


Algorithmics	Student information	Date	Number of session
	UO: 283928	03/14	Session 3
	Surname: Suárez Losada	 Escuela de Ingeniería Informática Universidad de Oviedo	
	Name: Gonzalo		



## Activity 1. Basic recursive models

### Complexities of every class

#### Division1.java

This method uses the Divide and Conquer strategy by division as every class called so. Then, the idea for getting the complexities is similar among them all so it will not be repeated.

There are three variables in the complexity of such a class:

- a: amount of recursive calls
- b: denominator in the recursive call
- k: overall complexity of the method without recursive calls

Then we must check the proportion of a with respect to  $b^k$ :

- If  $a < b^k$ :  $O(n^k)$
- If  $a = b^k$ :  $O(n^k \cdot \log n)$
- If  $a > b^k$ :  $O(n^{\log_b a})$

For this example:

```
{
    long cont = 0;
    if (n <= 0) cont++;
    else
    { for (int i=1; i<n; i++) cont++ ; //O(n)
      rec1(n/3);
    }
    return cont;
}
```

- a = 1 recursive call
- b = 3
- k = 1 (linear complexity)

Then, as  $1 < 3^1$ , the complexity will be linear  $O(n^1)$

#### Division2.java

```
public class Division2 {
    public static long rec2 (int n)
    {
        long cont = 0;
        if (n <= 0) cont++;
        else
        { for (int i=1; i<n; i++) cont++ ;
          rec2(n/2);
          rec2(n/2);
        }
        return cont;
    }
}
```

- a = 2 recursive call
- b = 2
- k = 1 (linear complexity)
- As  $2 = 2^1$ , complexity will be  $O(n \cdot \log n)$

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### Division3.java

```
public static long rec3 (int n)
{
    long cont = 0;
    if (n<=0) cont++;
    else
    {
        cont++ ; // 0(1)
        rec3(n/2);
        rec3(n/2);
    }
    return cont;
}
```

- a = 2 recursive call
- b = 2
- k = 0 (constant complexity)
- As  $2 > 2^0$ , complexity will be  $O(n^{\log_2(2)})$

### Division4.java

```
public static long rec4 (int n)
{
    long cont = 0;
    if (n<=0) cont++;
    else
    {
        for (int i=1;i<n;i++) cont++ ; //0(n)
        for ( int j = 1; j < n ; j++ ) cont++;
        rec4(n/2);
        rec4(n/2);
        rec4(n/2);
        rec4(n/2);
    }
    return cont;
}
```

- a = 4 recursive calls
- b = 2
- k = 1 (linear complexity)
- As  $4 < 2^1$  complexity will be  $O(n^{\log_2(4)}) = O(n^2)$

### fib1()

```
/* Iterative solution with temporal complexity O(n) */
public static int fib1(int n) {
    int n1 = 0;
    int n2 = 1;
    for (int i = 1; i <= n; i++) {
        int s= n1+n2;
        n1 = n2;
        n2 = s;
    }
    return n1;
}
```

In this case, no recursion is used. In fact, as stated, an iterative linear complexity approach is used.

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### fib2()

```
/* Second iterative solution with time complexity O(n)
 * using a vector. This algorithm will be viewed later
 * as a very simple case dynamic programming */
public static int fib2(int n, int[]v) {
    v[0] = 0;
    v[1] = 1;
    for (int i=2; i <= n; i++)
        v[i]=v[i-1]+v[i-2];
    return v[n];
}
```

Once again, this method uses an iterative linear complexity approach

From now on, the remaining methods will be implemented by a Divide and Conquer strategy based on subtraction. The concept is the same but modifying the complexity formulas:

- If  $a < 1$ :  $O(n^k)$
- If  $a = 1$ :  $O(n^{k+1})$
- If  $a > 1$ :  $O(a^{n/b})$

### fib3()

```
/* First recursive solution: T(n)=T(n-1)+O(1)
 * => O(n) */
public static int fib3(int n) {
    return aux(0, 1, n);
}

private static int aux(int n1, int n2, int n) {
    if (n < 1) return n1;
    return aux(n2, n1+n2, n-1);
}
```

In this case, we are told a linear complexity is achieved. Let us check so:

- $a = 1$  recursive call
- $b = 1$
- $k = 0$

Then, as  $1 == 1$ : complexity will be  $O(n^{0+1}) = O(n^1)$ . Then, it was correct.

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### fib4()

```

/* Second recursive solution: T(n)=T(n-1)+T(n-2)+O(1),
 * that is an exponential solution  $O(1.6^n)$ , that is,
 * each time n grows from n to n+1, the time is multiplied
 * by 1.6. Or what is the same, the amount of time it takes
 * for n is the sum of times for the preceding sizes
 * (n-1) and (n-2) */
public static int fib4(int n) {
    if (n<=1)
        return n;
    return fib4(n-1) + fib4(n-2);
}

```

By applying the D&C strategy by subtraction, we get:

- $a = 2$
- $b = 1$  or  $2$
- $k = 0$

So the final complexity will be somewhere in between  $O(2^n)$  and  $O(2^{n/2})$ .

### Subtraction1.java

```

public static long rec1(int n){
    long cont = 0;
    if (n<=0) cont++;
    else
    { cont++; //  $O(1)=O(n^0)$ 
      rec1(n-1);
    }
    return cont;
}

```

- $a = 1$  recursive call
- $b = 1$
- $k = 0$

Then, as  $1 == 1$ : complexity will be  $O(n^{0+1}) = O(n^1)$ .

### Subtraction2.java

```

public static long rec2(int n)
{
    long cont = 0;
    if (n<=0) cont++;
    else
    { for (int i=0;i<n;i++) cont++; //  $O(n)$ 
      rec2(n-1);
    }
    return cont;
}

```

- $a = 1$  recursive call
- $b = 1$
- $k = 1$

Then, complexity is  $O(n^{1+1}) = O(n^2)$

### Subtraction3.java

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```
public static long rec3(int n)
{
    long count = 0;
    if (n<=0) count++;
    else
    {
        count++; //O(1)
        rec3(n-1);
        rec3(n-1);
    }
    return count;
}
```

- a = 2 recursive calls
- b = 1
- k = 0

Then, the complexity will be  $O(a^{n/b}) = O(2^n)$ , which is exponential

#### Subtraction4.java

```
public static long rec4(int n){
    long count = 0;
    if (n<=0) count++;
    else
    {
        count++; // O(1)=O(n^0)
        rec4(n-2);
        rec4(n-2);
        rec4(n-2);
    }
    return count;
}
```

- a = 3 recursive calls
- b = 2
- k = 0

Then, the complexity will be  $O(a^{n/b}) = O(3^{n/2})$ , which is exponential and the one we are asked for

#### VectorSum1

```
/* This method iteratively calculates the sum of
 * the vector with linear complexity O(n) */
public static int sum1(int[]a) {
    int n= a.length;
    int s=0;
    for(int i=0;i<n;i++)
        s=s+a[i];
    return s;
}
```

As stated, the method has an iterative linear complexity approach

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## VectorSum2

```

/* This method recursively calculates the sum of
 * the vector with a complexity O(n) */
public static int sum2(int[]a) {
    return recSust(0,a);
}

private static int recSust(int i, int[]a) {
    if (i==a.length)
        return 0;
    else
        return a[i] + recSust(i+1,a);
}

```

• a = 1 recursive call

• b = 1

• k = 0

Then, as  $1 == 1$ : complexity will be  $O(n^{0+1}) = O(n^1)$ .

## VectorSum3

```

/* This method recursively calculates the sum of
 * the vector with a complexity O(n) */
public static int sum3(int[]a) {
    return recDiv(0,a.length-1,a);
}

private static int recDiv(int iz,int de,int[]a) {
    if (iz==de)
        return a[iz];
    else {
        int m= (iz+de)/2;
        return recDiv(iz,m,a)+ recDiv(m+1,de,a);
    }
}

```

• a = 2 recursive calls

• b = 2

• k = 0

Then, as  $2 > 2^0$ : complexity will be  $O(n^{\log_2 2}) = O(n)$