

# Randomized Algorithms

— P, NP, RP, PP, ZPP, BPP, ...

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# Outline

- 1 RAMs & Turing Machines
- 2 Complexity Classes
  - Deterministic Classes
  - Space Complexity Classes
  - Reduction & Completeness
  - Randomized Complexity Classes
- 3 Transformation of Probability Distributions

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# RAM (Random Access Machine)

- RAM is a model of computation used when describing and analyzing algorithms.
- A machine can perform operations involving registers and main memory.
- The *unit-cost* RAM: each instruction can be performed in one time step.
  - Too powerful; no known polynomial time simulation of this type of model by Turing machines.
- The *log-cost* RAM: each instruction requires time proportional to the logarithm of the size of its operands.

# Turing Machine

A physical Turing machine (with finite amount of tape).

## Deterministic Turing Machine

A deterministic Turing machine is a quadruple  $M = (S, \Sigma, \delta, s)$ .

- $S$ : a finite set of **states** ( $s$ : the initial state)
- $\Sigma$ : A finite set of **symbols** (including special symbols **BLANK** and **FIRST**).
- $\delta$ : the **transition function**.
  - $S \times \Sigma \mapsto (S \cup \{\text{HALT}, \text{YES}, \text{NO}\}) \times \Sigma \times \{\leftarrow, \rightarrow, \text{STAY}\}$ .
  - HALT, YES, NO: The three halting states not in  $S$ .

# Turing Machine (Input & Tape)

- The input to the TM: written on a tape.
- The TM, as an algorithm, may read from and write on this tape.
- Assume that HALT, YES, NO as well as the symbols  $\leftarrow$ ,  $\rightarrow$ , and STAY are not in  $S \cup \Sigma$ .
- The TM begins in the initial state  $s$  with its cursor at the first symbol FIRST of input  $x$ .
- The input is a string of  $(\Sigma \setminus \{\text{BLANK}, \text{FIRST}\})^*$ .
  - The left-most BLANK on the tape: the end of the input string.

# Turing Machine (Transition)

- The transition function  $\delta$ : can be thought as a *program*.
- In each step, the TM reads the symbol  $\alpha$  pointed by the cursor;
- Based on  $\alpha$  and the current state, choose:
  - a next state;
  - a symbol  $\beta$  to be overwritten on  $\alpha$ ;
  - a cursor motion direction from  $\{\leftarrow, \rightarrow, \text{STAY}\}$ .
- The cursor never falls off the left end of the input: FIRST.
- The BLANK symbol can be overwritten.

# Turing Machine (Acceptance & Reject)

- The TM has **accepted** the input  $x$ : if the TM halts in the YES state.
- The TM has **rejected** the input  $x$ : if the TM halts in the NO state.
- State HALT: for the computation of functions whose range is not Boolean (output of the function is written on the tape).



## Probabilistic Turing Machine

A probabilistic Turing machine is a Turing machine augmented with the ability to **generate an unbiased coin flip in one step**.

- This corresponds to a **randomized algorithm**.

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# SAT

An instance of satisfiability (SAT):

$$(x_1 \wedge \neg x_2 \wedge x_4) \vee (\neg x_3 \wedge \neg x_4 \wedge x_5) \vee (\neg x_1 \wedge x_2 \wedge x_4 \wedge \neg x_5)$$

- $x_1, x_2, \dots$ : variables
- $\neg x_1, x_2$ : literals
- $(\dots)$ : clauses

# Language Recognition Problems

## Language Recognition Problems

Any decision problem can be treated as a language recognition problem.

- $\Sigma^*$ : the set of all possible strings over  $\Sigma$ .
- $|S|$ : length of string  $s$ .

A language  $L \subseteq \Sigma^*$  is any collection of strings over  $\Sigma$ .

## A Language Recognition Problem

Decide whether a given string  $x \in \Sigma^*$  belongs to  $L$ .

## Complexity Class

A collection of languages all of whose recognition problems can be solved under prescribed bounds on the computational resources.

# $P$ & $NP$

## $P$

The class  **$P$**  consists of all languages  $L$  which has a polynomial time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

- $x \in L \Rightarrow A(x)$  accepts;
- $x \notin L \Rightarrow A(x)$  rejects.

## $NP$

The class  **$NP$**  consists of all languages  $L$  which has a polynomial time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

- $x \in L \Rightarrow \exists y \in \Sigma^*, A(x, y)$  accepts for  $|y| \leq \text{poly}(|x|)$ ;
- $x \notin L \Rightarrow \forall y \in \Sigma^*, A(x, y)$  rejects.

For example, given an instance of satisfiability (SAT):

$$\mathbf{x} = (x_1 \wedge \neg x_2 \wedge x_4) \vee (\neg x_3 \wedge \neg x_4 \wedge x_5) \vee (\neg x_1 \wedge x_2 \wedge x_4 \wedge \neg x_5)$$

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Suppose we have the “proof”  $\mathbf{y}$  as

$$x_1 = \text{True}, x_2 = \text{False}, x_3 = \text{True}, x_4 = \text{True}, x_5 = \text{True}$$

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We can check that  $(\mathbf{x}, \mathbf{y}) \in L$  (encoded as string of  $O(\log n)$  space in the tape) in polynomial time, where  $L$  denote the set of all satisfiable formula.



# A Useful, Alternative Viewpoint

The class  **$P$**  consists of all language  $L$  such that for any  $x \in L$ , a proof of  $x \in L$  (represented by the string  $y$ ) can be **found** and **verified** in polynomial time.

The class  **$NP$**  consists of all language  $L$  such that for any  $x \in L$ , a proof of  $x \in L$  (represented by the string  $y$ ) can be **verified** in polynomial time.

Obviously,

$$P \subseteq NP.$$

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# Complementary Classes

For any complexity class  $\mathcal{C}$ , the complementary class  $\text{co-}\mathcal{C}$  is the set of languages whose complement is in  $\mathcal{C}$ . That is,

$$\text{co-}\mathcal{C} = \{L \mid \bar{L} \in \mathcal{C}\}.$$

## Examples: $\text{co-}P$ & $\text{co-NP}$

### $\text{co-}P$

The class  $\text{co-}P$  consists of all languages  $L$  which has a polynomial time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

- $x \notin L \Rightarrow A(x)$  accepts;
- $x \in L \Rightarrow A(x)$  rejects.

### $\text{co-NP}$

The class  $\text{co-NP}$  consists of all languages  $L$  which has a polynomial time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

- $x \notin L \Rightarrow \exists y \in \Sigma^*, A(x, y)$  accepts for  $|y| \leq \text{poly}(|x|)$ ;
- $x \in L \Rightarrow \forall y \in \Sigma^*, A(x, y)$  rejects..

**Open Questions:**  $P = NP \cap \text{co-NP}$ ?  $NP = \text{co-NP}$ ?

Similarly, ...

# ***EXP & NEXP***

## EXP

The class **EXP** consists of all languages  $L$  which has an **exponential** time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

- $x \in L \Rightarrow A(x)$  accepts;
- $x \notin L \Rightarrow A(x)$  rejects.

## NEXP

The class **NEXP** consists of all languages  $L$  which has an **exponential** time algorithm  $A$  s.t. for any input  $x \in \Sigma^*$ ,

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The class **EXP** consists of all language  $L$  such that for any  $x \in L$ , a proof of  $x \in L$  (represented by the string  $y$ ) can be **found** and **verified** in **exponential** time.

The class **NEXP** consists of all language  $L$  such that for any  $x \in L$ , a proof of  $x \in L$  (represented by the string  $y$ ) can be **verified** in **exponential** time.

Obviously,

$$\mathbf{EXP} \subseteq \mathbf{NEXP}.$$

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# Space

- The space used by a TM: the number of distinct positions on the tape that are scanned during an execution.
  - For RAMs, its the number of words of memory required by an algorithm.
- **PSPACE** and **NPSPACE**: resembles the settings of **P** and **NP** but requiring polynomial space.
- A **PSPACE** algorithm may run for super-polynomial time (e.g.,  $2^{\text{poly}(n)}$ ).
- Known results: **PSPACE** = **NPSPACE**, **PSPACE** = co-**PSPACE**.
  - Savitch's theorem: a deterministic Turing machine can simulate a nondeterministic Turing machine without needing much more space.

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# Reduction

## Polynomial Reduction

A polynomial reduction from a language  $L_1 \subseteq \Sigma^*$  to a language  $L_2 \subseteq \Sigma^*$  is a function  $f : \Sigma^* \mapsto \Sigma^*$  such that

- $\exists$  a polynomial time algorithm that computes  $f$
- $\forall x \in \Sigma^*, x_1 \in L_1$  if and only if  $f(x) \in L_2$ .

# Completeness

## **NP**-hard

A language  $L$  is **NP**-hard if, for all  $L' \in \mathbf{NP}$ , there is a polynomial reduction from  $L'$  to  $L$ .

## **NP**-complete

A language  $L$  is **NP**-complete if it is in **NP** and is **NP**-hard.

The first **NP**-complete problem: SAT (Cook-Levin Theorem (1971)).

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# *RP*

## *RP*

The class **RP** (i.e., Randomized Polynomial time) consists of all languages  $L$  that have a randomized algorithm  $A$  which runs in **worst-case polynomial time** such that for any input  $x \in \Sigma^*$ :

- $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{1}{2}$ .
- $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] = 0$ .
- Err only when  $x \in L$ .  $\Rightarrow$  **one-sided error**.

# co-*RP*

## *RP*

The class co-***RP*** (i.e., complement Randomized Polynomial time) consists of all languages  $L$  that have a randomized algorithm  $A$  which runs in worst-case polynomial time such that for any input  $x \in \Sigma^*$ :

- $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{1}{2}$ .
- $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] = 0$ .
- Err only when  $x \notin L$ .  $\Rightarrow$  one-sided error.

# Exercise (3%)

Assume that we have the following class:

**$RP'$**

The class  **$RP'$**  consists of all languages  $L$  that have a randomized algorithm  $A$  which runs in worst-case polynomial time such that for any input  $x \in \Sigma^*$ :

- $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{1}{n^2}.$
- $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] = 0.$

Prove that  **$RP' = RP$** .



$$RP \cap co-RP$$

## **ZPP** (Zero-error Probabilistic Polynomial time)

The class **ZPP** is the class of languages that have Las Vegas algorithms running in expected polynomial time.

# Why *ZPP*?

- Suppose we have a language  $L \in \mathbf{RP} \cap \text{co-}\mathbf{RP}$ .
- $L$  can be recognized by an *RP* algorithm  $A$  and a *co-RP* algorithm  $B$ .

## A Las Vegas algorithm

Given the input  $x$ , perform the following procedure in iterations.

- 1 If  $A(x)$  accepts, then  $x$  must be a YES-instance;
- 2 Otherwise, if  $B(x)$  rejects, then  $x$  must be a NO-instance.
- 3 If neither of above occurs, continue to next iteration.

- The expected number of iterations is bounded!

# PP

## PP

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- $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] > \frac{1}{2}$ .
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- 
- To reduce the error probability of a two-sided error algorithm, we can perform several independent iterations on the same input.
  - Output the majority answer of these iterations.

# ***BPP***

## ***BPP***

The class **BPP** (i.e., Bounded-error Probabilistic Polynomial time) consists of all languages  $L$  that have a randomized algorithm  $A$  running in **worst-case polynomial time** such that for any input  $x \in \Sigma^*$ :

- $x \in L \Rightarrow \Pr[A(x) \text{ accepts}] \geq \frac{3}{4}$ .
- $x \notin L \Rightarrow \Pr[A(x) \text{ accepts}] \leq \frac{1}{4}$ .

# BPP

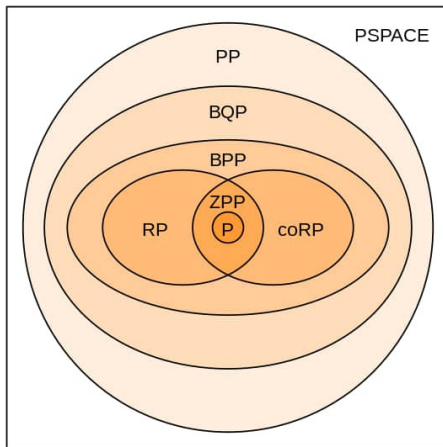
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# Randomized Complexity Classes

Source: Wikipedia



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## $p$ -coin & Transforming a Probability Distribution

### $p$ -coin

A coin is called a  $p$ -coin if it shows HEAD after one coin-flipping.

### Probability Distribution Transformations

A function that transforms a  $p$ -coin to get a  $q$ -coin, for  $0 < p, q < 1$ , is called a  $p$ -to- $q$  transformation.

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Easy cases:

- $p$ -to-0 and  $p$ -to-1..

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- $p$ -to- $(1 - p)$

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- $p$ -to-0 and  $p$ -to-1..
- $p$ -to- $(1 - p)$
- $p$ -to- $p^2$ .

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Easy cases:

- $p$ -to-0 and  $p$ -to-1..
- $p$ -to- $(1 - p)$
- $p$ -to- $p^2$ .
- $p$ -to- $\binom{n}{k} p^k (1 - p)^{n-k}$ , for  $n \in \mathbb{N}$  and  $k \in \{0, 1, \dots, n\}$ .

## Exercise (2%)

Given a  $p$ -coin, where  $0 < p < 1$ .

- Repeat the following steps until it returns YES or NO.
  - 1 flip the  $p$ -coin twice.
  - 2 if the results are HEAD-TAIL, return YES;
  - 3 else if the results are TAIL-HEAD, return NO;
  - 4 otherwise, continue to next iteration
- Please prove that the above procedure is a  $p$ -to- $\frac{1}{2}$  transformation (i.e., deriving a fair coin).
- Please compute the expected number of coin-flips of the above procedure.

# Concatenation Rule

## Concatenation Rule

Given a  $p$ -coin and a  $q$ -coin, we can derive a  $pq$ -coin as follows.

- 1 First, simulate the  $p$ -coin. If it is TAIL, output TAIL
- 2 Otherwise, simulate the  $q$ -coin and output the outcome.

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- ① First, simulate the  $p$ -coin. If it is TAIL, output TAIL
  - ② Otherwise, simulate the  $q$ -coin and output the outcome.
- By the Concatenation Rule & the exercise, we can derive a  $p/2$ -coin when we are given a  $p$ -coin.



# Discussions