V. Approx. Algorithms: Travelling Salesman Problem

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UNIVERSITY OF CAMBRIDGE

Outline

Introduction

General TSP

Metric TSP

The Traveling Salesman Problem (TSP)

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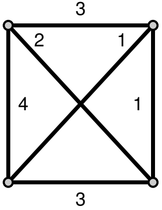

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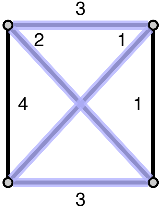

$3 + 2 + 1 + 3 = 9$

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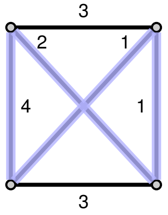

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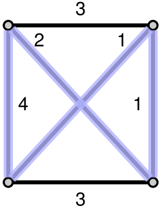

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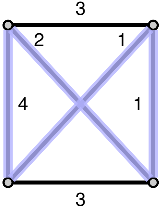

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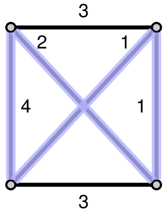

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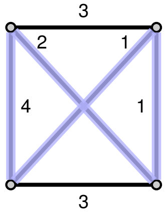

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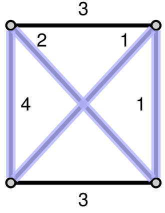

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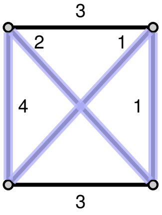

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Even this version is NP hard (Ex. 35.2-2)

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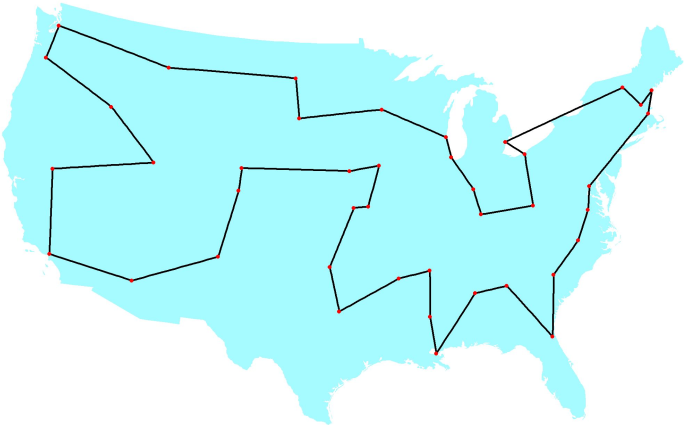
History of the TSP problem (1954)

Dantzig, Fulkerson and Johnson found an optimal tour through 42 cities.



`http://www.math.uwaterloo.ca/tsp/history/img/dantzig_big.html`

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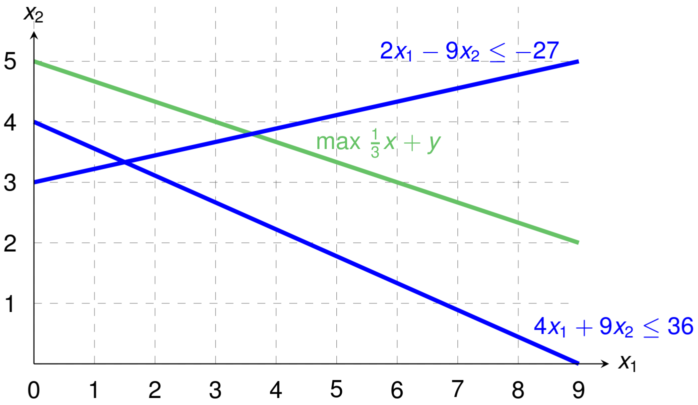

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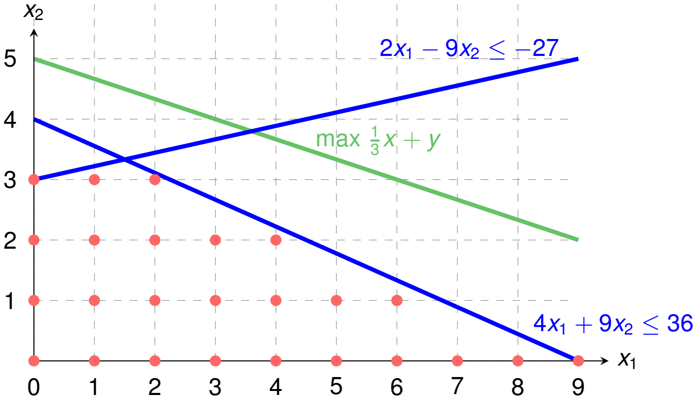

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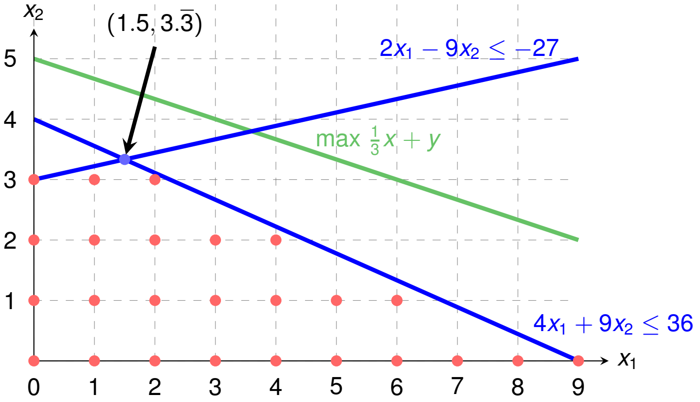

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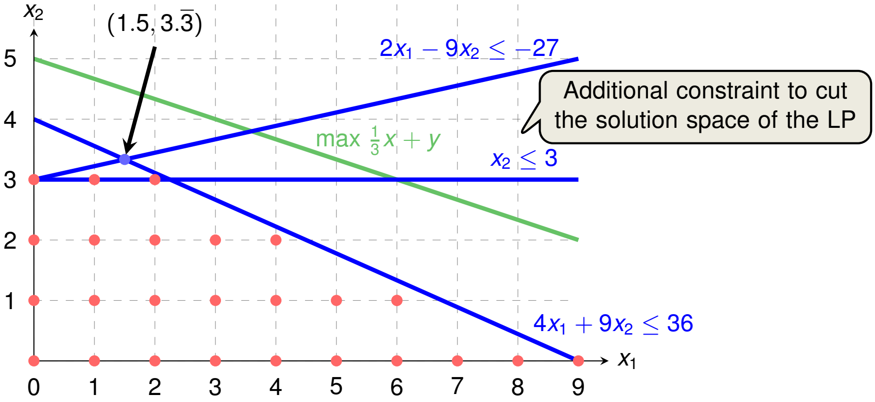

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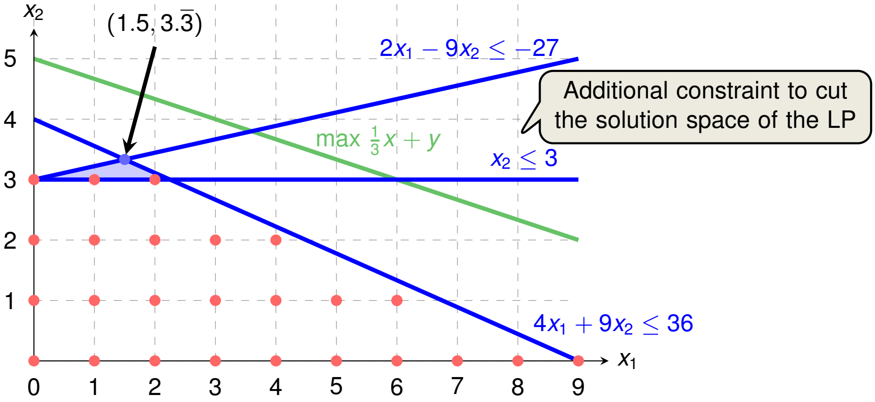

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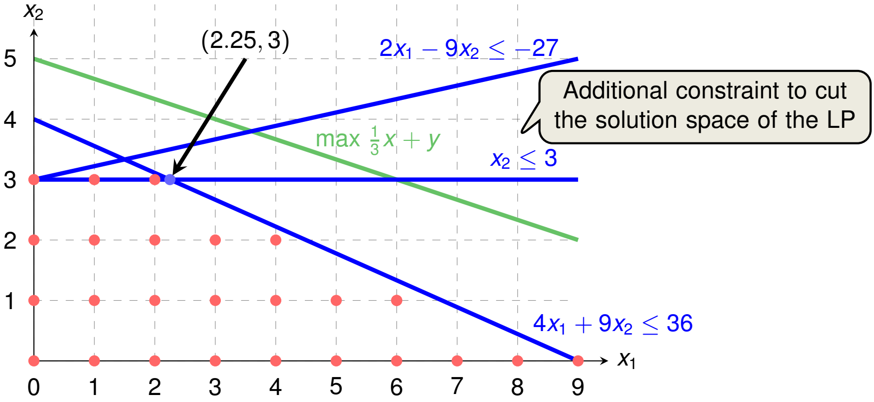

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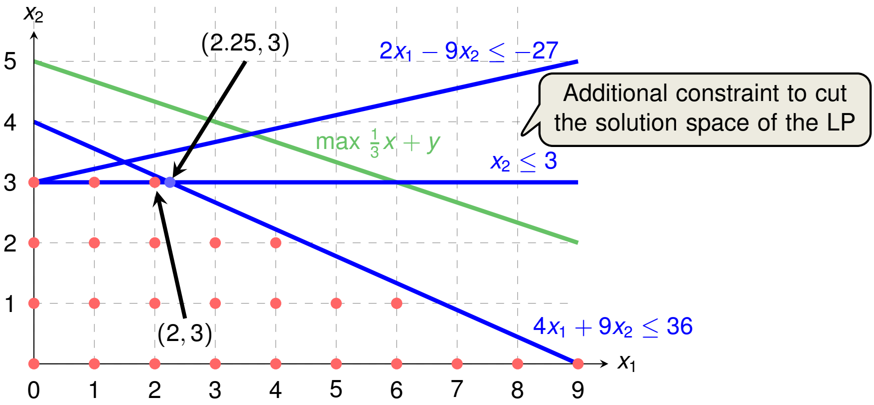

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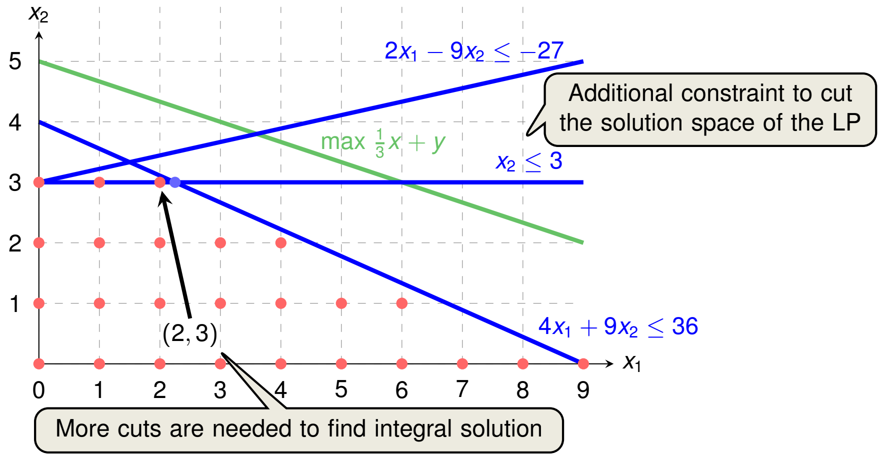

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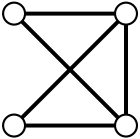

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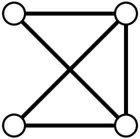

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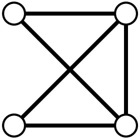

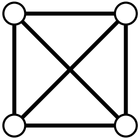

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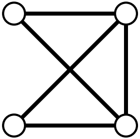
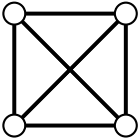
$G = (V, E)$   $G' = (V, E')$

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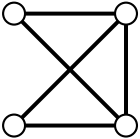

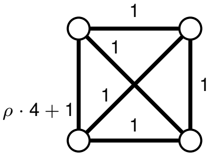

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Large weight will render this edge useless!

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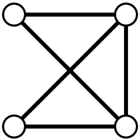

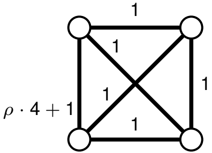

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Can create representations of G' and c in time polynomial in $|V|$ and $|E|$!

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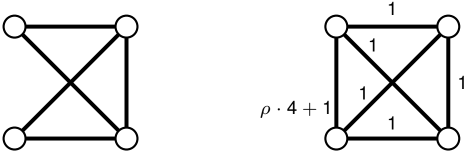

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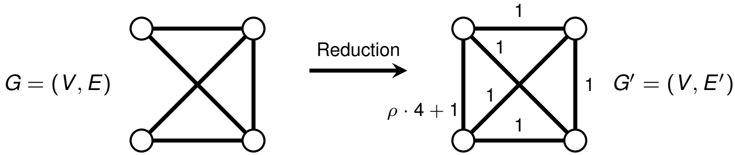

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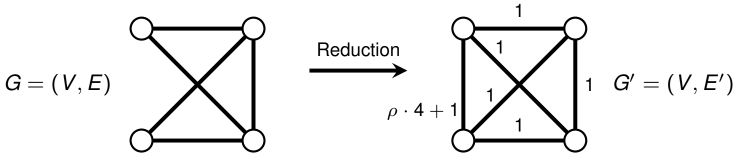

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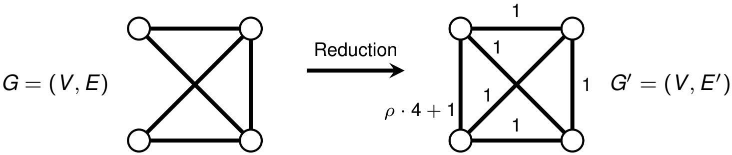

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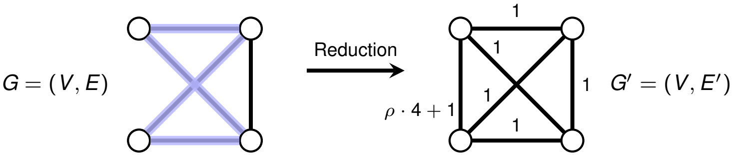

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If P $\neq$ NP, then for any constant $\rho \geq 1$, there is no polynomial-time approximation algorithm with approximation ratio ρ for the general TSP.

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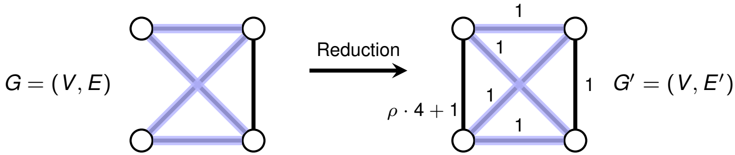

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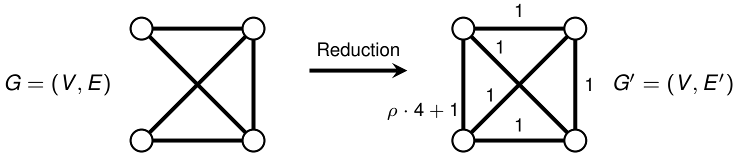

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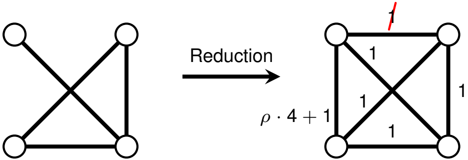

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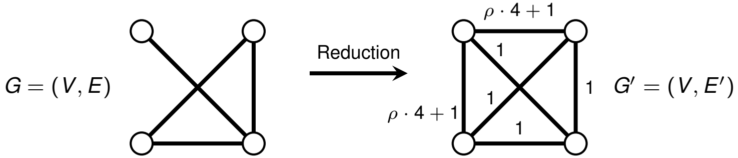

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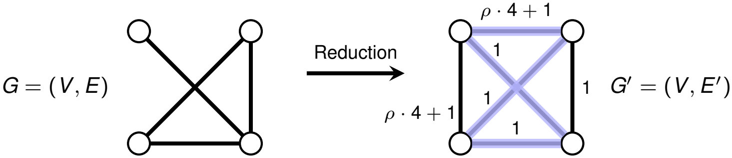

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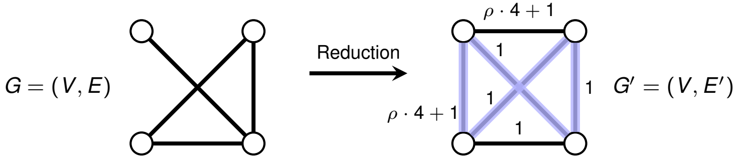

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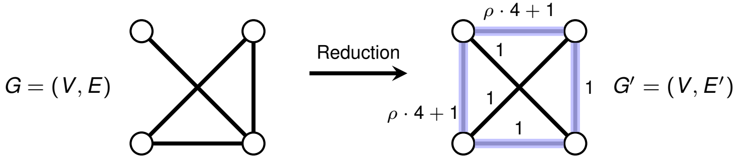

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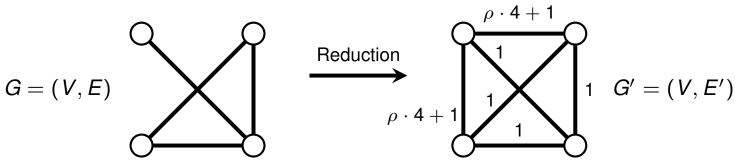

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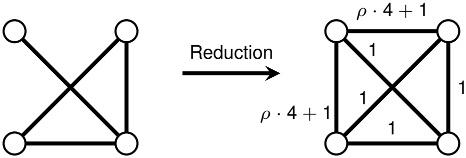

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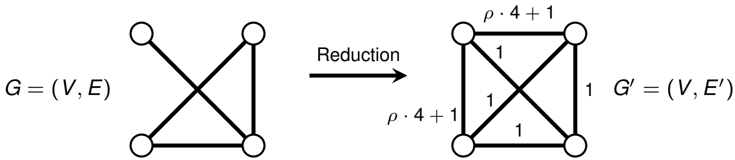

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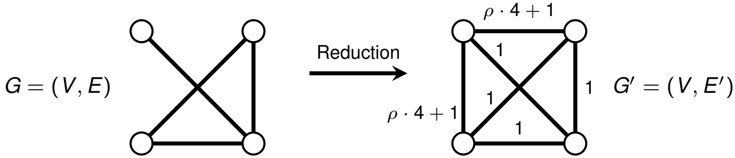

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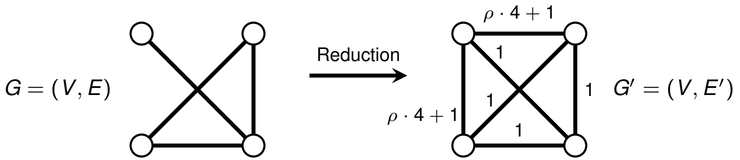
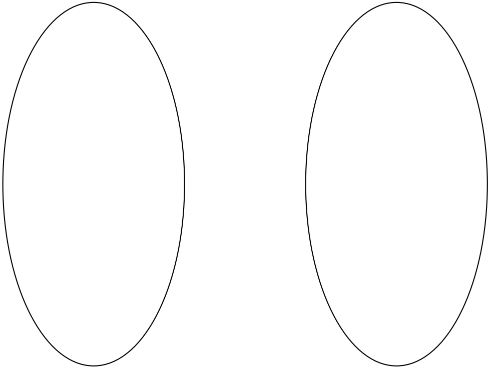


Proof of Theorem 35.3 from a higher perspective



instances of Hamilton instances of TSP

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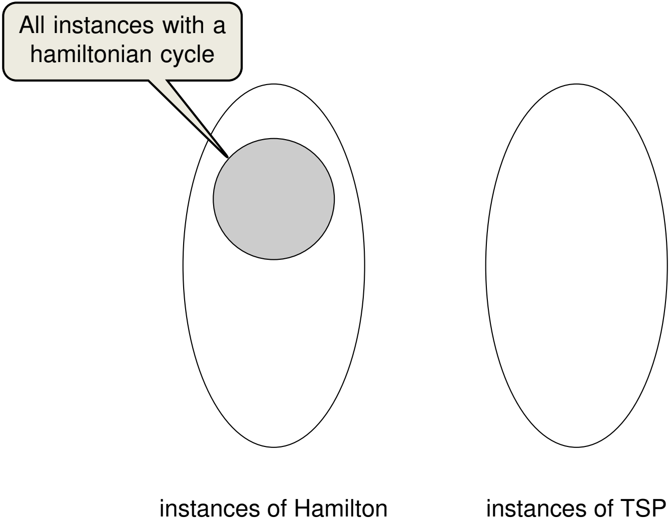

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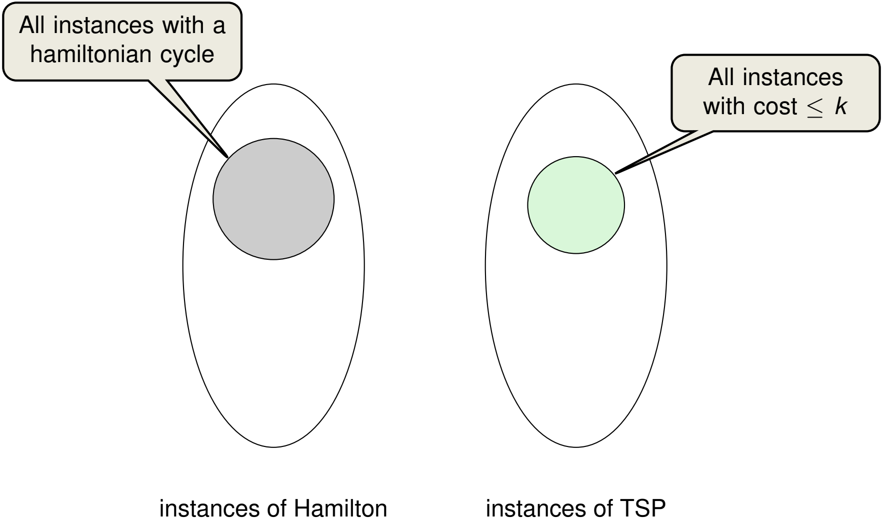

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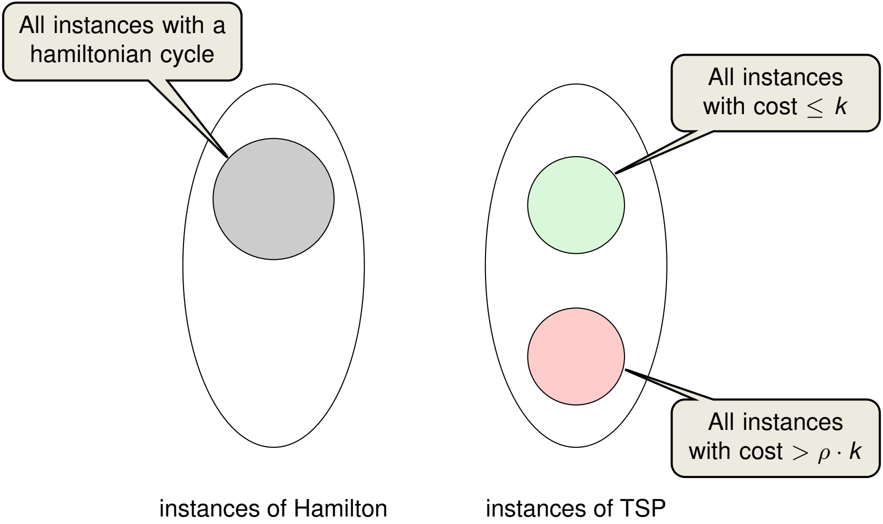

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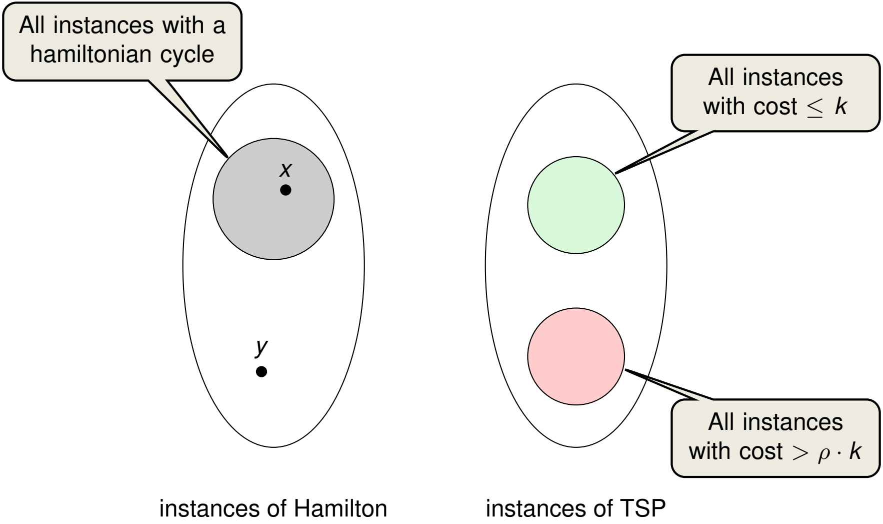

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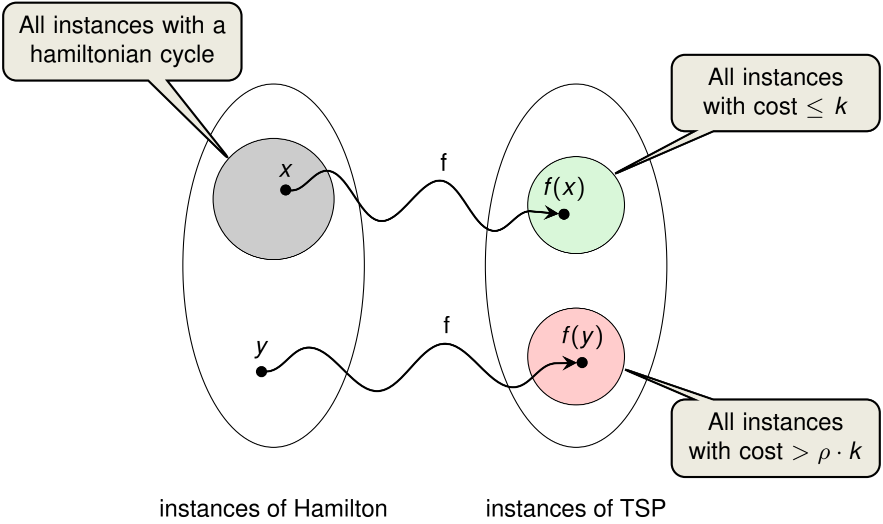

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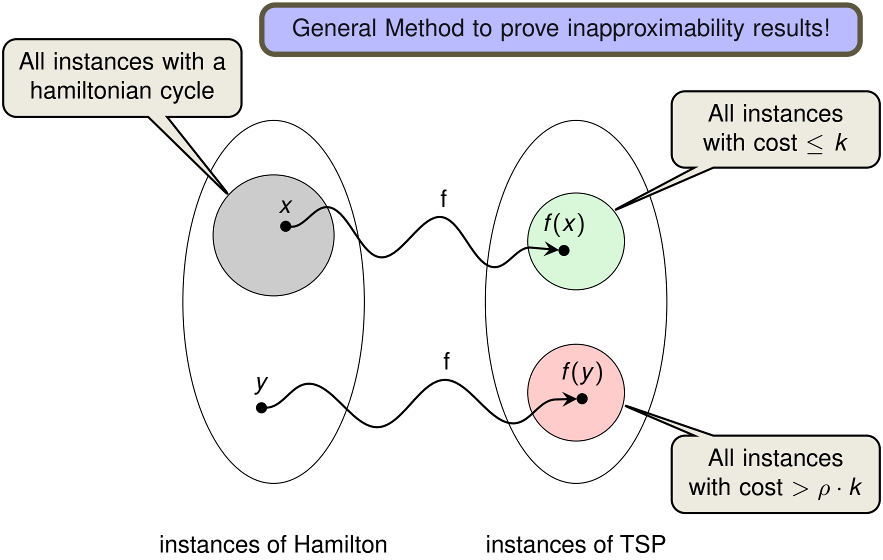

Outline

Introduction

General TSP

Metric TSP

Metric TSP (TSP Problem with the Triangle Inequality)

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APPROX-TSP-TOUR(G, c)
select a vertex $r \in G.V$ to be a "root" vertex
compute a minimum spanning tree T_{min} for G from root r
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let H be a list of vertices, ordered according to when they are first visited
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return the hamiltonian cycle H

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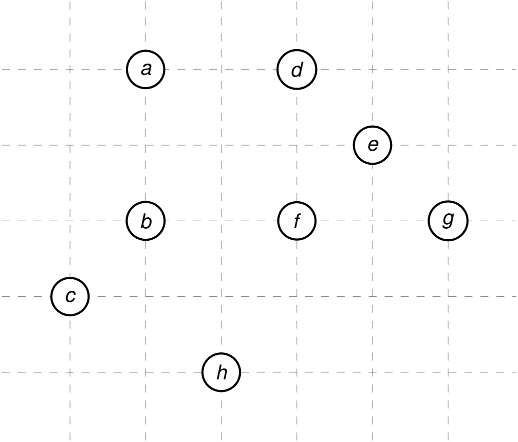
Remember: In the Metric-TSP problem, G is a complete graph.

Run of APPROX-TSP-TOUR



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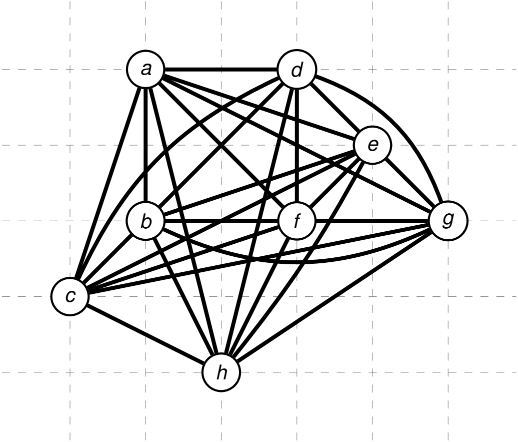
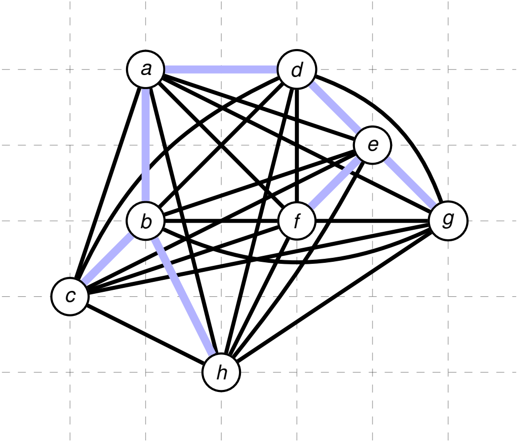


1. Compute MST $T_{\min}$

Run of APPROX-TSP-TOUR



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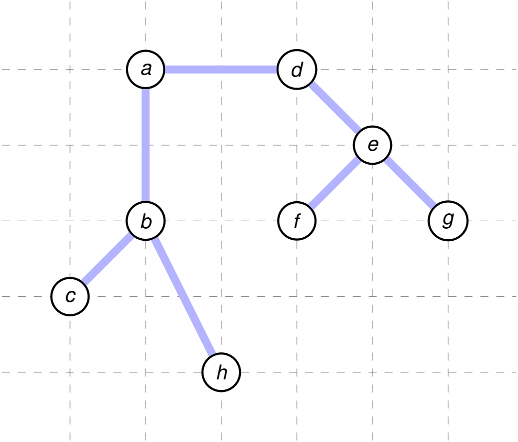

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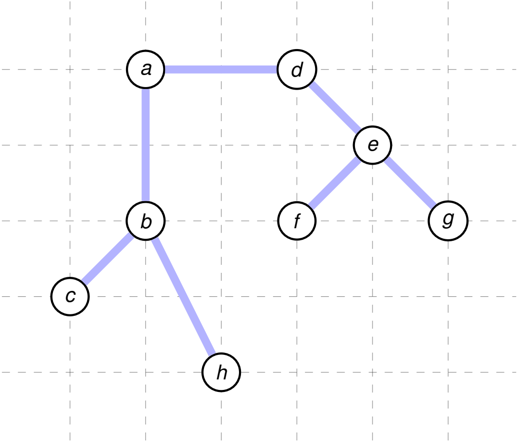

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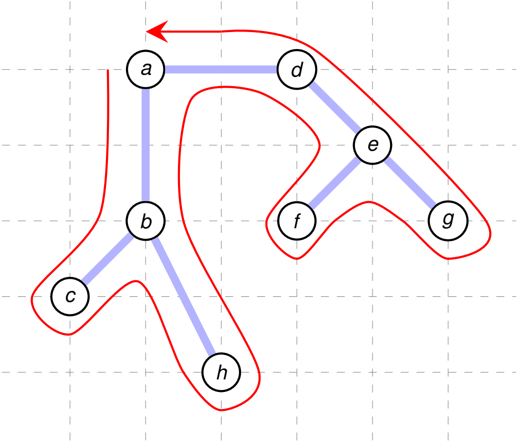

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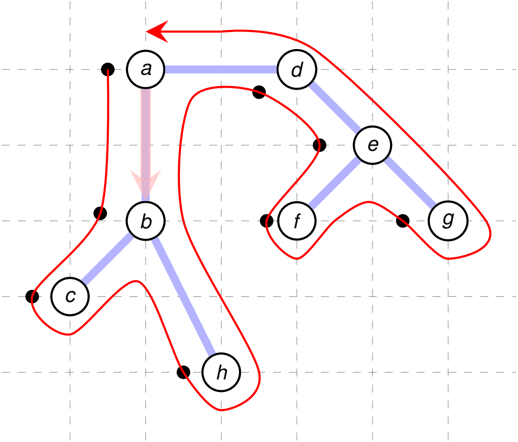

1. Compute MST $T_{\min}$ ✓
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3. Return list of vertices according to the preorder tree walk

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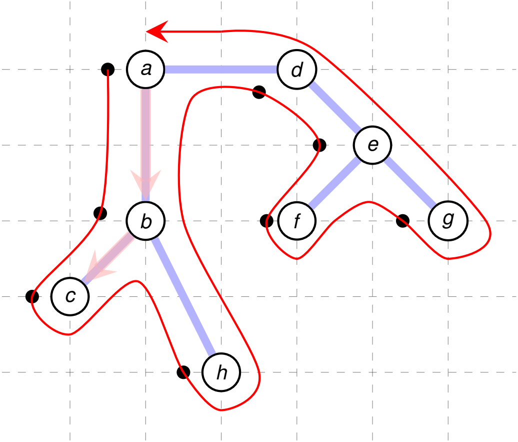

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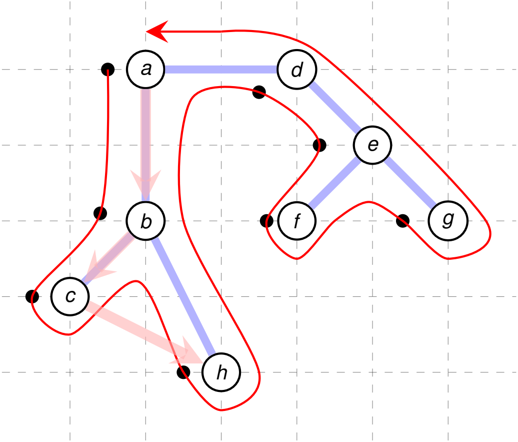

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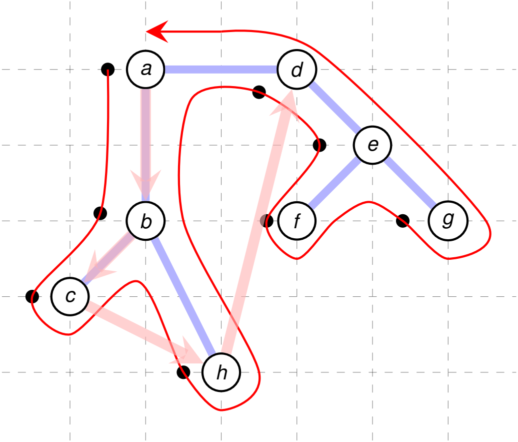

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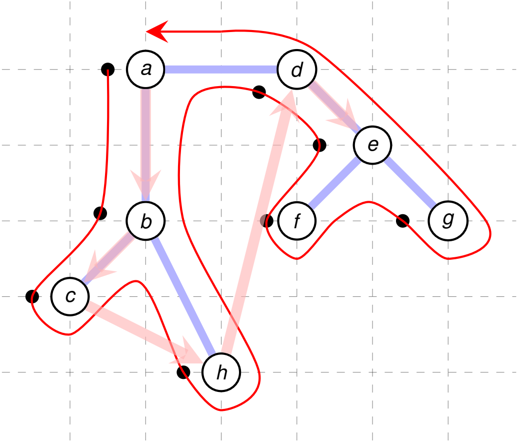

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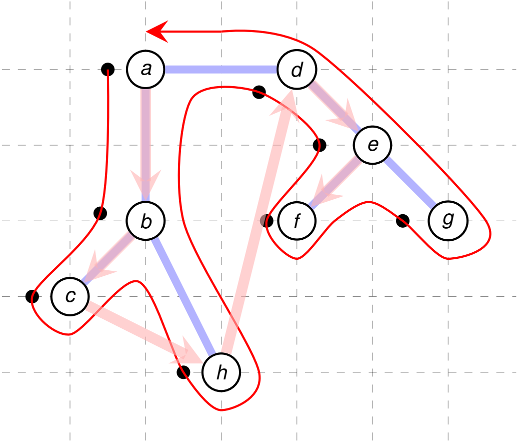

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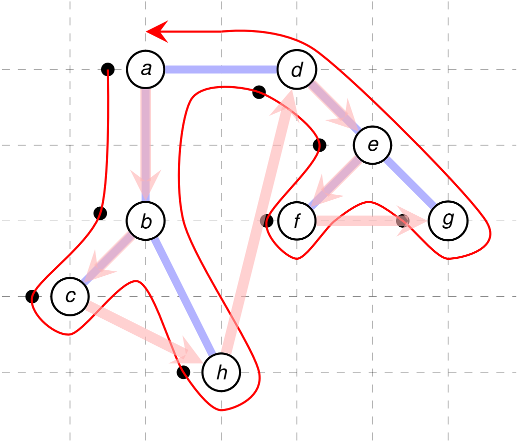

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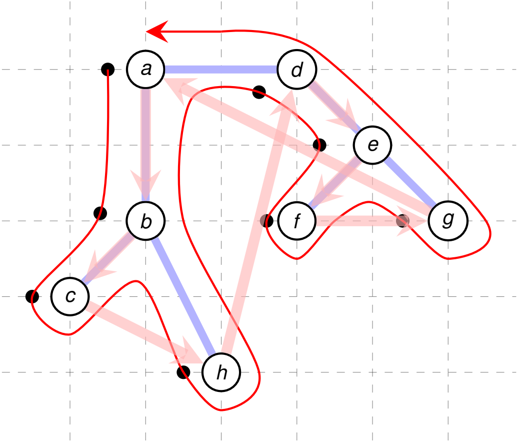

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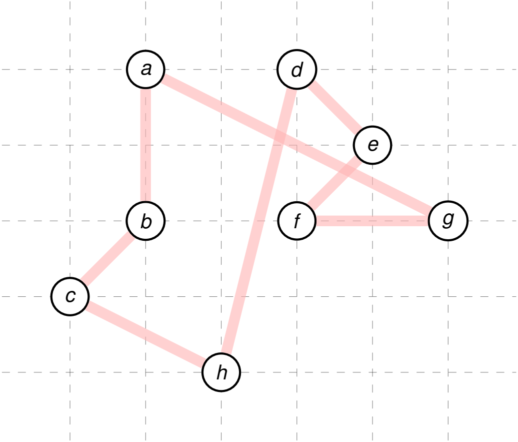
Run of Approx-Tsp-Tour



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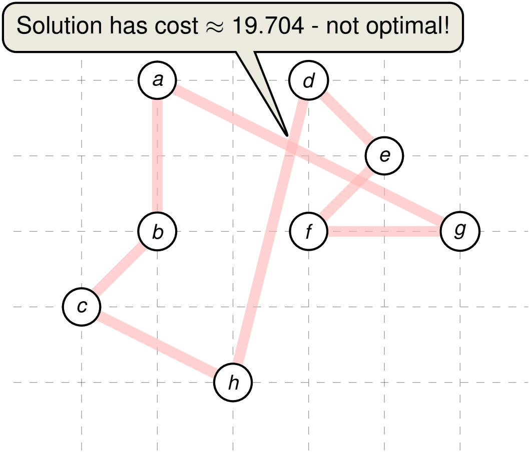
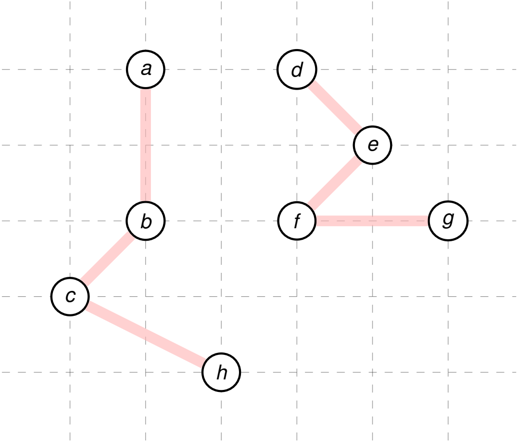
Run of APPROX-TSP-TOUR



1. Compute MST T_{min} ✓
2. Perform preorder walk on MST T_{min} ✓
3. Return list of vertices according to the preorder tree walk ✓

Run of APPROX-TSP-TOUR



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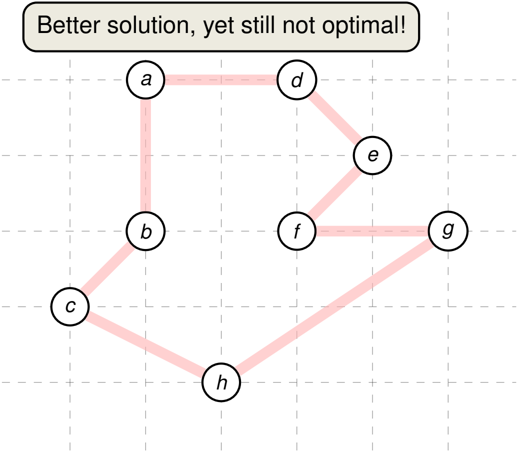
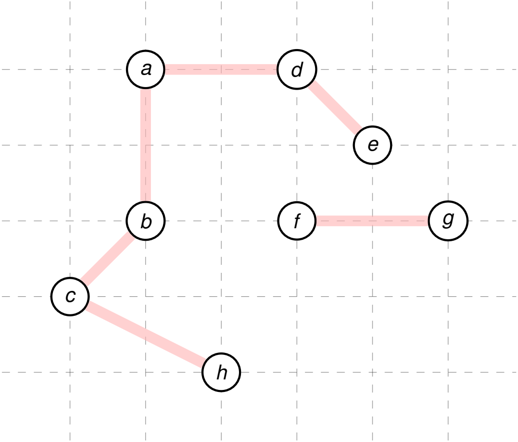
Run of APPROX-TSP-TOUR



1. Compute MST $T_{\min}$ ✓
2. Perform preorder walk on MST $T_{\min}$ ✓
3. Return list of vertices according to the preorder tree walk ✓

Run of APPROX-TSP-TOUR



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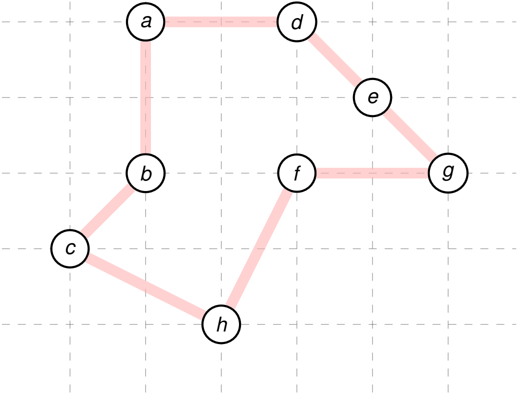
Run of Approx-Tsp-Tour

This is the optimal solution (cost $\approx$ 14.715).



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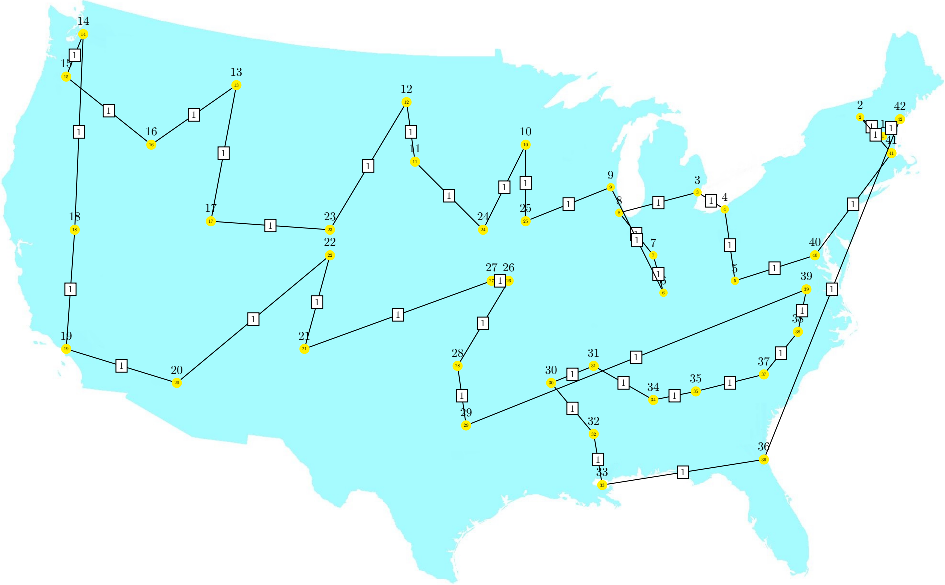
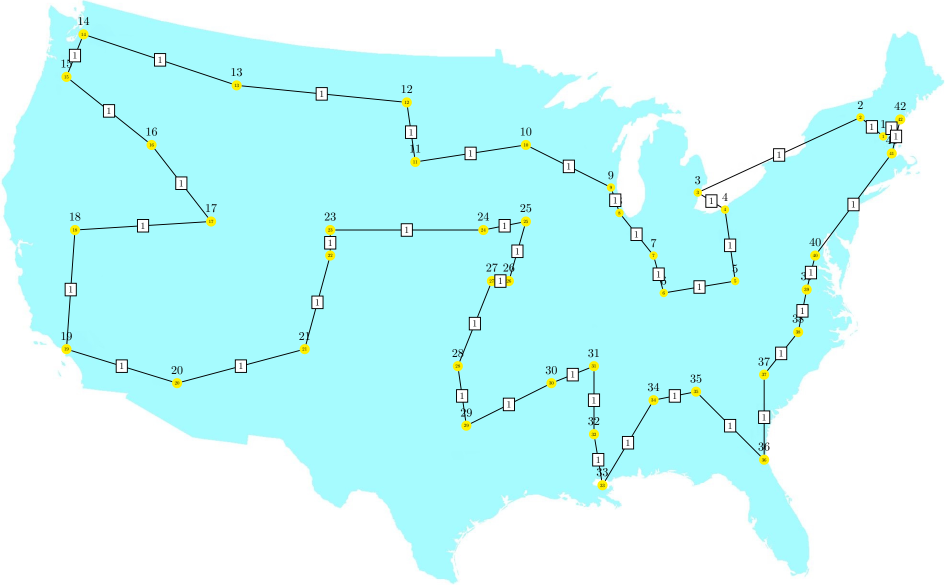
2. Perform preorder walk on MST $T_{\min}$ ✓
3. Return list of vertices according to the preorder tree walk ✓

Approximate Solution: Objective 921



Optimal Solution: Objective 699



Proof of the Approximation Ratio

Theorem 35.2

APPROX-TSP-TOUR is a polynomial-time 2-approximation for the traveling-salesman problem with the triangle inequality.

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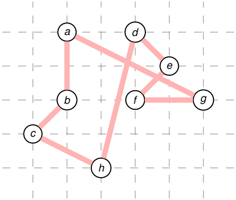

solution H of APPROX-TSP

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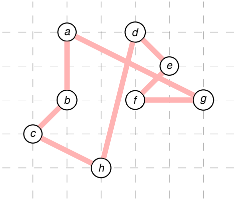

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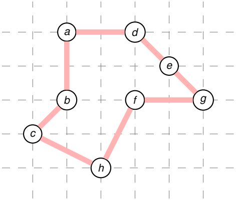

optimal solution H^*

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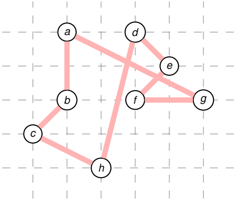

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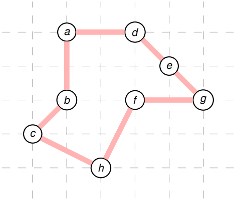

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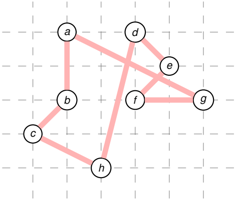

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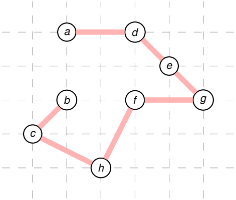

spanning tree T as a subset of H^*

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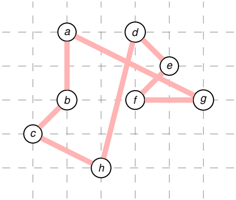

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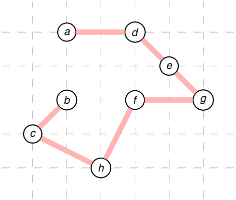

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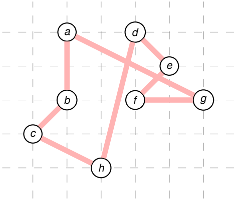

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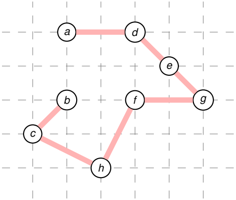

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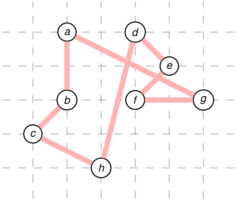
exploiting that all edge costs are non-negative!



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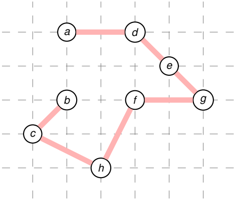

spanning tree T as a subset of H^*

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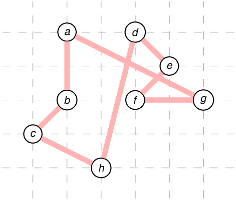

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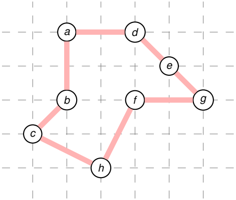

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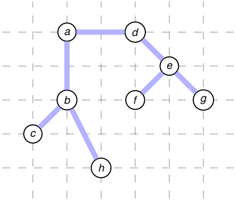
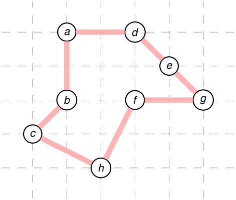


minimum spanning tree $T_{\min}$



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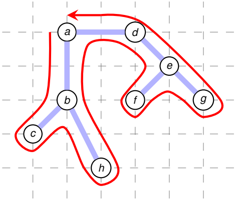
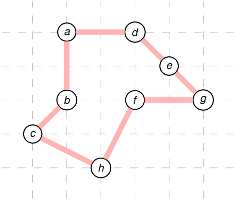


Walk $W = (a, b, c, b, h, b, a, d, e, f, e, g, e, d, a)$



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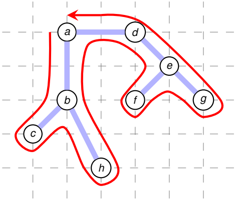
$\Rightarrow$ Full walk traverses every edge exactly twice, so



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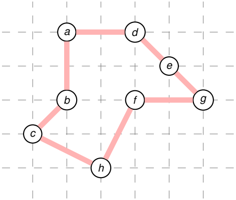

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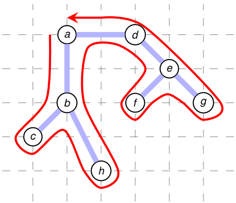

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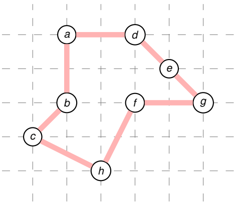

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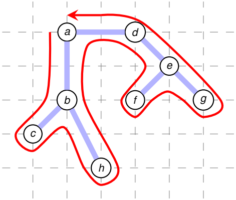

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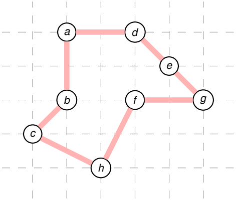

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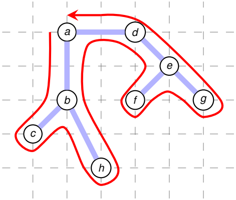
- Deleting duplicate vertices from W yields a tour H



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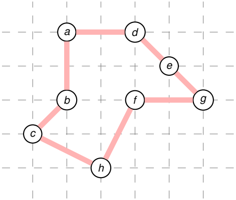

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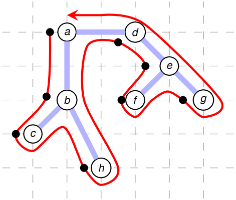

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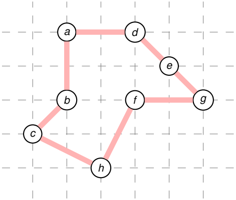

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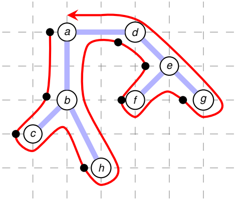

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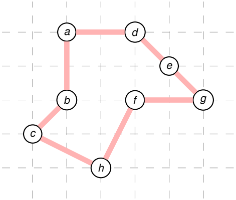

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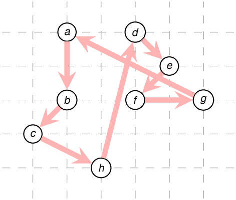
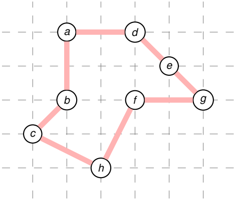


Tour $H = (a, b, c, h, d, e, f, g, a)$



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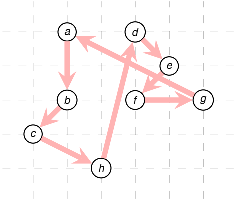
exploiting triangle inequality!

- Deleting duplicate vertices from W yields a tour H with smaller cost:



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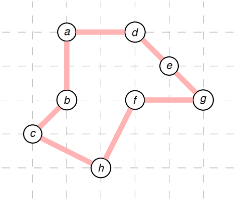

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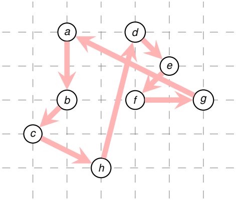

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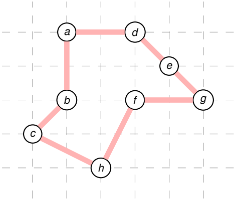

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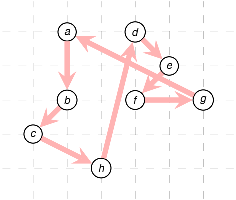
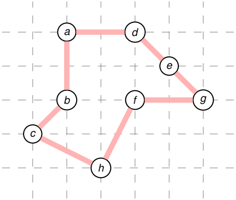


Tour $H = (a, b, c, h, d, e, f, g, a)$



optimal solution H^*

Proof of the Approximation Ratio

Theorem 35.2

APPROX-TSP-TOUR is a polynomial-time 2-approximation for the traveling-salesman problem with the triangle inequality.

Proof:

- Consider the optimal tour H^* and remove an arbitrary edge

$\Rightarrow$ yields a spanning tree T and $c(T) \leq c(H^*)$

- Let W be the full walk of the minimum spanning tree $T_{\min}$ (including repeated visits)

$\Rightarrow$ Full walk traverses every edge exactly twice, so

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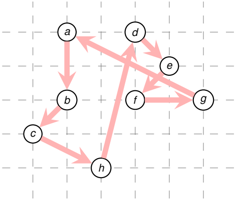

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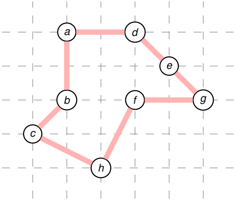

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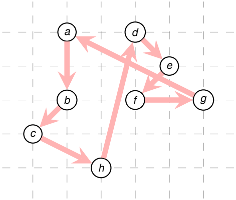

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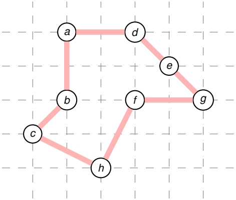

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Can we get a better approximation ratio?

Christofides Algorithm

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Can we get a better approximation ratio?

```
CHRISTOFIDES(G, c)
select a vertex r ∈ G.V to be a "root" vertex
compute a minimum spanning tree T_min for G from root r
        using MST-PRIM(G, c, r)
compute a perfect matching M_min with minimum weight in the complete graph
        over the odd-degree vertices in T_min
let H be a list of vertices, ordered according to when they are first visited
        in a Eulearian circuit of T_min ∪ M_min
return the hamiltonian cycle H
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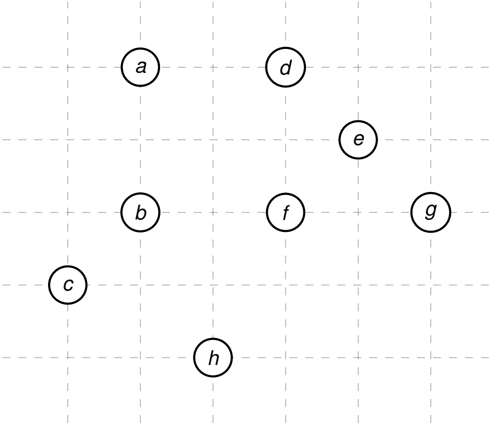
Theorem (Christofides'76)

There is a polynomial-time $\frac{3}{2}$-approximation algorithm for the travelling salesman problem with the triangle inequality.

Run of CHRISTOFIDES



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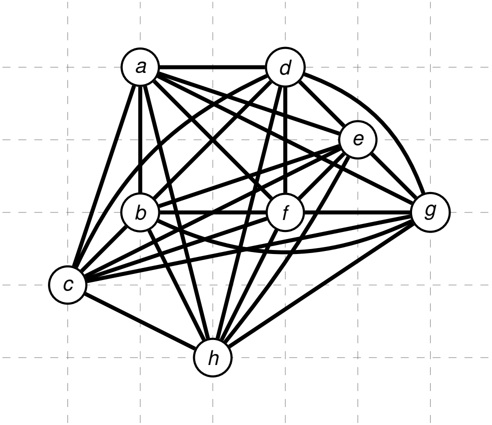

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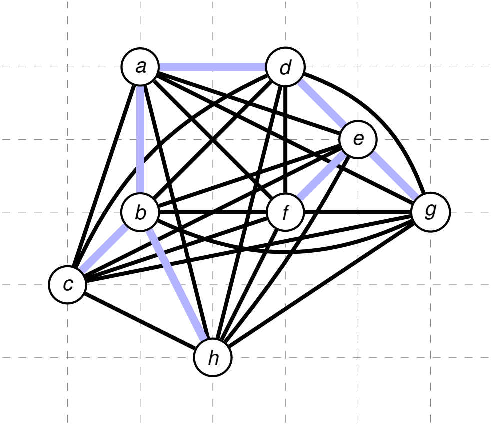
Run of CHRISTOFIDES



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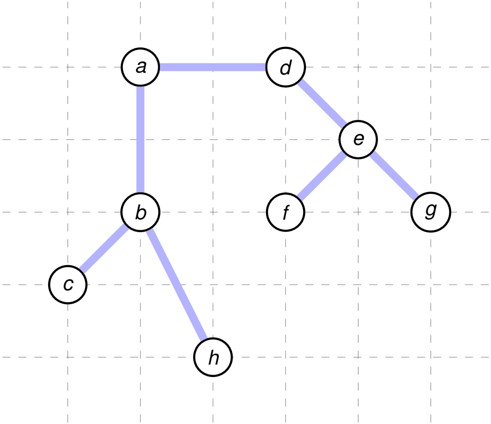

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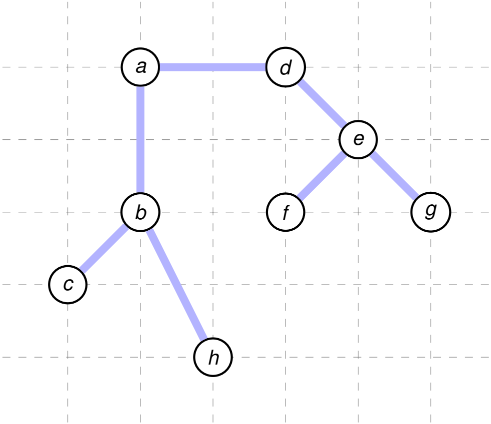

1. Compute MST $T_{\min}$ ✓
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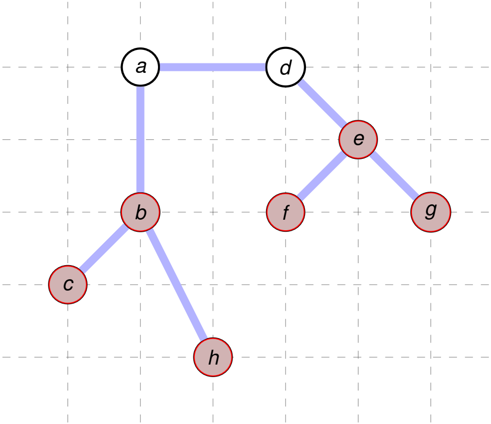

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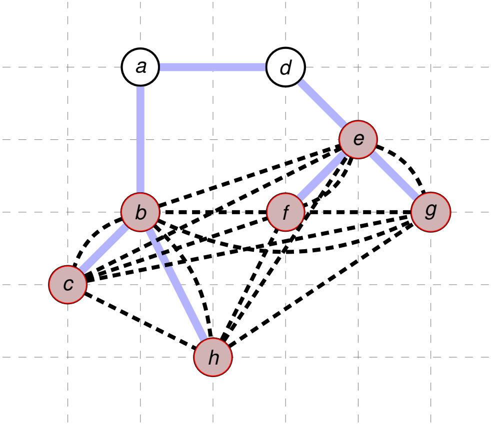

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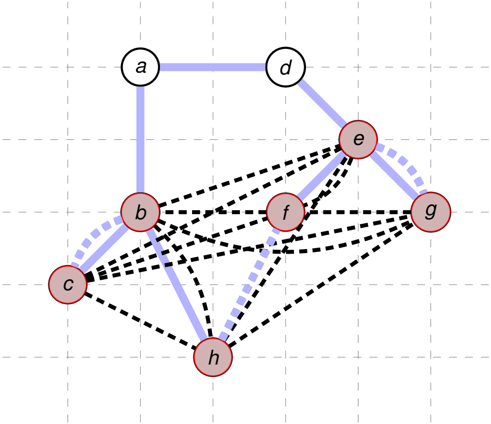

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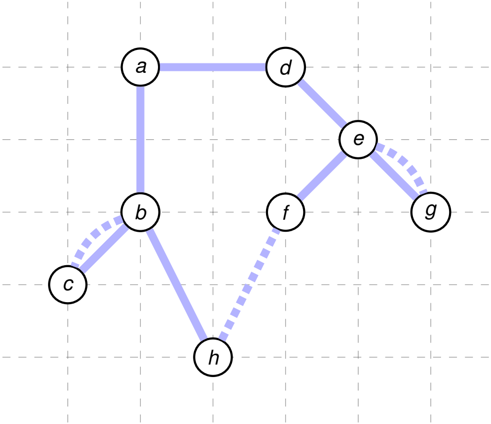

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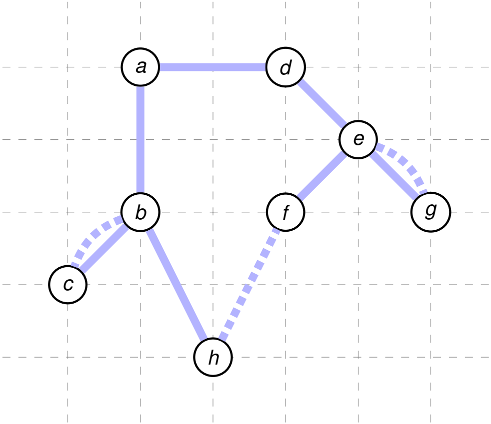

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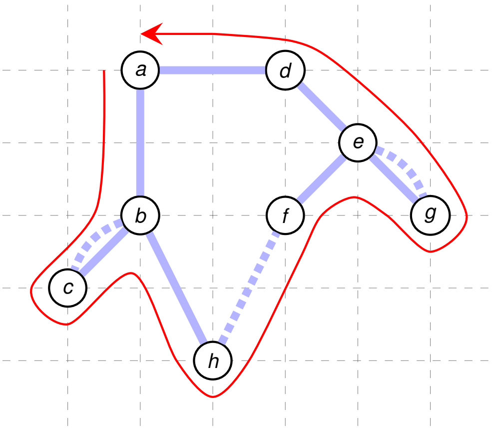

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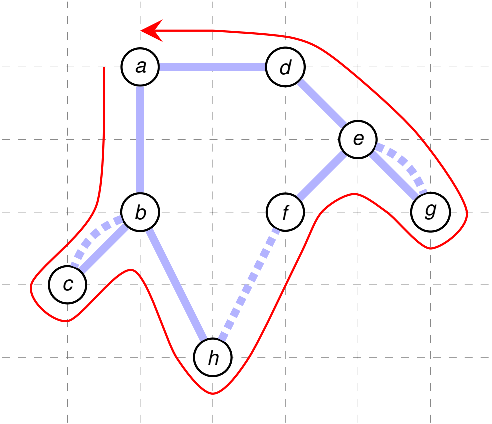

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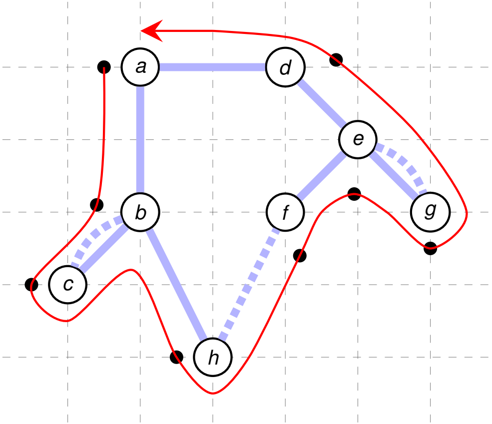

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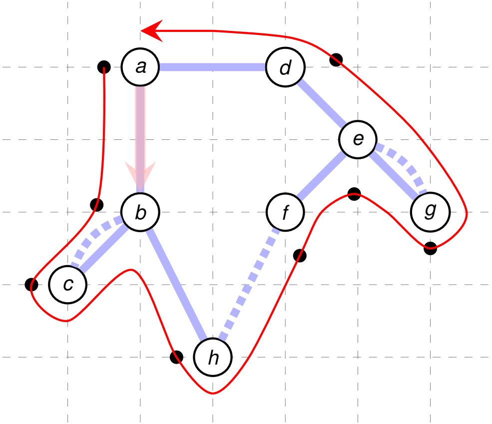

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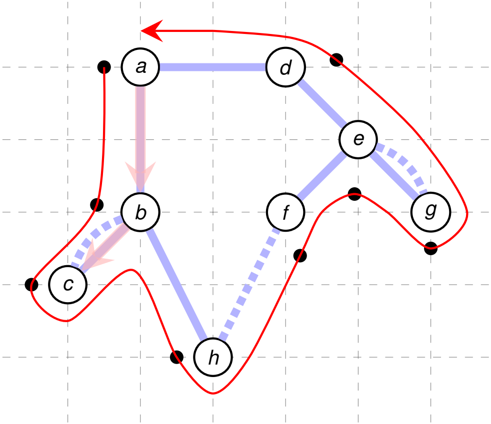

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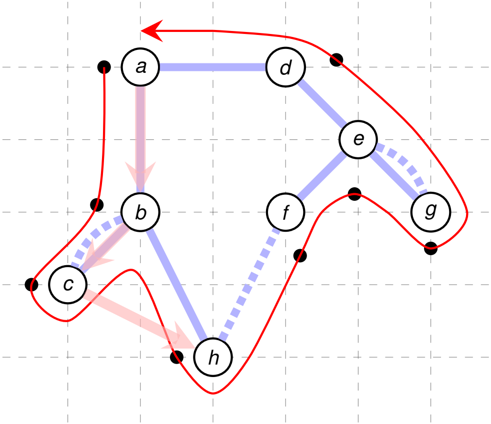

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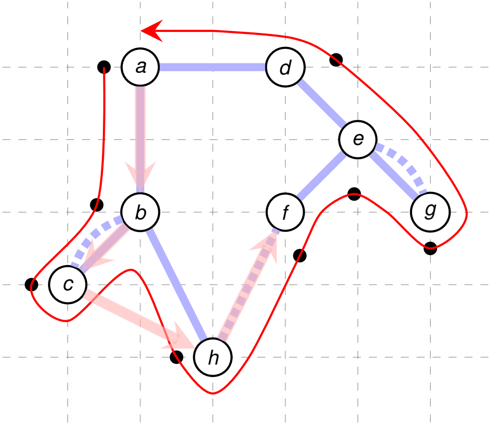

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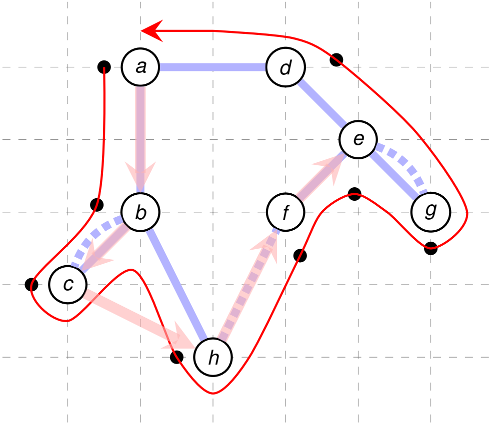

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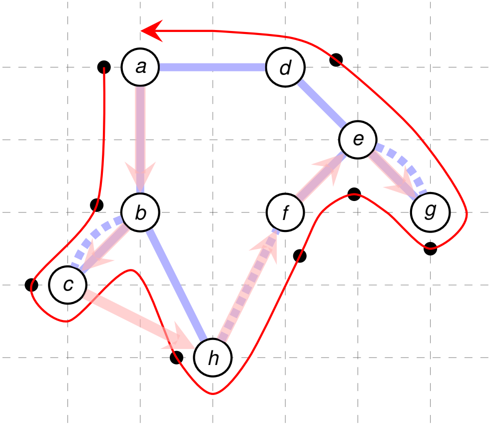

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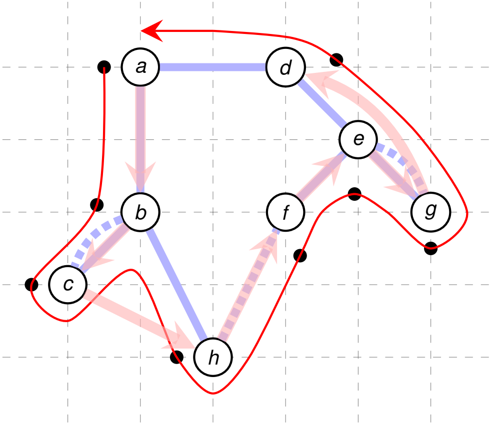

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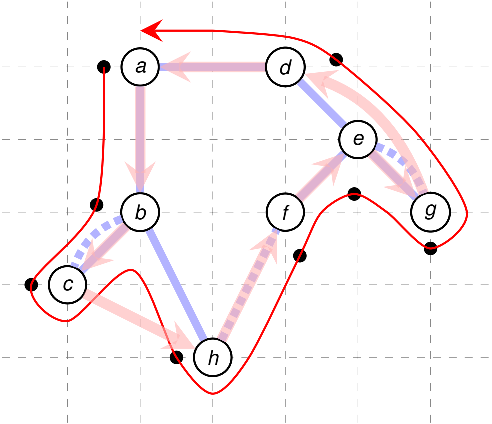

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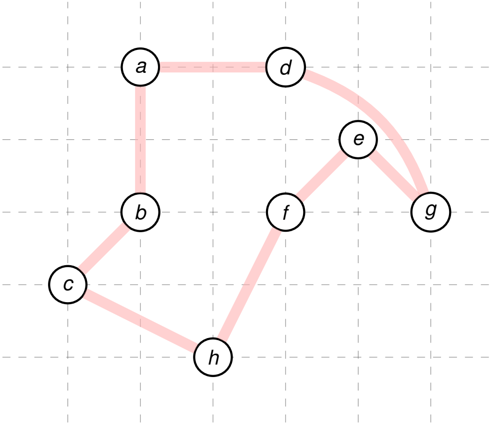

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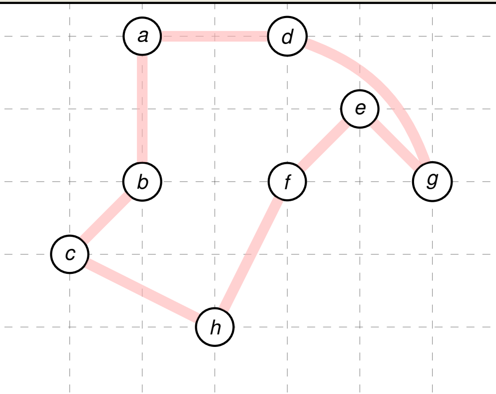
Solution has cost ≈ 15.54 - within 10% of the optimum!



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Theorem (Christofides'76)

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