Monash University Faculty of Information Technology

FIT2014 Theory of Computation

Lecture 24 Polynomial time, and the class P

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COMMONWEALTH OF AUSTRALIA

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Overview

- ► Decidability: it's not enough
- ▶ Time complexity
- Polynomial time
- ► The class P
- Properties of P
- Examples of languages in P

Decidable languages

Decidable languages are "solvable", in principle, by a computer.

It doesn't matter (much) which definition of "computer" you use. The set of decidable languages is the same.

BUT what about the resources required?

Which languages (or problems) can be solved efficiently, in practice?

The most important resource: <u>time</u>.

For any Turing machine M, define:

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t_M(x) := time taken by M for input x := \# steps M takes until it halts.
```

Time complexity:

$$t_M(n) := \max \{ t_M(x) : |x| = n \}.$$

It's a function of n, which can be any positive integer.

For it to be defined, M must halt on all inputs, i.e., M must be a decider.

It's a worst case measure.

Examples:

- Turing machine to decide whether a string ends in b
 - moves to the right until reaches first blank symbol, then does one step to the left and accepts if the symbol there is b, otherwise rejects
 - ▶ time complexity? n+1
- ► Turing machine to decide whether a string is a palindrome
 - time complexity? $\approx \frac{1}{2}(n+1)(n+2)$
- Turing machine to decide whether a string is empty
 - ► time complexity? 1

Exercise:

Consider any regular language. What can you say about its time complexity?

See Lecture 18 on Turing Machines:

Finite Automaton \longrightarrow Turing machine

Depends on the type of Turing machine (or computer) used.

Some details that affect time taken:

- number of symbols used,
- whether tape is infinite in both directions (or just one),
- number of tapes,
- ▶ dimensionality of "tape" (1-D, 2-D, ...?),
- whether allowed movement directions include sitting still,
- **.**...

In FIT1045 or FIT1008, you may have met time complexity of algorithms and programs.

Those time complexities still assumed some theoretical computer (maybe implicitly), and can be sensitive to the assumptions made.

We will use time complexity to define "efficiently solvable".

But setting a specific threshold (e.g., $3n^5$) is too arbitrary, and the meaning of "efficiently solvable" is then either tied to a specific type of Turing machine (computer) or can change with time (as technology improves).

Example time complexities

input size	time						
n	n	n^2	n^3	n^4	• • •	2 ⁿ	10 ⁿ
10	10	100	1000	10000		1024	10^{10}
20	20	400	8000	160000		1048576	10^{20}
30	30	900	27000	810000		1073741824	10^{30}
40	40	1600	64000	2560000		1099511627776	10^{40}
:							
100	10 ²	10^{4}	10^{6}	10 ⁸		$pprox 10^{30}$	10^{100}
:							
1000	10^{3}	10^{6}	10^{9}	10^{12}		$pprox 10^{300}$	10^{1000}
:							
10000	10 ⁴	10 ⁸	10^{12}	10^{16}		$pprox 10^{3000}$	10^{10000}

If you	<i>n^c</i> time	exponential time		
increase input size by a fixed amount (i.e., $n \longrightarrow n + k$)	then time increases by an amount $O(n^{c-1})$	then time increases by a fixed factor		
increase input size by a fixed factor $k \dots$ (i.e., $n \longrightarrow kn$)	then time increases by fixed factor k^c	then time is raised to power k		
double your computer's speed	then you increase feasible input size by some fixed factor.	then you increase feasible input size by some fixed amount.		
need to handle inputs which are twice as large as those you can solve now	then you must wait for 2 <i>c</i> years before computers are fast enough. *	then your wait is proportional to your current input size. *		

^{*} assumes Moore's Law: processor speed doubles every two years.

Polynomial time

A Turing machine M has **polynomial time complexity** if its time complexity is $O(n^k)$, for some fixed k.

$$t_{M}(n)=O(n^{k}).$$

This means that there is a constant c such that, for all sufficiently large n,

$$t_M(n) \leq c \cdot n^k$$
.

Symbolically:

$$\exists c \,\exists N \,\forall n : n > N \implies t_M(n) \leq c \cdot n^k$$
.

The power, k, is fixed.

It does not depend on the input.

But different polynomial time Turing machines often have different k.

The class P

A language is **polynomial time decidable** if it can be decided by a polynomial time Turing machine.

The class of all languages decidable in polynomial time is called P (which stands for \underline{P} olynomial time).

This is the simplest (and, historically, the first) formal notion of "efficiently solvable".

$\{ \text{strings that end in b} \}$	time complexity	=	n+1	=	O(n)	\checkmark
$\{palindromes\}$	time complexity	=	$\frac{1}{2}\cdot(n+1)(n+2)$	=	$O(n^2)$	✓
$\{\varepsilon\}$	time complexity	=	O(1)	=	$O(n^0)$	\checkmark
any regular language	time complexity	=				✓
any context-free language	time complexity	=				\checkmark

```
(two symbols, single one-way-infinite tape, moves one step left or right, ...)

But what if we used a different type of machine?
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has been defined using a particular type of Turing machine M

Or some other model of computation? (E.g., your laptop, a smartphone, CSIRAC, TaihuLight, LUMI, Fugaku, Frontier, . . .)

Would we get a different class P?

Suppose that

- ightharpoonup computer M_1 has time complexity t(n).
- computer M_2 can simulate machine M_1 . (E.g., M_2 is a UTM.)
- ▶ any computation that takes time t on M_1 takes time $\leq ct^k$ on M_2 . (polynomial slowdown)

Then:

if M_1 is polynomial time then M_2 takes polynomial time to simulate M_1 on input strings for M_1 .

Proof.

If M_1 takes polynomial time, then $t(n) = O(n^K)$ for some fixed K. In other words, there exist c_1 and K such that $t(n) \le c_1 n^K$ for sufficiently large n.

Time taken by M_2 to simulate M_1 on an input of size n is . . .

```
\leq c \cdot t(n)^k

\leq c (c_1 n^K)^k for sufficiently large n

\leq c c_1^k n^{Kk}

\leq c' n^{k'}, where c' = cc_1^k and k' = Kk (note, both constants)

= O(n^{k'}).
```

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It follows that, for such M_1 and M_2 :

If a language L can be decided in polynomial time using M_1 , then it can be decided in polynomial time using M_2 .

In fact, virtually any two computers can play the roles of M_1 and M_2 here.

This is because any "reasonable" computer can simulate any other computer with at most polynomial slowdown.

So the class P is independent of the particular model of computation used to define it.

This is reminiscent of the history of decidability: different paths can be taken to formulate the definition, using different models of computation, but it turns out that they all lead to the same class of decidable languages.

History of P : Alan Cobham (1965), Jack Edmonds (1965), Michael Rabin (1966), . . .

- the set of pairs of strings in lexicographic order
- the set of strings of matching parentheses
- ▶ the set of pairs of numbers that are coprime (a.k.a. relatively prime) [Euclidean algorithm]
- the set of square numbers
- the set of prime numbers
 (Agrawal, Kayal, Saxena, Annals of Mathematics, 2004)
- ▶ the set of invertible matrices [MAT1841/MAT2003]
- the set of trees
- the set of balanced binary trees

the set of connected graphs:

```
\{(G, s, t, k) : G \text{ has an } s-t \text{ path of length } \leq k\}
```

- ▶ the set of regular graphs (i.e., all vertices have the same degree)
- the set of 2-colourable graphs
 (graphs whose vertices can each be coloured Black or White,
 so that adjacent vertices receive different colours)
- ▶ the set of Eulerian graphs [MAT1830/FIT1045]
- ▶ the set of planar graphs (i.e., graphs which can be drawn in the plane so that no two edges cross, except possibly at their endpoints) [advanced]

2-SAT: the set of satisfiable Boolean expressions in Conjunctive Normal Form (CNF), with two literals per clause

The "2-" in the name indicates that every clause has exactly two literals.

E.g.:

$$(\neg x \vee \neg y) \wedge (x \vee \neg z) \wedge (y \vee z) \wedge (y \vee \neg y)$$

2-SAT (continued):

A **truth assignment** is an assignment of a truth value to each of the variables, i.e., a function $f : \{ \text{ variables } \} \rightarrow \{ \text{ True, False } \}$

E.g., f(x) = False, f(y) = False, f(z) = True is a truth assignment for the expression

$$(\neg x \vee \neg y) \wedge (x \vee \neg z) \wedge (y \vee z) \wedge (y \vee \neg y),$$

and it gives the expression the value

$$(\neg F \lor \neg F) \land (F \lor \neg T) \land (F \lor T) \land (F \lor \neg F) = \dots = False$$

Do all truth assignments make this expression False?

2-SAT (continued):

An expression is **satisfiable** if it has a truth assignment which makes the expression True.

E.g., the expression

$$(\neg x \vee \neg y) \wedge (x \vee \neg z) \wedge (y \vee z) \wedge (y \vee \neg y)$$

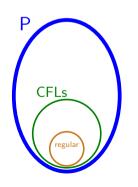
is satisfiable, since the truth assignment

$$g(x)$$
 = True
 $g(y)$ = False
 $g(z)$ = True

makes the expression True.

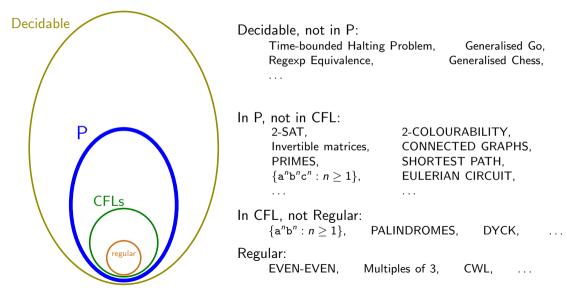
Challenge: show that 2-SAT is in P.

P and other language classes



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In P, not in CFL:
     2-SAT,
                            2-COLOURABILITY,
      Invertible matrices.
                            CONNECTED GRAPHS.
     PRIMES.
                            SHORTEST PATH.
      \{a^nb^nc^n: n > 1\},\
                            EULERIAN CIRCUIT,
      . . .
In CFL, not Regular:
     \{a^nb^n: n \geq 1\},
                      PALINDROMES.
                                          DYCK.
Regular:
     EVEN-EVEN.
                     Multiples of 3,
                                      CWL.
```

P and other language classes



Revision

- Time complexity
- Practical decidability
- Definition of P
- Relationship between P and model of computation (and argument using simulation and polynomial slowdown)
- Languages in P

Reading:

- ► Sipser, sections 7.1–7.2
- M. R. Garey and D. S. Johnson, Computers and Intractability: A Guide to the Theory of NP-Completeness, W. H. Freeman & Co., San Francisco, 1979, Chapter 1 and especially the first two sections of Chapter 2: §2.1, §2.2.