

• mailing list!

JAA2

(I.)

What will we do?

- What does it mean to prove something hard/intractable?
- How to
- Not a pure complexity course
- Approximation algorithms
- Online algorithms

→ "master techniques"

- key problems
- proof styles
- gadgets

Literature:

- Garey Johnson: Computers and Intractability; A Guide to the Theory of NP-completeness
- V. Vazirani: Approximation Algorithms
- Borodin / El-Yaniv: Online Computation and Competitive Analysis
- Course by Erik Demaine

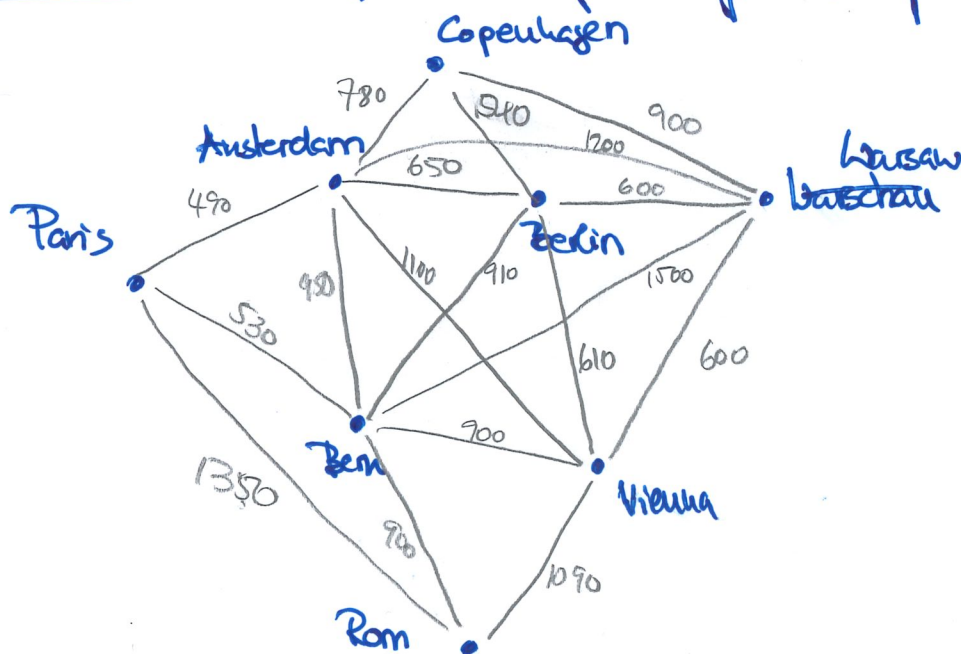
1. Problems, Instances etc, P, NP, EXP, hardness - Notation

* A problem Π gives a general description of parameters and statements of properties of a solution.

↳ example: TSP

- cities \leftrightarrow vertices (v_1, \dots, v_n)
- roads \leftrightarrow edges (e_1, \dots, e_m)
- distances \leftrightarrow weights $(c_{e_1}, \dots, c_{e_m})$
- We ask for a shortest tour that is a round trip visiting each vertex exactly once

* An instance $I \in \Pi$ define values for all problem parameters



* The input size of instance $I \in \Pi$: # of symbols of the underlying alphabet necessary to encode I

What are we looking for?

↳ Algorithm to solve our problem (instances)!

↳ any algorithm?

* An algorithm solves a problem Π in polynomial time if we can bound its running time for any $I \in \Pi$ by a polynomial of I 's input size. / n^c

Can we always find this? Probably not! (unless $P=NP$... more later)

Some more notation:

* A decision problem that allows for a polynomial time alg. belongs to P.

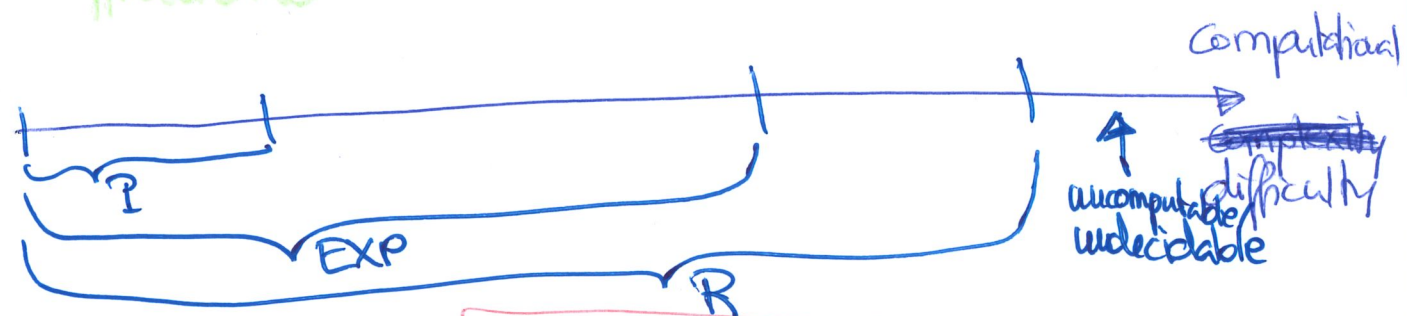
↳ answer is YES or NO

but: the TSP is an optimization problem? We still use the same notation: give a numerical parameter B as a parameter
↳ Is there a tour of length less or equal to B ?

$P = \{ \text{problems solvable in polynomial time} \}$

$EXP = \{ \text{problems solvable in exponential time} \}$

$R = \{ \text{problems solvable in finite time} \}$ ↳ 2^{n^c} [Turing 1936; Church 1941]
↳ "recursive"



$P \neq EXP \neq R$

Examples:

- negative-weight cycle detection $\in P$
- $n \times n$ Chess $\in EXP$ but $\notin P$
 - \hookrightarrow = who wins from given board configuration
- Tetris $\in EXP$, but not known whether $\in P$
 - \hookrightarrow = survive given pieces from given board
- halting problem $\notin R$ [Turing, 1936] \rightarrow Turing machine
 - \hookrightarrow decide: given: computer program + input
 - \hookrightarrow will it halt or run forever?
- "Most" decision problems $\notin R$
 - (# algorithms $\approx N$; # decision problems $\approx 2^N = R$)

* A problem belongs to NP if for every "YES"-instance there is a certificate in polynomial time.

NP = { decision problems solvable in polytime with a "lucky" algorithm }

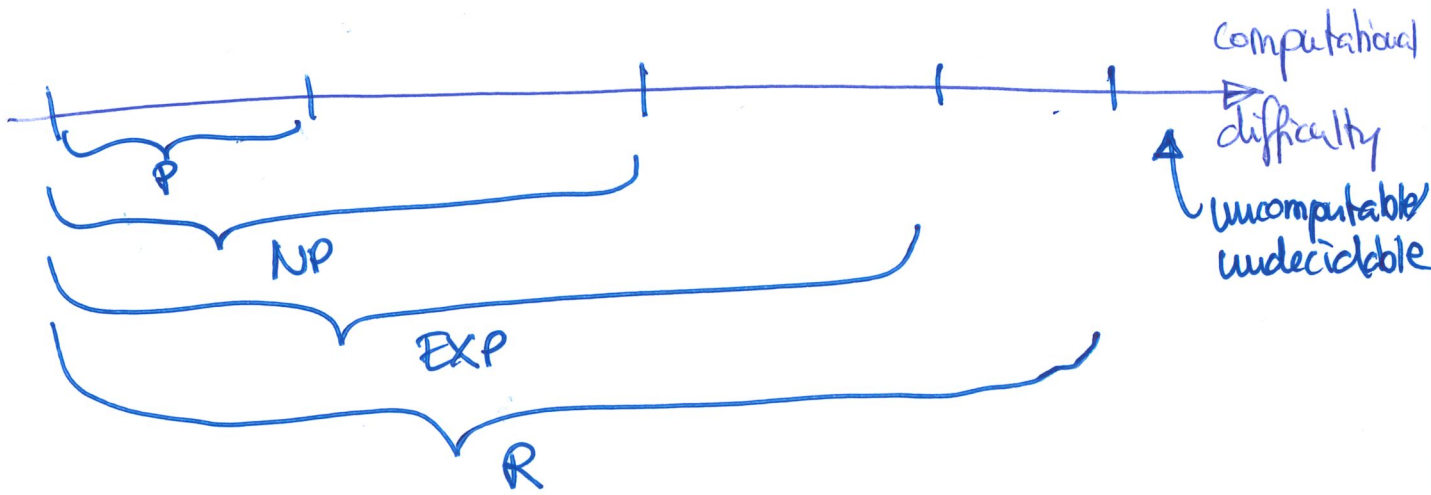


- Nondeterministic model: algorithm makes guesses and then says YES or NO
- guesses guaranteed to lead to YES outcome if possible (NO otherwise)

\hookrightarrow can make lucky guesses, always "right" w/o trying all options

= { decision problems with solutions that can be "checked" in polynomial time }

when answer = YES can "prove" it + polytime alg. can check proof



Example: Tetris $\in NP$

- nondeterministic alg: - guess each move - did I survive?
- proof of YES: list what moves to make (rules of Tetris easy)

P ≠ NP: big conjecture (worth \$ 1.000.000)

≈ can't engineer luck

≈ generating (proofs of) solutions can be harder than checking them

~~* Let's go~~

* A problem Π is NP-hard if $\Pi \in P$ implies $P = NP$.

* In general: X-hard = "as hard as" every problem $\in X$
↳ NP, EXP, etc

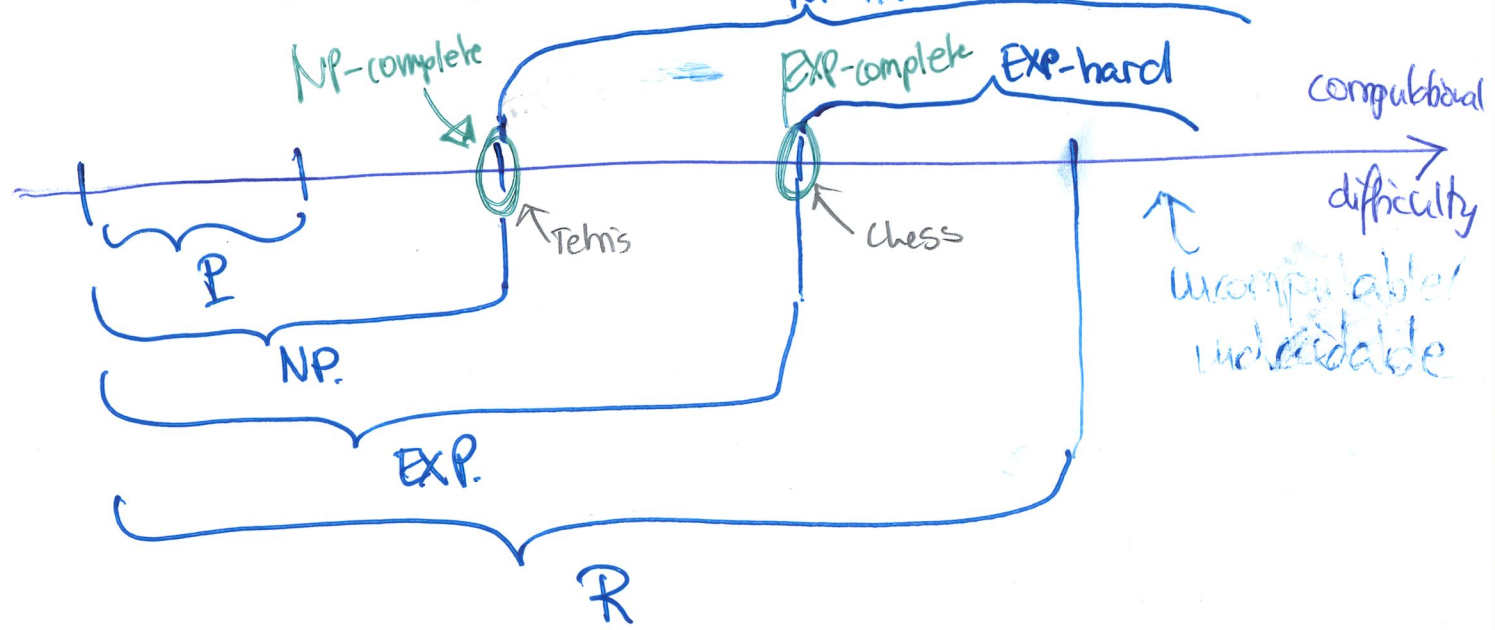
* A problem Π is NP-complete if it is NP-hard and in NP.

* X-complete = X-hard \cap X

What does that mean
↳ a bit later

Example: Tetris is NP-complete [Breukelaar et al., 2004]

⇒ if $P \neq NP$, then Tetris $\in NP \setminus P$
NP-hard



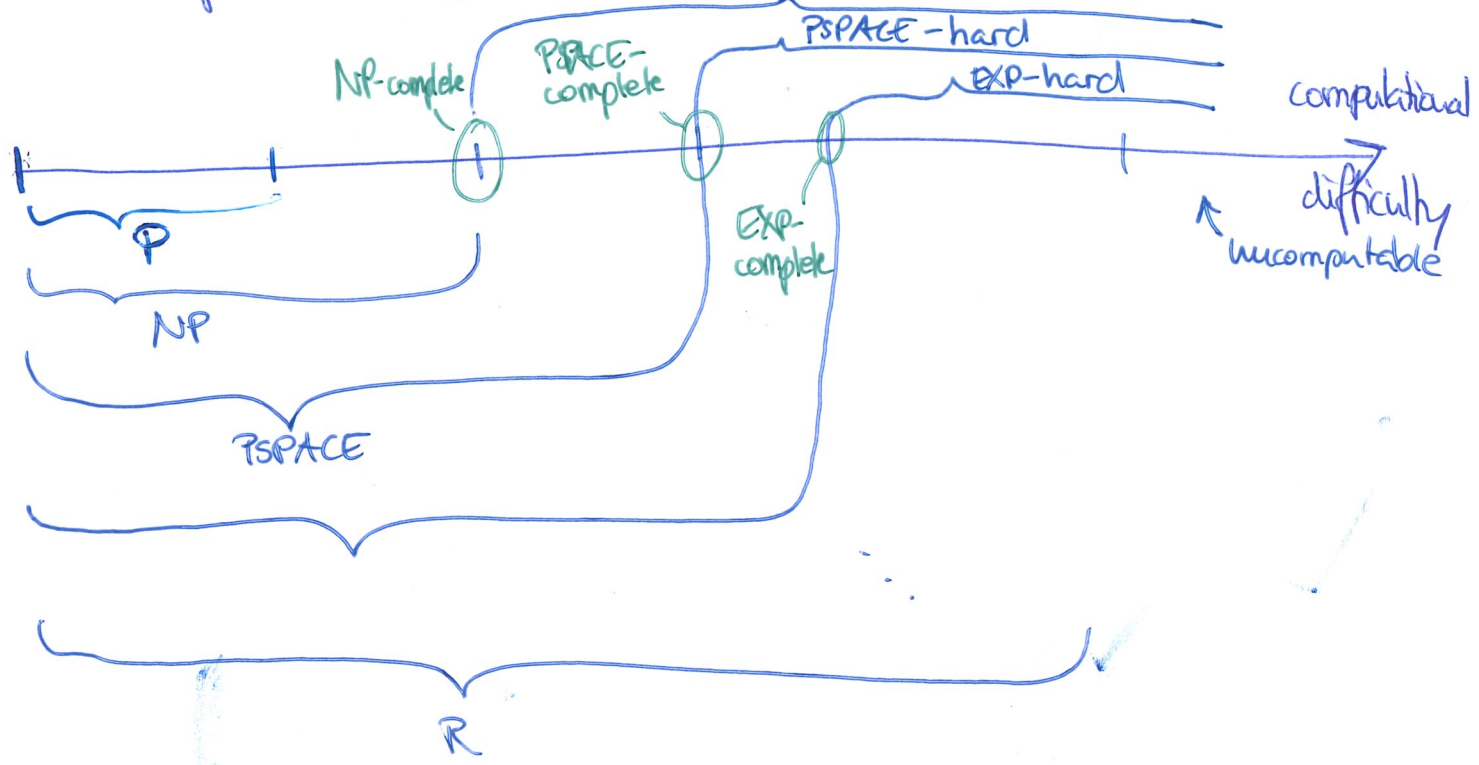
Chess is EXP-complete ⇒ $\notin P$

⇒ Chess $\in EXP \setminus NP$ if $NP \neq EXP$ (also open)



* PSPACE = {problems solvable in polynomial space}

- \subseteq EXP: only exp. many states
- \supseteq NP: simulate all executions, take min OR
- open whether either is strict NP-hard



Example: RushHour is PSPACE-complete [Flake & Baum, 2002]

* There exists stuff beyond exponential - not so important, and not for this course

2. How to prove a problem to be NP-hard? What does "as hard as" mean?

What does it mean:

~~* No known NP-hard~~

Reduction from problem A to problem B = poly-time algorithm to convert: instance of A \rightarrow instance of B

such that solution to A = solution to B

\Rightarrow if can solve B then can solve A

\Rightarrow B is at least as hard as A

(A is a special case of B)

- * E.g.: no NP-complete problem can be solved by any known polytime alg.
- * If there is a polytime alg. for any NP-complete problem, then there are polytime alg.s for all NP-complete problems

↳ Interest in approximation algorithms!

"if I cannot give an optimal solution in polytime, I like to give guarantees on how "bad" the solutions an algorithm gives can be - compared to the (unknown!) optimum" → interest in bounds

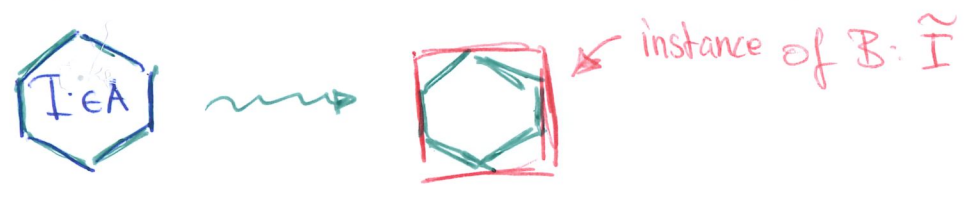
Proof of NP-hardness

Our problem: B

Some known NP-complete problem: A

Reduction? Which direction?

Idea: "dress up" every instance of A as an instance of B



If we had a polytime algorithm for B , we could solve I

Ensure with reduction! → This would give us the correct answer for I
 → We would have a polytime algorithm for A

reduction: every instance of $A \in NP$: input x

↓
instance of B

↓
input $T(x)$ such that answer (YES/NO) for $T(x)$ as an answer for B is the correct answer for x as an input for A

- This is a "one-call" reduction [Karp]
- "Multi-call" reduction [Turing] also possible:
solve A using an oracle that solves B (doesn't help much for problems we consider)

Almost all hardness proofs are by reduction from known hard problem to your problem.

3. Problems useful for reductions

Summary of recipe to show NP-completeness

- (1) Show B is in NP
- (2) Select a known NP-complete problem ~~B~~ A
- (3) Construct a reduction from A to B
- (4) Prove that reduction is polynomial

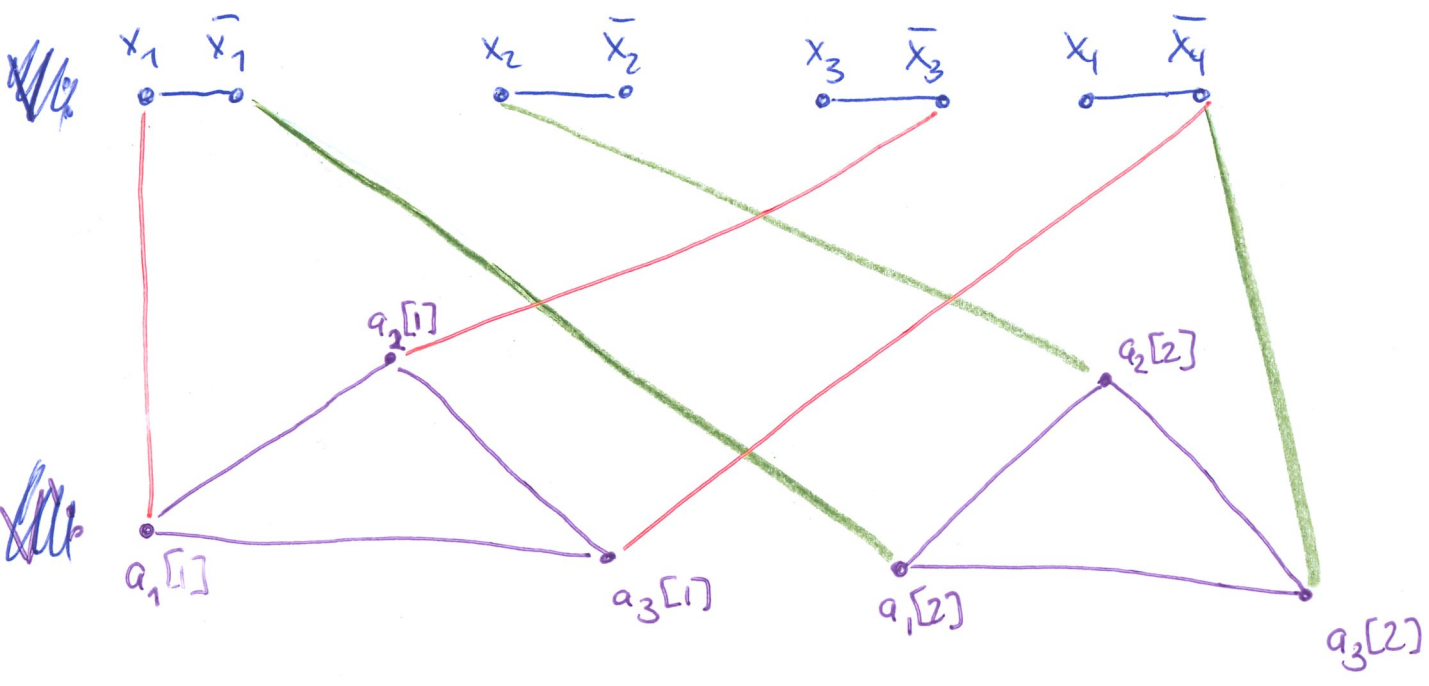
ad proof of Th. 3.1

$$X = \{x_1, x_2, x_3, x_4\}$$

~~$C = \{x_1, x_2, x_3\}$~~

also used \bar{x}_1

$$\begin{aligned}
C &= (x_1 \vee \neg x_3 \vee \neg x_4) \wedge (\neg x_1 \vee x_2 \vee \neg x_4) \\
&= (x_1 \vee \bar{x}_3 \vee \bar{x}_4) \wedge (\bar{x}_1 \vee x_2 \vee \bar{x}_4) \\
&= \{ \{x_1, \bar{x}_3, \bar{x}_4\}, \{\bar{x}_1, x_2, \bar{x}_4\} \}
\end{aligned}$$



ad number problems:

Assuming $P \neq NP$:

