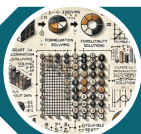


Integer programming & The Travelling Salesman Problem

Orr Zwebner
Baruch Hanya

Agenda



TSP and IP - Definitions



TSP Formulations – MTZ, DFJ



Implementation and results



Competitive quiz



Integer Programming

$IP \subset LP$ s.t all variables are integer, Unlike **M**ixed **L**inear **I**nteger **P**rogramming

IP **standard** form:

maximize $c^T x$

Subject to:

$$Ax + s \leq b$$

$$s \geq 0$$

$$x \geq 0$$

Where $x \in \mathbb{Z}^n, c \in \mathbb{R}^n, s, b \in \mathbb{R}^m, a \in \mathbb{R}^{m \times n}$



Integer Programming

Example - **Maximum independent set:**

Given a graph $G = (V, E)$, find a maximum size $S \subseteq V$ such that no 2 vertices in S have an edge between

$$(\forall v, u \in S \rightarrow (u, v) \notin E).$$

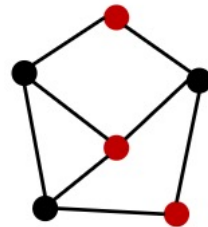
Maximize $\sum_{v \in V} x_v$

Subject to:

$$x_v \in \{0, 1\}$$

$$x_u + x_v \leq 1 \quad \forall (u, v) \in E$$

MAX IS \leq_P **IP** \rightarrow IP is NP-COMPLETE (**MAX IS** is NP-Hard)





Travelling Salesman Problem - TSP

- **Goal :** Find the **shortest route** visiting each city once and returning to the starting point
- **Challenge:** Number of possible routes grows fast with more cities
- **Approach:** Approximation methods offer quick and practical solutions
- **Important Note :** The starting point does not affect the optimality of the solution



In our case the running time is approximately:

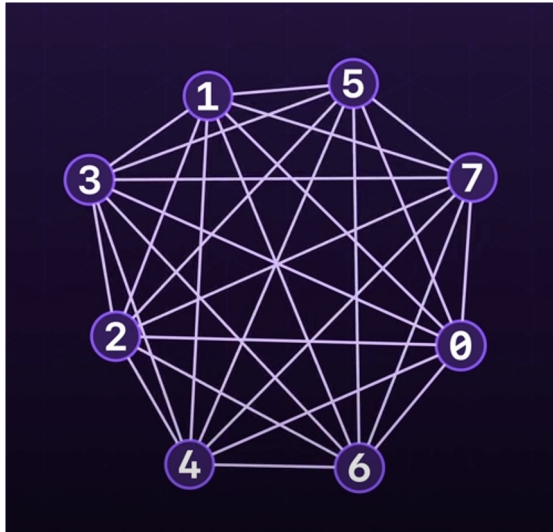
$\sim 10^{61}$

TSP is an NP-complete problem

Travelling Salesman Problem - TSP

Exponential time - it involves evaluating all possible permutations of the cities, which is

$(n-1)!$ permutations



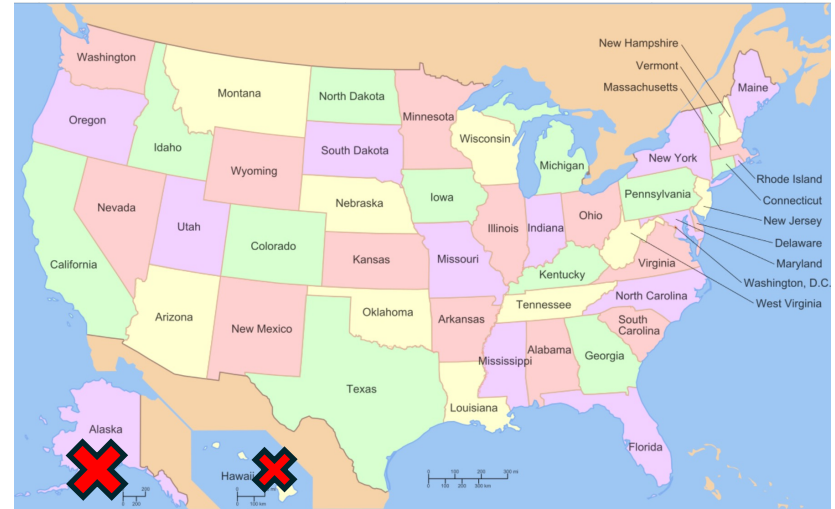
Naive Method

Number of Cities (n)	Possible Permutations $((n-1)!)!$	Approximate Running Time
3	2	Very fast
4	6	Very fast
5	24	Very fast
6	120	Very fast
7	720	Fast
8	5040	Moderate
9	40320	Moderate
10	362880	Slow
11	3628800	Very slow
12	39916800	Extremely slow



Our Dataset

- We took **48 Capitals of USA's states**
Excluding Alaska and Hawaii to **simplify the dataset**
- We calculated the distances using **Euclidean distance by KMs**





TSP - Formulation in IP

Optimization problem:

$$\text{Minimize } \sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$$

Where:

$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$

$$c_{ij} > 0 \quad \text{The distance from city } i \text{ to city } j$$





TSP - Formulation in IP

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\} \quad i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1 \quad i = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1 \quad j = 1 \dots n;$$



$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$
$$c_{ij} > 0 \quad \text{The distance from city } i \text{ to city } j$$



TSP - Formulation in IP

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

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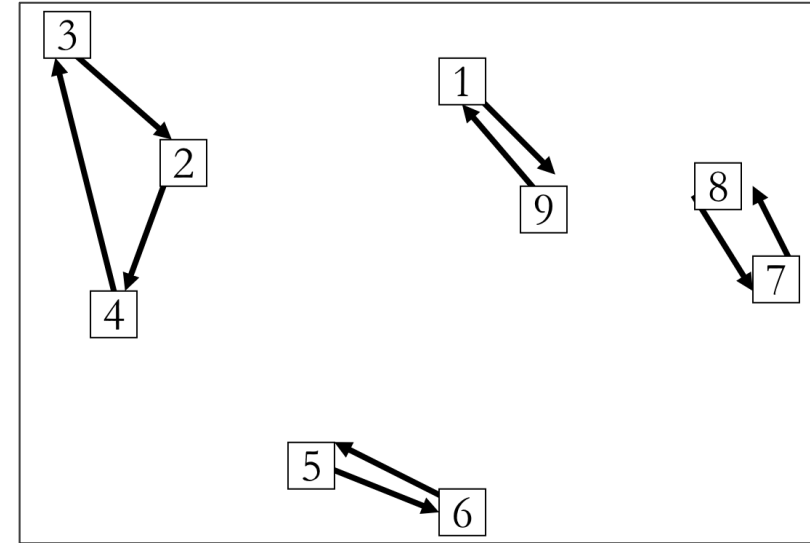
$$i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

$$j = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$i = 1 \dots n;$$



$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$

$$c_{ij} > 0 \quad \text{The distance from city } i \text{ to city } j$$



TSP – Miller Tucker Zemlin formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\}, u_i \in \mathbb{N} \quad i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1 \quad j = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1 \quad i = 1 \dots n;$$

$$\text{If } x_{ij} = 1: \quad i, j = 2 \dots n;^*$$

$$u_j = u_i + 1$$



$x_{ij} = \begin{cases} 1 \\ 0 \end{cases}$	The path goes from city i to city j Otherwise
$c_{ij} > 0$	The distance from city i to city j



TSP – Miller Tucker Zemlin formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

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$$x_{ij} \in \{0, 1\}, u_i \in \mathbb{N} \quad i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1 \quad j = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1 \quad i = 1 \dots n;$$

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad 2 \leq i \neq j \leq n$$

$$2 \leq u_i \leq n \quad 2 \leq i \leq n$$



$x_{ij} = \begin{cases} 1 \\ 0 \end{cases}$	The path goes from city i to city j Otherwise
$c_{ij} > 0$	The distance from city i to city j



TSP – Miller Tucker Zemlin formulation

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad \text{s.t.} \quad x_{ij} = 0:$$

$$u_i - u_j + 1 \leq (n - 1)(1 - 0) = n - 1 \qquad 2 \leq i \neq j \leq n$$

$$*2 \leq u_i \leq n \qquad 2 \leq i \leq n$$



TSP – Miller Tucker Zemlin formulation

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad \text{s.t.} \quad x_{ij} = 0:$$

$$u_i - u_j + 1 \leq (n - 1)(1 - 0) = n - 1 \qquad 2 \leq i \neq j \leq n$$

$$2 \leq u_i \leq n \qquad 2 \leq i \leq n$$

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad \text{s.t.} \quad x_{ij} = 1:$$

$$u_i - u_j + 1 \leq (n - 1)(1 - 1) = (n - 1) * 0 = 0 \qquad 2 \leq i \neq j \leq n$$

$$\text{If } x_{ij} = 1: \qquad i, j = 2 \dots n, *$$

$$u_j = u_i + 1$$

$$u_j = u_j + 1 \rightarrow u_i - u_j + 1 = 0 \rightarrow u_i - u_j + 1 \leq 0$$

$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$

$$c_{ij} > 0 \qquad \text{The distance from city } i \text{ to city } j$$



TSP – Miller Tucker Zemlin formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\}, u_i \in \mathbb{N} \quad i, j = 1 \dots n;$$

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$$\sum_{i=1, i \neq j}^n x_{ij} = 1 \quad i = 1 \dots n;$$

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad 2 \leq i \neq j \leq n$$

$$2 \leq u_i \leq n \quad 2 \leq i \leq n$$



$x_{ij} = \begin{cases} 1 \\ 0 \end{cases}$	The path goes from city i to city j Otherwise
$c_{ij} > 0$	The distance from city i to city j



Danzig Fulkerson Johnson formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\}$$

$$i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

$$j = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$j = 1 \dots n;$$

$$\sum_{i,j \in S, i \neq j} x_{ij} \leq |S| - 1$$

$$\forall S \subsetneq \{1, 2, \dots, n\}$$





Danzig Fulkerson Johnson formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\}$$

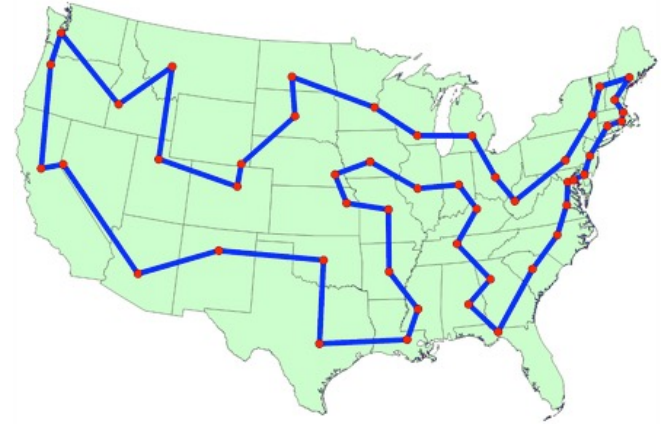
$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$i, j = 1 \dots n;$$

$$j = 1 \dots n;$$

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$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$

$$c_{ij} > 0 \quad \text{The distance from city } i \text{ to city } j$$



Danzig Fulkerson Johnson formulation

Minimize $\sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}$

Subject to:

$$x_{ij} \in \{0, 1\}$$

$$i, j = 1 \dots n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

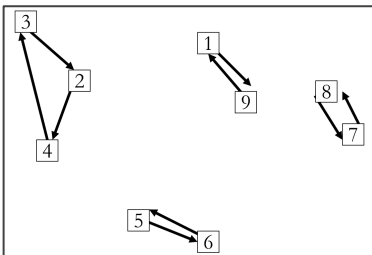
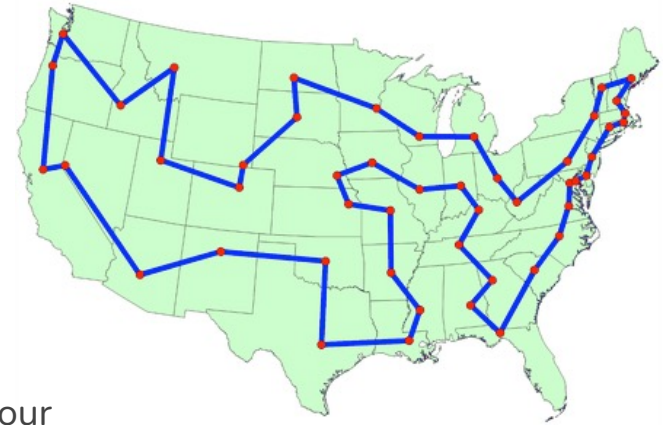
$$j = 1 \dots n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$j = 1 \dots n;$$

There is no sub-tour - The solution is a single tour

(not a union of smaller tours)



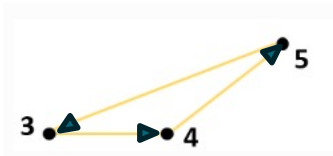
$$x_{ij} = \begin{cases} 1 & \text{The path goes from city } i \text{ to city } j \\ 0 & \text{Otherwise} \end{cases}$$
$$c_{ij} > 0 \quad \text{The distance from city } i \text{ to city } j$$



Danzig Fulkerson Johnson formulation

There is no sub-tour - The solution is a single tour =

The number of arcs between nodes in the subset should be less than the number of nodes in that subset

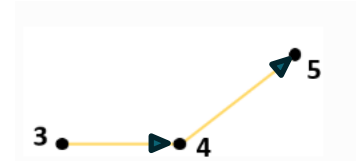
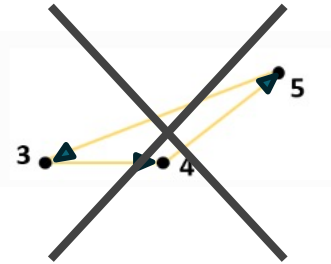




Danzig Fulkerson Johnson formulation

There is no sub-tour - The solution is a single tour =

The number of arcs between nodes in the subset should be less than the number of nodes in that subset

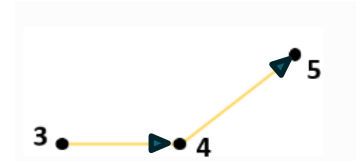
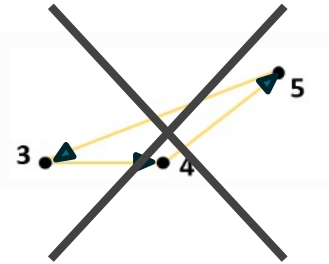




Danzig Fulkerson Johnson formulation

There is no sub-tour - The solution is a single tour =

The number of arcs between nodes in the subset should be less than the number of nodes in that subset



x_{ij} - An arc between nodes i and j

$$x_{34} + x_{45} + x_{53} \leq 2$$

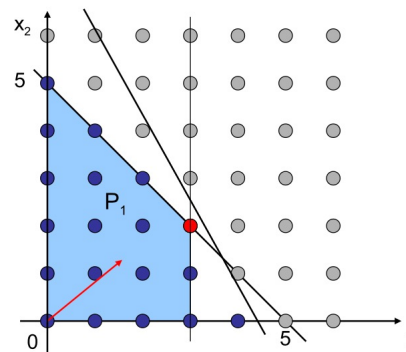
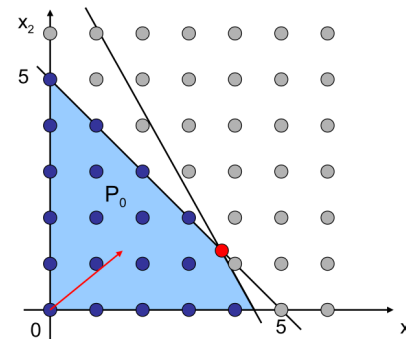
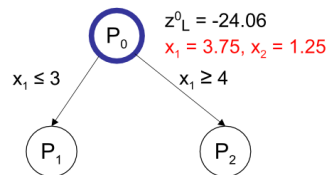


Branch and bound

$$\begin{aligned}
 z_I^1 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\
 \text{s.t.} \quad & x_1 + x_2 \leq 5 \\
 & 10x_1 + 6x_2 \leq 45 \\
 & x_1 \leq 3 \\
 & x_1, x_2 \geq 0 \\
 & x_1, x_2 \in \mathbb{Z}
 \end{aligned}
 \quad (P_1)$$

$$\begin{aligned}
 z_I^2 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\
 \text{s.t.} \quad & x_1 + x_2 \leq 5 \\
 & 10x_1 + 6x_2 \leq 45 \\
 & x_1 \geq 4 \\
 & x_1, x_2 \geq 0 \\
 & x_1, x_2 \in \mathbb{Z}
 \end{aligned}
 \quad (P_2)$$

This operation is called a *branching on variable x_1* . Note that the solution $(3.75, 1.25)$ does not belong to the linear relaxation of (P_1) or (P_2) . We can represent the subproblems and the corresponding bounds by means of a tree, called the Branch-and-Bound tree.

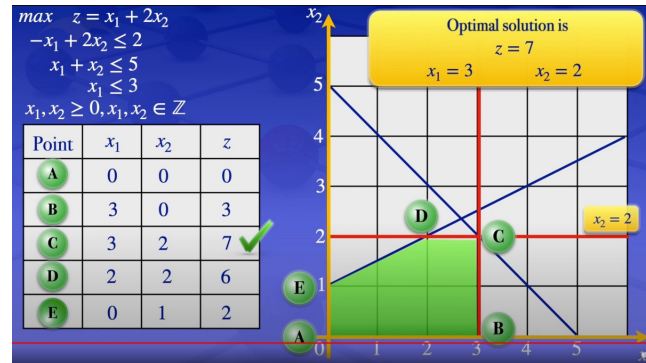
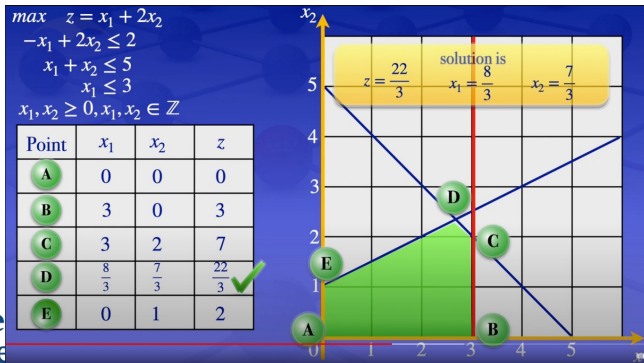


Cutting plane method

Cutting plane method

Start with the linear relaxation $\max\{c^T x \mid Ax \leq b, x \geq 0\}$.

1. Solve the current linear relaxation, and let x^* be a basic optimal solution;
2. If $x^* \in X$, then x^* is optimal for (P_I) ; STOP.
3. Otherwise, find an inequality $\alpha^T x \leq \beta$ that is valid for X and cuts off x^* ;
4. Add the inequality $\alpha^T x \leq \beta$ to the current linear relaxation and go to 1.





Solutions

NP HARD - Approximated solutions

1. Greedy randomized - Nearest neighbor
 - a. Generate a start city, always choose the closest city
 - b. Try N times, choose the best tour
 - i. Each iteration with a different start city





Solutions

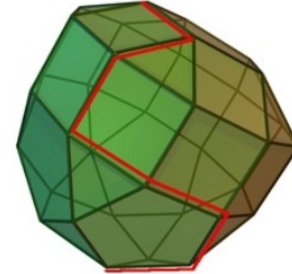
NP HARD - Approximated solutions

1. Greedy randomized - Nearest neighbor

- a. Generate a start city, always choose the closest city
- b. Try N times, choose the best tour

2. Solve LP and find a relaxed solution (simplex):

- a. Rounded
- b. Branch and bound
- c. Cutting plane method





Solutions

NP HARD - Approximated solutions

1. Greedy randomized - Nearest neighbor

a. Generate

b.

the closest city



coin-or/Cbc

COIN-OR Branch-and-Cut solver

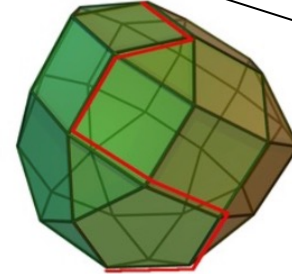
(randomized)

bound

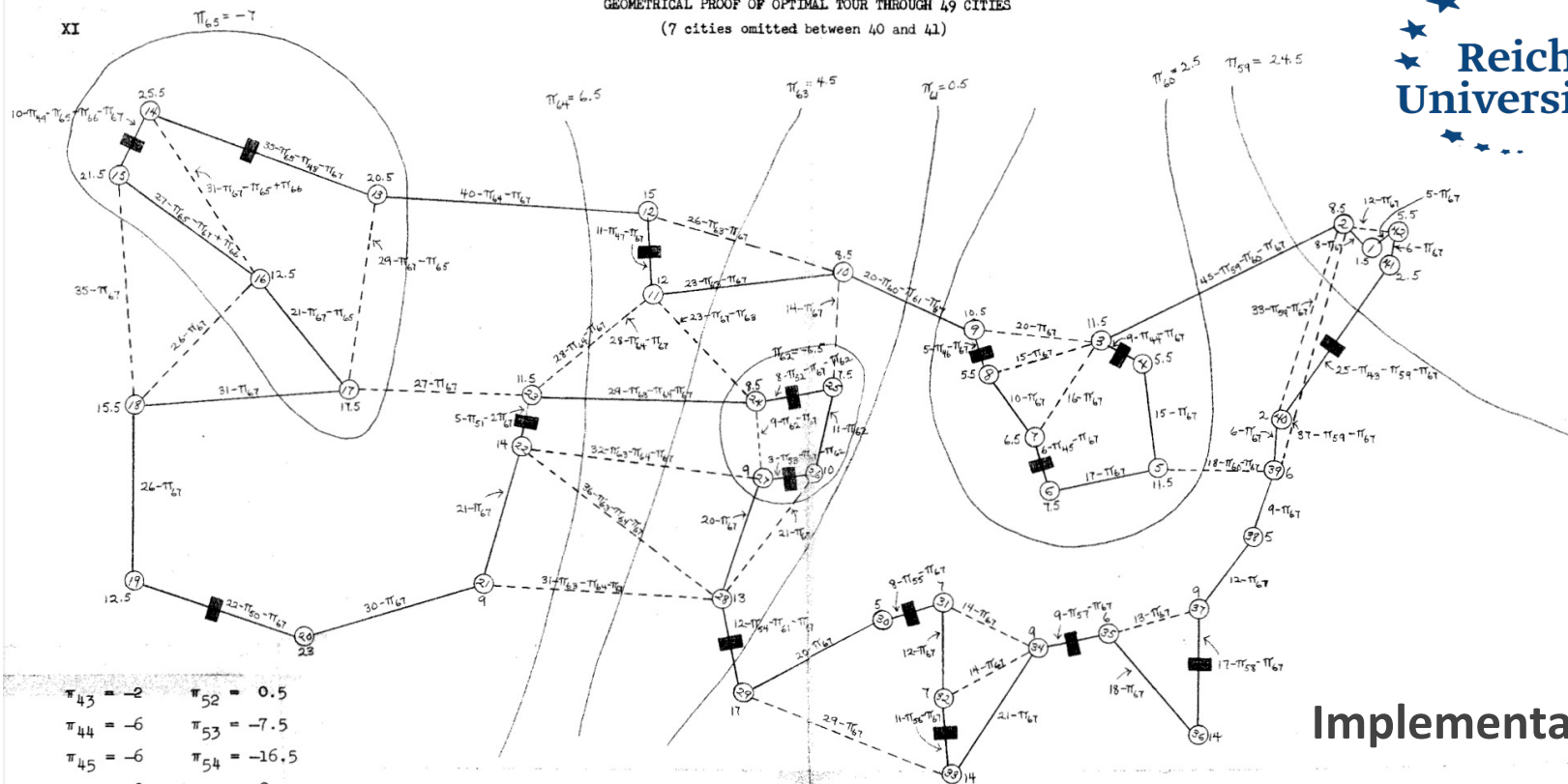
cutting plane method



GUROBI
OPTIMIZATION



GEOMETRICAL PROOF OF OPTIMAL TOUR THROUGH 49 CITIES
(7 cities omitted between 40 and 41)



$\pi_{43} = -2$	$\pi_{52} = 0.5$
$\pi_{44} = -6$	$\pi_{53} = -7.5$
$\pi_{45} = -6$	$\pi_{54} = -16.5$
$\pi_{46} = -9$	$\pi_{55} = -2$
$\pi_{47} = -14$	$\pi_{56} = -8$
$\pi_{48} = -2$	$\pi_{57} = -4$
$\pi_{49} = -26$	$\pi_{58} = -4$
$\pi_{50} = -11.5$	$\pi_{66} = -2$
$\pi_{51} = -16.5$	$\pi_{67} = -2$

Note: In comparing $\pi_{P_{1j}}$ with d_{1j} , remember that the components of P_{1j} , corresponding to the added relations must be considered. The added relations are pictured graphically above except for the last two, 66 and 67. A loop drawn about a subset of points corresponds to a relation of the kind $\sum_1 \leq n_1 - 1$; an arc separating a set of points from its complement corresponds to $\sum_{1,j} \geq 2$. For example, to compare $\pi_{P_{25,9}}$ with $d_{25,9} = 21$, observe that the segment (25,9) crosses two arcs, i.e., $P_{25,9}$ has a 1 in components 60 and 61; it has a zero in components 66 and 67 since $x_{25,9}$ does not appear in those relations, and of course, has zeros elsewhere, except for components 25 and 9. Hence $\pi_{P_{25,9}} = \pi_{25} + \pi_9 + \pi_{60} + \pi_{61} = 7.5 + 10.5 + 2.5 + 0.5 = 21 \leq d_{25,9}$. One more example: $\pi_{P_{27,25}} = \pi_{27} + \pi_{25} + \pi_{62}$, since (27,25) is contained within the loop representing the condition $\sum_{12} \leq 3$.

Implementation



Results

Model	Formula	Objective Value (KM)	Running time
Nearest neighbor	-	20,003.87	0.1 sec
CBC*	MTZ	17,585.43	1 hr
	DFJ	17,585.43	1 hr
GUROBI	MTZ	17,083.89	4 min 25.2 sec
	DFJ**	17,083.89	0.2 sec
Optimum***	-	17,083.89	-

* Time limited

** Using GUROBI lazy constraints

*** According to 2 online blogs



Conclusions

1. IP is a tool for formulating complex optimization problems
2. The simple solution is a good and efficient approximation
 - a. NN Objective value < 1.18 OPT
3. Solve IP or TSP in polynomial time and you will be a millionaire
4. Until then – Use GUROBI (with academic license)





Question?

