

Module 1: Algorithm Structures

- 1 With the pseudo-code below, determine the final value of `sum` when `n=23`.

Algorithm 1

```
1: procedure
2:   sum = 0
3:   for (i=2, i<n, i=i+3) do
4:     if i mod 2 == 0
5:       sum = sum + i
```

Solution: `sum = 44`

With `i` starting at 2 and incrementing by 3 every iteration, `i` will iterate through the values 2, 5, 8, 11, 14, 17, and 20. The `if` statement means `sum` will only be updated when `i` is even, adding 2, 8, 14, and 20 for a total of 44

- 2 Use the pseudo-code below to answer the following question.

- (a) Without working through the pseudo-code, how many times with `val` be updated in Line 7?

Solution: 24

The loop in line 5 will iterate 4 times. The loop in line 6 will iterate 6 times. Since they are nested, we multiply and `val` will be updated $4 \cdot 6 = 24$ times.

- (b) What is the final value of `val`.

Solution: `val = 210`

The algorithm loops through all possible combinations of x and y , multiplies them, and adds them to `val`. In the end, we get

$$\begin{aligned} &1 \cdot 1 + 1 \cdot 2 + 1 \cdot 3 + 1 \cdot 4 + 1 \cdot 5 + 1 \cdot 6 + \\ &2 \cdot 1 + 2 \cdot 2 + 2 \cdot 3 + 2 \cdot 4 + 2 \cdot 5 + 2 \cdot 6 + \\ &3 \cdot 1 + 3 \cdot 2 + 3 \cdot 3 + 3 \cdot 4 + 3 \cdot 5 + 3 \cdot 6 + \\ &4 \cdot 1 + 4 \cdot 2 + 4 \cdot 3 + 4 \cdot 4 + 4 \cdot 5 + 4 \cdot 6 = 210 \end{aligned}$$

Algorithm 2

```
1: procedure
2:   dFour = [1, 2, 3, 4]
3:   dSix = [1, 2, 3, 4, 5, 6]
4:   val = 0
5:   for (x in dFour) do
6:     for (y in dSix) do
7:       val = val + x*y
```

- 3 Using the pseudo-code below, what is the ending value for count when $n = 100$? When $n = 10,000$? When $n = 10^{42}$?

Algorithm 3

```
1: procedure
2:   count=0
3:   while (n>1) do
4:     count++
5:     n = n/10
```

Solution:

When $n = 100$, count will end at 2.

When $n = 10,000$, count will end at 4.

When $n = 10^{42}$, count will end at 42.

- 4 This question assumes familiarity with modular arithmetic discussed in Unit 2. In the pseudo-code below, Line 7 performs string addition (concatenation), for example:
"a" + "b" = "ab". Even if the addends look like numbers, it will still do string addition:
"2" + "3" = "23". Answer the questions below.

- a. What is the output of the algorithm when $x = 99$ and $b = 2$?

Solution: Context for writing a number in base b is presented in Module 6 of Unit 2.

Expand(99, 2) = "1100011"

- b. What is the output of the algorithm when $x = 1642$ and $b = 8$?

Solution: Expand(1642, 8) = "3152"

Algorithm 4

```
1: procedure EXPAND(x, b)
2:   Input:  integers x and b
3:   Output: outVal, x written in base b
4:
5:   outVal = an empty string
6:   while (x>0) do
7:     outVal = string(x mod b) + outVal
8:     x = x div b
9:   return(outVal)
```

Modules 2 & 3: Analyzing Algorithms and Big- \mathcal{O} Estimates

Clarification of notation

The following statements all mean the same thing:

$f(x)$ is $\mathcal{O}(g(x))$ OR

$f(x)$ is of $\mathcal{O}(g(x))$ OR

$f(x) = \mathcal{O}(g(x))$ OR

$f(x) \in \mathcal{O}(g(x))$

$\mathcal{O}(g(x))$ is a collection of functions (i.e. a set) so what we should say is $f(x) \in \mathcal{O}(g(x))$, but $f(x) = \mathcal{O}(g(x))$ is commonly used. This can be confusing since $\mathcal{O}(n) = \mathcal{O}(n^2)$ but $\mathcal{O}(n^2) \neq \mathcal{O}(n)!!$

See this interesting thread on StackExchange [Big O Notation is element of or is equal](#). Note the Wiki cites Donald Knuth.

Big- \mathcal{O} classification of common functions

Order of Asymptotic Behaviour. Remember that these are sets. So while it might be said that $2n + 3$ is $\mathcal{O}(n)$; actually its an element in that set: $2n + 3 \in \mathcal{O}(n) \subset \mathcal{O}(n^2) \subset \dots$

Here's a list of some functions listed from slowest to fastest function growth (shortest to longest runtime for an algorithm).

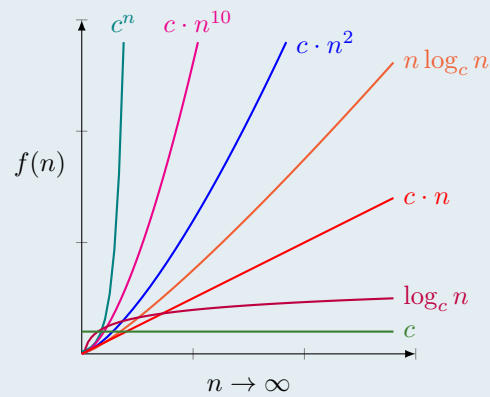
- 1 - Constant
- $\log(n)$ - Logarithmic
 - Ignore log bases: $\mathcal{O}(\log_{10}(n)) = \mathcal{O}(\log_2(n)) = \mathcal{O}(\log(n))$
 - Same growth as $\mathcal{O}(\log(n^k)) = \mathcal{O}(k \log(n)) = \mathcal{O}(\log(n))$
- n - Linear
- $n \log(n)$ - Linearithmic or Loglinear
- n^2 - Quadratic
- $n^2 \log(n)$
- n^3 - Polynomial (including n^4, n^5, n^6, \dots)
- 2^n - Exponential (including $3^n, 4^n, 5^n \dots$)
- $n!$ - Factorial

- 5 Order each function according to their growth from slowest to fastest (This is the same as execution speed from fastest to slowest for an algorithm).

- (a) $n \cdot \log n$
- (b) $\log_{10} n$
- (c) 10^n
- (d) n^{10}
- (e) $10n$
- (f) 100
- (g) n^2

Solution:

- (f) 100
- (b) $\log_{10} n$
- (e) $10n$
- (a) $n \cdot \log 2n$
- (g) n^2
- (d) n^{10}
- (c) 10^n



- 6 The provided expressions are the processing time of an algorithm for problems of size n . For each expression, find the *lowest possible* Big- \mathcal{O} complexity stated in simplest form.

Solution: Given two real functions $f(x)$ and $g(x)$ such that $g(x)$ is strictly positive for sufficiently large x , we say that

$$f(x) \in \mathcal{O}(g(x))$$

if and only if there exist $M, x_0 \in \mathbb{R}$ such that

$$|f(x)| \leq M \cdot g(x)$$

for all $x_0 \leq x$. This formal definition of Big- \mathcal{O} notation is not included in the text. We don't use it directly. Just remember that $f(x)$ is $\mathcal{O}(g(x))$ bound from above by $M \cdot g(x)$, i.e., $\mathcal{O}(g(x))$ is an upper bound.

(a) $100n^2 + 0.0002n^3 + 10,000$

Solution: $\mathcal{O}(n^3)$

From the definition above, we can see that coefficients are ignored. As $n \rightarrow \infty$, they have little effect. So we need only compare n^2 and n^3 . Since $n^3 = n \cdot n^2$, we have $\mathcal{O}(n^2) \subset \mathcal{O}(n^3)$ and our answer is:

$$\mathcal{O}(n^3)$$

(b) $20n + 2n \log_{10} n + 30$

Solution: $\mathcal{O}(n \log n)$

To find this we use the following property:

$$\text{If } f \in \mathcal{O}(f') \text{ and } g \in \mathcal{O}(g') \text{ then } f \cdot g \in \mathcal{O}(f' \cdot g')$$

So for the product of two functions we can evaluate the factors.

$2n \in \mathcal{O}(n)$ and $\log_{10} n \in \mathcal{O}(\log_2 n)$. So then our answer is:

$$\mathcal{O}(n \log_n)$$

As a matter of convention, the base is often excluded, however $\mathcal{O}(\log_2 n)$ is sometimes written.

Why do we ignore the base in logs but not exponents? Recall that

$$\log_2 n = \frac{\log_{10} n}{\log_{10} 2}$$

So then

$$\log_{10} n \leq (\log_{10} 2) \log_2 n$$

for all n . Here our M is $\log_{10} 2$. The situation is different for the bases of *exponential* functions, but remember that logs are the inverses of exponential functions.

(c) $\log_{10} 5 + \log_2 7$

Solution: $\mathcal{O}(1)$

Both terms are constants. So the answer is: $\mathcal{O}(1)$

(d) $5n + \sqrt{n}$

Solution: $\mathcal{O}(n)$ $\sqrt{n} = n^{1/2}$ and $5n^1 = 5n^{1/2}n^{1/2}$. Now we can see that $\mathcal{O}(\sqrt{n}) \subset \mathcal{O}(n)$. So the answer is: $\mathcal{O}(n)$

(e) $0.01(2^n) + n^5 + \ln n$

Solution: $\mathcal{O}(2^n)$ $\mathcal{O}(\log n) \subset \mathcal{O}(n^a) \subset \mathcal{O}(b^n)$ for $a, b \in \mathbb{R}$ and $b > 1$. That is, an exponential function is slower than a polynomial which is slower than a log. So the answer is

$$\mathcal{O}(2^n)$$

To see that $\mathcal{O}(n^a) \subset \mathcal{O}(b^n)$, note that $\log x < \log y \iff x < y$. So then,

$$\log n^a = a \log n < n \log b = \log b^n$$

for sufficiently large n as we know that $\mathcal{O}(\log n) \subset \mathcal{O}(n)$

(f) $100 \cdot 4^n + 200 \cdot 3^n$

Solution: The answer is:

$$\mathcal{O}(4^n)$$

The text incorrectly states that "*the base of the exponent is irrelevant*" in lesson 1.13. $2^n \leq 3^n$ for all n , but there is no constant M such that $3^n \leq M \cdot 2^n$ for all n . If there was then there would be an integer $x > 0$ such that $M < 3^x$, and then $3^x \cdot 3^{n-x} \leq 3^x \cdot 2^{n-x}$ for sufficiently large n . Which is a contradiction.**7** $f(x) = 6x \log_4 x$ is which of the following? **Mark all that apply.**☐ $\mathcal{O}(6)$ ☐ $\mathcal{O}(\log x)$ ☐ $\mathcal{O}(6 \log x)$ ☒ $\mathcal{O}(x \log x)$ ☒ $\mathcal{O}(x^2)$ ☒ $\mathcal{O}(2^x)$ **Solution:** Big- \mathcal{O} is an *upper bound*. So if $f(x)$ is no slower than $x \log x$ then it is also no slower than x^2 and 2^x . In set notation,

$$f(x) \in \mathcal{O}(x \log x) \subset \mathcal{O}(x^2) \subset \mathcal{O}(2^x)$$

8 Let $h(x) = 6x^4 + 2x^3 + x + 100$. Which of the following is true? **Mark all that apply.**☒ $h(x) \in \mathcal{O}(x^4)$ ☒ $h(x) \in \Theta(x^4)$ ☒ $h(x) \in \Omega(x^4)$ **Solution:** $\mathcal{O}(x^4)$ is an *upper bound*. $\Omega(x^4)$ is a *lower bound*. So then $h(x)$ is also $\Theta(x^4)$, a *tight bound*.

- 9 Let $h(x) = 6 \log_2 x$. Which of the following is true? **Mark all that apply.**

☒ $h(x) \in \mathcal{O}(4x^2)$ ☐ $h(x) \in \Theta(4x^2)$ ☐ $h(x) \in \Omega(4x^2)$

Solution: $6 \log_2 x$ is faster than $4x^2$. So then $h(x) \in \mathcal{O}(4x^2)$ and $h(x) \notin \Omega(4x^2)$. Which means $h(x) \notin \Theta(4x^2)$

- 10 Let $f(x) = x^2 \log_3 x$. Which of the following is true? **Mark all that apply.**

☐ $f(x) \in \mathcal{O}(2x \log_{10} x)$ ☐ $f(x) \in \Theta(2x \log_{10} x)$ ☒ $f(x) \in \Omega(2x \log_{10} x)$

Solution: Recall that the base of a logarithmic function can be ignored. So $\log_{10} x$ and $\log_3 x$ have the same asymptotic behavior, and we really only need to evaluate x^2 and $2x$. We know that $x^2 \notin \mathcal{O}(2x)$, and so neither can it be in $\Theta(2x)$ which gives us our answer.

- 11 Let $g(x) = 2x \log_{10} x$. Which of the following is true? **Mark all that apply.**

☒ $g(x) \in \mathcal{O}(x^2 \log_3 x)$ ☐ $g(x) \in \Theta(x^2 \log_3 x)$ ☐ $g(x) \in \Omega(x^2 \log_3 x)$

Solution: This is similar to question 10, only here $g(x)$ has a *faster* factor, $2x$. So then $g(x) \in \mathcal{O}(x^2 \log_3 x)$ (it's faster) and $g(x) \notin \Omega(x^2 \log_3 x) \Rightarrow g(x) \in \Theta(x^2 \log_3 x)$.

- 12 Using the provided pseudo-code, find the worst case performance in Big- \mathcal{O} notation.

Algorithm 5 Some more messy pseudocode

```
1: procedure SOMEPROCEDURE2
2:   j=0
3:   for i=0; i<n; i++ do
4:     j=i+j
```

- A. $\mathcal{O}(\log n)$
 B. $\mathcal{O}(n \log n)$
 C. $\mathcal{O}(n^2)$
 D. $\mathcal{O}(n)$

Solution: $\mathcal{O}(n)$

The for loop in Line 3 will iterate n times ($0, 1, 2, \dots, n-1$) so line 4 runs n times.

- 13 Using the provided pseudo-code, find the worst case performance in Big- \mathcal{O} notation.

Algorithm 6 Some messy pseudo-code

```
1: procedure SOMEPROCEDURE
2:   for i=1 and i<=n do
3:     j=1
4:     while j<n do
5:       j=j+2
```

- A. $\mathcal{O}(\log n)$
- B. $\mathcal{O}(n \log n)$
- C. $\mathcal{O}(n^2)$
- D. $\mathcal{O}(n)$

Solution: $\mathcal{O}(n^2)$

The for loop in Line 2 will iterate n times and the while loop will iterate (about) $\frac{n}{2}$ times. Since they are nested, we multiply them together and Line 5 will run $n \cdot \frac{n}{2} = \frac{n^2}{2}$ times.

- 14 Using the provided pseudo-code, find the worst case performance in Big- \mathcal{O} notation. Assume we know that **someMethod(n)** is $\mathcal{O}(\log n)$.

Algorithm 7 Some pseudocode with a method in it

```
1: procedure SOMEPROCEDURE3
2:   j=0
3:   for i=0; i<n; i++ do
4:     j=someMethod(n)
```

- A. $\mathcal{O}(\log n)$
- B. $\mathcal{O}(n \log n)$
- C. $\mathcal{O}(n^2)$
- D. $\mathcal{O}(n)$

Solution: $\mathcal{O}(n \log n)$

The for loop in line 3 will run n times. Line 4 calls **someMethod(n)** which runs in $\mathcal{O}(\log n)$ time. Since line 4 will run n times, we end up with $n \log n$.

- 15 Using the provided pseudo-code, find the worst case performance in Big- \mathcal{O} notation.

Algorithm 8 Some more messy pseudocode

```
1: procedure SOMEPROCEDURE4
2:   while n > 1 do
3:     n = n/2
```

- A. $\mathcal{O}(\log n)$
- B. $\mathcal{O}(n \log n)$
- C. $\mathcal{O}(n^2)$
- D. $\mathcal{O}(n)$

Solution: $\mathcal{O}(\log n)$

Each iteration of the `while` loop reduces n by half, so we want to know how many times can we divide n by 2 and stay above 1. If $n = 2^m$, then the power m is approximately the value we're seeking. To determine m , take \log_2 on both sides giving, $\log_2(n) = \log_2(2^m) = m \cdot \log_2(2) = m$. Since this `while` loop will run $m = \log_2(n)$ times, we get $\mathcal{O}(\log n)$. Recall that the base of the logarithm does not affect the Big- \mathcal{O} representation.

- 16 Using the provided pseudo-code, find the worst case performance in Big- \mathcal{O} notation.

Algorithm 9

```
1: procedure
2:   a = 1
3:   c = 0
4:   while a < n do
5:     for i = 0; i < n; i++ do
6:       c = c + 1
7:     a = a * 3
```

- A. $\mathcal{O}(\log n)$
- B. $\mathcal{O}(n \log n)$
- C. $\mathcal{O}(n^2)$
- D. $\mathcal{O}(n)$

Solution: $\mathcal{O}(n \log n)$

The index, a , in the `while` loop is multiplied by 3 each iteration causing it to run about $\log_3 n$ iterations. The `for` loop iterates n times for each iteration of the `while` loop. With the nested loops, we multiply to get $\mathcal{O}(n \log n)$