

Intro to complexity and asymptotic analysis

Textbook: 7.1

Running time

- ▶ A deterministic **halting** TM has a max number of steps it will ever take
- ▶ This is usually dependent on the input, most often the length
- ▶ We customarily let n be the length of the input

Def: Let M be a deterministic TM that always halts. The **running time** or **time complexity** of M is the function $f : \mathbb{N} \rightarrow \mathbb{N}$ where $f(n)$ is the maximal number of steps M will take on input w .

- ▶ It's often easier to only look at very large inputs, where we can neglect lower-order terms. This is called **asymptotic analysis**

Big “O”

- ▶ If we wanted to say $f(n)$ grows at a rate *no faster than* $g(n)$, we could say $f(n)$ **is** $O(g(n))$
 - ▶ In some texts, $f(n) = O(g(n))$ or $f(n) \in O(g(n))$
 - ▶ Colloquially, $f(n)$ is order $g(n)$

Def: Big-O notation. Formally, if there exist positive real numbers c and n_0 such that for all $n > n_0$

$$f(n) \leq cg(n)$$

then we say $g(n)$ **is an asymptotic upper bound on** $f(n)$ and $f(n)$ **is** $O(g(n))$.

- ▶ Means f is less than or equal to g if we disregard differences up to a constant factor

Big “O”

- ▶ Weirdly, we don't need to specify a base when logarithms are involved!
 - ▶ Since $\log_c(n)$ is always proportional to $\log_d(n)$
 - ▶ We usually use $\log(n)$, $\ln(n)$, or $\lg(n)$
- ▶ $f(n)$ being $O(1)$ means that $f(n)$ is bounded by some constant
- ▶ Big-O is the most common type of analysis, since it analyzes the “worst-case” scenario
 - ▶ It was popularized by Donald Knuth in the 70's

Little “o”

- ▶ Big-O gave us the option to say $f(n)$ is asymptotically *no more than* $g(n)$: What if we want to say $f(n)$ is *less than* $g(n)$? We use **little-o** notation!
- ▶ Little-o can be thought of as saying something grows insignificantly when compared to something else

Def: For functions f, g , we say f is $o(g(n))$ if

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

In other words, as $n \rightarrow \infty$, $g(n)$ grown infinitely larger and faster than $f(n)$.

- ▶ The difference is subtle, but far-reaching!

$$f(n) \text{ is } O(g(n)) \iff \exists c \exists n_0 \forall n [n_0 < n \rightarrow f(n) \leq cg(n)]$$

$$f(n) \text{ is } o(g(n)) \iff \forall c \exists n_0 \forall n [n_0 \leq n \rightarrow f(n) < cg(n)]$$

Polynomiality and nonpolynomiality

- ▶ Many algorithms run in time EG $O(n)$, $O(n^2)$, $O(n^{100})$
- ▶ In real life, anything as bad as or worse than $O(n^2)$ is too costly: We don't even bother with EG $O(2^n)$
- ▶ However, $O(2^n)$ is very theoretically interesting!

Def: An algorithm with time complexity of the form $O(n^c)$ for constant c is said to be **polynomial**. If the complexity is not of this form, it is said to be **nonpolynomial**. This is called the **polynomiality/nonpolynomiality** of the algorithm.

Complexity classes and the TIME operator

- ▶ There are many problems which can be decided by a deterministic TM in $f(n)$ time
- ▶ We introduce the TIME operator to find the set of all languages decidable in some time limit

Def: For some function f , the **time complexity class** $\text{TIME}(t(n))$ is the set of all **languages** decidable in $O(t(n))$ time by a **deterministic** Turing Machine.

Note: Computationally equivalent models solving the same problem **may have different time complexities!** Namely, dTMs and nTMs are computationally equivalent but have vastly different complexity classes.

Nondeterministic Running Time

- ▶ Running time makes sense on dTMs, but what about nTMs?
- ▶ Some branches of the nondeterminism may live longer than others
- ▶ We'll just use **the longest running time of any branch**

Def: Let N be a nondeterministic decider Turing machine on input of length n . The **running time** $f(n)$ of N is the maximum number of steps that N uses on any branch of its computation.

- ▶ We previously gave an algorithm for simulating an nTM with a dTM
- ▶ That algorithm ran in $O(2^n)$ deterministic timesteps for n nondeterministic timesteps
- ▶ Therefore, any algorithm running in $f(n)$ nondeterministic time runs in **at worst** $O(2^{f(n)})$.

The NTIME operator

- ▶ It is also interesting to study the set of all languages decidable in $f(n)$ **nondeterministic** time: We use the operator NTIME

Def: For some function f , the **time complexity class** $\text{NTIME}(f(n))$ is the set of all languages decidable by a **nondeterministic** TM in $O(f(n))$ time.

Note: All “reasonable” (textbook’s word) deterministic models of computation are polynomial-time convertible to one another

Next up: P and NP