

# Computational Model

Meng Ziyu

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## 1 Class $\mathcal{PP}$

$\mathcal{PP}$ , whose full name is **Probabilistic Polynomial Time**, is a class of computational model. What makes it different from classes  $\mathcal{P}, \mathcal{NP}, \mathcal{NPC}$  is that it uses Randomized Algorithm.

What's Randomized Algorithm? Take an example, given an binary tree, from the root node to the leaf nodes, for each step, we toss a coin to choose one of the two directions. During this process, since that for each time we toss a coin, it's probabilistic and independent from the prior choices, we say that this algorithm is randomized.

So, let's talk back to  $\mathcal{PP}$ , the definition of it is described below.

**Class  $\mathcal{PP}$ :**  *$\mathcal{PP}$  is a class of decision problems solvable by a Probabilistic Turing machine in polynomial time, with an error probability of less than  $1/2$  for all instances.*

A Probabilistic Turing machine is a kind of non-deterministic Turing machine that chooses between the available transitions at each point according to some probability distribution. Thus, for a given input, the Probabilistic Turing machine has stochastic result. More detailed speaking, for a deterministic Turing machine, it consists 7 elements:  $M = (Q, \Sigma, \Gamma, q_0, A, R, \delta)$ , but for a Probabilistic Turing machine, it has 8 elements:  $M = (Q, \Sigma, \Gamma, q_0, A, R, \delta_1, \delta_2)$ . In each step, the Probabilistic Turing machine probabilistically and independently choose the transition function  $\delta_1$  or  $\delta_2$ .

Because of this feature, a certain input may be accepted this time and be rejected in the next time. To accommodate this, a language  $L$  is said to be recognized with error probability  $\epsilon$  by a Probabilistic Turing machine  $M$  that: 1.  $\forall w \in L, \Pr[M \text{ accepts } w] \geq 1-\epsilon$ ; 2.  $\forall w \notin L, \Pr[M \text{ rejects } w] \geq 1-\epsilon$ . And we call this kind of Turing machines that are polynomially-bound and probabilistic as **Probabilistic Polynomial-Time machines** ( $\mathcal{PPT}$ ). Hence,  $\mathcal{PP}$  is the complexity class containing all problems solvable by a PPT machine with an error probability of less than  $1/2$  ( $\epsilon \leq 1/2$ ).

**Difference between non-deterministic and probabilistic.** The non-deterministic Turing machine is an unreal model that is useful for solving  $\mathcal{NP}$  problems. But for probabilistic model, it's a real model.

## 2 Class $\mathcal{BPP}$

$\mathcal{BPP}$ , whose full name is **Bounded-error Probabilistic Polynomial time**.  $\mathcal{BPP}$  is one of the largest practical classes of problems, meaning most problems of interest in  $\mathcal{BPP}$  have efficient probabilistic algorithms that can be run quickly on real modern machines

We give the definition of it.

**Class  $\mathcal{BPP}$ :**  *$\mathcal{BPP}$  is the class of decision problems solvable by a probabilistic Turing machine in polynomial time with an error probability bounded away from  $1/3$  for all instances.*

The phrase "bounded-probability" means the success probability is bounded away from the number  $1/2$ . In fact, replacing the  $2/3$  by any other constant greater than  $1/2$  will not change the class defined.

**Theorem:** *The following are both equivalent definitions of  $\mathcal{BPP}$ :*

1.  $L \in \mathcal{BPP}$  if there exists a  $\mathcal{PPT}$  machine  $M$  and a positive polynomial that

$$\Pr[M(x) = \chi_L(x)] \geq \frac{1}{2} + \frac{1}{|p(x)|}$$

2.  $L \in \mathcal{BPP}$  if there exists a  $\mathcal{PPT}$  machine  $M$  and a positive polynomial that

$$\Pr[M(x) = \chi_L(x)] \geq 1 - \frac{1}{2^{|p(x)|}}$$

\* $M(x) = 1$  iff  $M$  accepts  $x$ ,  $M(x) = 0$  iff  $M$  rejects  $x$

\* $\chi_L(x) = 1$  iff  $x \in L$ ,  $\chi_L(x) = 0$  iff  $x \notin L$

We could see that the main difference between  $\mathcal{PP}$  and  $\mathcal{BPP}$  is that they are bounded differently. And thus we have  $\mathcal{BPP} \subset \mathcal{PP}$ . And also,  $\mathcal{BPP}$  contains  $\mathcal{P}$ , since a deterministic machine is a special case of a probabilistic machine. So we have  $\mathcal{P} \subset \mathcal{BPP} \subset \mathcal{PP}$ .

### 3 Class $\mathcal{P}/poly$

In computational complexity theory,  $\mathcal{P}/poly$  is the complexity class of languages recognized by a polynomial-time Turing machine with a polynomial-bounded advice function.

**What is Advice Function?** An advice is an extra input to a Turing machine that is allowed to depend on the length  $n$  of the input, but not on the input itself. A decision problem is in the complexity class  $\mathcal{P}/f(n)$  if there is a polynomial time Turing machine  $M$  with the following property: *For any  $n$ , there is an advice string  $A$  of length  $f(n)$  such that, for any input  $x$  of length  $n$ , the machine  $M$  correctly decides the problem on the input  $x$ , given  $x$  and  $A$ .*

The formal Definition of  $\mathcal{P}/poly$  gives below:

**Non-uniform Polynomial Time:** *The complexity class non-uniform polynomial time (denoted  $\mathcal{P}/poly$ ) is the class of languages  $L$  that can be recognized by a non-uniform sequence of polynomial time "machines". Namely,  $L \in \mathcal{P}/poly$  if there exists an infinite sequence of machines  $M_1, M_2, \dots$  that satisfying the following:*

1. *There exists a polynomial  $p()$  such that for every  $n$ , the description of  $M_n$  has a length bounded above  $p(n)$ ;*
2. *There exists a polynomial  $q()$  such that for every  $n$ , the running time of  $M_n$  on each input of length  $n$  is bounded above  $q(n)$ ;*
3. *For every  $n$  and  $x \in \{0, 1\}^n$ , the machine  $M_n$  accepts  $x$  iff  $x \in L$ .*

The most common complexity class involving advice is  $\mathcal{P}/poly$  where advice length  $f(n)$  can be any polynomial in  $n$ .  $\mathcal{P}/poly$  is equal to the class of decision problems such that, for every  $n$ , there exists a polynomial size Boolean circuit correctly deciding the problem on all inputs of length  $n$ . It's easy to prove it from a high level. The size of a Boolean Circuit is number of its edges and it has size  $O(|<M_n>| + n + t^2(n))$  where  $M_n$  is the machine that accepts input with length  $n$  and  $t(n)$  is a bound on the running time of  $M_n$ , and since the size of circuit is polynomial, the running time of it is also polynomial.