

# Approximation Algorithms for some Important Geometric Problems

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# Outline

1 Travelling Salesman Problem

2 Rectangle Stabbing Problem

# Travelling Salesman Problem

## Problem

**Input:** An undirected graph  $G = (V, E)$ , with each edge  $e \in E$  attached with an integer cost  $w(e) > 0$ .

**Objective:** Find a Hamiltonian cycle of  $G$  with minimum cost.

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**Status of the problem:** The decision version is **NP-complete**.

# $\Delta$ -TSP

## A particular case

$\Delta$ -TSP: Nodes can be placed on a Euclidean plane.

Weight of each edge is equal to the distance between the corresponding pair of nodes.

# $\Delta$ -TSP

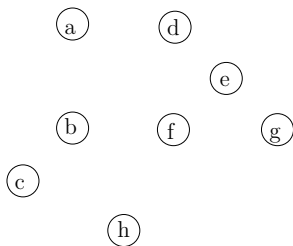
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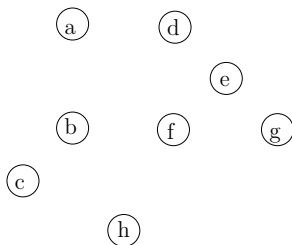
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## Status

The problem still remains **NP-complete**



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### Time Complexity

Computing the minimum spanning tree using Prim's Algorithm needs  $O(n^2)$  time (since  $G$  is a complete graph). All other works can be done in  $O(n)$  time.

# Analysis of Approximation Ratio

## Theorem

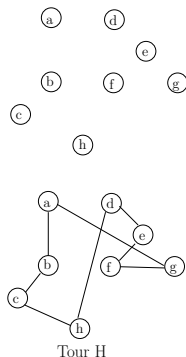
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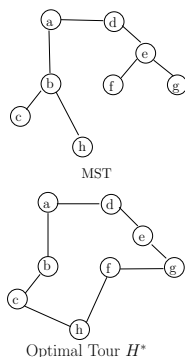
## Theorem

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$H^* \rightarrow$  Optimal Tour.



$H \rightarrow$  Tour produced by our algorithm.



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Thus  $w(\mathbf{A}) = 2 \times w(T) \leq 2 \times w(H^*)$ .

Thus,  $\mathbf{A}$  is within 2 factor of the optimal tour.

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We get a tour  $H$  from the walk  $\mathbf{A}$  as follows:

For each vertex, we remove its second occurrence in the walk except  $r$ .

By triangles inequality, we have  $w(H) \leq w(\mathbf{A})$ .

Thus we have  $w(H) \leq 2 \times w(H^*)$ . □

## Best Known Result on $\Delta$ -TSP

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## Theorem

*If  $P \neq NP$  and  $\rho \geq 1$ , there is no polynomial time apprimation algorithm for the general TSP with ratio bound  $\rho$ .*

## Proof [By contradiction]

Suppose there is a  $\rho$ -approximation algorithm  $\mathcal{A}$  for the general TSP problem. Let us assume that  $\rho$  is an integer.

We show that  $\mathcal{A}$  can solve the hamiltonian cycle problem for any arbitrary graph  $G = (V, E)$  in polynomial time.

Let  $G' = (V, E')$  be a complete graph, with  $E' = \{(u, v) \mid u, v \in V \text{ and } u \neq v\}$

Assign  $w(u, v) = 1$  if  $(u, v) \in E$ , and  
 $\rho \times |V| + 1$  otherwise.

If the original graph has a hamiltonian cycle, the optimal tour will be of cost  $|V|$ .

Any non-optimal tour will be of cost at least  
 $(\rho|V| + 1) + (|V| - 1) \geq \rho|V|$ .

### Proof (contd.)

Now we execute the algorithm  $\mathcal{A}$  on the graph  $G'$ .

If  $G$  has a Hamiltonian cycle, then  $\mathcal{A}$  must produce a tour in  $G'$  of cost at most  $\rho|V|$ .

But it is impossible unless it returns a tour in  $G'$  corresponding to the actual Hamiltonian cycle in  $G$ .

Thus we have a polynomial time algorithm for the Hamiltonian cycle problem.

# Rectangle Stabbing Problem<sup>1</sup>

## Problem Statement:

**Given a set  $\mathcal{R}$  of  $n$  axis-parallel rectangles, find the minimum number of axis-parallel lines to stab all the members in  $\mathcal{R}$ .**

**A rectangle  $r \in \mathcal{R}$  is given using a pair of coordinates  $[(a, b), (c, d)]$  corresponding to its (bottom-left, top-right) diagonal.**

For the sake of simplicity, we assume that the coordinates are integer valued.

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<sup>1</sup>Gaur, Ibaraki and Krishnamurthy, J. of Algorithms, 2002 

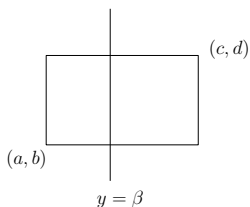


# Rectangle Stabbing Problem

## Definition

An axis-parallel (horizontal/vertical) line  $\ell$  stabs a rectangle  $r = [(a, b), (c, d)] \in \mathcal{R}$  if  $\ell$  passes through the interior of the rectangle  $r$ . if

## Example:



Here the line  $y = \beta$  stabs the rectangle.  
This implies  $a + 1 \leq \beta \leq c - 1$

# Status of the Problem and Our Objective

Status:

**The problem is NP-hard**

Reference: Hasin and Megiddo, Discrete Appl. Math., 1991

Objective: To design a constant factor approximation algorithm

Result available:

**A 2-factor approximation algorithm that runs in  $O(n^5)$  time.**

**Tools used:** LP-relaxation.

# Integer Programming Formulation

**Solution Space:**  $H \cup V$ .

- $H$  – Set of  $2n$  horizontal lines at  $y = b_i + \epsilon$  and  $y = d_i - \epsilon$ , where  $b_i$  and  $d_i$  are  $y$ -coordinates of bottom and top boundaries of  $i$ -th rectangle.
- $V$  – Set of  $2n$  vertical lines at  $x = a_i + \epsilon$  and  $x = c_i - \epsilon$ , where  $a_i$  and  $c_i$  are  $x$ -coordinates of left and right boundaries of  $i$ -th rectangle.

Take  $4n$  decision variables, namely  $x_1, x_2, \dots, x_{2n}, y_1, y_2, \dots, y_{2n}$ .

$x_i$  – corresponds to  $i$ -th vertical line, and

$y_j$  – corresponds to  $j$ -th horizontal line.

These decision variables can assume values 0 and 1 only.

# Integer Programming Formulation

$H_k$  – Set of horizontal lines that stab the rectangle  $r_k$ , and

$V_k$  – Set of vertical lines that stab the rectangle  $r_k$ .

## Integer Programming Problem – $P$

Objective Function:

$$\min \sum_{i \in V} x_i + \sum_{j \in H} y_j$$

Constraints:

For each rectangle  $r_k$ ,  $k = 1, 2, \dots, n$ , we have the constraint

$$\sum_{i \in V_k} x_i + \sum_{j \in H_k} y_j \geq 1$$

$$x_i \in [0, 1], \quad \text{for all } i = 1, 2, \dots, 2n, \text{ and}$$

$$y_j \in [0, 1], \quad \text{for all } j = 1, 2, \dots, 2n.$$

## LP Relaxation

Linear Programming Problem –  $\bar{P}$ 

Objective Function:

$$\min \sum_{i \in V} x_i + \sum_{j \in H} y_j$$

Constraints:

$$\sum_{i \in V_k} x_i + \sum_{j \in H_k} y_j \geq 1, \quad \text{for all } k = 1, 2, \dots, n$$

$$x_i \geq 0, \quad \text{for all } i = 1, 2, \dots, 2n, \text{ and}$$

$$y_j \geq 0, \quad \text{for all } j = 1, 2, \dots, 2n.$$

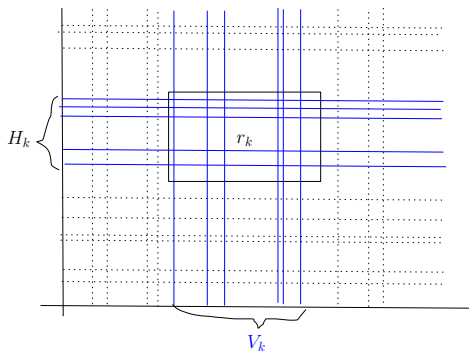
# Analysis of LP Solution

Let  $\bar{x}_i, i = 1, 2, \dots, 2n$  and  $\bar{y}_j, j = 1, 2, \dots, 2n$   
be an optimal fractional solution of the LP problem  $\bar{P}$ .

For each rectangle  $r_k, k = 1, 2, \dots, n$ , we have

either  $\sum_{i \in V_k} \bar{x}_i \geq \frac{1}{2}$

or  $\sum_{j \in H_k} \bar{y}_j \geq \frac{1}{2}$ .



## Analysis of LP Solution

Let  $R_H$  be the set of all  $k$  such that  $\sum_{i \in V_k} \bar{x}_i \geq \frac{1}{2}$ , and

$R_V$  be the set of all  $k$  such that  $\sum_{j \in H_k} \bar{y}_j \geq \frac{1}{2}$ .

### Implication:

- The set of rectangles in  $R_H$  will be stabbed by horizontal lines, and
- The set of rectangles in  $R_V$  will be stabbed by vertical lines.

Thus, we have the following two problems

$P_H$ : Compute the minimum clique cover of an interval graph with the vertical intervals corresponding to the rectangles in  $R_H$ , and

$P_V$ : Compute the minimum clique cover of an interval graph with the horizontal intervals corresponding to the rectangles in  $R_V$ .

$P_H$  and  $P_V$  can be optimally solved in polynomial time.

# Analysis of Approximation Factor

Let

$Q$ : Optimum solution of the integer programming problem  $P$ ,

$\hat{Q}$ : Optimum solution of the linear programming problem  $\hat{P}$ ,

$Q_H$ : Optimum solution of the clique cover problem  $P_H$ ,

$Q_V$ : Optimum solution of the clique cover problem  $P_V$ .

## Theorem

$$Q_H + Q_V \leq 2Q.$$



# Analysis of Approximation Factor

**Proof:** Let  $\hat{Q} = (\hat{x}, \hat{y})$  be an optimal fractional solution of  $\hat{P}$   
 $\implies Q_H^* = 2\hat{x}$  and  $Q_V^* = 2\hat{y}$  are feasible solutions of  $P_H$  and  $P_V$ .

**Reason:** For every  $k \in R_H$ , we have  $\sum_{i \in H_k} \hat{y}_i \geq \frac{1}{2}$   
 $\implies \sum_{i \in H_k} 2\hat{y}_i \geq 1$ .

We have,

- $Q_H + Q_V \leq Q_H^* + Q_V^*$   
 [since  $Q_H$  and  $Q_V$  are optimum solutions and  $Q_H^*$  and  $Q_V^*$  are feasible solutions of the same minimization problems.]
- $Q_H^* + Q_V^* = 2(\hat{x} + \hat{y}) = 2\hat{Q}$ .
- $\hat{Q} \leq Q$   
 [Since optimum solution of an LP minimization problem is less than the optimum solution of its corresponding IP problem]

Thus, we have the proof of the result.