

The traveling salesman problem

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Given finitely many “cities” along with the cost of travel between any two of them,

find the cheapest way of going through all the cities and coming back to the city you started out from.

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The *symmetric* TSP:

Travel from A to B costs as much as travel from B to A.

What are the origins of the problem?

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Der
Handlungsreisende
wie er sein soll

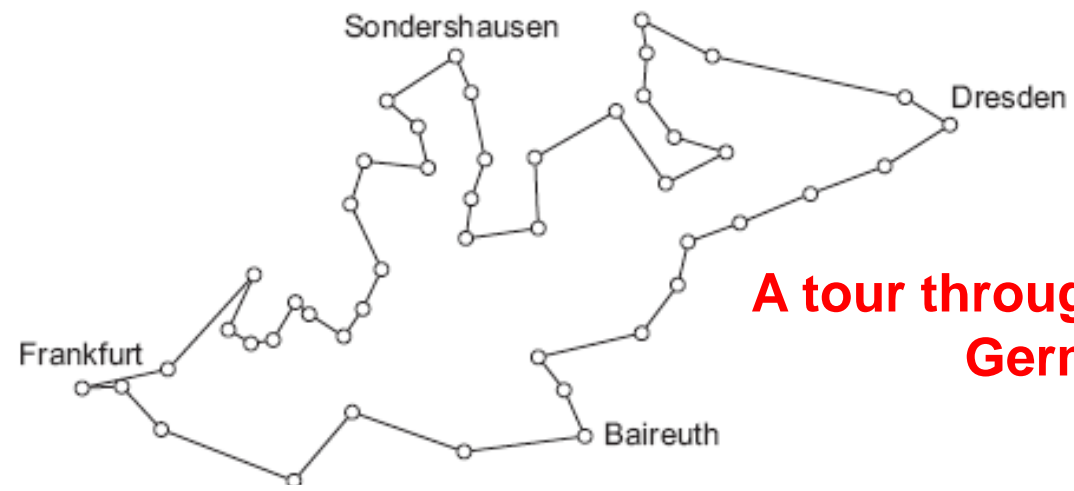
und was er zu thun hat, um Aufträge
zu erhalten und eines glücklichen Erfolgs
in seinen Geschäften gewiß zu sein.

Von
einem alten Commis - Voyageur.

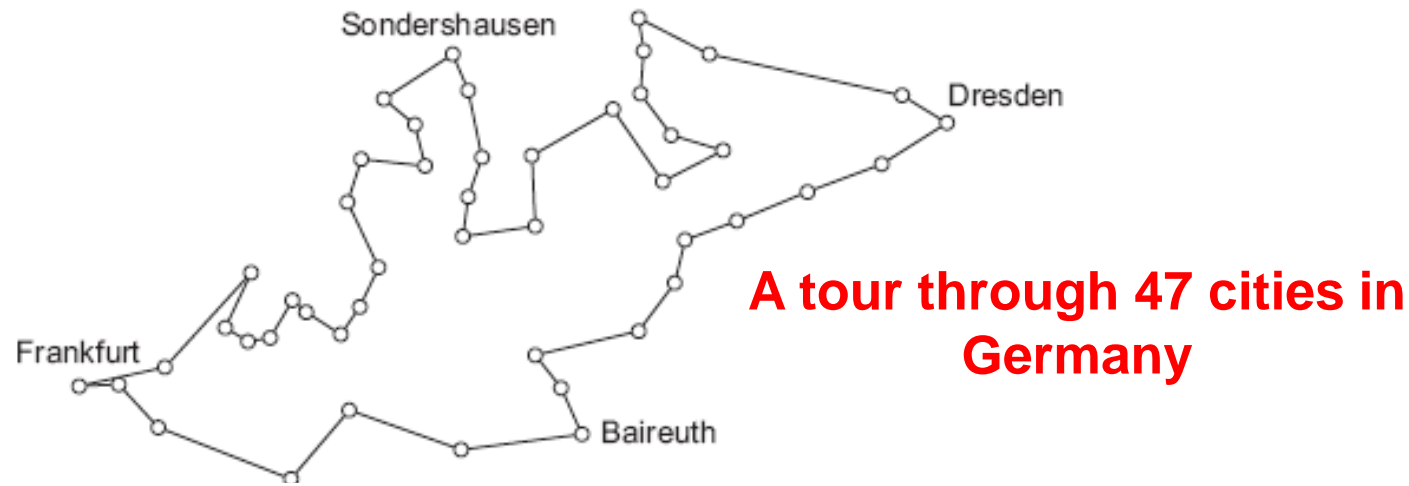


Mit einem Titellupfer.

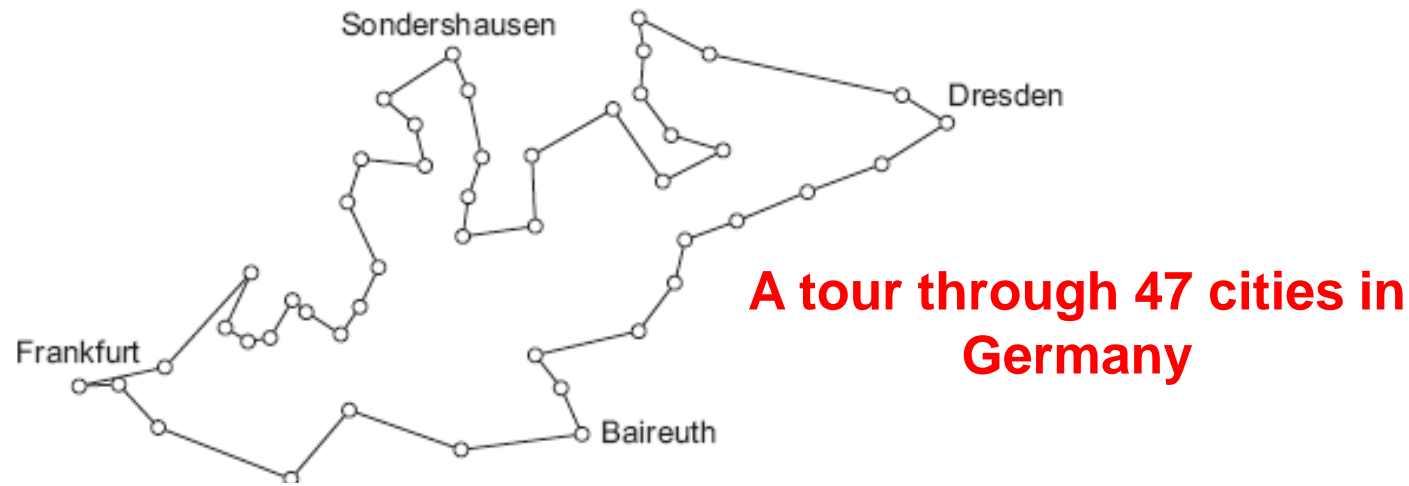
Stmenau 1832,
Druck und Verlag von W. Fr. Voigt.



**A tour through 47 cities in
Germany**



(...) Business leads the traveling salesman here and there, and there is not a good tour for all occurring cases; but **through an expedient choice and division of the tour so much time can be won** that we feel compelled to give guidelines about this. Everyone should use as much of the advice as he thinks useful for his application. We believe we can ensure as much that it will not be possible to plan the tours through Germany in consideration of the distances and the traveling back and forth, which deserves the traveler's special attention, with more economy.

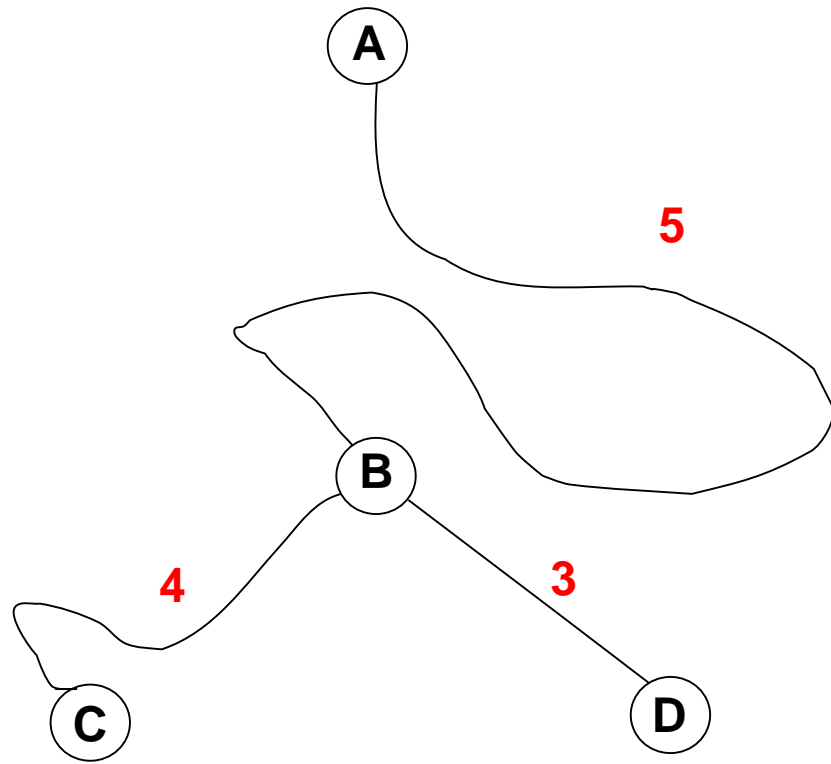


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The main thing to remember is always to visit as many localities as possible without having to touch them twice. (...)

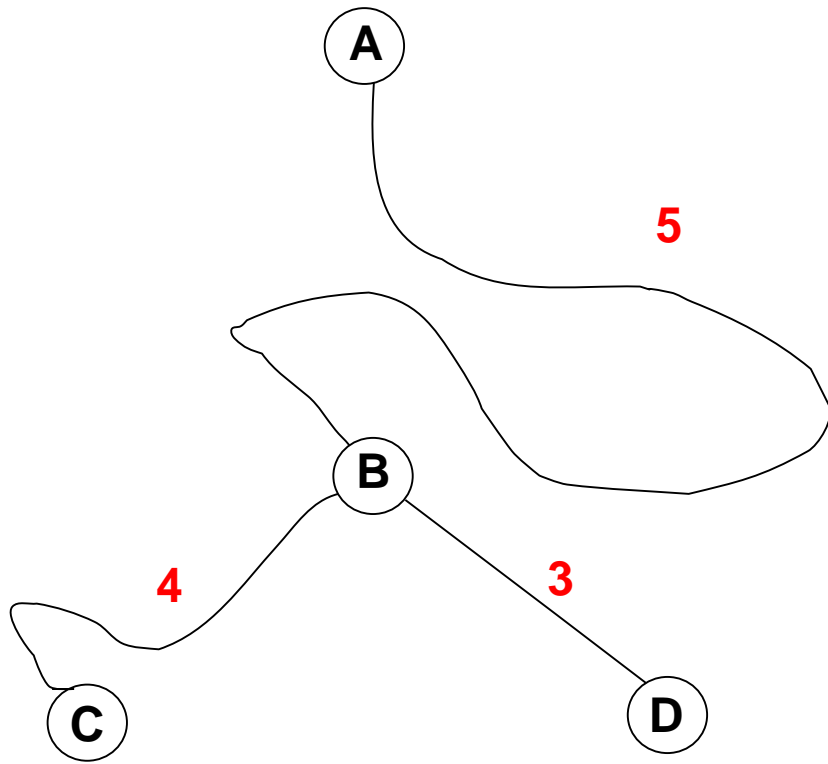
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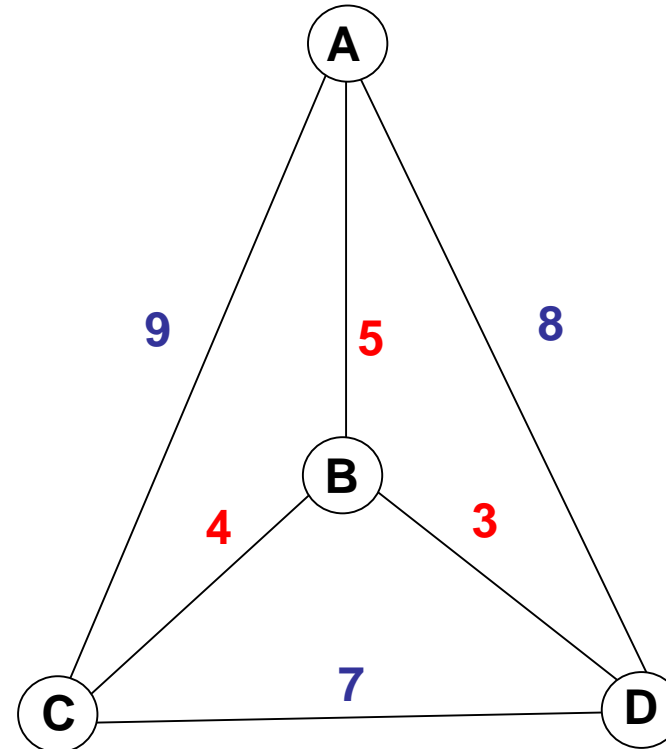


Road map

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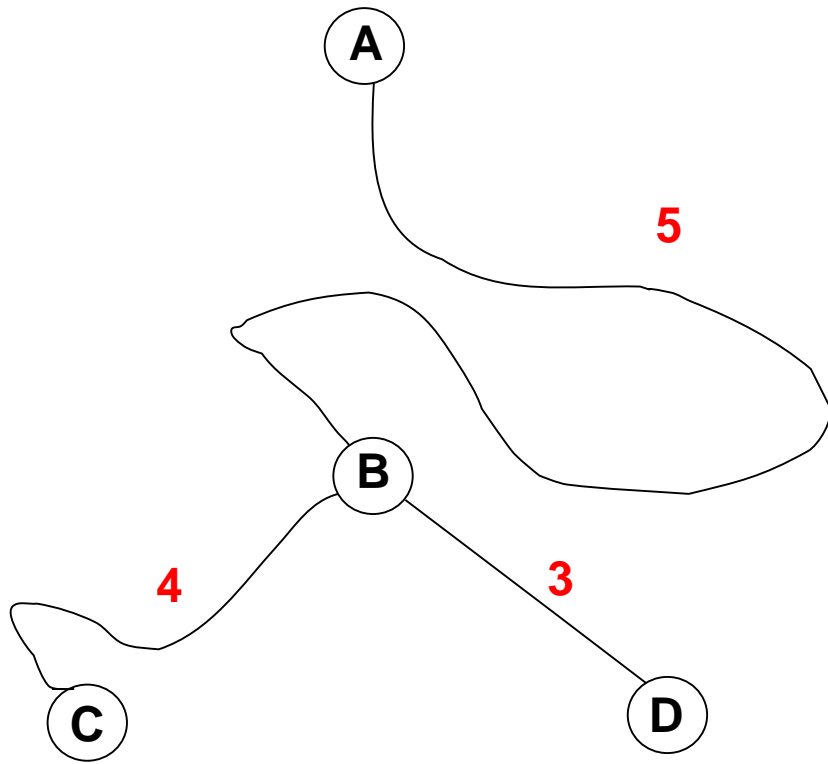


Road map

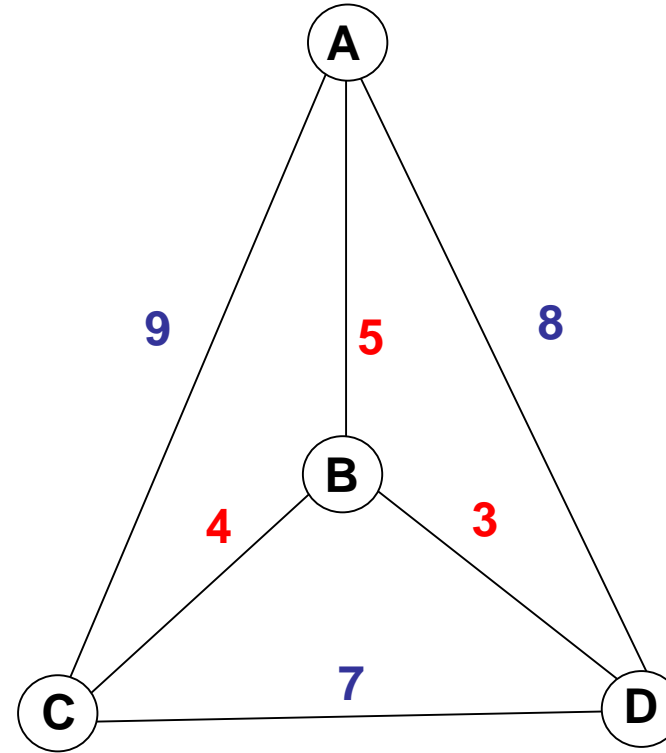


Travel costs

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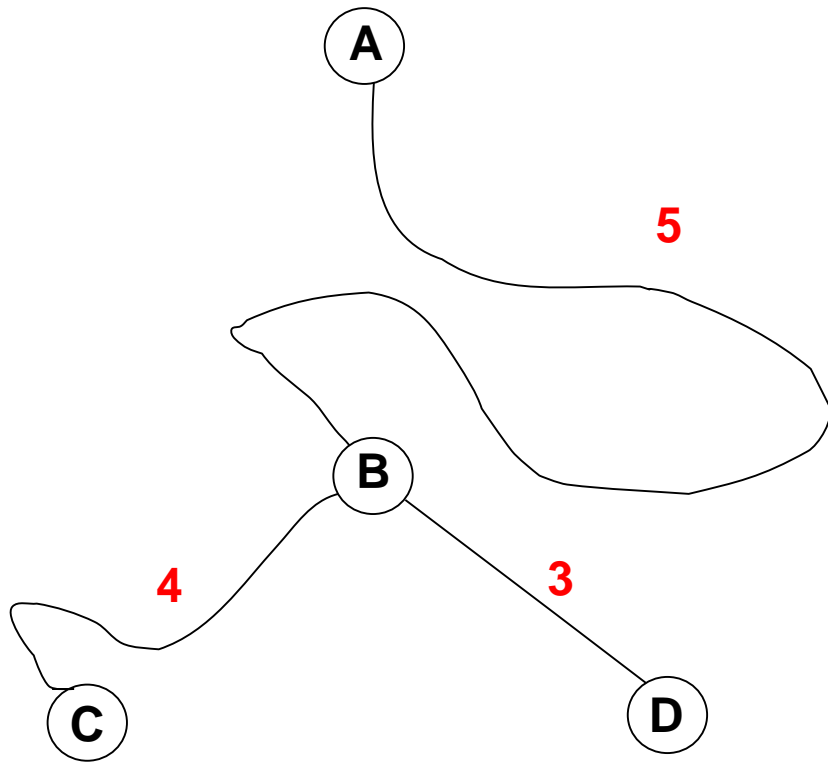
Road map



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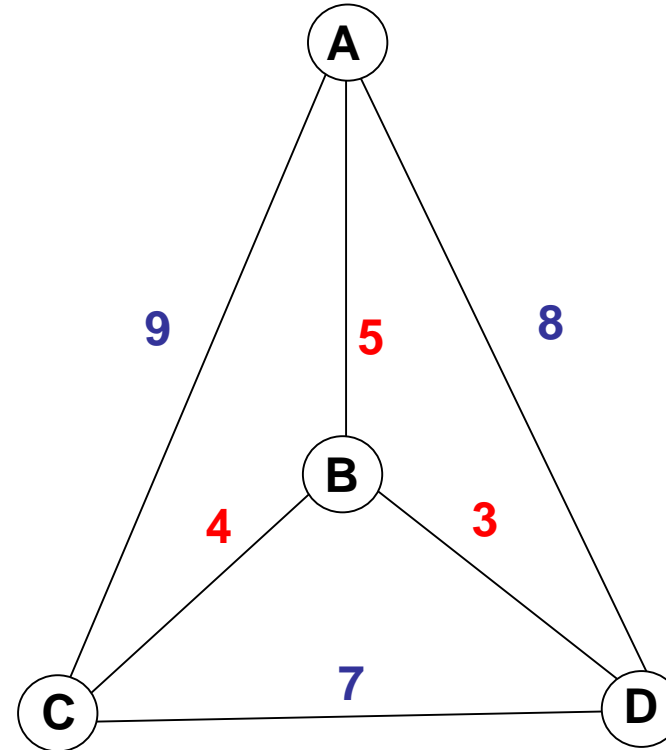
A-B-C-D-A

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A-B-C-B-D-B-A

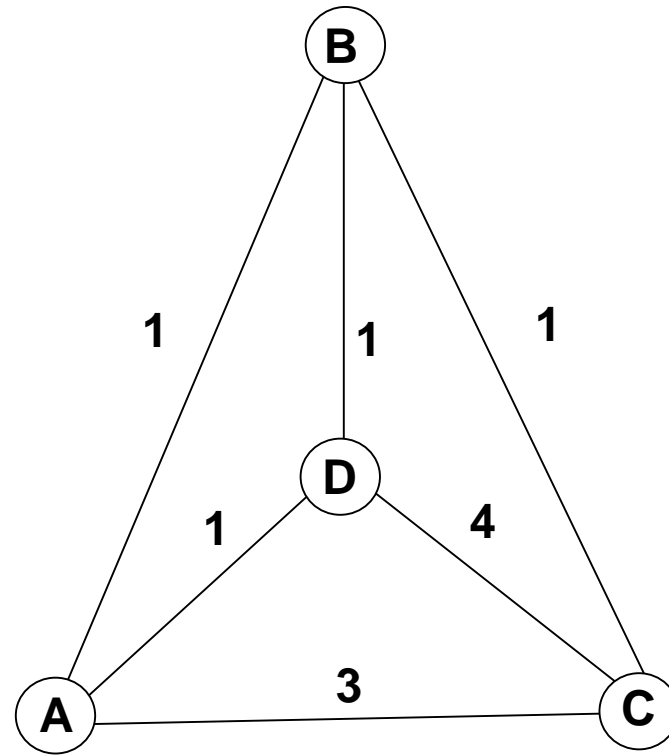


Travel costs

A-B-C-D-A

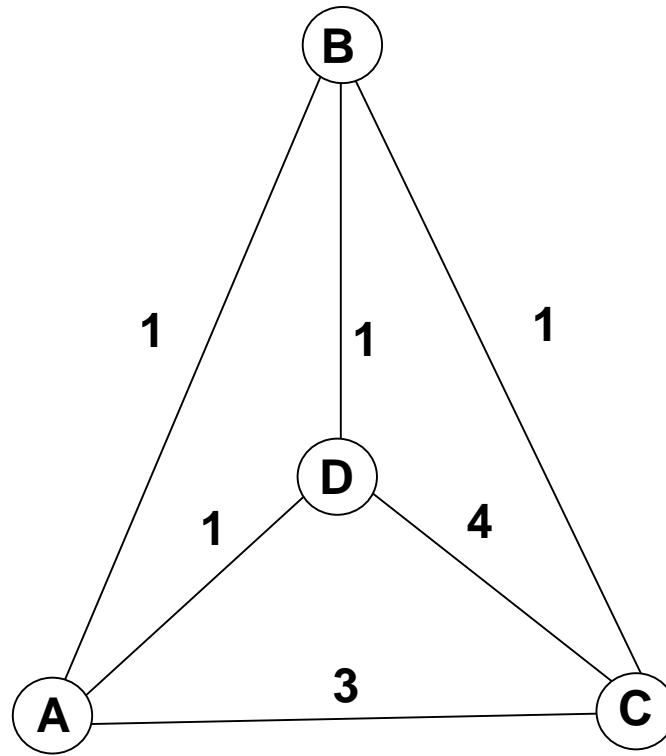
MISCONCEPTION #23:
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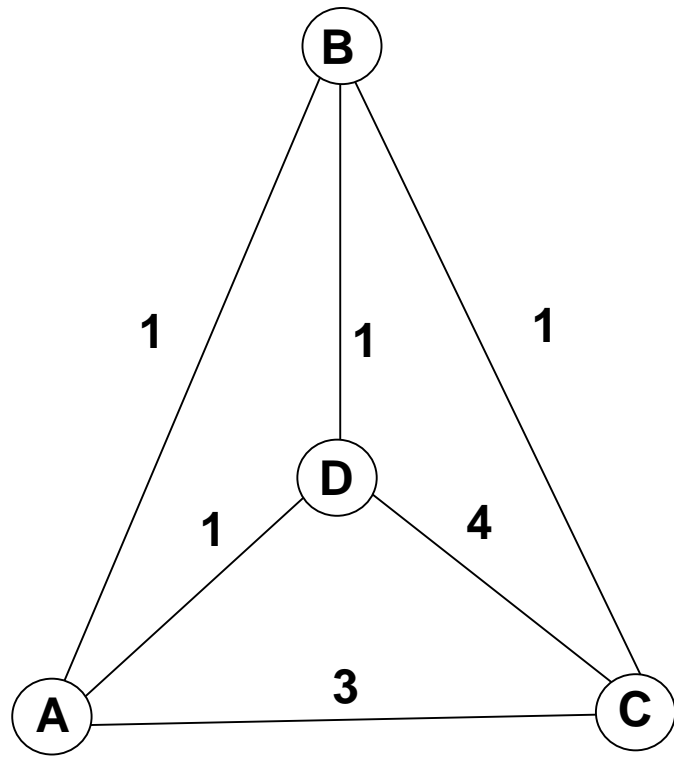
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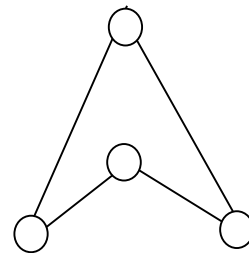
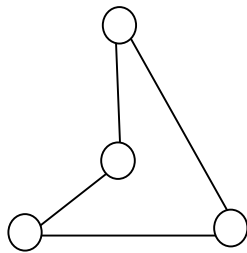
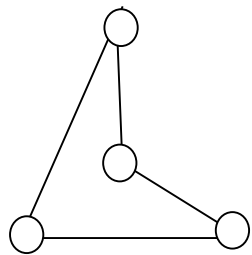
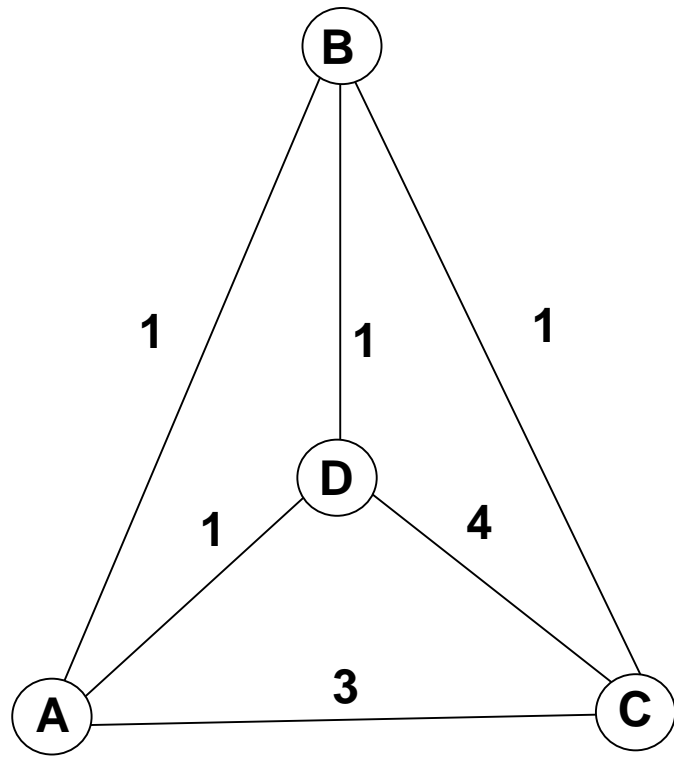
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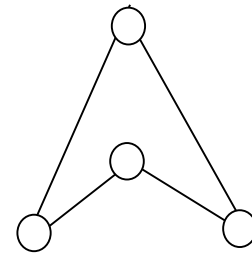
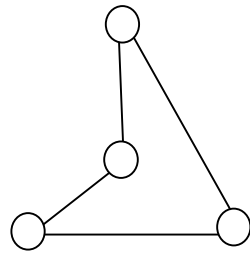
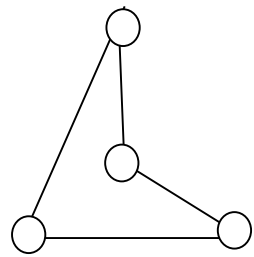
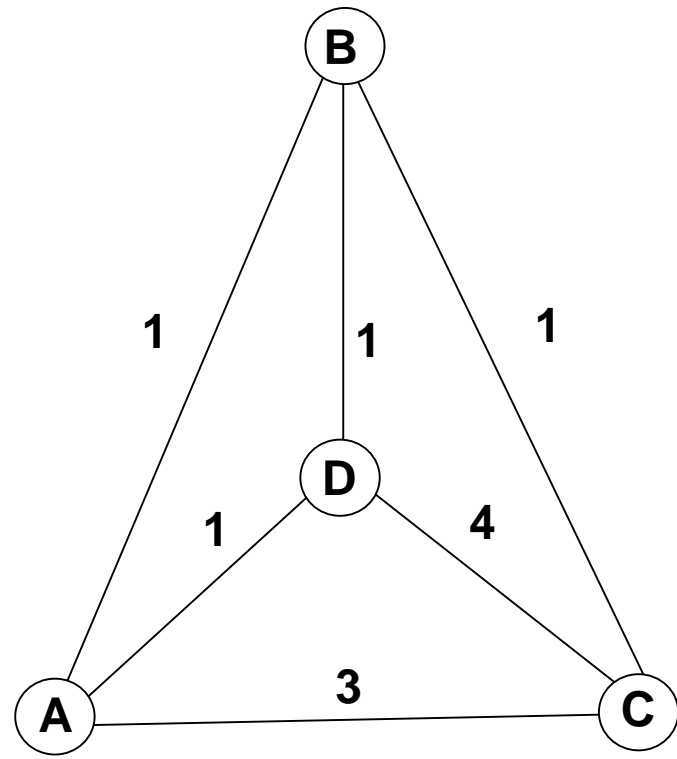


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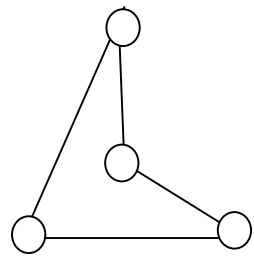
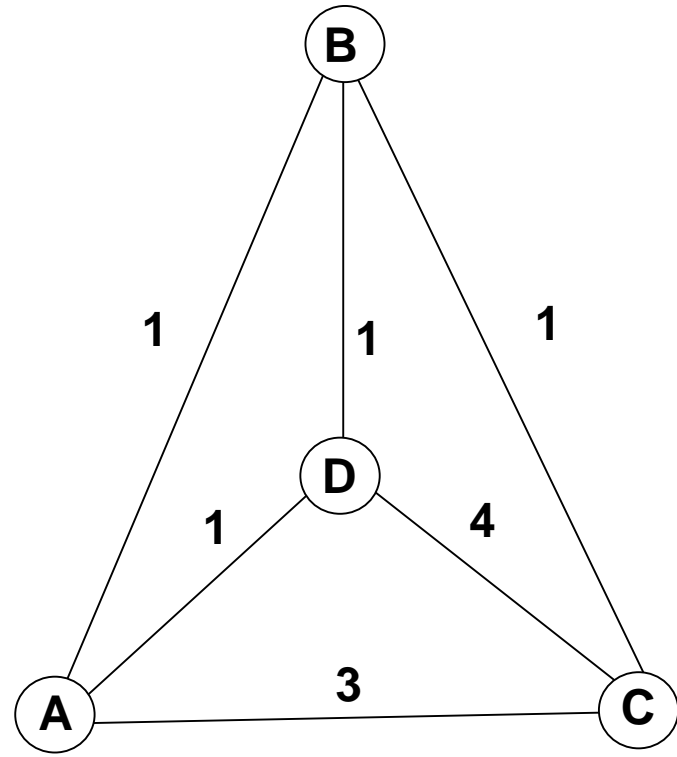
cost(A,C) exceeds cost(A,B)+cost(B,C)!



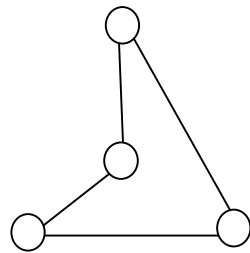




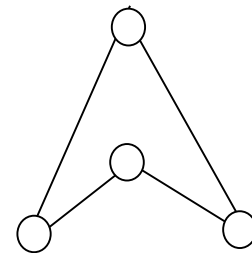
tour cost 9

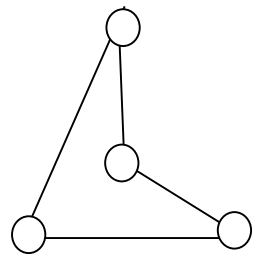
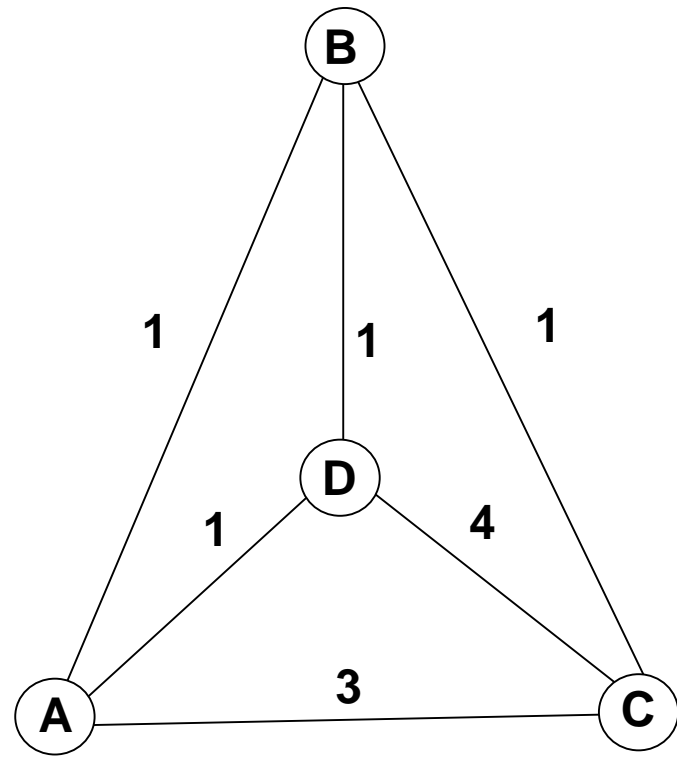


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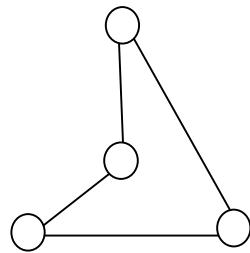


tour cost 6

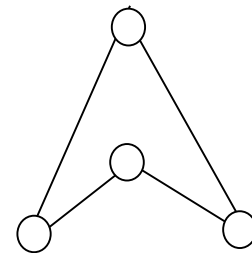




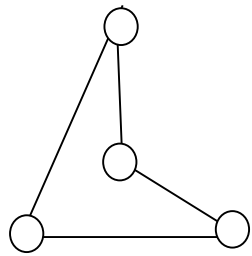
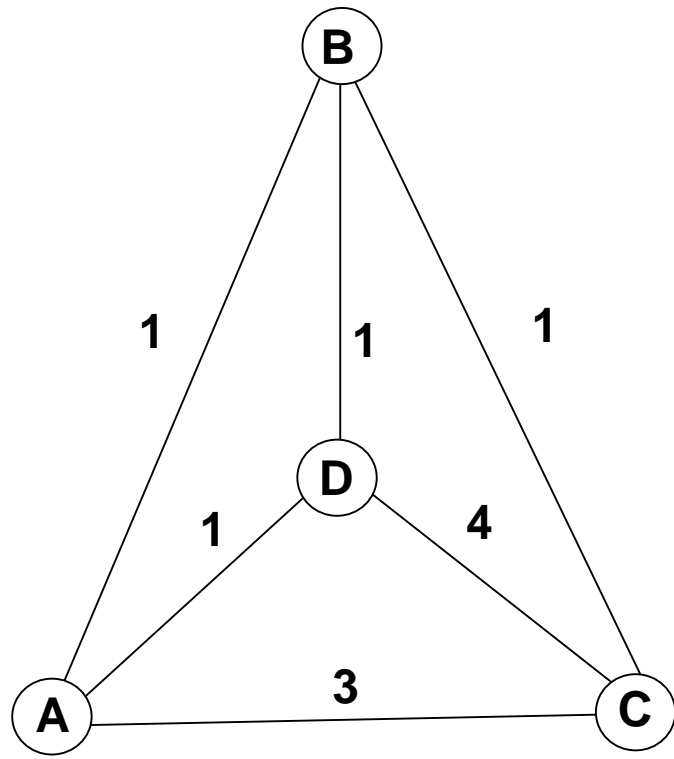
tour cost 9



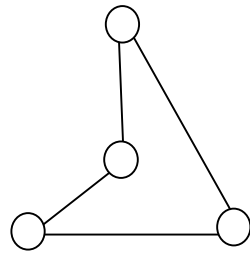
tour cost 6



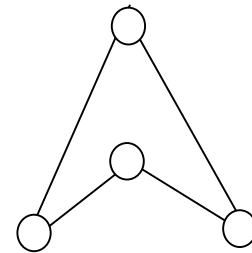
tour cost 7



tour cost 9



tour cost 6



tour cost 7

OPTIMAL

Great **terminological** confusion

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Mulder, S. A. and Wunsch, D. C. Million city traveling salesman problem **solution** by divide and conquer clustering with adaptive resonance neural networks. *Neural Networks* **16** (2003), 827-832.

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- etc. ...

Origins of the problem

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Karl Menger
Vienna 1930

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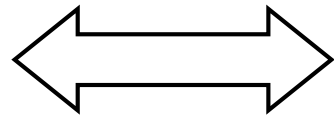


Hassler Whitney
Princeton 1934

Origins of the problem



Karl Menger
Vienna 1930



Harvard 1931

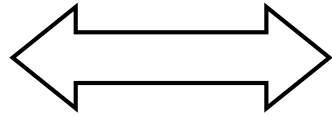


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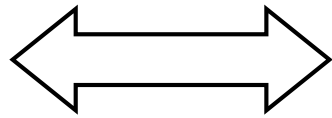


Merrill Flood
RAND Corporation 1948

Origins of the problem



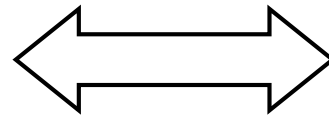
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Julia Robinson

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ARLINGTON 12, VIRGINIA
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Project **RAND**

ON THE HAMILTONIAN GAME
(A Traveling Salesmen Problem)
Julia Robinson
RM-303
5 December 1949 Copy No. 70

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REC-110

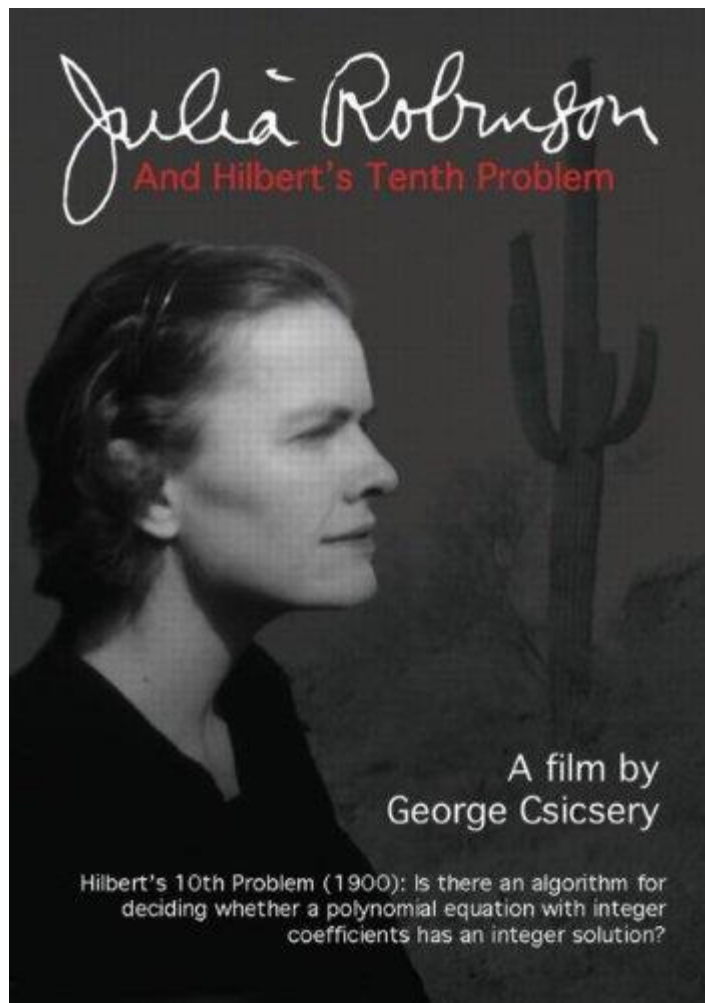
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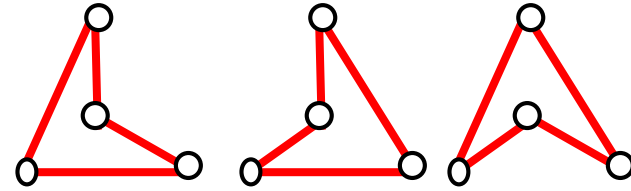
The RAND Corporation
SANTA MONICA · CALIFORNIA



<http://www.zalafilms.com/>

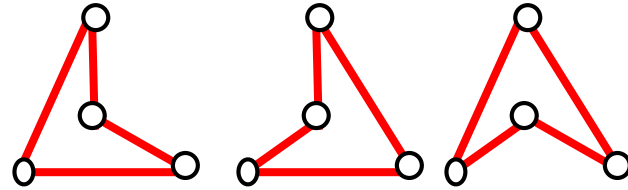
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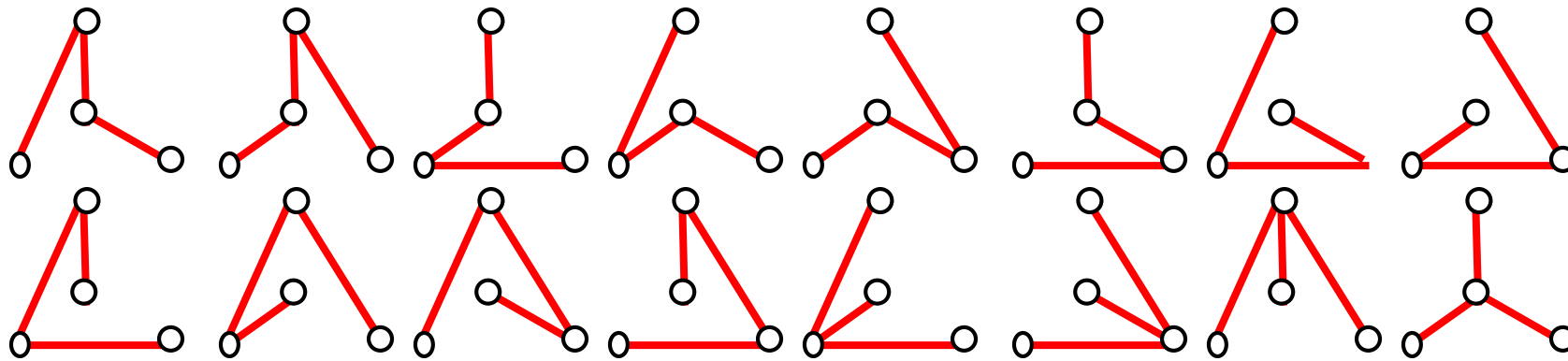


3 tours through 4 cities

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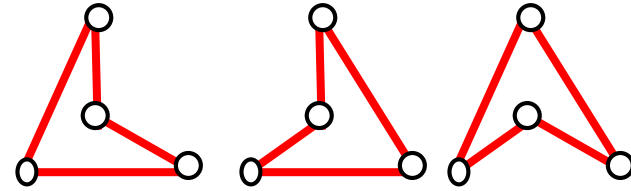


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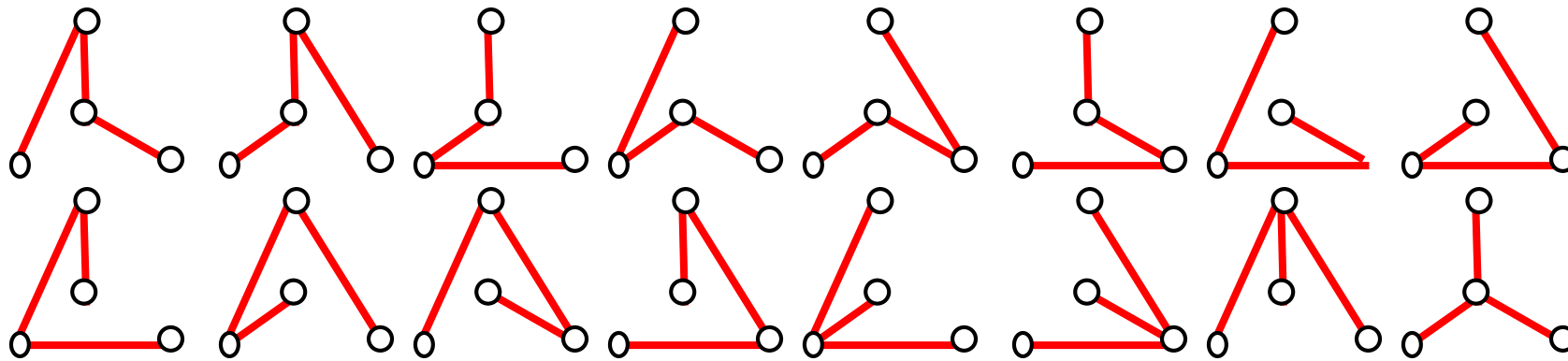
16 *spanning trees* on 4 cities

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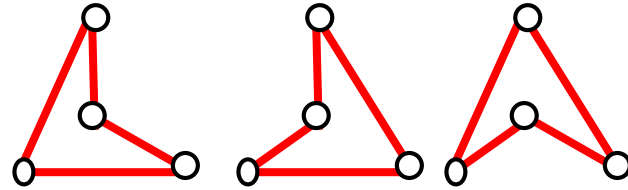
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$$\frac{1}{2} (n - 1)!$$



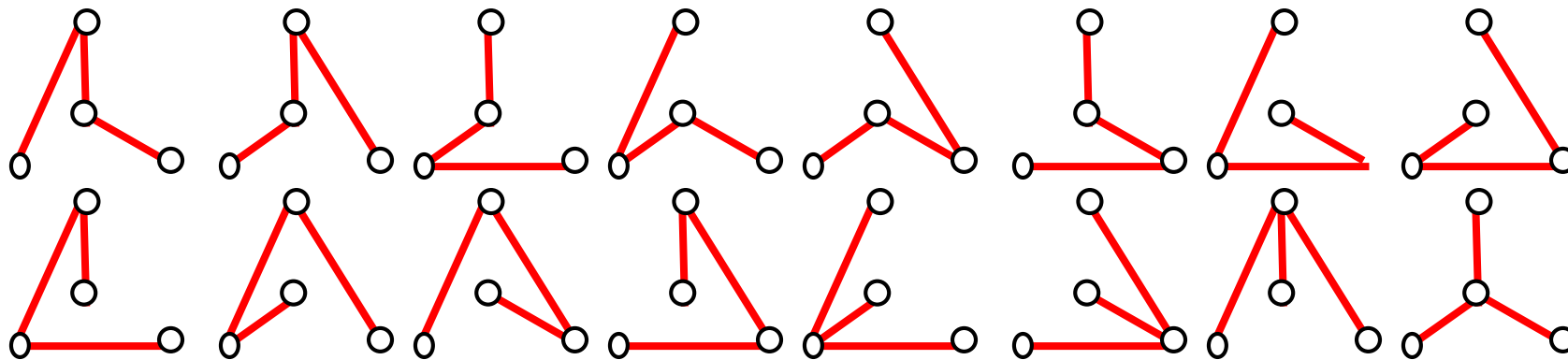
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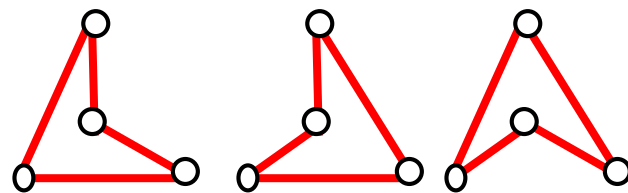
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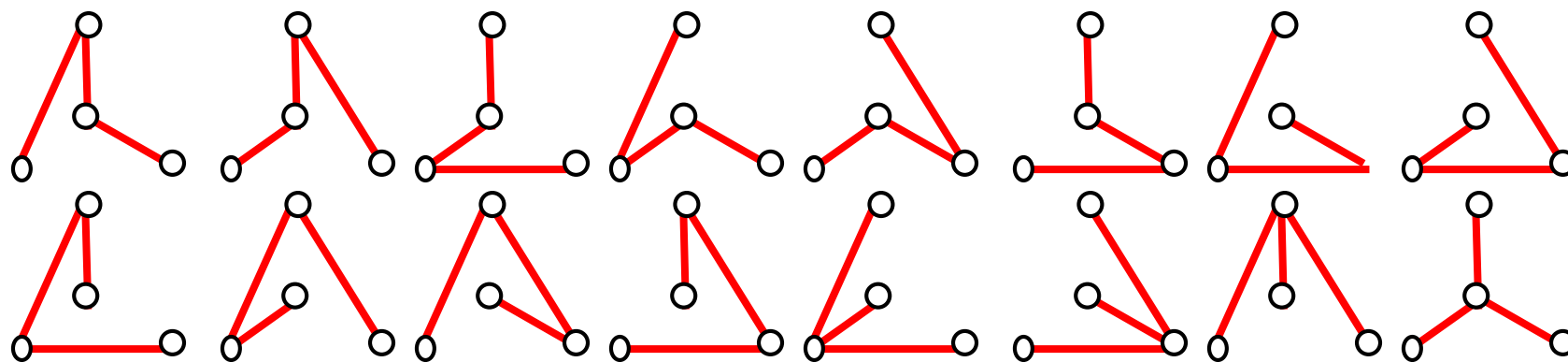
$$n^{n-2}$$

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3 tours through 4 cities

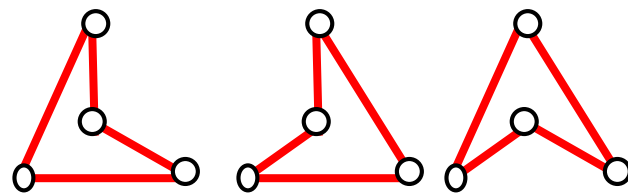
$$\frac{1}{2} (n - 1)!$$
$$= n^{n-2} e^{-n} O(n^{3/2})$$



16 spanning trees on 4 cities

$$n^{n-2}$$

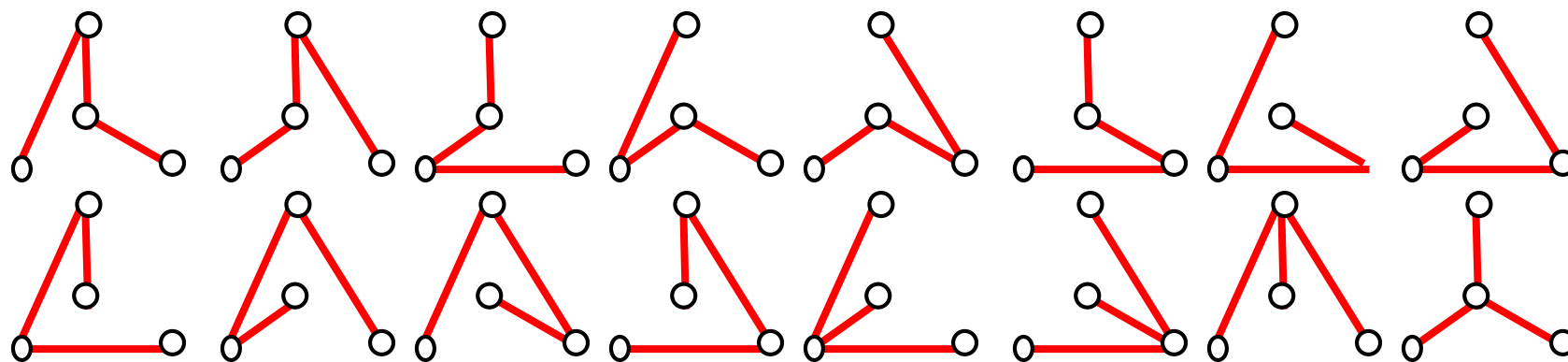
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THE MINIMUM-COST SPANNING TREE PROBLEM IS EASY!



16 *spanning trees* on 4 cities

$$n^{n-2}$$

WHAT MAKES THE TSP SO HARD?

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This question stimulated the study of *computational complexity* and, in particular, the development of the *theory of NP-completeness*

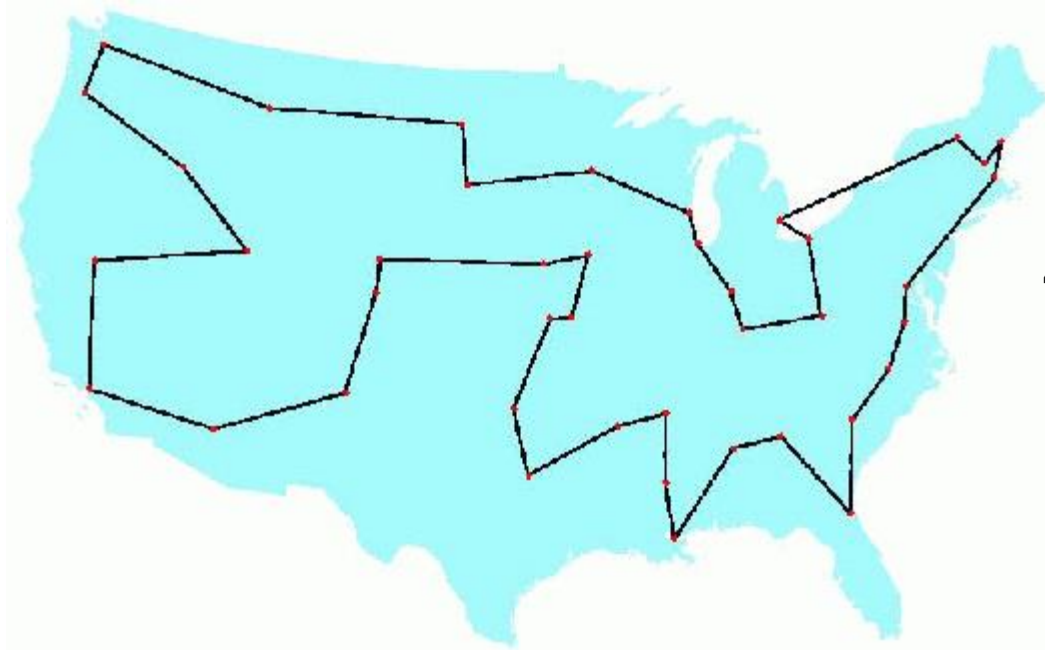
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Stephen A. Cook, *The complexity of theorem-proving procedures*. In Proc. 3rd Annual ACM Symposium on the Theory of Computing, 1971, pp.151—158.

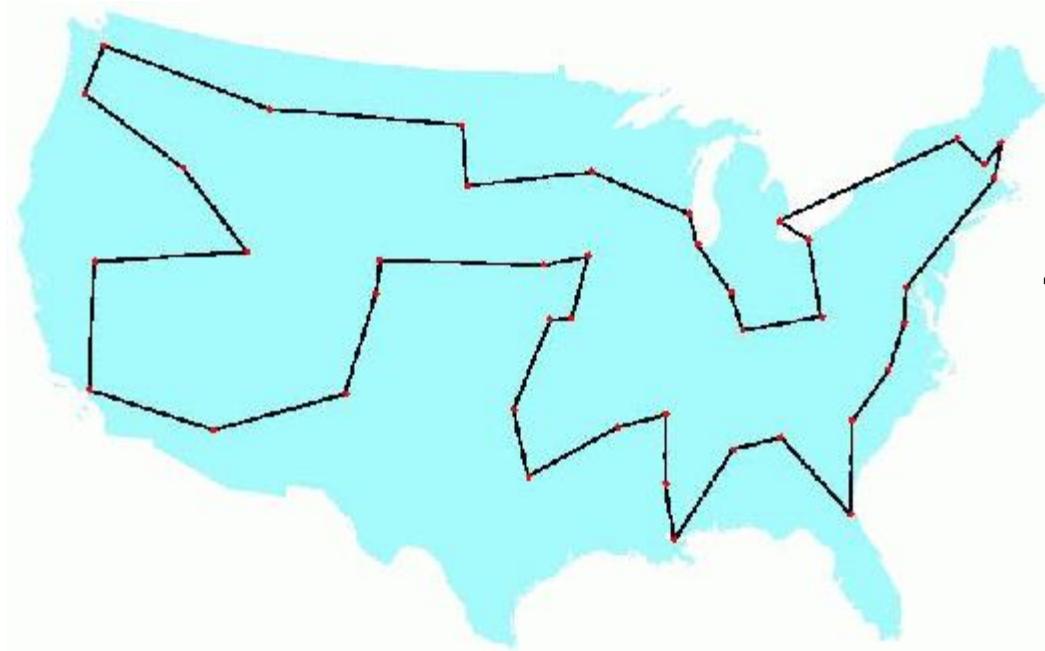
Richard M. Karp. *Reducibility among combinatorial problems*. In R.E. Miller and J.W.Thatcher, editors, *Complexity of Computer Computations*, Plenum Press, 1972, pp. 85--104.

G. Dantzig, R. Fulkerson, and S. Johnson,
"Solution of a large-scale traveling-salesman problem",
Operations Research **2** (1954), 393-410.



49 cities in the U.S.A.

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The Big Bang

The Dantzig-Fulkerson-Johnson cutting-plane method

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repeat  $x^* =$  an extreme point of  $\{x : Ax \leq b\}$  that minimizes  $c^T x$ ;
      if  $x^*$  belongs to  $\mathcal{S}$ 
      then return  $x^*$ ;
```



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end
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  end
end
```

 cutting plane

The Dantzig-Fulkerson-Johnson cutting-plane method

applies to any problem minimize $c^T x$ subject to $x \in \mathcal{S}$
such that \mathcal{S} is a finite subset of a Euclidean space

```
choose a system  $Ax \leq b$  of inequalities satisfied by all points of  $\mathcal{S}$ 
repeat  $x^* =$  an extreme point of  $\{x : Ax \leq b\}$  that minimizes  $c^T x$ ;
  if  $x^*$  belongs to  $\mathcal{S}$ 
  then return  $x^*$ ;
  else find a linear inequality satisfied by all points of  $\mathcal{S}$ 
        and violated by  $x^*$ ;
        add this inequality to  $Ax \leq b$ ;
  end
end
```

 cutting plane

The Dantzig-Fulkerson-Johnson initial TSP box:

$$\begin{aligned} 0 \leq x_{vw} \leq 1 & \quad \text{for all edges } vw \\ \sum_w x_{vw} = 2 & \quad \text{for all vertices } v \end{aligned}$$

1962 Procter & Gamble \$10,000 contest: a 33-city instance

HELP! WE'RE LOST!

HELP "CAR 54"...AND WIN CASH
54...\$1,000 PRIZES
ONE...\$10,000 GRAND PRIZE

START and FINISH

Help Toody and Muldoon find the shortest round trip route to visit all 33 locations shown on the map. All you do is draw connecting straight lines from location to location to show the shortest round trip route.

HERE'S THE CORRECT START...

Begin at Chicago, Illinois. From there, lines show correct route as far as Erie, Pennsylvania. Next, do you go to Carlisle, Pennsylvania or Wana, West Virginia? Check the easy instructions on back of this entry blank for details.

PROCTER & GAMBLE 1962

OFFICIAL RULES ON REVERSE SIDE

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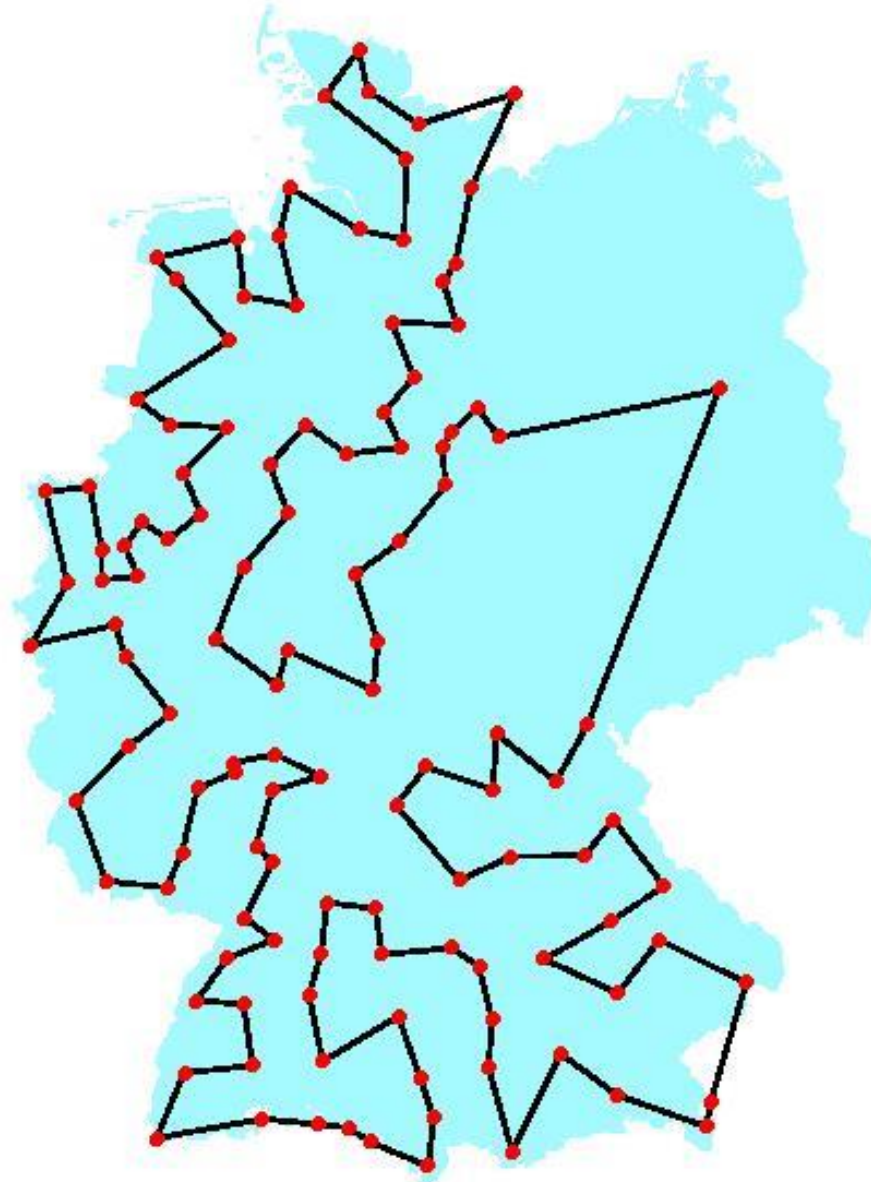
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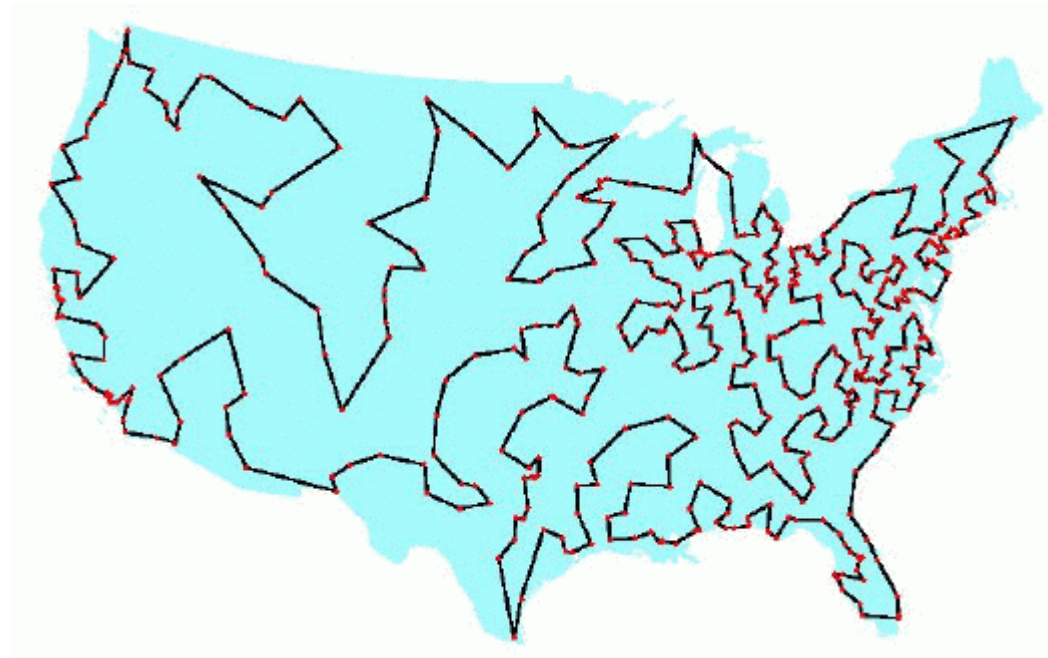
OFFICIAL RULES ON REVERSE SIDE

Tiebreaker: Write a short essay on one of Procter & Gamble's products

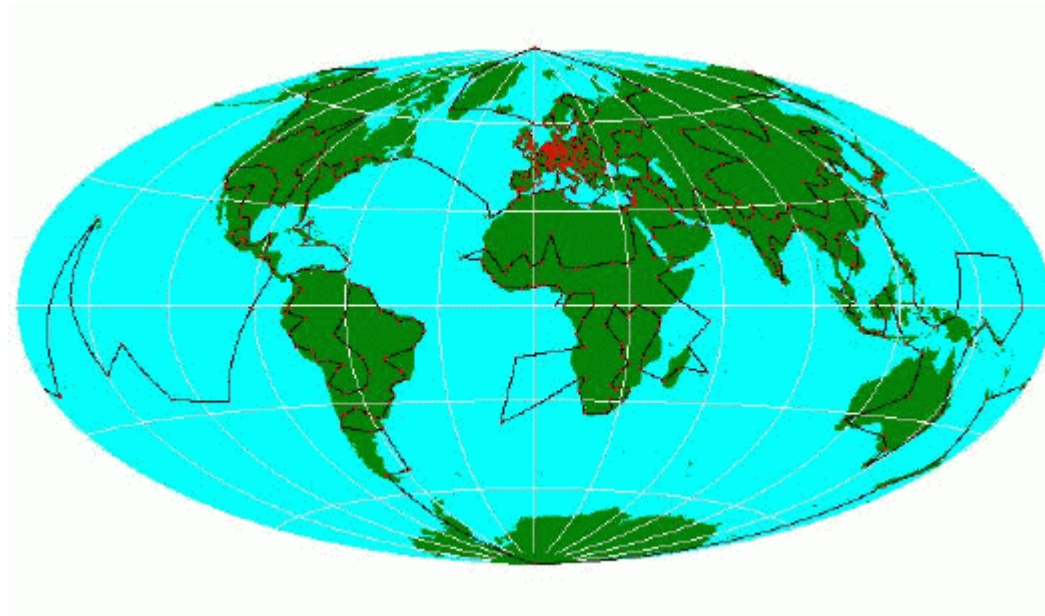
1977 Martin Grötschel: a 120-city instance



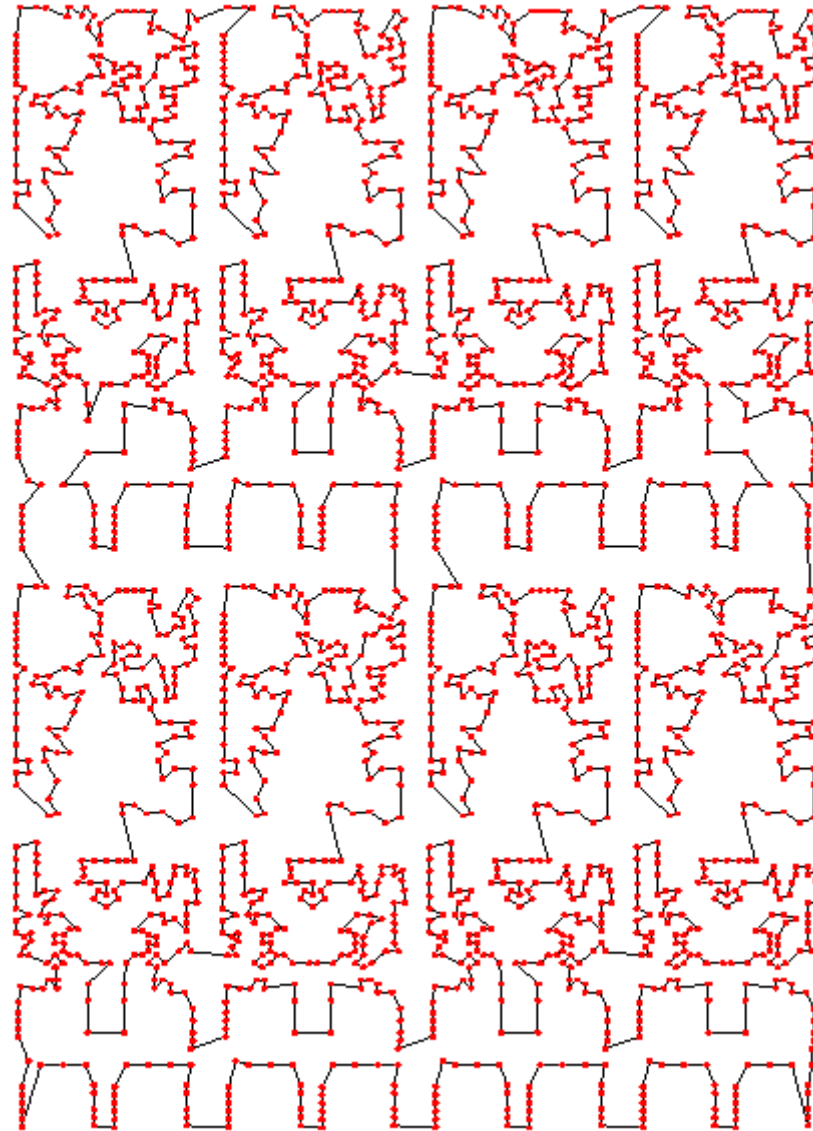
1986 Manfred Padberg and Giovanni Rinaldi: a 532-city instance

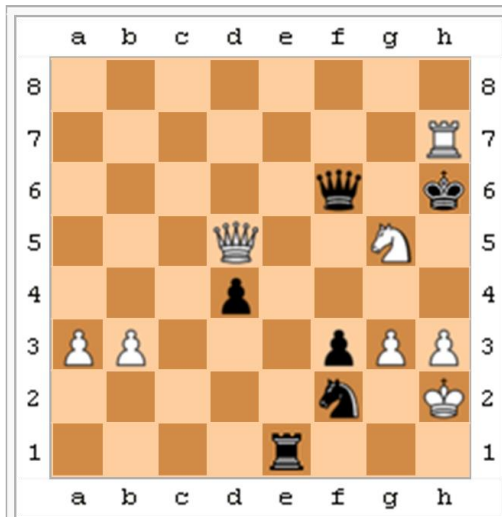


1987 Martin Grötschel and Olaf Holland: a 666-city instance

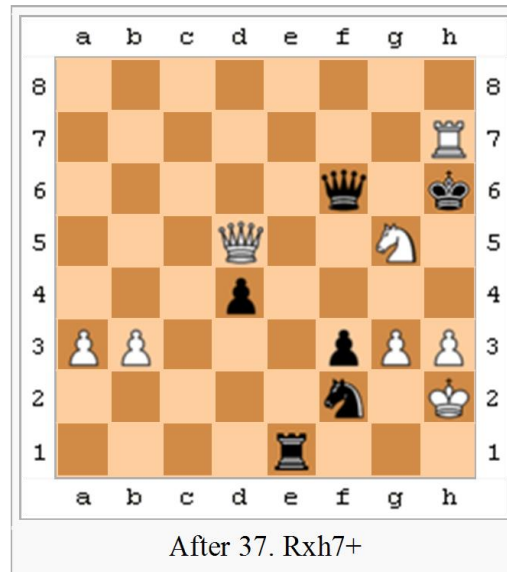


1987 Manfred Padberg and Giovanni Rinaldi: a 2,392-city instance





After 37. Rxd7+

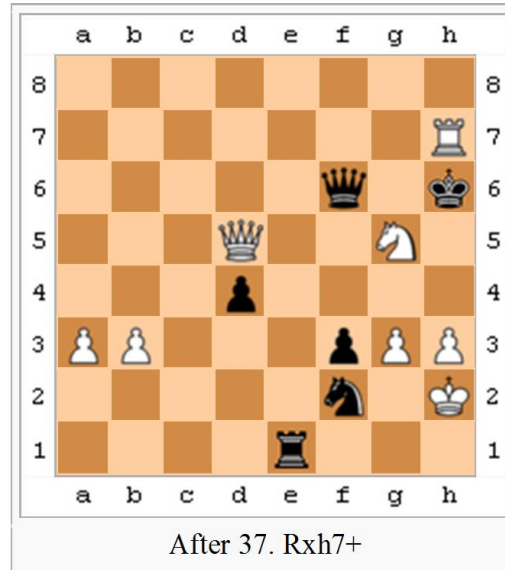


February 10, 1996



February 10, 1996

IBM's "supercomputer" *Deep Blue* beats Garry Kasparov



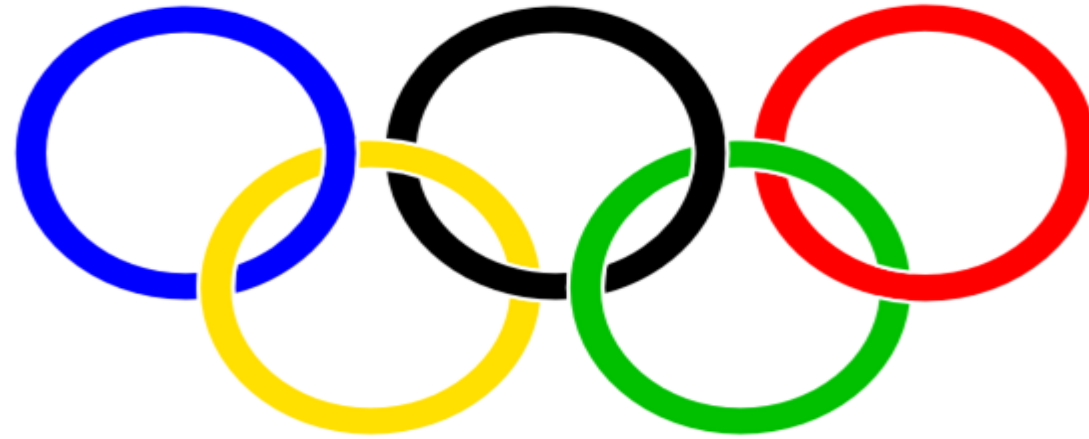
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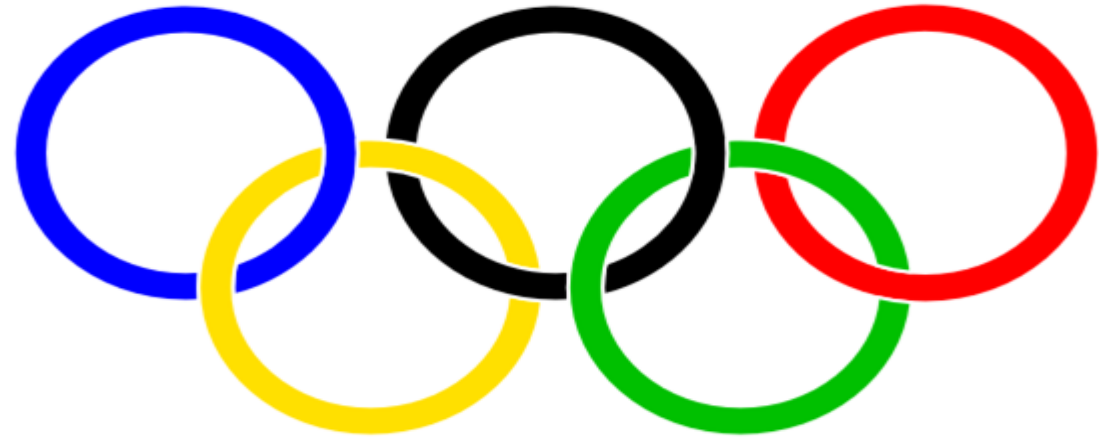
From an interview with V.C. published in 1996:

The traveling salesman problem is to mathematical programming what chess is to artificial intelligence: thoroughly useless and fiercely competitive sport that serves as a testing ground of your techniques.

The sport of solving the TSP



The sport of solving the TSP



My daddy can beat up your daddy

APPLICATION: Genome sequencing

APPLICATION: Genome sequencing



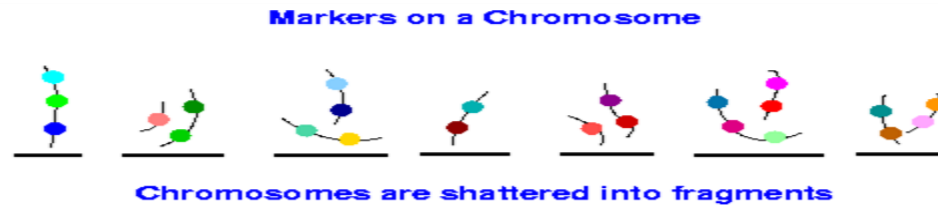
In which order are markers arranged on a genome?

APPLICATION: Genome sequencing



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Radiation hybrid technique: Break the DNA into pieces by X-rays and grow hybrid cells from these pieces



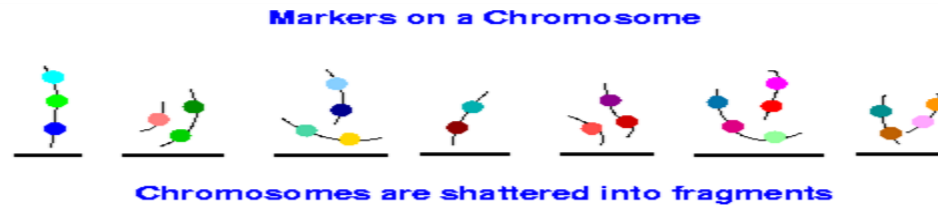
The closer two markers are to each other on the DNA, the more often they appear together in the hybrid cells.

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R. Karp, W. Ruzzo, and M. Tompa (1996):
To find a sequence that best fits the radiation hybrid data, solve a TSP

APPLICATION: NASA's StarLight mission

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The two StarLight mission spacecraft will orbit the sun, sort of tagging along behind Earth. Both spacecraft will carry telescope mirrors. Both telescope mirrors will be turned to look at the same star. Star light from the two mirrors will be combined to create a very good image, which is then sent back to Earth. Laser beams are used to keep the two spacecraft very precisely aligned.

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slew *also slue transitive verb 1* : to turn (as a telescope or a ship's spar) about a fixed point that is usually the axis

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Bailey, C., McLain, T., and Beard, R. Fuel Saving Strategies for Dual Spacecraft Interferometry Missions, *Journal of the Astronautical Sciences*, Volume 49, Number 3, pp. 469-488, July-September 2001.

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Bailey, C., McLain, T., and Beard, R. Fuel Saving Strategies for Dual Spacecraft Interferometry Missions, *Journal of the Astronautical Sciences*, Volume 49, Number 3, pp. 469-488, July-September 2001.

StarLight cancelled in 2002



OTHER APPLICATIONS

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- . postal deliveries
- . meals on wheels
- . inspection tours
- . school bus routing

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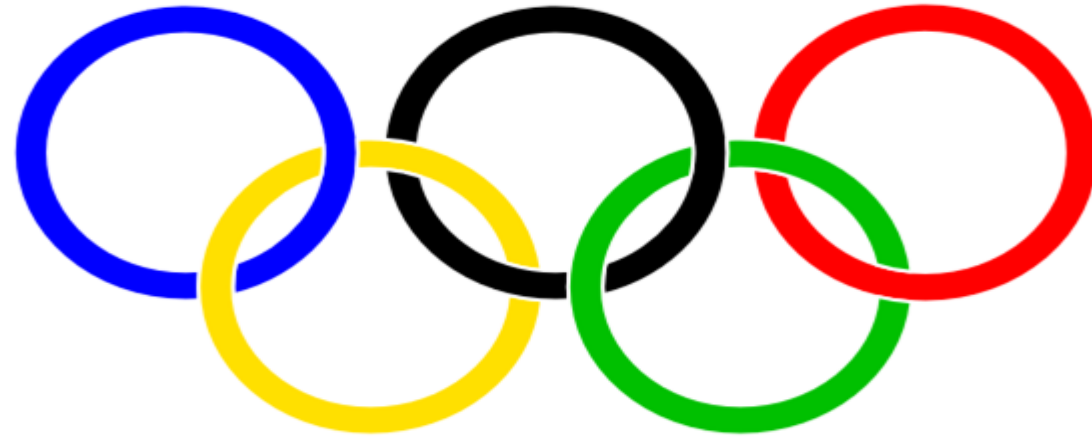
OTHER APPLICATIONS

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- post-manufacture computer chip testing (“scan chains”)
- data clustering

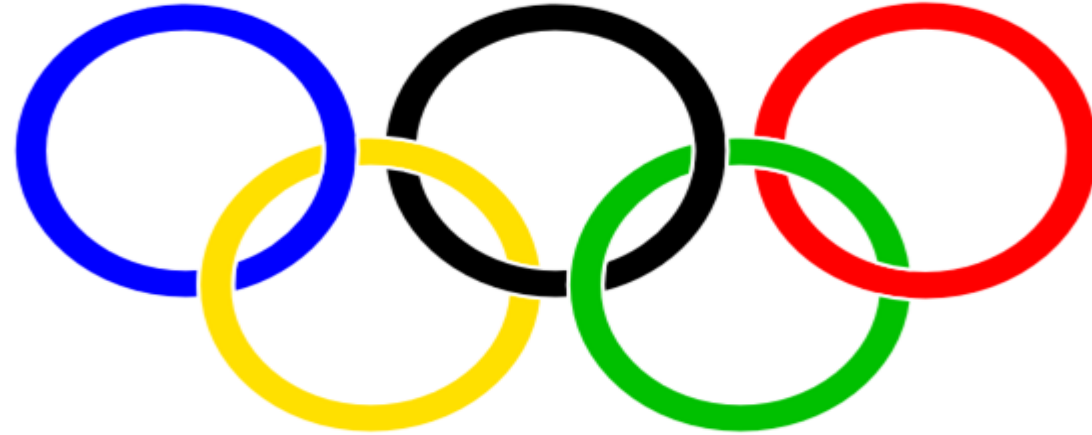
OTHER APPLICATIONS

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- . etc. etc. etc.

The sport of solving the TSP



The sport of solving the TSP



TSPLIB:

a library of 111 instances

collected by Gerhard Reinelt (Heidelberg) in 1991

TSPLIB

a280	d2103	gr48	lin318	pr264	si1032
ali535	d15112	gr96	linhp318	pr299	st70
att48	d18512	gr120	nrw1379	pr439	swiss42
att532	dantzig42	gr137	p654	pr1002	ts225
bayg29	dsj1000	gr202	pa561	pr2392	tsp225
bays29	eil51	gr229	pcb442	rat99	u159
berlin52	eil76	gr431	pcb1173	rat195	u574
bier127	eil101	gr666	pcb3038	rat575	u724
brazil58	fl417	hk48	pla7397	rat783	u1060
brd14051	fl1400	kroA100	pla33810	rd100	u1432
brg180	fl1577	kroB100	pla85900	rd400	u1817
burma14	fl3795	kroC100	pr76	rl1304	u2152
ch130	fnl4461	kroD100	pr107	rl1323	u2319
ch150	fri26	kroE100	pr124	rl1889	ulysses16
d198	gil262	kroA150	pr136	rl5915	ulysses22
d493	gr17	kroB150	pr144	rl5934	usa13509
d657	gr21	kroA200	pr152	rl11849	vm1084
d1291	gr24	kroB200	pr226	si175	vm1748
d1655		lin105		si535	

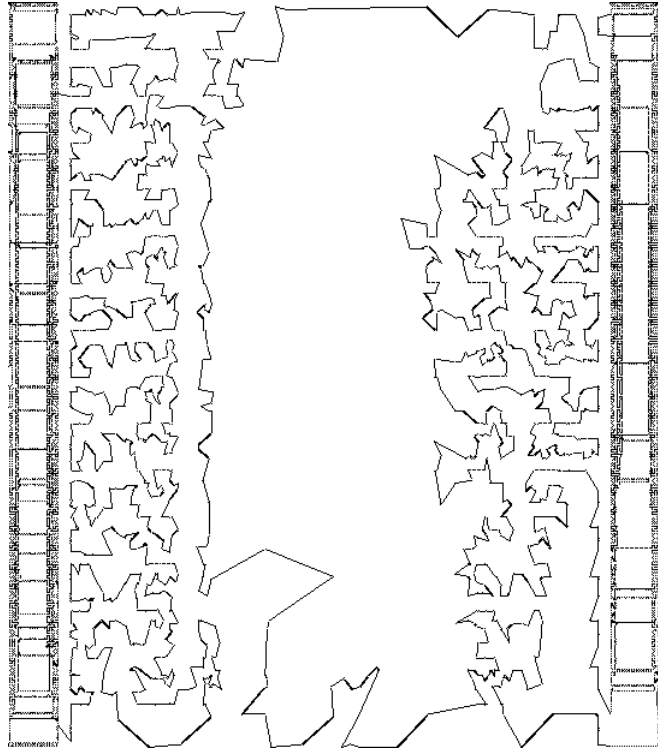
TSPLIB

a280	d2103	gr48	lin318	pr264	si1032
ali535	d15112	gr96	linhp318	pr299	st70
att48	d18512	gr120	nrw1379	pr439	swiss42
att532	dantzig42	gr137	p654	pr1002	ts225
bayg29	dsj1000	gr202	pa561	pr2392	tsp225
bays29	eil51	gr229	pcb442	rat99	u159
berlin52	eil76	gr431	pcb1173	rat195	u574
bier127	eil101	gr666	pcb3038	rat575	u724
brazil58	fl417	hk48	pla7397	rat783	u1060
brd14051	fl1400	kroA100	pla33810	rd100	u1432
brg180	fl1577	kroB100	pla85900	rd400	u1817
burma14	fl3795	kroC100	pr76	rl1304	u2152
ch130	fnl4461	kroD100	pr107	rl1323	u2319
ch150	fri26	kroE100	pr124	rl1889	ulysses16
d198	gil262	kroA150	pr136	rl5915	ulysses22
d493	gr17	kroB150	pr144	rl5934	usa13509
d657	gr21	kroA200	pr152	rl11849	vm1084
d1291	gr24	kroB200	pr226	si175	vm1748
d1655		lin105		si535	

The red instances were unsolved in 1991

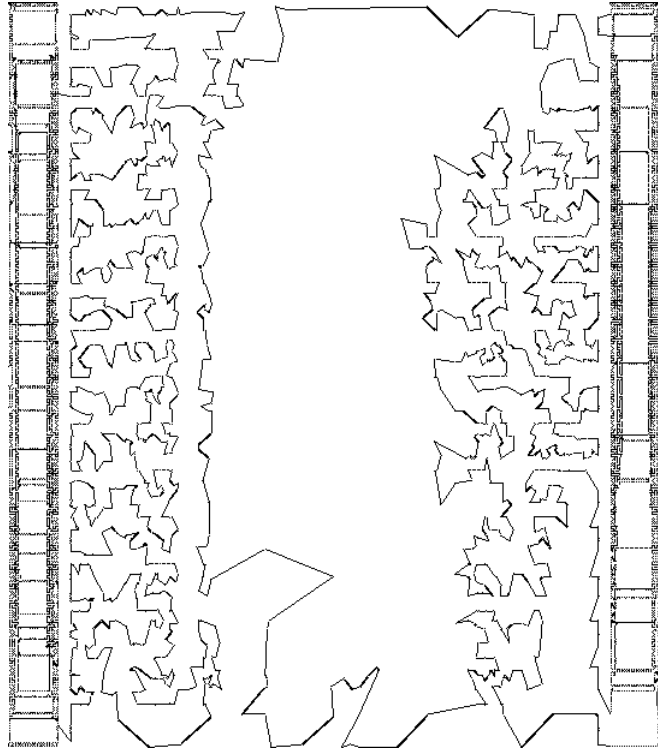
MISCONCEPTION #8: SIZE IS IMPORTANT

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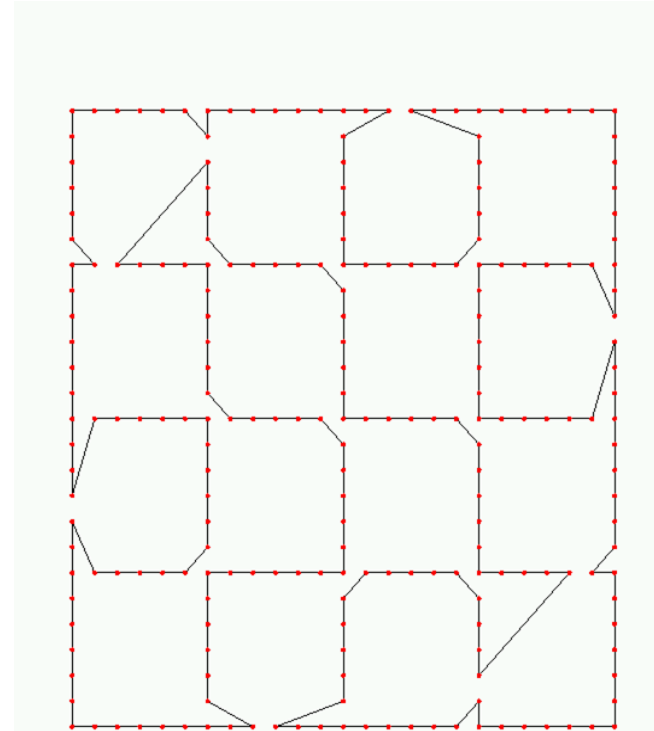


pla7397 solved
in October 1994

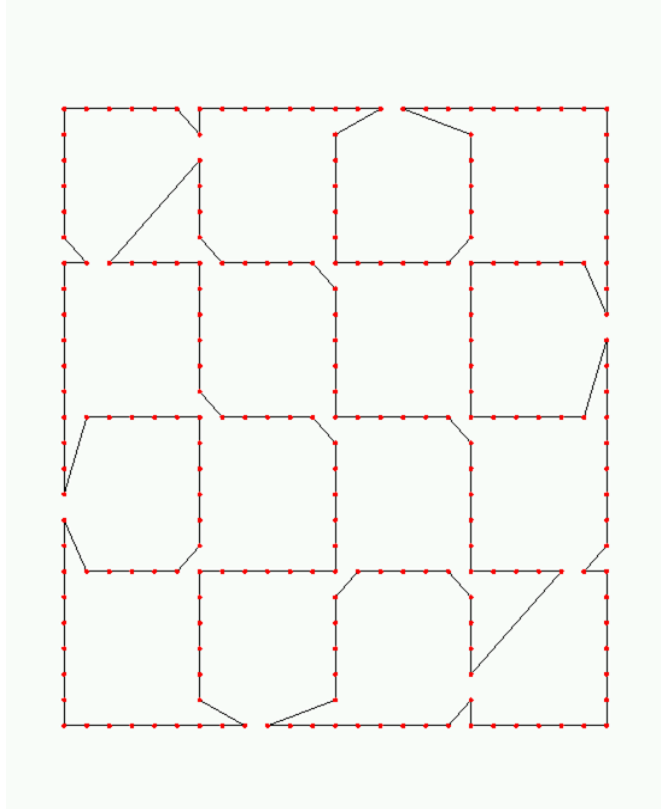
MISCONCEPTION #8: SIZE IS IMPORTANT



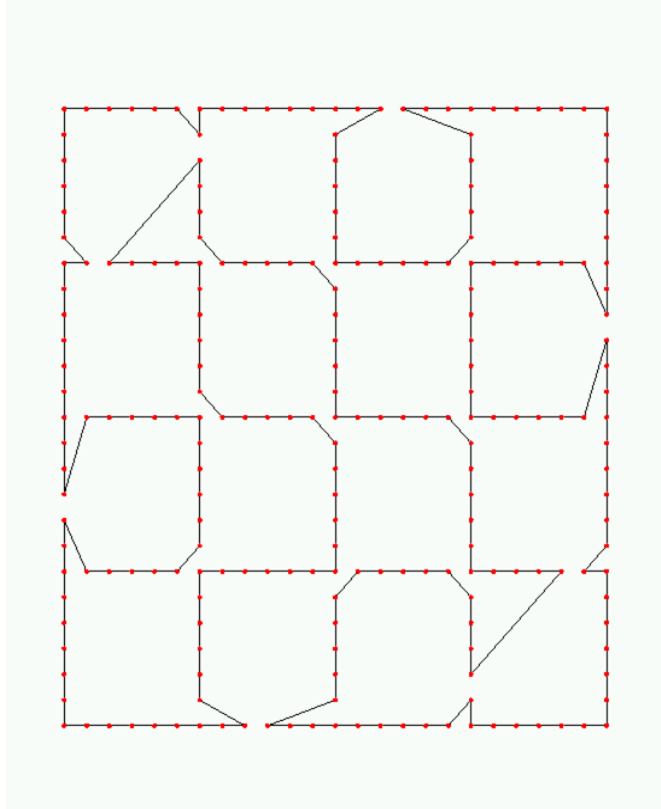
pla7397 solved
in October 1994



ts225 unsolved
in October 1994

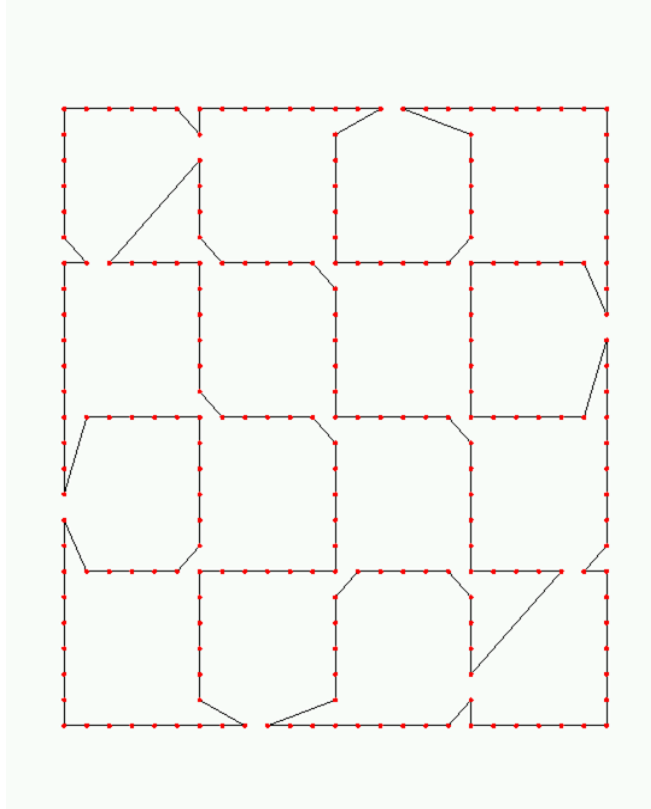


ts225 constructed by Stefan Tschöke



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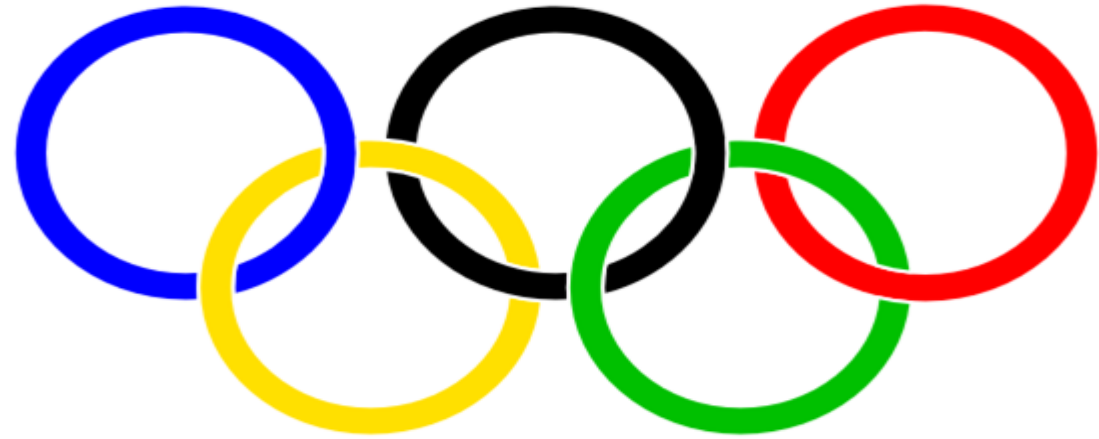
with malice aforethought



ts225 constructed by Stefan Tschöke
with malice aforethought

**LESSON #1:
MISCHIEF IS A POWERFUL ENGINE OF DISCOVERY**

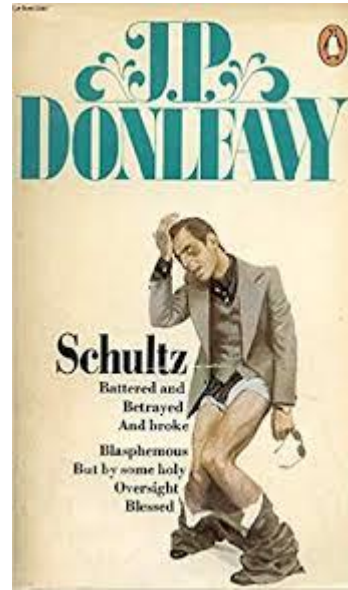
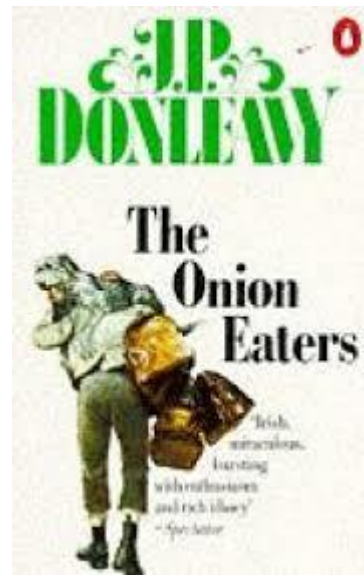
The sport of solving the TSP



My daddy can beat up your daddy

Bill Cook





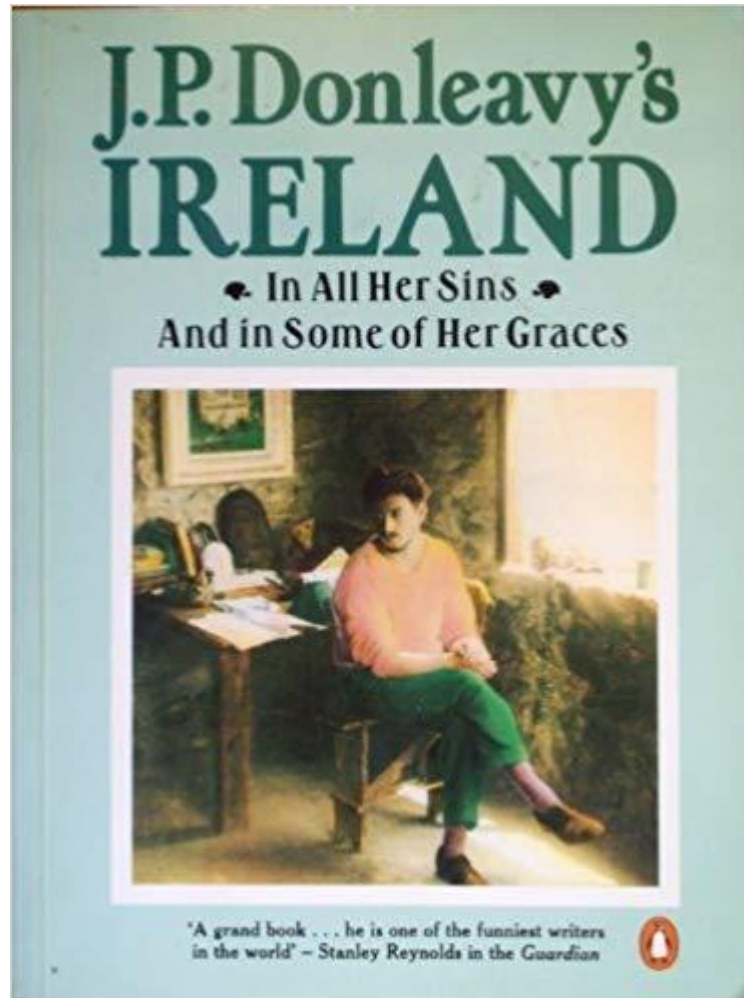
J.P. Donleavy's IRELAND

◆ In All Her Sins ◆
And in Some of Her Graces

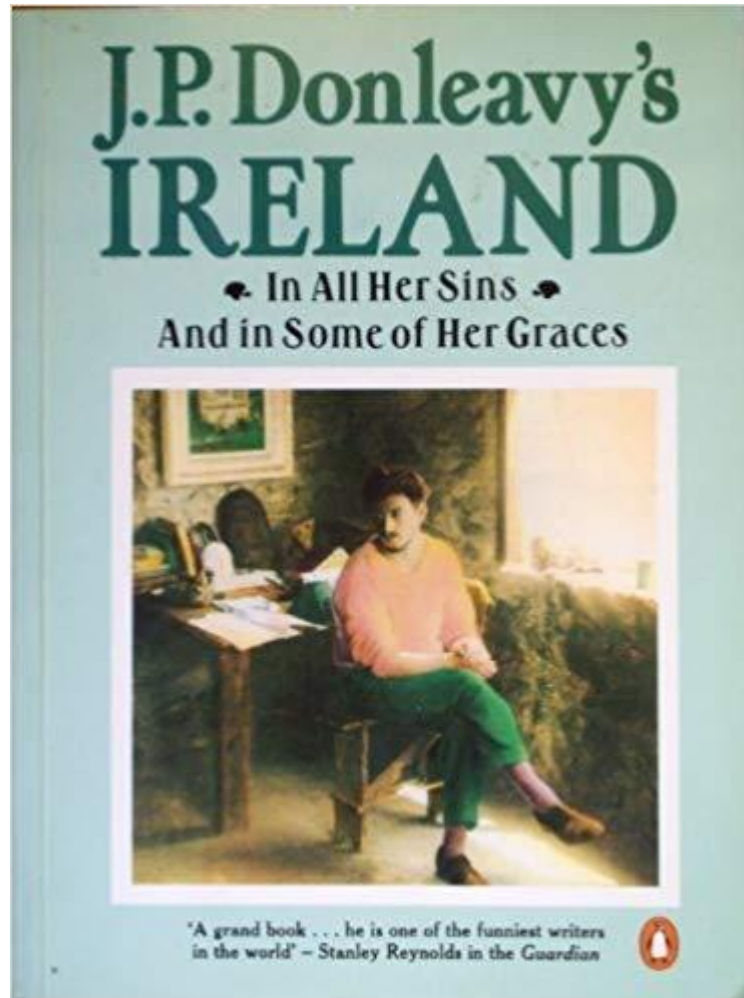


'A grand book . . . he is one of the funniest writers
in the world' – Stanley Reynolds in the *Guardian*



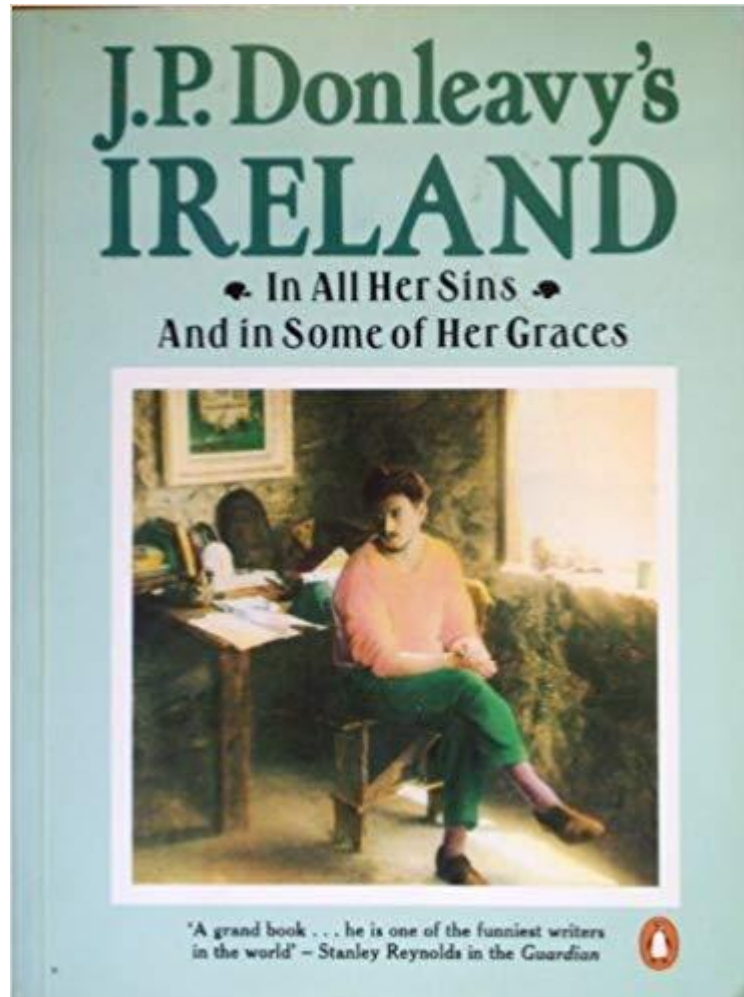


There is also the oft-repeated recollection in Donleavy's autobiographical writing that his Irish parents grew up "without a pot to piss in".



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J. P. writes in his memoir: "I had come to this peasant land with my nice big American pot to piss in."



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... the author's rise to artistic glory, when he finally found in his home at Levington Park, with eleven toilets, a pot to piss in.



McGill 1987: Chính Hoàng's Ph.D. defense

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McGill 1987: Chính Hoàng's Ph.D. defense



Professor Cook first from the left (in the doorway)

THE BONDAGE NUMBER OF A GRAPH

John Frederick FINK, Michael S. JACOBSON, Lael F. KINCH and
John ROBERTS

Department of Mathematics, University of Louisville, Louisville, KY 40292, USA

Received 2 December 1988

A set D of vertices in a graph G is a *dominating set* if each vertex of G that is not in D is adjacent to at least one vertex of D . The minimum cardinality among all dominating sets in G is called the *domination number* of G and denoted $\sigma(G)$. We define the *bondage number* $b(G)$ of a graph G to be the cardinality of a smallest set E of edges for which $\sigma(G - E) > \sigma(G)$. Sharp bounds are obtained for $b(G)$, and the exact values are determined for several classes of graphs.

THE DISCIPLINE NUMBER OF A GRAPH

V. CHVÁTAL

Department of Computer Science, Rutgers University, New Brunswick, NJ 08903, USA

W. COOK

Graduate School of Business, Columbia University, New York, NY 10027, USA

Received 2 December 1988

1. Introduction

The *domination number* $\gamma(G)$ of a graph G is the size of a smallest set D of vertices such that every vertex outside D has at least one neighbour in D ; Fink, Jacobson, Kinch, and Roberts [4] defined the *bondage number* $b(G)$ of a graph G as the least number of edges whose deletion makes $\gamma(G)$ increase. As we are about to point out, computing $b(G)$ amounts to solving an integer linear program.

Define a *whip* in a graph G as any spanning subgraph F of G such that each component of F is a star and F has precisely $\gamma(G)$ components; let $E(G)$ denote the set of edges of G and let $W(G)$ denote the set of all whips in G . Obviously, $b(G)$ is the optimal value of the problem

$$\begin{aligned} & \text{minimize } \sum \{x_e : e \in E(G)\} \\ & \text{subject to } \sum \{x_e : e \in E(F)\} \geq 1 \quad \text{for all } F \text{ in } W(G), \\ & \quad x_e \geq 0 \quad \text{for all } e \text{ in } E(G). \\ & \quad x_e = \text{integer} \quad \text{for all } e \text{ in } E(G). \end{aligned} \tag{1}$$

By the *fractional bondage number* $b^*(G)$ we shall mean the optimal value of the 'linear programming relaxation' of (1),

$$\begin{aligned} & \text{minimize } \sum \{x_e : e \in E(G)\} \\ & \text{subject to } \sum \{x_e : e \in E(F)\} \geq 1 \quad \text{for all } F \text{ in } W(G), \\ & \quad x_e \geq 0 \quad \text{for all } e \text{ in } E(G). \end{aligned} \tag{2}$$

By the duality theorem of linear programming, $b^*(G)$ equals the optimal value of

the dual of (2),

$$\begin{aligned} & \text{maximize } \sum \{y_F: F \in W(G)\} \\ & \text{subject to } \sum \{y_F: e \in E(F)\} \leq 1 \quad \text{for all } e \text{ in } E(G), \\ & \quad y_F \geq 0 \quad \quad \quad \text{for all } F \text{ in } W(G). \end{aligned} \tag{3}$$

Since (3) can be seen as the linear programming relaxation of

$$\begin{aligned} & \text{maximize } \sum \{y_F: F \in W(G)\} \\ & \text{subject to } \sum \{y_F: e \in E(F)\} \leq 1 \quad \text{for all } e \text{ in } E(G), \\ & \quad y_F \geq 0 \quad \quad \quad \text{for all } F \text{ in } W(G), \\ & \quad y_F = \text{integer} \quad \quad \quad \text{for all } F \text{ in } W(G), \end{aligned} \tag{4}$$

problems (1) and (4) are in a sense dual. Therefore we refer to the optimal value of (4) as the *discipline number* $\text{dis}(G)$ of G .

We have

$$1 \leq \text{dis}(G) \leq b^*(G) \leq b(G) \tag{5}$$

for all graphs G . Apart from establishing upper bounds on $b(G)$, Fink et al. computed the bondage number of cycles, paths, and complete multipartite graphs and studied the bondage number of trees (several of these results can also be found in Bauer, Harary, Nieminen, and Suffel [1]). The purpose of this paper is to provide ties with analogous results for the fractional bondage number and for the discipline number.

2. The fractional bondage number

The principle restraining device of this section goes as follows.

Theorem 1. *If G has n vertices and m edges then $b^*(G) \leq m/(n - \gamma(G))$.*

Proof. Observe that the constraints of (2) are satisfied by $x_e = 1/(n - \gamma(G))$ for all e . \square

As usual, let $\Delta(G)$ denote the largest degree of a vertex in G . Fink et al. conjectured that $b(G) \leq \Delta(G) + 1$.

Theorem 2. $b^*(G) \leq \Delta(G)$.

Claim 5. If $k \geq 3$, then $|Q| \geq 4$.

Proof of Claim 5. Assume the contrary: $k \geq 3$ but $|Q| \leq 3$. Since G has at least four vertices, some vertex w lies outside Q ; since F_1, F_2, \dots, F_k are edge-disjoint, w is adjacent to at least k distinct vertices in Q . Hence $|Q| = k = 3$. Now no S_j can include a vertex from Q and a vertex w outside Q (w has to be adjacent to at least three distinct vertices in Q); since $|S_j| \geq 2$ for all j , it follows that $Q = S_j$ for some j . Finally, this S_j includes some vertex w distinct from u_1 and v_1 , a contradiction: w must be adjacent to at least one of u_1 and v_1 . \square

Claim 6. If $k \geq 4$, then $k \leq a + \lfloor b/2 \rfloor$.

Proof of Claim 6. By virtue of Claim 4, we only need show that $|Q| = 2k$. For this purpose, assume the contrary: without loss of generality $u_1 = u_2$. Write

$$Q_0 = \{u_1, v_1, u_2, v_2, u_3, v_3, u_4, v_4\}$$

and consider the graph H_0 whose set of vertices is Q_0 , two vertices being adjacent in H_0 if and only if they are adjacent in some F_i with $1 \leq i \leq 4$. Since each F_i with $1 \leq i \leq 4$ contributes $|Q_0| - 2$ edges to H_0 , we have

$$4(|Q_0| - 2) \leq \binom{|Q_0|}{2};$$

observing that $|Q_0| \leq 7$ (since $u_1 = u_2$) and $|Q_0| \geq 4$ (by Claim 5), we conclude that $|Q_0| = 7$. Now H_0 has twenty edges, which is a contradiction: $\binom{7}{2} = 21$ and no u_i with $2 \leq i \leq 4$ is adjacent to v_i in H_0 . \square

Claim 7. If $k = 3$, then $k \leq a + \lfloor b/2 \rfloor$ or $a + b \geq 4$.

Proof of Claim 7. Claim 4 allows us to assume that $|Q| \leq 5$; Claim 5 guarantees that $|Q| \geq 4$. Defining H as in the proof of Claim 4, observe that H has $3(|Q| - 2)$ edges. It follows that H (and hence also G) contains four pairwise adjacent vertices. \square

Claim 8. If $a = 0$ and $b = 2$, then $k \neq 2$.

Proof of Claim 8. Assume the contrary: $k = 2$ but $a = 0$ and $b = 2$. Claim 4 implies that $|Q| \leq 3$ and so, without loss of generality, $u_1 = u_2 \in S_1$. Since S_1 includes a vertex distinct from both u_1, v_1 but adjacent to at least one of them, we must have $v_1 \in S_2$; a symmetric argument shows that $v_2 \in S_2$. But then S_2 includes a vertex outside Q and adjacent to only one vertex in Q , a contradiction. \square

This ties down the proof of Theorem 9. \square

The reader interested in additional results in a similar vein is directed to [2, Chapter 5].

Acknowledgement

We thank the two referees for their thoughtful comments which helped to improve the presentation of our results.

References

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- [7] D.A.F. de Sade, *Oeuvres Completes*, 2–3 (J.-J. Pauvert, Paris, 1955).

December 1987







LESSON #1:
MISCHIEF IS A POWERFUL ENGINE OF DISCOVERY

January 1988

David Applegate





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February 28, 1988

Compaq DeskPro 386

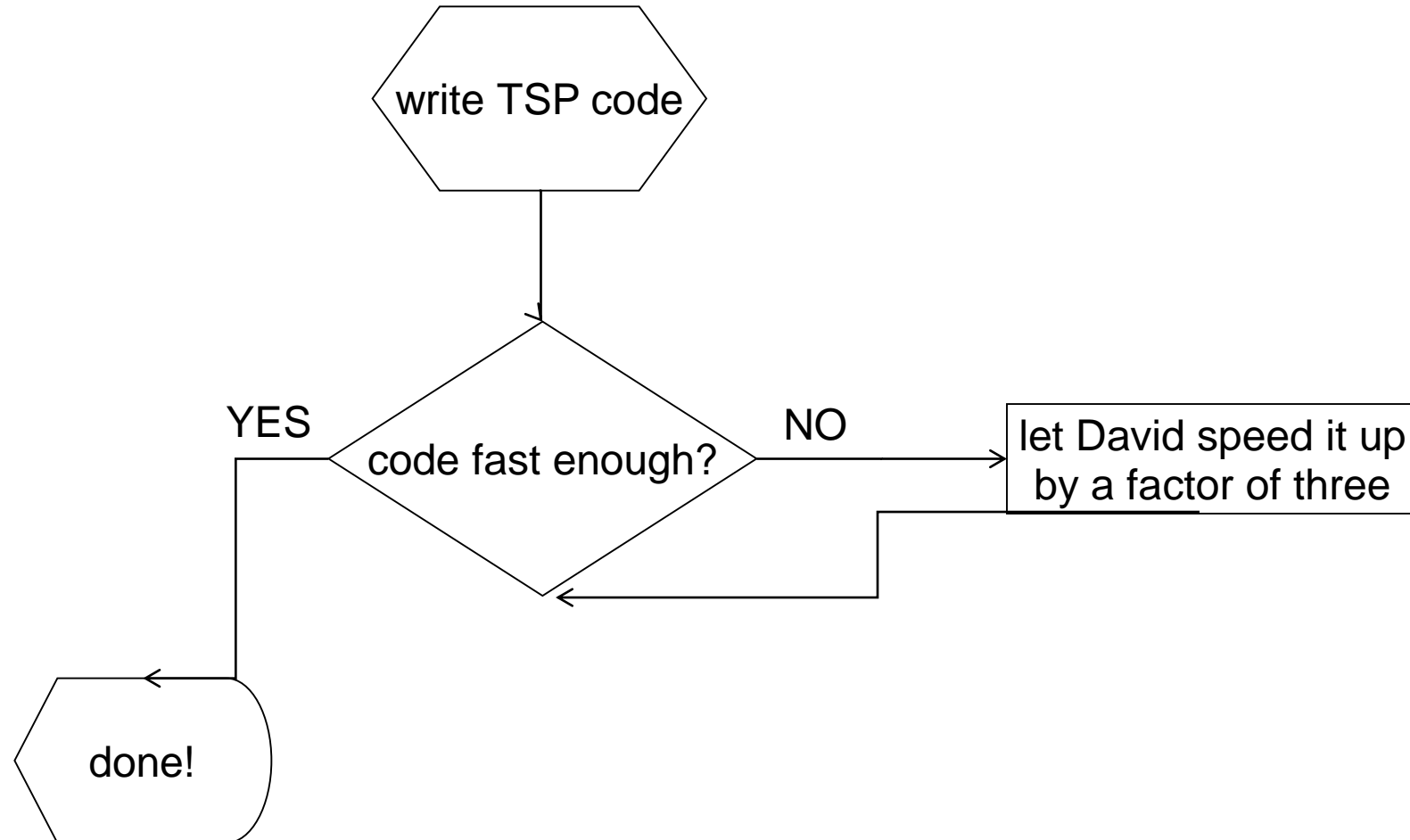
80387

DOS 3.3

\$7387.85

Bill Cook's original TSP algorithm

Bill Cook's original TSP algorithm



1988 – 1989: The Age of Innocence

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Battle cry:

If everyone does it in a certain way, we will do it differently!!

OUT WITH THE OLD, IN WITH THE NEW

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LESSON #3:

SOMETIMES THE OLD IS USED FOR A GOOD REASON

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Case in point:

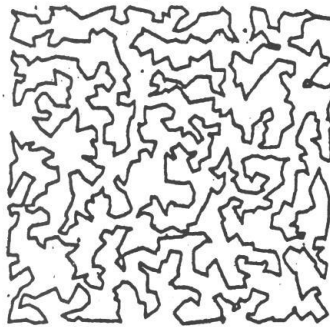
The simplex method of linear programming

The New York Times, March 12, 1991

Math Problem, Long Baffling, Slowly Yields

The traveling salesman problem still isn't solved, but computers can now get most answers.

By GINA KOLATA



A.T.&T.

Approximation of best route for 1,000 cities.

A CENTURY-OLD math problem of notorious difficulty has started to crumble. Even though an exact solution still defies mathematicians, researchers can now obtain answers that are good enough for almost all practical applications.

The traveling salesman problem, as it is known, crops up in many practical applications, from the design of computer

chips to the designation of work orders in factories.

Brute number-crunching by computers can now produce answers to most such problems, even though not an immaculate solution. "Everybody likes to point to the

The New York Times, March 12, 1991

Contests Among Investigators

Solving traveling salesman problems has become something of an obsession for aficionados. Last year, for example, Dr. Robert Bixby of Rice University invited about 50 groups of investigators to a meeting that focused only on traveling salesman problems. Investigators brought their computer programs to the meeting and Dr. William Pulleybank of I.B.M.'s Thomas J. Watson Research Center in Yorktown Heights, N.Y., challenged them with a 783-city problem that he had devised, he thought, to be especially difficult.

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The Texas Shootout: April 22-24, 1990

[Facsimile of a transparency used by V.C. in his talk in Texas]

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SURVEY OF PROGRESS IN TSP SOLVING

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1990	Applegate, Chvátal, and Cook	17 cities

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The winning team, made up of Dr. William Cook of Bell Communications Research, and Dr. David Applegate and Dr. Vasek Chvatal of Rutgers University, solved the problem in half an hour. "It turned out to be easy," Dr. Cook said.

Bob Bixby

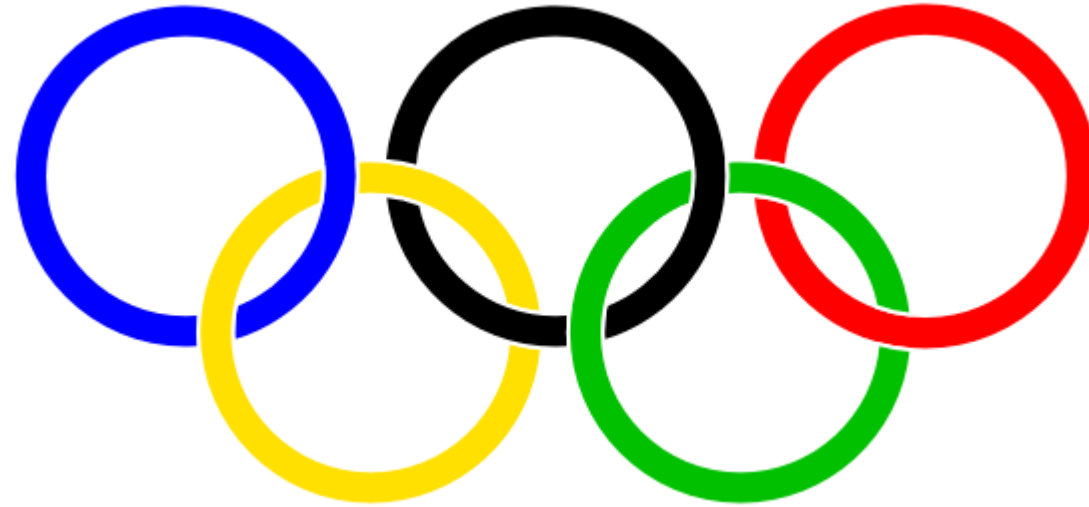


Bob Bixby



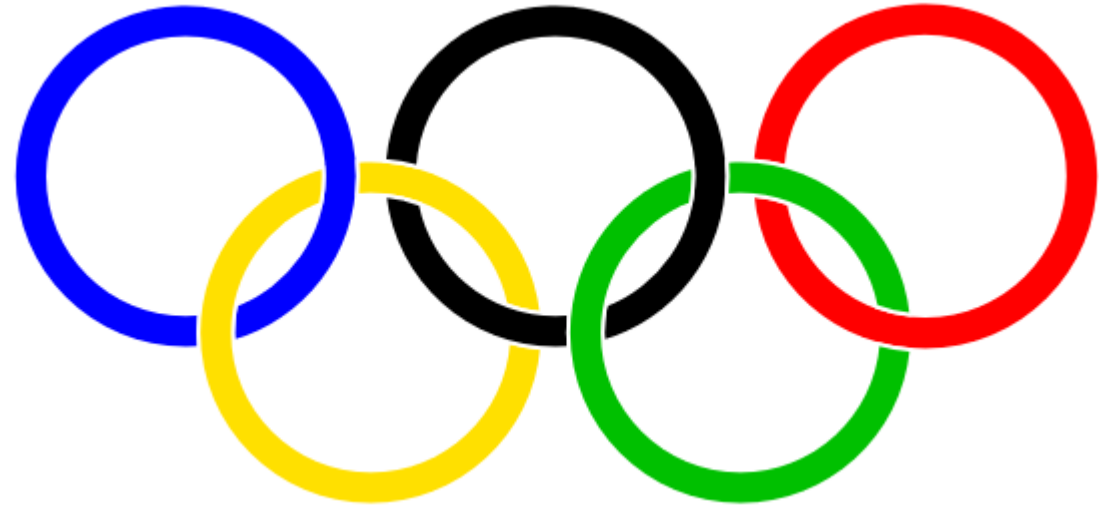
CPLEX

The sport of solving the TSP



My daddy can beat up your daddy

The sport of solving the TSP



The goal: pcb3038

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Essential constituents of research in computational mathematics:

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January 1992: The Great Dismay

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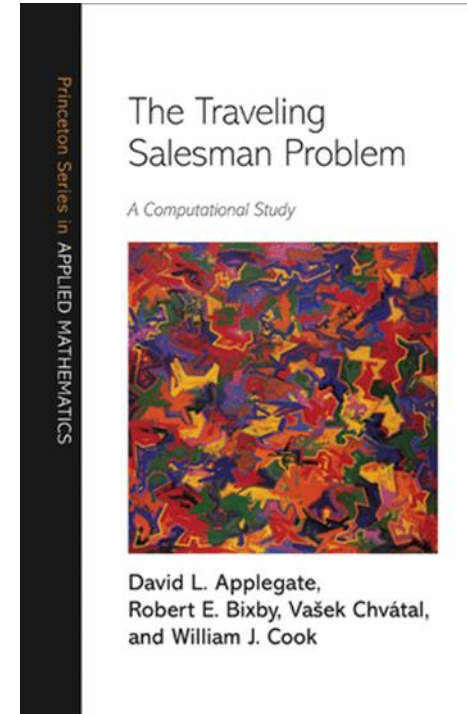
Desperate last-ditch efforts:

- combs from consecutive ones (Chapter 8)
- cut tightening (Section 10.1)
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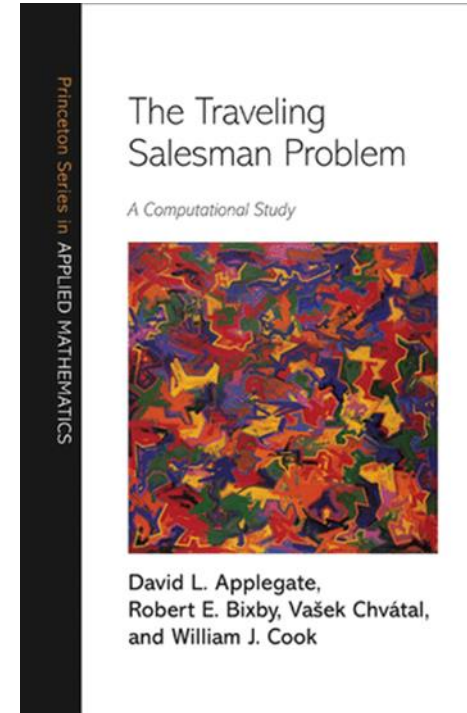


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April 1992: pcb3038 falls

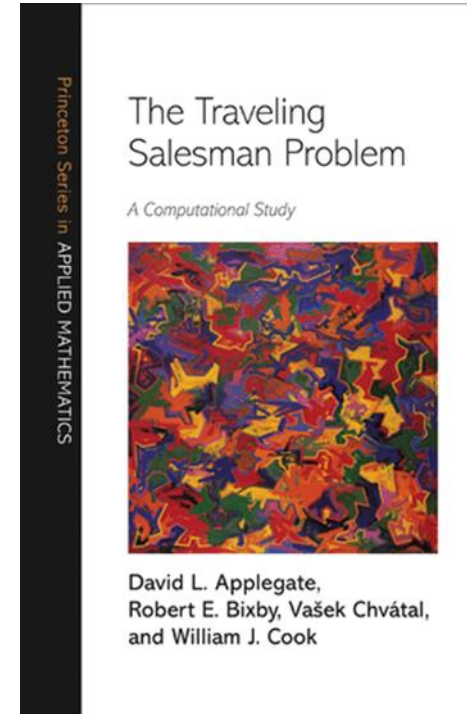


January 1992: The Great Dismay

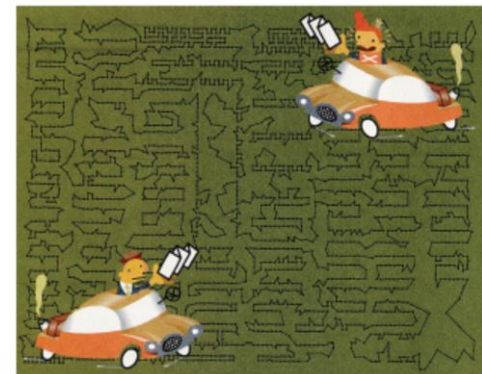
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The Discover Magazine, January 1993
“A new TSP record, 3,038 cities”



June 1993: fnl4461



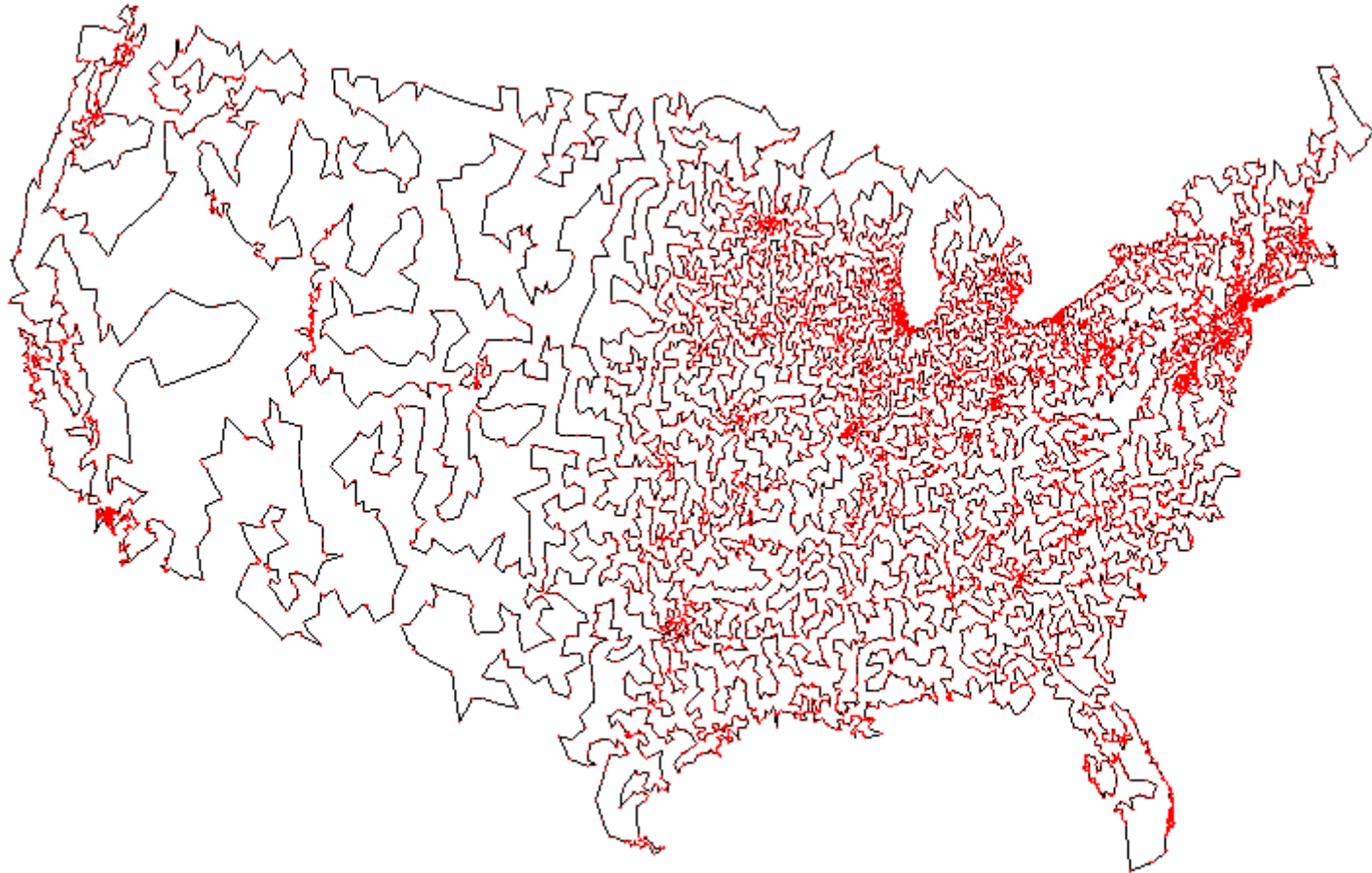
January 1996

Another breakthrough: local cuts (Chapter 11)

January 1996

Another breakthrough: local cuts (Chapter 11)

1998: usa13509



TSPLIB

a280	d2103	gr48	lin318	pr264	si1032
ali535	d15112	gr96	linhp318	pr299	st70
att48	d18512	gr120	nrw1379	pr439	swiss42
att532	dantzig42	gr137	p654	pr1002	ts225
bayg29	dsj1000	gr202	pa561	pr2392	tsp225
bays29	eil51	gr229	pcb442	rat99	u159
berlin52	eil76	gr431	pcb1173	rat195	u574
bier127	eil101	gr666	pcb3038	rat575	u724
brazil58	fl417	hk48	pla7397	rat783	u1060
brd14051	fl1400	kroA100	pla33810	rd100	u1432
brg180	fl1577	kroB100	pla85900	rd400	u1817
burma14	fl3795	kroC100	pr76	rl1304	u2152
ch130	fnl4461	kroD100	pr107	rl1323	u2319
ch150	fri26	kroE100	pr124	rl1889	ulysses16
d198	gil262	kroA150	pr136	rl5915	ulysses22
d493	gr17	kroB150	pr144	rl5934	usa13509
d657	gr21	kroA200	pr152	rl11849	vm1084
d1291	gr24	kroB200	pr226	si175	vm1748
d1655		lin105		si535	

The red instances were unsolved in 1991

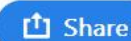
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bayg29	dsj1000	gr202	pa561	pr2392	tsp225
bays29	eil51	gr229	pcb442	rat99	u159
berlin52	eil76	gr431	pcb1173	rat195	u574
bier127	eil101	gr666	pcb3038	rat575	u724
brazil58	fl417	hk48	pla7397	rat783	u1060
brd14051	fl1400	kroA100	pla33810	rd100	u1432
brg180	fl1577	kroB100	pla85900	rd400	u1817
burma14	fl3795	kroC100	pr76	rl1304	u2152
ch130	fnl4461	kroD100	pr107	rl1323	u2319
ch150	fri26	kroE100	pr124	rl1889	ulysses16
d198	gil262	kroA150	pr136	rl5915	ulysses22
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d657	gr21	kroA200	pr152	rl11849	vm1084
d1291	gr24	kroB200	pr226	si175	vm1748
d1655		lin105		si535	

The red instances were unsolved in 2006

Large TSP instances solved in the new millennium

April 2001	d15112	34 CPU years
December 2002	it16862	11 CPU years
March 2004	brd14051	5 CPU years
May 2004	sw24978	85 CPU years
October 2004	pla33810	16 CPU years
March 2005	d18512	57 CPU years
April 2006	pla85900	136 CPU years



年の9月に日本オペレーションズリサーチ学会の会員に中央大学の伊理正大を連れて
Chvátal からの協力依頼の電子メールが流れたんだ。一部割愛して引用しよう。

Date: Tue, 28 Sep 93 18:00:00

Subject: ohisashiburi desu ne

Dear Professor Iri,

you may be surprised to get this message out of the blue. I wonder if you would do me a favor. David Applegate, Bob Bixby, Bill Cook, and I have written a computer code to solve travelling salesman problems. Last year we have solved one with 3038 cities (from a database called TSPLIB), which was the world record at that time; a couple of months ago we have broken the record by solving a problem with 4461 cities, and now we would like to try 7397 cities. We solve those problems in parallel on different workstations; currently we are using about 65 of them. The way it works is that people open accounts for us on their machines and we run parts of our program on them whenever no one else is using them (we have a clever program that kills our computations as soon as somebody touches the keyboard or logs into the machine from home). The 65 workstations are sitting around here in New Jersey, but with our sights on the 7397 cities we would like to recruit as many new machines as we can. David will try to get us some at McGill, we are asking a few other friends in North America to do the same, and thought it would be nice if we had a few helpers in Japan as well. If you know anybody who would be willing to lend his or her machine to contribute to breaking the record again, would you, please, pass this message on to him or her? (This is not terribly important, so please, don't go to too much trouble to help us.)

Thank you very much!

Ogenki de,

Vasek Chvatal

何台ものワークステーションをつなげて世界記録を目指しているが、日本でも協力し

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CONCORDE'S YOUNGER HELPERS

- An implementation of Adam Letchford's *domino parity constraints* developed by Bill Cook, Daniel Espinoza, and Marcos Goycoolea. Used in the solution of d18512, pla33810, and pla85900.
- Keld Helsgaun's refinement of the Lin-Kernighan heuristic. Used in the solution of sw24978 and pla85900.

In February 2009, Robert Bosch created a 100,000-city problem.

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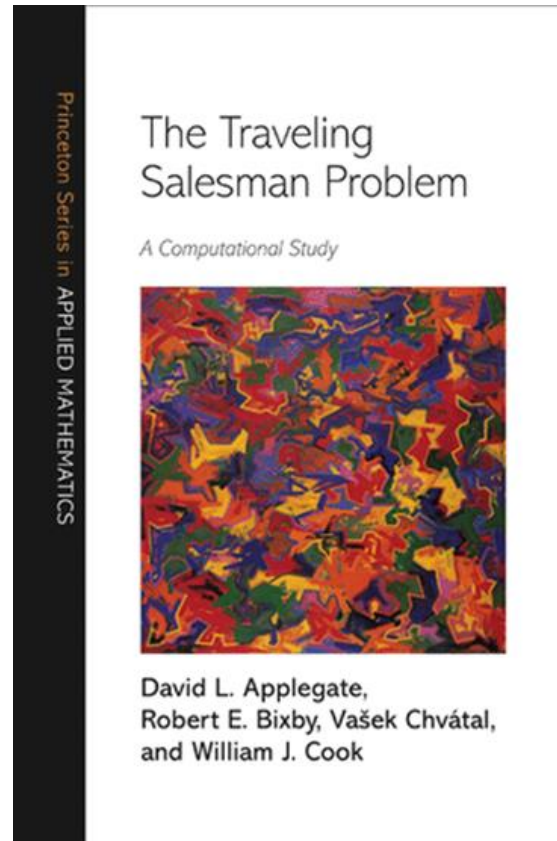
Bill Cook offers \$1000 for a shorter tour.

Bill Cook's enchanting website:

<http://www.math.uwaterloo.ca/tsp/>

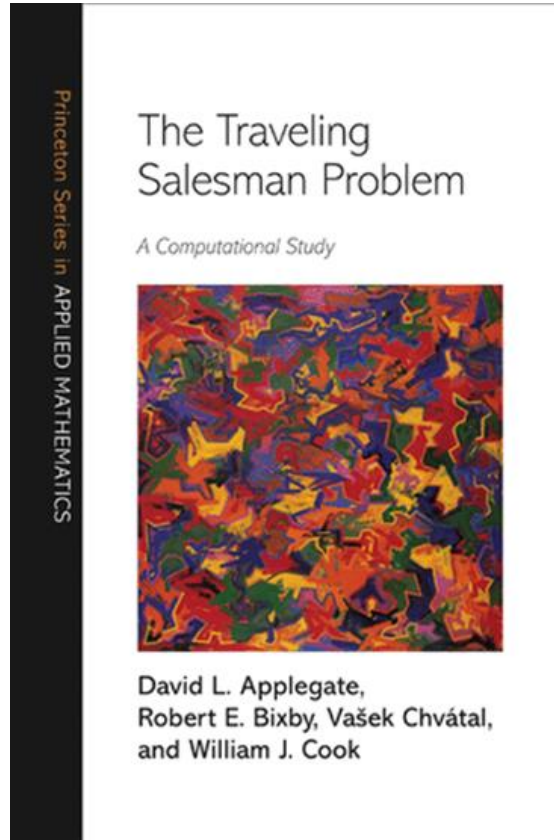
It is never too early to think of Christmas gifts

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