## Lecture V - NP Problems and Graphs

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Combinatorial Optimization

#### Previously on.

Complexity class P

Complexity class NP

TSP -Travelling Salesman Problem

# Previously on..



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Matching Problems:

- Weighted Matching;
- Maximum Matching.

# PREVIOUSLY ON ....

So far, all problems either have **algorithm** or **a mixed integer** formulation.

But what happens when **an algorithm** is not achievable nor efficient? A mixed integer formulation **might still** be possible, but would it be enough?



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## Complexity class P



They differ from **optimization problem**, because the former requires an answer that have an **optimal configuration**.

For instance: "What is the shortest path between two nodes?" vs "Is a particular path P the shortest path between these two nodes?".

**Remark**: An optimization problem has a **corresponding** decision problem.



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Class P

- *P* is the **class** of all decision problems (*X*, *Y*) for which there is a **polynomial** time algorithm.
- Given  $x \in A^*$ : compute  $f(x) \in \{0,1\}$  with time $(x) \le p(\text{size}(x))$ .

## Examples

. . .

- linear inequalities;
- shortest path;
- maximum matching;
- minimum cost flow.

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# Complexity class NP

Class NP

- decision problem (X, Y) belongs to class NP if for each  $y \in Y$  a certificate c can be verified in **polynomial time**
- usually c is a **feasible solution** to the problem
- name NP = **nondeterministic** polynomial: "guessing" certificates long enough would work
- (X, Y) can be solved by nondeterministic in polynomial time

## Examples

- integer linear inequalities  $\rightarrow c$  is feasible vector x
- knapsack  $\rightarrow c$  is a feasible set of items to take



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P vs. NP



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# Theorem $P \subseteq NP$ .

Proof.

 $(X,Y) \in P$  can be decided in polynomial time.  $\Rightarrow x$  can be used as certificate.  $\mathsf{P}$  vs.  $\mathsf{N}\mathsf{P}$ 







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TSP -Travelling Salesman Problem

Definition



Imagine a scenario, where a set of cities are expecting a visit from a **travelling** merchant.

As part of their visit, this **salesman** has to start and finish their travel in the same city, cannot visit the same city more than once and every city has to be visited in a single trip.



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Example



Figure: A lone traveller about to make important decisions

 $\rightarrow$  What is the shortest possible route?



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TSP -Travelling Salesman Problem

- Combinatorial Optimization

Such problem is called Travelling Salesman Problem, very important to the fields of theoretical computer science and operations research.

It was first described by Irish mathematician W.R. Hamilton and British mathematician **Thomas Kirkman** in the 1800s through the description of a game where the solution involved a cycle without overlapping nodes.

# History







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Initial Approach

At first glance, the first solution is to try all possibilities and choose the best solution.

 $\longrightarrow$  enumeration process

Initial <u>challenge</u>: for an instance with n cities, there are  $2^n$  possible combinations.  $\longrightarrow$  impractical.

Is there any better alternative?

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First, let us assume that the set of cities can be modelled as graph G = (V, E, f), where:

- V is the set of individual cities;
- *E* represent the paths between a pair of cities and;
- $f_{ij}$  is the cost to travel from city *i* to city *j*, for all  $(i, j) \in E$ .



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# Modelling - Variable and Objective

The choice to travel from city i to city j using a path (edge in our modelling) connecting them is modelled by our decision variable.

$$x_{ij} = \left\{ \begin{array}{ll} 1, & \text{if path goes from city } i \text{ to city } j \\ 0, & \text{otherwise} \end{array} \right\}$$

Our objective function can be also derived from the problem description:

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

where n is the number of cities (|V| = n).



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## Modelling - Constraints

Two constraints can also be derived directly from description:

**Singular** incoming degree:

$$\sum_{i=1, i\neq j}^{n} x_{ij} = 1 \quad \forall j \in \{1, \cdots, n\}$$

Singular outgoing degree:

$$\sum_{j=1, j\neq i}^{n} x_{ij} = 1 \quad \forall i \in \{1, \cdots, n\}$$

Those constraints are characterize **paths**.



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# Modelling - Split version

These constraints imposed that every city is visited only once.

However, they do not guarantee that there is a single trip will connecting all cities.

For instance:



Figure: Two solutions that do not violate the previous constraints, but only one has a single trip.



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# Modelling Decisions - MTZ vs DFJ



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There are two main strategies to prevent separate tour (subtour) from our potential solution: Miller-Tucker-Zemlin and Dantzig-Fulkerson-Johnson.

Both impose the presence of a single tour using **linear constraints**.



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TSP -Travelling Salesman Problem

Requires an additional variable to track which city has been visited starting from initial city i = 1.

 $\longrightarrow$  By setting  $u_j > u_i$ , it determines the **order** of visiting each city (city j will be visited after city i).

This leads to following requirement:

$$u_j \ge u_i + 1$$
 if  $x_{ij} = 1$ 

which can encapsulated as the following constraints:

$$\begin{aligned} u_i - u_j + 1 &\leq (n-1)(1 - x_{ij}) & \text{for } 2 &\leq i \neq j \leq n \\ 2 &\leq u_i \leq n & \text{for } 2 \leq i \leq n \end{aligned}$$



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## Miller-Tucker-Zemlin



Subject to:

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

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

$$u_i - u_j + 1 \le (n-1)(1 - x_{ij})$$

$$2 \le u_i \le n$$

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

 $\forall i \in \{1, \cdots, n\}$  $2 \le i \ne j \le n$  $2 \le i \le n$  $i, j \in \{1, \cdots, n\}$ 

 $\forall j \in \{1, \cdots, n\}$ 



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# Dantzig-Fulkerson-Johnson

Impose an extra requirement that eliminates all subset of nodes to create a subtour.

$$\sum_{i \in Q} \sum_{j \neq i, j \in Q} x_{ij} \le |Q| - 1 \qquad \forall Q \subset \{1, \cdots, n\}, |Q| \ge 2$$



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# Dantzig-Fulkerson-Johnson

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

Subject to:

$$\sum_{i=1, i \neq j}^{n} x_{ij} = 1 \qquad \forall j \in \{1, \cdots, n\}$$

$$\sum_{j=1, j \neq i}^{n} x_{ij} = 1 \qquad \forall i \in \{1, \cdots, n\}$$

$$\sum_{i \in Q} \sum_{j \neq i, j \in Q} x_{ij} \leq |Q| - 1 \qquad \forall Q \subset \{1, \cdots, n\}, |Q| \geq 2$$

$$x_{ij} \in \{0, 1\} \qquad i, j \in \{1, \cdots, n\}$$



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