WIA2005 Algorithm Design & Analysis Semester 2 Tutorial 1

1. The following is an insertion sort algorithm.

```
def InsertionSort(A):
1    for j in range(1,len(A),1):
2    key = A[j]
3     # insert A[j] into the sorted sequence A[l..j-l]
4    i=j-1
5    while (i>=0 and A[i]> key):
6     A[i+1]=A[i]
7    i=i-1
8    A[i+1]= key
```

Illustrate the insertion sort operation on array A = 41, 51, 69, 36, 51, 68.

41	51	69	36	51	68
41	51	69	36	51	68
36	41	51	69	51	68
36	41	51	51	69	68
36	41	51	51	68	69

j	i	key	A[i]	A[0]	A[1]	A[2]	A[3]	A[4]	A[5]
1	0	51	41	41	51	69	36	51	68
2	1	69	51	41	51	69	36	51	68
3	2	36	69	41	51	69	69	51	68
3	1	36	51	41	51	51	69	51	68
3	0	36	41	41	41	51	69	51	68
3	-1	36	-	36	41	51	69	51	68
4	3	51	69	36	41	51	69	69	68
4	2	51	51	36	41	51	51	69	68
5	4	68	69	36	41	51	51	69	69
5	3	68	51	36	41	51	51	68	69

2. Modify the insertion sort algorithm to sort array into decreasing order.

```
idef InsertionSort(A):
    for j in range (1, len(A), 1):
        key = A[j]
        i = j-1
        while(i>=0 and A[i] < key):
              A[i+1] = A[i]
              i = i-1
              A[i+1] = key</pre>
```

 $3. \ \ Write a pseudocode for linear search for the following requirement:$

Input: A sequence of *n* numbers $A = \langle a_1, a_2, ..., a_n \rangle$ and a value *v*. **Output:** An index *i* such that v = A[i] or the special value NIL if *v* does not appear in A.

Prompt input value v

4. Express the function n^3 / $1000 - 100n^2 - 100n + 3$ in terms of Θ -notation.

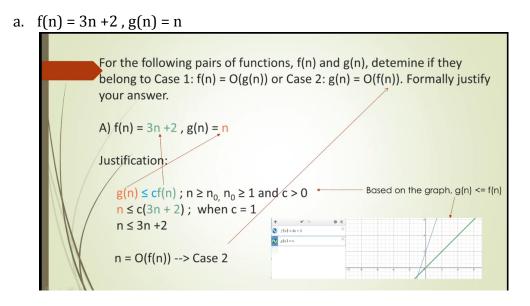
 $\Theta(n^3)$

Asymptotic complexity only cares about the fastest growing term. This is because it analyses what the function approaches as it tends towards infinity. In this case, n^3

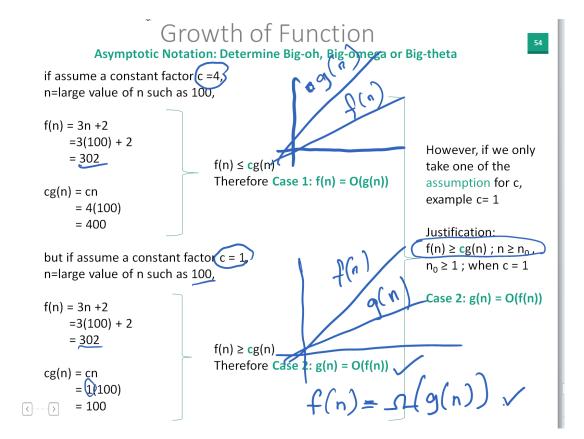
5. For the following pairs of functions, f(n) and g(n), determine if they belong to Case 1: f(n) = O(g(n)) or Case 2: g(n) = O(f(n)). Formally justify your answer.

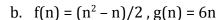
O(f(n)) represents the set of all functions upper bounded by asymptotic f(n). e.g. O(n) contains all functions upper bounded by asymptotic n. $O(e^n)$ contains all functions upper bounded by asymptotic e^n . It is more formally defined as:

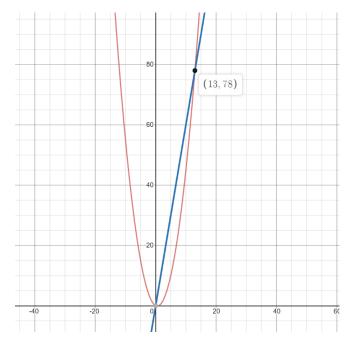
 $O(g(n)) = \{f(n) \mid 0 < f(n) \le c*g(n), \text{ and } c \text{ is a constant real number} \}$ $f(n) = O(g(n)) \text{ in asymptotic notation represents } f(n) \subseteq O(g(n)) \}$ https://www.youtube.com/watch?v=whit N9uYFI



Case 2, both functions have same order term but coefficient of f(n) is higher than g(n)







The graph shows c = 1

(since f(n) intersect with g(n) at 2 points (0,0) and (13,78), for $n_{\scriptscriptstyle 0}$ we will take the largest intersection which is (13,78). So given , $n \ge n_{\scriptscriptstyle 0}$, $n \ge 1$ and c>0, which in this case $n_{\scriptscriptstyle 0}$ =13 and c = 1, we have

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proven that g(n) will always be upper bounded by f(n). Hence, it is
case 2, g(n) = O(f(n)).)
Proof:
Case 2: g(n) = O(f(n))
f(n) \ge cg(n), n \ge no, n \ge 1 when c = 1
Case 1: False, Case 2: True
f is in O(n^2) while g is in O(n).
Functions that grow quadratically are not in the set O(n).
Case 1 is thus false
O(n^2) is a superset of O(n). g(n) is in the set O(n).
It must also be in O(n^2)
Case 2 is true
6n \le c(n^2 - n)/2
12n \le c(n^2 - n)
12 \le c(n-1)
12 <= cn-c
c >= 12/(n-1)
As n approaches infinity, right term approaches 0
c will always win at some point
n!=1
n > 1
6cn \le (n^2 - n)/2
12cn \le n^2 - n
12c \le n - 1
c \le (n-1)/12
As n approaches infinity, right term approaches infinity
c can never win. sad nia
n >= 1
f(n) = (n^2 - n)/2, g(n) = 6n
Assume n = 500, c = 3
f(n) = (500^2 - 500) / 2 > 3 * 6 (500)
Case 2
f(n) = 124750, g(n) = 3000
Assume n = 500, c = 5
12470 \le (5)(3000)
Case 1
```

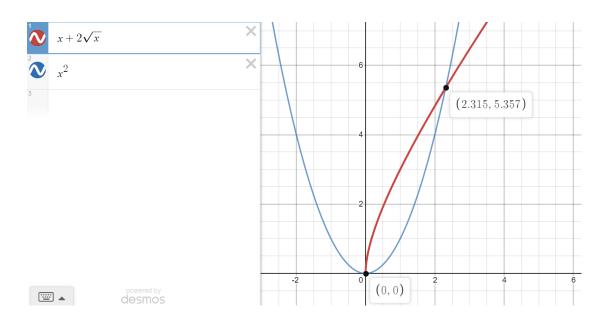
Case 2, because f(n) has higher order term compared to g(n)

c.
$$f(n) = n+2\sqrt{n}$$
, $g(n) = n^2$
Case 1: True, Case 2: False
f is in $O(n)$ while g is in $O(n^2)$
This is a reflection case of $S(b)$

Case 1, because g(n) has higher order term compared to f(n)

let x be n

$$\lim_{x \to \infty} \left(\frac{x^2}{x + 2\sqrt{x}} \right) = \infty$$
 , as x tends to infinity,



```
case 1: f(n) \le cg(n); n \ge n0, and c > 0 n+2\sqrt{n} \le cn^2 let choose C=10 n+2\sqrt{n} \le 10n^2 so n^2 greater than n, f(n)=O(g(n)) so case 1 true case 2: cf(n) \ge g(n); n \ge n0, and c > 0 c(n+2\sqrt{n}) \ge n^2 let choose C=10 10n+20\sqrt{n} \ge 10n^2
```

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so n<sup>2</sup> greater than n, case 2 is false
   Case 1: f(n) = O(g(n))
   f(n) \le cg(n), n \ge no, n \ge 1 when c = 3
d. f(n) = n^2 + 3n + 4, g(n) = n^3
   Case 1: True, Case 2: False
   f is in O(n^2) while g is in O(n^3)
   O(n^3) is a superset of O(n^2). g(n) is in the set O(n^2).
   It must also be in O(n^3)
   Case 1 is true
   Functions that grow cubically are not in the set O(n^2).
   Case 2 is thus false
   Case 1, because g(n) has higher order term compared to f(n)
   case 1:
   f(n) \le cg(n); n \ge n0, and c > 0
   n^2 + 3n + 4 \le c n^3
   let choose C=10
   n^2 + 3n + 4 \le 10n^3
   so n<sup>3</sup> greater than n<sup>2</sup>,
   f(n)=O(g(n))
   so case 1 true
   case 2:
   c f(n) \ge g(n); n \ge n0, and c > 0
    c(n^2 + 3n + 4) >= n^3
   let choose C=10
   10n^2 + 30n + 40 >= 10n^3
   so n<sup>3</sup> greater than n<sup>2</sup>,
    so case 2 is false
   Case 1: f(n) = O(g(n))
   f(n) \le cg(n), n \ge no, n \ge 1 when c = 8
```

6. Given the iterative function below (in Java), calculate their time complexity.

```
a. function1 (){
        for (int i = 1; i <= n; i ++) {
             printf("Hello world");
  }
  O(n)
b. function2(){
        for (int i = 1; i <= n; i ++) {
                  for (int j = 1; j <=n; j ++) {
                       printf("Hello world");
         }
     }
  O(n^2)
  n*n
c. function3 (){
        for (int i = 1; i^2 <= n; i ++) {
             printf("Hello world");
  O(\sqrt{n}) = O(n^{0.5})
  you can also move the square term to the
```

other side and get i $\leq \sqrt{n}$.

My another proving style

K	į	Generalise K
1	1	$\sqrt{1}$
2	4	$\sqrt{4}$
3	9	√9
4	16	√16
5	25	$\sqrt{25}$
6	36	√36
k	n	√n

```
Let said n = 36
   When i = 6, n = 36
   K = \sqrt{n}
   Answer: O(\sqrt{n})
d. function4 (){
        for (int i = 1; i <= n; i = i*2) {
              printf("Hello world");
        }
   }
  O(logn)
  the loop actually doing 2°
  where c is the number of iteration
  2^{c} \le n
  c \le \log_2 n
e. function3(){
        for (int i = n/2; i <=n; i ++) {
              for (int j <= 1; j <=n/2; j = 2*j) {
                    for (int k = 1; k <= n; k = k*2) {
                         printf("Hello world");
                    }
```

```
}
}
O(nlog²n) = O(n[logn]²)
```