# **Mathemathical Formulas**

1. 
$$\sum_{i=1}^{n} 1 = n$$

2. 
$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2}$$

3. 
$$\sum_{i=1}^{n} i^2 = \frac{n(n+1)(2n+1)}{6}$$

4. 
$$\sum_{i=1}^{n} i^3 = (\sum_{i=1}^{n} i)^2$$

5. 
$$\sum_{i=0}^{k} 2^i = 2^{k+1} - 1$$

6. 
$$\sum_{i=0}^{k} \frac{1}{2^i} = 2 \text{ when } k \to +\infty$$

7. 
$$\sum_{i=0}^{n} x^i = \frac{x^{n+1}-1}{x-1}$$
 with  $x \neq 1$ 

8. 
$$\sum_{k=1}^{n} \frac{1}{k} = 1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \sim \ln(n)$$

9. 
$$\log(n!) \sim n \log(n)$$

10. 
$$\lim_{n \to +\infty} \sum_{k=0}^{n} x^k = \frac{1}{1-x}$$
 if  $|x| < 1$ 

11. 
$$\sum_{i=0}^{n} ix^i = \frac{1}{(1-x)^2}$$
 if  $|x| < 1$ 

12. 
$$\log(ab) = \log(a) + \log(b)$$

13. 
$$\log(\frac{a}{b}) = \log(a) - \log(b)$$

14. 
$$\log(a^b) = b \log(a)$$

15. 
$$\log_b(a) = \frac{\ln(a)}{\ln(b)}$$

16. 
$$a^{\log_b(n)} = n^{\log_b(a)}$$

17. 
$$a^n \times a^m = a^{n+m}$$

18. 
$$(a^m)^n = a^{m*n}$$

19. 
$$a^{(m)^n} \neq (a^m)^n$$

# **How To Calculate Sums**

## 1 Nested Sum

### Calculate Nested Sum

We always start from the most inner sum until we finish calculating

### Example:

$$\sum_{i=1}^{n} \sum_{j=1}^{i} \sum_{k=1}^{j} 1 = \sum_{i=1}^{n} \sum_{j=1}^{i} j$$

$$= \sum_{i=1}^{n} \frac{i(i+1)}{2}$$

$$= \frac{1}{2} \sum_{i=1}^{n} i(i+1)$$

$$= \frac{1}{2} (\sum_{i=1}^{n} i^{2} + \sum_{i=1}^{n} i)$$

$$= \frac{1}{2} (\frac{n(n+1)(2n+1)}{6} + \frac{n(n+1)}{2})$$

$$= \frac{n(n+1)(2n+4)}{12}$$

### 2 Adjusting Boundaries To Match A Known Sum

### **Adjusting Boundaries**

We might come across a known sum where the boundaries aren't the same as the rule. All we need to do is calculate the sum with the boundaries of the rule and then add or subtract the term accordingly.

### Example:

$$\sum_{i=0}^{k-1} 2^i = \sum_{i=0}^k 2^i - \sum_{i=k}^k 2^i$$
$$= 2^{k+1} - 1 - 2^k$$

We have a sum that, by the rule, goes till the k-th term, but here it goes up to the (k-1)-th term. So, to get the sum up to (k-1), we take the sum up to k and remove the last term.

$$term(0) + term(1) + \dots + term(k-1) + term(k) = \sum_{i=0}^{k} 2^{i}$$
$$\sum_{i=0}^{k-1} 2^{i} + term(k) = \sum_{i=0}^{k} 2^{i}$$
$$\sum_{i=0}^{k-1} 2^{i} = \sum_{i=0}^{k} 2^{i} - term(k)$$

# Big Notation

### Notations

 $\bullet$  **Big O** : O upper bound

• Big Omega:  $\Omega$  lower bound

• Big Theta :  $\Theta$  average

### Limit Definition

$$\text{if} \quad \lim_{n \to +\infty} \left| \frac{f(n)}{g(n)} \right| = k \quad , \quad k > 0 \quad => \quad f(n) \in \Theta(g(n))$$

if 
$$\lim_{n \to +\infty} \left| \frac{f(n)}{g(n)} \right| = +\infty$$
  $\Longrightarrow f(n) \in \Omega(g(n))$ 

## **Enequalities Definition**

$$f(n) = \Omega(g(n))$$
  $\exists c > 0$ ,  $c.g(n) \le f(n)$  ,  $\forall n \ge n_0$ 

$$f(n) = O(g(n))$$
  $\exists c > 0$ ,  $c.g(n) \ge f(n)$  ,  $\forall n \ge n_0$ 

$$f(n) = \Theta(g(n)) \quad \exists c_1 > 0 \;,\; \exists c_2 > 0 \;, \quad c_1.g(n) \leq f(n) \leq c_2.g(n) \quad, \forall n \geq n_0$$

### Note

- $\bullet$  c is a strictly positive real number.
- $\bullet$  *n* is a strictly positive integer.

### $\mathbf{Example}$

$$\sqrt{2\log_2 n + 3 + 7n} \sim O(\sqrt{n})$$

$$\sqrt{2\log_2 n + 3 + 7n} \le \sqrt{n \cdot c}$$

$$\sqrt{\frac{2\log_2 n + 3 + 7n}{n}} \le c$$

$$\sqrt{\frac{2\log_2 n}{n} + \frac{3}{n} + 7} \le c$$

$$\frac{2\log_2 n}{n} \le 1 \qquad \Longrightarrow \quad n \ge 4$$

$$\frac{3}{n} \le 1 \qquad \Longrightarrow \quad n \ge 3$$

$$\sqrt{1+1+7} \le c$$

$$\sqrt{9} \le c \longrightarrow \boxed{c = \sqrt{9}}$$

$$2^n \sim \Omega(5^{\log_e(n)})$$

$$2^{n} \ge 5^{\log_{e}(n)} \cdot c$$
$$2^{n} \ge n^{\log_{e}(5)} \cdot c$$
$$2^{n} \ge n^{1.6} \cdot c$$

$$\frac{2^n}{n^{1.6}} \ge c$$

$$\frac{2^n}{n^{1.6}} \ge 1 \Longrightarrow n \ge 1 \longrightarrow \boxed{n_0 = 1}$$
 and  $\boxed{c = 1}$ 

$$n^{2^n} + 6 \cdot 2^n \sim \Theta(n^{2^n})$$

$$c_1 \cdot n^{2^n} \le n^{2^n} + 6 \cdot 2^n \le c_2 \cdot n^{2^n}$$
$$c_1 \le 1 + \frac{6 \cdot 2^n}{n^{2^n}} \le c_2$$
$$\frac{6 \cdot 2^n}{n^{2^n}} \le 1 \Longrightarrow n \ge 3 \longrightarrow \boxed{n_0 = 3}$$

$$c_1 \le 2 \le c_2 \longrightarrow \boxed{c_1 = 1} \text{ and } \boxed{c_2 = 2}$$

# **Induction**

## Induction's Steps

- 1. Verify for n = term
- 2. Assume true for n
- 3. Prove for n+1

### $\underline{\mathbf{Example}}$

$$a_1 = 1 \qquad a_2 = 1$$

$$a_n = a_{n-1} + a_{n-2}$$
 if  $n \ge 3$ 

Prove That : 
$$a_n = \frac{(\frac{1+\sqrt{5}}{2})^n - (\frac{1-\sqrt{5}}{2})^n}{\sqrt{5}}$$

Verify For n = 1

$$a_1 = 1$$

$$a_1 = \frac{(\frac{1+\sqrt{5}}{2})^1 - (\frac{1-\sqrt{5}}{2})^1}{\sqrt{5}} = 1$$
True For  $n = 1$ 

#### Assume True For n And Prove For n+1

$$a_{n+1} = a_n + a_{n-1} \quad \text{if } n \ge 3$$

Prove That : 
$$a_{n+1} = \frac{\left(\frac{1+\sqrt{5}}{2}\right)^{n+1} - \left(\frac{1-\sqrt{5}}{2}\right)^{n+1}}{\sqrt{5}}$$

$$a_{n+1} = a_n + a_{n-1}$$

$$= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^n - \left(\frac{1-\sqrt{5}}{2}\right)^n}{\sqrt{5}} + \frac{\left(\frac{1+\sqrt{5}}{2}\right)^{n-1} - \left(\frac{1-\sqrt{5}}{2}\right)^{n-1}}{\sqrt{5}}$$

$$= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^n \cdot \left(1 + \frac{2}{1+\sqrt{5}}\right) - \left(\frac{1-\sqrt{5}}{2}\right)^n \cdot \left(1 + \frac{2}{1-\sqrt{5}}\right)}{\sqrt{5}}$$

$$= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^n \cdot \left(\frac{(3+\sqrt{5})(1-\sqrt{5})}{(1+\sqrt{5})(1-\sqrt{5})}\right) - \left(\frac{1-\sqrt{5}}{2}\right)^n \cdot \left(\frac{(3-\sqrt{5})(1+\sqrt{5})}{(1-\sqrt{5})(1+\sqrt{5})}\right)}{\sqrt{5}}$$

$$= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^n \cdot \left(\frac{1+\sqrt{5}}{2}\right) - \left(\frac{1-\sqrt{5}}{2}\right)^n \cdot \left(\frac{1-\sqrt{5}}{2}\right)}{\sqrt{5}}$$

$$= \frac{\left(\frac{1+\sqrt{5}}{2}\right)^{n+1} - \left(\frac{1-\sqrt{5}}{2}\right)^{n+1}}{\sqrt{5}}$$

$$= a_{n+1}$$

True For n+1

Conclusion True For n

# Order Of Complexity

Order

$$1 \le \log(n) \le \sqrt{n} \le n \le n \log(n) \le n^2 \le n^3 \le 2^n \le n! \le n^n$$

# Complexity

## Some Terminology

- Frequency of Execution  $(F_r)$ : The number of times an instruction is executed.
- Execution Time of Basic Instructions ( $\Delta t$ ): We assume that all basic instructions (such as print, read, assignment, arithmetic operations, etc.) have the same execution time, denoted as  $\Delta t$ .
- Function (f(n)): Represents the total frequency of the program's instructions in relation to the data size n.
- Exact Execution Time Function (T(n)): The execution time function in relation to the data size n.
- Approximate Theoretical Complexity (Asymptotic): It approximates T(n) by omitting all constants and taking the term with the highest growth rate.

#### Example

#### Algorithm Sum of First N Integers

```
1: Var
2: n, sum, i integer;

3: Begin
4: sum ← 0;
5: i ← 1;
6: print('Input Integer N : ')
7: Read(n);
8: while i <= n do
9: sum ← sum + i;
10: i ← i + 1;
11: end while
12: print('Sum is ', sum);
13: End
```

### Time Complexity:

we have  $f(n) = \sum fr$  since f(n) is sum of frequency of execution (fr) we need to figure out the fr of each instruction and sum them:

```
sum \leftarrow 0
                                        fr = 1 (one affectation)
                                       fr = 1 (one affectation)
i \leftarrow 1
                                       fr = 1 (one print)
print('Input Integer N : ')
Read(n)
                                       fr = 1 (one read)
while i <= n do
                                       fr = n + 1 (check the while condition n+1 times)
                                       fr = 2n (one affectation and one arithmetic operation + (2) inside a while that loops n times (2n))
sum \leftarrow sum + i
i \leftarrow i + 1
                                       fr = 2n (one affectation and one arithmetic operation + (2) inside a while that loops n times (2n))
print('Sum is ',sum)
                                       fr = 1 (one print)
```

$$f(n) = \sum fr$$
= 1 + 1 + 1 + 1 + (n + 1) + 2n + 2n + 1  
= 5n + 6  
=  $5n + 6$ 

now that we have the complexity function f(n) we need to find the T(n) we have  $T(n) = f(n) \times \Delta t$ 

$$T(n) = f(n) \times \Delta t$$

$$= (5n+6) \times \Delta t$$

$$= \underbrace{5\Delta t}_{a} n + \underbrace{\Delta t 6}_{b}$$

$$= \boxed{an+b}$$

In this example the exact theoritical complexity is T(n) = an + b and its approximate theoritical complexity is  $an + b \sim O(n)$  we notice that its time complexity is linear

# Iterative Algorithm

## Iterative Algorithm

Iterative algorithm complexity is calculated by summing the frequency of all instructions.

## For Loop

- If the index is decremented or incremented by just 1:
  - The number of iterations is upper bound lower bound + 1
  - Each loop can be represented as a sum and used to calculate complexity
- If the index is decremented or incremented by a constant  $c \geq 2$ :
  - The number of iterations is  $c \cdot k = n \rightarrow k = \frac{n}{c}$
  - Cannot be represented as a sum
- If the index is multiplied or divided by a constant  $c \geq 2$ :
  - The number of iterations is  $c^k = n \to k = \log_c(n)$
  - Cannot be represented as a sum

# While Loop

- Can never be represented as a sum
- If the index is incremented or decremented by just 1, the number of iterations is the difference between the upper and lower bounds.
- If the index is incremented or decremented by a constant  $c \geq 2$ , the number of iterations is  $\frac{n}{c}$ .
- If the index is multiplied or divided by a constant  $c \geq 2$ , the number of iterations is  $\log_c(n)$ .

### Example

#### Algorithm

```
for (i = n ; i \ge 0 ; i \leftarrow i/2) do

print('Index i : ',i)

end for
```

```
Iteration: \frac{n}{2^0}, \frac{n}{2^1}, \frac{n}{2^2}, \frac{n}{2^3}, \dots, \frac{n}{2^k}
```

Loop stops when  $i < 0 \Rightarrow \frac{n}{2^k} < 0$  since it's an natural division it means :  $2^k > n \Longrightarrow k > \log_2(n) \Longrightarrow k = \lceil \log_2(n) \rceil = \boxed{\log_2(n) + 1}$ In conslusion  $O(\log(n))$ 

#### Algorithm

```
 \begin{array}{l} r \leftarrow 0; \\ \textbf{for} \ \ (i=1 \ ; \ i \leq n^2 \ ; \ i \leftarrow i+1) \ \textbf{do} \\ \textbf{for} \ \ (j=1 \ ; \ j \leq 2n-1 \ ; \ j \leftarrow j+1) \ \textbf{do} \\ r \leftarrow r+1; \\ \textbf{end for} \\ \textbf{end for} \end{array}
```

$$\sum_{i=1}^{n^2} \sum_{j=1}^{2n-1} 1 = \sum_{i=1}^{n^2} 2n - 1$$
$$= (2n-1) \cdot n^2$$
$$= 2n^3 - n^2$$

In conslusion  $O(n^3)$ 

#### Algorithm

```
\begin{array}{c} r \leftarrow 0; \\ \textbf{for} \ \ (i=1 \ ; \ i \leq n \ ; \ i \leftarrow i+1) \ \textbf{do} \\ \textbf{for} \ \ (j=i+1 \ ; \ j \leq n \ ; \ j \leftarrow j+1) \ \textbf{do} \\ \textbf{for} \ \ (k=1 \ ; \ k \leq j \ ; \ j \leftarrow j+1) \ \textbf{do} \\ r \leftarrow r+1; \\ \textbf{end for} \\ \textbf{end for} \\ \textbf{end for} \end{array}
```

$$\begin{split} \sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum_{k=1}^{j} 1 &= \sum_{i=1}^{n} \sum_{j=i+1}^{n} j \\ &= \sum_{i=1}^{n} (\sum_{j=1}^{n} j - \sum_{j=1}^{i} j) \\ &= \sum_{i=1}^{n} (\frac{n(n+1)}{2} - \frac{i(i+1)}{2}) \\ &= \frac{1}{2} (\sum_{i=1}^{n} n(n+1) - \sum_{i=1}^{n} i^{2} - \sum_{i=1}^{n} i) \\ &= \frac{1}{2} (n^{2}(n+1) - \frac{n(n+1)(2n+1)}{6} - \frac{n(n+1)}{2}) \\ &= \frac{4n^{3} - 3n^{2} - 7n}{12} \end{split}$$

In conslusion  $O(n^3)$ 

#### Algorithm

```
r\leftarrow 0; i\leftarrow 1
while (i\leq n) do
for (j=n^2\;;\;j\leq 5\;;\;j\leftarrow j-1) do
r\leftarrow r+1;
end for
i\leftarrow i+2;
end while
```

For Loop iterates :  $n^2 - 5 + 1 = \boxed{n^2 - 4}$ 

While Loop: 
$$2k > n \Rightarrow k > \frac{n}{2} \Rightarrow k = \lceil \frac{n}{2} \rceil = \boxed{\frac{n}{2} + 1}$$

In conslusion  $O((n^2 - 4)(\frac{n}{2} + 1)) = O(n^3)$ 

### Algorithm

```
\begin{array}{l} i \leftarrow 2; j \leftarrow 1 \\ \textbf{while} \quad (i \leq n) \ \textbf{do} \\ i \leftarrow i * i; \\ \textbf{end while} \\ \textbf{while} \quad (j \leq i) \ \textbf{do} \\ j \leftarrow 4 * j; \\ \textbf{end while} \\ \end{array}
```

```
First While Iteration : 2, 2^{2^1}, 2^{2^2}, 2^{2^3}, \dots, 2^{2^k}
2^{2^k} > n \Rightarrow k > \log(\log(n)) \Rightarrow k = \lceil \log(\log(n)) \rceil = \boxed{\log(\log(n)) + 1}
```

Second While Iteration : 
$$1, 4^1, 4^2, 4^3, \dots, 4^k$$
  
$$4^k > i \Rightarrow 4^k > n \Rightarrow k > \log(n) \Rightarrow k = \lceil \log(n) \rceil = \boxed{\log(n) + 1}$$

In Conclusion  $O(\log(\log(n)) + 1 + \log(n) + 1) = O(\log(n))$