## 14 Computational Complexity and Polynomial Time

A program is only said to compute, or decide, a problem if it eventually stops on every input. For computability, it does not matter how long the program takes to produce an answer.
But, from a practical standpoint, time matters. The data switches that form the backbone of the internet process a data packet roughly every 1020 nanoseconds. They can only afford to run extremely fast algorithms. For most everyday purposes, an algorithm that gets its answer in a few seconds is fast enough, and for some problems, a program that takes a few minutes or hours is fast enough. But a program that takes years is usually not acceptable.
With this section we start to look at what can be computed efficiently. To do that, we need to choose a reasonable definition of an "efficient" algorithm, and to study which problems can be solved efficiently under that definition.
A definition of efficiency cannot be based on any fixed amount of time. The larger the input is, the longer we expect a program to take, so a reasonable notion of efficiency should be concerned with the time that a program takes as a function of the length of the input. Also, a faster computer will produce an answer faster even for the same algorithm, and we need a way to get processor speed out the way.

### 14.1 The Class P

A program performs a sequence of instructions, and the time that it uses is just a count of the number of instructions that it performs before it stops. In this view, time has no units; it is a pure number. We assume that it takes at least one instruction to look at a symbol in the input and at least one instruction to write one symbol in the output.

Definition 14.1. $\operatorname{Time}(p, x)$ is the number of instructions that program $p$ takes when it is run on input $x$. If $p(x) \uparrow$, then $\operatorname{Time}(p, x)=\infty$.

Definition 14.2. Let $f: \mathcal{N} \rightarrow \mathcal{N}$. A program $p$ runs in time $O(f(n))$ if there exists constants $a$ and $c$ so that, for all $n>a$ and all strings $x$ of length $n, \operatorname{Time}(p, x) \leq c \cdot f(n)$.

Definition 14.3. Program $p$ runs in polynomial time if there exists a positive integer $k$ so that $p$ runs in time $O\left(n^{k}\right)$. When $p$ runs in polynomial time, we say that $p$ is a polynomial-time algorithm.

Definition 14.4. $P$ is the class of all decision problems that have polynomialtime algorithms.
Notice that P is a set of problems, not a set of programs. It makes no sense to say that a program is in P .

### 14.2 Examples of Problems that Are In P

### 14.2.1 Example: is $x$ a Palindrome?

A palindrome is a string such as "aabaa" that is the same forwards and backwards. The Palindrome Problem is the following decision problem.

Input. String $x$.
Question. Is $x$ a palindrome?
The Palindrome Problem is in P. You should be able to figure out an algorithm that solves the Palindrome Problem in time $O(n)$.

### 14.2.2 Example: Does DFA $M$ Accept $x$ ?

We have seen that it is computable to determine if a given DFA $M$ accepts a given string $x$. It should also be clear that an algorithm can do that in time that is proportional to the product of the length of the description of $M$ and the length of $x$. (You should be able to do much better than that, but it is fast enough for our purposes.) If the input has length $n$, that is surely $O\left(n^{2}\right)$ time.

### 14.2.3 Example: Is $x \cdot y=z$ ?

How might you solve the following decision problem?
Input. Three positive integers $x, y$ and $z$.
Question. Is $x \cdot y=z$ ?
The obvious thing to do is to multiply $x$ and $y$ and check whether the answer is $z$. It is important to notice that the time needed to do that is not a constant, since $x, y$ and $z$ can be very large. When a number occurs in the input, its length is the number of digits needed to write it down. For example, 490 has length 3 . The algorithm that you learned in elementary school multiplies an $i$-digit number by a $j$-digit number in time $O(i j)$. The length of the input is the total number of symbols that it contains. Clearly, if $x$ has length $i$ and $y$ has length $j$ and the total length of the input if $n$, then $i<n$ and $j<n$, and it is possible to check an integer product in time $O\left(n^{2}\right)$. That is polynomial time.

### 14.2.4 Example: Is $x$ a Prime Number?

The Primality Problem is the following decision problem.
Input. A positive integer $x$.
Question. Is $x$ prime?
Here is an algorithm that solves the Primality Problem.

```
" \(\{\) prime \((x)\) :
    \(i=2\)
    while \(i<x\)
            if \(n \bmod i==0\)
            return 0
            \(i=i+1\)
        return 1
\}"
```

How much time does that algorithm take? It goes around the loop $x-2$ times. If $x$ is $n$ digits long, then $x$ is in the rough vicinity of $10^{n}$. (Assuming
no leading 0 s, $10^{n-1} \leq x<10^{n}$.) The division algorithm that you learned in elementary school divides an $m$-digit number by an $n$-digit number in time $O(m n)$. Puting that all together, we find that our algorithm for testing whether an integer is prime takes time $O\left(n^{2} 10^{n}\right)$. But function $f(n)=10^{n}$ grows faster than any polynomial. That is, for every $k$,

$$
\lim _{n \rightarrow \infty} \frac{10^{n}}{n^{k}}=\infty
$$

Our primality-testing algorithm is not a polynomial-time algorithm.
What does that tell us about whether the Primality Problem is in P? Absolutely nothing! You can write a bad algorithm for any computable problem. The issue is not whether there is a bad algorithm to solve the Primality Problem, but whether there is a polynomial-time algorithm for it.
As it turns out, the primality problem is in P . It was long conjectured to be in P , and was shown to be in P in 2003.

### 14.3 The Validity Problem for Propositional Logic

Chapter 1 defines the notion of a valid propositional formula. The Validity Problem for Propositional Logic (or simply the Validity Problem) is the following decision problem.

Input. A propositional formula $\phi$.
Question. Is $\phi$ valid?
You already know an algorithm that solves that problem: truth tables. Suppose that $\phi$ has $v$ variables. Then a truth table for $\phi$ has $2^{v}$ rows, and determining validity takes time at least $2^{v}$.
Determining whether an algorithm runs in polynomial time requires determining the algorithm's running time in terms of the length $n$ of the input. The number of variables $v$ is surely shorter than the total length of $\phi$, but how close to $n$ can $v$ be? As long as we allow long variable names (such as $x_{1}, x_{2}, \ldots$ ), it is easy to write an interesting propositional formula of length $n$ with at least $\sqrt{n}$ variables. A little calculus shows that

$$
\lim _{n \rightarrow \infty} \frac{2^{\sqrt{n}}}{n^{k}}=\infty
$$

for every $k$, so the truth table algorithm does not run in polynomial time.
What does that have to say about whether the Validity Problem is in P? Absolutely nothing! Nobody knows a polynomial-time algorithm for the Validity Problem, but that lack of knowledge also is not convincing evidence that the Validity Problem is not in P .

Beginning with the next section, we begin to address the question of whether the Validity Problem is in P.

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