

# Coping with NP-completeness: Approximation Algorithms

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Advanced Algorithms and Complexity  
Data Structures and Algorithms

# Outline

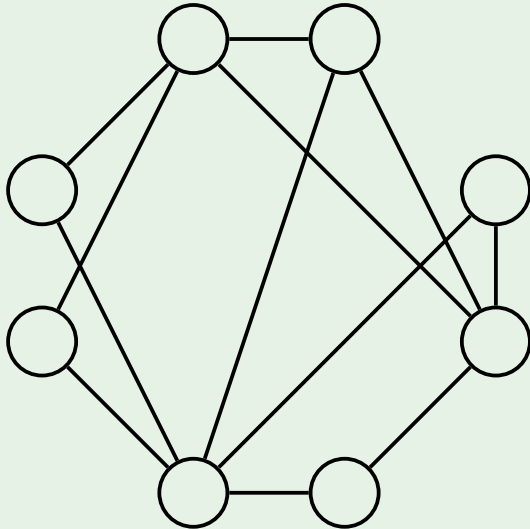
- 1 Vertex cover
- 2 Traveling salesman
  - Metric TSP
  - Local search

## Vertex cover (optimization version)

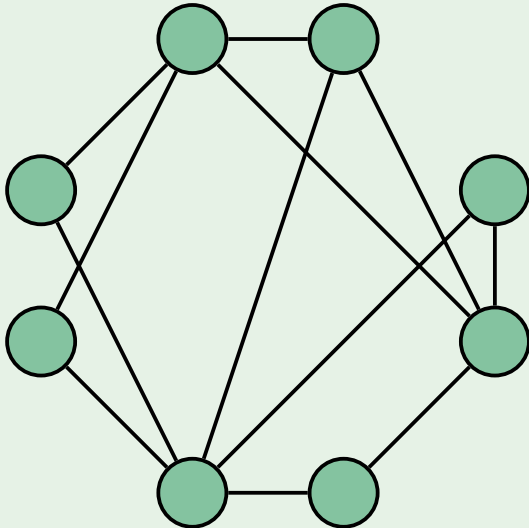
Input: A graph.

Output: A subset of vertices of minimum size that touches every edge.

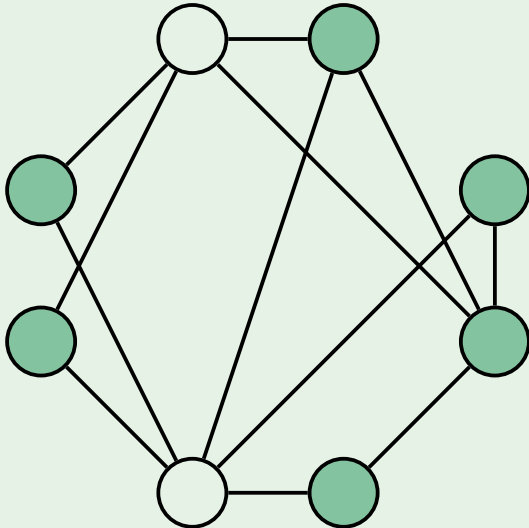
# Example



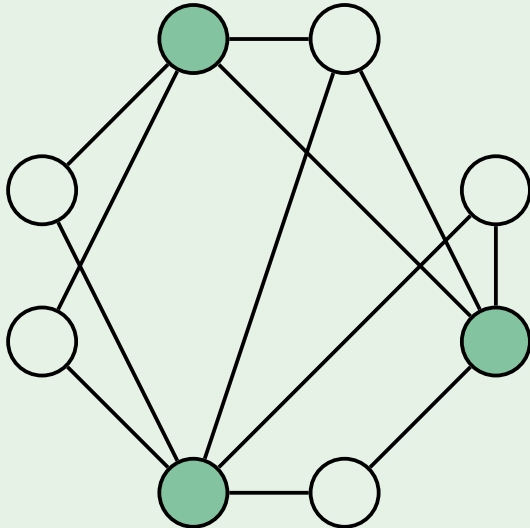
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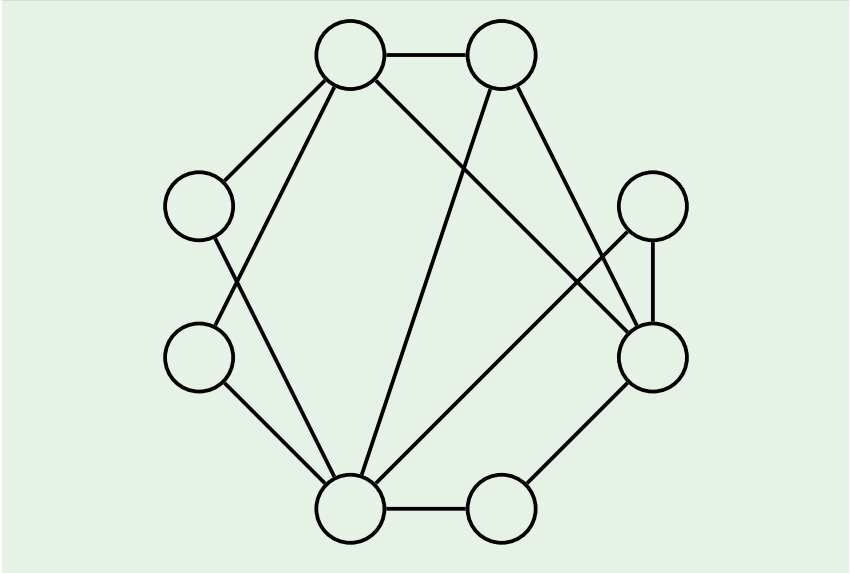


## ApproxVertexCover( $G(V, E)$ )

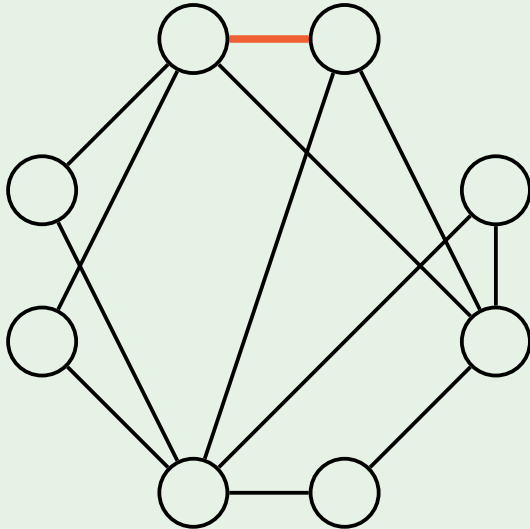
```
 $C \leftarrow$  empty set  
while  $E$  is not empty:  
     $\{u, v\} \leftarrow$  any edge from  $E$   
    add  $u, v$  to  $C$   
    remove from  $E$  all edges incident to  $u, v$   
return  $C$ 
```



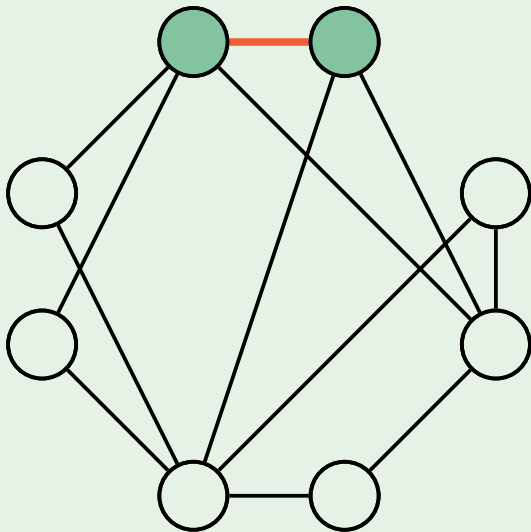
## Example



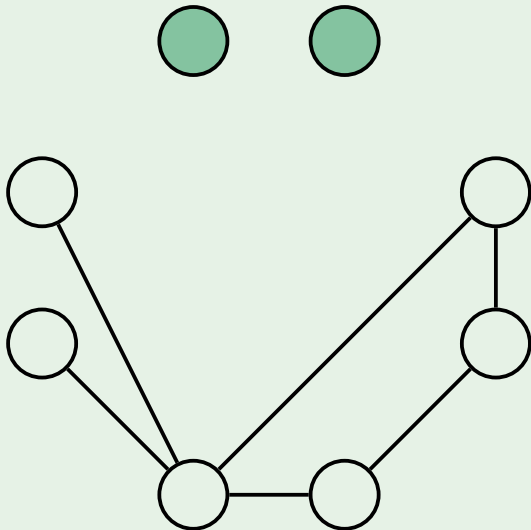
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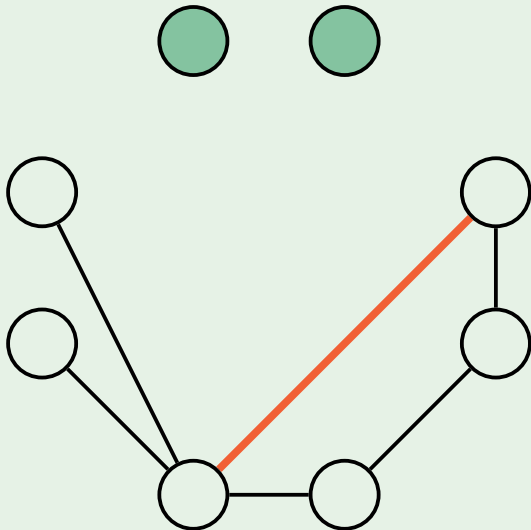
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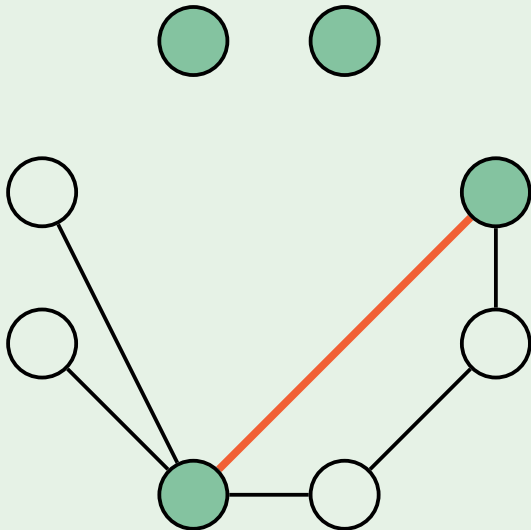
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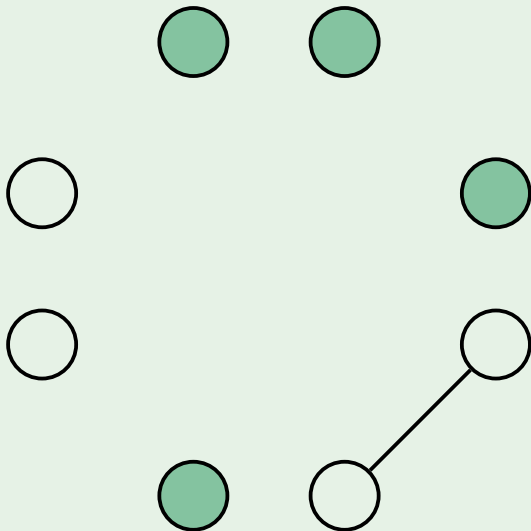
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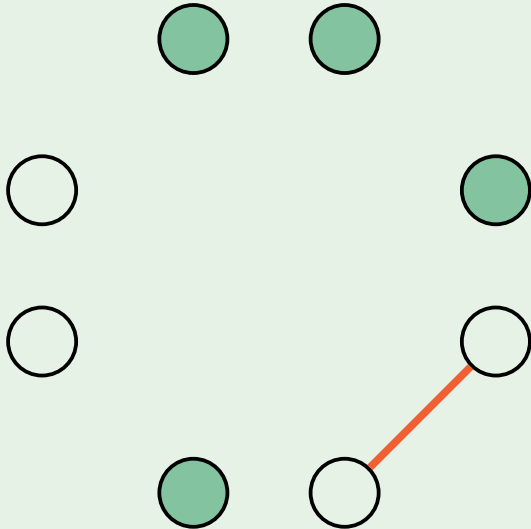
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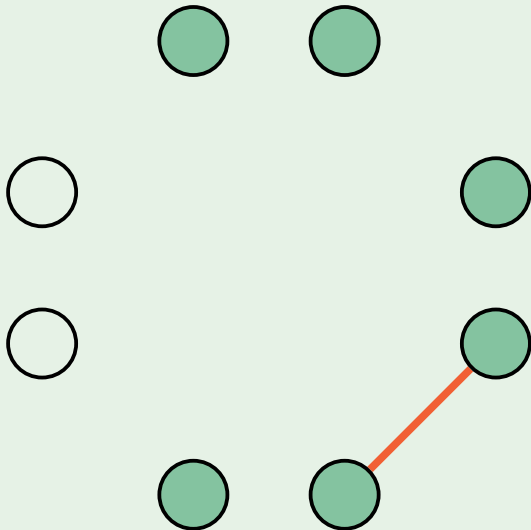


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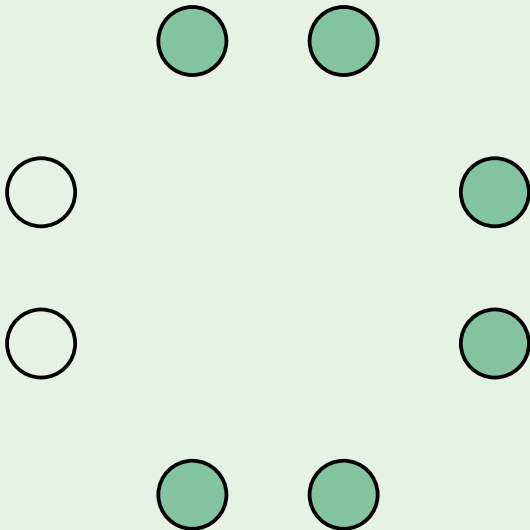




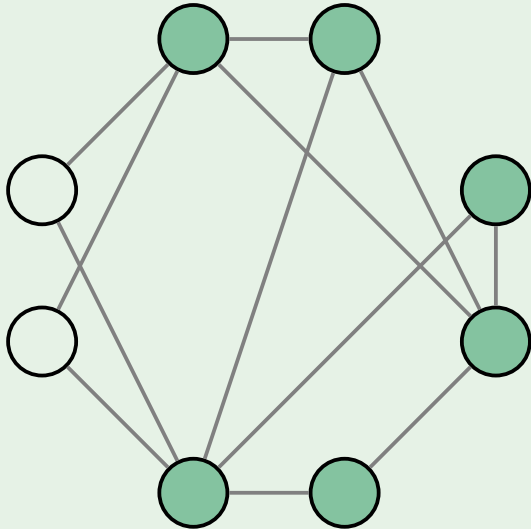
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## Lemma

The algorithm `ApproxVertexCover` is 2-approximate: it returns a vertex cover that is at most twice as large as an optimal one and runs in polynomial time.

# Proof

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- Any vertex cover of the graph has size at least  $|M|$

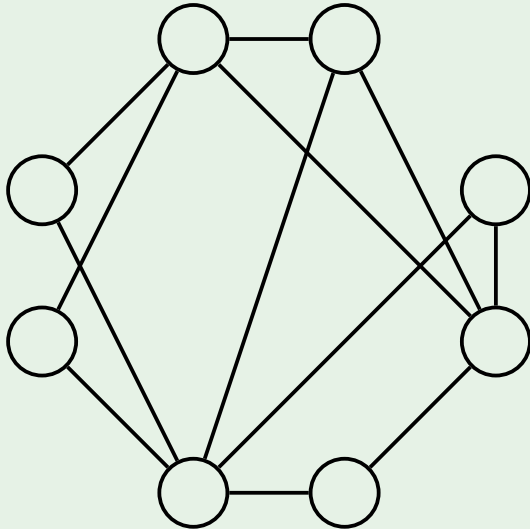
# Proof

- The set  $M$  of all edges selected by the algorithm forms a matching
- Any vertex cover of the graph has size at least  $|M|$
- The algorithm returns a vertex cover  $C$  of size  $2|M|$ , hence

$$|C| = 2 \cdot |M| \leq 2 \cdot \text{OPT}$$

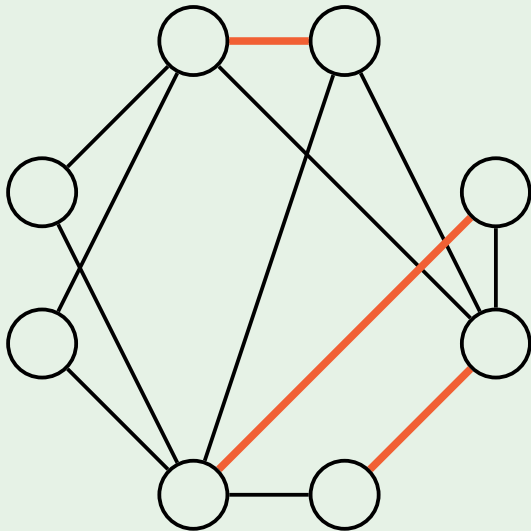


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- This is because we know a **lower bound** on OPT: it is at least the size of any matching

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# Final Remarks

- The bound is tight: there are graphs for which the algorithm returns a vertex cover of size twice the minimum size.

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- No 1.99-approximation algorithm is known.

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## Metric TSP (optimization version)

**Input:** An undirected graph  $G(V, E)$  with non-negative edge weights satisfying the triangle inequality: for all  $u, v, w \in V$ ,  
$$d(u, v) + d(v, w) \geq d(u, w).$$

**Output:** A cycle of minimum total length visiting each vertex exactly once .



# Lower Bound

- We are going to design a 2-approximation algorithm: it returns a cycle that is at most twice as long as an optimal cycle:  $C \leq 2 \cdot \text{OPT}$

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- Since we don't know the value of  $\text{OPT}$ , we need a good lower bound  $L$  on  $\text{OPT}$ :

$$C \leq 2 \cdot L \leq 2 \cdot \text{OPT}$$

# Minimum Spanning Trees

## Lemma

Let  $G$  be an undirected graph with non-negative edge weights. Then  $\text{MST}(G) \leq \text{TSP}(G)$ .

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## Proof

By removing any edge from an optimum TSP cycle one gets a spanning tree of  $G$ .  $\square$

## ApproxMetricTSP( $G$ )

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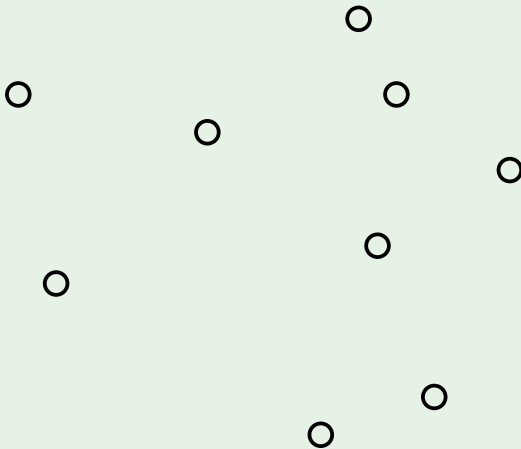
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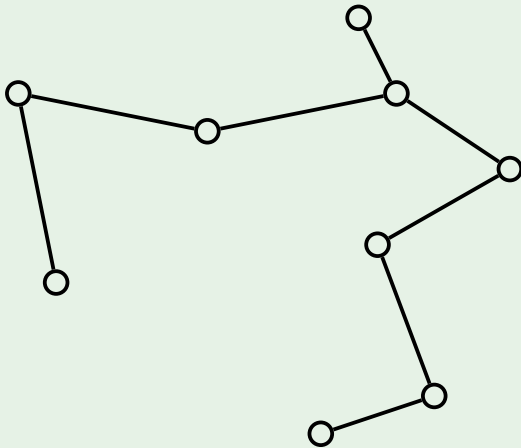
```
 $T \leftarrow$  minimum spanning tree of  $G$   
 $D \leftarrow T$  with each edge doubled  
find an Eulerian cycle  $C$  in  $D$   
return a cycle that visits vertices in  
the order of their first appearance in  $C$ 
```



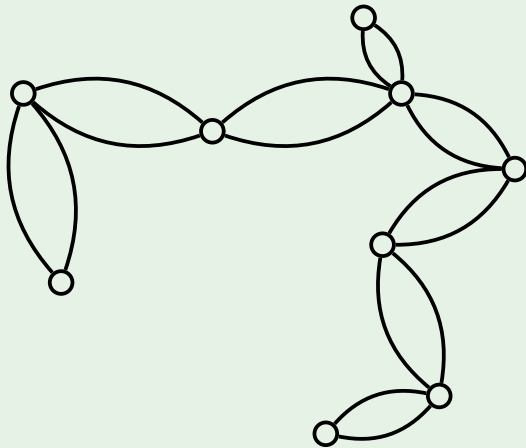
## Example: points on a plane



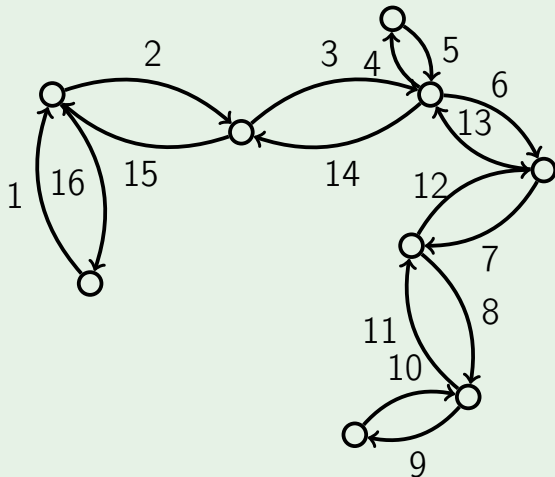
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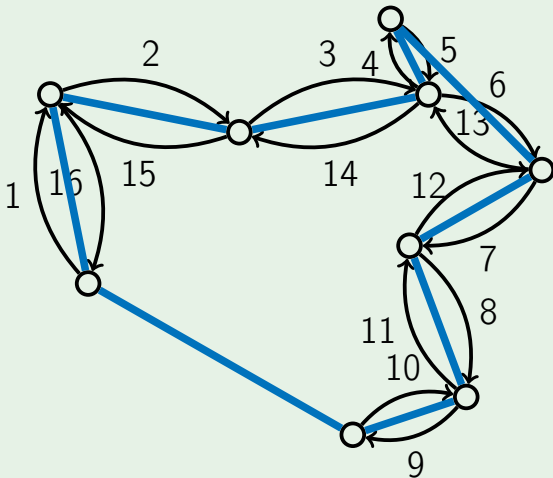
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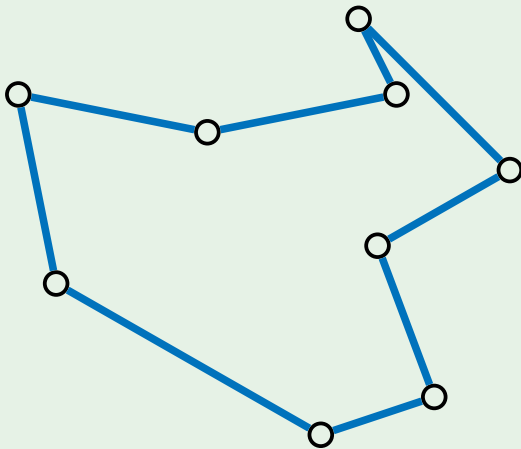
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- The total length of the MST  $T$  is at most  $\text{OPT}$ .
- Bypasses can only decrease the total length.



# Final Remarks

- The currently best known approximation algorithm for metric TSP is Christofides' algorithm that achieves a factor of 1.5

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- If  $\mathbf{P} \neq \mathbf{NP}$ , then there is no  $\alpha$ -approximation algorithm for the general version of TSP for any polynomial time computable function  $\alpha$

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# LocalSearch

```
s ← some initial solution
while there is a solution  $s'$  in the
neighborhood of  $s$  which is better than  $s$ :
    s ←  $s'$ 
return s
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```

- Computes a local optimum instead of a global optimum
- The larger is the neighborhood, the better is the resulting solution and the higher is the running time

# Local Search for TSP

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# Local Search for TSP

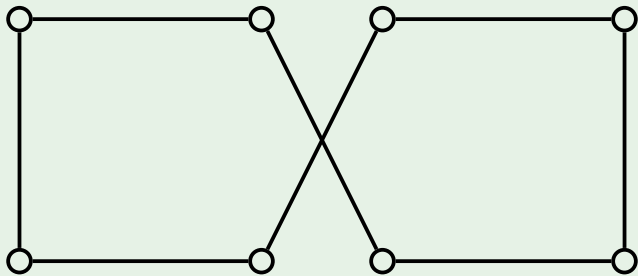
- Let  $s$  and  $s'$  be two cycles visiting each vertex of the graph exactly once
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- Neighborhood  $N(s, r)$  with center  $s$  and radius  $r$ : all cycles with distance at most  $r$  from  $s$

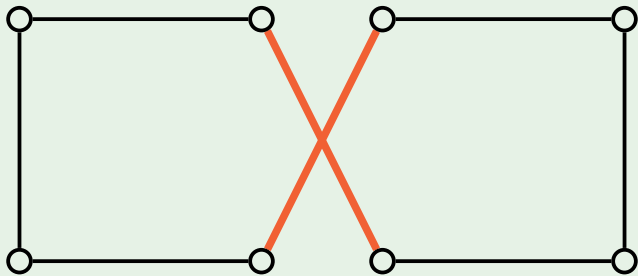
## Example

Changing two edges in a suboptimal solution:



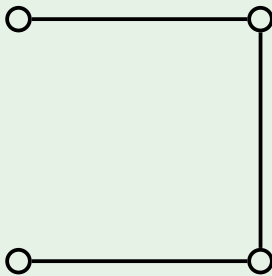
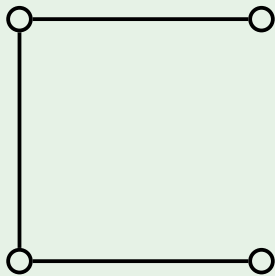
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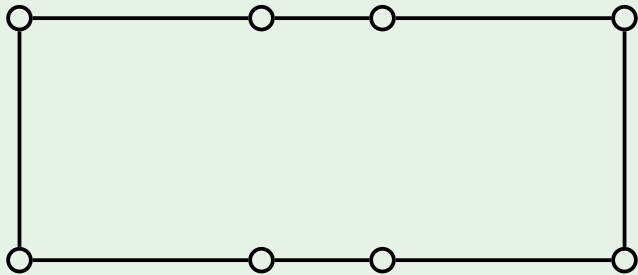
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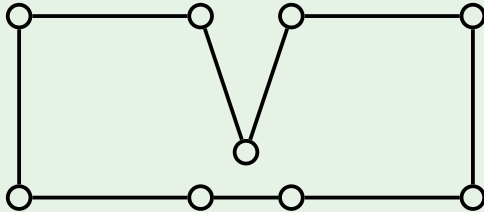
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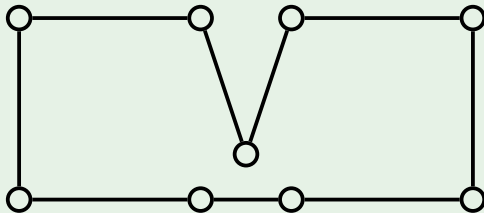
## Example

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Need to allow changing three edges to improve this solution



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- Still, the number of iterations may be exponential and the quality of the found cycle may be poor
- But works well in practice

# Coping with NP-completeness

- special cases
- intelligent exhaustive search
- approximation algorithms