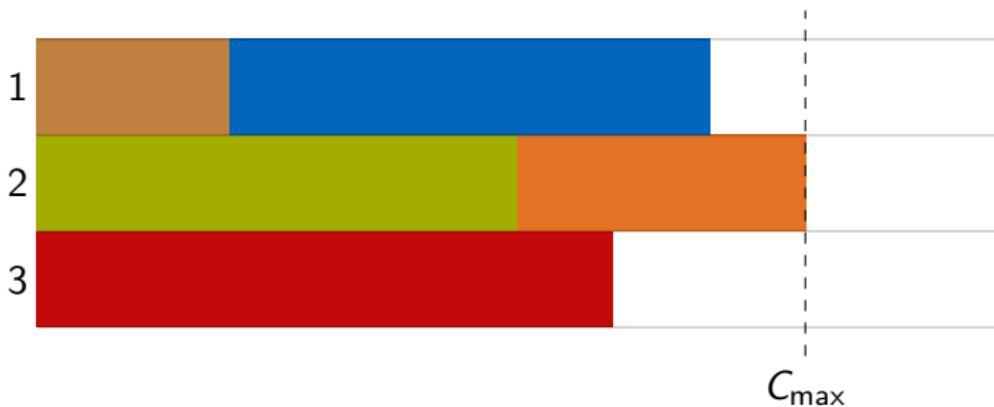
Example II: Scheduling on  
Identical Parallel Machines

# Scheduling on Parallel Machines

**Input:**  $m$  identical machines,

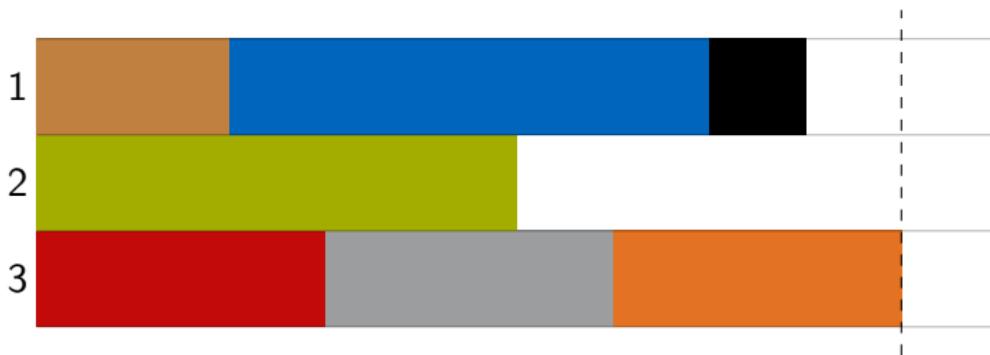
$n$  jobs with processing times  $p_1, \dots, p_n$

**Task:** assign each job  $j \in [n]$  to a machine  $\sigma(j) \in [m]$   
minimizing  $C_{\max} := \max_{i \in [m]} \sum_{j : \sigma(j)=i} p_j$



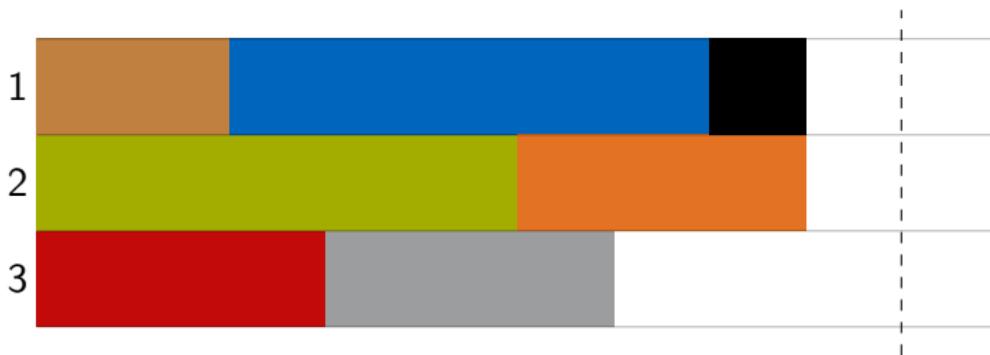
## Algorithm:

- 1 Let  $\sigma$  be some assignment.
- 2 while ( $\exists i \in [m], j \in [n] : \text{load}_\sigma(i) + p_j < \text{load}_\sigma(\sigma(j))$ )  
    Set  $\sigma(j) := i$ .
- 3 Return  $\sigma$ .



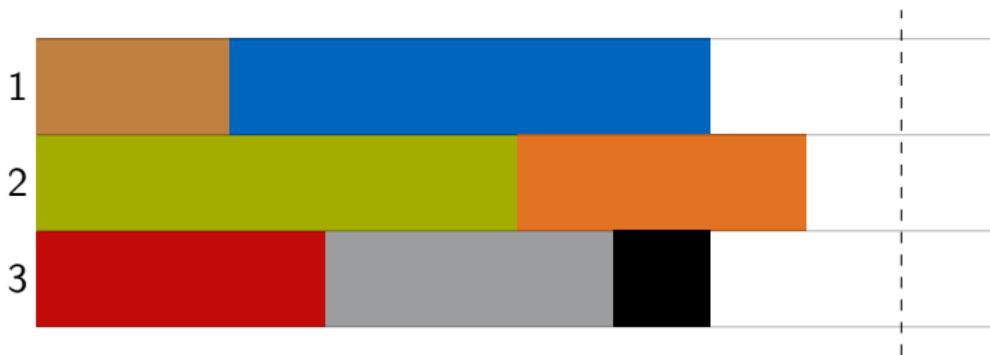
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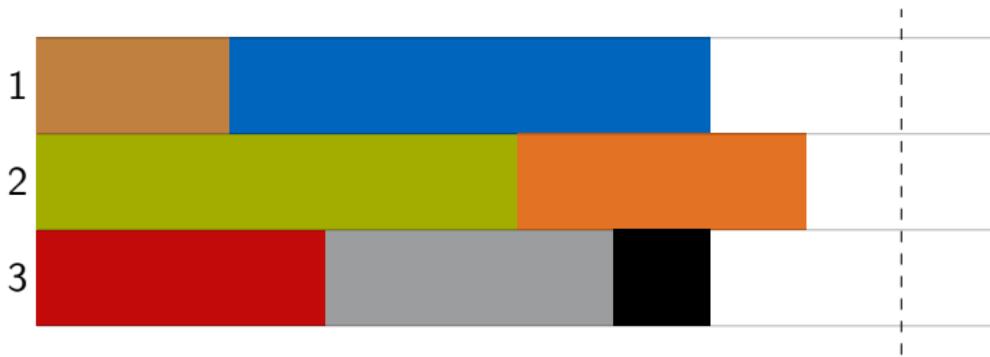
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## Theorem 3.4

Local search is a 2-approximation for  $P||C_{\max}^*$ .

# List Scheduling

## Algorithm:

1 For  $j := 1$  to  $n$

    Assign  $j$  to machine  $i$  with lowest load.



1

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2

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3

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## Theorem 3.5

List Scheduling is a 2-approximation for  $P||C_{\max}$ .

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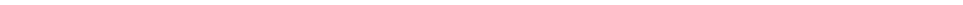
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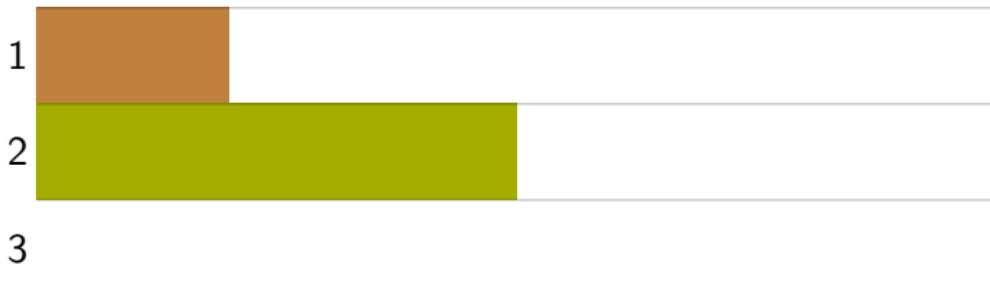
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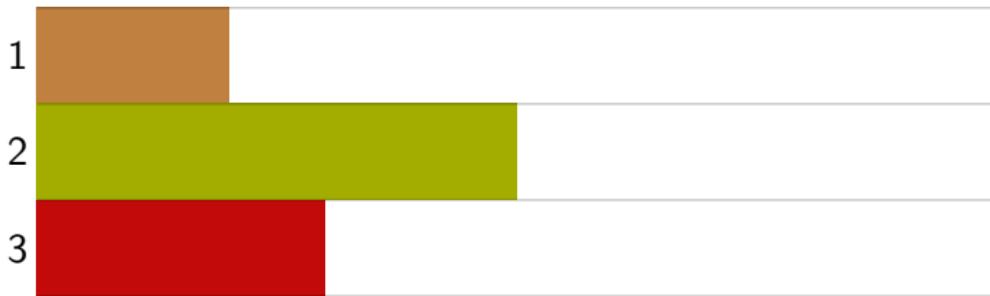
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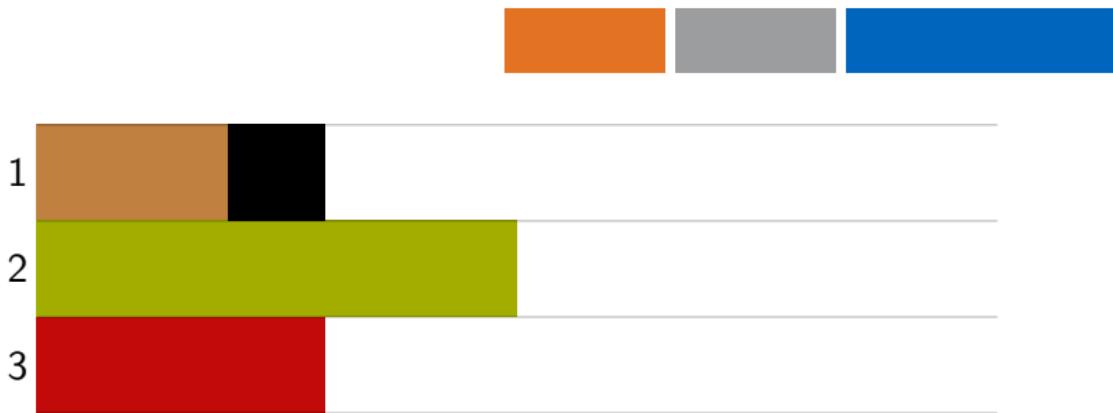
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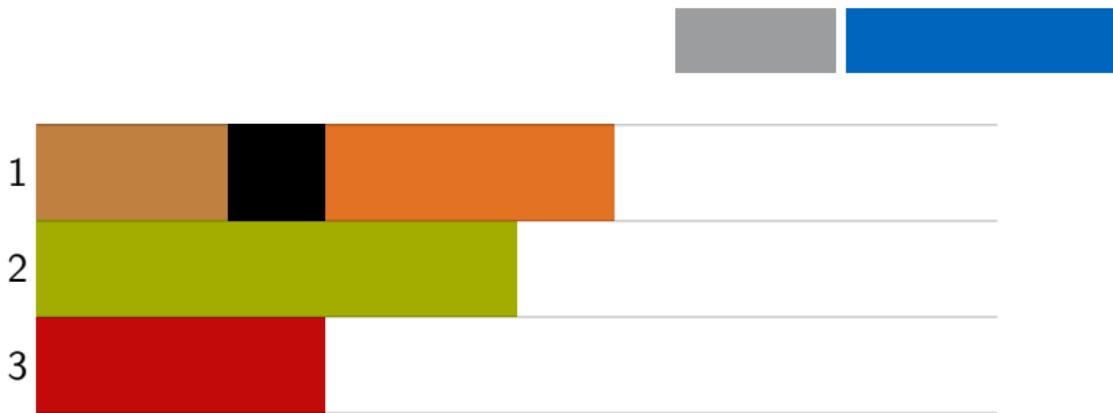
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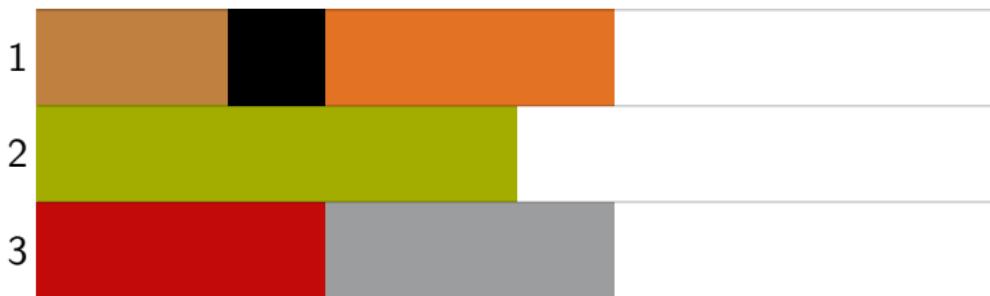
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# Longest Processing Time First

## Algorithm:

- 1 Order jobs such that  $p_1 \geq p_2 \geq \dots \geq p_n$ .
- 2 For  $j := 1$  to  $n$   
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## Theorem 3.6

LPT List Scheduling is a  $4/3$ -approximation for  $P||C_{\max}$ .

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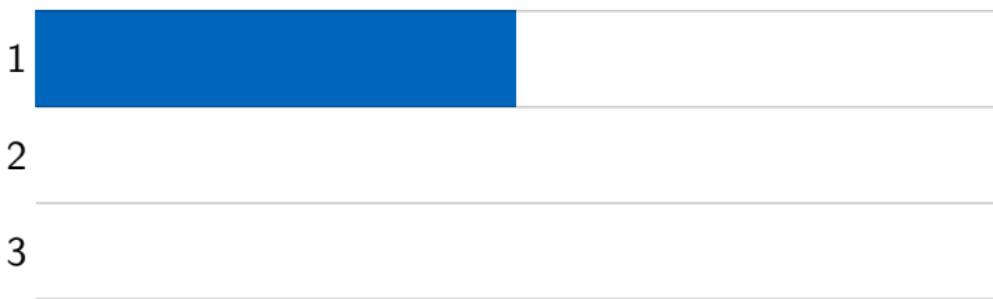
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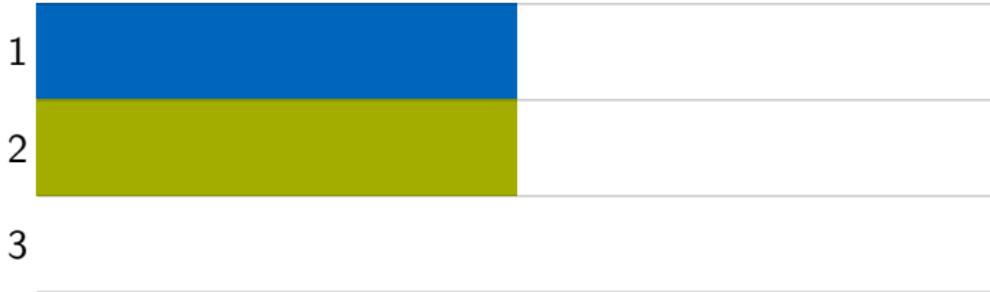
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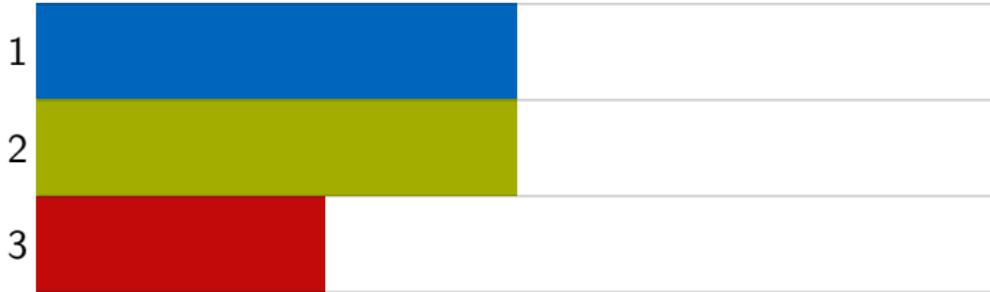
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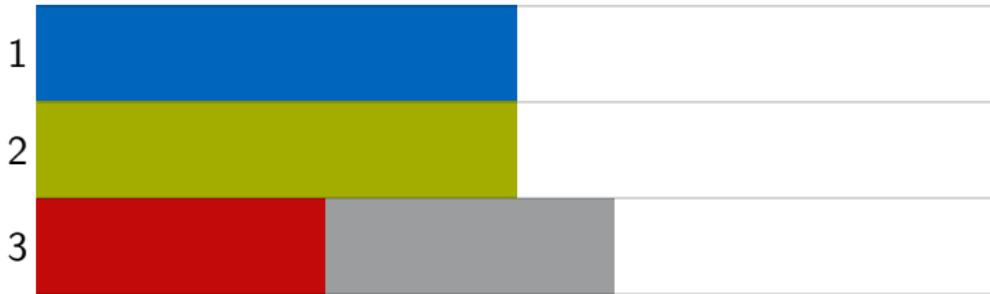
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# Greedy Algorithms

Example III: The  $k$ -Center Problem

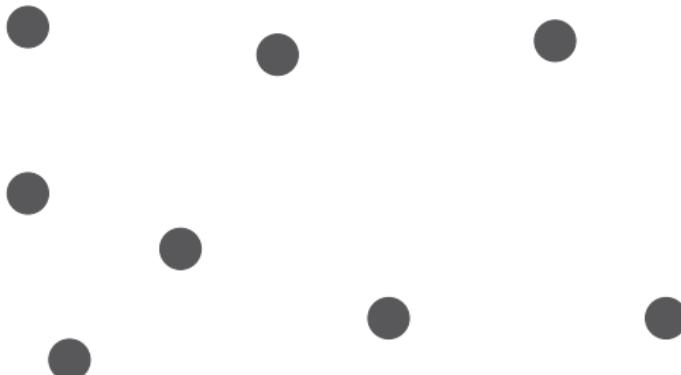
# The $k$ -Center Problem

**Input:** clients  $V$ , metric  $d : V \times V \rightarrow \mathbb{R}_+$

**Task:** find  $S \subseteq V$  with  $|S| \leq k$ ,

$$\text{minimizing } \max_{v \in V} d(v, S)$$

$$\text{where } d(v, S) := \min_{s \in S} d(v, s)$$



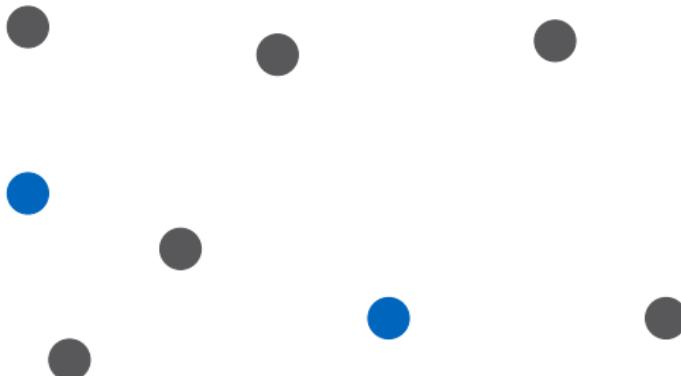
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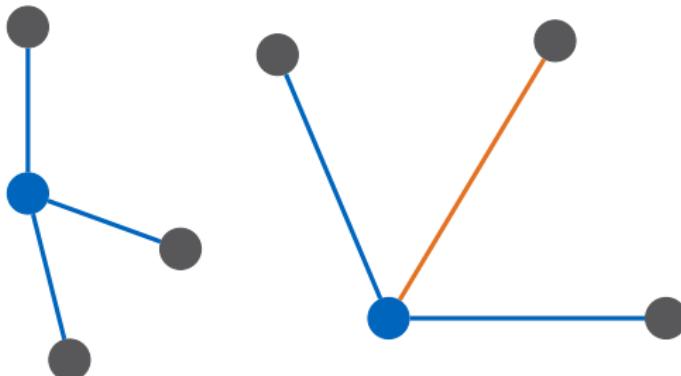
# The $k$ -Center Problem

**Input:** clients  $V$ , metric  $d : V \times V \rightarrow \mathbb{R}_+$

**Task:** find  $S \subseteq V$  with  $|S| \leq k$ ,

minimizing  $\max_{v \in V} d(v, S)$

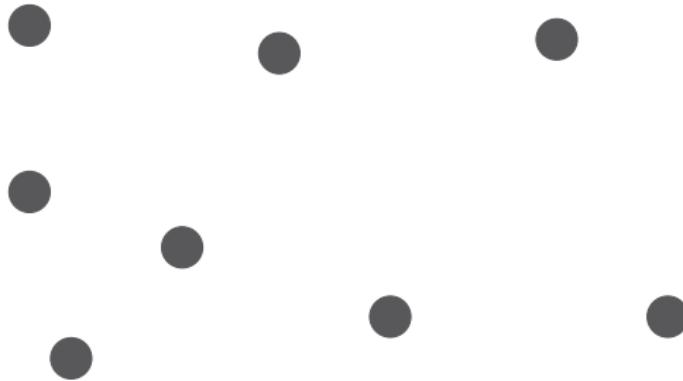
where  $d(v, S) := \min_{s \in S} d(v, s)$



# Greedy algorithm

## Algorithm:

- 1 Let  $S := \{v_0\}$  for some  $v_0 \in V$ .
- 2 while ( $|S| < k$ )
  - Let  $v \in \operatorname{argmax} d(v, S)$ .
  - Add  $v$  to  $S$ .
- 3 Return  $S$ .



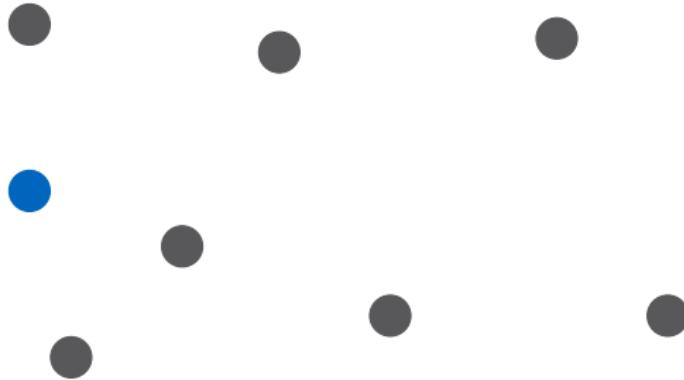
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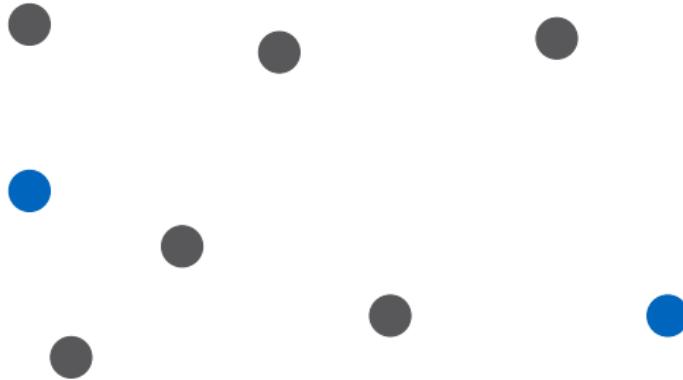
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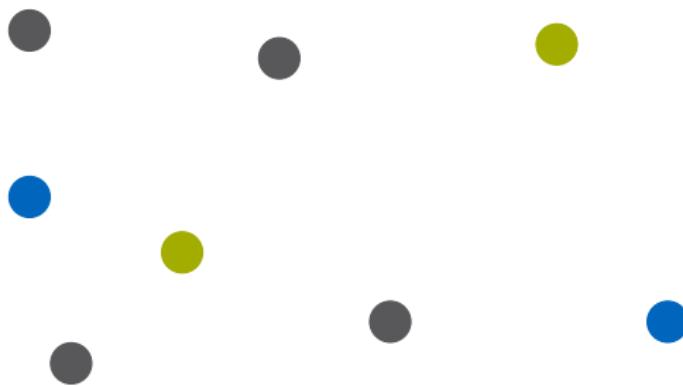
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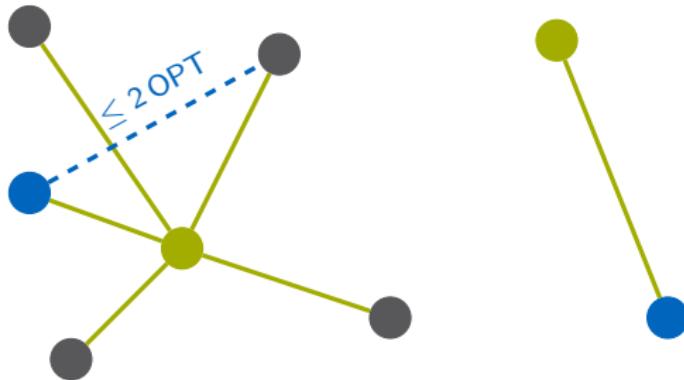
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# Hardness of Approximation

## Theorem 3.8

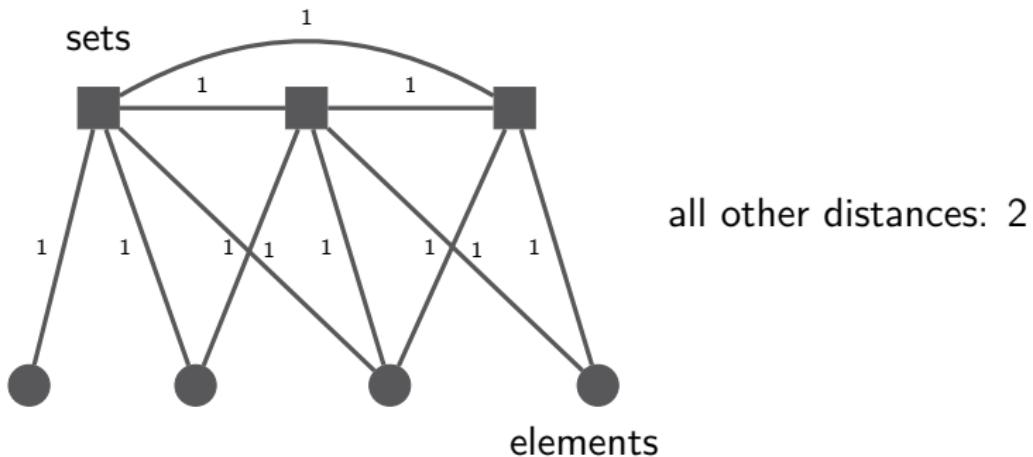
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### Reduction from SET COVER:

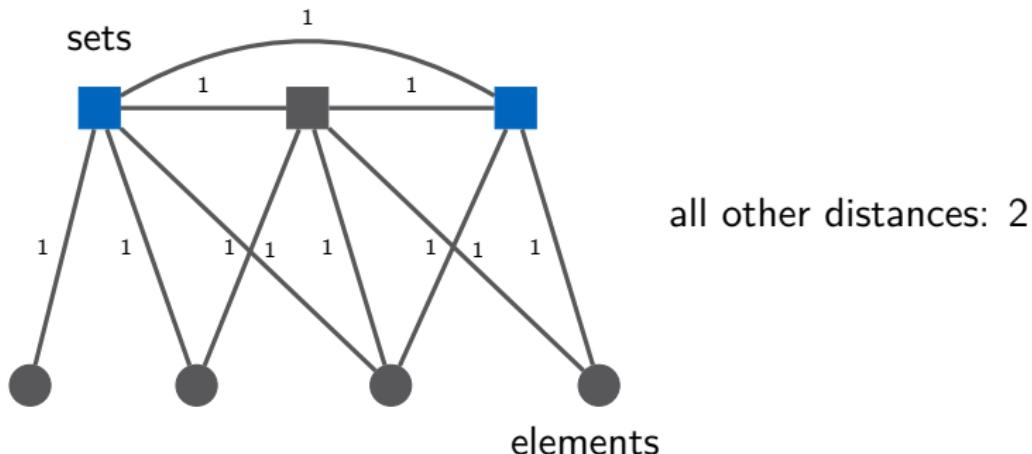


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**Reduction from SET COVER:**



$OPT = 1$  if and only if there is set cover of size  $\leq k$ .

Greed can be good.\*

\*If you know how to analyze it.

Local Search:

If the grass is greener on the other side,  
move into your neighbor's house.  
(and repeat)