



Carthagène



A brief introduction to combinatorial
optimization:
The Traveling Salesman Problem

Simon de Givry

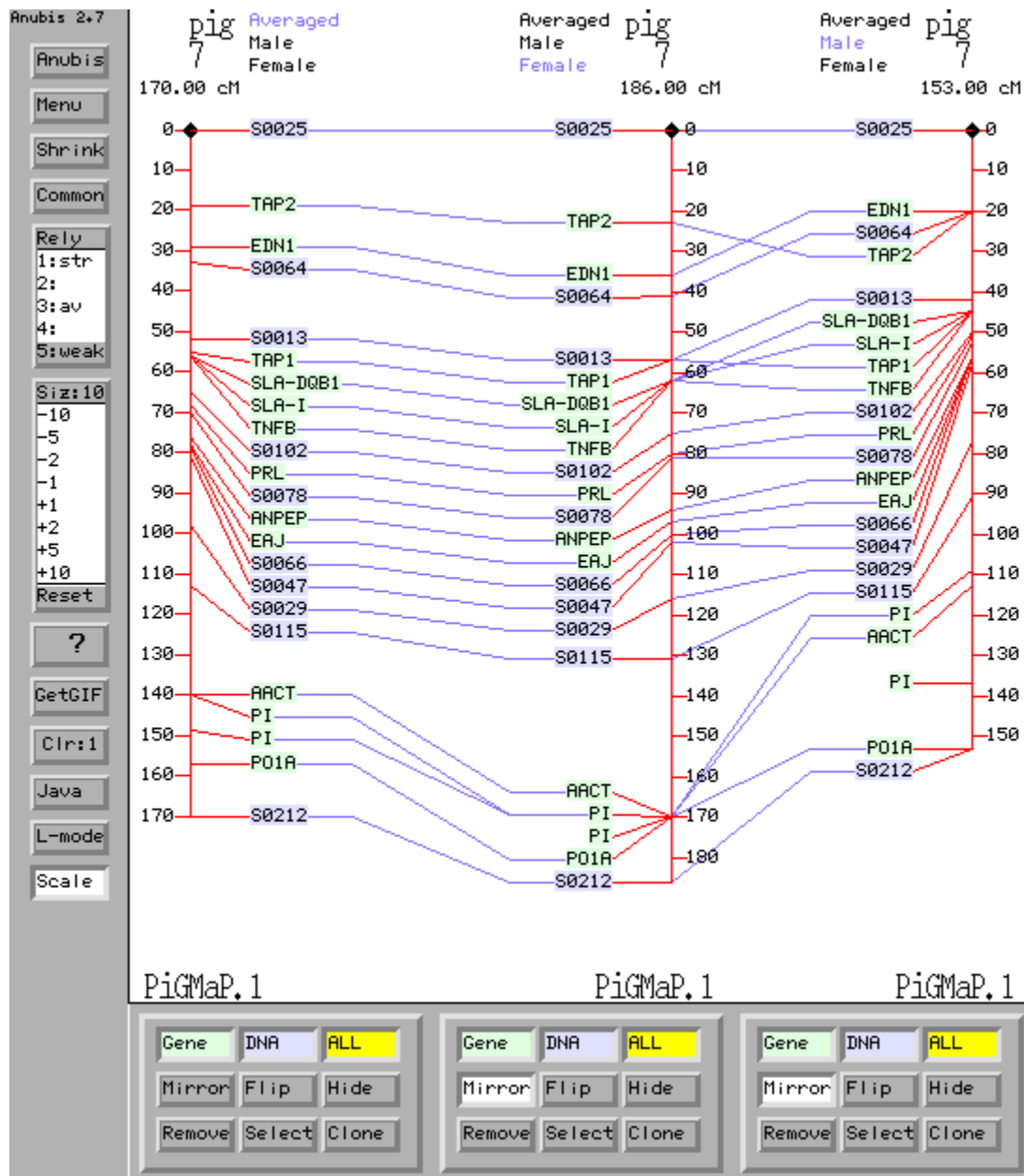


Find a tour with minimum distance, visiting every city only once



Distance matrix (miles)

Distances	Camara	Canico	Funchal	...
Camara	0	15	7	...
Canico	15	0	8	...
Funchal	7	8	0	...
...



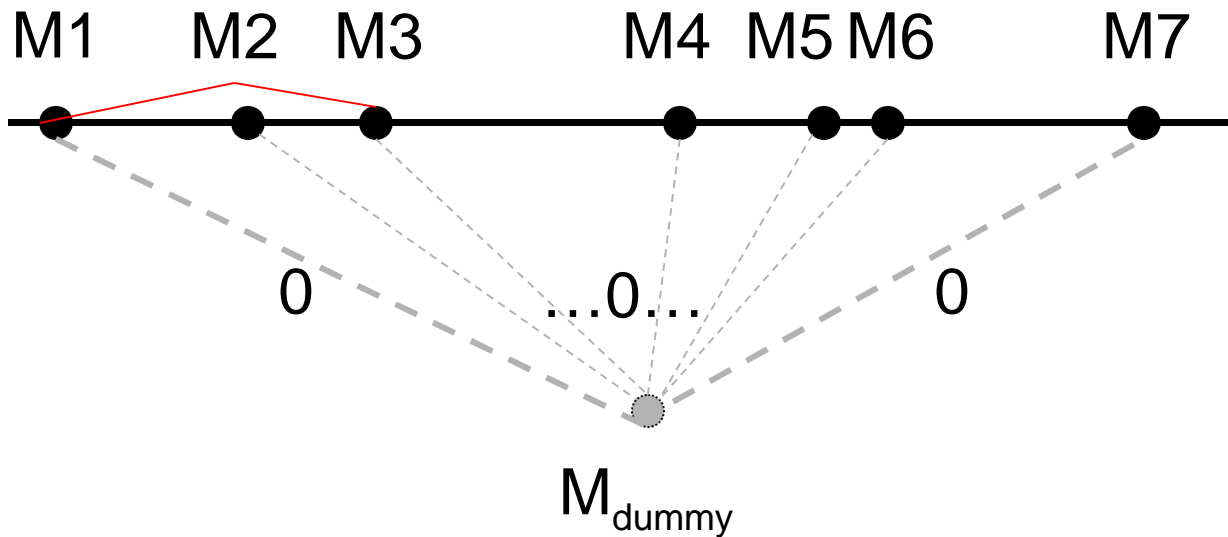
Find an order of all the markers with maximum likelihood



2-point distance matrix (Haldane)

Distances	M1	M2	M3	...
M1	0	14	7	...
M2	14	0	8	...
M3	7	8	0	...
...

Link



$$\forall j, k \quad \text{distance}(i, j) \stackrel{?, \leq}{=} \text{distance}(i, k) + \text{distance}(k, j)$$

Multi-point likelihood (with unknowns) \Leftrightarrow
the distance between two markers depends on the order



Traveling Salesman Problem

Complete graph

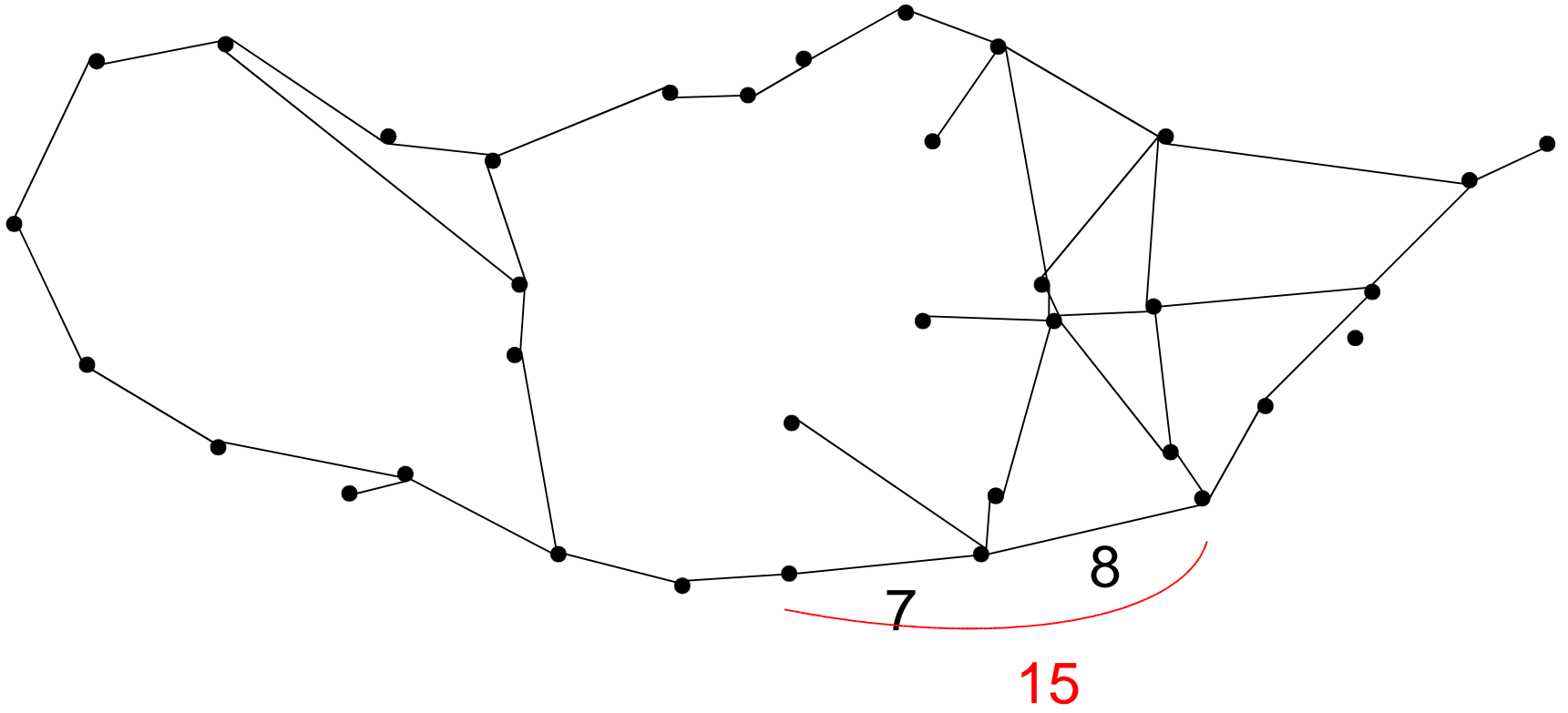
Positive weight on every edge

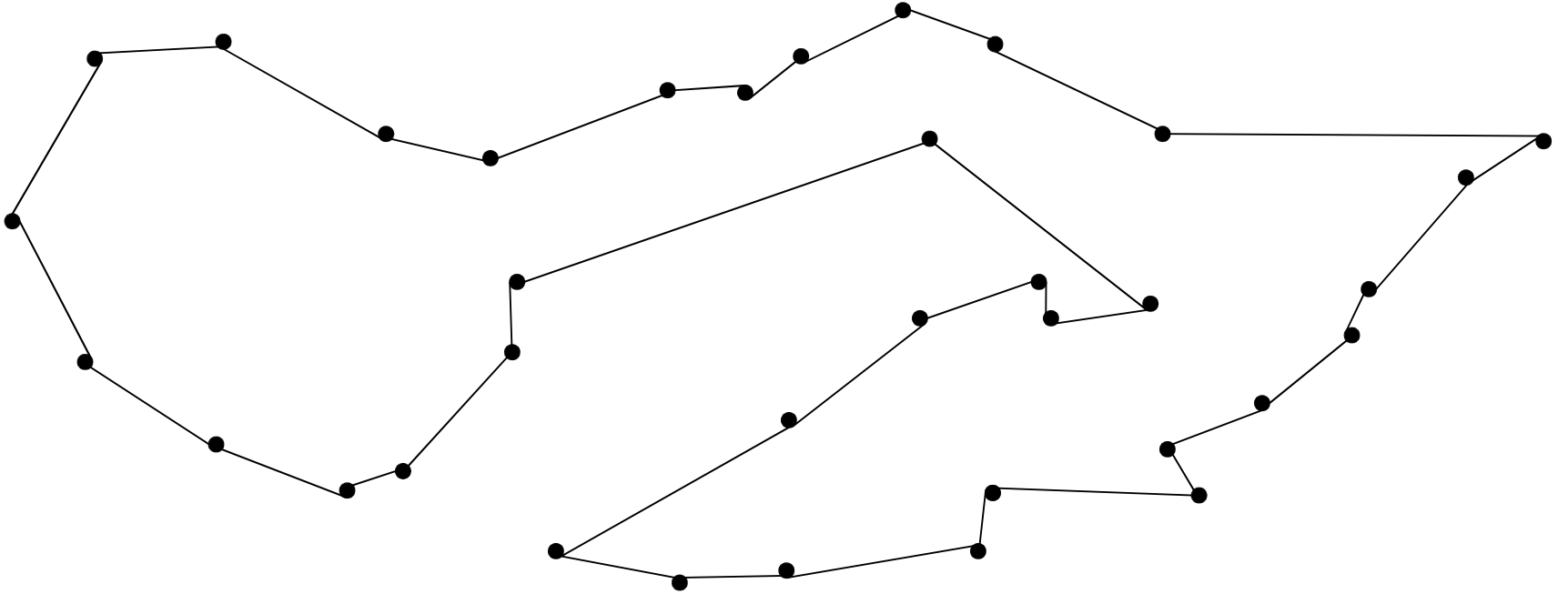
Symmetric case: $\text{dist}(i,j) = \text{dist}(j,i)$

Triangular inequality: $\text{dist}(i,j) \leq \text{dist}(i,k) + \text{dist}(k,j)$

Euclidean distance

Find the shortest Hamiltonian cycle





Total distance = xxx miles



Traveling Salesman Problem

Theoretical interest

NP-complete problem

1993-2001: +150 articles about TSP in
INFORMS & Decision Sciences databases

Practical interest

Vehicle Routing Problem

Genetic/Radiated Hybrid Mapping Problem

NCBI/Concorde, Carthagène, ...



Variants

Euclidean Traveling Salesman Selection Problem

Asymmetric Traveling Salesman Problem

Symmetric Wandering Salesman Problem

Selective Traveling Salesman Problem

TSP with distances 1 and 2, TSP(1,2)

K-template Traveling Salesman Problem

Circulant Traveling Salesman Problem

On-line Traveling Salesman Problem

Time-dependent TSP

The Angular-Metric Traveling Salesman Problem

Maximum Latency TSP

Minimum Latency Problem

Max TSP

Traveling Preacher Problem

Bipartite TSP

Remote TSP

Precedence-Constrained TSP

Exact TSP

The Tour Cover problem



Plan

Introduction to TSP

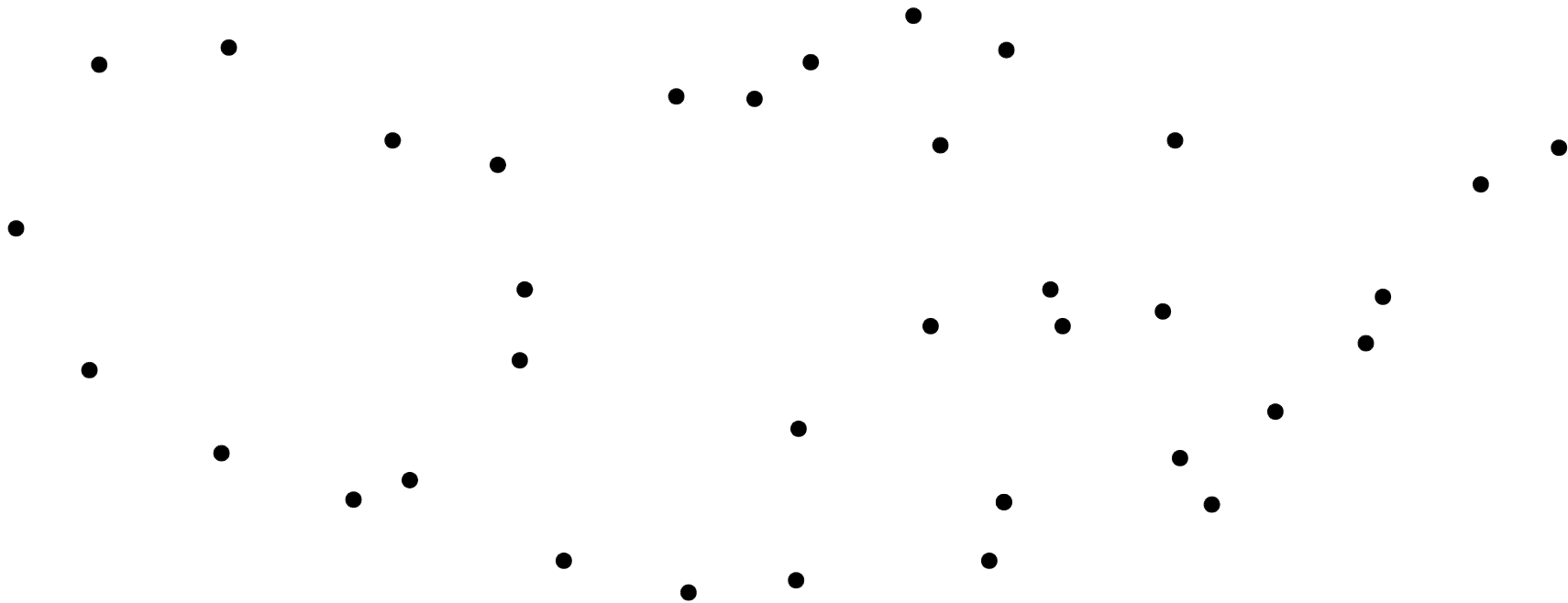
Building a new tour

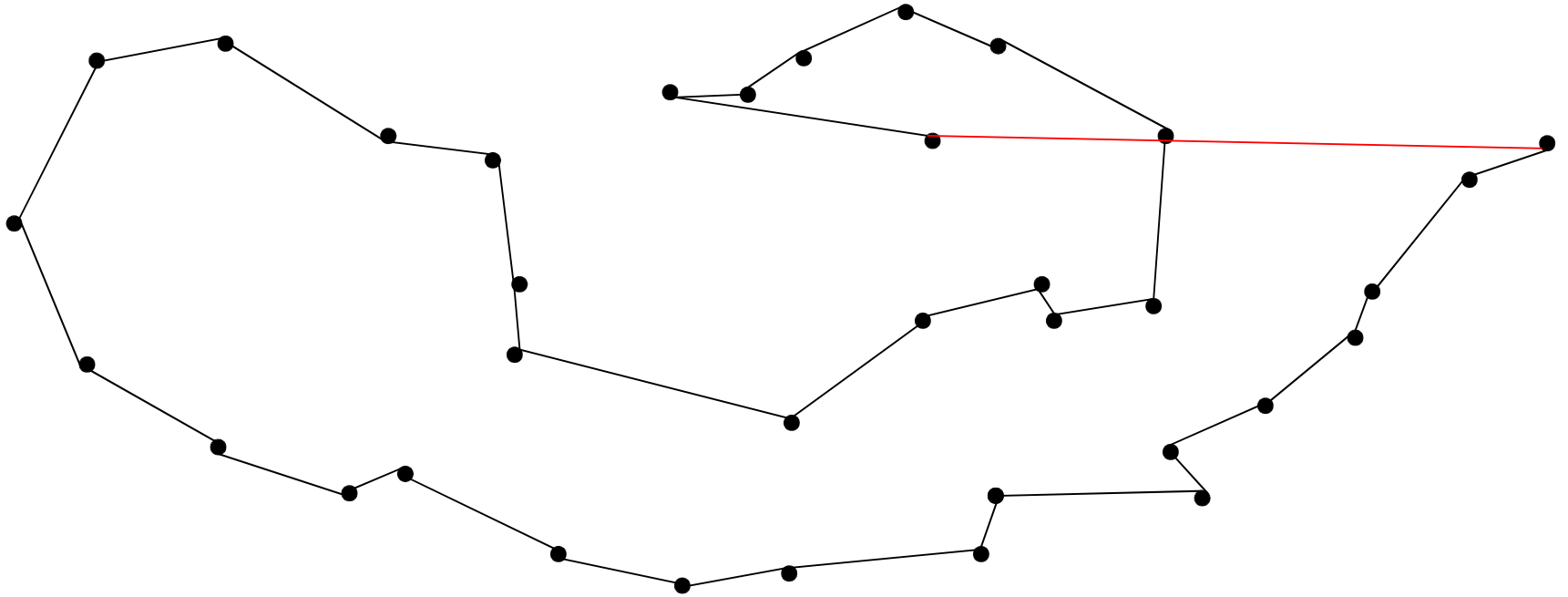
Improving an existing tour

Finding the best tour

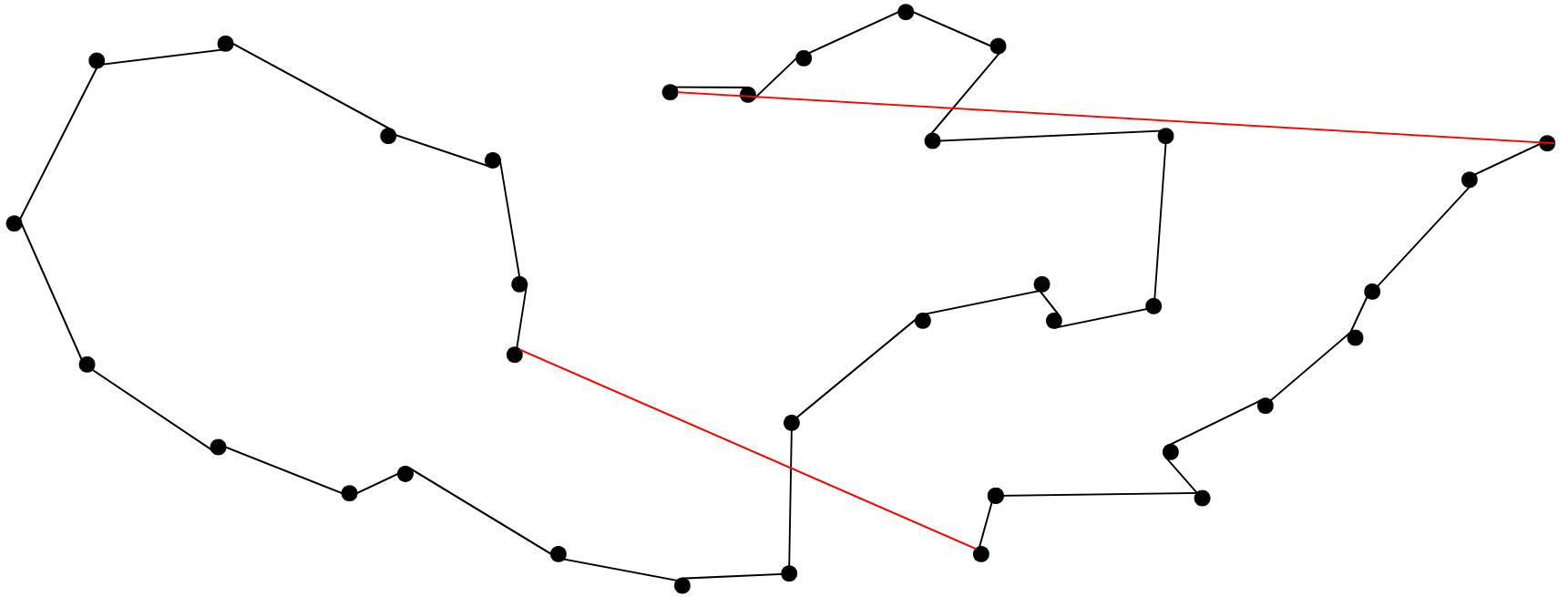


Building a new tour

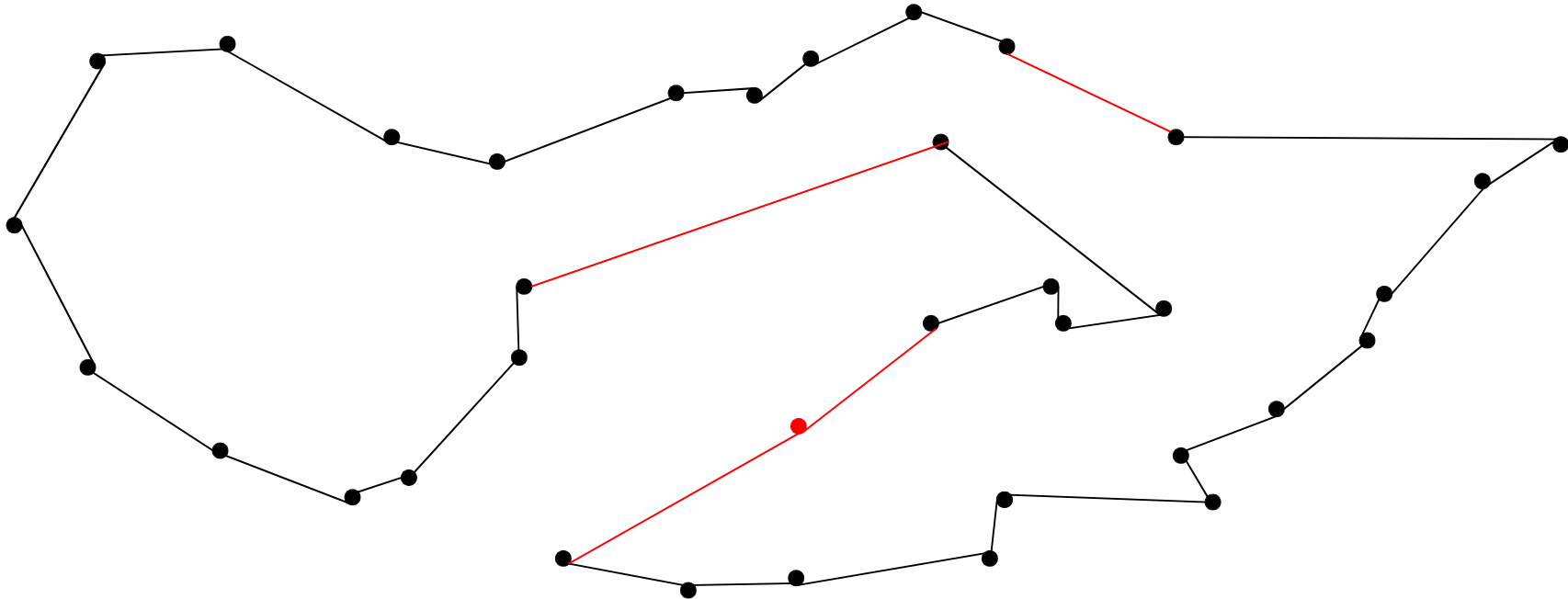




Nearest Neighbor heuristic



Greedy (or multi-fragments) heuristic



Savings heuristic (Clarke-Wright 1964)



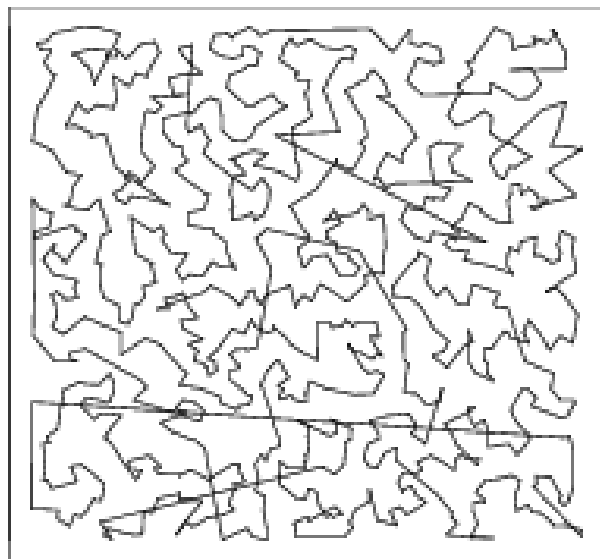
Heuristics

Mean distance to the optimum

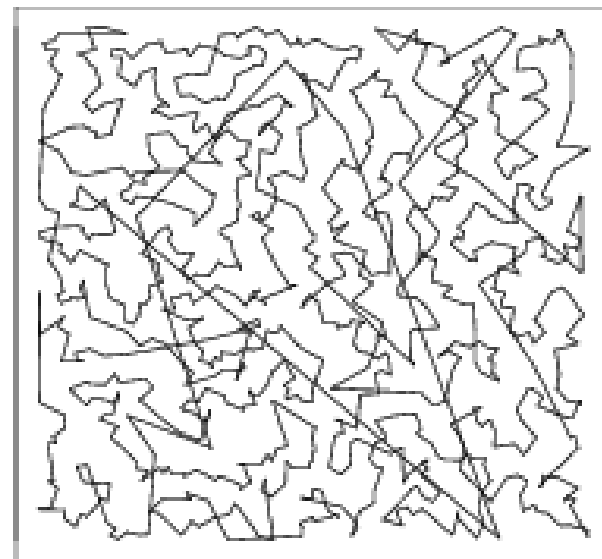
Savings: 11%

Greedy: 12%

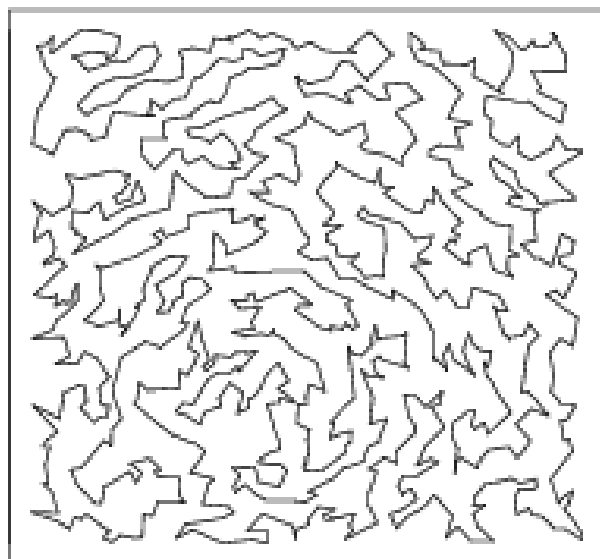
Nearest Neighbor: 26%



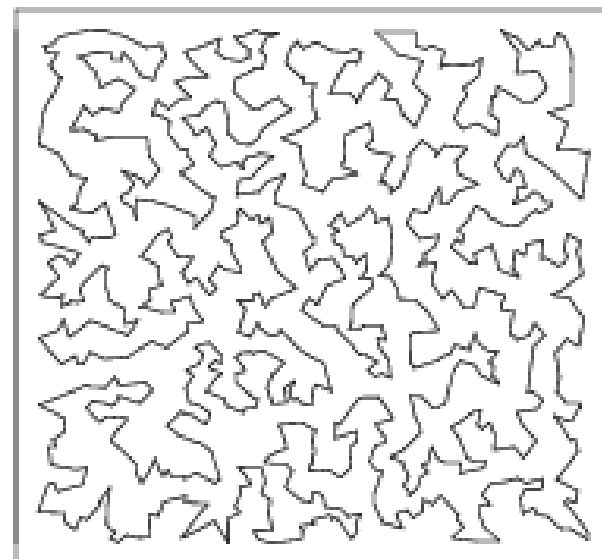
Greedy Tour



Nearest Neighbor Tour



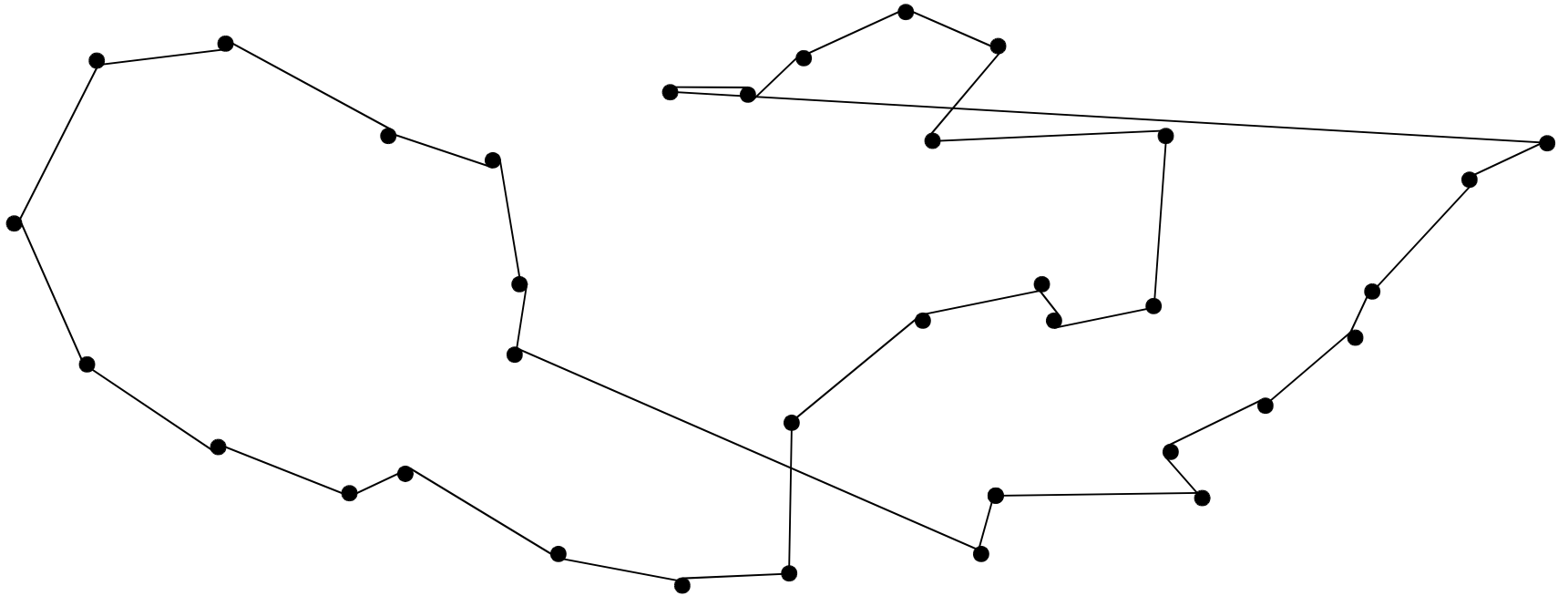
Savings Tour



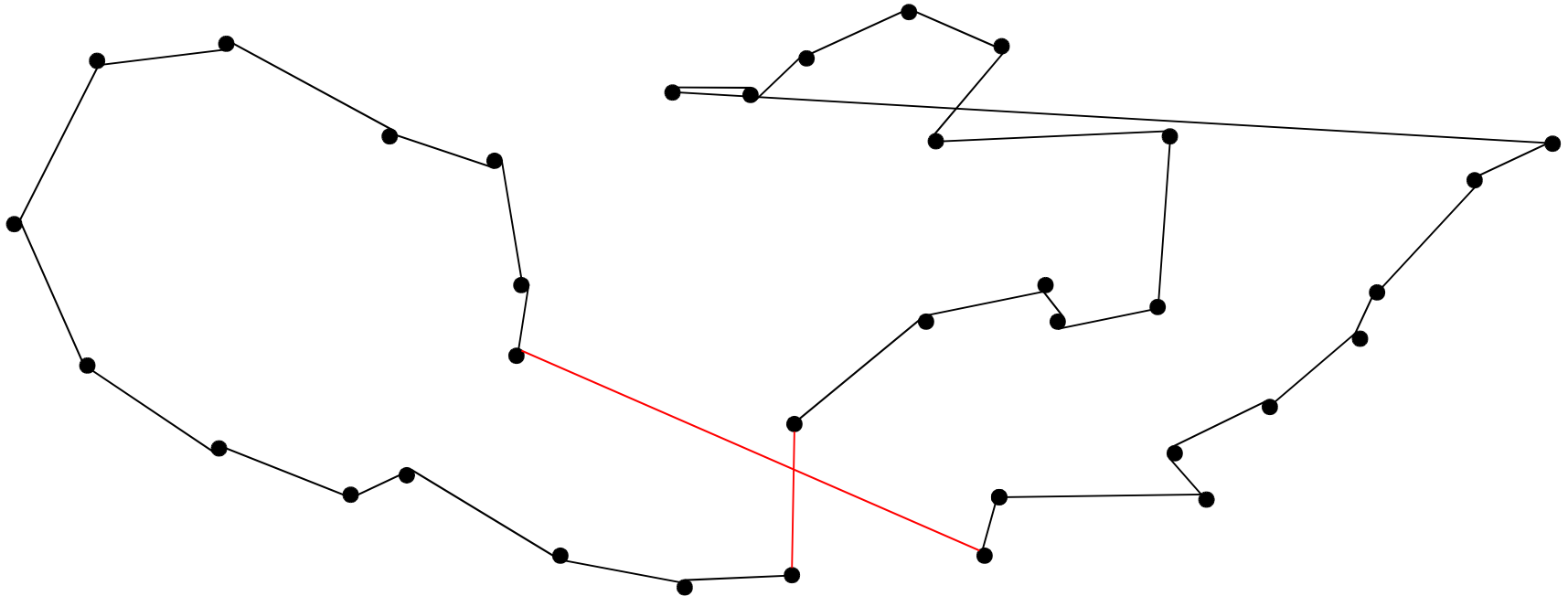
Optimal Tour



Improving an existing tour



Which local modification can improve this tour?

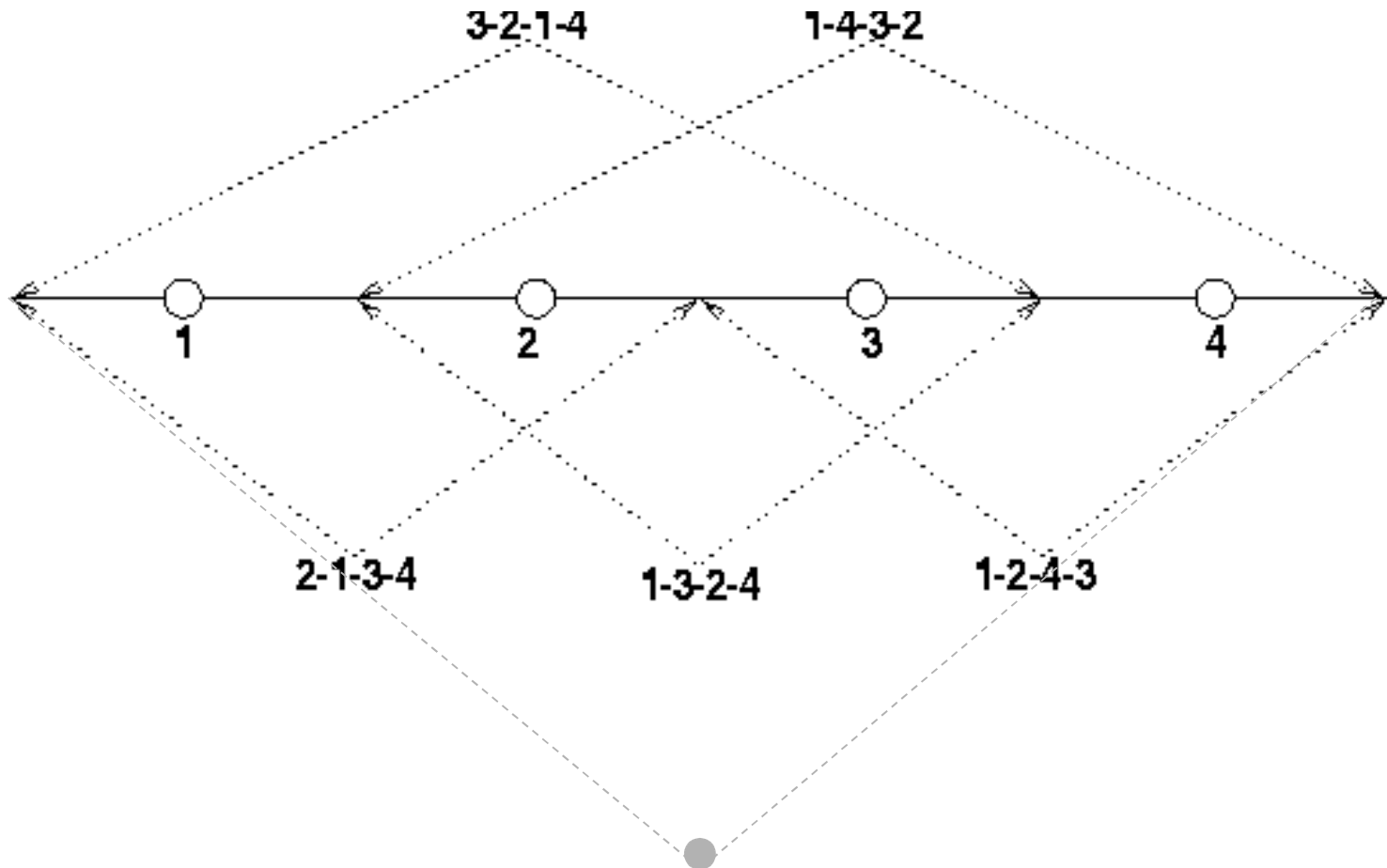


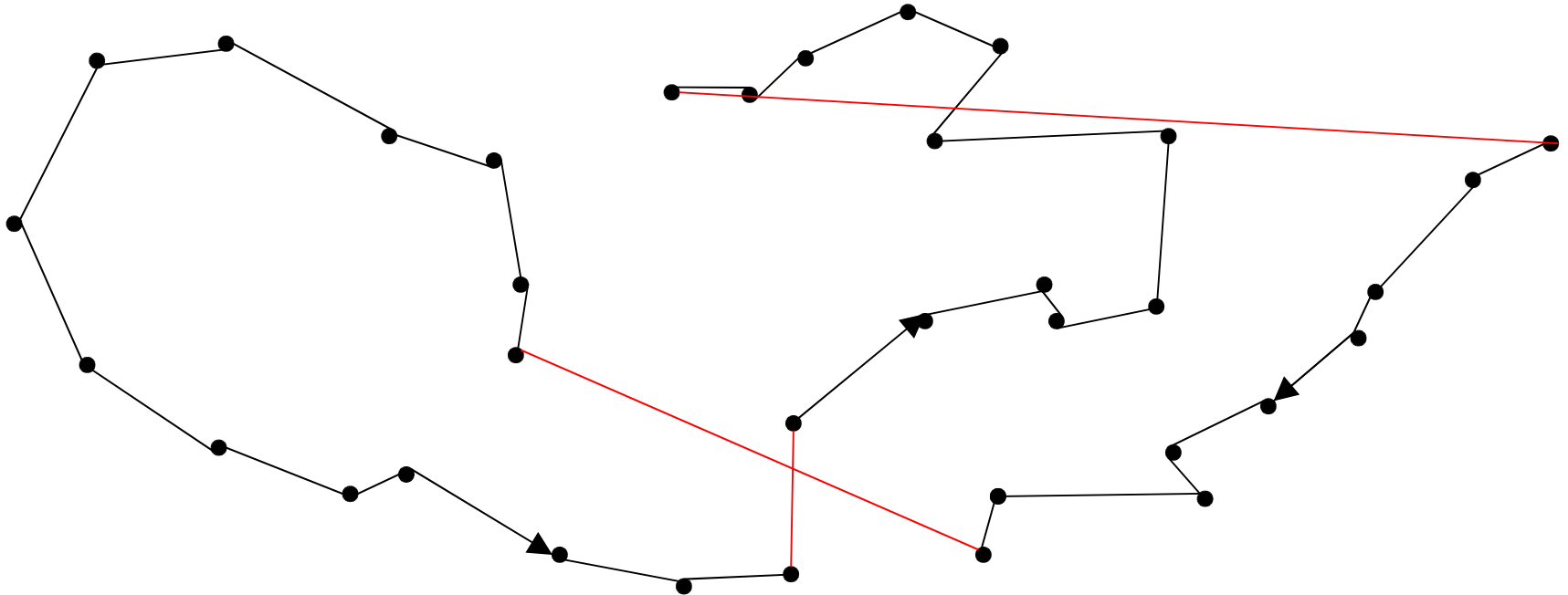
Remove two edges and rebuild another tour

⇔ Invert a given sequence of markers



2-change





Remove three edges and rebuild another tour (7 possibilities)

⇒ Swap the order of two sequences of markers



« greedy » local search

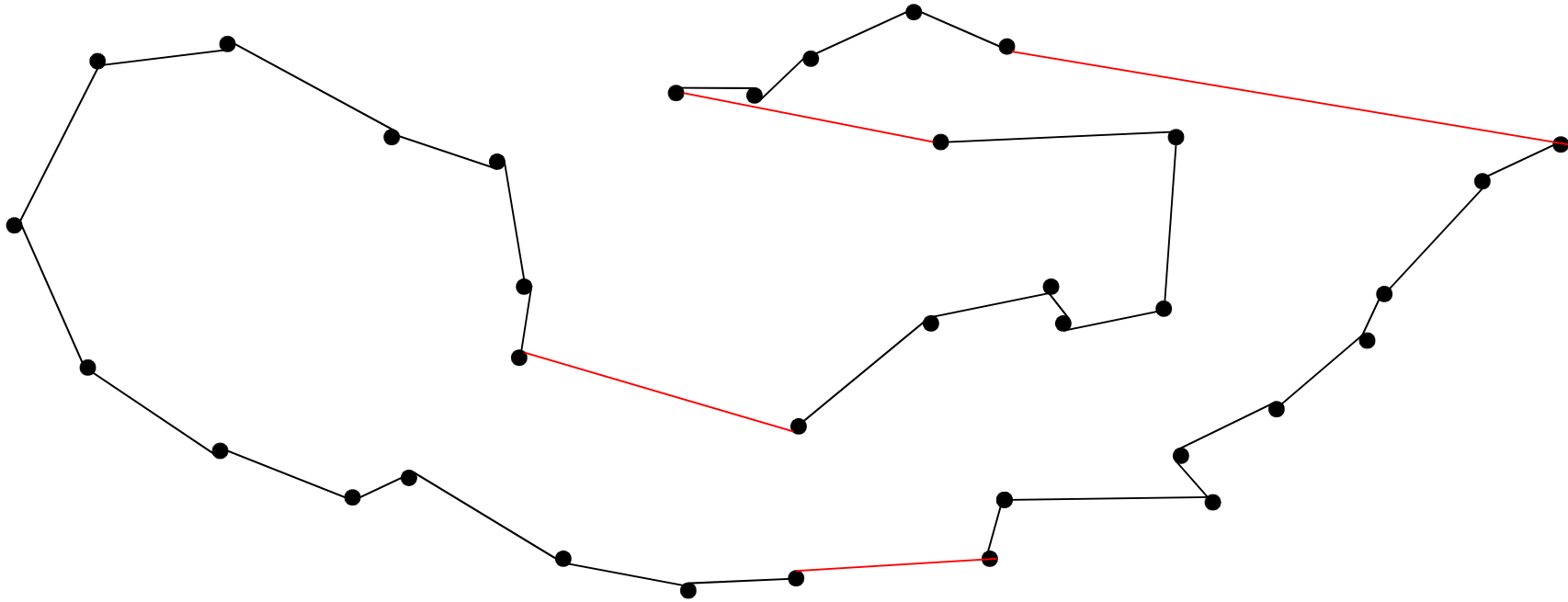
2-opt

Note: a finite sequence of « 2-change » can reach any tour, including the optimum tour

Strategy:

Select the best 2-change among $N*(N-1)/2$ neighbors (2-move neighborhood)

Repeat this process until a fix point is reached (i.e. no tour improvement was made)



2-opt



Greedy local search

Mean distance to the optimum

2-opt : 9%

3-opt : 4%

LK (limited k-opt) : 1%

Complexity

2-opt : $\sim N^3$

3-opt : $\sim N^4$

LK (limited k-opt) : $< N^4$?

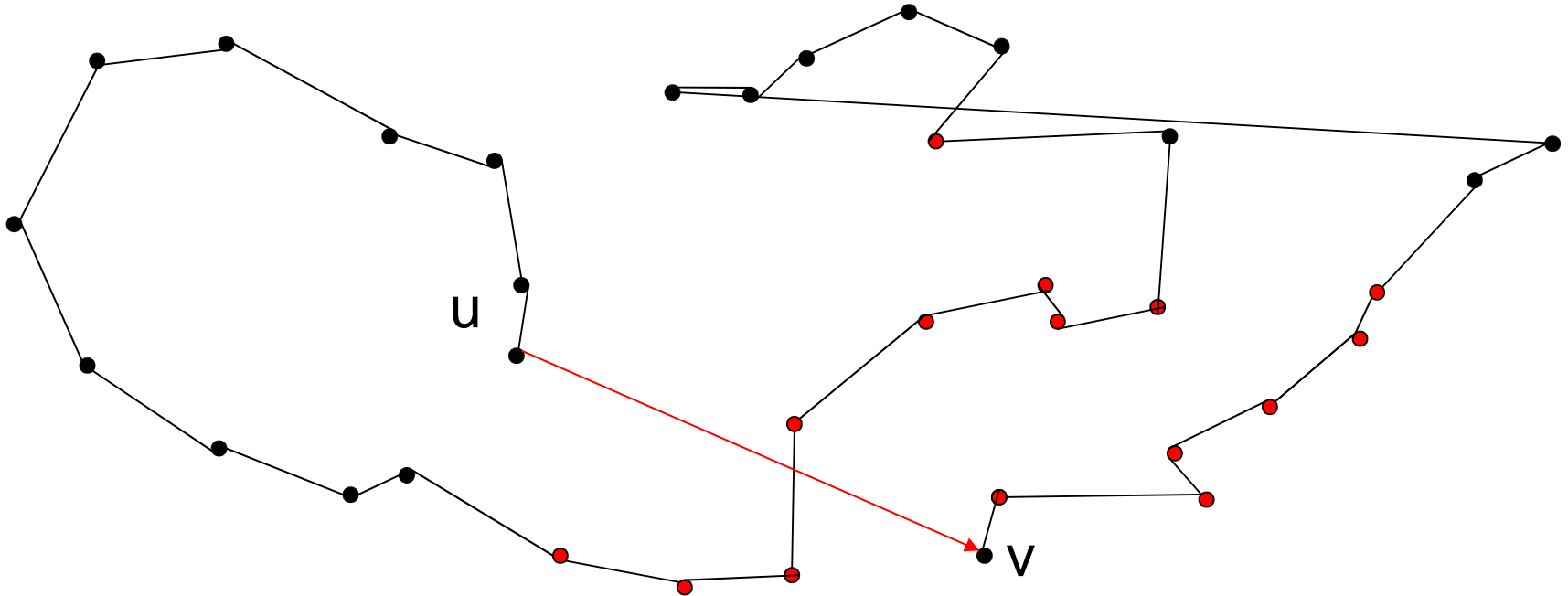


Complexity

n = number of vertices

Algorithm	Complexity
A-TSP	$(n-1)!$
S-TSP	$(n-1)! / 2$
2-change	1
3-change	7
k-change	$(k-1)! \cdot 2^{k-1}$
k-move	$(k-1)! \cdot 2^{k-1} \cdot n! / (k! \cdot (n-k)!)$
	$\sim O(n^k) \quad k \ll n$
In practice:	$o(n)$
2-opt et 3-opt	$\sim O(n^{k+1})$
In practice:	$o(n^{1.2})$
	$\text{time}(3\text{-opt}) \sim 3 \times \text{time}(2\text{-opt})$

2-opt implementation trick:



For each edge (uv) , maintain the list of vertices w such that $\text{dist}(w,v) < \text{dist}(u,v)$



Lin & Kernighan (1973)

k-change : $e_1 \rightarrow f_1, e_2 \rightarrow f_2, \dots$

$$\Rightarrow \text{Sum}_{i=1}^k (\text{dist}(e_i) - \text{dist}(f_i)) > 0$$

There is an order of i such that all the partial sums are positives:

$$S_l = \text{Sum}_{i=1}^l (\text{dist}(e_i) - \text{dist}(f_i)) > 0$$

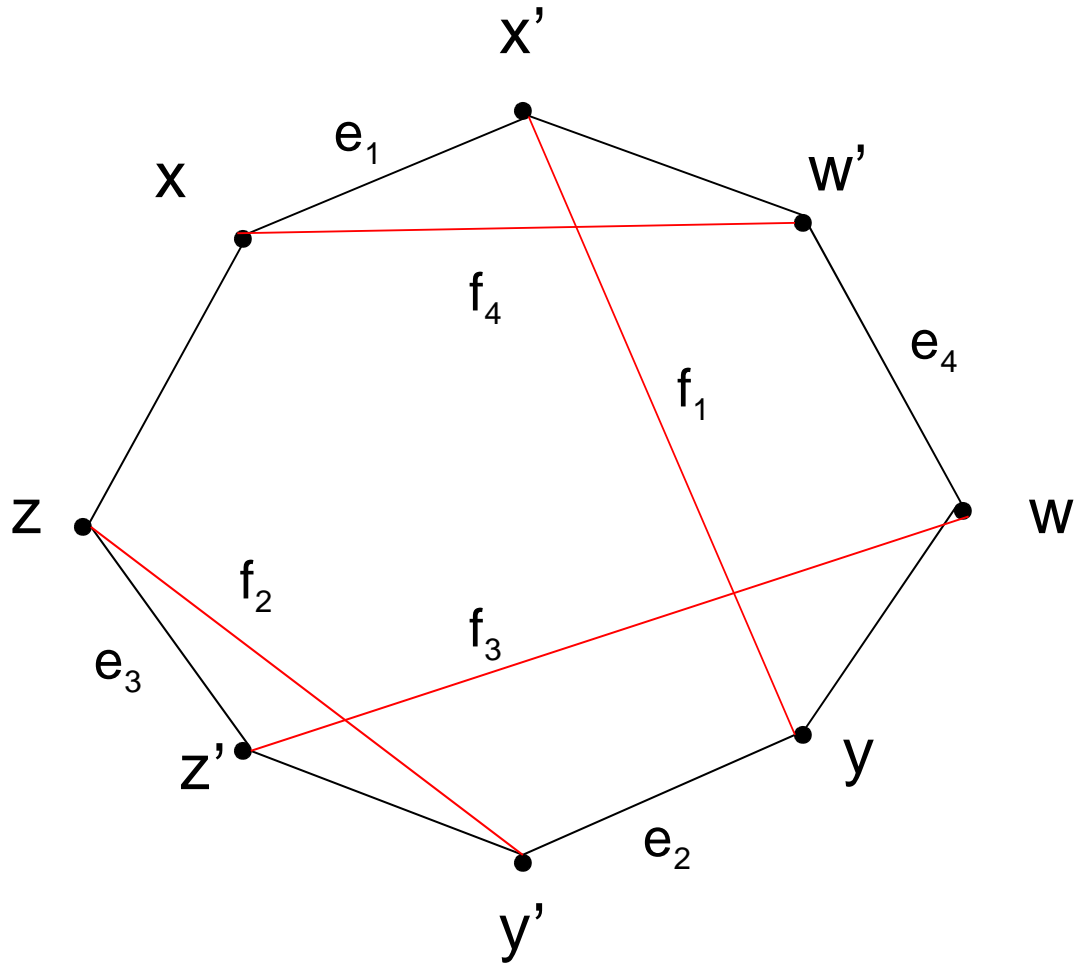
\Rightarrow Build a valid increasing alternate cycle:

$xx' \rightarrow yx', yy' \rightarrow zy', zz' \rightarrow wz', \dots$

$\text{dist}(f_1) < \text{dist}(e_1), \text{dist}(f_1) + \text{dist}(f_2) < \text{dist}(e_1) + \text{dist}(e_2), \dots$

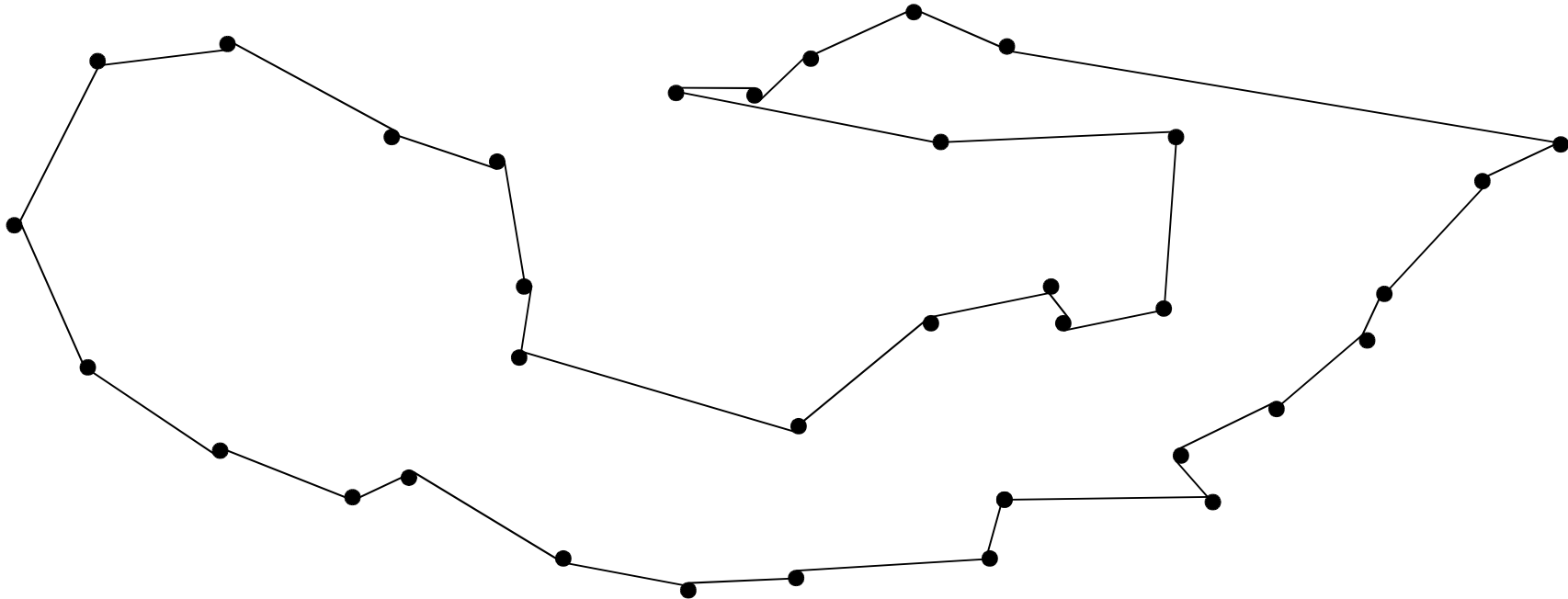
+ Backtrack on y and z choices + Restart

(in maximization)

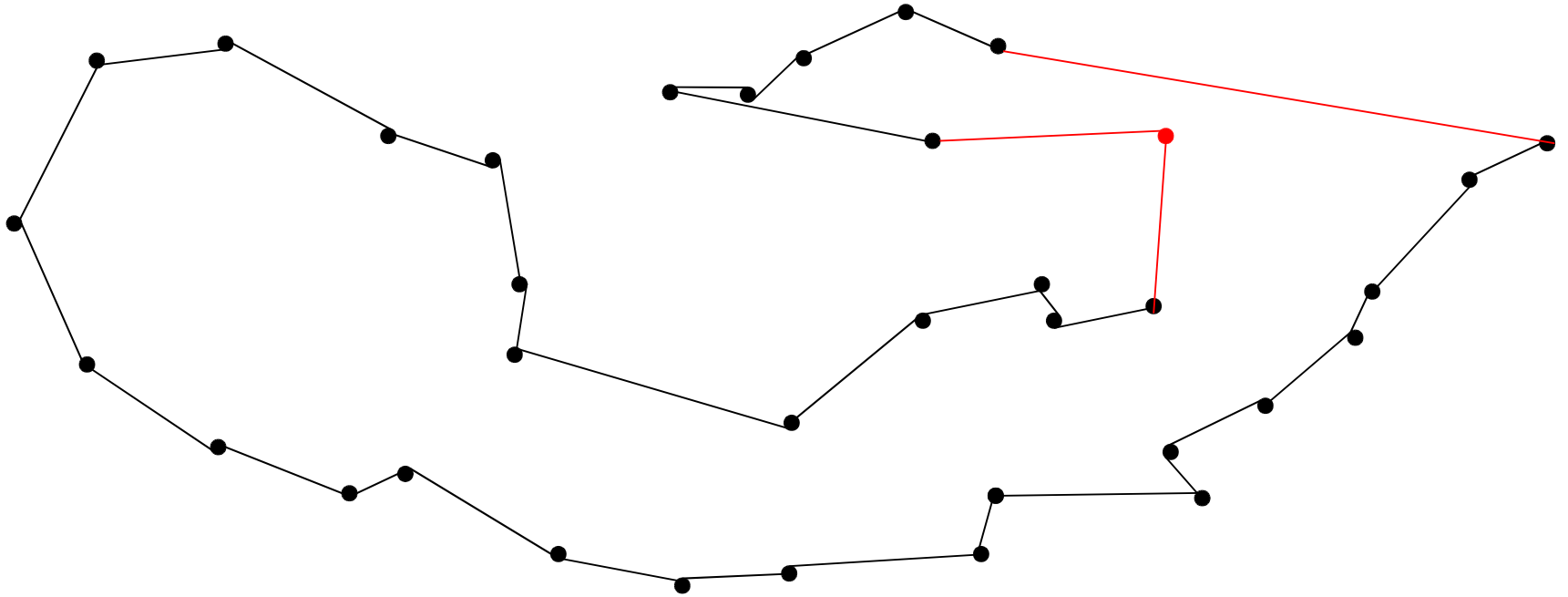


$$\{x, y, z, w, \dots\} \wedge \{x', y', z', w', \dots\} = 0$$

y is among the 5 best neighbors of x' , the same for z' and w



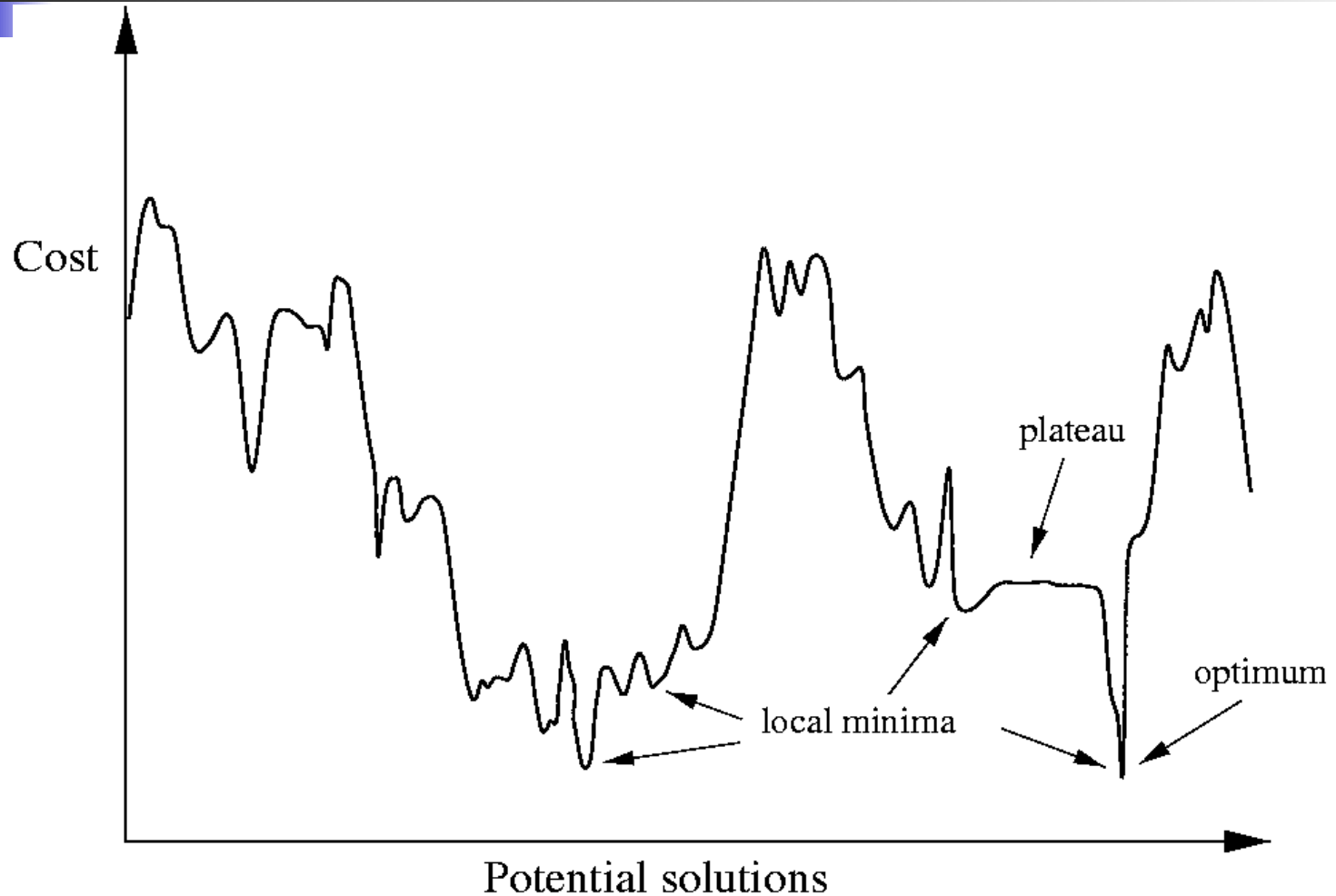
Is this 2-opt tour optimum?

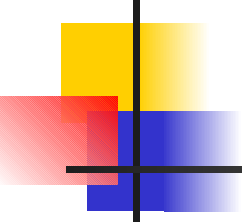


2-opt + vertex reinsertion



local versus global optimum





Local search & « meta-heuristics »

Tabu Search

Select the best neighbor even if it decreases the quality of the current tour

Forbid previous local moves during a certain period of time

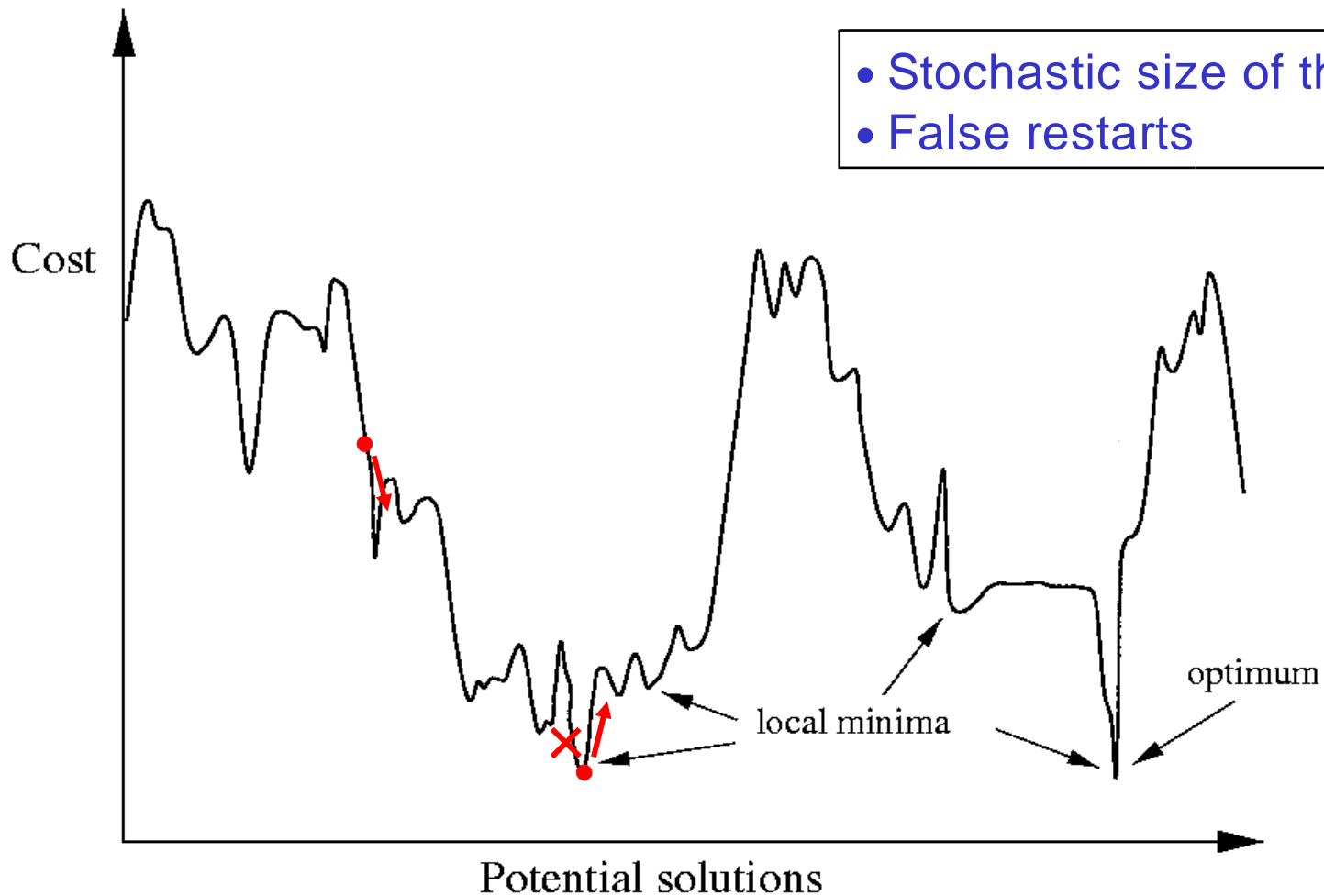
- List of tabu moves

Restart with new tours

- When the search goes to a tour already seen

- Build new tours in a random way

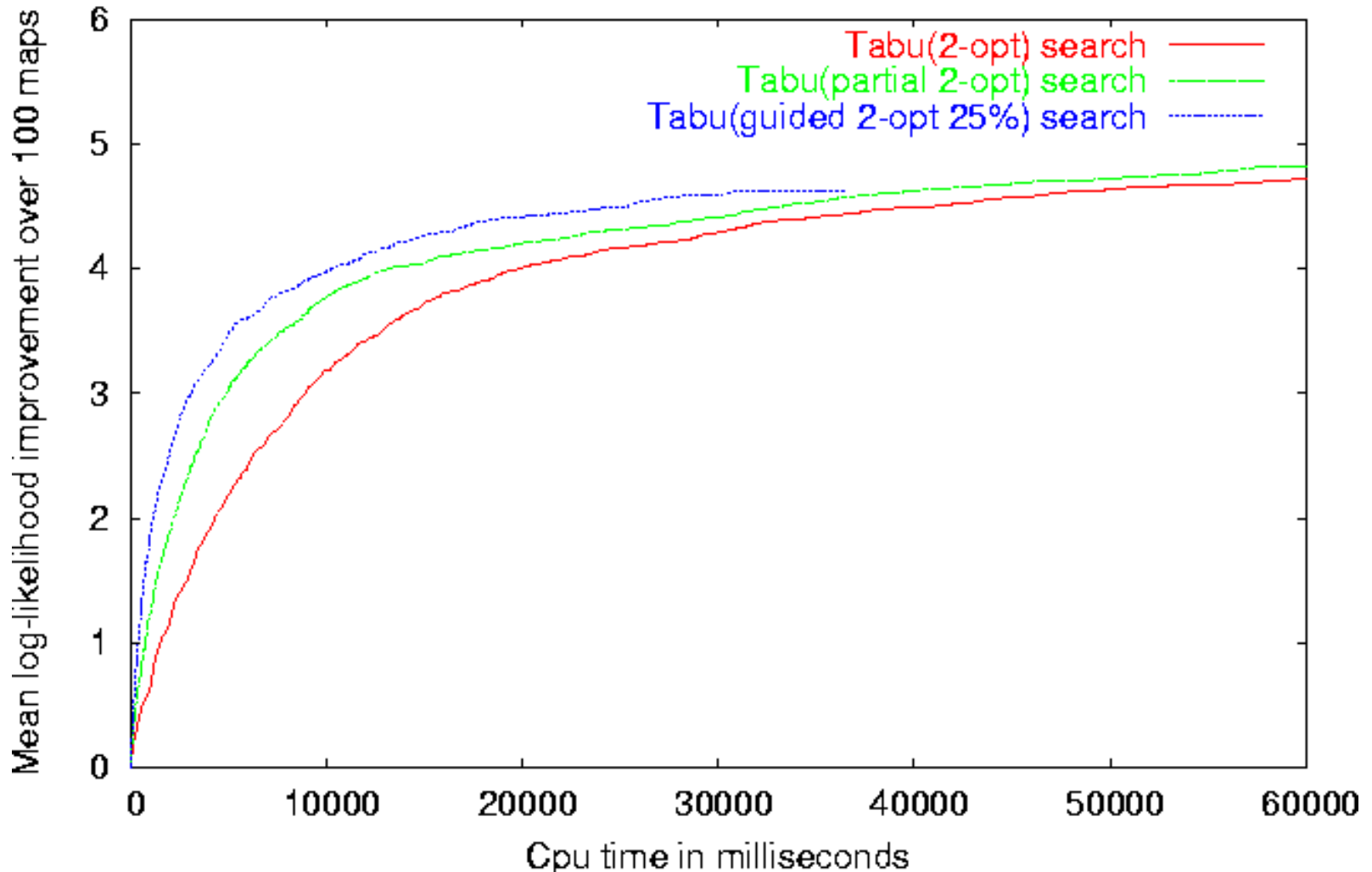
Tabu Search



Experiments with CarthaGène

N=50 K=100 Err=30% Abs=30%

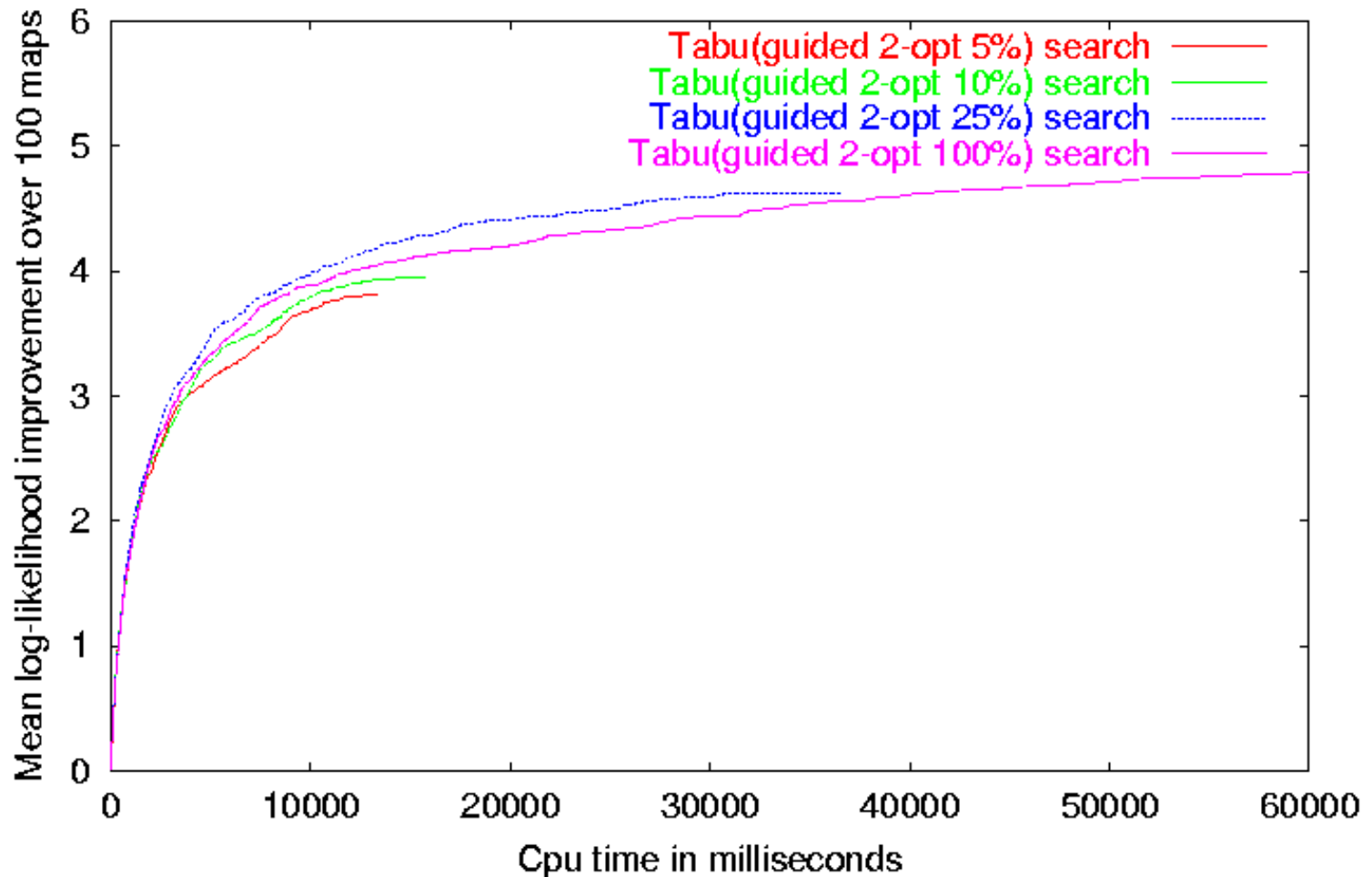
Log-likelihood improvement of genetic maps with 50 markers on simulated data



Legend: partial 2-opt = early stop , guided 2-opt 25% = early stop & sort with X = 25%

Experiments - next

Log-likelihood improvement of genetic maps with 50 markers on simulated data





Other meta-heuristics

Simulated Annealing

- Local moves are randomly chosen

- Neighbor acceptance depends on its quality

- Acceptance process is more and more greedy

Genetic Algorithms

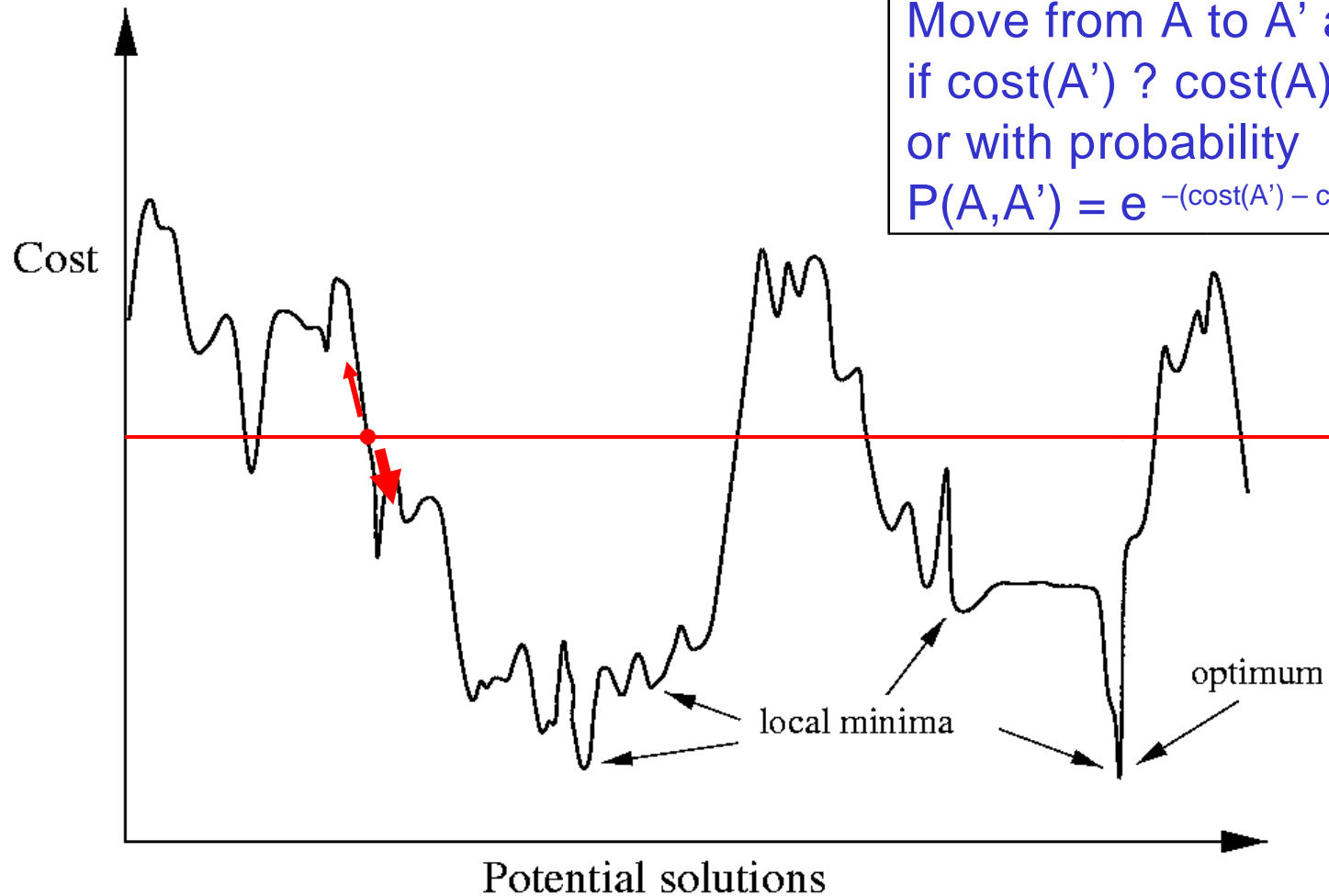
- Population of solutions (tours)

- Mutation, crossover,...

Variable Neighborhood Search

...

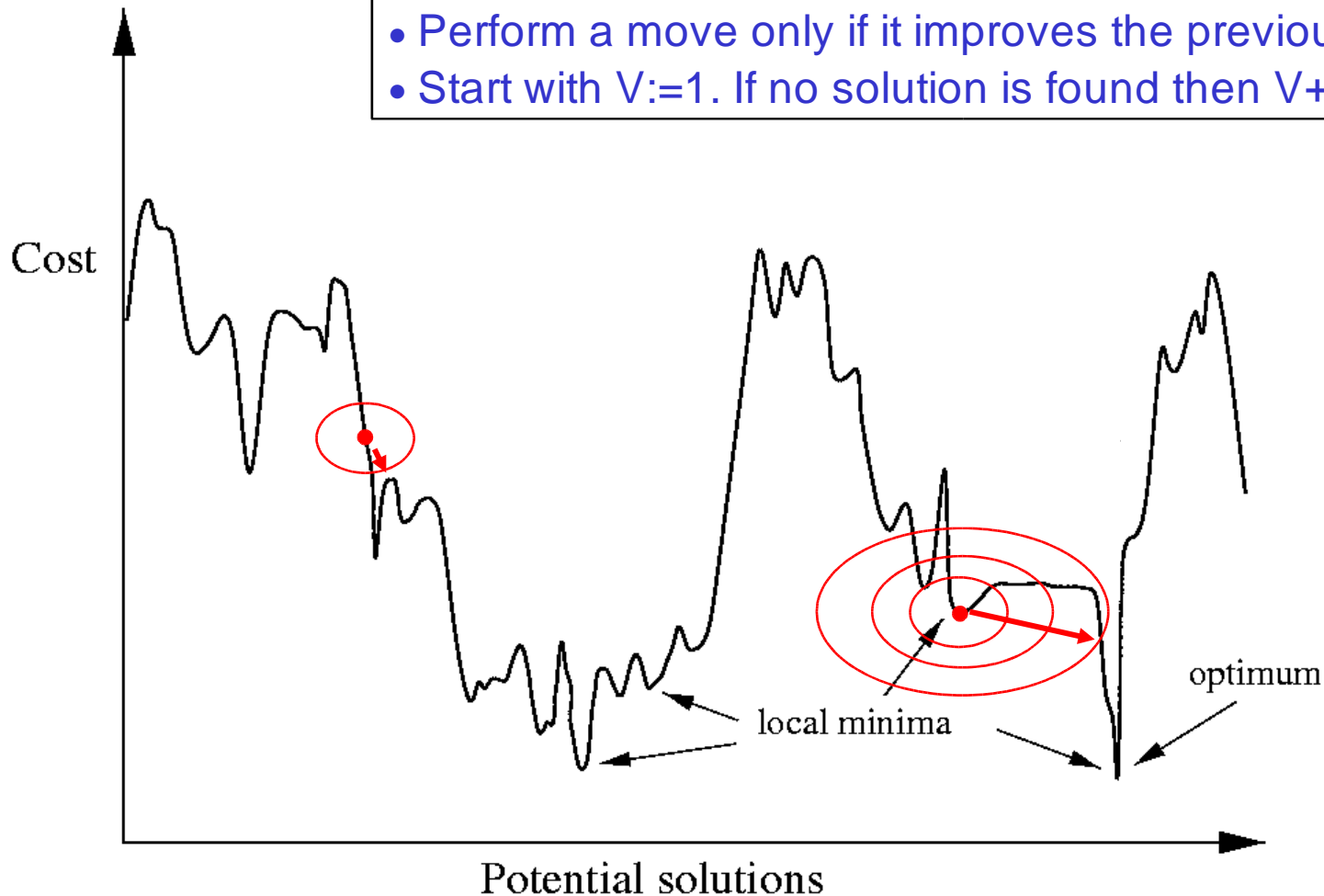
Simulated Annealing



Move from A to A' accepted
if $\text{cost}(A') \leq \text{cost}(A)$
or with probability
 $P(A, A') = e^{-(\text{cost}(A') - \text{cost}(A))/T}$

Variable Neighborhood Search

- Perform a move only if it improves the previous solution
- Start with $V:=1$. If no solution is found then $V++$ else $V:=1$





Local Search

Demonstration

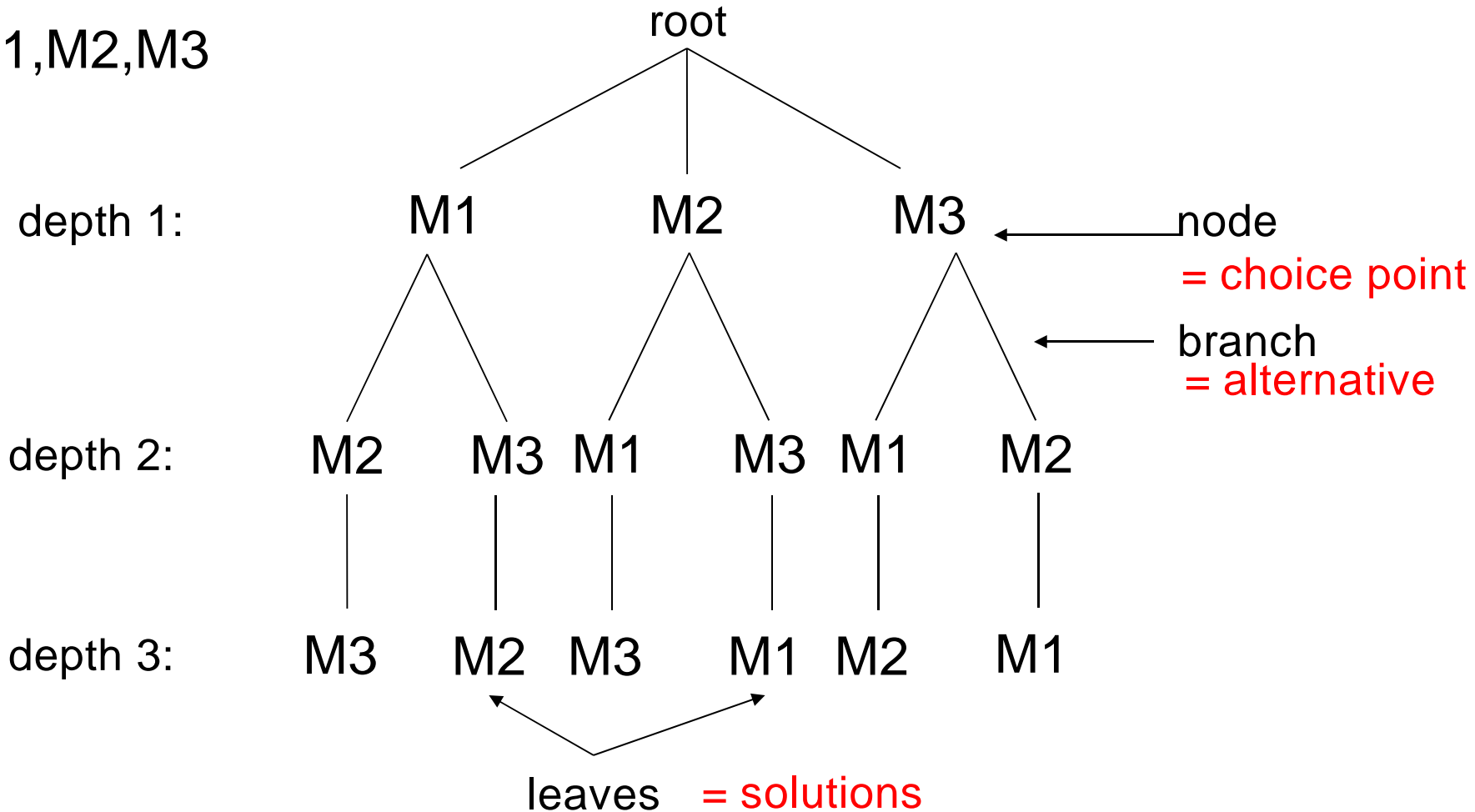


Finding the best tour

Search tree



M1, M2, M3





Tree search

Complexity : $n!/2$ different orders

Avoid symmetric orders (first half of the tree)

Can use *heuristics* in choice points to order possible alternatives

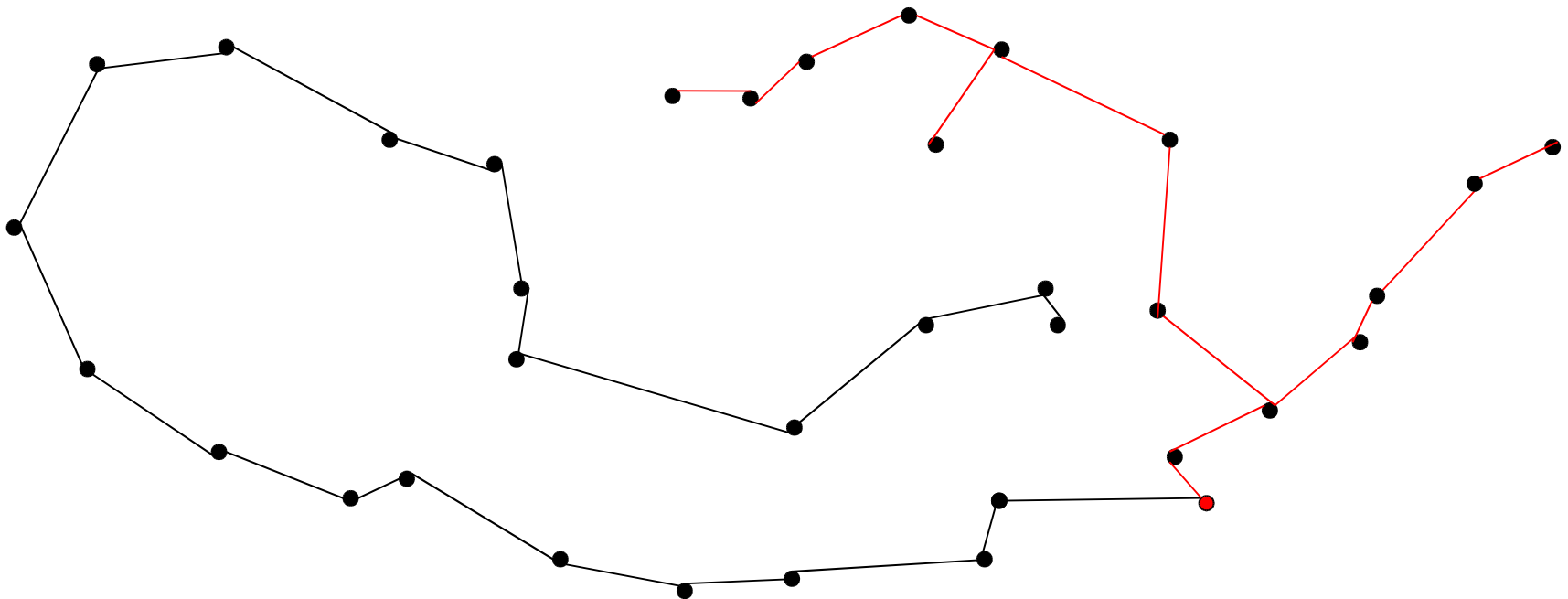
Branch and bound algorithm

Cut all the branches which cannot lead to a better solution

Possible to combine local search and tree search

Branch and bound

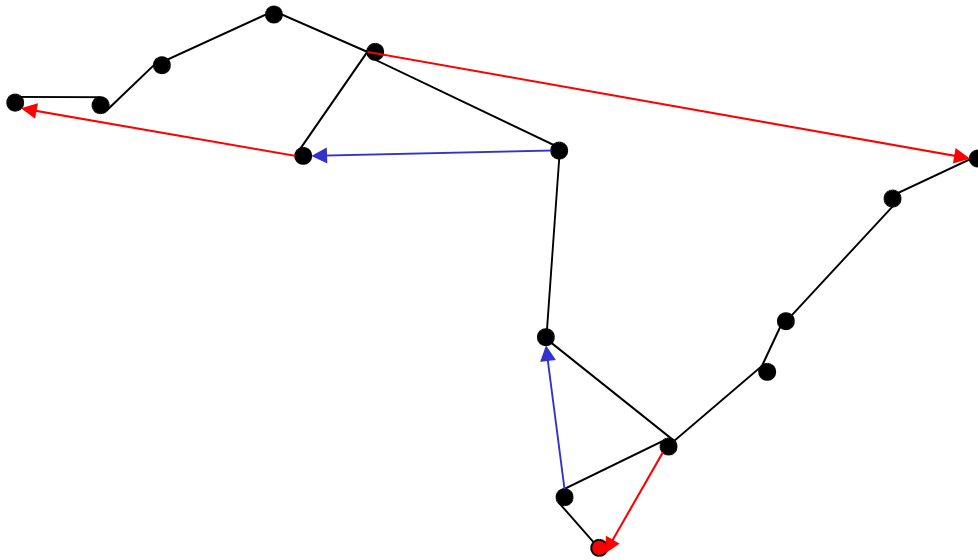
Minimum weight spanning tree



Prim algorithm (1957)

Held & Karp algorithm (better spanning trees) (1971)

\leq linear programming relaxation of TSP, $LB(I)/OPT(I) \geq 2/3$



Christofides heuristic (1976)

$\Rightarrow A(I) / \text{OPT}(I) \leq 3/2$ (with triangular inequalities)



Complexity

Complexity	Standard computer	Computer 100 times faster	Computer 1000 times faster
N	$N1$	$100*N1$	$1000*N1$
N^2	$N2$	$10*N2$	$31,6*N2$
N^3	$N3$	$4,64*N3$	$10*N3$
2^N	$N4$	$N4+6,64$	$N4+9,97$
3^N	$N5$	$N5+4,19$	$N5+6,29$



Complete methods

1954 : 49 cities

1971 : 64 cities

1975 : 100 cities

1977 : 120 cities

1980 : 318 cities

1987 : 2,392 cities

1994 : 7.397 cities

1998 : 13.509 cities

2001 : 15.112 cities (585936700 sec. \approx 19 years of CPU!)