

Complexity Theory

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Outline

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Time Complexity of Deterministic Turing Machine

Let M be a deterministic Turing machine that halts on all inputs. The *running time* or *time complexity* of M is the function $f: \mathcal{N} \rightarrow \mathcal{N}$, where $f(n)$ is the maximum number of steps that M uses on any input of length n . If $f(n)$ is the running time of M , we say that M runs in time $f(n)$ and that M is an $f(n)$ time Turing machine. Customarily we use n to represent the length of the input.

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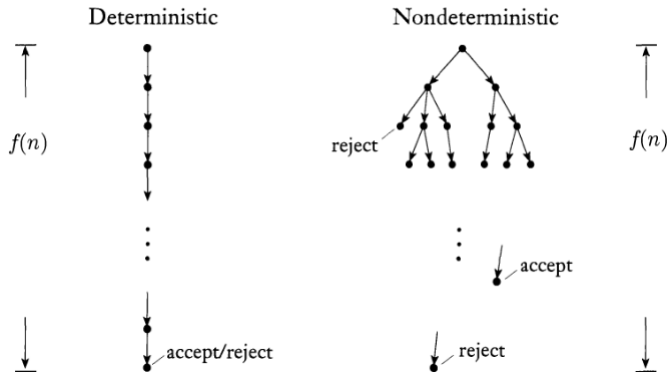
Let $t: \mathcal{N} \rightarrow \mathcal{R}^+$ be a function. Define the *time complexity class*, $\mathbf{TIME}(t(n))$, to be the collection of all languages that are decidable by an $O(t(n))$ time Turing machine.

Time Complexity of Nondeterministic Turing Machine

Let N be a nondeterministic Turing machine that is a decider. The *running time* of N is the function $f: \mathcal{N} \rightarrow \mathcal{N}$, where $f(n)$ is the maximum number of steps that N uses on any branch of its computation on any input of length n , as shown in the following figure.

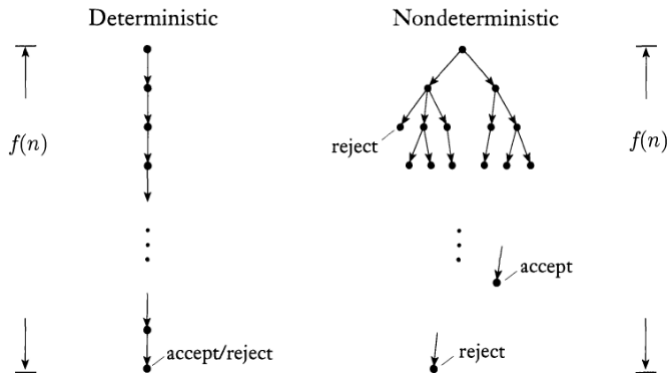
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$\text{NTIME}(t(n)) = \{L \mid L \text{ is a language decided by a } O(t(n)) \text{ time nondeterministic Turing machine}\}.$

Complexity Relationships among Models

Theorem

Let $t(n)$ be a function, where $t(n) \geq n$. Then every $t(n)$ time multitape Turing machine has an equivalent $\mathcal{O}(t^2(n))$ time single-tape Turing machine.

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The Class P and NP

P is the class of languages that are decidable in polynomial time on a deterministic single-tape Turing machine. In other words,

$$P = \bigcup_k \text{TIME}(n^k).$$

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A *verifier* for a language A is an algorithm V , where

$$A = \{w \mid V \text{ accepts } \langle w, c \rangle \text{ for some string } c\}.$$

We measure the time of a verifier only in terms of the length of w , so a *polynomial time verifier* runs in polynomial time in the length of w . A language A is *polynomially verifiable* if it has a polynomial time verifier.

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NP is the class of languages that have polynomial time verifiers.

Examples

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- ▶ $\text{SUBSET-SUM} = \{\langle S, t \rangle \mid S = \{x_1, \dots, x_k\} \text{ and for some } \{y_1, \dots, y_l\} \subseteq \{x_1, \dots, x_k\}, \text{ we have } \sum y_i = t\}$ is in NP.

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$N =$ “On input w of length n :

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V = “On input $\langle w, c \rangle$, where w and c are strings:

1. Simulate N on input w , treating each symbol of c as a description of the nondeterministic choice to make at each step (as in the proof of Theorem 3.16).
2. If this branch of N 's computation accepts, *accept*; otherwise, *reject*.”



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- ▶ To prove $P \neq NP$, we need to show that there exists a problem in NP which can't be solved in polynomial time;
- ▶ To prove $P = NP$, we need to show that all the problems in NP can be solved in polynomial time.

Polynomial Time Reducibility

A function $f: \Sigma^* \longrightarrow \Sigma^*$ is a *polynomial time computable function* if some polynomial time Turing machine M exists that halts with just $f(w)$ on its tape, when started on any input w .

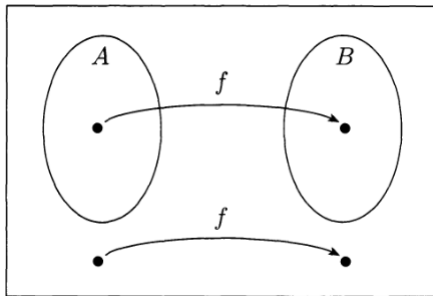
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Language A is ***polynomial time mapping reducible***,¹ or simply ***polynomial time reducible***, to language B , written $A \leq_P B$, if a polynomial time computable function $f: \Sigma^* \longrightarrow \Sigma^*$ exists, where for every w ,

$$w \in A \iff f(w) \in B.$$

The function f is called the ***polynomial time reduction*** of A to B .



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If B is NP -complete and $B \leq_P C$ for $C \in NP$, then C is NP -complete.

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- ▶ In 1972, Richard Karp showed 21 diverse combinatorial and graph theoretical problems, each infamous for its computational intractability, are NP -complete;
- ▶ Garey and Johnson's 1979 book, "Computers and Intractability: A Guide to NP-Completeness" collected much more NP -complete problems.