

# An Exact ETH-Tight Algorithm for Euclidean TSP

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Shonan Workshop March 8, 2019

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# The Traveling Salesman Problem

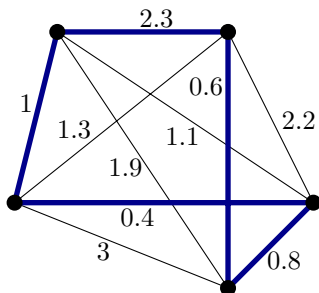
## TSP

Given a complete graph with edge weights, find the shortest round trip that visits all vertices exactly once.

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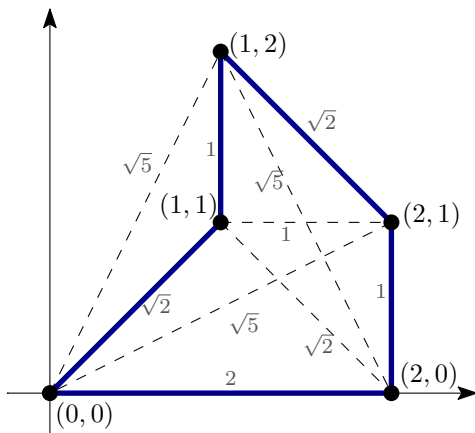
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# Applications

- Logistics
- Microchips (printing/drilling)
- Astronomy (pointing the telescope)
- Robotics
- ...

Princeton Series in APPLIED MATHEMATICS

## The Traveling Salesman Problem

*A Computational Study*



David L. Applegate,  
Robert E. Bixby, Vašek Chvátal,  
and William J. Cook



# Solving TSP

- Exact methods in practice (e.g., ILP's, matching upper and lower bound heuristics, . . . )
- Approximation: PTAS by Arora and by Mitchell, improved by Rao and Smith ('98-'99)
- This talk focus on worst case time of exact algorithms

# Computational model

How hard is it to test for integers  $a_1, \dots, a_r, b_1, \dots, b_s$  if

$$\sum_{i=1}^r \sqrt{a_i} \leq \sum_{j=1}^s \sqrt{b_j}$$

??

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Smith–Wormald	'98	ETSP $\mathbb{R}^d$	$n^{O(\sqrt{n})}, (n^{O(n^{1-1/d})})$

# Contribution

Earlier:  $n^{O(n^{1-1/d})} = 2^{O(n^{1-1/d} \log n)}$  algorithm and  $2^{\Omega(n^{1-1/d-\epsilon})}$  lower bound.

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## Theorem (Main)

*For any fixed  $d$ , there is a  $2^{O(n^{1-1/d})}$  algorithm for Euclidean TSP in  $\mathbb{R}^d$ . All algorithms need  $2^{\Omega(n^{1-1/d})}$  time under ETH.*



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## Theorem

*There is a  $2^{O(\sqrt{n})}$  algorithm for Euclidean TSP in  $\mathbb{R}^2$ . All algorithms need  $2^{\Omega(\sqrt{n})}$  time under ETH.*

# On the lower bounds

## Theorem (Main)

*For any fixed  $d$ , all algorithms for Euclidean TSP in  $\mathbb{R}^d$  need  $2^{\Omega(n^{1-1/d})}$  time under ETH.*

Follows from B-B-K-Marx-v.d.Zanden '18 for Ham. Cycle in  $\mathbb{Z}^d$ .

1. ETH with sparsification
2. Embedding of 3-SAT formula in  $d$ -dimensional space, and
3. modifying existing NP-hardness proof of Hamiltonian Circuit: 3-SAT in  $d$ -dimensional space  $\rightarrow$  HC in  $d$ -dim space
4. Building / modifying gadgets

# Contribution

## Theorem (Main)

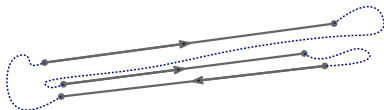
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Main ideas:

1. Balanced separating point set with a square (or cube)
2. Recursively separating gives tree structure
3. Packing property guarantees that 'few edges in solution cross cube boundary'
4. Bounding the number of candidate sets of edges across a separator: twiggling square
5. Bounding the number of ways endpoints of these edges are connected (matchings): small representative set with rank based approach

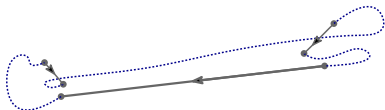
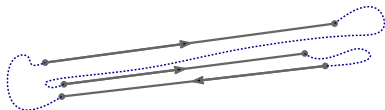
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Could this tour be optimal?



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Could this tour be optimal? → No, it can be shortened.



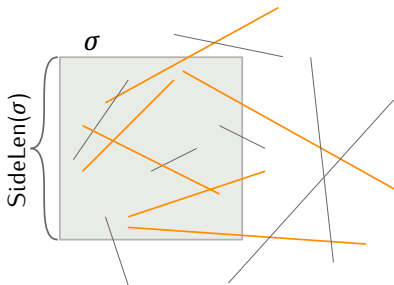
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## Definition

A segment set has the *packing property* if for any square  $\sigma$ , there are only  $O(1)$  segments of length at least  $\text{SideLen}(\sigma)/2$  intersected by  $\text{int}(\sigma)$ .



# Using the packing property

## Lemma

The segments of an optimal TSP tour in  $\mathbb{R}^d$  have the packing property.

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Already observed by Kann ('92) and Smith-Wormald ('98).

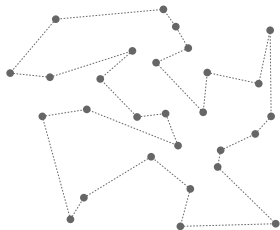
Idea behind algorithm:

- Find separator square  $\sigma$  intersected by  $O(\sqrt{n})$  tour segments
- Solve subproblems recursively

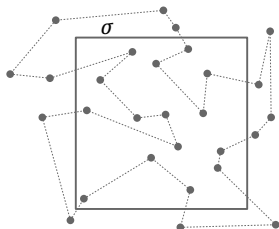
From Packing property: such separator exists!



# The separator approach (in $\mathbb{R}^2$ )

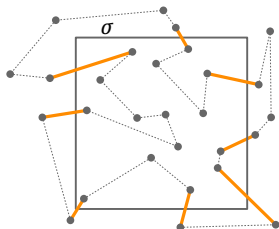


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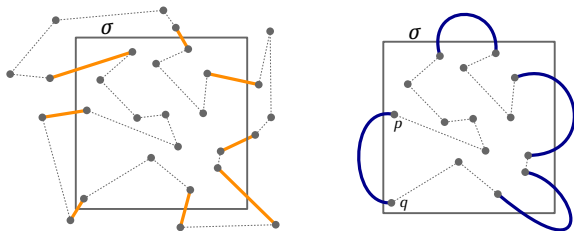
1. Find a square  $\sigma$  such that
  - (a)  $\sigma$  partitions  $P$  into subsets  $P_{\text{in}}$  and  $P_{\text{out}}$  in a balanced way and
  - (b)  $\sigma$  intersects  $O(\sqrt{n})$  segments of the (unknown) optimal tour

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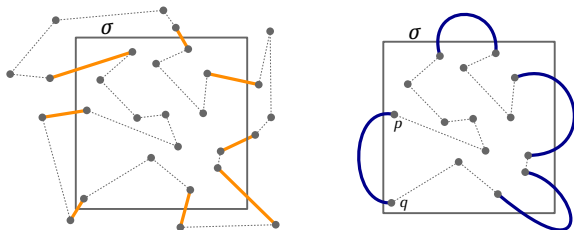
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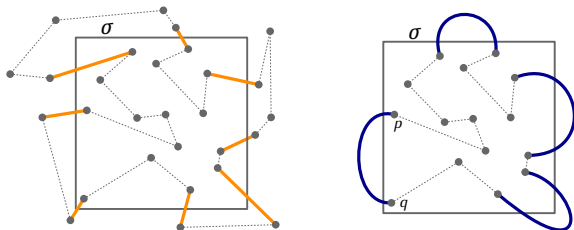
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Running time:

**# of candidate sets**

$\times$

**# of matchings**

## Bottleneck 1: number of candidate sets

Running time: # of candidate sets × # of matchings

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We have

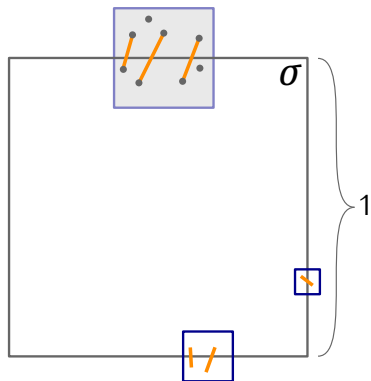
$$\simeq \binom{\binom{n}{2}}{c\sqrt{n}} = 2^{\Theta(\sqrt{n} \log n)} \text{ candidate sets...}$$

This is tight for known separator theorems.

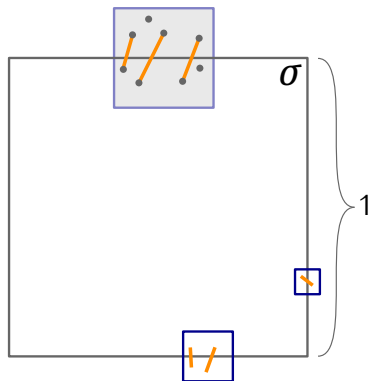


# Resolving Bottleneck 1: Pushing the packing property further

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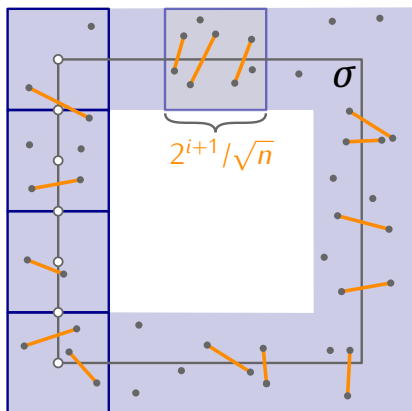
$S$  has the packing property.

Split  $S$  into length classes:

$$S_i := \left\{ s \in S \mid \frac{2^{i-1}}{\sqrt{n}} \leq s < \frac{2^i}{\sqrt{n}} \right\}$$

Guess each  $S_i$  separately.

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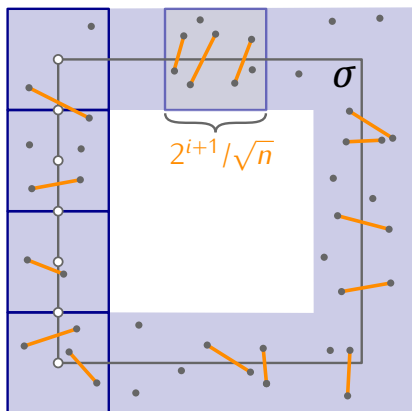
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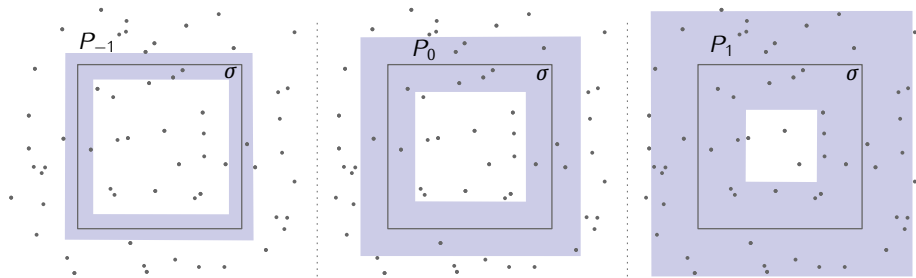
$S_i$  is inside **annulus** of width  $\frac{2^{i+1}}{\sqrt{n}}$ .

Few guesses for  $S_i \Leftrightarrow$  few pts in the  $i$ -th annulus.

We need sparse annuli around  $\sigma$ .

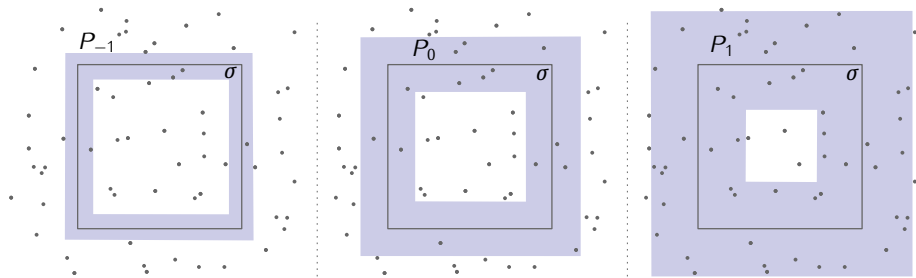
# The separator theorem in $\mathbb{R}^2$

$P_i :=$  pts of  $P$  at distance  $\leq 2^i/\sqrt{n}$  from  $\sigma$



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## Theorem

Given  $P \subset \mathbb{R}^2$ , there is a balanced separator  $\sigma$  such that  $|P_i(\sigma)| \leq c^i \sqrt{n}$ , and  $\sigma$  can be found in polynomial time.

Only  $2^{O(\sqrt{n})}$  candidates

### Theorem

*For any set of  $n$  points in  $\mathbb{R}^2$ , there is a balanced separator  $\sigma$  such that*

- (i) each candidate set  $S$  contains  $O(\sqrt{n})$  segments*

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*Moreover,  $\sigma$  and the candidates can be computed in  $2^{O(\sqrt{n})}$  time.*

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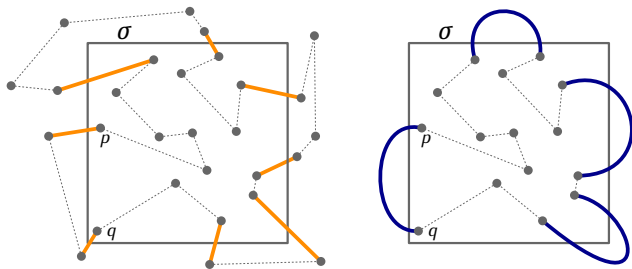
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Bottleneck 1 ✓

## Bottleneck 2: number of matchings

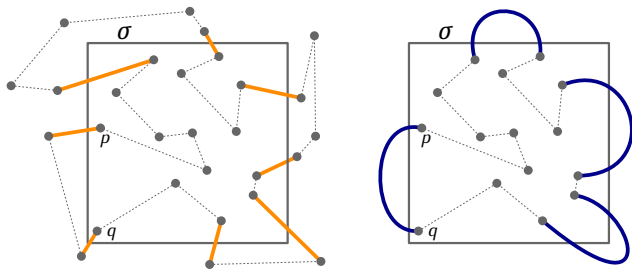
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There are  $2^{\Theta(\sqrt{n} \log n)}$  matchings on  $c\sqrt{n}$  points...

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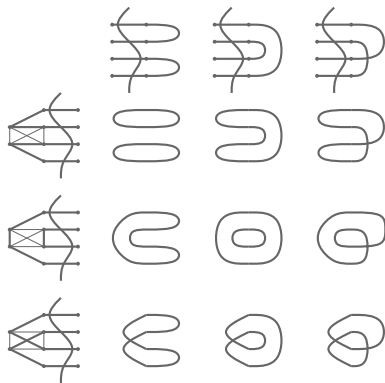
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Resolution: adapt **Rank Based Approach** ('15) by Bodlaender *et al.*

# Rank based approach








- Introduced for solving connectivity problems like Hamiltonian Circuit, Steiner Tree, Connected Dominating Set, ... in  $O(2^{O(tw)}n)$  time on graphs of small treewidth  $tw$  by B, Cygan, Nederlof, Kratsch (2015)
- Application, better rank bound for HC-like problems by Cygan, Nederlof, Kratsch (2018)
- Experimental evaluation for Steiner tree by Fafianie, B, Nederlof (2015)
- Experimental evaluation for Hamiltonian Circuit by Pilipczuk, Ziobp (2019)

# Back to Hamiltonian Circuit on Graphs



If we have two of these in a table, we do not need the third!

# The rank based approach

			
	0	1	1
	1	0	1
	1	1	0

We can drop a row, when for each 1 in the row, another row has also a 1 in that column

# The rank based approach

The connectivity matrix

	○	1	1
	1	○	1
	1	1	○

We can drop a row, when for each 1 in the row, another row has also a 1 in that column: a sufficient condition is that the row is a linear combination mod2 of other rows



# Rank based approach scheme for HC on graphs of small treewidth

- Do a 'usual' DP on the tree decomposition, BUT
- If a table has more rows than the rank of the connectivity matrix, then REDUCE

REDUCE:

Build the part of the connectivity matrix with

rows: the entries in the current table

columns: a basis of the connectivity matrix

Sweep with Gauss elimination (compute mod2)

Remove every row with only 0's

Gives 'representative set', and we end with at most  $\text{rank}(M)$  number of table entries

# Ranks

Connectivity matrix: columns and rows are partitions; 1 if closure of both partition connects all elements, 0 otherwise

Connectivity matrix for matchings: rows and columns are a matching; 1 if combination gives one cycle, 0 otherwise

## **Theorem (BCKN (see also Lovasz), CKN)**

*The rank of the connectivity matrix for  $k$  elements is  $2^{k-1}$ . The rank of the connectivity matrix for matchings on  $k$  elements is  $2^{k/2-1}$ .*

# Picture from paper by Cygan, Nederlof, Kratsch































Nr.		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	LC
																	
1		0	0	0	0	1	1	0	1	1	1	1	0	1	1	0	1
2		0	0	0	1	0	1	1	1	0	0	1	1	1	0	1	2
3		0	0	0	1	1	0	1	0	1	1	0	1	0	1	1	1+2
4		0	1	1	0	0	0	0	1	1	1	0	1	1	0	1	4
5		1	0	1	0	0	0	1	0	1	0	1	1	1	1	0	5
6		1	1	0	0	0	0	1	1	0	1	1	0	0	1	1	4+5
7		0	1	1	0	1	1	0	0	0	0	1	1	0	1	1	1+4
8		1	1	0	1	0	1	0	0	0	1	0	1	1	1	0	2+4+5
9		1	0	1	1	1	0	0	0	0	1	1	0	1	0	1	1+2+5
10		1	0	1	1	0	1	0	1	1	0	0	0	0	1	1	2+5
11		1	1	0	0	1	1	1	0	1	0	0	0	1	0	1	1+4+5
12		0	1	1	1	1	0	1	1	0	0	0	0	1	1	0	1+2+4
13		1	1	0	1	1	0	0	1	1	0	1	1	0	0	0	1+2+4+5
14		1	0	1	0	1	1	1	1	0	1	0	1	0	0	0	1+5
15		0	1	1	1	0	1	1	0	1	1	0	0	0	0	0	2+4

Figure 1: The matrix  $\mathcal{H}_6$ . Letting the baseset be  $\{0, \dots, 5\}$  matching 1 indexing row and column 1 equals  $\{\{0, 1\}, \{2, 3\}, \{4, 5\}\}$ . The set  $\mathbf{X}_\ell = \{1, 2, 4, 5\}$  from Definition 3.1 is easily seen to be a row basis: the linear combinations are depicted in the last column.

# Weighted solutions

Sort the rows with respect to non-decreasing cost

Gaussian elimination top-to-bottom: eliminate rows that are a linear combination of 'cheaper' rows

## Using the rank based approach here

At each step in the recursion:

- For each candidate set of edges across the separating square:
- We have  $n^{1-1/d}$  endpoints of the candidate set that can be matched inside and outside
  - Recursively, build representative set of matchings inside
  - Recursively, build representative set of matchings outside
- Make all combinations of inside and outside

In 2d, one can also use that matchings are non-overlapping and use Catalan structures

This resolves Bottleneck 2.

# Conclusion

We can solve Euclidean TSP exactly for constant  $d$  in  $2^{O(n^{1-1/d})}$  time. This is tight under ETH.

- Rank= $\infty$ -based approach can give practical and theoretical faster algorithms on tree decompositions and similar structures

# Conclusion

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