

Discrete Structures

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Text book

Discrete Mathematics and Its Application, 7th Edition
Kenneth H. Rosen

References

Chapter 3

1. Discrete Mathematics and Its Application, 7^h Edition

by

Kenneth H. Rose

2. Discrete Mathematics with Applications

by

Thomas Koshy

3. Discrete Mathematical Structures, CS 173

by

Cinda Heeren, Siebel Center

4. <https://www.cs.cornell.edu/courses/JavaAndDS/files/constantTime.pdf>

5. Discrete Mathematics for Computer Science by Gary Haggard

These slides contain material from the above resources.

Constant time

Constant time: An operation or method takes constant time if the time it takes to carry it out does not depend on the **size of its operands**.

For example, an **array element reference $b[i]$** takes **constant time**, but **printing out all elements of array b** is **not constant time** but instead takes time proportional to the **size of b** .

Also, this assignment takes **constant time**:

$b[i] = b[i] + 2$

Example Give **big-O estimates** for the **factorial function** and the **logarithm of the factorial function**, where the factorial function **$f(n) = n!$** is defined by

$n! = 1 \times 2 \times 3 \dots n$, whenever n is a positive integer, and $0! = 1$.

For example,

$1! = 1$, $2! = 1 \times 2 = 2$, $3! = 1 \times 2 \times 3 = 6$, $4! = 1 \times 2 \times 3 \times 4 = 24$.

Note that the function **$n!$ grows rapidly**.

For instance, **$20! = 2,432,902,008,176,640,000$**

Solution:

We have to show $f(n) = n!$ is $O(n^n)$

$|f(n)| \leq C|g(n)|$ whenever $n > k$.

$$n! = 1 \times 2 \times \dots \times n$$

$$n! \leq n \times n \times n \dots n, \text{ where } k > \mathbf{1}$$

$$0 \leq n! \leq n \times n \times n \dots n = n^n$$

$$\Rightarrow n! \leq n^n$$

$$\Rightarrow n! \leq \mathbf{1} \times n^n \\ = Cg(n)$$

Here **$f(n) = n!$** and **$g(n) = n^n$**

Consequently, we can take $C = 1$ and $k = 1$ as witnesses to show that $f(n)$ is $O(n^n)$

n	$n! \leq n^n$
1	$1! \leq 1^2$ (true)
2	$2! \leq 2^2$ (True)
3	$3! \leq 3^2$ (True)
\vdots	\vdots

$$\because n! \leq n^n$$

Taking log on both sides

$$\Rightarrow \log n! \leq \log n^n$$

$$\Rightarrow \log n! \leq n \log n$$

$$\because \log m^n = n \times \log m$$

This implies that **$\log n!$** is **$O(n \log n)$** , again taking $C = 1$ and $k = 1$ as witnesses.

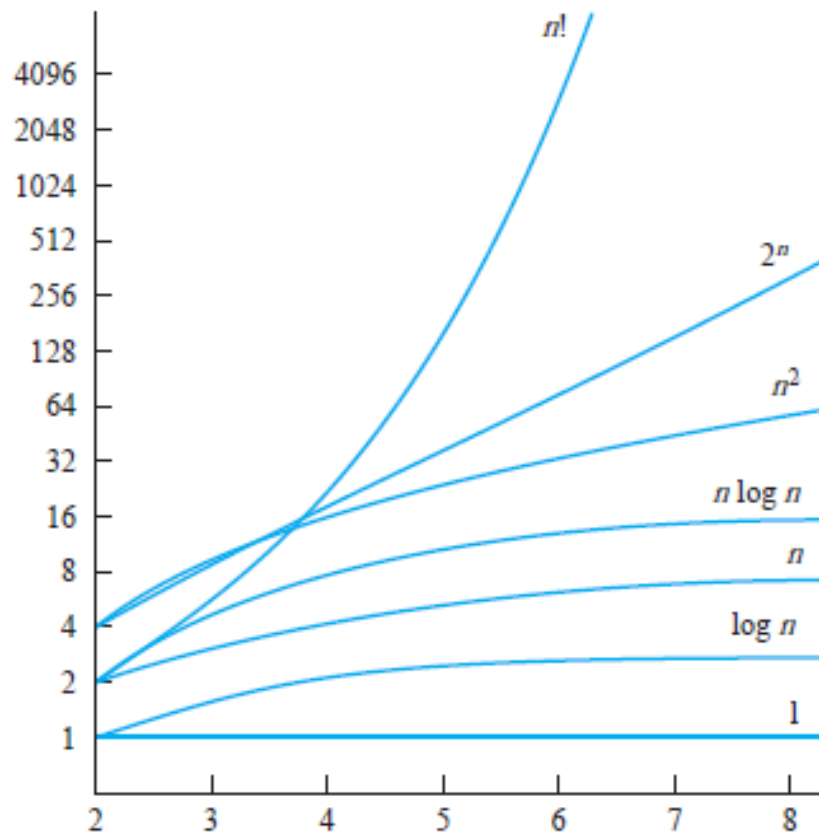
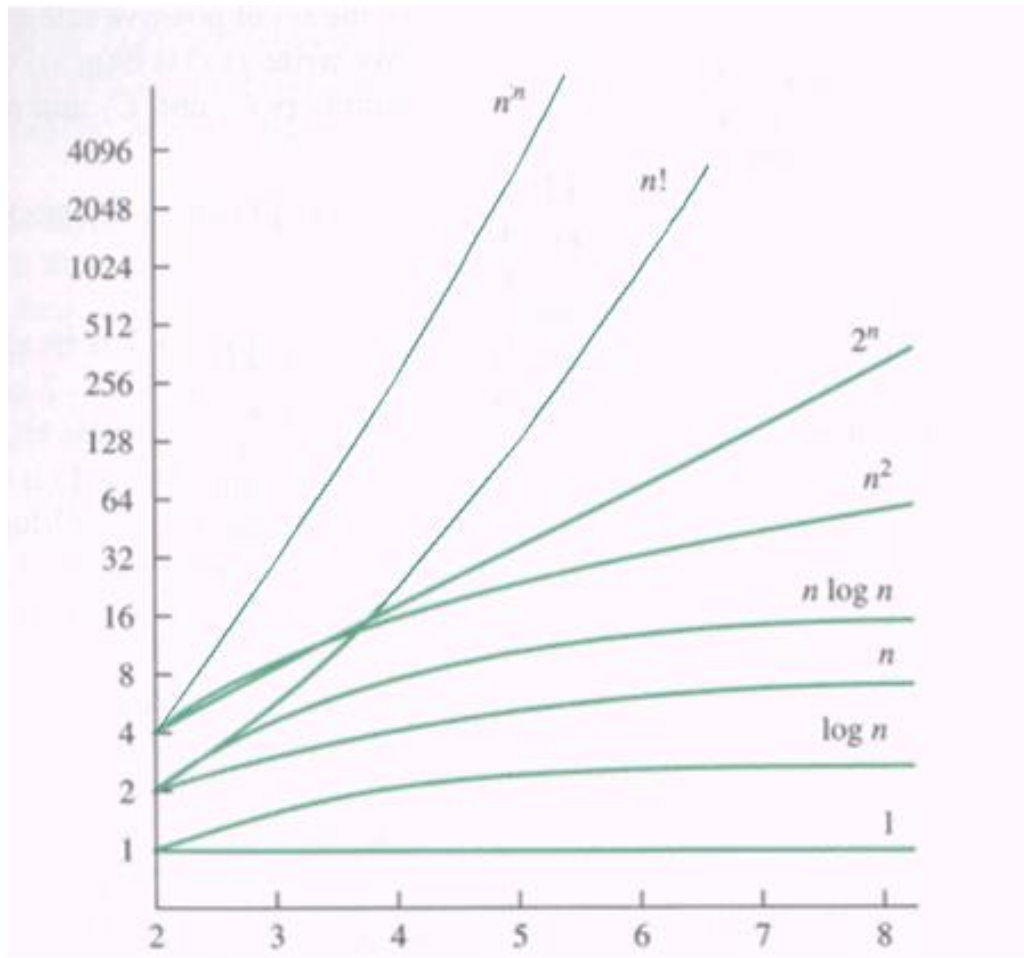


FIGURE 3 A Display of the Growth of Functions Commonly Used in Big- O Estimates.



Complexity Comparisons for Various Functions

- To see the difference in the **time requirement** for processing **data sets of arbitrary size**, we will assume **a single machine cycle** will require **10^{-6} seconds** to be completed.
- **Table** on the next slide gives the **time required to process a data set of size n** , for six different values of n , if it takes $\log_2(n)$ (n , n^2 , n^5 , and 2^n , respectively) machine cycles to make the computation.
- For example, in the column labeled n^2 for the row labeled $n = 100$, 100^2 operations are needed to complete execution. The time is **$(10^2)^2 10^{-6}$ seconds = 10^{-2} seconds**.

Complexity Comparisons for Various Functions

$F(n)$	$\log_2(n)$	n	n^2	n^5	2^n
$n = 10$	3×10^{-6} sec	10^{-5} sec	10^{-4} sec	0.1 sec	10^{-3} sec
$n = 20$	4×10^{-6} sec	2×10^{-5} sec	4×10^{-4} sec	3 sec	1 sec
$n = 50$	6×10^{-6} sec	5×10^{-5} sec	3×10^{-3} sec	5 min	36 yrs
$n = 100$	7×10^{-6} sec	10^{-4} sec	10^{-2} sec	3 hrs	4×10^{16} yrs
$n = 1000$	1×10^{-5} sec	10^{-3} sec	1 sec	32 yrs	3.9×10^{287} yrs
$n = 100,000$	2×10^{-5} sec	0.1 sec	2.7 hrs	3×10^{11} yrs	$> 10^{30,089}$ yrs

What Does Machine Cycle Mean?

- A machine cycle consists of the steps that a computer's **processor executes whenever it receives a machine language instruction**.
- It is the most basic CPU operation, and modern CPUs are able to perform millions of machine cycles per second.
- The cycle consists of three standard steps: fetch, decode and execute. In some cases, store is also incorporated into the cycle.

Show that $7x^2$ is $O(x^3)$.

Solution:

$7x^2$ is $O(x^3)$.

$|f(x)| \leq C|g(x)|$ whenever $x > k$.

$$f(x) = 7x^2$$

$$g(x) = x^3$$

We observe that we can readily estimate the size of $f(x)$ when **$x > 7$**

$$\because 7x^2 < x^3. \quad \text{when } x > 7.$$

$$0 \leq 7x^2 \leq x^3$$

$$\Rightarrow 7x^2 \leq \mathbf{1} \times x^3$$

$$\Rightarrow f(x) \leq Cg(x)$$

Consequently, we can take **$C = 1$** and **$k = 7$** as witnesses to establish

Show that $7x^2$ is $O(x^3)$.

Alternative solution:

$7x^2$ is $O(x^3)$.

$|f(x)| \leq C|g(x)|$ whenever $x > k$.

$$f(x) = 7x^2$$

$$g(x) = x^3$$

We observe that we can readily estimate the size of $f(x)$ when $x > 1$

$$\because 7x^2 < 7x^3.$$

when $x > 1$.

$$0 \leq 7x^2 \leq 7x^3$$

$$\Rightarrow 7x^2 \leq \mathbf{7 \times x^3}$$

$$\Rightarrow f(x) \leq Cg(x)$$

Consequently, we can take **$C = 7$** and **$k = 1$** as witnesses to establish

Show that n^2 is $O(n)$.

Solution:

n^2 is $O(n)$.

$|f(n)| \leq C|g(n)|$ whenever $n > k$.

$$f(n) = n^2$$

$$g(n) = n$$

We have to show that

$$n^2 \leq Cn$$

Dividing both sides by n

$$\Rightarrow n \leq C$$

We cannot find any C and k as witnesses

Note: C and k are constants, whereas k is a positive real number and C is a real number

Example: Find Big-oh notation of $1 + 2 + 3 + \dots + n$

Let $f(n) = 1 + 2 + 3 + \dots + n$

We have to show that $f(n) = 1 + 2 + 3 + \dots + n$ is $O(n^2)$

$\therefore |f(n)| \leq C|g(n)|$ whenever $n > k$.

$1 < n^2, 2 < n^2, 3 < n^2, \dots$ so on when $n > 1$

$$f(n) = 1 + 2 + 3 + \dots + n$$

$$\leq n + n + n + \dots + n$$

$$\leq n \times n$$

$$\leq 1 \times n^2$$

$$\Rightarrow f(n) \leq Cg(n)$$

Consequently, we can take $C = 1$ and $k = 1$ as witnesses to establish

Show that $\lg n$ is $O(n)$

We will prove by mathematical induction

$$n < 2^n$$

$$\Rightarrow f(x) \leq Cg(x)$$

Consequently, we can take $C = 1$ and $k = 1$ as witnesses to establish

We conclude that n is $O(2^n)$

$$n < 2^n$$

Taking log on both sides with base 2

$$\lg n < \lg 2^n$$

$$\lg n < n \lg 2 \qquad \because \lg 2 = 1$$

$$\lg n < n$$

$$\Rightarrow f(x) \leq Cg(x)$$

Consequently, we can take $C = 1$ and $k = 1$ as witnesses to establish. We conclude that $\lg n$ is $O(n)$

Cont.

If we have logarithms to a base b , where b is different from 2, we still have $\log_b n$ is $O(n)$

Because

$$\because \lg n < n$$

$$\log_b n = \frac{\log n}{\log b} < \frac{n}{\log b}$$

$$\because \log_b n = \frac{\log_c n}{\log_c b}$$

$$\log_b n = \frac{\log n}{\log b} < \frac{1}{\log b} \times n$$

whenever n is a positive integer. We take $C = 1/\log b$ and $k = 1$ as witnesses.

- $n! = O(n^n)$
- $\log(n!) = O(n \log n)$.
- $\log_b n = O(n)$
- $n = O(2^n)$
- $f(n) = 1 + 2 + 3 + \dots + n$ is $O(n^2)$

The Growth of Combinations of Functions

Theorem Suppose that $f_1(x)$ is $O(g_1(x))$ and that $f_2(x)$ is $O(g_2(x))$. Then $(f_1 + f_2)(x)$ is $O(\max(|g_1(x)|, |g_2(x)|))$.

Corollary Suppose that $f_1(x)$ and $f_2(x)$ are both $O(g(x))$. Then $(f_1 + f_2)(x)$ is $O(g(x))$.

Theorem Suppose that $f_1(x)$ is $O(g_1(x))$ and $f_2(x)$ is $O(g_2(x))$. Then $(f_1 f_2)(x)$ is $O(g_1(x)g_2(x))$.

Example Give a **big-O estimate** for $f(n) = 3n \log(n!) + (n^2 + 3) \log n$, where n is a positive integer.

Solution: for $f(n) = 3n \log(n!) + (n^2 + 3) \log n$

First, the product $3n \log(n!)$ will be estimated.

$$\log(n!) = O(n \log n).$$

$$3n \text{ is } O(n)$$

Suppose that $f_1(x)$ is $O(g_1(x))$ and $f_2(x)$ is $O(g_2(x))$. Then $(f_1 f_2)(x)$ is $O(g_1(x)g_2(x))$.

$$\Rightarrow 3n \log(n!) = O(n^2 \log n).$$

$(n^2 + 3) \log n$ will be estimated:

Because $(n^2 + 3) < 2n^2$ when $n > 2$, it follows that

$$n^2 + 3 = O(n^2).$$

$$\Rightarrow (n^2 + 3) \log n = O(n^2 \log n).$$

Cont.

Theorem Suppose that $f_1(x)$ is $O(g_1(x))$ and that $f_2(x)$ is $O(g_2(x))$.
Then $(f_1 + f_2)(x)$ is $O(\max(|g_1(x)|, |g_2(x)|))$.

$$\therefore f(n) = 3n \log(n!) + (n^2 + 3) \log n$$

is $O(\max(n^2 \log n, n^2 \log n))$ is $O(n^2 \log n)$.

Example: Let $f(n) = 6n^2 + 5n + 7\lg n!$ Estimate the growth of $f(n)$

Solution:

$$f(n) = 6n^2 + 5n + 7\lg n!$$

$$6 = O(1)$$

$$6n^2 = O(n^2)$$

$$5 = O(1)$$

$$5n = O(n)$$

$$\because (f_1 + f_2)(x) \text{ is } O(\max(|g_1(x)|, |g_2(x)|))$$

$$6n^2 + 5n = O(\max(n^2, n)) = O(n^2)$$

$$7 = O(1)$$

$$\lg n! = O(n \lg n)$$

$$\because (f_1 f_2)(x) \text{ is } O(g_1(x) g_2(x))$$

$$7\lg n! = O(1 \cdot n \lg n) = O(n \lg n)$$

$$\because (f_1 + f_2)(x) \text{ is } O(\max(|g_1(x)|, |g_2(x)|))$$

$$6n^2 + 5n + 7\lg n! \text{ is } O(\max(n^2, n \lg n)) = O(n^2)$$

big-Omega (Ω)

Let f and g be functions from the set of integers or the set of real numbers to the set of real numbers. We say that $f(x)$ is $\Omega(g(x))$ if there are positive constants C and k such that

$|f(x)| \geq C|g(x)|$ whenever $x > k$. [This is read as “ $f(x)$ is big-Omega of $g(x)$.”]

Note: There is a strong connection between big-O and big-Omega notation. In particular, $f(x)$ is $\Omega(g(x))$ if and only if $g(x)$ is $O(f(x))$

Example The function $f(x) = 8x^3 + 5x^2 + 7$ is $\Omega(g(x))$, where $g(x)$ is the function $g(x) = x^3$.

$$|f(x)| \geq C|g(x)| \text{ whenever } x > k.$$

This is easy to see because $f(x) = 8x^3 + 5x^2 + 7 \geq 8x^3$ for all positive real numbers x .

This is equivalent to saying that $g(x) = x^3$ is $O(8x^3 + 5x^2 + 7)$, which can be established directly by turning the inequality around.

big-Theta(Θ)

- Let f and g be functions from the set of integers or the set of real numbers to the set of real numbers. We say that $f(x)$ is $\Theta(g(x))$ if $f(x)$ is $O(g(x))$ and $f(x)$ is $\Omega(g(x))$. When $f(x)$ is $\Theta(g(x))$ we say that “ f is big-Theta of $g(x)$ ”, that $f(x)$ is of order $g(x)$, and that $f(x)$ and $g(x)$ are of the same order.
- When $f(x)$ is $\Theta(g(x))$, it is also the case that $g(x)$ is $\Theta(f(x))$. Also note that $f(x)$ is $\Theta(g(x))$ if and only if $f(x)$ is $O(g(x))$ and $g(x)$ is $O(f(x))$.
- $f(x)$ is $\Theta(g(x))$ if and only if there are real numbers C_1 and C_2 and a positive real number k such that $C_1|g(x)| \leq |f(x)| \leq C_2|g(x)|$ whenever $x > k$. The existence of the constants C_1 , C_2 , and k tells us that $f(x)$ is $\Omega(g(x))$ and that $f(x)$ is $O(g(x))$, respectively

Example Show that $3x^2 + 8x \log x$ is $\Theta(x^2)$.

Solution:

$$f(x) = 3x^2 + 8x \log x$$

$$C_1 |g(x)| \leq |f(x)| \leq C_2 |g(x)| \text{ whenever } x > k$$

For O:

$$|f(x)| \leq C_2 |g(x)| \quad \text{whenever } x > k.$$

$$x \log x < x^2 \quad \text{whenever } x > 1$$

$$0 \leq 3x^2 + 8x \log x \leq 3x^2 + 8x^2 \leq 11x^2 \quad \text{whenever } x > 1$$

$$3x^2 + 8x \log x \leq 11x^2$$

$$3x^2 + 8x \log x \leq Cg(x) \quad \text{whenever } x > 1$$

$$C_2 = 11, k = 1$$

Note: C_2 and k are constants.

We have to show that $3x^2 + 8x \log x$ is $\Theta(x^2)$.

$$\mathbf{f(x) = 3x^2 + 8x \log x}$$

For Ω :

$$\mathbf{|f(x)| \geq C_1 |g(x)|} \qquad \mathbf{\text{whenever } x > k.}$$

$$3x^2 + 8x \log x \geq \mathbf{3}x^2 \quad \text{whenever } x > 1.$$

$$3x^2 + 8x \log x \geq C_2 |g(x)|$$

$$C_1 = 3, k = 1$$

Note: C_1 and k are constants.

$$3x^2 + 8x \log x \text{ is } \Theta(x^2).$$

Theorem Let $f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$, where a_0, a_1, \dots, a_n are real numbers with $a_n \neq 0$. Then $f(x)$ is of order x^n .

Example The polynomials

$$3x^8 + 10x^7 + 221x^2 + 1444,$$

$$x^{19} - 18x^4 - 10,112,$$

and

$$-x^{99} + 40,001x^{98} + 100,003x$$

are of orders x^8 , x^{19} , and x^{99} , respectively.

Suggested Readings

Chapter 3

3.1 Algorithms

3.2 The Growth of Functions