# Complexity Theory 

Lecture 3

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http://www.cl.cam.ac.uk/teaching/2122/Complexity

## Complexity Classes

A complexity class is a collection of languages determined by three things:

- A model of computation (such as a deterministic Turing machine, or a nondeterministic TM, or a parallel Random Access Machine).
- A resource (such as time, space or number of processors).
- A set of bounds. This is a set of functions that are used to bound the amount of resource we can use.


## Polynomial Bounds

By making the bounds broad enough, we can make our definitions fairly independent of the model of computation.

The collection of languages recognised in polynomial time is the same whether we consider Turing machines, register machines, or any other deterministic model of computation.

The collection of languages recognised in linear time, on the other hand, is different on a one-tape and a two-tape Turing machine.

We can say that being recognisable in polynomial time is a property of the language, while being recognisable in linear time is sensitive to the model of computation.

## Polynomial Time

$$
\mathrm{P}=\bigcup_{k=1}^{\infty} \operatorname{TIME}\left(n^{k}\right)
$$

The class of languages decidable in polynomial time.
The complexity class P plays an important role in our theory.

- It is robust, as explained.
- It serves as our formal definition of what is feasibly computable

One could argue whether an algorithm running in time $\theta\left(n^{100}\right)$ is feasible, but it will eventually run faster than one that takes time $\theta\left(2^{n}\right)$.
Making the distinction between polynomial and exponential results in a useful and elegant theory.

## Example: Reachability

The Reachability decision problem is, given a directed graph $G=(V, E)$ and two nodes $a, b \in V$, to determine whether there is a path from $a$ to $b$ in $G$.

A simple search algorithm as follows solves it:

1. mark node a, leaving other nodes unmarked, and initialise set $S$ to \{a\};
2. while $S$ is not empty, choose node $i$ in $S$ : remove $i$ from $S$ and for all $j$ such that there is an edge $(i, j)$ and $j$ is unmarked, mark $j$ and add $j$ to $S$;
3. if $b$ is marked, accept else reject.

## Analysis

This algorithm requires $O\left(n^{2}\right)$ time and $O(n)$ space.
The description of the algorithm would have to be refined for an implementation on a Turing machine, but it is easy enough to show that:

$$
\text { Reachability } \in P
$$

To formally define Reachability as a language, we would have to also choose a way of representing the input ( $V, E, a, b)$ as a string.

## Example: Euclid's Algorithm

Consider the decision problem (or language) RelPrime defined by:

$$
\{(x, y) \mid \operatorname{gcd}(x, y)=1\}
$$

The standard algorithm for solving it is due to Euclid:

1. Input $(x, y)$.
2. Repeat until $y=0: x \leftarrow x \bmod y$; Swap $x$ and $y$
3. If $x=1$ then accept else reject.

## Analysis

The number of repetitions at step 2 of the algorithm is at most $O(\log x)$.

This implies that RelPrime is in P .
If the algorithm took $\theta(x)$ steps to terminate, it would not be a polynomial time algorithm, as $x$ is not polynomial in the length of the input.

## Primality

Consider the decision problem (or language) Prime defined by:

$$
\{x \mid x \text { is prime }\}
$$

The obvious algorithm:
For all $y$ with $1<y \leq \sqrt{x}$ check whether $y \mid x$.
requires $\Omega(\sqrt{x})$ steps and is therefore not polynomial in the length of the input.

Is Prime $\in P$ ?

## Boolean Expressions

Boolean expressions are built up from an infinite set of variables

$$
X=\left\{x_{1}, x_{2}, \ldots\right\}
$$

and the two constants true and false by the rules:

- a constant or variable by itself is an expression;
- if $\phi$ is a Boolean expression, then so is $(\neg \phi)$;
- if $\phi$ and $\psi$ are both Boolean expressions, then so are $(\phi \wedge \psi)$ and $(\phi \vee \psi)$.


## Evaluation

If an expression contains no variables, then it can be evaluated to either true or false.

Otherwise, it can be evaluated, given a truth assignment to its variables.

## Examples:

```
(true \veefalse) }\wedge(\negfalse
( }\mp@subsup{x}{1}{}\vee\mathrm{ false ) }\wedge((\neg\mp@subsup{x}{1}{})\vee\mp@subsup{x}{2}{}
( }\mp@subsup{x}{1}{}\veef\mathrm{ false )}\wedge(\neg\mp@subsup{x}{1}{}
(x, \vee (\neg\mp@subsup{x}{1}{}))\wedge true
```


## Boolean Evaluation

There is a deterministic Turing machine, which given a Boolean expression without variables of length $n$ will determine, in time $O\left(n^{2}\right)$ whether the expression evaluates to true.

The algorithm works by scanning the input, rewriting formulas according to the following rules:

## Rules

- (true $\vee \phi) \Rightarrow$ true
- $(\phi \vee$ true $) \Rightarrow$ true
- (false $\vee \phi) \Rightarrow \phi$
- $(\phi \vee \mathrm{false}) \Rightarrow \phi$
- (false $\wedge \phi) \Rightarrow$ false
- $(\phi \wedge$ false $) \Rightarrow$ false
- (true $\wedge \phi) \Rightarrow \phi$
- $(\phi \wedge$ true $) \Rightarrow \phi$
- ( $\neg$ true $) \Rightarrow$ false
- $(\neg$ false $) \Rightarrow$ true


## Analysis

Each scan of the input ( $O(n)$ steps) must find at least one subexpression matching one of the rule patterns.

Applying a rule always eliminates at least one symbol from the formula. Thus, there are at most $O(n)$ scans required.

The algorithm works in $O\left(n^{2}\right)$ steps.

## Satisfiability

For Boolean expressions $\phi$ that contain variables, we can ask Is there an assignment of truth values to the variables which would make the formula evaluate to true?

The set of Boolean expressions for which this is true is the language SAT of satisfiable expressions.

This can be decided by a deterministic Turing machine in time $O\left(n^{2} 2^{n}\right)$. An expression of length $n$ can contain at most $n$ variables.
For each of the $2^{n}$ possible truth assignments to these variables, we check whether it results in a Boolean expression that evaluates to true.

Is $S A T \in P ?$

## Composites

Consider the decision problem (or language) Composite defined by:

$$
\{x \mid x \text { is not prime }\}
$$

This is the complement of the language Prime.
Is Composite $\in \mathrm{P}$ ?
Clearly, the answer is yes if, and only if, Prime $\in P$.

## Hamiltonian Graphs

Given a graph $G=(V, E)$, a Hamiltonian cycle in $G$ is a path in the graph, starting and ending at the same node, such that every node in $V$ appears on the cycle exactly once.

A graph is called Hamiltonian if it contains a Hamiltonian cycle.
The language HAM is the set of encodings of Hamiltonian graphs.
Is $H A M \in P$ ?

## Examples



The first of these graphs is not Hamiltonian, but the second one is.

## Graph Isomorphism

Given two graphs $G_{1}=\left(V_{1}, E_{1}\right)$ and $G_{2}=\left(V_{2}, E_{2}\right)$, is there a bijection

$$
\iota: V_{1} \rightarrow V_{2}
$$

such that for every $u, v \in V_{1}$,

$$
(u, v) \in E_{1} \quad \text { if, and only if, } \quad(\iota(u), \iota(v)) \in E_{2} .
$$

Is Graph Isomorphism $\in P$ ?

## Polynomial Verification

The problems Composite, SAT, HAM and Graph Isomorphism have something in common.

In each case, there is a search space of possible solutions. the numbers less than $x$; truth assignments to the variables of $\phi$; lists of the vertices of $G$; a bijection between $V_{1}$ and $V_{2}$.

The size of the search is exponential in the length of the input.
Given a potential solution in the search space, it is easy to check whether or not it is a solution.

