Algorithms and Data Structures 1 (Assignment 1)

1.

a.

```
a) # n > 0

for i in range(n*n):#O(n²)
    # Statements
    for j in range(i, n): # Partially runs hence it becomes a constant
        # Statements
# Statements
# Statements
```

This would be $O(n^2)$ because the outer loop dominates the inner loop. This is because the inner_loop stops at i == n, because having for example range(5,5) is not possible hence it stops if i == n. Even if i == n stop running the outer loop will keep on running until n * n is full filled. Since the outer loop dominates over the inner loop, we will take the time complexity of the outer loop making it $O(n^2)$.

b.

```
b) # n, k, i, j > 0

for i in range(n): #O(n)
  for j in range(n): #O(n)
    k = n
    # Statements
  while k > 1: #O(log_2(n))
    k = k / 2
    # Statements
```

This would be $O(n^2 * log_2(n))$ because the outer for loop is O(n) then there is a nested for loop hence making it $O(n^2)$ because we do O(n) * O(n) then the while loop has a complexity of $O(log_2(n))$. Since the while loop is inside the for loop, we have to multiply it making it $O(n^2 * log_2(n))$.

C.

```
c) # m, n > 0, n < m < n^2
i = 1
while i < n: \#O(log_10(n))
    i = 1
    while j < n : \#O(n)
         # Statements
        j += 1
    k = m
    while k > m: #Doesn't run hence ignored
         # Statements
        k = 1
    i *= 10
i = 1
while i < n: \#O(n)
    # Statements
    i += 10
```

The overall complexity of the code is $O(log_{10}(n)*(n)+n)$, the first while loop is $O(log_{10}(n))$ because at the "i" will be multiplied by ten hence the log and since it runs up until n it is $O(log_{10}(n))$. The second while loop is inside the first while loop hence we multiply the time complexity with $O(log_{10}(n))$. The second while loop is O(n) because it j starts with 1 and always adds 1 hence it will go for the full n hence it is O(n). Since k=m, there will never be a case where k>m hence that condition will never be met which means that the while loop will never run. The last while loop is O(n) however since the first loop has a worse complexity, we will use the first while loop time complexity as the worse case scenario hence the time complexity is $O(log_{10}(n)*(n))$

d.

```
d) # a, b, c > 0
if a < b and b < c:
    for i in range(a): \#O(a)
        # Statements
    if c < a:
        for j in range(c): #Doesn't run hence ignored
            # Statements
    else:
        for k in range(b): \#O(b)
            # Statements
elif a > b and b > c:
    for i in range(c, b): \#O(b)
        # Statements
else:
    for i in range(a, a + 5): \#O(5)
        # Statements
```

The time complexity would be O(b). The if statement has two notable time complexities the for i-loop and the for k-loop. The k-loop would run more times because of the the if statement condition. The condition states that a < b hence the k-loop will have a higher range causing it to run more times. The elif statement also has a O(b) time complexity however it won't exactly be O(b) because it needs to take into account "c" however it can still be considered as O(b). Lastly the else statement has a time complexity of O(5) since it will always run 5 times no matter what because of the a + 5. This means due to the if statement condition a < b, the worse time complexity it can run is O(b).

2.

a.

$$T(1) = 1$$

 $T(n) = 100T(n/10) + n^{2}$

T (n) = 100T (
$$\frac{a}{10}$$
)+ $\frac{a}{n^2}$

T (1) = 1

Build Solution

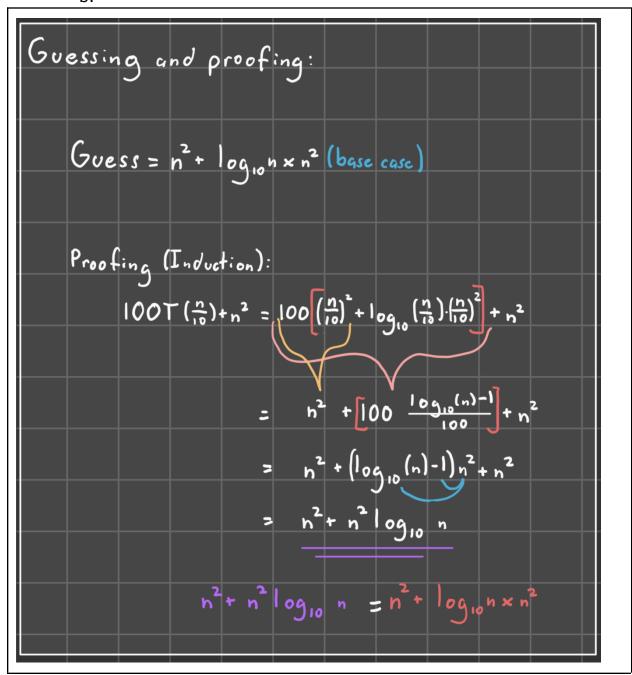
T (n) = 100T ($\frac{a}{10}$)+ $\frac{a}{n^2}$

= 100 100T ($\frac{a}{10}$)+ $\frac{a}{100}$ + $\frac{a}{100}$

= 100 100T ($\frac{a}{10}$)+ $\frac{a}{100}$ + $\frac{a}{100}$

= 100 100T ($\frac{a}{10}$)+ $\frac{a}{100}$ + $\frac{a}{10$

h.



3.

a.

$$T(n) = 8T(n/2) + n^{3}$$

$$T(n) = 8T(\frac{n}{3}) + n^{3}$$

$$\log_{2} = 3$$

$$\int_{-\infty}^{\infty} f(n) \text{ has the same growth rate as } n^{\log_{2} n}$$

$$\text{Hence } \text{ Case } 2:$$

$$\therefore \text{ Time complexity } = \theta(n^{3} \log n)$$

b.

$$T(n) = T(n/2) + n * log n$$

$$T(n)=|T(\frac{\pi}{2})+n'|\log n$$

$$|\log_{2}|=0$$

$$|\log_{2}|=0$$

$$|\log_{2}|=n'$$

$$f(n) \text{ grows}$$

$$f(n)=|\Omega_{2}(n^{0+\epsilon})|=|\epsilon|, |\epsilon| > 0$$

$$f(n) \text{ grows at least as fast as } n^{0+\epsilon} \text{ if constant is } 1$$

$$Additional \text{ Proof: } f(\frac{\pi}{2})^{\epsilon} = \frac{n}{2} = \frac{n}{2} = fn \cdot \frac{1}{2} - 2 = \frac{1}{2} < 1$$

$$\text{Hence } (\text{ase } 3:$$

$$\therefore \text{ Time complexity } = \Theta(n'|\log_{2}n)$$

c.

$$T(n) = 3T(\frac{n}{3}) + \log n$$

$$T(n) = 3T(\frac{n}{3}) + \log n$$

$$\log_{2} = 1 \quad f(n) = \log n$$

$$f(n) \text{ grows slower than } \log_{2} = 1$$

$$\text{Hence (ase 1:}$$

$$\therefore \text{Time Complexity} = \Theta(n^{\log_{2} 3})$$

$$\Theta(n)$$