

# Computational Complexity: Classes

AA-GEI FIB, UPC

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# Problem types

- **Decision**

*Input  $x$  in  $I$*

*Is  $P(x)$  true?*

**Example:** Given a graph and two vertices, is there a path joining them?

- **Function**

*Input  $x$  in  $I$*

*Compute  $y$  such that  $Q(x, y)$*

**Example:** Given a graph and two vertices, compute the minimum distance between them.

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- We need to guarantee that  $\{\langle x \rangle \mid x \in I\}$  is decidable.

# Decision problem classes

- **Undecidable**  
No algorithm can solve the problem.
- **Decidable**  
There is an algorithm solving them.

# Classifying decidable problems

- Decidable problems, admit algorithmic solutions.
- The algorithms solving such problems can use more or less amount of resources.
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- As usual we take a worst case analysis on inputs with the same size.
- For a TM, time correspond to the **length of the computation** and space to the portion of the tape **accessed** during the computation.
- Observe that sometimes the space used might be smaller than the input.

# Complexity classes: definitions

Let  $f : \mathbb{N} \rightarrow \mathbb{N}$ ,

- The class  $\text{TIME}(f(n))$  is the class of decision problems for which an algorithm exists that solves instances of size  $n$  in time  $O(f(n))$ .
- The class  $\text{SPACE}(f(n))$  is the class of decision problems for which an algorithm exists that solves instances of size  $n$  using space  $O(f(n))$ .

# Some complexity classes

- $P = \cup_k TIME(n^k) = TIME(poly(n))$
- $EXP = TIME(2^{poly(n)})$
- $2EXP = TIME(2^{2^{poly(n)}})$
- $L = LOGSPACE = SPACE(\lg n)$
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$$L \subseteq P \subseteq EXP \quad PSPACE \subseteq EXP$$

# Non deterministic machines

- We can use a generalized, although unrealistic, model of computation.
- A program has a built-in operation allowing to branch to different positions in the code.
- When the code is executed, the machine follows the the “best” continuation.

# Complexity classes: definitions

Let  $f : \mathbb{N} \rightarrow \mathbb{N}$ ,

- The class  $\text{NTIME}(f(n))$  is the class of decision problems that a non deterministic TM recognizes in time  $O(f(n))$ .
- The class  $\text{NSPACE}(f(n))$  is the class of decision problems for which a non deterministic TM solves instances of size  $n$  using space  $O(f(n))$ .

# Some complexity classes

- $NP = NTIME(poly(n))$
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$$L \subseteq NL \subseteq P$$

$$P \subseteq NP \subseteq PSPACE = NPSPACE \subseteq EXP$$

# Logic and NP

- A **verifier** for a language  $L$  is an algorithm  $\mathcal{A}$ , where  $L = \{x \mid \mathcal{A} \text{ accepts } \langle x, c \rangle \text{ for some word } c\}$ .

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- A **verifier** for a language  $L$  is an algorithm  $\mathcal{A}$ , where  $L = \{x \mid \mathcal{A} \text{ accepts } \langle x, c \rangle \text{ for some word } c\}$ .
- A **polynomial time verifier** runs in polynomial time in the length of  $x$ .
- A language  $L$  is **polynomially verifiable** if it has a polynomial time verifier.
- A verifier uses additional information, represented by the word  $c$ , to verify that a string  $x$  is a member of  $L$ . This information is called a **certificate**, or **proof**, of membership in  $L$ .
- A polynomial verifier can access only polynomial space. Therefore, we can assume that  $|c|$  is polynomial in  $|x|$ .

# Decision Problems in NP: Examples

## Theorem

$L \in NP$  iff  $L = \{x \mid \exists y R(x, y)\}$  where  $|y| = poly(|x|)$  and  $R$  can be decided in polynomial time.

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NP is the class of polynomially verifiable languages.

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- **SUBSET-SUM:** Given a set  $S = \{x_1, \dots, x_n\}$  of  $\mathbb{Z}^+$  and  $t \in \mathbb{Z}$ , exists  $S' \subseteq S$  with  $\sum_{x_j \in S'} x_j = t$ ?

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- **HAMILTONIAN CYCLE:** Given a graph  $G$ , is  $G$  Hamiltonian?
  - The certificate is a permutation of  $V(G)$ , it can be represented by a vector, so it has polynomial size. The verifier also runs in polynomial time.
  - So,  $\text{HAMILTONIAN CYCLE} \in \text{NP}$ .

# Classes of complements

- Let  $\mathcal{C}$  be a class of decision problems.  $\text{co-}\mathcal{C}$  is formed by the complements of languages in  $\mathcal{C}$ , i.e.,  $L \in \text{co-}\mathcal{C}$  iff  $\bar{L} \in \mathcal{C}$ .

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$$P = \text{co-}P \quad \text{EXP} = \text{co-}EXP \quad \text{PSPACE} = \text{co-}PSPACE$$

- Not so clear for classes defined by NTMs.  
 $\text{co-NP} = \{x \mid \forall y R(x, y)\}$  where  $|y| = \text{poly}(|x|)$  and  $R$  can be decided in polynomial time.
- It seems different to assess that a property holds for a leaf in a tree than on all their leaves.

# Decision Problems in co-NP: Examples

- **PRIME**: Given an integer  $n$ , is  $n$  prime?
  - $\text{PRIME} = \{n \mid \forall a \leq n \ n \bmod a \neq 0\}$ .
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  - A polynomial time algorithm  $O(\lg n^6)$  was devised by Agrawal, Kayal and Saxena in 2002.

# Decision Problems in NP: Examples

- **FACTORIZATION:** Given an integer  $n$ , are there primes  $p_1, \dots, p_k$  s.t.  $x = p_1 \times \dots \times p_k$ .

# Decision Problems in NP: Examples

- **FACTORIZATION:** Given an integer  $n$ , are there primes  $p_1, \dots, p_k$  s.t.  $x = p_1 \times \dots \times p_k$ .
  - $p_1, \dots, p_k$  is the certificate.
  - $k$  and all the possible factors  $p$  are  $\leq n$ .
  - The verifier, on input  $n$  and  $p_1, \dots, p_k$ , computes  $p_1 \times \dots \times p_k$ , checks if the result is  $n$ . If so, checks that all the values are prime numbers.
  - This takes time  $O(k \lg n^2 + k \lg n^6)$ .
  - So, FACTORIZATION  $\in$  NP.

# Open questions

- Is  $P = NP$ ?
- $P \neq EXP$ , but is  $NP = EXP$ ?
- $P \subseteq NP \cap co-NP$ , but is  $P = NP \cap co-NP$ ?
- Is  $L = NL$ ?

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*The Determinant Problem:* Given an  $n \times n$  matrix  $M = (m_{ij})$ , compute

$$\det(M) = \sum_{\pi \in S_n} (-1)^\pi \prod_{i=1}^n m_{i,\pi(i)},$$

where  $(-1)^\pi = \begin{cases} -1 & \text{if } \pi \text{ has odd parity,} \\ +1 & \text{if } \pi \text{ has even parity.} \end{cases}$

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Can be solved in  $O(n^3)$ , using the LU decomposition.

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*The Permanent Problem:* Given an  $n \times n$  matrix  $M = (m_{ij})$ , compute

$$\text{perm}(M) = \sum_{\pi \in S_n} \prod_{i=1}^n m_{i,\pi(i)}.$$

$$\text{Perm} \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = aei + bfg + cdh + ceg + bdi + afh$$

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Main difference  $\det(M_1 \times M_2) = \det(M_1) \times \det(M_2)$  but  
 $\text{perm}(M_1 \times M_2) \neq \text{perm}(M_1) \times \text{perm}(M_2)$

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- In practice, it is synonym of a low degree polynomial time.
- Nevertheless, we some times use efficient as synonym of **best known** cost.
- How to differentiate strict exponential time from polynomial time?
- We cannot do that in most of the cases, **we relate such answers to some of the open complexity questions.**

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Classes: #P

# Function Problems: Examples

- **Reachability:** Given a directed graph  $G$  and two vertices  $u$  and  $v$ , find a path from  $u$  to  $v$ , if one exists.
- **Eulerian cycle:** Given a graph  $G$ , find an Eulerian cycle that traverses all the edges exactly once, if such a cycle exists.
- **Factoring:** Given an integer  $n$ , find its prime factors.
- **Subset Sum:** Given a sequence of positive integers  $S = \{a_1, \dots, a_n\}$  and  $t \in \mathbb{Z}$ , obtain  $I \subseteq \{1, \dots, n\}$  s.t.  $\sum_{i \in I} a_i = t$ .

# Optimization Problems: Examples

- **Max Cut:** Given a graph  $G$ , find a partition a partition  $V$  into  $V_1, V_2$ , such that it maximizes the number of crossing edges between  $V_1$  and  $V_2$ .
- **Min Cut:** Given a digraph  $G$ , find a partition  $V$  into  $V_1, V_2$ , such that it minimizes the number of edges going from  $V_1$  to  $V_2$ .
- **Shortest path:** Given a connected and edge weighted digraph  $G$  and  $s, t \in V(G)$ , find a path between  $s$  and  $t$  minimizing the sum of the weights in the path.

# Counting problems

- How many perfect matchings are there for a given bipartite graph?
- How many graph colorings using  $k$  colors are there for a particular graph  $G$ ?
- What is the value of the permanent of a given matrix whose entries are 0 or 1?
- How many different variable assignments will satisfy a given general boolean formula?

# More complexity classes?

Sure, plenty!

Look into the [Complexity zoo](#).

547 classes the last time I've had a look!