

Problem 1.1

```

x = 0; // O(1)
y = 0; // O(1)
for (i=1; i<n; i++)
    ... // O(n-1)
return (x - y); // O(1)

```

In this loop, what's shown is the value i will go from 1 to n . Where n is the size of the array, the value n will increase linearly. Therefore, analyzing the running time complexity will be $O(n)$.

Problem 1.2

```

x = 0;
i = 0;
while (i < n) {
    y = 0;
    j = 0;
    while (j < n) {
        k = 0
        while (k <= j) {
            y = y + a[k];
            k = k + 1;
        }
        j = j + 1;
    }
    if (b[i] == y) {
        x++;
    } i = i + 1;
}
return x;

```

There are three while-loops, and they can be described as a nested loop.

The outer first while loop, the value of i will run from 0 to $n-1$ so, it will run for n times.

The inner second while loop will also run for n times but for every iteration of the first loop. So, for each value of i , the value of j will move from 0 to $n-1$.

The inner third while loop will for j times for each iteration of the second loop. For each value of j , the value k moves from 0 to j depending on the value of j .

The third loop will run for 1 time when $j=0$, 2 times when $j=1$, 3 times when $j=2$, ..., n times when $j=n$. It will run for: $(1+2+3+4+5+\dots+n)$ times = $n*(n+1)/2$ times.

Being that the first loop will run for n iteration, the second and third loop will combinedly run for $n*(n+1)/2$. Therefore, the analyzed running time complexity will be $n*n*(n+1)/2$ which is $O(n^3)$.

Problem 1.3

```

if (i == 0) { //O(1)
    p[0] = a[0];
    p[1] = a[0];
}
else {
    method3(a, i-1, p); //O(n-1)
    if (a[i] < p[0]) { //O(1)
        p[0] = a[i];
    }
    if (a[i] > p[1]) { //O(1)
        p[1] = a[i];
    }
}
}

```

The starting value of i is $n-1$. With the new i value as $i-1$, the recursive calls are being made. This function is being called recursively n times before reaching the base case so the analyzing running time complexity will be $O(n)$.

Problem 1.4

```

if (x >= y) { //O(1)
    return a[x];
}
else {
    z = (x + y) / 2; //O(1)
    u = method4(a, x, z); //O(n)
    v = method4(a, z+1, y); //O(n)
    if (u < v) return u; //O(1)
    else return v;
}
}

```

The analyzed running time complexity will be $O(n)$ because each node will be split into 2 nodes also known as the child nodes. So, it will be equal to the number of the internal nodes ($2n-1$) as a function will be called for $2n-1$ times. Therefore, using the Master theorem, $T(n) = a.T(n/b) + O(n^d)$ where $a=2$, $b=2$, $d=0$ as $a > b^d$ there by making the running time complexity $O(n^{\log_2 2})$ which is $O(n)$.

Problem 2.1

Operation	Return Value	Stack Contents
push(10)	-	(10)
pop()	10	()
push(12)	-	(12)
push(20)	-	(12,20)
size()	2	(12,20)
push(7)	-	(12,20,7)
pop()	7	(12,20)
top()	20	(12,20)
pop()	20	(12)
pop()	12	()
push(35)	-	(35)
isEmpty()	false	(35)

Problem 2.2

Operation	Return Value	Queue Contents (first \leftarrow Q \leftarrow last)
enqueue(7)	-	(7)
dequeue()	7	()
enqueue(15)	-	(15)
enqueue(3)	-	(15,3)
first()	15	(15,3)
dequeue()	3	(15)
dequeue()	15	()
first()	null	()
enqueue(11)	-	(11)
dequeue()	11	()
isEmpty()	true	()
enqueue(5)	-	(5)