

# Traveling Salesman

Given a set of cities  $(\{1, \dots, n\})$  and a symmetric matrix  $C = (c_{ij})$ ,  $c_{ij} \geq 0$  that specifies for every pair  $(i, j) \in [n] \times [n]$  the cost for travelling from city  $i$  to city  $j$ . Find a permutation  $\pi$  of the cities such that the round-trip cost

$$c_{\pi(1)\pi(n)} + \sum_{i=1}^{n-1} c_{\pi(i)\pi(i+1)}$$

is minimized.

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## Theorem 2

*There does not exist an  $O(2^n)$ -approximation algorithm for TSP.*

### Hamiltonian Cycle:

For a given undirected graph  $G = (V, E)$  decide whether there exists a simple cycle that contains all nodes in  $G$ .

- Given an instance to HAMILTONIAN PATH we create an instance for TSP.
- If  $G = (V, E)$  then set  $c_{ij} = |v_i - v_j|$  for every  $c_{ij} \in E$ . This is a polynomial reduction.
- Does there exist a Hamiltonian Path of length  $|V|$  in  $G$  if and only if there exists a TSP tour with cost  $|V|$  (only larger than  $|V|$  if there are edges that are not in  $E$ ).
- If  $O(2^n)$ -approximation algorithm could solve this then there would exist an  $O(2^n)$  algorithm for  $P = NP$ .

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- ▶ Given an instance to HAMPATH we create an instance for TSP.
  - ▶ If  $(i, j) \notin E$  then set  $c_{ij}$  to  $n2^n$  otw. set  $c_{ij}$  to 1. This instance has polynomial size.
  - ▶ There exists a Hamiltonian Path iff there exists a tour with cost  $n$ . Otw. any tour has cost strictly larger than  $2^n$ .
  - ▶ An  $O(2^n)$ -approximation algorithm could decide btw. these cases. Hence, cannot exist unless  $P = NP$ .

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# Metric Traveling Salesman

In the metric version we assume for every triple  
 $i, j, k \in \{1, \dots, n\}$

$$c_{ij} \leq c_{ij} + c_{jk} .$$

It is convenient to view the input as a complete undirected graph  
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# TSP: Lower Bound I

## Lemma 3

*The cost  $\text{OPT}_{TSP}(G)$  of an optimum traveling salesman tour is at least as large as the weight  $\text{OPT}_{MST}(G)$  of a minimum spanning tree in  $G$ .*

Proof:

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- ▶ Start with a tour on a subset  $S$  containing a single node.
- ▶ Take the node  $v$  closest to  $S$ . Add it  $S$  and expand the existing tour on  $S$  to include  $v$ .
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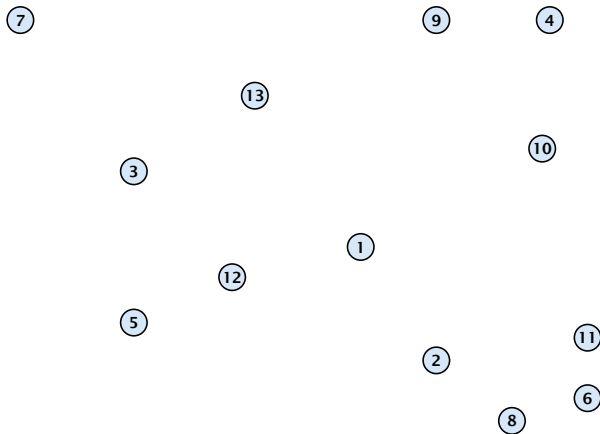
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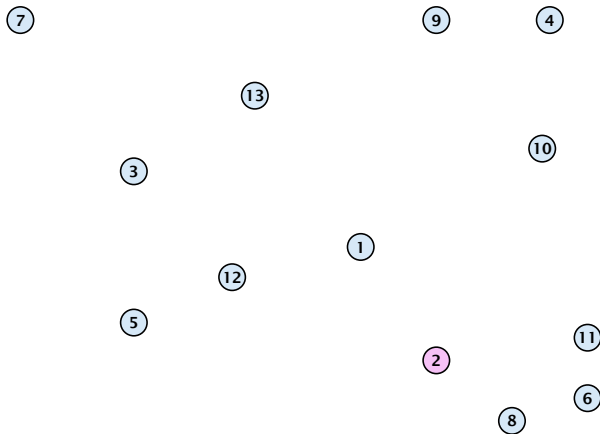


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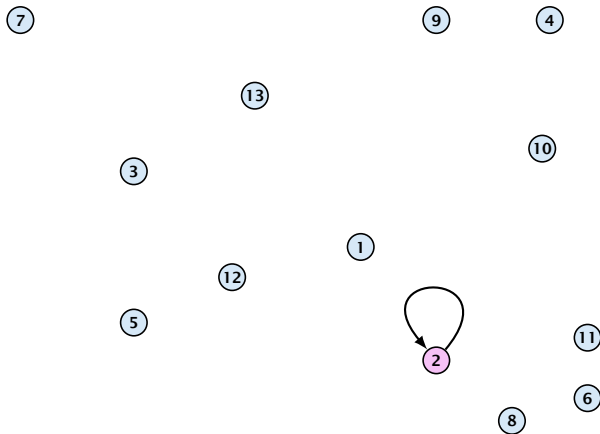
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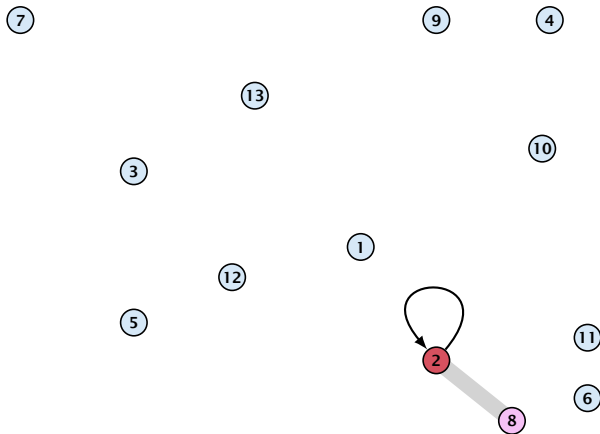
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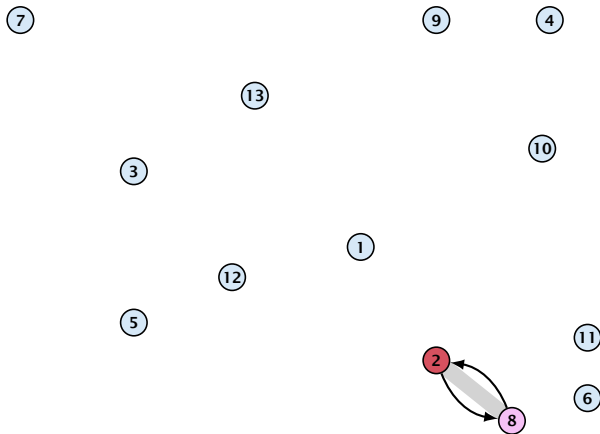
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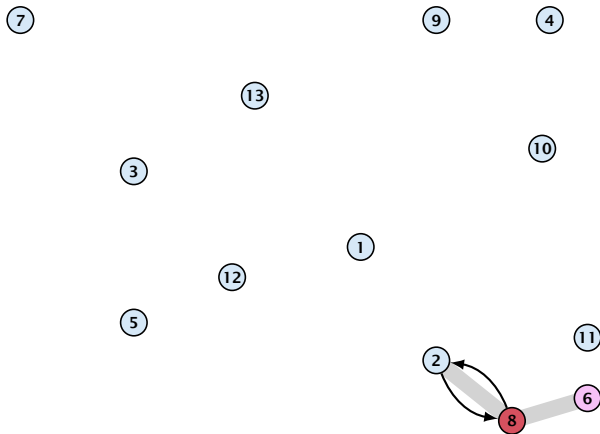
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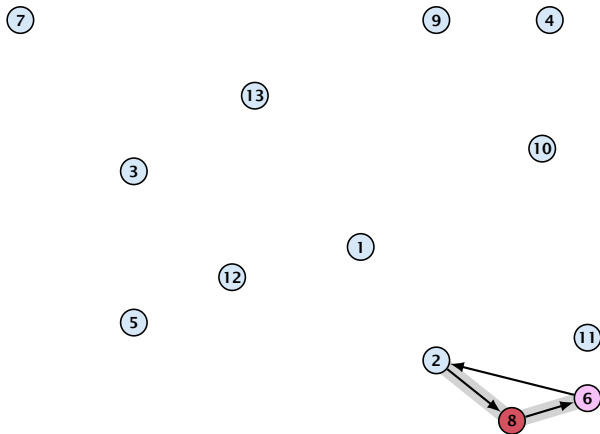
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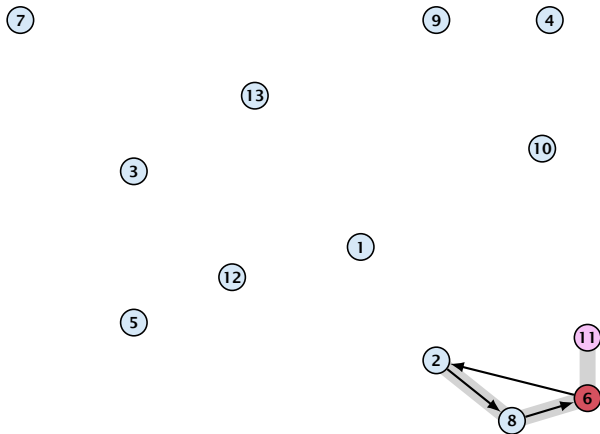
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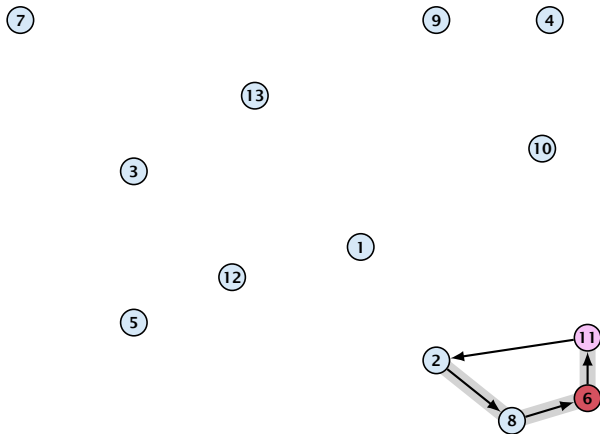
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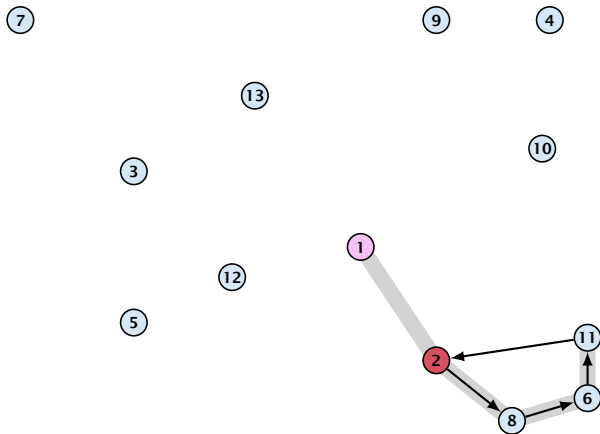


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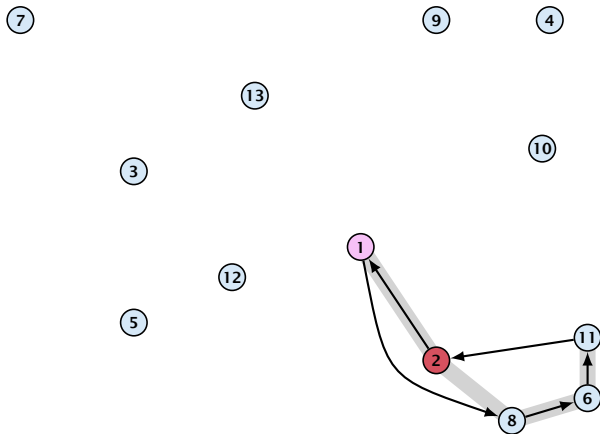
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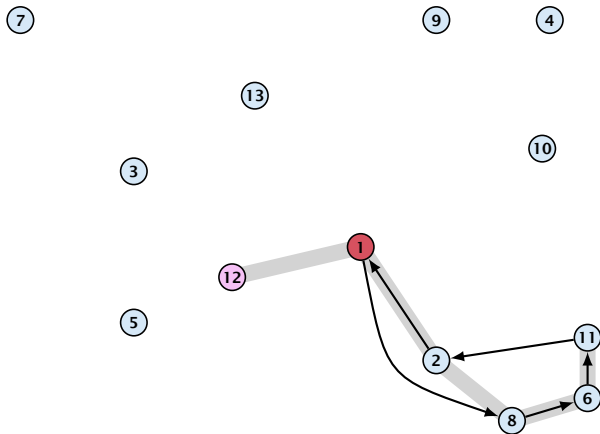
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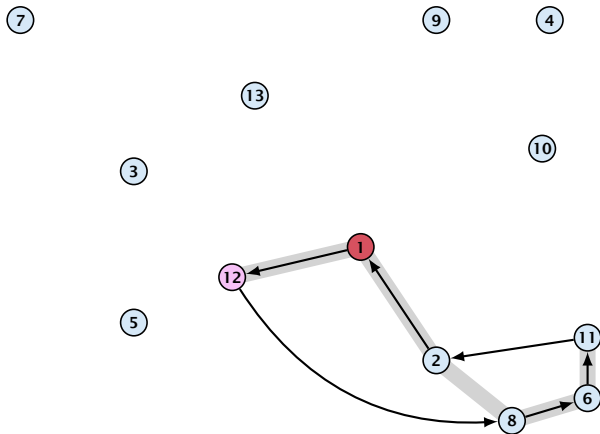
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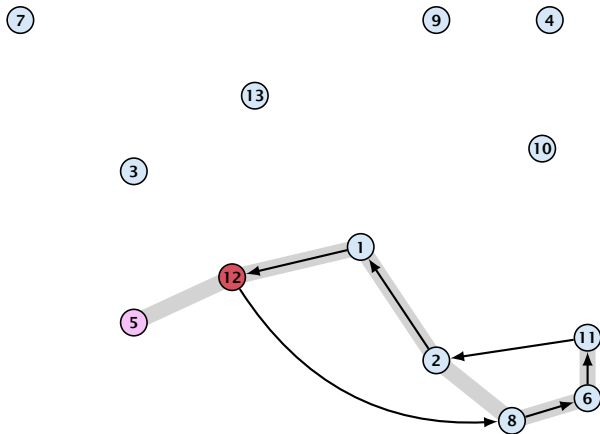
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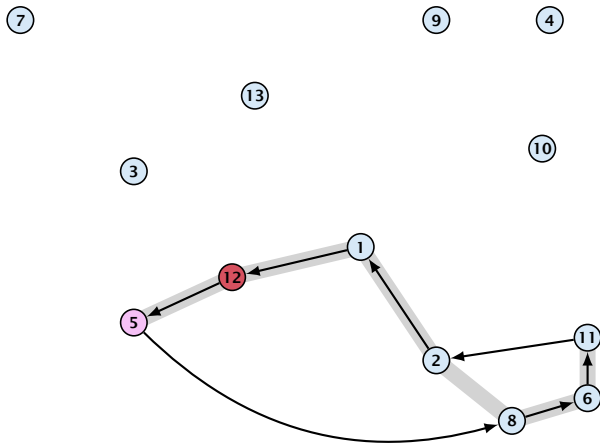
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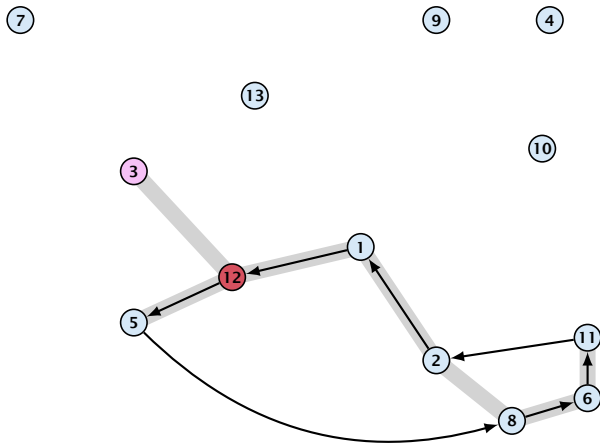
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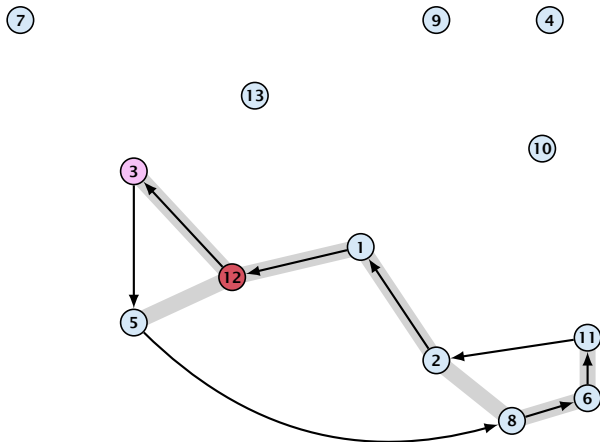
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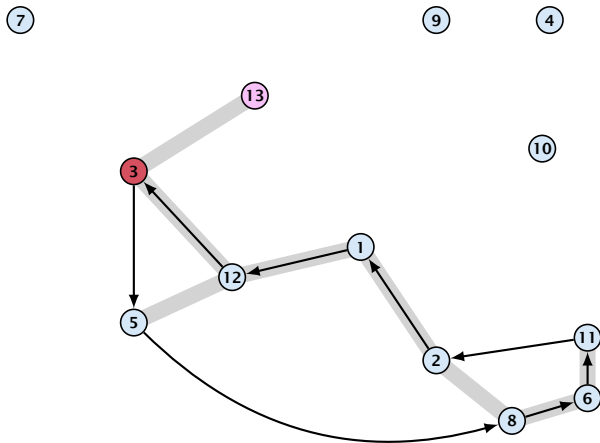


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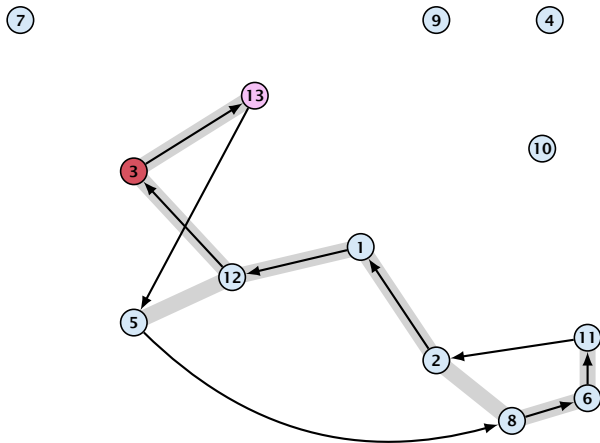
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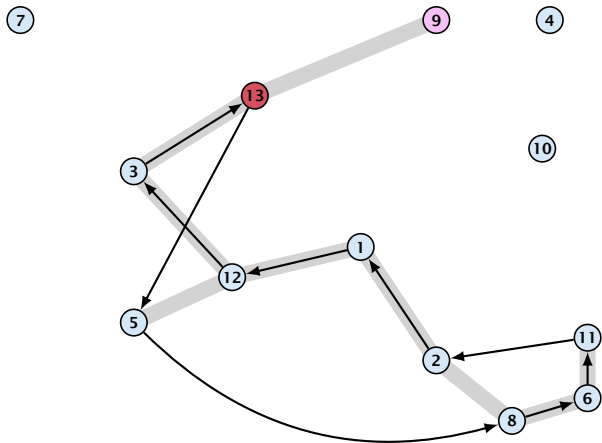
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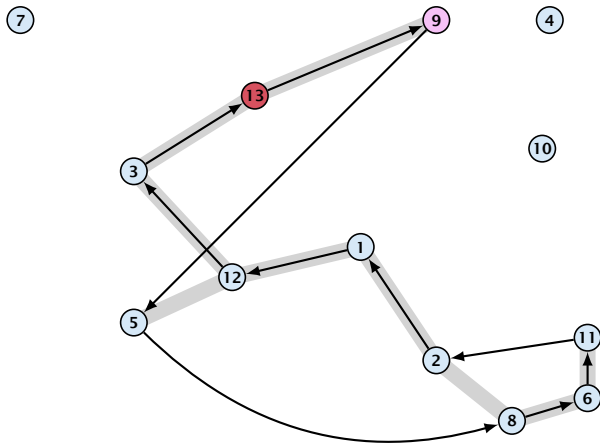
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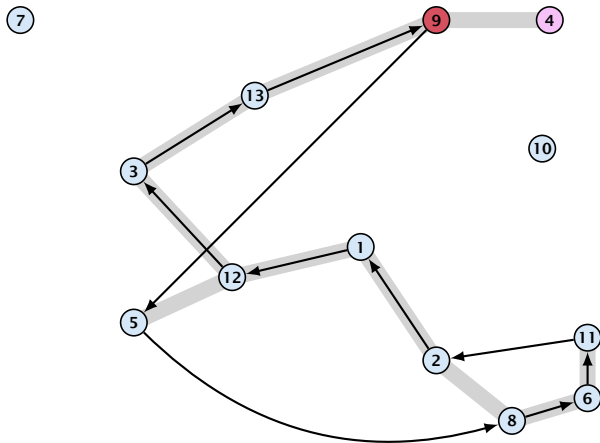
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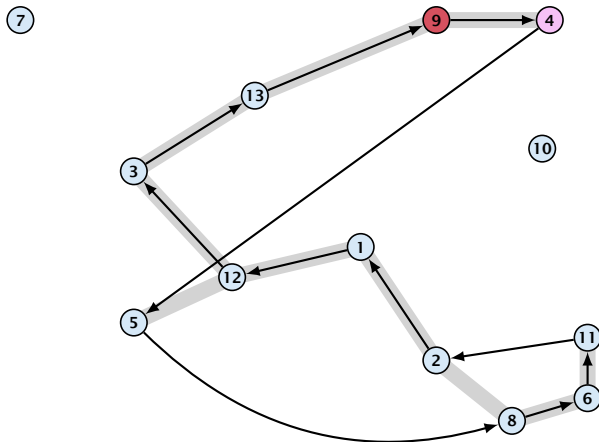
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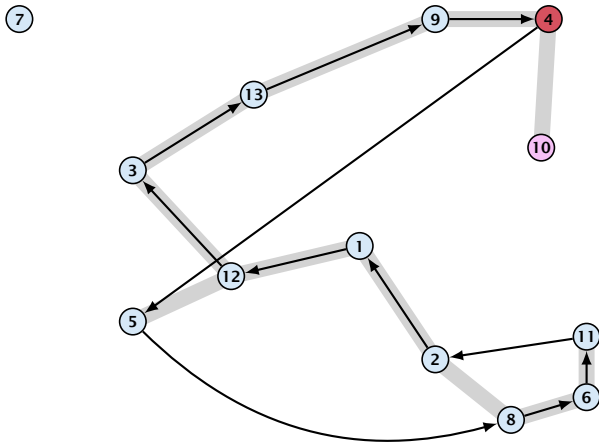
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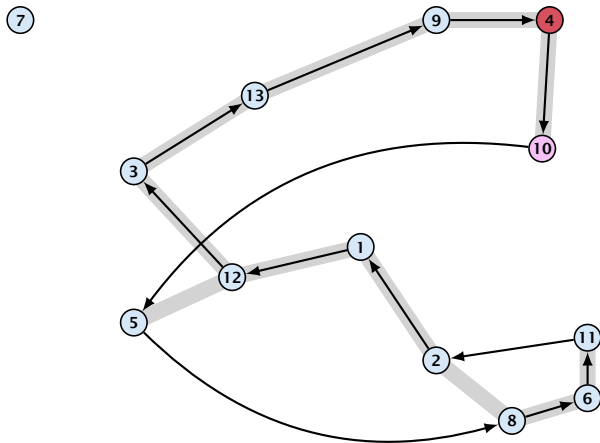
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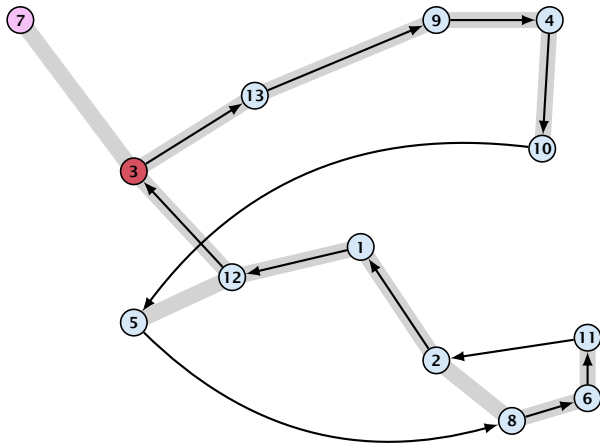


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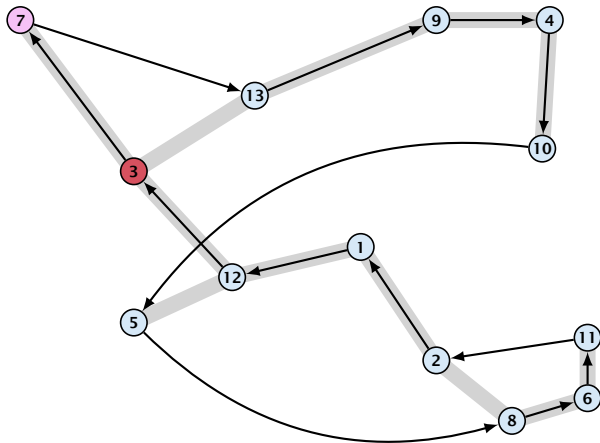
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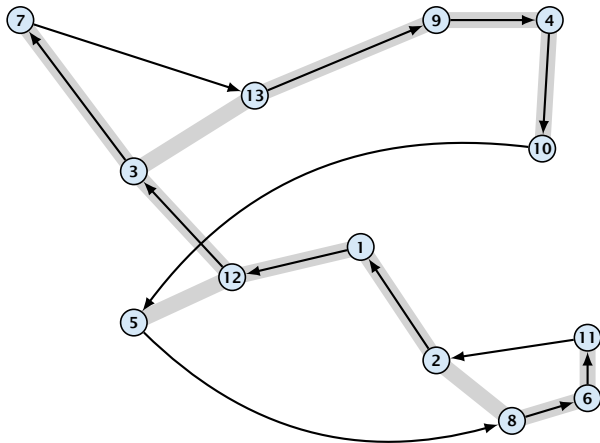
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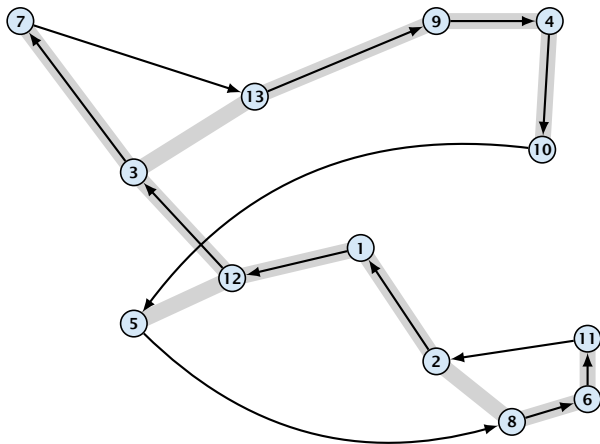
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*The Greedy algorithm is a 2-approximation algorithm.*

Let  $S_i$  be the set at the start of the  $i$ -th iteration, and let  $v_i$  denote the node added during the iteration.

Further let  $s_i \in S_i$  be the node closest to  $v_i \in S_i$ .

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$$c_{s_i, v_i} + c_{v_i, r_i} - c_{s_i, r_i} \leq 2c_{s_i, v_i}$$

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Suppose that we are given an Eulerian graph  $G' = (V, E', c')$  of  $G = (V, E, c)$  such that for any edge  $(i, j) \in E'$   $c'(i, j) \geq c(i, j)$ .

Then we can find a TSP-tour of cost at most

$$\sum_{e \in E'} c'(e)$$

Find an Euler tour of  $G'$ .

Fix a permutation of the edges (i.e., a TSP-tour) by traversing the Euler tour and only note the first occurrence of a city.

The cost of this TSP-tour is at most the cost of the Euler tour because of triangle inequality.

This technique is known as **short cutting** the Euler tour.

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This tour is a TSP-tour because of the triangle inequality.

(The triangle inequality is not needed here.)

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This technique is known as **short cutting** the Euler tour.

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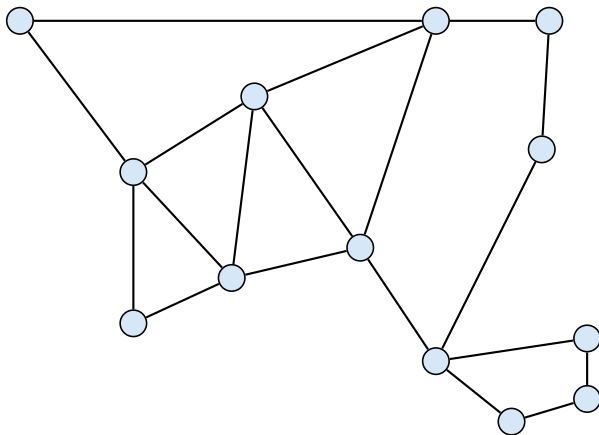
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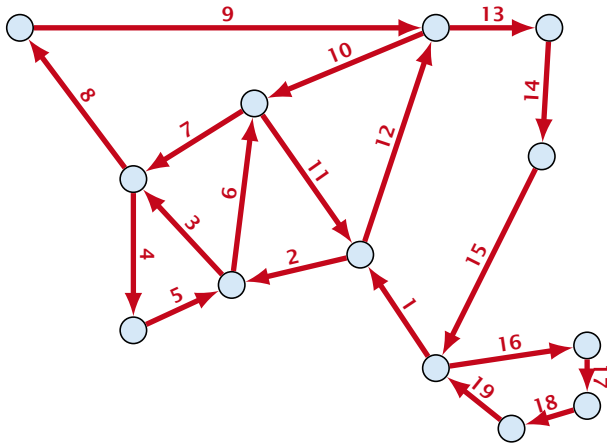
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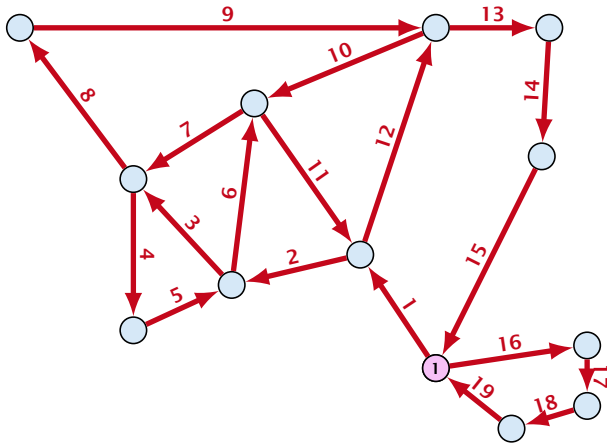
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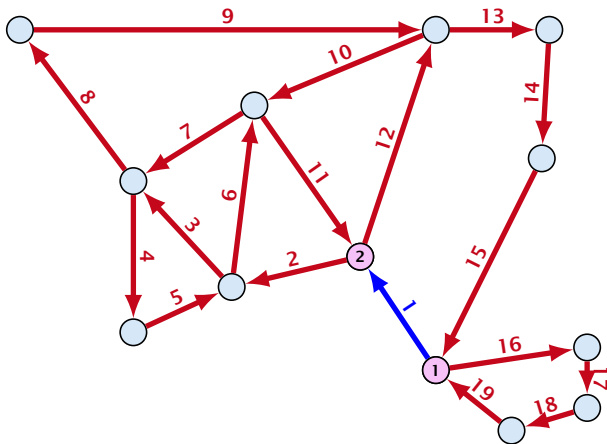
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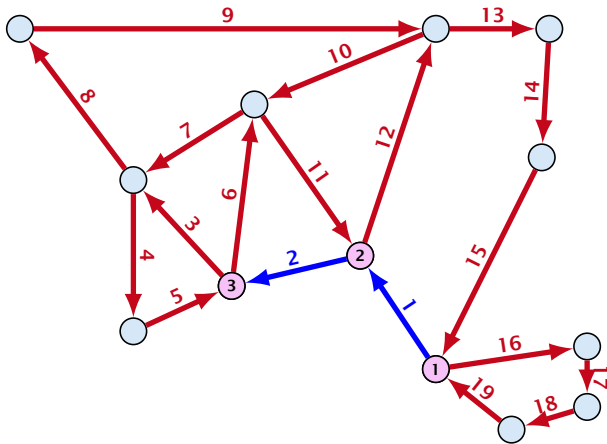


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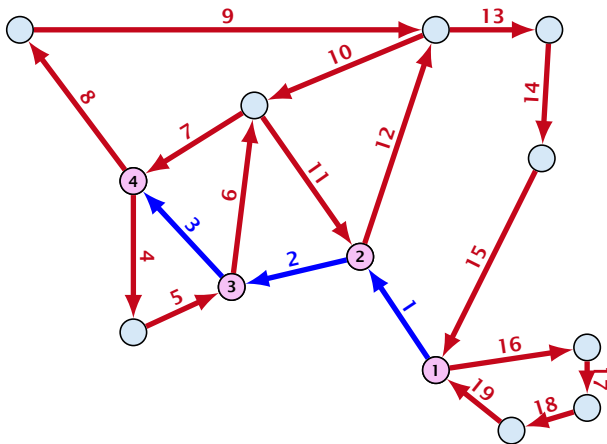




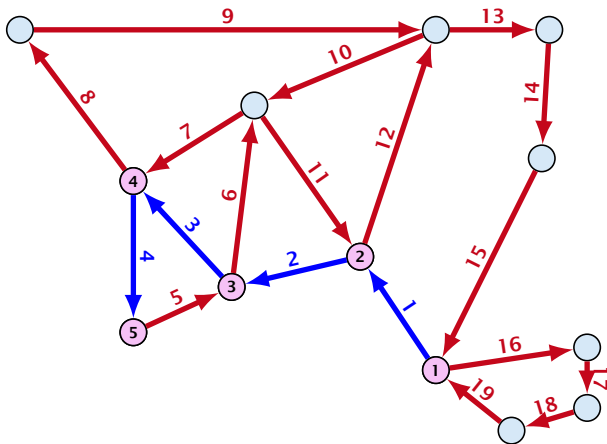
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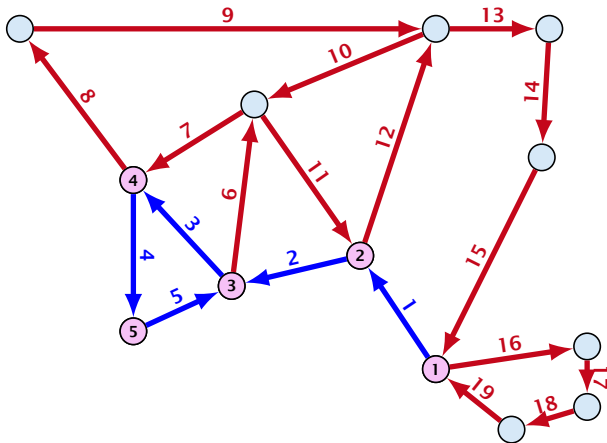
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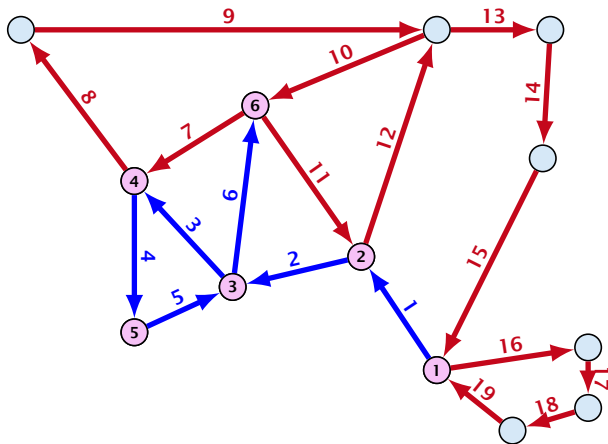
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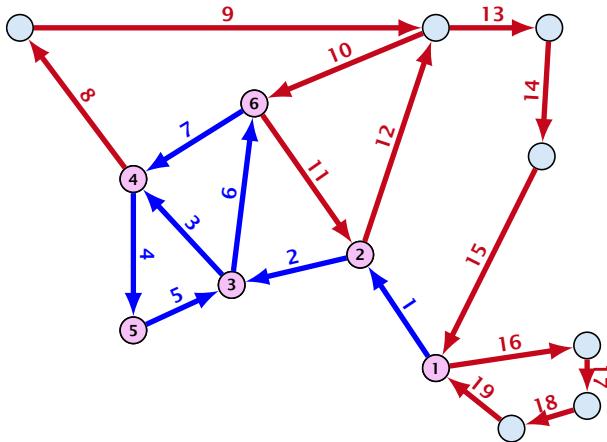
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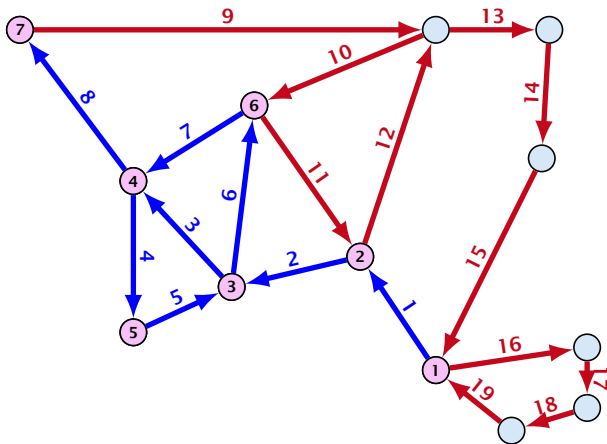
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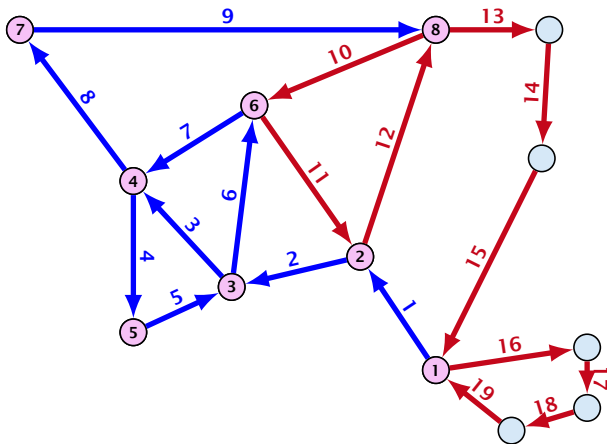
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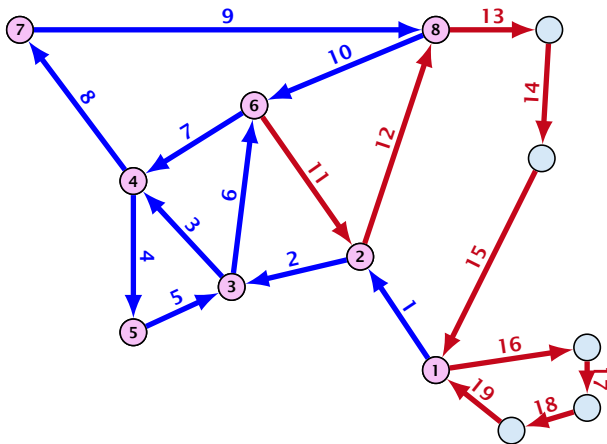


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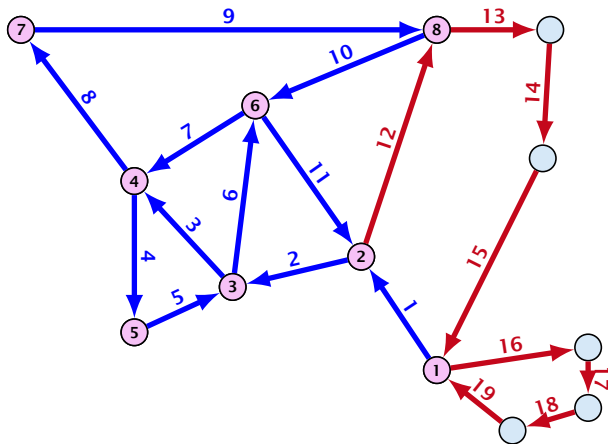




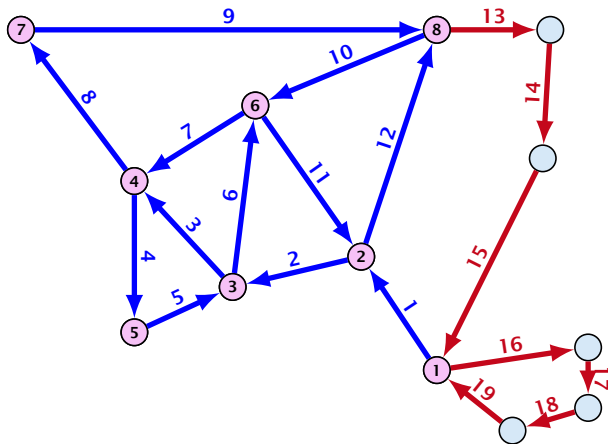
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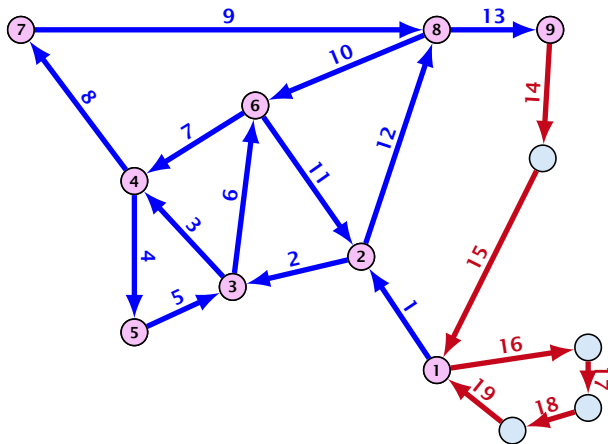
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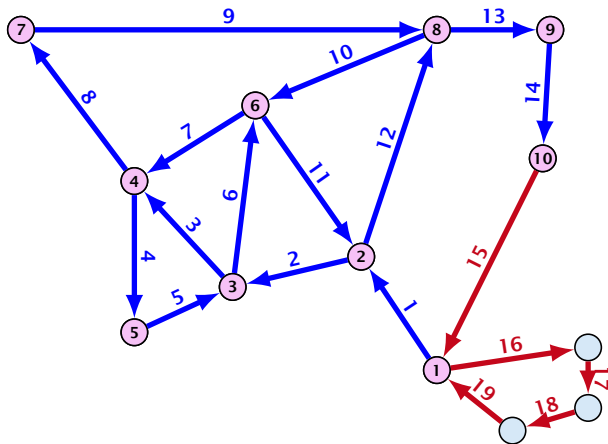
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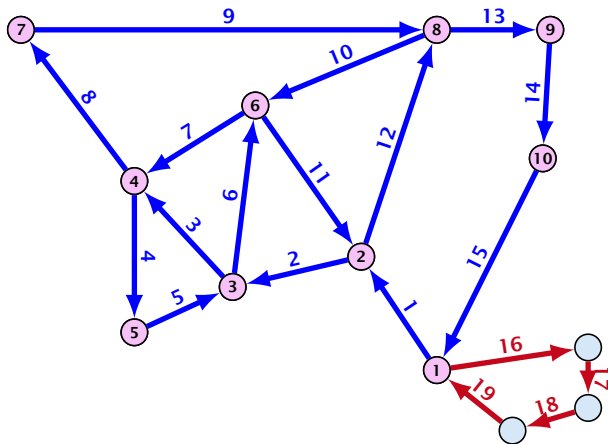
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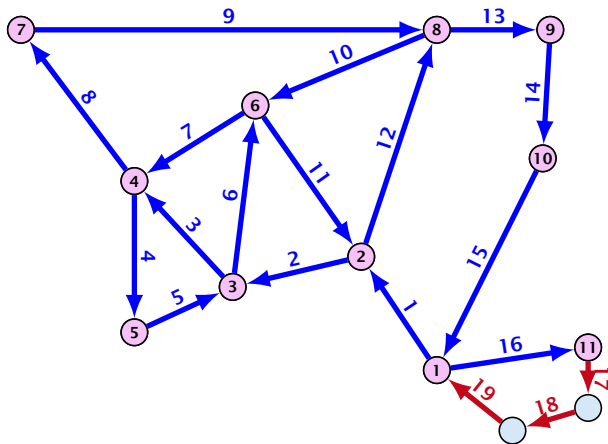
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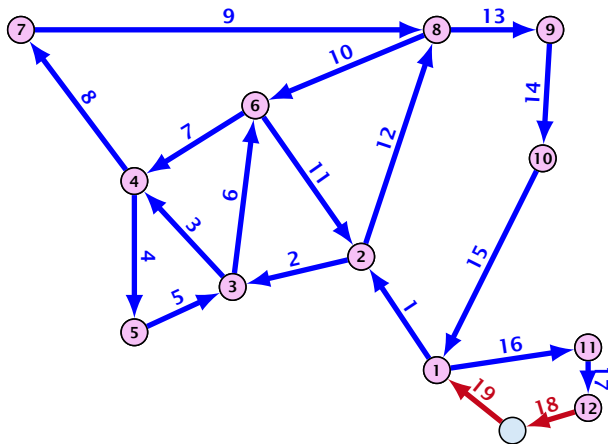
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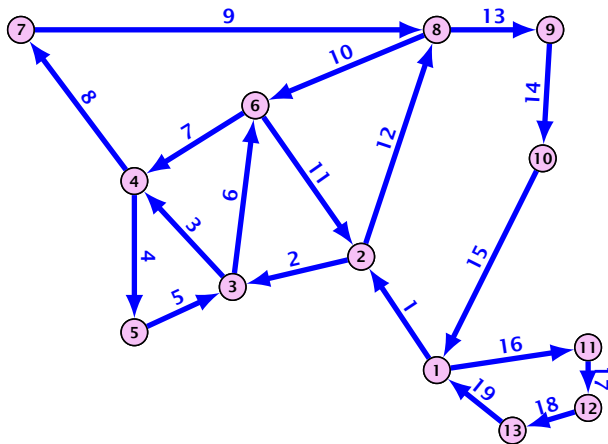


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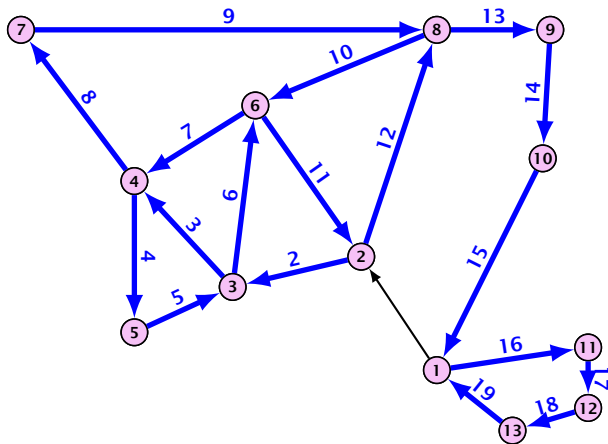




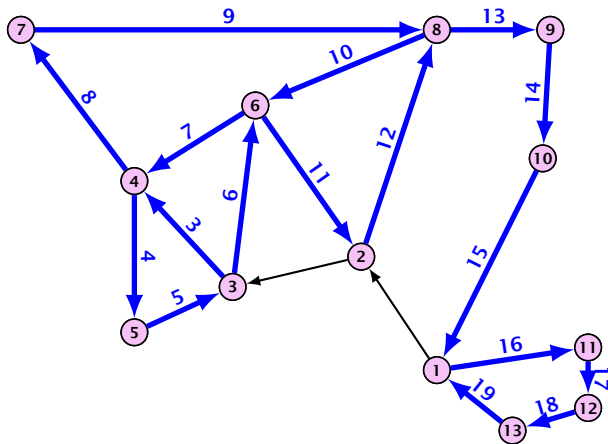
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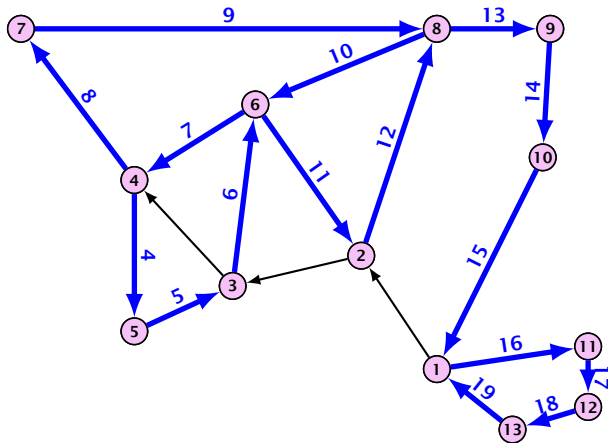
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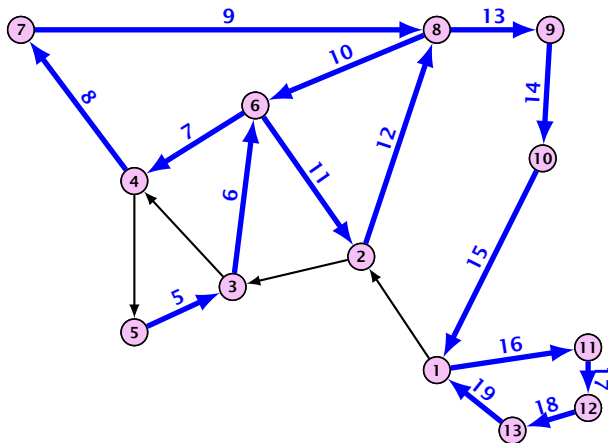
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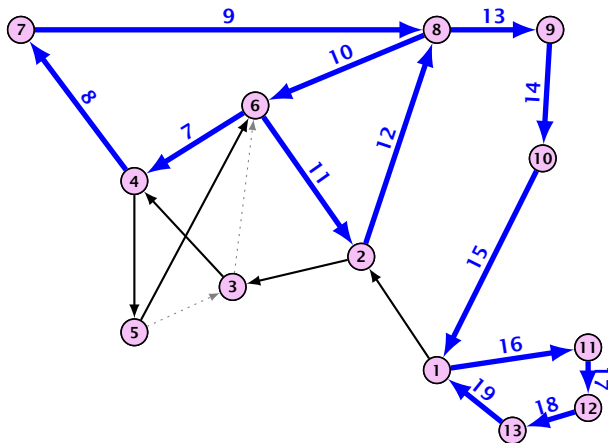
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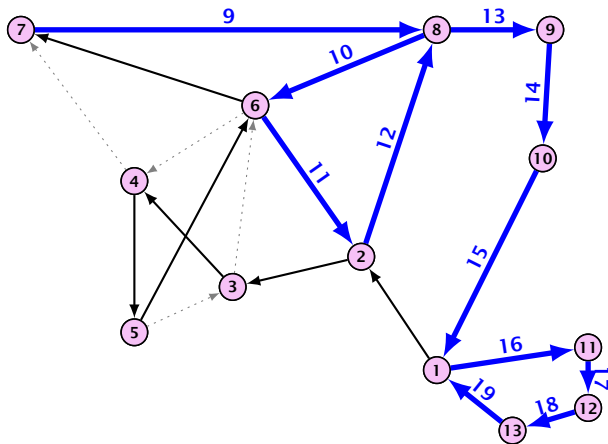
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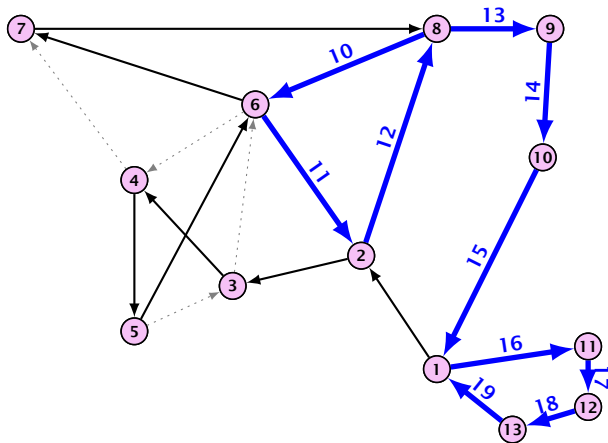
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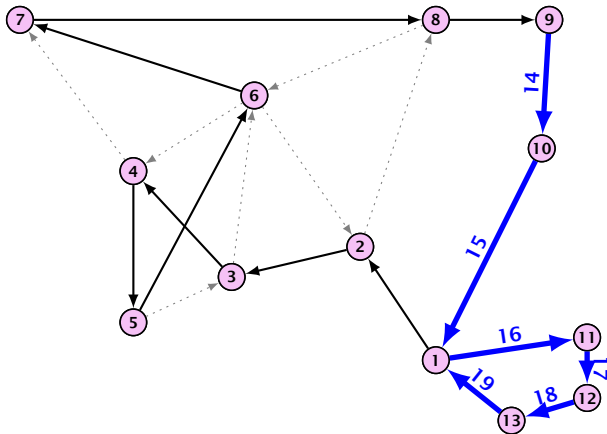


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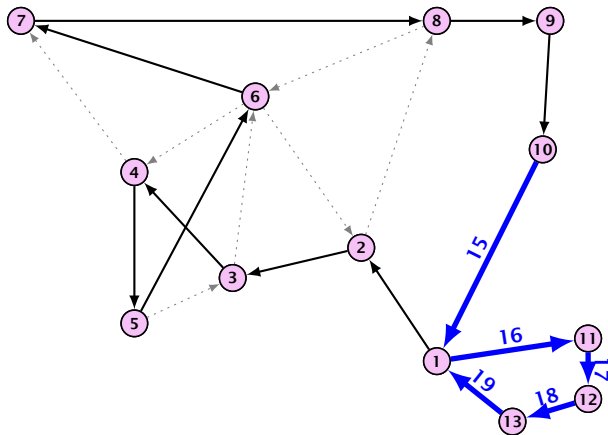




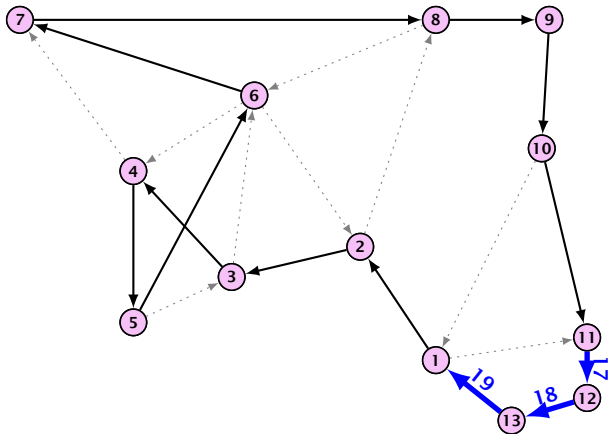
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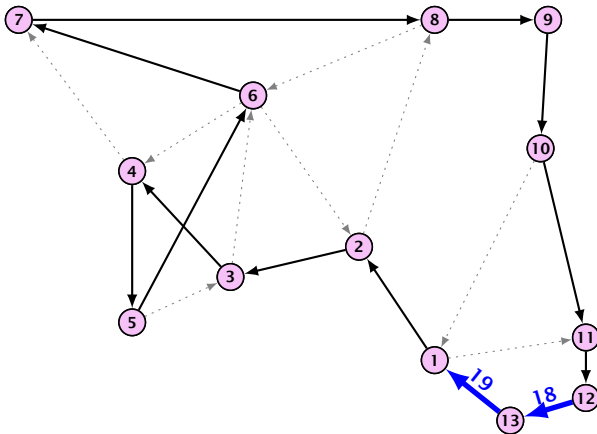
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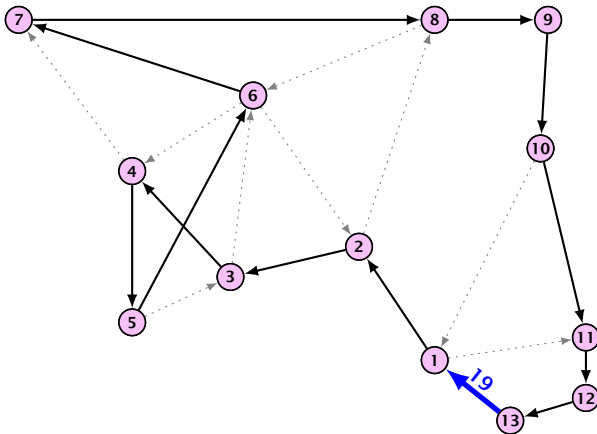
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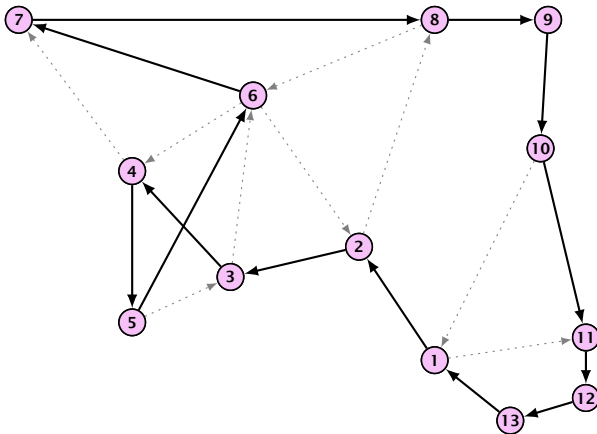
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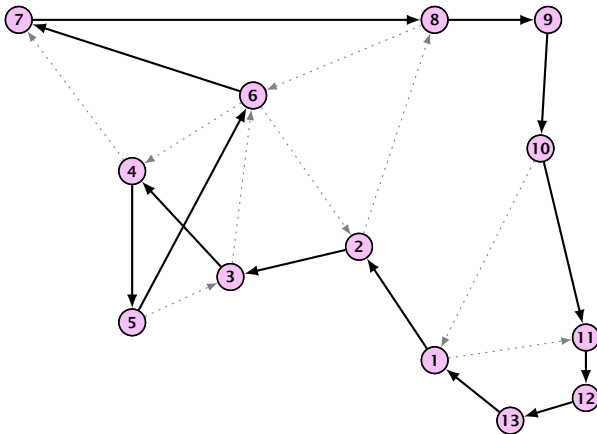
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This graph is Eulerian, and the total cost of all edges is at most  $2 \cdot \text{OPT}_{\text{MST}}(G)$ .

Hence, short-cutting gives a tour of cost no more than  $2 \cdot \text{OPT}_{\text{MST}}(G)$  which means we have a 2-approximation.



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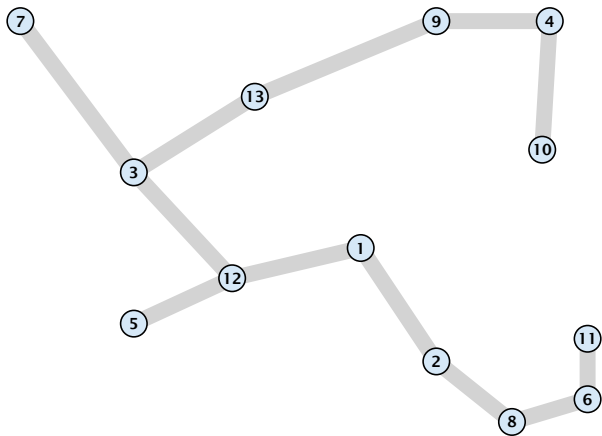
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An optimal tour on the odd-degree vertices has cost at most  $\text{OPT}_{\text{TSP}}(G)$ .

However, the edges of this tour give rise to two disjoint matchings. One of these matchings must have weight less than  $\text{OPT}_{\text{TSP}}(G)/2$ .

Adding this matching to the MST gives an Eulerian graph with edge weight at most

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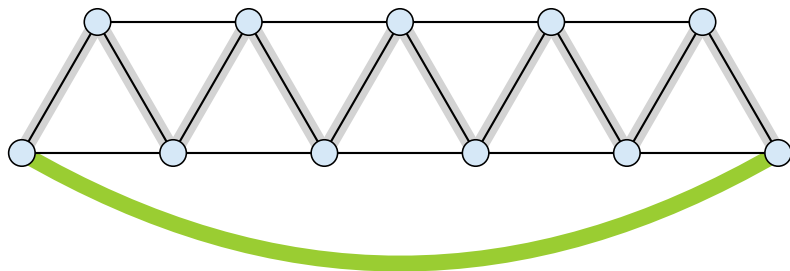
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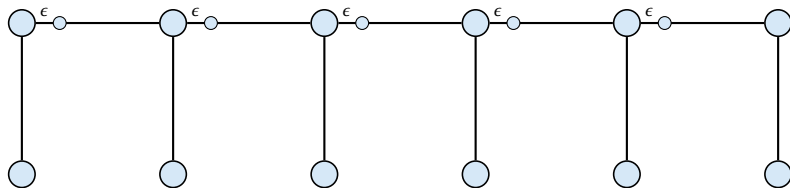
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## Christofides. Tight Example



- ▶ optimal tour:  $n$  edges.
- ▶ MST:  $n - 1$  edges.
- ▶ weight of matching  $(n + 1)/2 - 1$
- ▶ MST+matching  $\approx 3/2 \cdot n$

## Tree shortcutting. Tight Example



- ▶ edges have Euclidean distance.