

# Theory of Algorithms

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# Topics

Functional programming

Turing machines

Computational complexity

NP complete problems

Permutations

Timing Algorithms

Functional Programming

Turing Machines

Complexity Classes

# Functional programming

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# Turing machines

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## Computational complexity

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## How many numbers can we represent with $k$ bits?

- Consider the case of four bits.
- Imagine four placeholders: 

?	?	?	?
---	---	---	---
- Each placeholder can contain either 0 or 1.
- There are  $2 \times 2 \times 2 \times 2 = 2^4 = 16$  different numbers.
- Add another bit, how many numbers is it now?
- It's  $2 \times 2 \times 2 \times 2 \times 2 = 2^5 = 32$ .
- Generally  $k$  bits can represent  $2^k$  numbers.
- This grows **exponentially** relative to  $k$ .

## NP complete problems

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# Permutations

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# Permutations

**Permutations** are rearrangements of ordered collections of items.

**Example:** “abcd” is a rearrangement of “bacd”.

**Think** of having a four boxes, where we have to place one of the items in each box.

**What** are all the different ways of doing this?

**What** are all the different ways of associating items with boxes?

We can consider permutations in abstraction. For instance, if we have four items to rearrange, we can label the first item 1, the second 2, and so on. Then we can represent the various permutations, in terms of the numbers associated with the items. The permutations “abcd” and “bacd” could be represented by:

$$\begin{pmatrix} 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 1 & 3 & 4 \end{pmatrix}$$

In this way we can consider permutations in their own right.

# Counting permutations

**With four items, how many distinct permutations are there?**

Consider having four placeholders where we can place the items:



When we place an item in the first box, we have four choices, then we are left with only three choices for the second, two choices for the third, and one choice (i.e. not a choice at all) for the last.

So for there are  $4 \times 3 \times 2 \times 1 = 4!$  choices in total.

## Repetition

What happens when we consider two of our items to be the same? For instance, what if we are looking for all distinct rearrangements of “aacd” as opposed to “abcd”? In that case we need to account for the rearrangements of those items by dividing by the factorial of the number of times each item is repeated:  $\frac{4!}{2!}$ .

If more than one item is repeated a number of times we just keep dividing by the factorials of the numbers of repetitions. The distinct number of rearrangements of “aaabbcd” is  $\frac{7!}{3!2!}$ .

## Exercise

Calculate the number of distinct rearrangements of the word “Mississippi”.

# Heap's algorithm

**Heap** published an algorithm in November 1963 for generating permutations.

**Published** in The Computer Journal.

**Read** the article in the link below – it's an easy read.

**Pairs** of items are interchanged to generate each new permutation.

**Induction** is used to show the algorithm works.

## Heap's algorithm description

- Suppose we know how to permute  $(n - 1)$  items.
- That is, we know all of the different ways of slotting  $(n - 1)$  items in  $(n - 1)$  boxes.
- Let's add another,  $n^{th}$  item, and another box, an  $n^{th}$  box.
- First, place the  $n^{th}$  item in the  $n^{th}$  box.
- Permute all the the other items, which you know how to do.
- Then swap another item with the  $n^{th}$  item.
- Again permute the items in boxes 1 to  $(n - 1)$ .
- Swap the item in the  $n^{th}$  box with another, different item.
- Repeat until all items have been in box  $n$ .

# Steinhaus-Johnson-Trotter algorithm

**Johnson** published another algorithm in 1963 for generating permutations.

**Attributed** to three people: Steinhaus, Johnson and Trotter.

**See** the article in the link below – it's a trickier read.

**Pairwise** – the algorithm can be done pair-wise, like Heap's.

**Induction** is used to show the algorithm works.

# Steinhaus-Johnson-Trotter algorithm description

- Start with two items, 1 and 2, and generate their list of permutations 12 and 21.
- Use the two-item list to generate the three item list in the following way:
  - Place 3 at the right of the first element in the two-item list.
  - Then move 3 one place to the left continuously until its on the left.
  - Then place 3 at the left of the next element in the two-item list.
  - move 3 one place to the right continuously until it reaches the right.
- Use the three item list to generate the four item list in the same way, and so on.



## Conundrum – Naive method

```
for permutation in permutations(letters):  
    checkIfWord(permutation)
```

## Conundrum – Quick method preparation

```
worddict = {}  
  
for word in dictionaryOfWords:  
    sortword = sorted(list(word))  
    hashword = hash(sortword)  
    allWords = worddict.get(hashword, set())  
    allWords.update({word})  
    worddict[hashword] = allWords
```

## Conundrum – Quick method checking

```
word = "conundrum"  
sortword = sorted(list(word))  
hashword = hash(sortword)  
  
worddict.get(hashword, None)
```

# Timing Algorithms

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## timeit command line

```
$ python -m timeit '"-".join(str(n) for n in range(100))'  
10000 loops, best of 3: 30.2 usec per loop  
$ python -m timeit '"-".join([str(n) for n in range(100)])'  
10000 loops, best of 3: 27.5 usec per loop  
$ python -m timeit '"-".join(map(str, range(100)))'  
10000 loops, best of 3: 23.2 usec per loop
```

```
import timeit

def test():
    """Stupid test function"""
    L = [i for i in range(100)]

if __name__ == '__main__':
    import timeit
    print(timeit.timeit("test()",
                        setup="from __main__ import test"))
```

# Functional Programming

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- John McCarthy while at MIT – late 1950s.
- Created Lisp.
- Lisp generally considered first functional programming language (not really though).
- Lots of dialects exist today, such as Scheme and Common Lisp.



## Wasteful for loops

*# How do you parallelise this?*

```
inds = list(range(10))  
total = 0  
for i in inds:  
    total = total + (i * 7)
```

```
def byseven(i):  
    return i * 7
```

```
inds = list(range(10))  
sevens = map(byseven, inds)  
total = sum(sevens)
```

# State

**Imperative** programming is a programming paradigm where statements are used to change the *state*.

**State** is the name given to the current data/values related to an executing process, including internal stuff like the call stack.

**Processes** begin with an initial state and (possibly) have (a number of) halt states.

**Statements** change the state.

**Functions** in imperative programming languages might return different values for the same input at different times, because of the state.

**Functional** programming languages (try to) not depend on state.

**Functions** are said to have side effects if they modify the state (on top of returning a value).

**Static and global** variables are often good examples of side effects in action.

**Functional** programming tries to avoid side effects.

**It's tricky** to avoid them – such as when we need user input.

## Basic operators

```
> (+ 3 4)
```

```
7
```

```
> (* 3 2)
```

```
6
```

```
> (- 5 3)
```

```
2
```

```
> (/ 6 3)
```

```
2
```

## More arguments

```
> (+ 3 4 5)
```

```
(+ 3 4 5)
```

```
> (- 3 4 5)
```

```
-6
```

```
> (* 2 3 4)
```

```
24
```

```
> (/ 6 3 3)
```

```
2/3
```

```
> (/ 6 3 3 3)
```

```
2/9
```

# Functions and values

*; Define a value called foo with value 3.*

```
>(define foo 3)
```

*; Define a function f.*

```
>(define (f x)
  (+ (* 3 x) 12))
```

```
>(define (g x)
  (* 3 (+ x 4)))
```

```
>(g 2)
18
```

# Conditionals

```
> (if (< 1 2) '(y e s) '(n o))  
(y e s)
```

```
>(define (abs x)  
  (if (< x 0)  
      (- x)  
      x))
```

```
>(list 1 2 3)  
(1 2 3)
```

```
>(list 'a 'b 'c)  
(a b c)
```

```
> (length (list 1 2 3))  
3
```



## car and cdr

```
> (car (list 1 2 3))
```

```
1
```

```
> (cdr (list 1 2 3))
```

```
(2 3)
```

```
> (define l (list 1 2 3))
```

```
> (car l)
```

```
1
```

```
> (cdr l)
```

```
(2 3)
```

```
> (car (cdr l))
```

```
2
```

```
> (cadr l)
```

```
2
```

# Recursion

```
> (define (sum lv)
  (if (null? lv)
      0
      (+ (car lv) (sum (cdr lv)))))
> (sum (list 1 2 3))
6
> (define (derange n)
  (if (= 0 n)
      '()
      (cons n (derange (- n 1)))))
> (derange 12)
(12 11 10 9 8 7 6 5 4 3 2 1)
```

## Looping recursively

```
> (let loop ((i 5))  
  (print "i is " i ".\n")  
  (if (> i 0) (loop (- i 1))))  
i is 5.  
i is 4.  
i is 3.  
i is 2.  
i is 1.  
i is 0.
```

```
> (define (swap3-1-2 x)
  (list (cadr x) (car x) (caddr x)))
```

```
> (swap3-1-2 (list 1 2 3))
(2 1 3)
```

```
> (define four-over-two (list 4 '/ 2))
```

```
> four-over-two
(4 / 2)
```

```
> (eval (swap3-1-2 four-over-two))
```

## More on functions

*; Printing stuff to terminal.*

```
> (print "Ay" "-" "yo.\n")
```

*; Proper way to define a function.*

```
> (define foo (lambda (bar) (print "Bar is " bar ".\n")))
```

*; Shorthand.*

```
> (define (foo bar) (print "Bar is " bar ".\n"))
```

*; Local variables.*

```
> (define (foo bar) (let ((thing "Bar"))
```

```
  (print thing " is " bar ".\n")))
```

```
> (foo "open")
```

```
Bar is open.
```

## Function example

```
> (define l
  (let
    ((d 4) (e 5))
    (lambda (a b c) (list a b c d e))
  )
)
> (1 1 2 3)
(1 2 3 4 5)
```

```
> (cons 1 '())  
(1)  
> (cons 1 (cons 2 null))  
(1 2)  
> (cons 1 (cons 2 (cons 3 null)))  
(1 2 3)  
> (define mylist (cons 1 (cons 2 (cons 3 null))))  
> mylist  
(1 2 3)  
> (car mylist)  
1  
> (cdr mylist)  
(2 3)
```

## More on lists

```
> (list "a" "b" "c")  
("a" "b" "c")  
> (list a b c)  
reference to undefined identifier: a  
> (list 'a 'b 'c)  
(a b c)  
  
> (equal?  
(list 1 2 3)  
(cons 1 (cons 2 (cons 3 '()))))  
#t
```



# Quoting

```
> (list a b c)
*** ERROR IN (console)@1.7--Unbound variable: a
> (quote (a b c))
(a b c)
> (quote a b c)
*** ERROR IN (console)@2.1--Ill-formed special form: quote
> '(a b c)
(a b c)
> (define forty-two '(* 6 9))
> forty-two
(* 6 9)
> (eval forty-two)
```

54

## Null list

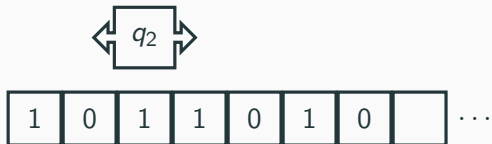
```
> ()  
missing procedure expression  
> (list)  
( )  
> '()  
( )  
> null  
( )  
> 'null  
null
```

```
> (define (container value)
  (lambda ()
    (string-append "This container contains " value ".")))
> (define apple (container "an apple"))
> (define pie (container "a pie"))
> (apple)
"This container contains an apple."
> (apple)
"This container contains an apple."
> (pie)
"This container contains a pie."
```

# Turing Machines

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# Visualisation



$Q$  Set of states (finite).

$G$  Tape alphabet (finite).

$B$  Blank symbol, element of  $G$ .

$S$  Input alphabet, subset of  $G \setminus \{B\}$ .

$\delta$  Transition function.

$q_0$  Initial state,  $\in Q$ .

$q_a$  Accept state,  $\in Q$ .

$q_r$  Reject state,  $\in Q$ .

$M$  Turing Machine:  $[Q, G, B, S, \delta, q_0, q_a, q_r]$ .

## State Table

State	Input	Write	Move	Next
0	B	B		Accept
	0	0	L	0
	1	1	L	1
1	B	B		Fail
	0	0	L	1
	1	1	L	0

$$\delta(q_i, g_n) \rightarrow (q_j, g_m, L/R)$$

## Turing's Second Example

### A slightly more difficult example

We can construct a machine to compute the sequence

00101101110111101111...

The machine is to be capable of five  $m$ -configurations, viz.  $o$ ,  $q$ ,  $p$ ,  $f$ ,  $b$  and of printing  $a$ ,  $x$ ,  $0$ ,  $1$ . The first three symbols on the tape will be  $aaO$ ; the other figures follow on alternate squares. On the intermediate squares we never print anything but  $x$ . These letters serve to keep the place for us and are erased when we have finished with them. We also arrange that in the sequence of figures on alternate squares there shall be no blanks.



## Turing's Second Example: Table

<i>Configuration</i>		<i>Behaviour</i>	
<i>m-config.</i>	<i>symbol</i>	<i>operations</i>	<i>final m-config.</i>
b		$P\emptyset, R, P\emptyset, R, P0, R, R, P0, L, L$	o
o	1	$R, Px, L, L, L$	o
	0		q
q	Any (0 or 1)	$R, R$	q
	None	$P1, L$	p
p	$x$	$E, R$	q
	$\emptyset$	$R$	f
	None	$L, L$	p
f	Any	$R, R$	f
	None	$P0, L, L$	o

## Turing's Second Example: JavaScript 1

```
// The contents of the tape.  
var tape = []  
s// The current position of the machine on the tape.  
var pos = 0  
// The current state;  
var state = b;
```

## Turing's Second Example: JavaScript 2

*// Writes a symbol to the current cell on the tape.*

```
function write(sym) {  
    tape[pos] = sym;  
}
```

*// Returns true iff the current cell contains sym.*

```
function read(sym) {  
    return sym == tape[pos] ? true : false;  
}
```

## Turing's Second Example: JavaScript 3

*// Erases the symbol in the current cell of the tape.*

```
function erase() {  
    delete tape[pos];  
}
```

*// Returns true iff the current cell is blank.*

*// Returns true iff the current cell is blank.*

```
function blank() {  
    return typeof(tape[pos])  
        == 'undefined' ? true : false;  
}
```

## Turing's Second Example: JavaScript 4

```
function b() {  
  write('e');  
  pos++;  
  write('e');  
  pos++;  
  write('0');  
  pos++;  
  pos++;  
  write('0');  
  pos--;  
  pos--;  
  state = o;  
}
```

**Alphabet** Finite set of symbols, denoted  $\Sigma$ .

**Word** Sequence of symbols,  $w$  from  $\Sigma$ .

**Language** Set of words, denoted  $L$ .

**Length** Of a word, denoted  $|w|$ .

**Empty word** Unique word of length 0, denoted  $\lambda$ .

**Words**  $w_1 w_2$  is the concatenation of words  $w_1$  and  $w_2$ .

**Languages**  $L_1 L_2$  is the language resulting from the concatenation of all words in  $L_1$  and all words in  $L_2$ , in that order.

**Powers**  $L^0 = \{\lambda\}$ ,  $L^1 = L$  and  $L^{n+1} = L^n L$  for all  $n > 1$ .

## Kleene Star

$$L^* = \bigcup_{i=0}^{\infty} L^i$$

Note that treating the alphabet  $\Sigma$  as a language in itself, we get that  $\Sigma^*$  is the set of all words over  $\Sigma$ .

## Example

$$\Sigma \{0, 1\}$$

$$L \{00, 01, 10, 11\}$$

$$w_1 \ 01$$

$$w_3 \ 11$$

$$w_1 w_3 \ 0111$$

$$\Sigma^* \{\lambda, 0, 1, 00, 01, 10, 11, 001, 010, \dots\}$$

$$L^* \{\lambda, 00, 01, 10, 11, 0000, 0001, \dots\}$$

$$L^+ \{00, 01, 10, 11, 0000, 0001, \dots\}$$



# Complexity Classes

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## Definition

An algorithm is said to be solvable in *polynomial time* if the number of steps required to complete the algorithm for a given input is  $O(n^k)$  for some nonnegative integer  $k$ , where  $n$  is the complexity of the input.

## **$P$ complexity class**

The  $P$  complexity class is the set of problems for which there exists, for each such problem, at least one algorithm to solve that problem in polynomial time.

# Non-deterministic polynomial time

## Definition

A problem is in the  $NP$  complexity class if it is solvable by a non-deterministic Turing Machine in polynomial time. A non-deterministic Turing Machine is one which may not have a unique action to take for some or all states and inputs.

### **$P$ is a subset of $NP$**

The  $P$  complexity class is a subset of  $P$  because all polynomial time solvable problems can be modelled using Nondeterministic Turing Machines.

# Decision problems

**Decision problems** are problems where the answer is 0 or 1.

**Restricting** ourselves to decision problems is convenient and fair.

$f : \{0, 1\}^n \rightarrow \{0, 1\}$  is useful notation for considering decision problems.

**Other problems** can be easily (polynomial time) adapted into decision problems.

## Definition

An problem is *NP*-hard if each problem in *NP* can be reduced to it in polynomial time.

### *NP*-hard problems are hard

*NP*-hard problems are at least as hard to solve as the hardest *NP* problems. Note that *NP*-hard problems don't have to be *NP*.

## Definition

An problem is *NP*-complete if it's in *NP* and in *NP*-hard.

# Subset sum problem

## Problem

Given a set of integers  $S$ , is there a non-empty subset whose elements sum to zero?

## Example

Does  $\{1, 3, 7, -5, -13, 2, 9, -8\}$  have such a subset?

## Note

If somebody suggests a solution, it is very quick to check it. Being able to quickly verify a solution is a characteristic of *NP* problems.

# Propositional logic

**Literals** are Boolean variables (can be True or False), and their negations. They are represented by lower case letters like  $a$  and  $x_i$ .

**Clause** are expressions based on literals, that evaluate as True or False based on the literals. We use not, and and or on the literals. We'll sometimes call them expressions.

**Not** is depicted by  $\neg$ . So “not  $a$ ” is denoted by  $\neg a$ .

**Or** is depicted by  $\vee$ . So “ $a$  or  $b$ ” is denoted by  $a \vee b$ .

**And** is depicted by  $\wedge$ . So “ $a$  and  $b$ ” is denoted by  $a \wedge b$ .

**CNF** A clause is in Conjunctive Normal Form if it is a  
“conjunction of disjunctions”:  $(a \vee b) \wedge (\neg a \vee c) \wedge d$ .

**DNF** A clause is in Disjunctive Normal Form if it is a  
“disjunction of conjunctions”:  $(a \wedge b) \vee (\neg a \wedge c) \vee d$ .

### Converting to CNF and DNF

Every Boolean expression can be converted to CNF, and every Boolean expression can also be converted to DNF.



## Four laws

The following four laws can be used to convert expressions to CNF and DNF. The first two are known as De Morgan's laws, and the latter two are called the distributivity laws.

### Conversion laws

$$\neg(a \vee b) = \neg a \wedge \neg b$$

$$\neg(a \wedge b) = \neg a \vee \neg b$$

$$c \wedge (a \vee b) = (c \wedge a) \vee (c \wedge b)$$

$$c \vee (a \wedge b) = (c \vee a) \wedge (c \vee b)$$

# Boolean Satisfiability Problem (SAT)

## The problem

We're often interested in knowing if there is any setting of the variables in a Boolean expression that makes the expression true. Another way of asking the question is: is the expression satisfiable? The real question is, is there a polynomial time algorithm that takes as input *any* Boolean expression and outputs true if there is any setting, and false otherwise.

The problem is the prototypical NP-complete problem. Note that it's quick to check the correctness of a solution for a given expression.

## Only using CNF

All Boolean expressions can be converted in CNF. However, it's not always possible to do that in polynomial time. We can though, in polynomial time, create new expressions using some extra(neous) variables that are satisfiable if and only if the original expression is.

## ***k*-SAT**

*k*-SAT is like SAT except that all expressions must be in CNF and each clause must be a disjunction of *k* literals.

**2-SAT**  $(a \vee b) \wedge (\neg c \vee d) \wedge \dots$

**3-SAT**  $(a \vee b \vee c) \wedge (\neg c \vee d \vee \neg a) \wedge \dots$

**2-SAT** is not NP-complete. There are polynomial time algorithms that solve it.

**3-SAT** is NP complete.

## 3-SAT is NP-Complete

### Reduction

3-SAT is a special case of SAT, so 3-SAT must be in NP. We can reduce SAT to 3-SAT in polynomial time. First take the expression and convert it to a CNF expression (in polynomial time). Then we just need to convert each clause into a CNF expression with 3 literals per clause.

Suppose we have a clause with 1 literal:  $a$ . Convert this to  $(a \vee u_1 \vee u_2) \wedge (a \vee u_1 \vee \neg u_2) \wedge (a \vee \neg u_1 \vee u_2) \wedge (a \vee \neg u_1 \vee \neg u_2)$ .  
Suppose we have a clause with 2 literals:  $a \vee b$ . Convert this to  $(a \vee b \vee u_1) \wedge (a \vee b \vee \neg u_1)$ .

Now suppose we have a clause with  $n$  literals:  $a \vee b \vee c \vee \dots$   
Convert this to  $(a \vee b \vee u_1) \wedge (c \vee \neg u_1 \vee u_2) \wedge \dots \wedge (i \vee \neg u_{n-4} \vee u_{n-3}) \wedge (j \vee k \vee \neg u_{n-3})$ .