

# Time Complexity

## 1 Time complexity

Reading: Sipser chapter 7.

From now on we will be interested not only in whether a Turing machine can compute a function, but also in how many resources (space, time) uses in order to do so.

**Definition 1** (Running time). *Let  $M$  be a deterministic Turing machine that halts on all inputs. The running time (time complexity) of  $M$  is the function  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of steps that  $M$  takes on an input of length  $n$ . We say that  $M$  runs in time  $f(n)$  and  $M$  is an  $f(n)$  time Turing machine.*

We will measure running time asymptotically.

**Definition 2** (Upper bounds, Big-O). *Let  $f, g : \mathbb{N} \rightarrow \mathbb{R}^+$ . We say that  $f(n) = O(g(n))$  if there exist  $c, n_0 \in \mathbb{N}$  such that for all  $n \geq n_0$ , we have  $f(n) \leq c \cdot g(n)$ . In this case the function  $g$  is said to be an asymptotic upper bound on  $f$ .*

Upper bounds of the form  $n^c$ , where  $c > 0$  is a constant, are called polynomial bounds, while upper bounds of the form  $2^{(n^c)}$  are called exponential bounds.

**Definition 3** (Small-o). *Let  $f, g : \mathbb{N} \rightarrow \mathbb{R}$ . We say that  $f(n) = o(g(n))$  if*

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$$

*That is, for every  $c > 0$  there exists  $n_0$  such that  $f(n) < c \cdot g(n)$  for all  $n \geq n_0$ .*

**Example 1.** *What is the runtime of the machine that decides the language  $A = \{0^k 1^k \mid k \geq 0\}$ ? Consider a TM  $M$  that solves this problem. On input  $w$ :*

1. *Check that the input has the desired form  $0^* 1^*$ .*
2. *While there are both 0s and 1s left on the tape:*
  - *Move the tape head from left to right, crossing off one zero and one one. Then go back to the leftmost end of the tape.*
3. *If no 0s and 1s are left, accept. Otherwise, reject.*

The first operation of checking the form of the input takes  $O(n)$  steps. In the second step, the tape head moves from the left end of the tape to the right end of the input, and then back, which gives  $2n$  steps. Each iteration of step 2 crosses off at most 2 symbols, so there will be at most  $n/2$  iterations of this type. In the third step, the tape head moves across the tape one more time, for a total of  $n$  steps. Thus in total we have  $O(n^2)$  steps.

Recall that a Turing machine  $M$  decides a language  $L \in \{0, 1\}^*$  if for any word  $w \in \{0, 1\}^*$ , the machine  $M$  accepts if  $w \in L$  and rejects if  $w \notin L$ .

**Definition 4** (The class DTIME). Let  $f : \mathbb{N} \rightarrow \mathbb{N}$  be a function. A language  $L$  is in  $DTIME(t(n))$  if there exists a Turing machine running in  $O(t(n))$  steps and decides  $L$ .

## 2 The class P

**Definition 5** (The class P).  $P$  is the class of languages decidable in polynomial time (on deterministic, one-tape) Turing machines:

$$P = \bigcup_k DTIME(n^k)$$

The class  $P$ :

- remains invariant for all models of computation that are polynomially-time equivalent to deterministic one-tape Turing machines
- corresponds to problems that are realistically solvable on a real computer.

Some examples of problems in  $P$  are as follows.

**Example 2** (PATH). Consider the language

$$PATH = \{\langle G, s, t \rangle \mid G \text{ is a directed graph and there is a path from } s \text{ to } t\}$$

**Theorem 2.1.**  $PATH \in P$ .

*Proof.* Consider the following TM for solving  $PATH$ . On input  $\langle G, s, t \rangle$ ,

1. Place a mark on node  $s$ .
2. Repeat until no new nodes are marked:
  - Scan all the edges of  $G$ . If an edge  $(a, b)$  is found outgoing from a marked node  $a$  to an unmarked node  $b$ , mark node  $b$ .
3. If  $t$  is marked, accept. Otherwise, reject.

□

### 3 The class NP

Next we will look at problems where coming up with a solution is (intuitively) harder than verifying the solution. For instance, when solving a crossword puzzle, it's harder to solve it than to check that a solution given by someone else. Similarly with a math problem where the hard part is getting the right idea and getting it to work, while checking that a solution is correct (i.e. tracing the sequence of derivations and making sure they follow logically from each other) is easier.

Thus in contrast to  $P$ , which is the class of “efficiently solvable problems”, NP is the class of “efficiently verifiable solutions”.

**Definition 6** (Verifier). *A verifier for a language  $L$  is an algorithm  $V$  that takes inputs of the form  $\langle w, u \rangle$  and*

$$w \in L \iff \text{there exists } u \text{ such that } V \text{ accepts } \langle w, u \rangle$$

*The string  $u$  with this property is called a certificate (or witness) for  $w$ . That is,*

$$L = \{w \mid V \text{ accepts } \langle w, u \rangle \text{ for some string } u\}$$

**Definition 7** (Polynomial-time verifier). *A polynomial-time verifier for a language  $L$  is an algorithm  $V$  that takes inputs of the form  $\langle w, u \rangle$  and runs in time  $p(|w|)$  for some polynomial  $p : \mathbb{N} \rightarrow \mathbb{N}$ , such that*

$$w \in L \iff \text{there exists } u \text{ such that } V \text{ accepts } \langle w, u \rangle$$

**Definition 8** (The class NP). *NP is the class of languages  $L \in \{0, 1\}^*$  that have polynomial time verifiers.*

**Example 3** (Independent Set). *The language*

$$INDSET = \{\langle G, k \rangle \mid \text{there exists } S \subseteq V(G) \text{ such that } |S| \geq k \text{ and } \forall u, v \in S, (u, v) \notin E(G)\}$$

*Consider the following TM  $M$ , which on input  $\langle G, k, u \rangle$ , outputs 1 if and only if  $u$  encodes a list of  $k$  vertices of  $G$  with no edges between them. Thus  $\langle G, k \rangle \in INDSET$  if and only if there exists a string  $u$  such that  $M(\langle G, k \rangle u) = 1$ , so  $INDSET \in NP$ .*

*The list of vertices  $u$  represents the certificate that  $\langle G, k \rangle$  is in  $INDSET$ . Also note that if  $n$  is the number of vertices in  $G$ , then a list of  $k$  vertices can be encoded using  $O(k \log n)$  bits. So  $|u| = O(n \log n)$ , which is polynomial in the size of the representation of  $G$ .*

**Example 4** (Graph Isomorphism). *Given two graphs  $G$  and  $H$ , is there a bijective function  $f : V(G) \rightarrow V(H)$  such that  $(u, v) \in E(G)$  if and only if  $(f(u), f(v)) \in E(H)$ ?*

*To show that Graph Isomorphism is in NP, we need to find a polynomial time verifier  $V$  for it. The verifier should receive as input two graphs  $(G, H)$  and a certificate  $u$ , and accept  $(G, H, u)$  if and only if  $G$  and  $H$  are isomorphic. Moreover, the runtime of  $V$  must be bounded by  $p(|(G, H)|)$  for some polynomial  $p$ .*

*Consider the following algorithm  $V$ , which receives as input two graphs  $G$  and  $H$  on  $n$  vertices and a string  $u$ :*

1. *If  $u = (i_1, \dots, i_n)$  is not a permutation of  $(1, \dots, n)$ , reject.*
2. *Else, permute the vertices of  $G$  as indicated by  $u$ . Check that the permuted graph  $G$  is identical to  $H$ .*

Step 1 runs in  $O(n^2)$  and step 2 runs in  $O(n + m)$ , where  $m$  is the number of edges. So the verifier  $V$  runs in time  $O(n^2)$ , thus the Graph-Isomorphism problem is in NP.

**Example 5** (Subset-Sum). Given natural numbers  $U = \{w_1 \dots w_n\}$  and a number  $W$ , is there a subset  $S \subseteq U$  such that  $\sum_{a \in S} a = W$ ?

A certificate for this problem is a subset of numbers that sum up to  $W$ .