## Introduction to Operations Research

CERMICS, ENPC

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Traveling salesman problem


Mister Supersales must plan his tour

Traveling salesman problem


Mister supersale has planned his tour

## Traveling salesman problem

Is there an optimal solution?

## Traveling salesman problem

Is there an optimal solution?
Yes: finite set of solution
Can we enumerate all the solutions?

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Can we enumerate all the solutions?
With 25 cities, we have 24 ! $=24 \times 23 \times 22 \times \cdots \times 2 \times 1$ possibilities, that is, around $6.204 \times 10^{25}$ possibilities.

Using paper and pencil, testing 1 possibility per second, requires around $1.976 \times 10^{16}$ years.

Testing 1 million possibilities per second with a computer, requires 19 billion years.

# 1. What is Operations Research 

2. Syllabus
3. Graphs
4. Complexity

## Operations Research

The traveling salesman problem is one of the most famous Operations Research problem.

Operations Research (OR): mathematical discipline that deals with the optimal allocation of resources (typically in firms).

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Why this name? What was it invented for?

And for those who search fundings for their start-up

Data mining is descriptive analytics:

- My data contains A and B

Machine Learning is predictive analytics:
$>$ if $A$ happens, then $B$ will happen

Operations Research is prescriptive analytics:

- if I want B to happen, then I must do A

As such, OR is a key tool of artificial intelligence

More prosaically, Big Data have multiplied the fields of applications of OR, and there are too few experts today

## Examples of Operations Research problems

- Find the best tour
- Plan the best timetable
- Find the most resilient network
- Fill a container optimally

- Locate facilities/warehouses optimally M
- Schedule jobs on machines

The founding fathers

Monge, Blackett, Dantzig.


Fast growth since the fifties
academia (maths, computer science),
industry (supply chain, transport, telecommunications, etc.)

Two keywords: Modeling et Optimization.

Performances on the traveling salesman problem

| 1954 | Dantzig, Fulkerson, Johnson | 49 cities |
| :--- | :---: | :---: |
| 1971 | Held, Karp | 64 cities |
| 1975 | Camerini, Fratta, Maffioli | 67 cities |
| 1977 | Grötschel | 120 cities |
| 1980 | Crowder, Padberg | 318 cities |
| 1987 | Padberg, Rinaldi | 532 cities |
| 1987 | Grötschel, Holland | 666 cities |
| 1987 | Padberg, Rinaldi | $2^{\prime} 392$ cities |
| 1994 | Applegate, Bixby, Chvátal, Cook | $7^{\prime} 3^{397}$ cities |
| 1998 | Applegate, Bixby, Chvátal, Cook | $13^{\prime} 509$ cities |
| 2001 | Applegate, Bixby, Chyátal, Cook | $15^{\prime} 112$ citites |
| 2004 | Applegate, Bixby, Chvátal, Cook, Helsgaun | $24^{\prime} 978$ cities |

And it is not a question of computer performances. Initial algorithms on today's computers would not deal with 100 cities.

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## Syllabus: tools

L1, sep. 26 Graphs and complexity

L2, oct. 3 Shortest paths and dynamic programming


L3, oct. 10 Network flows


L4, oct. 17 Bipartite graphs and Linear Programming . . .

L5, oct. 24 Mixed Integer Linear Programming
L6, nov. 14 Heuristics $e^{-\frac{z}{k_{B} T}}$
L7, nov. 28 Exercises + Introduction to Implementation (Julia)

## Groups

3 groups.

## Syllabus: Applications

L8 dec 5 Project (self-learning)
L9 dec 12 Facility location and bin-packing + Industrial presentation (Frédéric Gardi - Localsolver) 1h

L10 dec 19 Network design + Industrial presentation (Thibault Corneloup Air France)

L11 jan 9 Routing + Intervention 1h (Mathieu Sanchez)
L12 jan 16 Scheduling (self-learning)
L13 jan 23 Exam

## Evaluation

- Multiple choice tests (mandatory) on Educnet
- Project
- Final Exam

$$
\text { Note }=\frac{2}{5} \cdot \text { Project }+\frac{3}{5} \cdot \text { Exam }
$$

No re-sit exam if more than 1 unjustified absence or more than 1 undone multiple choice test

## Resources

Frédéric Meunier's monograph
New monograph (progressively updated on Educnet):
Please indicate me the typos: axel.parmentier@enpc.fr up to two bonus points for those who identify

Three previous exams

## Hackathon

The problem of the miniproject will be the same as the one of the Hackathon
Agenda :
L7, nov. 28 Introduction to Implementation (Julia)
nov. 29 Hackathon (voluntary) + subject available on Educnet
L8, dec. 5 "Séance en autonomie" on the project
jan. 16 End of the project

## 1. What is Operations Research

2. Syllabus
3. Graphs
3.1 Modeling
3.2 Undirected graphs
3.3 Optimization
4. Complexity

# In the textbook 

Complexity: Chapter 2
Graphs: Chapter 3

## Modeling

Model $=$
A mathematical transcription of reality that enables to apply mathematical theory and tools, and translate their results into prediction and decisions in the real world.

Two kinds of models

- Epistemology: model to understand a complicated phenomenon
- Praxeology: model to decide

Operations Research: models to find a good/the best decision in a huge set of potential ones

## Graphs

Graph : $G=(V, E)$
$V$ : set of sommets
$E$ : set of edges $=$ unordered pairs of vertices

degree $\operatorname{deg}(v)$ of a vertex $v$ : number of incident edges.

Simple graphs, complete graphs, bipartite graphs

Simple graph: at most one edge between two vertices
Complete graph: simple graph where every pair of vertices is an edge
Graphe biparti: vertices partitioned into two subsets such that there is no edge between two vertices of the same subset


Complete graphs, bipartite graphs
$K_{n}$ : complete graph with $n$ vertices
$K_{m, n}$ : bipartite complete graph with $m$ and $n$ vertices

How many edges in $K_{n}$ ? And in $K_{m, n}$ ?

## Paths

Path: sequence of the form

$$
v_{0}, e_{1}, v_{1}, \ldots, e_{k}, v_{k}
$$

$v_{i} \in V, e_{j} \in E$ with $e_{j}=v_{j-1} v_{j}$.
Simple path: crosses at most once an edge.
Elementary path: crosses at most once a vertex
Connected graph: a path between any pair of vertices

Cycles

Cycle : path such that $v_{0}=v_{k}$ and all the other vertices are contained at most once

Eulerian path/cycle: simple path/cycle containing all the edges
Hamiltonian path/cycle: elementary path/cycle containing all the vertices

Hamiltonian cycle


Hamilton (1805-1865) invented the "Icosian Game", which was not a commercial success.


Icosian game

Find a Hamiltonian cycle


Icosian game

Find a Hamiltonian cycle


## Modeling with cycles

Give examples of real-life problems whose solutions are Hamiltonian / Eulerian cycles.

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Give examples of real-life problems whose solutions are Hamiltonian / Eulerian cycles.

- Traveling salesman
- Post office


# Königsberg bridges (1736) - Euler (1707-1783) 



Is it possible to go through all the bridge of
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without crossing
twice the same bridge?

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A graph is Eulerian $\Leftrightarrow$ has at most two vertices of odd degree.

What is the minimum number of bridges crossed?

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What is the minimum number of bridges crossed?
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## Graphs : coloring

Coloring : $c: V \rightarrow \mathbb{N}$ ( $\mathbb{N}=$ colors $)$.

Proper coloring : for any neighbor $u, v$, we have $c(u) \neq c(v)$.

Chromatic number $\chi(G)$ : minimum numbers of colors in a proper coloring

Modeling with colorings

A set $F$ of formations must be given to employees of a firm. Each employee $i$ must follow a subset $F_{i}$ of formations. The firm wants to find the minimum number of formation slots it must schedule so that each employee can attend to its formations. Model this problem as a coloring problem.

## Optimization

$\begin{array}{cc}\text { Min } & f(x) \\ \text { s.c. } & x \in X .\end{array}$
$f$ : criteria / objective.
" s.t. " = "subject to"
$X$ : set of feasible solutions
" $x \in X$ " : constraints of the Optimization program.
Among the feasible solutions, we seek an optimal solution $x^{*}$, i.e., a feasible solution that minimizes the criteria.

Minimization: on the importance of lower bounds

$$
\begin{aligned}
& \text { OPT }=\operatorname{Min} f(x) \\
& \text { s.c. } x \in X \text {. }
\end{aligned}
$$

Always ask if there is a simple and good quality lower bound to OPT.
Enables to evaluate the quality of a solution $\rightarrow$ identify when to stop searching, and possibly to the optimality of the solution

## Coloration



Graphe coloré avec cinq couleurs: $\mathrm{r}, \mathrm{b}, \mathrm{j}, \mathrm{v}, \mathrm{n}$.
Can we color this graph with fewer than five colors?

Coloring and complete subgraph: an inequality
cardinal of a complete subgraph $\leq$ number of colors in a proper coloring.

We denote by $\omega(G)$ the maximum cardinality of a complete subgraph of $G$.

$$
\omega(G) \leq \chi(G) .
$$

## Matchings and covers

Set of edges two by two disjoint: matching

Set of vertices $S$ such that each edge contains a vertex in $S$ : cover


Give example of problems modeled by matching / covers.

## Matching and covering: an inequality

$\tau(G)$ : minimum cardinality of a cover
$\nu(G)$ : maximum cardinality of a matching

Prove that

$$
\nu(G) \leq \tau(G)
$$

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Let $M$ be a matching $C$ a vertex cover. Then

$$
|M| \leq|C| .
$$



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Problem

Problem:
Input
Question / task

## Example of problems

EUlerian path problem
Input. A graph G
Question. Is there an Eulerian path in $G$ ?

Maximum weight matching
Input. A graph $G=(V, E)$, a weight function $w: E \rightarrow \mathbb{Q}_{+}$.
Output. A maximum weight matching in $G$.

Decision problem: answer by yes or no to a question.
Optimization problem: find the optimum of a function (under some constraints)

## Algorithm

Algorithm: a sequence of elementary operations that can be implemented on a computer.

Given a problem $\mathcal{P}$ and an algorithm $\mathcal{A}$ solving it, we can ask how efficient it is.

Complexity theory

## Complexity function

Time complexity $f(n)$ of an algorithm: number of elementary operation that must be realized if the input is of size $n$.
Example:

1. Sorting $n$ integers?
2. Testing if an Eulerian cycles exists?

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3. $O(n \log (n))$
4. $O(m+n)$

Time complexity $f(n)$ of an algorithm: number of elementary operation that must be realized if the input is of size $n$.

## Example:

1. Sorting $n$ integers?
2. Testing if an Eulerian cycles exists?
3. $O(n \log (n))$
4. $O(m+n)$
"Clean" definition of algorithm, size of the input, and time complexity.
requires to formalize what is an algorithm on a computer

- See textbook for more details
- Informal understanding sufficient for this lecture


## Polynomial vs exponential algorithm

Polynomial algorithm: time complexity in $=O\left(n^{a}\right)$ with a fixed.
Otherwise, exponential algorithm.

|  | Size $n$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| Time complexity | 10 | 20 | 50 | 60 |
| $n$ | $0,01 \mu \mathrm{~s}$ | $0,02 \mu \mathrm{~s}$ | $0,05 \mu \mathrm{~s}$ | $0,06 \mu \mathrm{~s}$ |
| $n^{2}$ | $0,1 \mu \mathrm{~s}$ | $0,4 \mu \mathrm{~s}$ | $2,5 \mu \mathrm{~s}$ | $3,6 \mu \mathrm{~s}$ |
| $n^{3}$ | $1 \mu \mathrm{~s}$ | $8 \mu \mathrm{~s}$ | $125 \mu \mathrm{~s}$ | $216 \mu \mathrm{~s}$ |
| $n^{5}$ | $0,1 \mathrm{~ms}$ | $3,2 \mathrm{~ms}$ | $312,5 \mathrm{~ms}$ | $777,6 \mathrm{~ms}$ |
| $2^{n}$ | $\sim 1 \mu \mathrm{~s}$ | $\sim 1 \mathrm{~ms}$ | $\sim 13$ jours | $\sim 36.5$ years |

Table: Comparison of different time complexity functions on a computer executing 1 billion operations per second.

## A question of computer speed?

Let $\mathcal{A}$ be an algorithm solving a problem $\mathcal{P}$ in $2^{n}$ operations. We have a computer that solved $\mathcal{P}$ with $\mathcal{A}$ in 1 hour for instances of size up to $n=438$.

With a computer 1000 times faster, instances of up to which size wan we solve in 1 hour?

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Let $\mathcal{A}$ be an algorithm solving a problem $\mathcal{P}$ in $2^{n}$ operations. We have a computer that solved $\mathcal{P}$ with $\mathcal{A}$ in 1 hour for instances of size up to $n=438$.

With a computer 1000 times faster, instances of up to which size wan we solve in 1 hour?
448

Size of the largest instance that we can solved in 1 hour

| Complexity <br> function | Present day <br> computer | Computer <br> $100 \times$ faster | Computer <br> $1000 \times$ faster |
| :---: | :---: | :---: | :---: |
| $n$ | $N_{1}$ | $100 N_{1}$ | $1000 N_{1}$ |
| $n^{2}$ | $N_{2}$ | $10 N_{2}$ | $31.6 N_{2}$ |
| $n^{3}$ | $N_{3}$ | $4.64 N_{3}$ | $10 N_{3}$ |
| $n^{5}$ | $N_{4}$ | $2.5 N_{4}$ | $3.98 N_{4}$ |
| $2^{n}$ | $N_{5}$ | $N_{5}+6.64$ | $N_{5}+9.97$ |
| $3^{n}$ | $N_{6}$ | $N_{6}+4.19$ | $N_{6}+6.29$ |

Table: Comparison of different complexity functions

## Complexity classes

A decision problem is polynomial or in $\mathscr{P}$ if there exists a polynomial algorithm that solves it.

A decision problem is non-deterministically polynomial or in $\mathscr{N} \mathscr{P}$ if, if the answer is yes, there exists a certificate and a polynomial algorithm that enables to check that the solution is yes.

## Example of problem in $\mathscr{N P}$

Hamiltonian cycle
Input. A graph $G=(V, E)$
Question. Does $G$ have a Hamiltonian cycle

Show that the Hamiltonian cycle problem is in $\mathscr{N} \mathscr{P}$ (Give a certificate)

For $F \subseteq E$, we can test in polynomial if $F$ is a Hamiltonian cycle. $F$ is a certificate.
The Hamiltonian cycle problem is in $\mathscr{N} \mathscr{P}$.

Class $\mathfrak{N P}$

## Proposition

$$
\mathscr{P} \subseteq \mathscr{N} \mathscr{P} .
$$

## Proof.

If the answer is yes, the input is a certificate.
$\mathcal{N} \mathscr{P}$-complete problem

A problem $\mathcal{P}$ is $\mathscr{N} \mathscr{P}$-complete if
$>\mathcal{P}$ is in $\mathscr{N} \mathscr{P}$
$>\mathcal{P}$ is at least as difficult as any problem in $\mathscr{N} \mathscr{P}$.

If there exists a polynomial algorithm solving an $\mathscr{N} \mathscr{P}$-complete problem, then there is a polynomial algorithm solving any problem in $\mathscr{N} \mathscr{P}$.

Theorem (Cook, 1970)
There exists $\mathscr{N} \mathscr{P}$-complete problems.

What does "at least as difficult" mean

## Example of $\mathscr{N} \mathscr{P}$-complete problem

Hamiltonian cycle
Input. A graph $G=(V, E)$
Question. Does $G$ have a Hamiltonian cycle

A polynomial reduction of a decision problem $\mathcal{P}^{\prime}$ to a problem $\mathcal{P}$ is a function $f$ that transforms an instances $x^{\prime}$ of $\mathcal{P}^{\prime}$ into an instance $x$ of $\mathcal{P}$ such that
$\Rightarrow \operatorname{size}\left(x^{\prime}\right)=O\left(\operatorname{size}\left(P\left(f\left(x^{\prime}\right)\right)\right)\right)$ where $P$ is a polynomial
$\Rightarrow$ the answer of $\mathcal{P}^{\prime}$ for $x^{\prime}$ is yes if and only if the answer of $\mathcal{P}$ for $f\left(x^{\prime}\right)$ is yes.

How to prove that a problem $\mathcal{P}$ is $\mathscr{N} \mathscr{P}$-complete?
$\Rightarrow$ Prove that $\mathcal{P}$ is in $\mathscr{N} \mathscr{P}$
$\Rightarrow$ Prove that an $\mathscr{N} \mathscr{P}$-complete problem $\mathcal{P}^{\prime}$ reduces to (a polynomial number of instances of) $\mathcal{P}$
$\mathscr{N} \mathscr{P}$-hard problem

A problem is $\mathscr{N} \mathscr{P}$-hard if
$>\mathcal{P}$ is at least as difficult as any problem in $\mathscr{N} \mathscr{P}$.

Only decision problems can be $\mathscr{N} \mathscr{P}$-complete. Optimization and decisions problems can be $\mathscr{N} \mathscr{P}$-hard.

How to prove that a problem $\mathcal{P}$ is $\mathscr{N} \mathscr{P}$-hard?
Prove that an $\mathscr{N} \mathscr{P}$-complete problem $\mathcal{P}^{\prime}$ reduces to $\mathcal{P}$

## Complexity classes

 Paris Tech

## 1 million \$ question

$$
\mathscr{P} \stackrel{?}{=} \mathscr{N} \mathscr{P}
$$

funded by the Clay institute

## Exercise

1. Prove that the problem of existence of a Hamiltonian path is $\mathscr{N} \mathscr{P}$ -complete, knowing that the problem of existence of a Hamiltonian cycle is $\mathscr{N} \mathscr{P}$-complete
2. Prove that the problem of existence of a Hamiltonian cycle is $\mathscr{N} \mathscr{P}$ -complete, knowing that the problem of existence of a Hamiltonian path is $\mathscr{N} \mathscr{P}$-complete
