

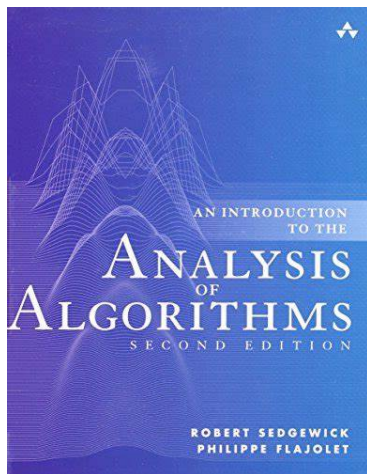
## 2-2 The Efficiency of Algorithms

Hengfeng Wei

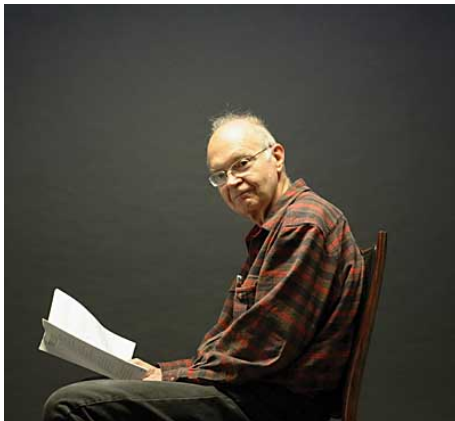
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March 05, 2020





## The Analysis of Algorithms



Donald E. Knuth (1938 ~)



Donald E. Knuth (1974)



Donald E. Knuth (1974)

*“For his major contributions to **the analysis of algorithms**  
and **the design of programming languages**,  
and in particular for his contributions to  
the **“art of computer programming”** through  
his well-known books in a continuous series by this title.”*

*“People who analyze algorithms have double happiness.”*

*“People who **analyze algorithms** have **double happiness**.*

*First of all they experience the sheer beauty of elegant **mathematical patterns** that surround elegant **computational procedures**.*

## Fibonacci numbers in the analysis of Euclid's GCD algorithm

*“People who analyze algorithms have double happiness.*

*First of all they experience the sheer beauty of elegant mathematical patterns that surround elegant computational procedures.*



Fibonacci numbers in the analysis of Euclid's GCD algorithm  
 $H_n$  in the analysis of FIND-MAX @ Stanford Lecture by Knuth

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*“People who **analyze algorithms** have **double happiness**.*

*First of all they experience the sheer beauty of elegant **mathematical patterns** that surround elegant **computational procedures**.*

*Then they receive a **practical payoff** when their theories make it possible to get other jobs done **more quickly and more economically**.”*

## How Fast is It?



## How Fast is It?



Time (and Space) Complexity of Algorithms

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Time (and Space) Complexity of Algorithms

$O$   $\Omega$   $\Theta$

$o$   $\omega$

# Space Complexity of Algorithms

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INSERTION-SORT( $A, n$ ) :  $O(1)$  (constant)

Is it the Fastest?



Is it the Fastest?



Complexity of Problems

Is it the Fastest?



Complexity of Problems

This is much harder and is not our focus today.





Whenever you design an algorithm,



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you provide an **upper bound** for the **complexity of the problem**.



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When the gap is gone, you get the **optimal** algorithm.



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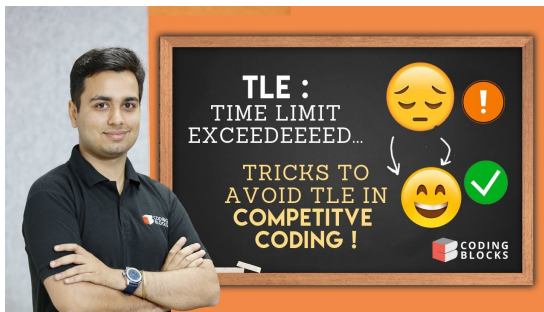
When the gap is gone, you get the **optimal** algorithm.

$$\text{sorting}(A, n) : \Theta(n \log n) = O(n \log n) \cap \Omega(n \log n)$$

$Q$  : How fast is your algorithm?

Q : How fast is your algorithm?

A : It runs 3.1415926 seconds.



Disadvantages:

## Disadvantages:

- ▶ On different machines



## Disadvantages:

- ▶ On different machines
- ▶ At different time

## Disadvantages:

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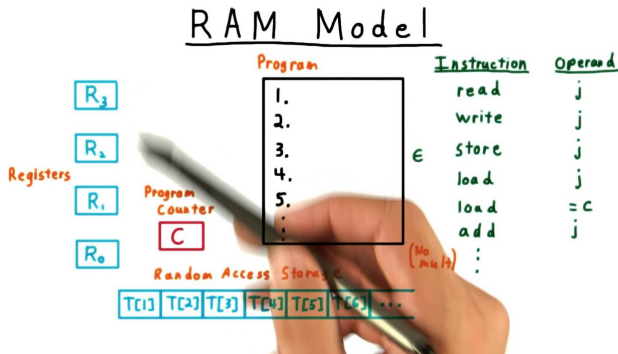
- ▶ On different machines
- ▶ At different time
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No Standards.

We need a uniform **model of computation**.

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## The RAM (Random Access Machine) Model of Computation



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Counting up the number of time units  
as a function of the input size  
in typical cases.

# INSERTION-SORT( $A$ )

	<i>cost</i>	<i>times</i>
1 <b>for</b> $j = 2$ <b>to</b> $A.length$	$c_1$	$n$
2 $key = A[j]$	$c_2$	$n - 1$
3       // Insert $A[j]$ into the sorted sequence $A[1..j - 1]$ .	0	$n - 1$
4 $i = j - 1$	$c_4$	$n - 1$
5 <b>while</b> $i > 0$ and $A[i] > key$	$c_5$	$\sum_{j=2}^n t_j$
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$$\begin{aligned}
 T(n) = & c_1 n + c_2(n - 1) + c_4(n - 1) \\
 & + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n - 1)
 \end{aligned}$$

... as a function of the input size ...

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$T(n)$ : Depends on *which* input of size  $n$

... in typical cases.

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Problem  $P$       Algorithm  $A$



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Inputs:  $\mathcal{X}_n$  of size  $n$

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... in typical cases.

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$$B(n) = (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8)$$



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$$A(n) =$$

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$$A(n) = 2.25n^2 + 7.75n - 3H_n - 6 \quad (H_n = \sum_{k=1}^n \frac{1}{k} \approx \ln n)$$

$Q$  : How fast is your algorithm?

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**listen carefully.**

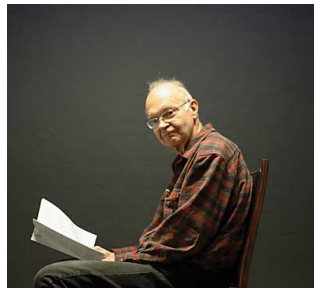
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## BIG OMICRON AND BIG OMEGA AND BIG THETA

Donald E. Knuth  
Computer Science Department  
Stanford University  
Stanford, California 94305

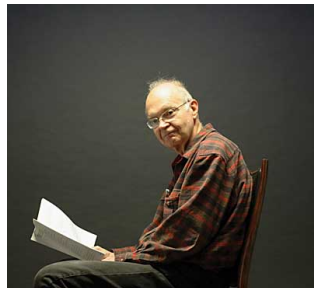


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*“Big Omicron and Big Omega and Big Theta”*, Donald E. Knuth, 1976.

## BIG OMICRON AND BIG OMEGA AND BIG THETA

Donald E. Knuth  
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## Asymptotics



Q : How fast is your algorithm?

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$$W(n) = O(n^2)$$

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“Order at most  $n^2$ ”

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$$W(n) = O(n^2)$$

“Order at most  $n^2$ ”

“ $W(n)$  is a function whose **order of magnitude** is **upper-bounded** by a **constant times  $n^2$** , for all large  $n$ .”

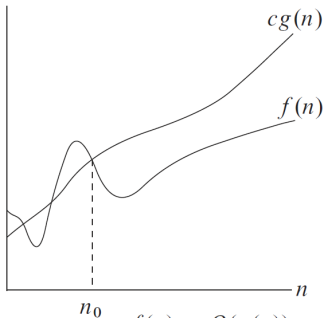
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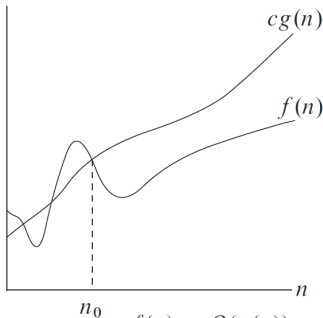
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$$O(g(n)) = \left\{ f(n) \mid \exists c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq f(n) \leq cg(n) \right\}$$



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{ }

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$$\{ \}$$

It is a tradition to write  $f(n) = O(g(n))$  instead of  $f(n) \in O(g(n))$ .

$$42n^2 + 2020n = O(n^2)$$

$$42n^2 + 2020n = O(n^2) = O(n^3)$$

$$42n^2 + 2020n = O(n^2) = O(n^3)$$

$$42n^2 + 2020n \in O(n^2) \subseteq O(n^3)$$

$$O(f(n)) + O(g(n)) \triangleq$$

$$O(f(n)) + O(g(n)) \triangleq \{h + l \mid h \in O(f(n)), l \in O(g(n))\}$$



$$O(f(n)) + O(g(n)) \triangleq \{h + l \mid h \in O(f(n)), l \in O(g(n))\}$$

$$O(f(n))O(g(n)) \triangleq \{hl \mid h \in O(f(n)), l \in O(g(n))\}$$

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$$O(f(n)) - O(g(n)) \triangleq$$

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$$42n = O(0.50n^2)$$

$$O(g(n)) = \left\{ f(n) \mid \exists c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq f(n) \leq cg(n) \right\}$$

$$42n = O(0.50n^2) \quad 42n^2 = O(0.50n^2)$$

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*Q* : What does  $O(1)$  mean?

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$$42n = O(0.50n^2) \quad 42n^2 = O(0.50n^2)$$

*Q* : What does  $O(1)$  mean?

*A* : It means constants.

$$\Omega(g(n)) = \left\{ f(n) \mid \exists c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq cg(n) \leq f(n) \right\}$$



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$$0.50n^2 = \Omega(42n)$$

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$$0.50n^2 = \Omega(42n) \quad 0.50n^2 = \Omega(42n^2)$$

$$\Theta(g(n)) = \left\{ f(n) \mid \exists c_1 > 0, \exists c_2 > 0, \exists n_0 > 0, \forall n \geq n_0 : \right. \\ \left. 0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \right\}$$

$$\Omega(g(n)) = \left\{ f(n) \mid \exists c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq cg(n) \leq f(n) \right\}$$

$$0.50n^2 = \Omega(42n) \quad 0.50n^2 = \Omega(42n^2)$$

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$$0.50n^2 = \Theta(42n^2)$$

$$o(g(n)) = \left\{ f(n) \mid \forall c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq f(n) < cg(n) \right\}$$

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$$42n = o(0.50n^2)$$

$$\omega(g(n)) = \left\{ f(n) \mid \forall c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq cg(n) < f(n) \right\}$$

$$o(g(n)) = \left\{ f(n) \mid \forall c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq f(n) < cg(n) \right\}$$

$$42n = o(0.50n^2)$$

$$\omega(g(n)) = \left\{ f(n) \mid \forall c > 0, \exists n_0 > 0, \forall n \geq n_0 : 0 \leq cg(n) < f(n) \right\}$$

$$0.50n^2 = \omega(42n)$$



$O$     $\Omega$     $\Theta$

$o$     $\omega$     $\theta$

$O \quad \Omega \quad \Theta$

$o \quad \omega \quad \theta$

$$f(n) \sim g(n) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$$

$O \quad \Omega \quad \Theta$

$o \quad \omega \quad \theta$

$$f(n) \sim g(n) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$$

$$42n^2 + 2020n \sim 42n^2 + 2019n$$

$$f(n) = \Theta(g(n)) \iff f(n) = O(g(n)) \wedge f(n) = \Omega(g(n))$$

$$f(n) = O(g(n)) \iff g(n) = \Omega(f(n))$$

$$f(n) = o(g(n)) \iff g(n) = \omega(f(n))$$

$$O(f(n)) + O(g(n)) = O(f(n) + g(n))$$

$$O(f(n)) + O(g(n)) = O(f(n) + g(n))$$

$$O(f(n))O(g(n)) = O(f(n)g(n))$$

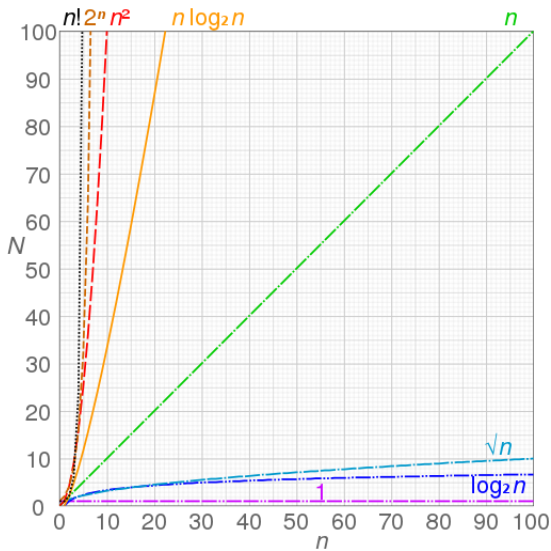
$Q$  : How to compare functions in terms of  $O/\Omega/\Theta$ ?

*Q* : How to compare functions in terms of  $O/\Omega/\Theta$ ?

$$\begin{aligned} O(1) &= O(\log \log n) = O(\log n) = O((\log n)^c) \\ &= O(n^\epsilon) = O(n^c) \\ &= O(n^c \log n) = O(n^{\log n}) = O(c^n) = O(n^n) \end{aligned}$$

$(0 < \epsilon < 1 < c)$





Stirling Formula (by *James Stirling*):

$$n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$



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$$\log(n!) = \Theta(n \log n)$$

$$H_n = \sum_{k=1}^n \frac{1}{k} = \Theta(\log n)$$



$$A[0, \dots, n-1] \quad 1 \leq l \leq n$$

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ROTATE( $A, n, l$ ) : Rotate  $A$  left by  $l$  places

$$A[0, \dots, n-1] \quad 1 \leq l \leq n$$

ROTATE( $A, n, l$ ) : Rotate  $A$  left by  $l$  places

0	1	2	3	4
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---



$$A[0, \dots, n-1] \quad 1 \leq l \leq n$$

ROTATE( $A, n, l$ ) : Rotate  $A$  left by  $l$  places

0	1	2	3	4
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

Critical Operation: copy

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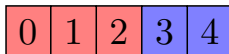
```
1: procedure ROTATE( $A, n, l$ )  
2:   for  $i = 1 \dots l$  do  
3:     ROTATE-BY-ONE( $A, n$ )
```

---

---

```
1: procedure ROTATE( $A, n, l$ )
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```

---



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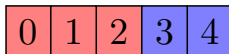
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```

1: procedure ROTATE( $A, n, l$ )
2:   for  $i = 1 \dots l$  do
3:     ROTATE-BY-ONE( $A, n$ )

```

---



Algorithm	Time	Space
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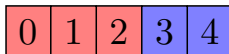
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```

1: procedure ROTATE( $A, n, l$ )
2:   for  $i = 1 \dots l$  do
3:     ROTATE-BY-ONE( $A, n$ )

```

---



Algorithm	Time	Space
rotate-one-by-one	$nl = O(n^2)$	$O(1)$

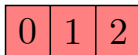
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```
1: procedure ROTATE( $A, n, l$ )  
2:   copy  $A[0 \dots l - 1]$  into  $v$   
3:   move  $A[l \dots n - 1]$  left  $l$  places  
4:   copy  $v$  to  $A[l \dots n - 1]$ 
```

---

- 
- 1: **procedure** ROTATE( $A, n, l$ )
  - 2:     copy  $A[0 \dots l - 1]$  into  $v$
  - 3:     move  $A[l \dots n - 1]$  left  $l$  places
  - 4:     copy  $v$  to  $A[l \dots n - 1]$
- 



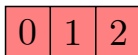
- 
- 
- 1: **procedure** ROTATE( $A, n, l$ )
  - 2:     copy  $A[0 \dots l - 1]$  into  $v$
  - 3:     move  $A[l \dots n - 1]$  left  $l$  places
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Algorithm	Time	Space
-----------	------	-------



- 
- 1: **procedure** ROTATE( $A, n, l$ )
  - 2:     copy  $A[0 \dots l - 1]$  into  $v$
  - 3:     move  $A[l \dots n - 1]$  left  $l$  places
  - 4:     copy  $v$  to  $A[l \dots n - 1]$
- 



Algorithm	Time	Space
rotate-copy	$O(n)$	$l = O(n)$

$$n = 5, \quad l = 3$$

0	1	2	3	4
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

$$n = 5, \quad l = 3$$

0	1	2	3	4
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

(0, 2, 4, 1, 3)

$$n = 5, \quad l = 3$$

0	1	2	3	4
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

(0, 2, 4, 1, 3)

$$n = 9, \quad l = 6$$

0	1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---	---

6	7	8	0	1	2	3	4	5
---	---	---	---	---	---	---	---	---

(0, 3, 6)      (1, 4, 7)      (2, 5, 8)

# Correctness Proof?

## Correctness Proof?

Permutations as **Product** of  
**Disjoint Cycles**



## Correctness Proof?

Permutations as **Product** of  
**Disjoint Cycles**



Algorithm	Time	Space
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## Correctness Proof?

Permutations as **Product** of  
**Disjoint Cycles**



Algorithm	Time	Space
rotate-cyclic	$O(n)$	$O(1)$



$$B \cdot A = (A^R \cdot B^R)^R$$

0	1	2	3	4
---	---	---	---	---

2	1	0	4	3
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

$$B \cdot A = (A^R \cdot B^R)^R$$

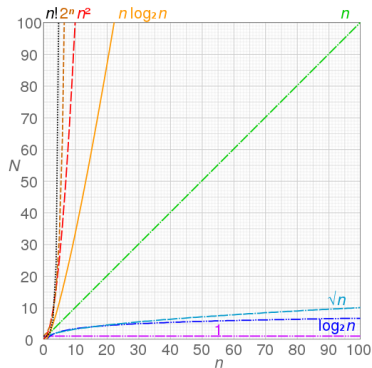
0	1	2	3	4
---	---	---	---	---

2	1	0	4	3
---	---	---	---	---

3	4	0	1	2
---	---	---	---	---

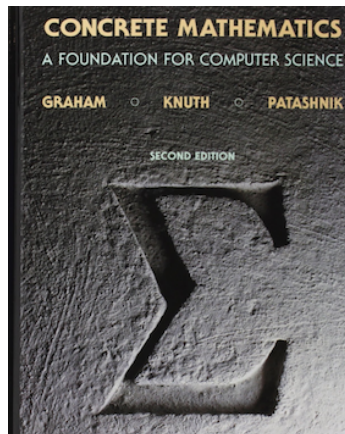
Algorithm	Time	Space
rotate-reverse	$O(n)$	$O(1)$

Algorithm	Time	Space
rotate-one-by-one	$O(n^2)$	$O(1)$
rotate-copy	$O(n)$	$O(n)$
rotate-cyclic	$O(n)$	$O(1)$
rotate-reverse	$O(n)$	$O(1)$



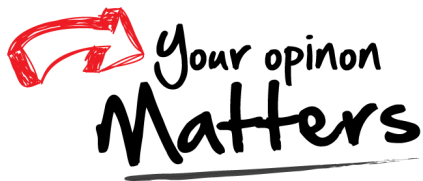
$O \quad \Omega \quad \Theta$

$o \quad \omega$



## Chapter 9: Asymptotics

Thank  
You!



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