

# MST based Pyramid Model of TSP

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# Multiresolution Pyramid Representation

## Approach

- Local to global: **bottom-up** reduction of resolution
- Global to local: **top-down** solution refinement

Approximate the TSP solution using pyramid representation

The idea

- 1 partition the input space
  - preserve approximate location
- 2 reduce number of cities
- 3 repeat until solution becomes trivial
- 4 refine solution top down to the base level

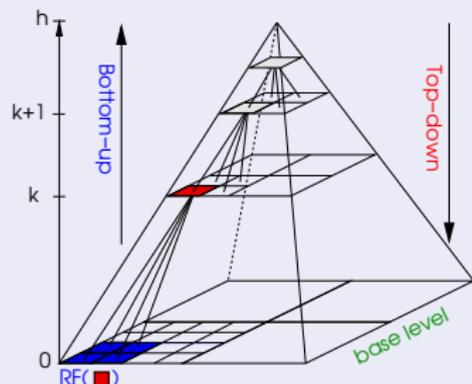


# Image Pyramids

## Introduction

- Hierarchical structures in computer vision
  - ⇒ image pyramids, wavelets, quad-trees ...
- Characteristics of pyramids:
  - Structure
    - horizontal and vertical relations
  - Content of the cells
    - numeric, symbolic or both
  - Processing of a cells

### $2 \times 2/4$ Regular pyramid



# Image Pyramids

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# Outline of the Talk

- 1 Image Pyramids
- 2 Minimum Spanning Tree
- 3 Approximate Traveling Salesman Problem
- 4 Summary



# How to Organize/Partition City-Space?

- 1 Raster cell with/without city
- 2 Graph  $G = (V, E)$ : city = vertex  $v \in V$ ; edges  $e \in E$ ?
  - Complete graph:  $E = V \times V$
  - Delaunay triangulation  $E \subset V \times V$  and Voronoi Diagram



# Traveling Salesman Problem (TSP)

- Fully connected graph  $G = (V, E, w)$  and attributed by weight  $w$ 
  - costs  $w : e \in E \rightarrow R^+$
- **Goal** : Find the tour  $\tau$  with the smallest weight  
 $\sum_{e \in \tau} w(e) \rightarrow \min.$ 
  - If the weights are (2D) Euclidean distances  $\rightarrow$  (2D) E-TSP
  - TSP (E-TSP) is **hard optimization problem**  
 $\rightarrow$  solution: **approximation algorithms**



# TSP with triangle inequality

Fakts:

- MST is a natural **lower bound** for the length of the optimal route.
- In TSP with triangle inequality, it is possible to prove **upper bounds** in terms of the minimum spanning tree → 'Christofides Heuristics' [Christofides, 1976] ...



# Minimum Spanning Tree (MST)

- Graph  $G = (V, E, w)$  connected and attributed by weight  $w$ 
  - $w : e \in E \rightarrow R^+$
- **Goal** : Find the spanning tree  $T$  with the smallest weight  $\sum_{e \in T} w(e) \rightarrow \min.$ 
  - **Easy optimization problem**  $\rightarrow$  solution: **greedy algorithms**
  - Kruskal's [Kruskal, 1956] and Prim's algorithms [Prim, 1957]
  - **Borůvka's algorithm** [Borůvka, 1926] ( $\mathcal{O}(|E| \log |V|)$ )



# MST Algorithm

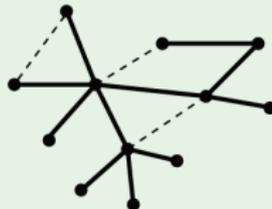
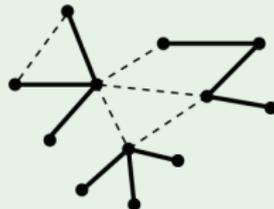
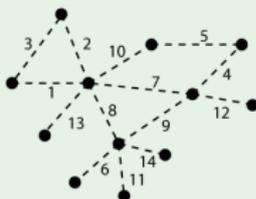
Borůvka's Version

## MST Algorithm [Borůvka, 1926]

**Input:** graph  $G = (V, E, w)$

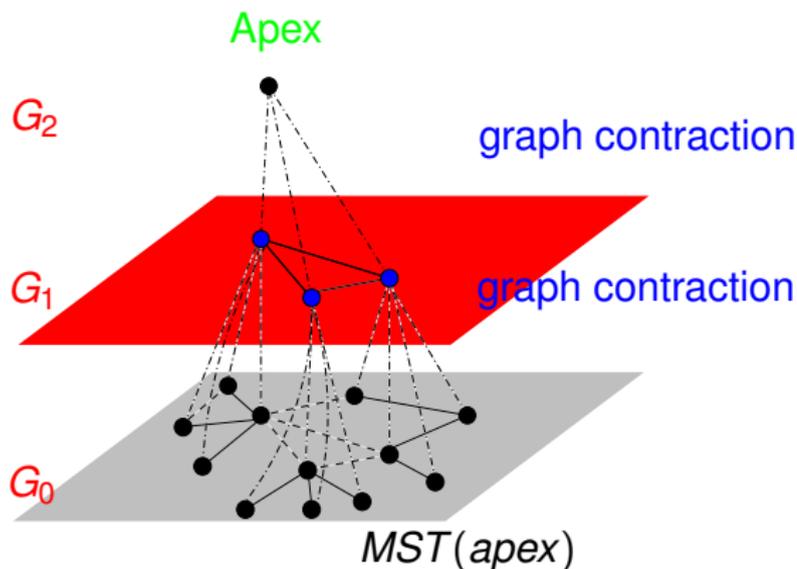
- 1:  $MST :=$  empty edge list
- 2:  $\forall v \in V$  make a list of trees  $L$
- 3: **while** there is more than one tree in  $L$  **do**
- 4:   each tree  $T \in L$  finds the edge  $e$  with the minimum weight which connects  $T$  to  $G \setminus T$  and add edge  $e$  to  $MST$
- 5:   using edge  $e$  merge pairs of trees in  $L$
- 6: **end while**

**Output:** minimum spanning tree of  $G$



# Borůvka's Algorithm and Dual Graph Pyramid

- Graph contraction merges all trees  $T \in L$  in step 3.
- Step 4: called Borůvka's step



# Pyramid Solution

## MST based approximate Algorithm for TSP

**Input:** graph  $G = (V, E, w)$

- 1: **while** there is more than **3** cities **do**
- 2:     merge no more than  $k$  cities using Borůvka's step
- 3: **end while**
- 4: Find the trivial tour  $\tau^*$
- 5: **repeat**
- 6:     refine  $\tau^*$ :  $\tau \leftarrow \tau^*$
- 7: **until** the bottom of the pyramid

**Output:** (approximated) tour  $\tau$

$k \in \mathcal{N}$  and  $k \geq 2$

- + invariant to shifts of the input
- Borůvka's step on fully connected graph  $\rightarrow$   
 $\mathcal{O}(|V|^2)$

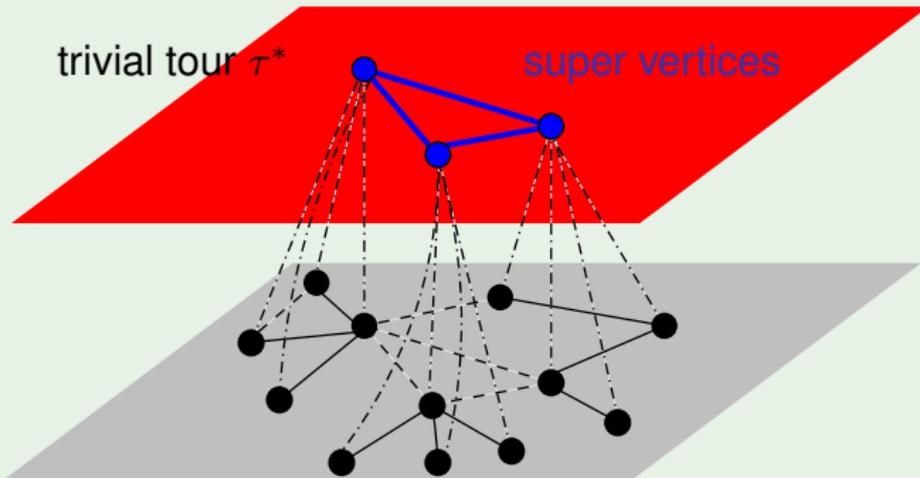


# Top-down flow

## Tour Refinement

trivial tour  $\tau^*$

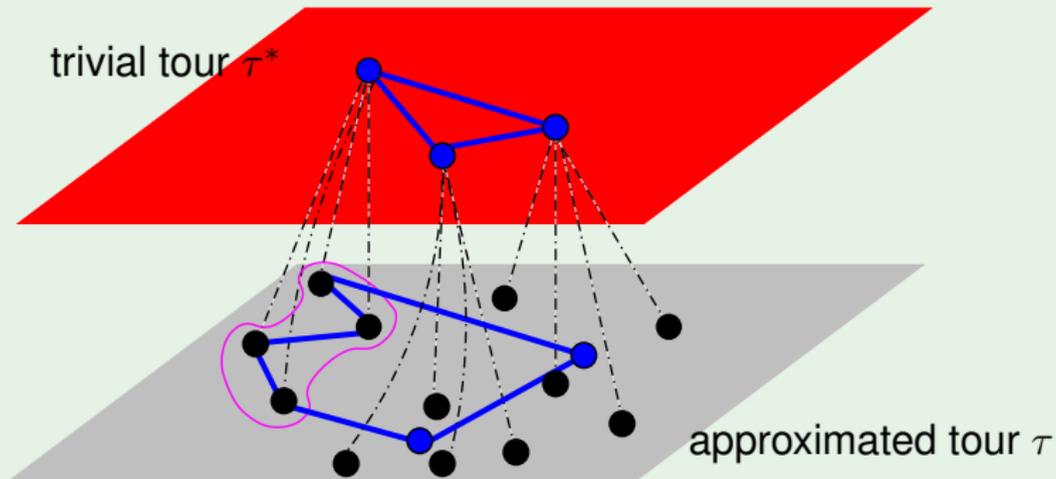
super vertices



# Top-down flow

## Tour Refinement

trivial tour  $\tau^*$



# Solution Errors on Random Instances

## MST Pyramid implementation issues

- $k = 7$
- super vertices on gravitational center of clusters

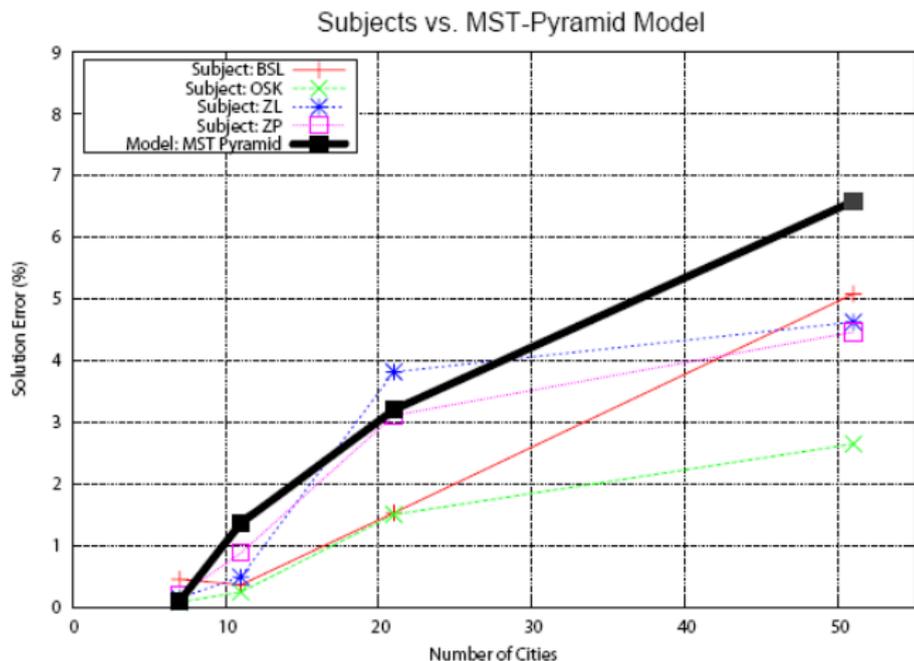
## Tested on 6, 10, 20, 50 random instances

- Human subjects
- MST pyramid



# Solution Errors on Random Instances

## Subjects vs MST-based Pyramid



# Comparing Solutions of Pyramid Algorithms

## Compared

- Concorde algorithm [Applegate et al., 2001]
- Adaptive binary pyramid [Pizlo et al., 2006]
- MST-based pyramid

## Input

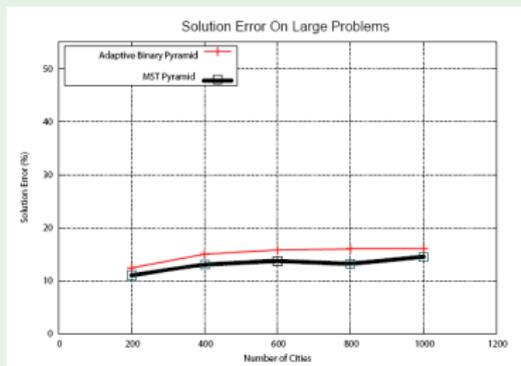
- Random instances of 200, 400, 600, 800, 1000

Code: <http://bigbird.psych.purdue.edu/~pizlo/> and  
<http://www.tsp.gatech.edu/concorde/index.html>

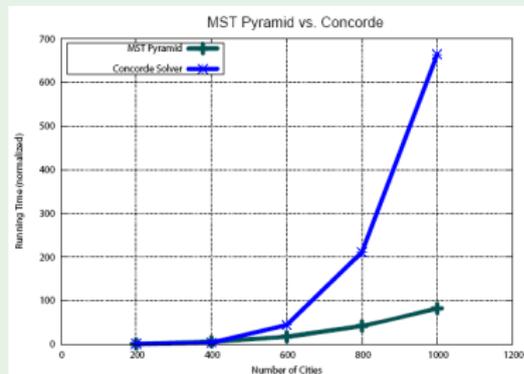


# Comparison of the Algorithms

## Pyramid Approximation



Solution Error



Running Time



# Special TSP Instances

When subjects are tested on random instances they always produce close to optimal solutions

## Hypothesis 1

Subjects minimize the total length of the tour

## Hypothesis 2

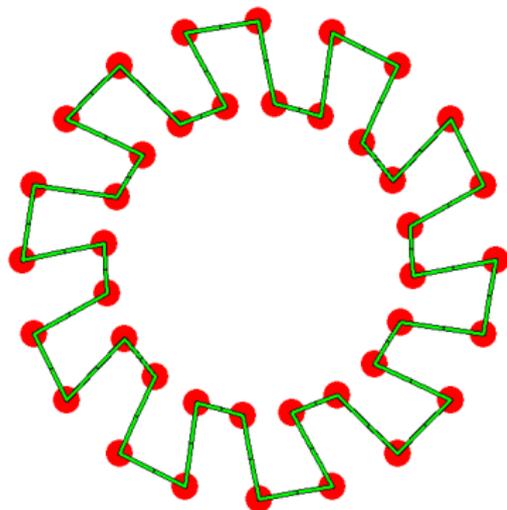
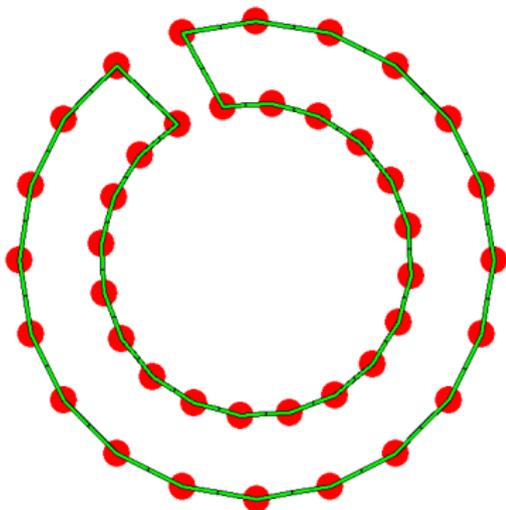
Subjects optimize something else than the length  
(this is how pyramid models work)

To tell between these two hypotheses nonrandom problems should be used?



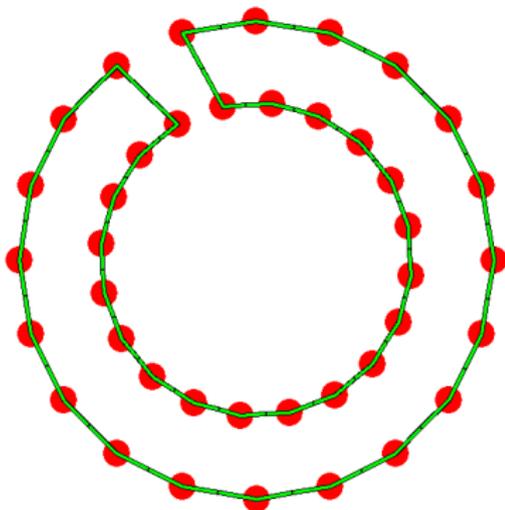
# Two Rings ZigZag

40 city instance

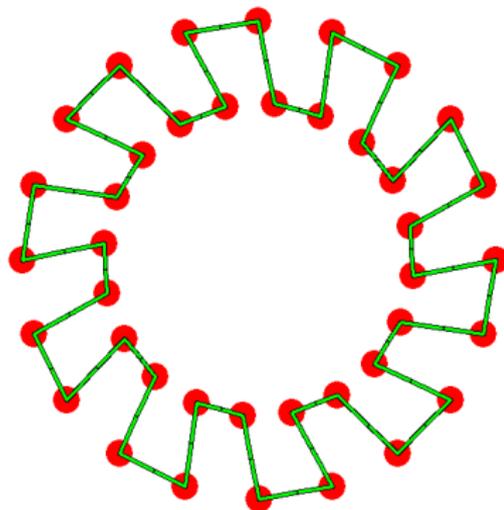


# Two Rings < ZigZag

40 city instance

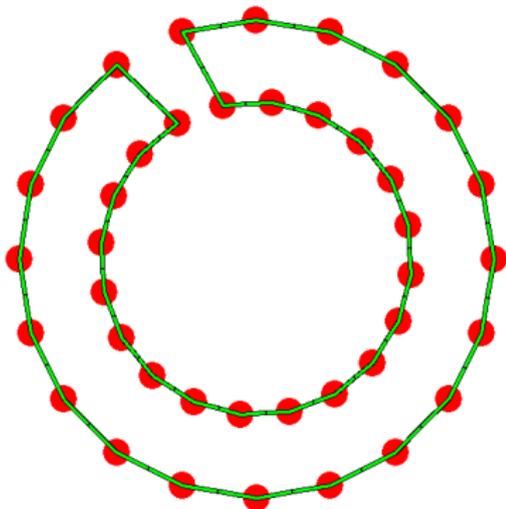


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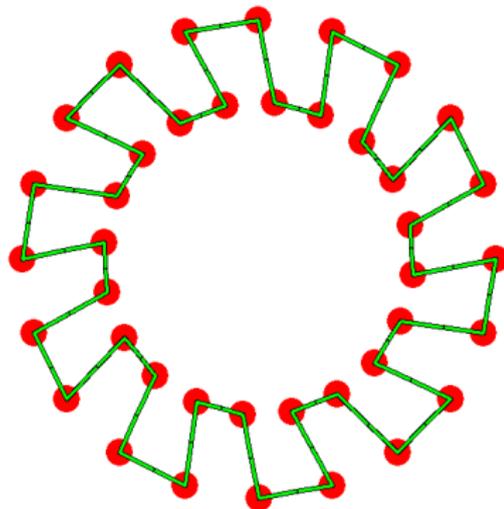
# Two Rings < ZigZag

40 city instance



2596.42

<



3013.6

16% difference  
Optimal: 2594.07



# Two Ring's length

- Two concentric circles with radii  $R > r > 0$
- If density of cities along circle is high:
  - 1 the optimal tour follows one circle,
  - 2 switches then to the other circle
  - 3 which it follows in the opposite orientation
  - 4 and returns then back to the starting city.
- tour length (approx.):  $L_R = 2\pi R + 2\pi r + 2(R - r)$
  
- Solution by nearest neighbor possible



# ZigZag's length

- If density of cities along circle is low:
  - 1 the shorter tour follows the two circles
  - 2 jumping forth and back between the 2 circles in a zigzag fashion.
- tour length with  $n$  cities (approx.):  $L_Z = \pi(R + r) + \frac{n}{2}(R - r)$



# Two Rings = ZigZag

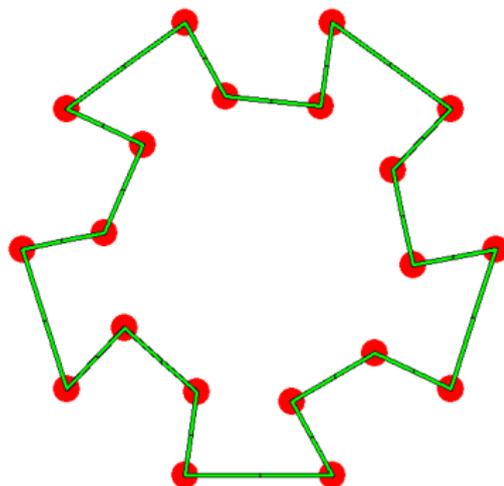
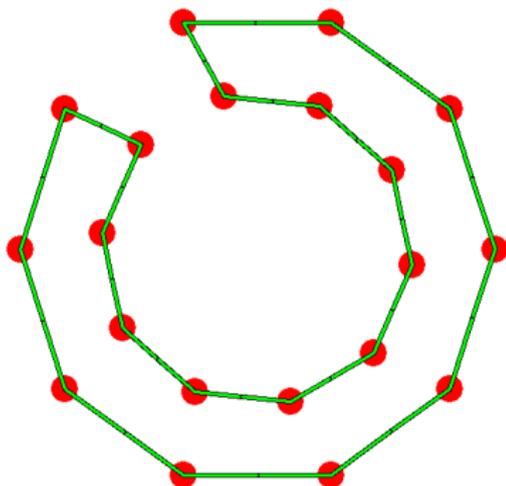
- If  $L_R = L_Z$  both solutions have the same (optimal) tour length:
- let  $n_0$  denote the number of cities for this case.
- Then  $n_0 = 2 + 2\pi \frac{R+r}{R-r}$

- $$\frac{L_Z < L_R \quad | \quad L_R < L_Z}{n < n_0 \quad | \quad n = n_0 \quad | \quad n > n_0}$$



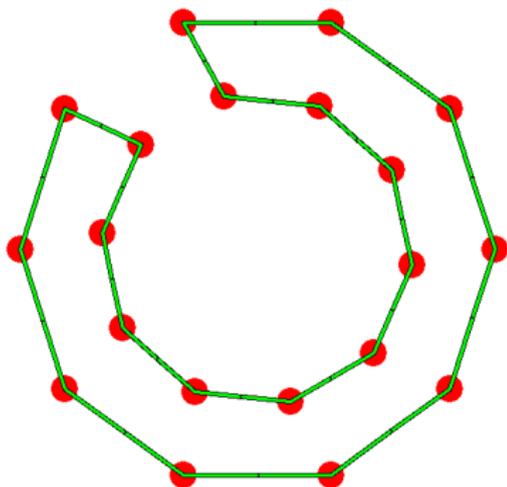
# Two Rings ZigZag

20 city instance

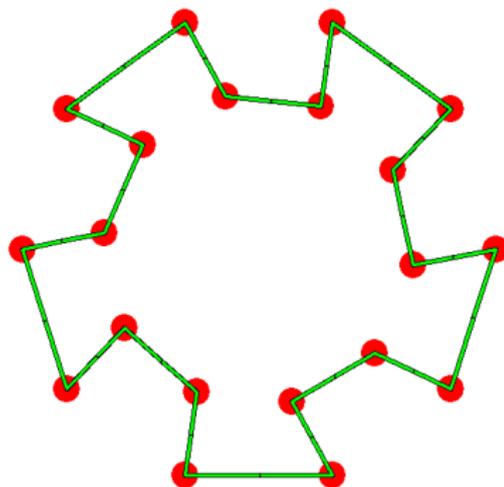


# Two Rings > ZigZag

20 city instance

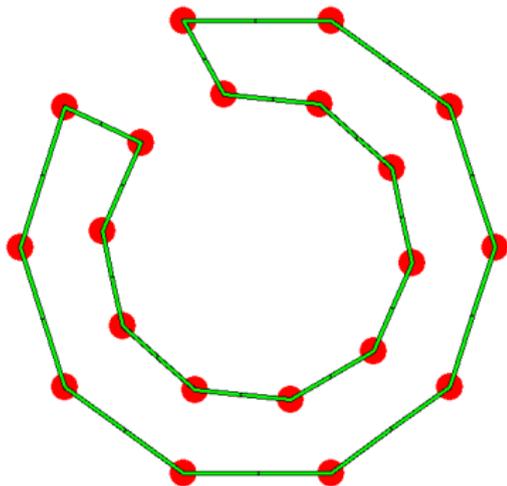


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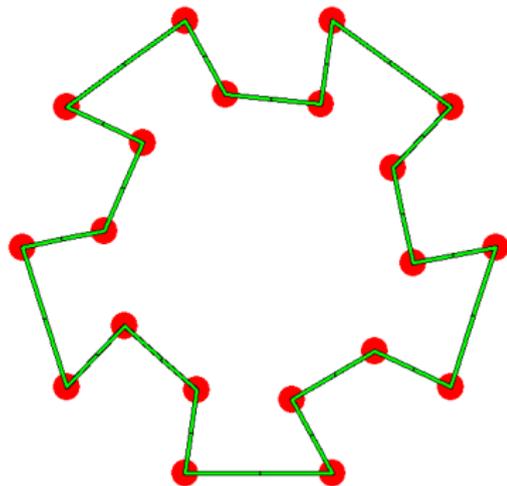
# Two Rings > ZigZag

20 city instance



2440.28

>



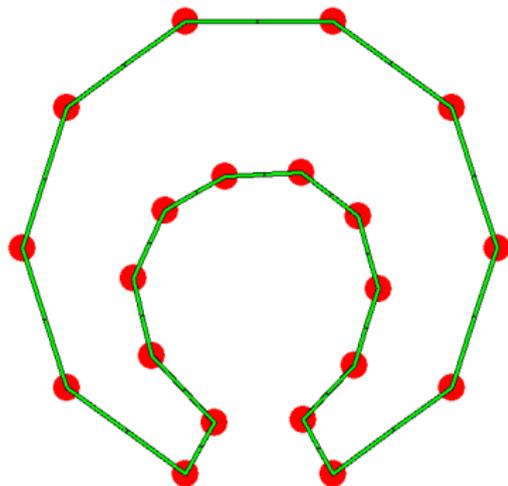
2127.77

14% difference



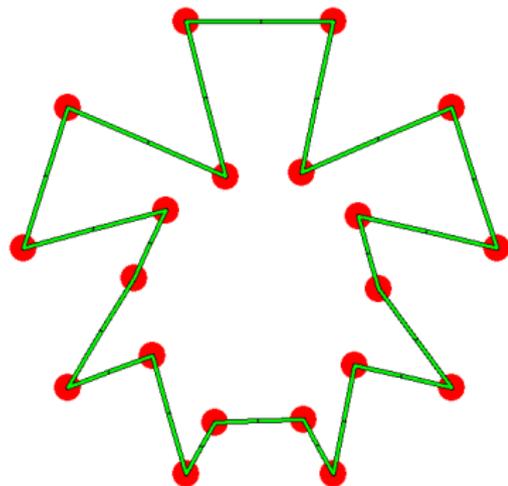
# Two non-concentric Rings $<$ ZigZag

20 city instance



2209.15

$<$



2501.91

13% difference



# Summary

- MST based TSP approximation algorithm
  - shows similar results as the other pyramid models on random instances
- Non-random instances suggest that humans do not minimize the total length of the tour
- Subjects and models will be tested on **non-random** instances
  - e.g. two circles with parameters: Radii  $R, r$ , number of cities  $n$ , circle offset



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*Human Problem Solving Symposium  
Vancouver, BA, Canada 1st of August 2006*



# References I



Jolion, J.-M. and Rosenfeld, A. (1994).  
*A Pyramid Framework for Early Vision.*  
 Kluwer.



Bister, M., Cornelis, J., and Rosenfeld, A. (1990).  
 A critical view of pyramid segmentation algorithms.  
*Pattern Recognition Letters*, 11(9):605–617.



Kropatsch, W. G. (1995a).  
 Building irregular pyramids by dual graph contraction.  
*IEE-Proc. Vision, Image and Signal Processing*, 142(6):366–374.



Felzenszwalb, P. F. and Huttenlocher, D. P. (2004).  
 Efficient graph-based image segmentation.  
*International Journal of Computer Vision*, 59(2):167–181.



S. M. Graham, Z. Pizlo and A. Joshi.  
 Problem Solving in Human Beings and Computers.  
*Annual Meeting of the Society for Mathematical Psychology*, Irvine, CA, 1995.



Z. Pizlo, E. Stefanov, J. Saalweachter, Z. Li, Y. Haxhimusa and W. G. Kropatsch.  
 Traveling Salesman Problem: a Foveating Pyramid Model.  
*Journal of Problem Solving*, in press, 2006.



Borůvka, O. (1926).  
 O jistém problému minimálním (about a certain minimal problem).  
*Práce Moravské Přírodovědecké Společnosti v Brně (Acta Societ. Scienc. Natur. Moraviae)*, 3(3):37–58.



# References II



Kruskal, J. B. J. (1956).

On the shortest spanning subtree of a graph and the travelling salesman problem.  
*In Proc. Am. Math. Soc.*, volume 7, pages 48–50.



Prim, R. C. (1957).

Shortest connection networks and some generalizations.  
*The Bell System Technical Journal*, 36:1389–1401.



A. Aggrawal, L. J. Guibas, J. Saxe, and P.W. Shor.

A Linear-Time Algorithm for Computing the Voronoi Diagram of a Convex Polygon.  
*Discrete & Comput. Geometry*, 4(6):591–604, 1989.



C. H. Papadimitriou and S. Vempala.

On the approximability of the traveling salesman problem).  
*Proceedings of STOC'2000*, extended abstract, 2000.



N. Christofides.

Worst-case analysis of a new heuristic for the travelling salesman problem.  
 Graduate School of Industrial Administration, Carnegie-Mellon University, 1976, 388 .



D. Applegate, R. Bixby, V. Chvatal, and W. Cook

TSP cuts which do not conform to the template paradigm.  
*Computational combinatorial optimization: optimal or provably near optimal solutions*, Lecture Notes in Computer Science, 2241, M. Jünger and D. Naddef, eds., Springer, 2001, pp. 261–303.



Z. Pizlo, E. Stefanov, J. Saalweachter, Z. Li, Y. Haxhimusa and W. G. Kropatsch.

Adaptive Pyramid Model for the Traveling Salesman Problem.  
*Workshop on Human Problem Solving*, June 2005.

