# Introduction to computational complexity 

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Computational Aspects of Optimization

## Software

General mathematical and statistical software:

- Matlab and (free) OPTI toolbox: http://www.i2c2.aut.ac.nz/Wiki/OPTI/index.php/Main/HomePage
- SAS: OR package
- R: suitable packages
- Mathematica: findminimum, minimize, ...


## Software

Modelling tools for optimization:

- GAMS: http://www.gams.com
- AIMMS: http://www.aimms.com
- Gurobi: http://www.gurobi.com
- AMPL: http://ampl.com
- CPlex Studio:
http://www-03.ibm.com/software/products/cs/ibmilogcpleoptistud
- MPL: http://www.maximalsoftware.com/mpl


## Software

## Open Source libraries <br> - COIN-OR: http://www.coin-or.org/

## Software

Solvers

- Baron (LP, NLP, MIP, MINLP, ...)
- Bonmin (NLP, MIP, MINLP, ...)
- Conopt (LP, NLP, ...)
- CPlex (LP, MIP, MIQCP, ...)
- Dicopt (MIQCP, MINLP, ...)
- Gurobi
- Knitro (LP, MINLP, MIQCP, NLP, ...)
- Lindo (LP, MINLP, MIP, MIQCP, NLP, ...)
- Minos (LP, NLP, ...)
- Mosek (LP, MIP, MIQCP, NLP, ...)
- Xpress (LP, MIP, MIQCP, ...)


## Introduction to complexity theory

Wolsey (1998): Consider decision problems having YES-NO answers. Optimization problem

$$
\max _{x \in M} c^{T} x
$$

can be replaced by (for some $k$ integral)
Is there an $x \in M$ with value $c^{T} x \geq k$ ?
For a problem instance $X$, the length of the input $L(X)$ is the length of the binary representation of a standard representation of the instance. Instance $X=\{c, M\}, X=\{c, M, k\}$

## Example: Knapsack decision problem

For an instance

$$
X=\left\{\sum_{i=1}^{n} c_{i} x_{i} \geq k, \sum_{i=1}^{n} a_{i} x_{i} \leq b, x \in\{0,1\}^{n}\right\}
$$

the length of the input is

$$
L(X)=\sum_{i=1}^{n}\left\lceil\log c_{i}\right\rceil+\sum_{i=1}^{n}\left\lceil\log a_{i}\right\rceil+\lceil\log b\rceil+\lceil\log k\rceil
$$

## Running time

## Definition

- $f_{A}(X)$ is the number of elementary calculations required to run the algorithm $A$ on the instance $X \in P$.
- Running time of the algorithm $A$

$$
f_{A}^{*}(I)=\sup _{X}\left\{f_{A}(X): L(X)=I\right\}
$$

- An algorithm $A$ is polynomial for a problem $P$ if $f_{A}^{*}(I)=O\left(I^{p}\right)$ for some $p \in \mathbb{N}$.


## Classes $\mathcal{N P}$ and $\mathcal{P}$

## Definition

- $\mathcal{N P}$ (Nondeterministic Polynomial) is the class of decision problems with the property that: for any instance for which the answer is YES, there is a polynomial proof of the YES.
- $\mathcal{P}$ is the class of decision problems in $\mathcal{N P}$ for which there exists a polynomial algorithm.
$\mathcal{N} \mathcal{P}$ may be equivalently defined as the set of decision problems that can be solved in polynomial time on a non-deterministic Turing machine ${ }^{1}$.

[^0]
## Alan Turing



The Imitation Game (2014)

## Polynomial reduction and the class $\mathcal{N P C}$

## Definition

- If problems $P, Q \in \mathcal{N} \mathcal{P}$, and if an instance of $P$ can be converted in polynomial time to an instance of $Q$, then $P$ is polynomially reducible to $Q$.
- $\mathcal{N P C}$, the class of $\mathcal{N} \mathcal{P}$-complete problems, is the subset of problems $P \in \mathcal{N P}$ such that for all $Q \in \mathcal{N} \mathcal{P}, Q$ is polynomially reducible to $P$.

Proposition: Suppose that problems $P, Q \in \mathcal{N} \mathcal{P}$.

- If $Q \in \mathcal{P}$ and $P$ is polynomially reducible to $Q$, then $P \in \mathcal{P}$.
- If $P \in \mathcal{N} \mathcal{P C}$ and $P$ is polynomially reducible to $Q$, then $Q \in \mathcal{N} \mathcal{P C}$.

Proposition: If $\mathcal{P} \cap \mathcal{N} \mathcal{P C} \neq \emptyset$, then $\mathcal{P}=\mathcal{N} \mathcal{P C}$.

## Open question \& Euler diagram

Is $\mathcal{P}=\mathcal{N} \mathcal{P}$ ?


# $\mathcal{N} \mathcal{P}$-hard optimization problems 

## Definition

An optimization problem for which the decision problem lies in $\mathcal{N P \mathcal { P }}$ is called $\mathcal{N} \mathcal{P}$-hard.

## Literature

- G.L. Nemhauser, L.A. Wolsey (1989). Integer Programming. Chapter VI in Handbooks in OR \& MS, Vol. 1, G.L. Nemhauser et al. Eds.
- L.A. Wolsey (1998). Integer Programming. Wiley, New York.
- L.A. Wolsey, G.L. Nemhauser (1999). Integer and Combinatorial Optimization. Wiley, New York.


[^0]:    ${ }^{1}$ NTM writes symbols one at a time on an endless tape by strictly following a set of rules. It determines what action it should perform next according to its internal state and what symbol it currently sees. It may have a set of rules that prescribes more than one action for a given situation. The machine "branches" into many copies, each of which follows one of the possible transitions leading to a "computation tree""

