



# COMP36111: Advanced Algorithms I

## Lecture 6: Problems, algorithms and complexity measures

Ian Pratt-Hartmann

Room KB2.38: email: [ipratt@cs.man.ac.uk](mailto:ipratt@cs.man.ac.uk)

2016–17



- Reading for this lecture:
  - Sipser Ch. 3 (TMs)
  - Sipser Ch. 4 (Halting problem)
  - Sipser Ch. 7.1 (Time complexity)
  - Sipser Ch. 9.1 (Hierarchy theorem)



# Outline

## Basic notions

Turing machines

Computability and recognition

An undecidable problem

## Time and space

Basic notions

Classes of functions

## Complexity of problems

Basic definitions

Some more classes of functions

## Complexity classes

A cornucopia of complexity classes

A simple complexity-theoretic result

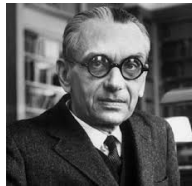
Complement classes



Alan Turing

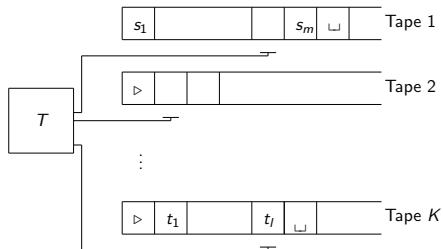


Alonzo Church



Kurt Gödel

- From the point of view of the rest of this course, an *algorithm* is a *multi-tape Turing Machine*:



- We think of Tape 1 as the *input*, Tape  $K$  as the *output*, and Tapes  $2-(K-1)$  as *work tapes*.



- Formally, a *Turing Machine* is a quintuple

$$M = \langle K, \Sigma, Q, q_0, T \rangle,$$

where

- $K \geq 2$  (number of *tapes*)
- $\Sigma$  is a non-empty, finite set (*alphabet*)
- $Q$  is a non-empty, finite set (set of *states*)
- $q_0 \in Q$  (*initial state*)
- $T$  is a set of *transitions* (for  $K$ ,  $\Sigma$  and  $Q$ )—see below.
- A *symbol* is an element of  $\Sigma \cup \{\sqcup, \triangleright\}$ 
  - We pronounce  $\sqcup$  as *blank* and  $\triangleright$  as *start*.



- A *transition* (for  $K$ ,  $\Sigma$  and  $Q$ ) is a quintuple

$$\langle p, \bar{s}, q, \bar{t}, \bar{d} \rangle$$

where

- $p \in Q$  and  $q \in Q$
- $\bar{s}$  and  $\bar{t}$  are  $K$ -tuples of symbols
- $\bar{d}$  is a  $K$ -tuple whose elements from  $\{\text{left}, \text{right}, \text{stay}\}$
- It has the informal meaning:

*If you are in state  $p$ , and the symbols written on the squares currently being scanned on the  $K$  tapes are given by  $\bar{s}$ , then set the new state to be  $q$ , write the symbols of  $\bar{t}$  on the  $K$  tapes and move the heads as directed by  $\bar{d}$ .*



- We insist that
  - the tape never moves left past  $\triangleright$ ,
  - Tape 1 is read-only (input) and Tape  $K$  write-only (output) . . .
- $M$  is *deterministic* if, for every  $p$  and every  $\bar{s}$ , there is at most one transition  $\langle p, \bar{s}, q, \bar{t}, \bar{d} \rangle$ .





- A configuration of  $M$ : a  $K$ -tuple of strings

$$\triangleright, s_{k,1}, \dots, s_{k,i-1}, q, s_{k,i}, \dots, s_{k,n(k)}.$$

representing the situation in which the  $k$ th tape of  $M$  reads  $\triangleright, s_{k,1}, \dots, s_{k,n(k)}$ , the head is over square  $i$ , and the current state is  $q$  (same for all  $K$  strings).

- A run of machine  $M$  on input  $x$  is a sequence of configurations (finite or infinite) in which successive configurations conform to some transition in  $T$ .
- Just to be clear:  $M$  must make some transition if one is available.
- The run is *terminating* if it is finite. In this case, and for deterministic  $M$ , we write  $M \downarrow x$ ; otherwise  $M \uparrow x$ .

## Definition

Let  $M$  be a **deterministic** Turing machine over alphabet  $\Sigma$ , and let  $x \in \Sigma^*$ . If  $M \downarrow x$ , then the output tape of  $M$  will contain a definite string  $y \in \Sigma^*$ ; and we can define the partial function  $f_M : \Sigma^* \rightarrow \Sigma^*$  as follows.

$$f_M(x) = \begin{cases} y & \text{if } M \downarrow x \\ \text{undefined} & \text{otherwise} \end{cases}$$

We say that  $M$  *computes* the function  $f_M$ . A partial function  $f : \Sigma^* \rightarrow \Sigma^*$  is *computable* (if it is computed by some (deterministic) Turing machine).



- An important feature of Turing machines is that they they are **finite** objects of the form  $\langle K, \Sigma, Q, q_0, T \rangle$ .
- As such, they can be encoded (in some standard way) in any alphabet  $\Sigma$ . used as input to other Turing machines.
- In particular, there exists a **universal** Turing machine:

## Theorem

*Fix some alphabet  $\Sigma$ . There exists a Turing machine  $U$  with the following property. For any Turing Machine  $M$  with alphabet  $\Sigma$ , and any strings  $x, y \in \Sigma^*$ ,  $U$  has a terminating run on input  $(M; x)$  leaving  $y$  on the output tape if and only if  $M$  has a terminating run on input  $x$  leaving  $y$  on the output tape; moreover,  $U$  has a non-terminating run on input  $(M; x)$  if and only if  $M$  has a non-terminating run on input  $x$ .*



## Definition (Acceptance and recognition)

Let  $M$  be a Turing machine over alphabet  $\Sigma$ , and let  $x \in \Sigma^*$ .

Assume WLOG  $\Sigma$  contains the symbol Yes. We say  $M$  *accepts*  $x$  if  $M$  has a halting run on input  $x$ , with output Yes.

The set of strings accepted by  $M$  is called the language *recognized* by  $M$ .

## Theorem

If a language is recognized by some Turing machine  $M$ , then it is recognized by some *deterministic* Turing Machine  $M'$ .



- A language is *recognizable* if there is a (deterministic) Turing Machine which recognizes it.
- A language is *computable* if there is a deterministic Turing Machine  $M$  recognizing it, such that  $M$  always halts.
- Sometimes, we use the vocabulary of *problems* and *decidability*
  - language  $\iff$  problem
  - computable  $\iff$  decidable
- Sometimes, we use the following older terminology:
  - recognizable  $\iff$  recursively enumerable (r.e.)
  - computable  $\iff$  recursive

- It is easy to see from the above that a problem  $\mathcal{P} \subseteq \Sigma^*$  is decidable if and only if the total function

$$f_{\mathcal{P}} : \Sigma^* \rightarrow \{\text{Yes}, \text{No}\}$$

$$f_{\mathcal{P}}(x) = \begin{cases} \text{Yes} & \text{if } x \in \mathcal{P}; \\ \text{No} & \text{otherwise.} \end{cases}$$

is computable.

- Typically, we present problems in the form

### PROBLEM NAME

Given: a string  $x$  (coding some object we are interested in);

Return: Yes if  $x$  has some property  $\mathcal{P}$ , and No otherwise.



## Definition

The *Halting problem* is the following problem:

### HALTING

Given: a Turing Machine description,  $M$  and a string,  $x$   
in the alphabet of  $M$ ,

Return: Yes if  $M \downarrow x$ , and No otherwise.

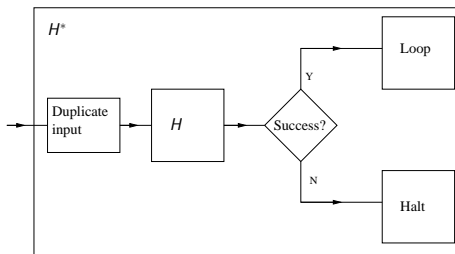
## Theorem (Turing, 1936)

*The Halting problem is not decidable.*



## Proof.

Suppose such a Turing machine exists, say  $H$ . Consider the Turing machine, say  $H^*$ :



What happens if  $H^*$  is given as input  $H^*$ —i.e. a description of itself as input? The embedded  $H$  receives input  $(H^*, H^*)$ . Hence:

$$H^* \downarrow H^* \Rightarrow H^* \uparrow H^*$$

$$H^* \uparrow H^* \Rightarrow H^* \downarrow H^*$$





# Outline

## Basic notions

Turing machines

Computability and recognition

An undecidable problem

## Time and space

Basic notions

Classes of functions

## Complexity of problems

Basic definitions

Some more classes of functions

## Complexity classes

A cornucopia of complexity classes

A simple complexity-theoretic result

Complement classes



- In these lectures, we are interested in the *resources* required by Turing machines to recognize languages (= decide problems)
- There are two main types of machine:
  - deterministic
  - non-deterministic
- There are two main types of resource:
  - time
  - space
- Warning: this four-way classification is not meant to exhaust the possible types of complexity analysis!
- There now follow some rather dreary definitions . . .



## Definition

Let  $M$  be a Turing machine with alphabet  $\Sigma$ , and let  $g : \mathbb{N} \rightarrow \mathbb{N}$ . We say  $M$  *runs in time*  $g$  if, for all but finitely many strings  $x \in \Sigma^*$ , any run of  $M$  on input  $x$  halts within at most  $g(|x|)$  steps. Similarly,  $M$  *runs in space*  $g$  if, for all but finitely many strings  $x \in \Sigma^*$ , any run of  $M$  on input  $x$  uses at most  $g(|x|)$  squares on any of its work-tapes.



- Thus, it makes sense to say, for example, that a given Turing machine runs in time (or space)  $n^2$ , or  $3n^3 - 13n + 42$ .
- The problem is: this sort of complexity-measure is not very *robust*:
- Suppose  $M$  is a Turing machine running in time/space  $g$ , and let  $c > 0$ .
- Provided  $g$  is moderately fast-growing, there exists a TM  $M'$ , running in time  $c \cdot g$ , halting exactly when  $M$  does, and writing the same results on its output tape. (“Linear speed-up”.)

## Definition

Let  $M$  be a Turing machine, and  $G$  a set of functions from  $\mathbb{N}$  to  $\mathbb{N}$ . We say that  $M$  runs in time  $G$  if, for some  $g \in G$ ,  $M$  runs in time  $g$ . Similarly, we say that  $M$  runs in space  $G$  if, for some  $g \in G$ ,  $M$  runs in space  $g$ .

## Definition

Let  $g : \mathbb{N}^k \rightarrow \mathbb{N}$  be a function. Denote by  $O(g)$  the set of functions

$$O(g) = \{g' : \mathbb{N}^k \rightarrow \mathbb{N} \mid \text{there exist } c \in \mathbb{N}, n'_1, \dots, n'_k \in \mathbb{N} \text{ s.t.}$$

for all  $n_1 > n'_1 \dots$  for all  $n_k > n'_k, g'(n_1, \dots, n_k) \leq cg(n_1, \dots, n_k)\}$ .



- Thus, it makes sense to say, for example, that a given Turing machine runs in time (or space)  $O(n^2)$ , or  $O(n^3)$ , or  $O(2^n)$ .
- This reflects the fact that additive and multiplicative constants are irrelevant: we can always speed up Turing machines arbitrarily.
- It also enables us to abstract away from irritating details of TMs
- Thus, we can say, for example, that the algorithm for MATCHING described in Lecture 0 runs in  $O(n^3)$  time, without having to worry about how, exactly, all the data-structures are encoded.

# Outline

## Basic notions

Turing machines

Computability and recognition

An undecidable problem

## Time and space

Basic notions

Classes of functions

## Complexity of problems

Basic definitions

Some more classes of functions

## Complexity classes

A cornucopia of complexity classes

A simple complexity-theoretic result

Complement classes

## Definition

Let  $L$  be a language over some alphabet, and let  $G$  be a set of functions from  $\mathbb{N}$  to  $\mathbb{N}$ . We say that  $L$  is in  $\text{TIME}(G)$  (or  $\text{SPACE}(G)$ ) if there exists a deterministic Turing machine  $M$  recognizing  $L$ , such that  $M$  runs in time (respectively, space)  $G$ .

If  $G = \{g\}$ , we write  $\text{TIME}(g)$  instead of  $\text{TIME}(\{g\})$ ; similarly for  $\text{SPACE}$ .



## Definition

Let  $L$  be a language over some alphabet, and let  $G$  be a set of functions from  $\mathbb{N}$  to  $\mathbb{N}$ . We say that  $L$  is in  $\text{NTIME}(G)$  (or  $\text{NSPACE}(G)$ ) if there exists a Turing machine  $M$  recognizing  $L$ , such that  $M$  runs in time (respectively, space)  $G$ .

If  $G = \{g\}$ , we write  $\text{NTIME}(g)$  instead of  $\text{NTIME}(\{g\})$ .



- Thus, we showed in Lectures 2 and 3 that
  - REACHABILITY is in  $\text{TIME}(n)$ ;
  - MATCHING is in  $\text{TIME}(n^3)$ .
- To think about: why did we not say:
  - REACHABILITY is in  $\text{TIME}(O(n))$ ;
  - MATCHING is in  $\text{TIME}(O(n^3))$ ?



- When talking about the complexity of problems, we typically consider *larger* classes of functions than those of the form  $O(f)$ .
- Here are some such classes:

$$P = \{n^c \mid c > 0\}$$

$$E = \{2^{n^c} \mid c > 0\}$$

$$E_2 = \{2^{2^{n^c}} \mid c > 0\}$$

$$E_k = \{2^{\underbrace{2^{\dots^2}}_{n^c \text{ times}}} \mid c > 0\}$$

- A function  $g : \mathbb{N} \rightarrow \mathbb{N}$  which is in  $E_k$  for some  $k$  is said to be *elementary*.
- It is easy to define a computable function which is non-elementary:

$$f(n) = 2^{\underbrace{2^{\dots^2}}_n \text{ times}} .$$

It grows rapidly.



# Outline

## Basic notions

Turing machines

Computability and recognition

An undecidable problem

## Time and space

Basic notions

Classes of functions

## Complexity of problems

Basic definitions

Some more classes of functions

## Complexity classes

A cornucopia of complexity classes

A simple complexity-theoretic result

Complement classes



- Thus we have the following deterministic complexity classes

$$\begin{array}{ll}
 \text{PTIME} & = \text{TIME}(P) \\
 \text{EXPTIME} & = \text{TIME}(E) \\
 k\text{-EXPTIME} & = \text{TIME}(E_k)
 \end{array}
 \qquad
 \begin{array}{ll}
 \text{LOGSPACE} & = \text{SPACE}(\log n) \\
 \text{PSPACE} & = \text{SPACE}(P) \\
 \text{EXPSpace} & = \text{SPACE}(E) \\
 k\text{-EXPSpace} & = \text{SPACE}(E_k).
 \end{array}$$



- And likewise their non-deterministic counterparts:

$$\text{NPTIME} = \text{NTIME}(P)$$

$$\text{NEXPTIME} = \text{NTIME}(E)$$

$$\text{N}k\text{-EXPTIME} = \text{NTIME}(E_k)$$

$$\text{NLOGSPACE} = \text{NSPACE}(\log n)$$

$$\text{NPSPACE} = \text{NSPACE}(P)$$

$$\text{NEXPSPACE} = \text{NSPACE}(E)$$

$$\text{N}k\text{-EXPSPACE} = \text{NSPACE}(E_k).$$



- Thus, we showed in earlier lectures that
  - DIRECTED GRAPH CYCLICITY
  - GRAPH CONNECTEDNESS
  - PERFECT MATCHING
  - STRING OCCURRENCE
  - LINEAR PROGRAMMING FEASIBILITY (well, we **stated** this)

are all in  $P_{TIME}$

- Exercise: try writing out definitions of the above problems in the standard form:

PROBLEM NAME

Given: ... ;

Return: ...

- We will encounter many examples of problems in other complexity classes in the coming lectures.

- Complexity classes fit inside one another obvious ways:

$$\text{PTIME} \subseteq \text{EXPTIME} \subseteq 2\text{-EXPTIME} \subseteq \dots$$

$$\text{LOGSPACE} \subseteq \text{PSPACE} \subseteq \text{EXPSpace} \subseteq 2\text{-EXPSpace} \subseteq \dots$$

$$\text{NPTIME} \subseteq \text{NEXPTIME} \subseteq 2\text{-NEXPTIME} \subseteq \dots$$

$$\text{NLOGSPACE} \subseteq \text{NPSpace} \subseteq \text{NEXPSpace} \subseteq 2\text{-NEXPSpace} \subseteq \dots$$

- Actually, the non-deterministic classes can be interleaved:

$$\text{PTIME} \subseteq \text{NPTIME} \subseteq \text{EXPTIME} \subseteq \text{NEXPTIME} \subseteq \dots$$

- There are some non-obvious inclusions, which we will discuss later.
- Most obvious problem: is any of these inclusions strict?





## Definition

The  $f$ -bounded *Halting problem* is the following problem:

### HALTING<sub>f</sub>

Given: a Turing Machine description,  $M$  and a string,  $x$   
in the alphabet of  $M$ ,

Return: Yes if  $M \downarrow x$  in time  $\leq f(|x|)$ , and No otherwise.

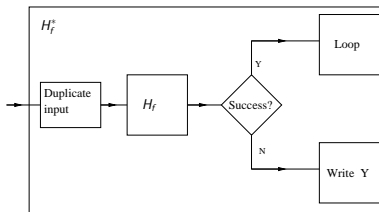
## Theorem

$\text{HALTING}_f \notin \text{TIME}(f(\lfloor n/2 \rfloor))$ .

## Proof.

Suppose  $\text{HALTING}_f$  is recognized by a Turing machine  $H_f$ , guaranteed to terminate in time  $f(\lfloor n/2 \rfloor)$ .

Consider the Turing machine, say  $H_f^*$ :



What happens if  $H_f^*$  is given as input  $H_f^*$ —i.e. a description of itself as input? The embedded  $H$  receives input  $(H_f^*, H_f^*)$ , and terminates (if at all) in time  $f(|H_f^*|)$ . Hence:

$$H_f^* \downarrow H_f^* \Rightarrow H_f^* \uparrow H_f^*$$

$$H_f^* \uparrow H_f^* \Rightarrow H_f^* \downarrow H_f^*$$



- However, if  $f(n)$  is a ‘proper’ complexity function, we *can* decide the problem  $H_f$  in time  $(f(n))^3$  using a version,  $U_f$  of the universal Turing machine,  $U$ .
- The machine  $U_f$  works as follows given input  $(M, x)$ .
  - writes  $f(|x|)$  symbols  $\star$  on an ‘alarm-clock’ (work)tape;
  - simulate the steps of  $M$  in the usual way, advancing a counter on the alarm clock tape by 1 for each step;
  - abandon the computation if the alarm clock rings, and just output No.
- This machine can be made to run in time  $O(f(n)^3)$ , and so can be sped-up to run in time  $f(n)^3$ .



Define  $f'(n) = f(\lfloor n/2 \rfloor)$ . Now,  $M_f$  decides  $\text{HALTING}_f$ , and runs in time  $f(n^3) = f'(2n+1)^3$ . On the other hand,  $\text{HALTING}_f$ , is not computable in time  $f(\lfloor n/2 \rfloor) = f'(n)$ . Moreover, if  $f'$  is 'proper', so is  $f$ . Hence:

### Theorem

For all 'proper' functions  $f$ ,  $\text{TIME}(f(n)) \subsetneq \text{TIME}((f(2n+1))^3)$ .

Using similar reasoning:

### Theorem

For all 'proper' functions  $f$ ,  $\text{SPACE}(f(n)) \subsetneq \text{SPACE}(f(n) \log f(n))$ .

This yields a very important corollary:

### Theorem

$\text{PTIME} \subsetneq \text{EXPTIME}$ .

### Proof.

Since any polynomial is dominated by  $2^n$ ,

$$\begin{aligned} \text{PTIME} &\subseteq \text{TIME}(2^n) \\ &\subsetneq \text{TIME}(2^{3(2n+1)}) \\ &\subseteq \text{EXPTIME}. \end{aligned}$$



Similarly

### Theorem

$\text{NPTIME} \subsetneq \text{NEXPTIME}$ , *and*  $\text{PSPACE} \subsetneq \text{EXPSPACE}$ .



- Notice the asymmetry involved in the notion of (non-deterministic) computation:

*$M$  recognizes  $L \subseteq \Sigma^*$  just in case, for each string  $x \in \Sigma^*$ ,  $x \in L$  if and only if there exists a terminating run of  $M$  on input  $x$ .*

- This asymmetry prompts us to define the *complement* classes as follows.

*If  $\mathcal{C}$  is a class of languages, then  $\text{Co-}\mathcal{C}$  is the class of languages  $L$  such that  $\Sigma^* \setminus L$  is in  $\mathcal{C}$ , where  $\Sigma$  is the alphabet of  $L$ .*



- Trivially,

$$\begin{aligned}\text{TIME}(G) &= \text{CO-TIME}(G) \\ \text{SPACE}(G) &= \text{CO-SPACE}(G).\end{aligned}$$

- For non-deterministic classes, some of these equations are not known to hold:

$$\text{NPTIME} \stackrel{?}{=} \text{CO-NPTIME}$$

- But there are some surprises to come ...