

Operations research: Mathematics and algorithmics for solving decision-making problems

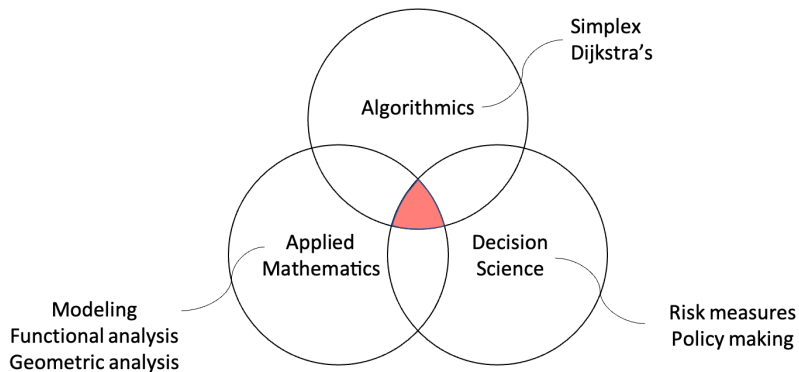
Ayşe N. Arslan

Chargée de recherche
Centre Inria de l'Université de Bordeaux
Equipe : EDGE

Soirée EDM I
04/12/2023

Operations research (OR)

- Operations research and optimization are at the intersection of multiple disciplines.



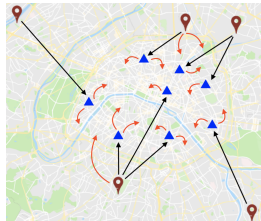
Operations research (OR)

- ▶ OR solves optimization problems that decision-makers (managers, politicians, engineers, etc.) encounter.
- ▶ Problems are typically expressed in terms of decisions, costs and constraints.
- ▶ Tools coming from mathematics, informatics, economics and industrial engineering are often used in their solution.
- ▶ The end result is a decision-making tool.

$$\min_{x \in X} f(x)$$

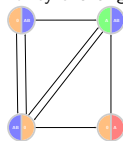
Some examples

- ▶ Route optimization
- ▶ Planning and scheduling
- ▶ Network design (telecommunications, distribution, electricity, etc.)
- ▶ Supply chain management



Some projects from our team¹

Kidney exchange



Maintenance planning



Retail network design



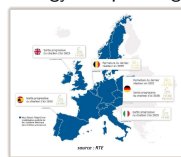
Phytosanitary treatments



Maritime transportation



Energy mix planning



¹Team EDGE-Centre Inria de l'Universite de Bordeaux

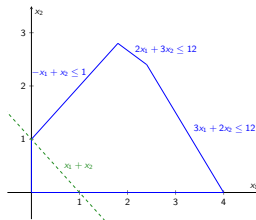
Classical tools of operations research²

- ▶ **Mathematical programming**
 - ▶ Linear programming (LP/PL)
 - ▶ Mixed-integer linear programming (MILP/PLNE)
 - ▶ ..
- ▶ **Graph theory and algorithms**
- ▶ Constraint programming (CP/PPC)
- ▶ Convex analysis
- ▶ Approximation algorithms
- ▶ Heuristics, metaheuristics
- ▶ Queueing theory, simulation, statistics
- ▶ ...

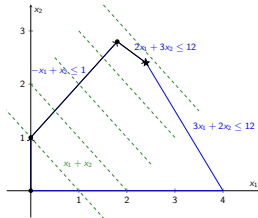
²Most topics are covered in the master MAS-ROAD

Mathematical programming: Linear programming

$$\begin{array}{ll} \min_{x \in \mathbb{R}_+^n} & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} \leq \mathbf{b} \end{array}$$



- ▶ Well-known algorithms: Simplex, ellipsoid, interior point, etc.
- ▶ Solvers: CPLEX, Gurobi, Clp, HiGHS, Excel-Solver etc.



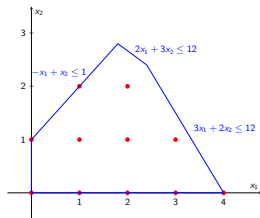
$$z = \mathbf{c}_B^\top \mathbf{A}_B^{-1} \mathbf{b} + \left(\mathbf{c}_N - \mathbf{c}_B^\top \mathbf{A}_B^{-1} \mathbf{A}_N \right)^\top \mathbf{x}_N$$

$$\mathbf{x}_B = \mathbf{A}_B^{-1} \mathbf{b} - \mathbf{A}_B^{-1} \mathbf{A}_N \mathbf{x}_N$$

Mathematical programming: Mixed-integer programming

$$\begin{array}{ll} \min & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} \leq \mathbf{b} \end{array}$$

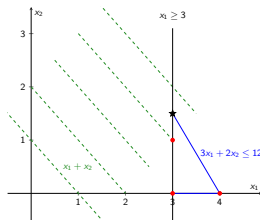
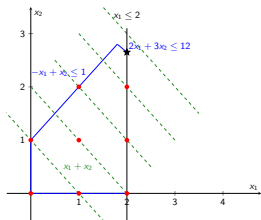
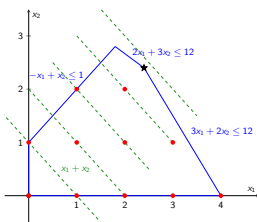
$\mathbf{x} \in \mathbb{Z}_+^{n_1} \times \mathbb{R}_+^{n_2}$



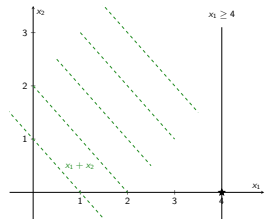
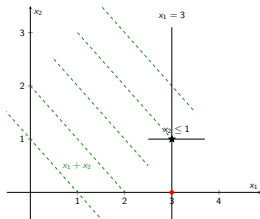
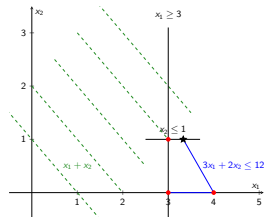
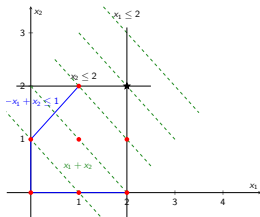
- ▶ Integer/binary decisions that cannot be rounded from fractions:
 - ▶ Do we open facility i or not?
 - ▶ How many trucks do we need to send to client j from facility i ?
 - ▶ If facility i is not open then it cannot be used to satisfy demand.
- ▶ Well-known algorithms: Branch & Bound, Branch & Cut etc.
- ▶ Solvers: CPLEX, Gurobi, HiGHS, GLPK etc.

Mathematical programming: Mixed-integer programming

- Branch & Bound: Solve relaxations and successively partition the feasible region



Mathematical programming: Mixed-integer programming



Difficulty of mixed-integer programming problems³

- ▶ Mixed-integer programming problems are NP-Complete in the general case.
- ▶ In practice, considerable progress has been made since the beginning.

A classical problem: Traveling salesman

- ▶ Find the shortest cycle passing through N cities given the pairwise distances.
- ▶ In theory finding the best cycle requires testing $N!$ possibilities.
 - ▶ For 10 cities ($N=10$) : < 1 milliseconds
 - ▶ For 30 cities ($N=30$) : 35000 billion years



³Source : <http://www.math.uwaterloo.ca/tsp/>

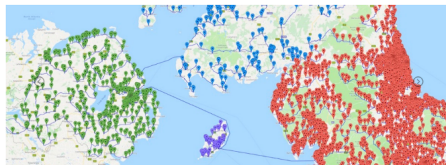
Difficulty of mixed-integer programming problems³

- ▶ Mixed-integer programming problems are NP-Complete in the general case.
- ▶ In practice, considerable progress has been made since the beginning.

1954	1977	1987	1994	1998	2001	2004
N = 49	N = 120	N=2392	N=7397	N=13509	N=15112	N=24978

UK49687

Shortest possible tour to nearly every pub in the United Kingdom.



Optimal 49,687-stop pub crawl. [Click.](#)

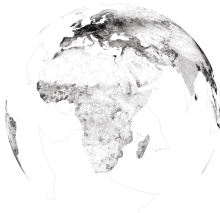
• • • • •

³Source : <http://www.math.uwaterloo.ca/tsp/>

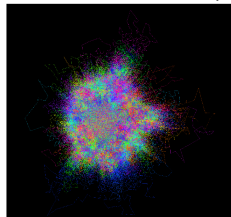
Difficulty of mixed-integer programming problems³

- ▶ Mixed-integer programming problems are NP-Complete in the general case.
- ▶ In practice, considerable progress has been made since the beginning.

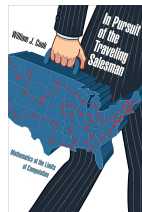
1.9 millions cities to 0.0473% optimality



1 331 906 450 stars to 0.37% optimality

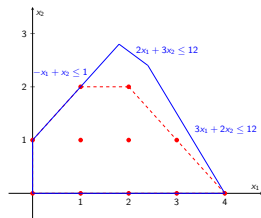
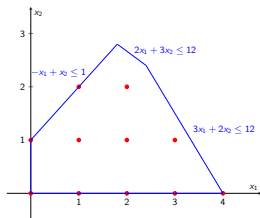


- ▶ Powerful heuristics coupled with branch & cut
- ▶ Many results dedicated to the problem
- ▶ Months of computation in parallel processing



³Source : <http://www.math.uwaterloo.ca/tsp/>

Advanced techniques in mixed-integer programming

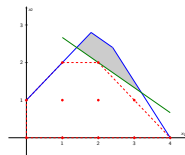
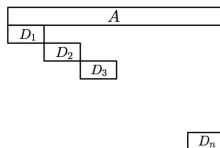
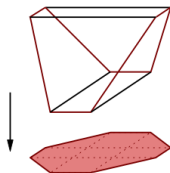
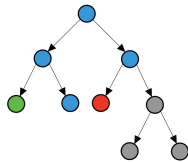


Implicit enumeration

Extended formulations

Decomposition algorithms

Geometric analysis



Conclusions

- ▶ OR is an interdisciplinary field that studies decision-making problems from a mathematical and algorithmic perspective.
- ▶ It develops tools to help decision makers.
- ▶ Significant algorithmic progress has been made in recent years.
- ▶ Further research is needed in order to extend classical results to more realistic contexts.

Questions?



*Thank you for your attention!
Any questions?*

*ayse-nur.arslan@inria.fr
<https://www.inria.fr/fr/edge>*

Appendix:
Optimisation under uncertainty

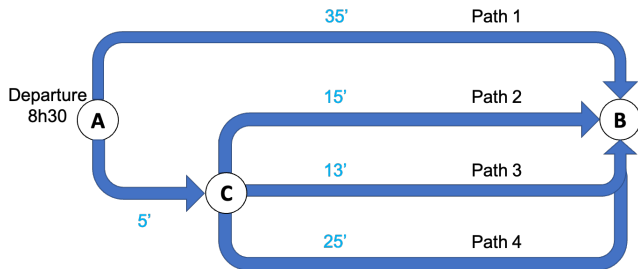
Presence of uncertainty

- ▶ New challenges:
 - ▶ Renewable energy production.
 - ▶ Resilient network design.
 - ▶ Healthcare/disaster management.
 - ▶ Circular economy.
 - ▶ Security and defense.



- ▶ Uncertainty:
 - ▶ Stochastic nature of systems.
 - ▶ Long duration of decision processes.
 - ▶ Difficulty of precise measurements.
 - ▶ Lack of historical information.
 - ▶ Presence of adversarial participants.

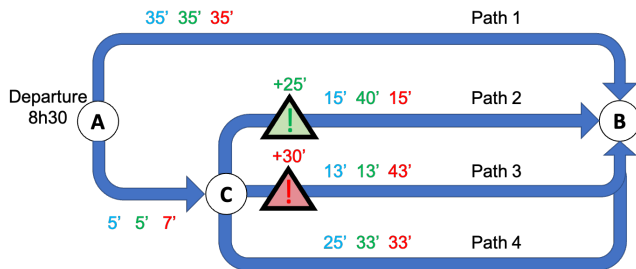
A practical example: Shortest path under uncertainty



Path 1	35'
Path 2	20'
Path 3	18'
Path 4	30'

- What is the shortest path from point A to point B?

A practical example: Shortest path under uncertainty

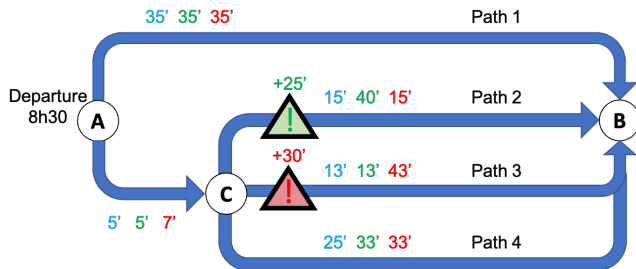


Scenario	Probability
1	0,5
2	0,25
3	0,25

- What is the shortest path from point A to point B?

Our first order of business is to characterize the uncertain data.

A practical example: Shortest path under uncertainty



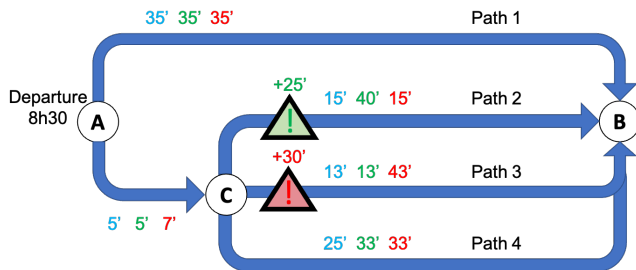
Scenario	Probability
1	0,5
2	0,25
3	0,25

Scenario	1	2	3
Path 1	35'	35'	35'
Path 2	20'	45'	22'
Path 3	18'	18'	50'
Path 4	30'	38'	40'

- What is the shortest path from point A to point B?

Second order of business is to characterize what constitutes a "good" solution.

A practical example: Shortest path under uncertainty



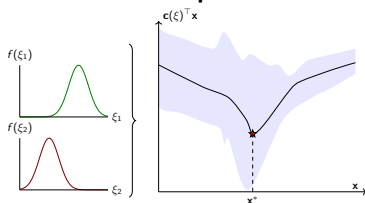
Scenario	Probability
1	0,5
2	0,25
3	0,25

Scenario	1	2	3	Esp	Max
Path 1	35'	35'	35'	35'	35'
Path 2	20'	45'	22'	26,75'	45'
Path 3	18'	18'	50'	26'	50'
Path 4	30'	38'	40'	34,5'	38'

In optimization under uncertainty the notion of a good solution depends on the risk preferences of the decision-maker.

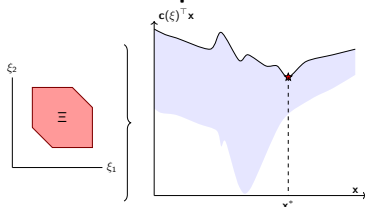
Optimization under uncertainty paradigms

Stochastic optimization



- Distribution \mathbb{P} is known.
- Consequences are observed repeatedly.
- Risk level is low or moderate.
- Example: Distribution network design.

Robust optimization



- Distribution is not known (or no distribution).
- Consequences are observed once.
- Risk level is high.
- Example: Disaster management.

Optimization under uncertainty paradigms

Stochastic optimization

$$\begin{array}{ll} \min_{\mathbf{x} \in X} & \mathbb{E}_{\xi \in \Xi}^{\mathbb{P}} \left[\mathbf{c}(\xi)^{\top} \mathbf{x} \right] \\ \text{s.t.} & \mathbf{Ax} \leq \mathbf{b} \end{array}$$

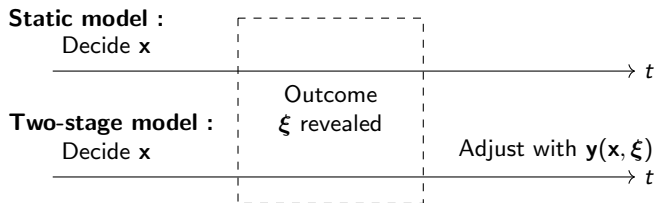
- ▶ Distribution \mathbb{P} is known.
- ▶ Consequences are observed repeatedly.
- ▶ Risk level is low or moderate.
- ▶ Example: Distribution network design.

Robust optimization

$$\begin{array}{ll} \min_{\mathbf{x} \in X} & \max_{\xi \in \Xi} \left[\mathbf{c}(\xi)^{\top} \mathbf{x} \right] \\ \text{s.t.} & \mathbf{Ax} \leq \mathbf{b} \end{array}$$

- ▶ Distribution is not known (or no distribution).
- ▶ Consequences are observed once.
- ▶ Risk level is high.
- ▶ Example: Disaster management.

Sequential decision-making under uncertainty



In optimization under uncertainty the timing of decisions is important.

The difficulty of solution can increase with the number of decision stages.