

# Worst Case and Probabilistic Analysis of the 2-Opt Algorithm for the TSP

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# Traveling Salesperson Problem

## Traveling Salesperson Problem (TSP)



- Input: weighted (complete) graph  $G = (V, E, d)$  with  $d : E \rightarrow \mathbb{R}$ .

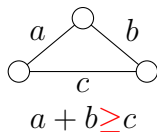
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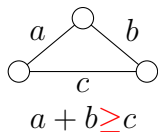


- Input: weighted (complete) graph  $G = (V, E, d)$  with  $d : E \rightarrow \mathbb{R}$ .
- Goal: Find **Hamiltonian cycle of minimum length**.

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  - ▶ Strongly NP-hard.
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  - ▶ **APX-hard**: lower bound of  $220/219$  [Papadimitriou, Vempala (2000)]



$d(P_1, P_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$

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- Euclidean TSP
  - ▶ Cities  $\subset \mathbb{R}^d$
  - ▶ Strongly NP-hard ( $\Rightarrow$  **no FPTAS**) [Papadimitriou (1977)]
  - ▶ **PTAS** exists [Arora (1996), Mitchell (1996)].

- Numerous experimental studies.
  - ▶ TSPLIB contains “real-world” and random (Euclidean) instances.
  - ▶ DIMACS Implementation Challenge [Johnson and McGeoch (2002)].

# Experimental Results

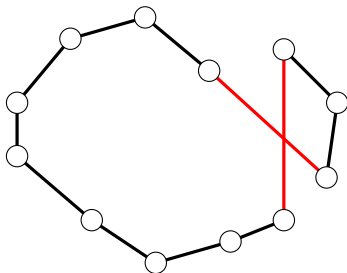
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Some conclusions:

- Worst-case results are often too pessimistic.
- The **PTAS is too slow** on large scale instances.
- The most successful algorithms (w. r. t. quality and running time) in practice rely on **local search**.

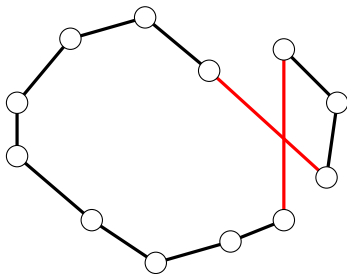


# 2-Opt Heuristic



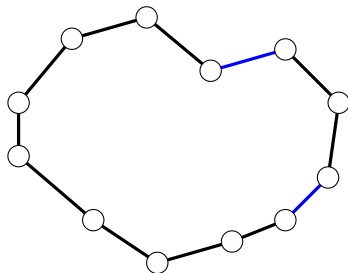
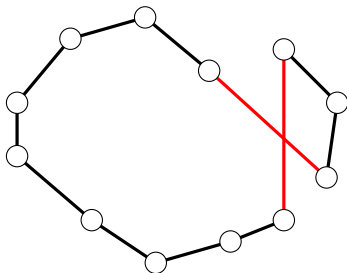
- 1 Start with an arbitrary tour.

# 2-Opt Heuristic



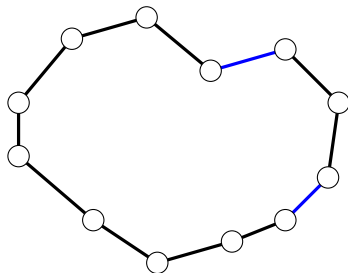
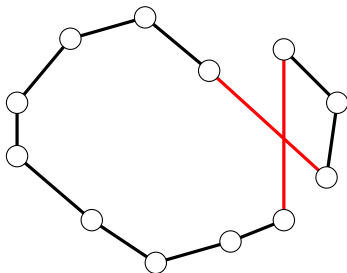
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- 2 Remove two edges from the tour.

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# 2-Opt Heuristic



- 1 Start with an arbitrary tour.
- 2 Remove two edges from the tour.
- 3 Complete the tour by two other edges.
- 4 Repeat steps 2 and 3 until no local improvement is possible anymore.

# Why 2-Opt?

Experiments on **Random Euclidean Instances**

[Johnson and McGeoch (2002)]

## Approximation Ratio

- Christofides (for  $n \leq 10^5$ ):  $\approx 1.1$
- 2-Opt (for  $n \leq 10^6$ ):  $\approx 1.05$

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## Number of Local Improvements of 2-Opt

- Greedy Starts: Probably  $O(n)$
- Random Starts: Probably  $O(n \log n)$

# Running time of 2-Opt: Known and New Results

	General TSP	Euclidean metric	Manhattan metric
average			
smoothed			
worst-case	$2^{\Omega(n)}$		

Average-case results: [Chandra, Karloff, Tovey (1999)].

Worst-case results: [Lueker (1975)].

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Our results.



# Running time of 2-Opt: Known and New Results

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# Running time of 2-Opt: Known and New Results

	General TSP	Euclidean metric	Manhattan metric
average	$n^{3+o(1)}$	$\tilde{O}(n^{10})$ $\tilde{O}(n^{4.33})$ [ $\tilde{O}(n^{3.83})$ ]	$\tilde{O}(n^6)$ $\tilde{O}(n^4)$ [ $\tilde{O}(n^{3.5})$ ]
smoothed			
worst-case	$2^{\Omega(n)}$	$2^{\Omega(n)}$	$2^{\Omega(n)}$

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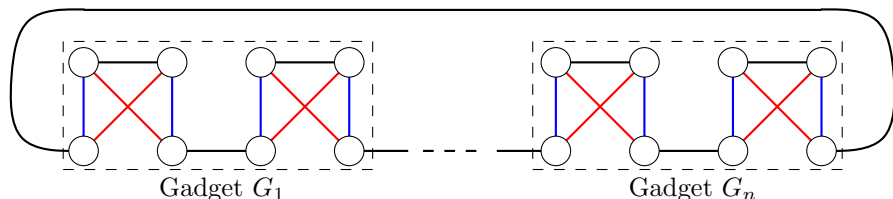
Our results.

- 1 Introduction
- 2 Lower Bound**
- 3 Upper Bound
- 4 Extensions and Open Problems

# Lower Bound

## Theorem

For every  $n \in \mathbb{N}$ , there is a graph in the Euclidean plane with  $8n$  vertices on which 2-Opt can make  $2^{n+3} - 14$  steps.



Possible States of a Gadget:

(Long, Long), (Long, Short), (Short, Long), (Short, Short)

# Lower Bound

$$(\text{Long}, \text{Long}) = 0$$

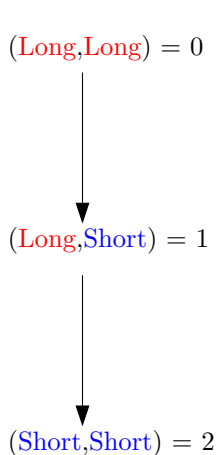


$$(\text{Long}, \text{Short}) = 1$$



$$(\text{Short}, \text{Short}) = 2$$

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0

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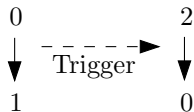
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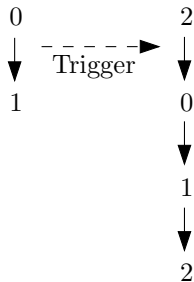
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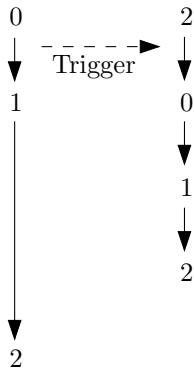
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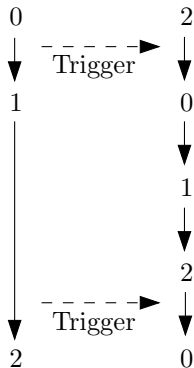
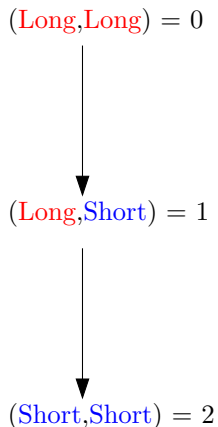
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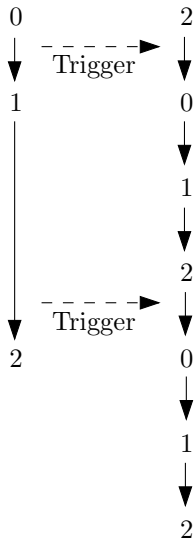
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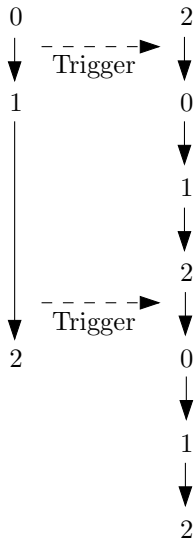
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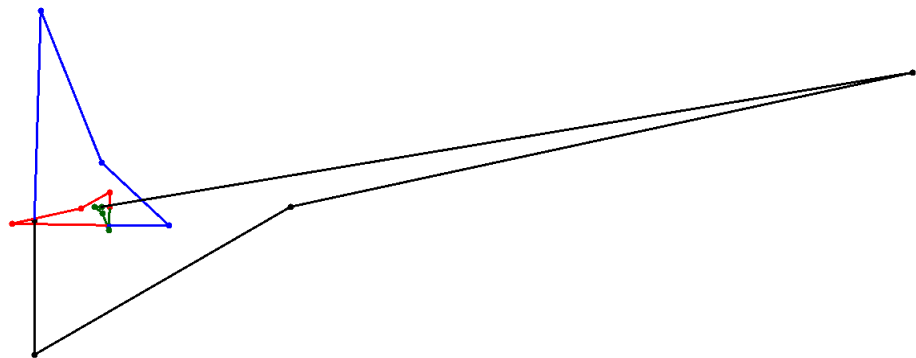


(Short, Short) = 2



Gadget  $G_i$  is reset  $2^{i-1}$  times to (Long, Long) = 0.

# Euclidean Embedding of the Gadgets



# Upper Bound

- 1 Introduction
- 2 Lower Bound
- 3 Upper Bound**
- 4 Extensions and Open Problems

## Theorem

Assume that  $n$  points are placed *independently, uniformly* at random in the unit square  $[0, 1]^2$ . The expected number of 2-Opt steps is bounded by  $O(n^{4+1/3} \cdot \log n)$  (for *every initial tour* and *every pivot rule*).



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- $\Delta(e_1, e_2, e_3, e_4) = l(e_1) + l(e_2) - l(e_3) - l(e_4)$ .



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- Bound  $\Delta$  by a **union bound**: There are  $O(n^4)$  different 2-Opt steps, analyze  $\Delta(e_1, e_2, e_3, e_4)$  for one of them.  $\Rightarrow \Delta \approx 1/(n^4 \log n)$ .

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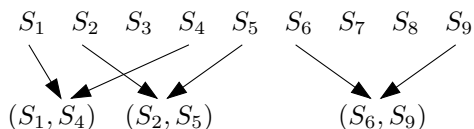
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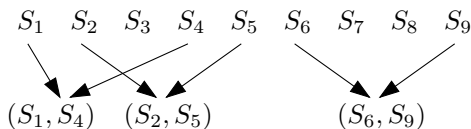
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- Sequence of  $t$  consecutive steps, contains  $\Omega(t)$  linked pairs:



- $\Delta_{\text{Linked}} \approx 1/(n^{3+1/3} \log^{2/3} n)$ .

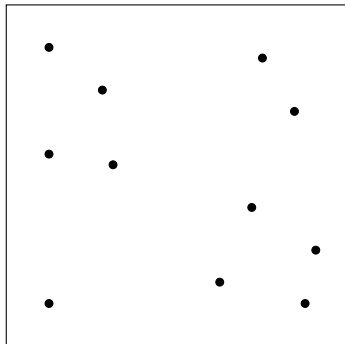
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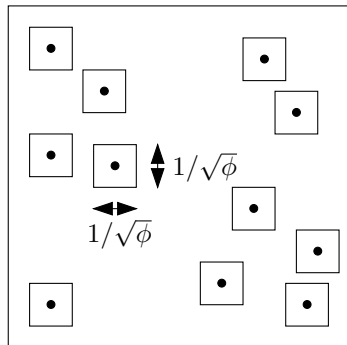
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# Smoothed Analysis

	General TSP	Euclidean metric	Manhattan metric
average	$n^{3+o(1)}$	$\tilde{O}(n^{4.33})$	$\tilde{O}(n^4)$
smoothed	$m \cdot n^{1+o(1)} \cdot \phi$	$\tilde{O}(n^{4.33} \cdot \phi^{2.67})$	$\tilde{O}(n^4 \cdot \phi)$
worst-case	$2^{\Omega(n)}$	$2^{\Omega(n)}$	$2^{\Omega(n)}$

# Approximation Ratio

- Worst Case:  $O(\log n)$
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- Average Case:  $O(1)$
- Smoothed:  $O(\sqrt{\phi})$

## Worst-Case Analysis

- Analyze the **diameter** of the 2-Opt state graph.
- Analyze **particular pivot rules** like “largest improvement”.

## Probabilistic Analysis

- Show **exact bounds** on the running time of 2-Opt and  $k$ -Opt.
- Show **small constant approximation ratio** for 2-Opt on random Euclidean instances.

Thanks!  
Questions?