

## Task-1

R-3.2

Firstly, we need to find an intersection point.  
(For example in base 2 in logarithmic function)

$$8n \log_2 n = 2n^2$$

$$\hookrightarrow \log_2 n = n$$

For all  $n \geq n_0 = 16$ ,  $8n \log_2 n \leq 2n^2$

So,  $A = O(B)$  implying that A is better in time consuming.

R-3.9

$$d(n) = O(f(n)) \Rightarrow \text{for all } n \geq n_0 \quad d(n) \leq f(n)$$

By multiplying both sides by a constant  $a > 0$ ,  
we get  $a \cdot d(n) \leq a \cdot f(n)$

By ignoring  $a$ , we have  $O(a \cdot f(n)) = O(f(n))$ . Then,

$$a \cdot d(n) = O(a \cdot f(n)) = O(f(n)) \Rightarrow a \cdot d(n) = O(f(n))$$

R-3.17

Take  $n_0 = 1$  and  $k = 33$ . Then, for all  $n \geq n_0 = 1$

$$(n+1)^5 \leq 33n^5. \text{ This means } (n+1)^5 = O(n^5)$$

Also note that we can ignore all terms except  $n^5$  in

$$(n+1)^5 = n^5 + 5n^4 + 10n^3 + 10n^2 + 5n + 1. \text{ This means } O(n^5) = O((n+1)^5)$$

### R-3.18

Take  $k=3$ . For all  $n \geq n_0$

since  $2^{n+1} = 2 \cdot 2^n \leq 3 \cdot 2^n$ , we have  $2^{n+1} = O(2^n)$

### R-3.20

To find sufficiently large  $n \geq n_0$  and constant  $k$

$$n^2 = n \cdot \log n$$

$$n = \log n$$

For example in base 2:

$$n = k \cdot \log_2 n$$

Taking  $n_0 = 4$  and  $k=2$ , we have  $4 = 2 \cdot \log_2 4$

For all  $n \geq n_0 = 2$ ,  $n \geq 2 \log_2 n$  ( $n=8 \Rightarrow 8 > 2 \log_2 8$ )

$$n^2 = \Omega(n \log n)$$

## Task-2

### R-3.15

( $\Rightarrow$ ) Assume that  $f(n) = O(g(n))$ . This means that

for all  $n \geq n_0$   $f(n) \leq k \cdot g(n)$  for some  $k > 0$

for all  $n \geq n_0$   $g(n) \geq \frac{1}{k} \cdot f(n)$  note that  $\frac{1}{k} > 0$

This means  $g(n) = \Omega(f(n))$

( $\Leftarrow$ ) Assume that  $g(n) = \Omega(f(n))$ . This means that

for all  $n \geq n_0$   $g(n) \geq k \cdot f(n)$  for some  $k > 0$

for all  $n \geq n_0$   $f(n) \leq \frac{1}{k} \cdot g(n)$  note that  $\frac{1}{k} > 0$

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

## Task-3

R-3.25

def example3(s):

n = len(s)

total = 0

for j in range(n):

for k in range(1+j):

total += s[k]

return total

cost:

$c_1$

$c_2$

$c_3$

$c_4 \cdot n$

$c_5 \cdot \frac{n(n+1)}{2}$

$c_6$

Let the time consuming function in terms of  $n$  be  $f(n)$

$$f(n) = c_5 \frac{n^2}{2} + (2c_4 + c_5) \frac{n}{2} + c_1 + c_2 + c_3$$

By ignoring all terms except  $n^2$

$$O(f(n)) = O\left(\frac{c_5}{2} n^2\right) = O(n^2) \text{ note that } \frac{c_5}{2} \text{ is also ignored}$$

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

R-3.27

def example5(A, B):

n = len(A)

count = 0

for i in range(n):

total = 0

for j in range(n):

for k in range(1+j):

total += k

if B[i] == total:

count += 1

return count

cost:

$c_1$

$c_2$

$c_3$

$c_4 n$

$c_5 n$

$c_6 n^2$

$c_7 n \left(\frac{n(n+1)}{2}\right)$

$c_8 n$

Time consuming in terms of  $n$ :

$$f(n) = \frac{c_7}{2} (n^3 + n^2) + c_6 n^2 + (c_4 + c_5 + c_8) n + c_1 + c_2 + c_3$$

Ignoring all terms except  $n^3$  and ignoring  $\frac{c_7}{2}$ ,

$$\text{we have } O(f(n)) = O(n^3)$$

Time consuming is associated with  $n^3 \Rightarrow \boxed{O(n^3)}$

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#### Task-4

R-3.33

Claim:  $O(n \log n)$  is always faster than  $O(n^2)$  for sufficiently large  $n \geq n_0 = 100$ .

To check this:

When  $n < 100$ , claim is false:

Assume that  $O(n^2) = k n^2$  for some  $k > 0$

Take  $n = 10^2$

$$10^2 \log 10^2 = k 10^4 \Rightarrow k = \frac{1}{50}$$

When Bob's time method is  $\frac{1}{50} n^2$

$$10 \log 10 > \frac{1}{50} 10 \quad (n = 10)$$

Bob's method runs faster, so the claim is false.

When  $n > 100$ , claim is true:

Take  $n = 10^3$

$$10^3 \log 10^3 < \frac{1}{50} 10^6$$

Al's method runs faster, so the claim is true.