

Design and Analysis of Algorithms (203105301)

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CHAPTER-1

Introduction



Outline

- Characteristics of algorithm.
- Analysis of algorithm: Asymptotic analysis of complexity bounds—best, average and worst-case behavior;
- Performance measurements of Algorithm,
- Time and space trade-offs,
- Analysis of recursive algorithms through recurrence relations:
Substitution method



Computational problems?

- **A computational problem specifies an input-output relationship**
 - What does the input look like?
 - What should the output be for each input?
- **Example:**
 - Input: an integer number n
 - Output: Is the number even?
- **Example:**
 - Input: A list of names of people
 - Output: The same list sorted alphabetically



Problems and Solution as Algorithm

- For example, we need to solve a computational problem

“Convert a distance in kilometer to meter”

An algorithm specifies how to solve it,

- I. 1. Read distance in kilometer
- II. 2. Calculate distance-in-meter = distance-in-kilometer*1000
- III. 3. Print distance-in-meter





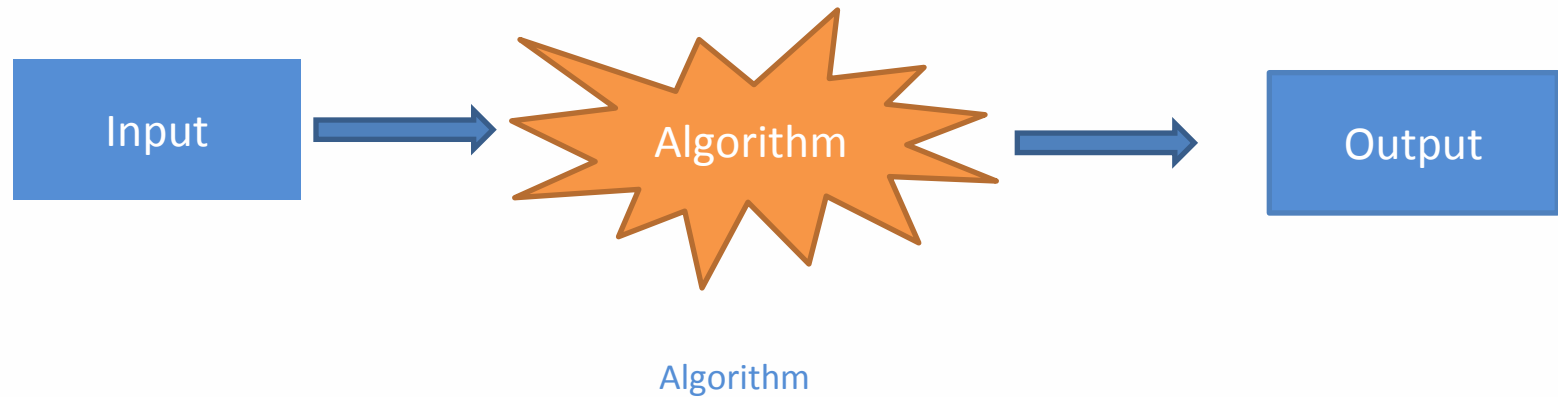
What is Algorithm?

- The word algorithm comes from the name of a Persian mathematician Abu Ja'far Mohammed ibn-i Musa al Khowarizmi.
- Algorithm is a finite set of instructions used to accomplish particular task.
- **An algorithm** takes some value, or set of values, as input and produces some value, or set of values, as output.





What is Algorithm?



Characteristics of algorithm

Input

- Input may be Zero or more

Output

- At least one output will be produced..

Definiteness

- Each instruction should be cleared and unambiguous.

Finiteness

- the steps should be finite.

Characteristics of algorithm

Effectiveness

- each step must be performed in a finite amount of time

Non-ambiguous

- more than one interpretation should not be generated.

Correctness

- produce an relevant and correct output For each input produce.

Generality

- applicable to all homogeneous problems.

Analysis of algorithm





Why algorithm analysis?

- For one problem one or more solutions are available.
- Which one is better ? How can we choose?
- The analysis of an algorithm can help us in understanding of solution in better way.
- **Time & Space analysis**





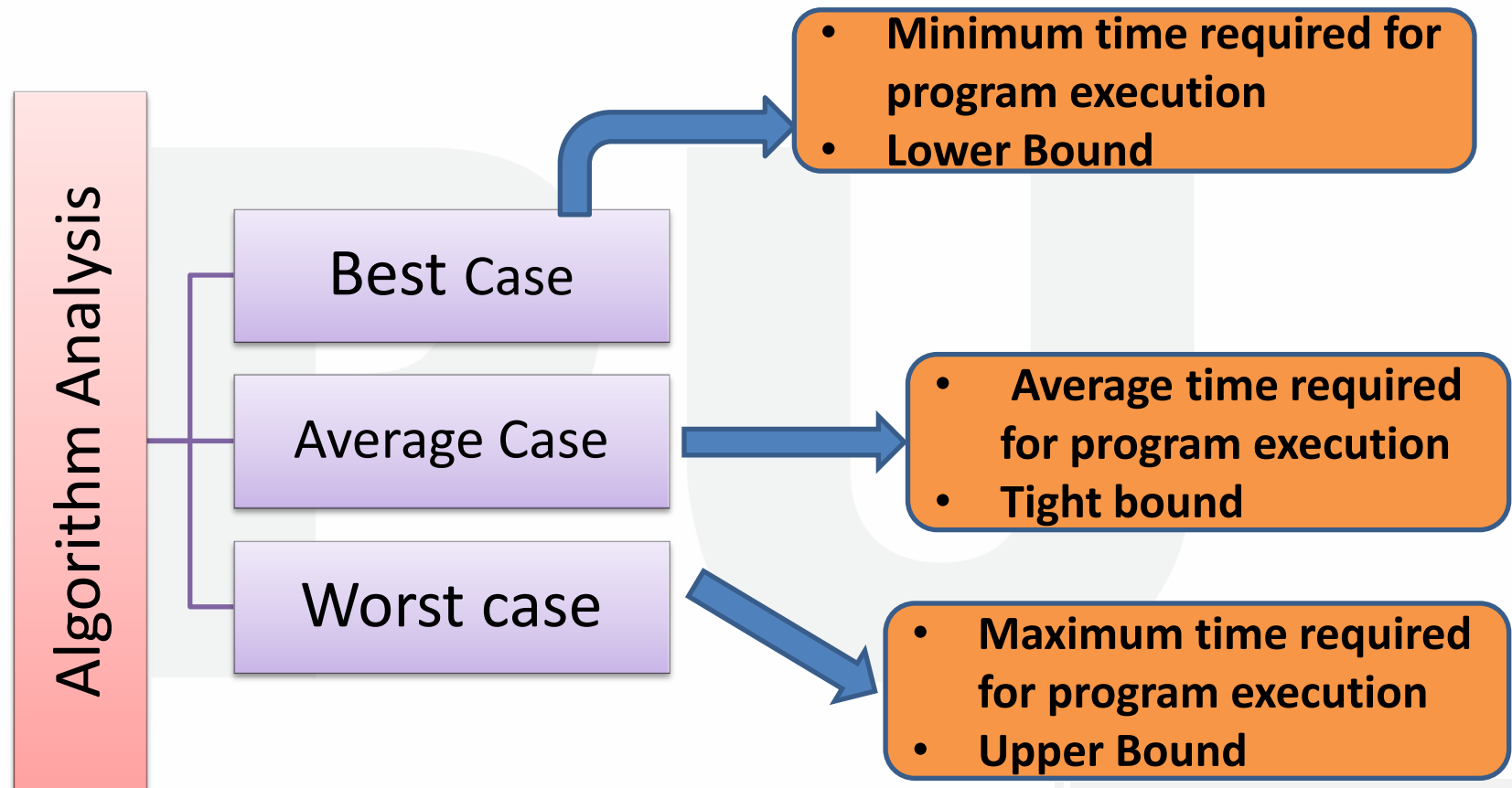
Order of Growth

- Any algorithm is expected to work fast for any input size.
- Smaller input size algorithm will work fine but for higher input size execution time is much higher.
- So how the behavior of algorithm changes with the no. of inputs will give the analysis of the algorithm and is called the Order of Growth.





Algorithm falls under three types





Rate of Growth

- Rate at which the running time increases as a function of input is called rate of growth.

Time Complexity	Name	Performance
1	Constant	Best
$\log n$	Logarithmic	Very good
n	Linear	Good
$n \log n$	Linear Logarithmic	Fair
n^2	Quadratic	Acceptable
n^3	Cubic	Poor
2^n	Exponential	Bad

Asymptotic Complexity

- Refers to defining the mathematical bound of its run-time performance.
- Running time of an algorithm as a function of input size n for large n .
- asymptotic means approaching a value or curve arbitrarily.

Asymptotic Notations

O Notation

- express the tight upper bound of an algorithm
- $f(n)=O(g(n))$ implies: $O(g(n)) = \{ f(n) : \text{there exists positive constants } c>0 \text{ and } n_0 \text{ such that } f(n) \leq c.g(n) \text{ for all } n > n_0. \}$



Asymptotic analysis

Ω Notation

- express the lower bound of an algorithm
- $f(n) = \Omega(g(n))$ implies: $\Omega(g(n)) = \{f(n) : \text{there exists positive constants } c > 0 \text{ and } n_0 \text{ such that } f(n) \geq c \cdot g(n) \text{ for all } n > n_0.\}$

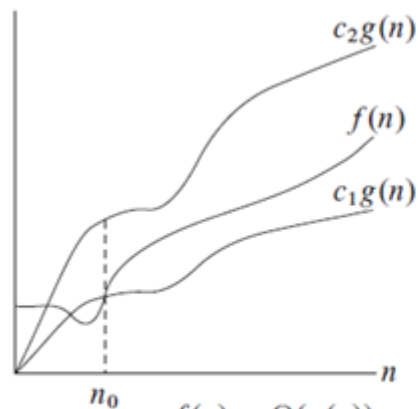
Θ Notation

- express both the lower bound and the upper bound (tight bound)
- $f(n) = \Theta(g(n))$ implies: $\Theta(g(n)) = \{f(n) : \text{there exists positive constants } c_1 > 0, c_2 > 0 \text{ and } n_0 \text{ such that } c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n) \text{ for all } n > n_0.\}$

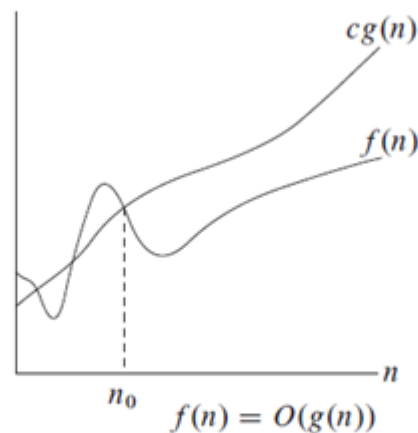
OR, if & only if $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$ for all $n > n_0$



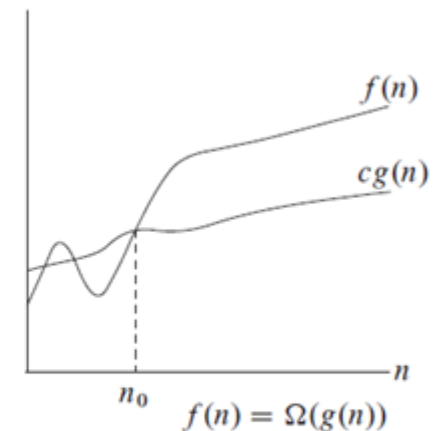
Asymptotic analysis



(a)



(b)



(c)

Fig: Asymptotic average bound (a), upper bound (b) and lower bound (c)

Image source: <http://staff.ustc.edu.cn/~csl/graduate/algorithms/book6/chap02.htm>

Asymptotic analysis

Examples on Big-O Notation(Upper bound/Worst case)

1)find upper bound for $f(n)=2n+2$

solution: $2n+2 \leq 4n$, for all $n \geq 1$

so $2n+2=O(n)$ with $c=4$ and $n_0 = 1$

2)find upper bound for $f(n)=2n^2+1$

solution: $2n^2+1 \leq 3n^2$, for all $n \geq 1$

so $2n^2+1=O(n^2)$ with $c=3$ and $n_0 = 1$



Asymptotic analysis

Examples on Big-O Notation(Upper bound/Worst case)

3) Find upper bound for $f(n)=5n^4+4n+1$

Solution : $5n^4+4n+1 \leq 7n^4$, for all $n \geq 2$

so $5n^4+4n+1 = O(n^4)$ with $c=7$ and $n_0 = 2$



Now Try to solve below Questions

- I. Find upper bound for $f(n)=200$
- II. Find upper bound for $f(n)=n^3+n^2$
- III. Show that $20n^3=O(n^4)$ for appropriate c and n_0 .





Asymptotic analysis

Examples on Big- Ω Notation(Lower bound/ Best case)

1)find lower bound for $f(n)=2n+2$

solution:

$$2n \leq 2n+2, \text{ for all } n \geq 1$$

$$\text{so } 2n+2 = \Omega(n) \text{ with } c=2 \text{ and } n_0 = 1$$

2)find lower bound for $f(n)=2n^2+1$

solution:

$$n^2 \leq 2n^2+1, \text{ for all } n \geq 1$$

$$\text{so } 2n^2+1 = \Omega(n^2) \text{ with } c=1 \text{ and } n_0 = 1$$



Asymptotic analysis

Examples on Big- Ω Notation(Lower bound/ Best case)

3)Find lower bound for $f(n)=5n^4+4n+1$

Solution :

$4n^4 \leq 5n^4+4n+1$, for all $n \geq 1$

so $5n^4+4n+1 = \Omega(n^4)$ with $c=4$ and $n_0 = 1$





Now Try to solve below Questions

- I. For $\sqrt{n+54} = \Omega(\lg n)$. Choose c and n_0
- II. Any linear *function* $an + b$ is in $\Omega(n)$. How?



Asymptotic analysis

Examples on Big- θ Notation(Tight bound/ Average case)

1)find tight bound for $f(n)=2n+2$

solution:

$$2n \leq 2n+2 \leq 4n, \text{ for all } n \geq 1$$

$$\text{so } 2n+2 = \theta(n) \text{ with } c_1 = 2, c_2 = 4 \text{ and } n_0 = 1$$

2)find lower bound for $f(n)=2n^2+1$

solution:

$$n^2 \leq 2n^2+1 \leq 3n^2, \text{ for all } n \geq 1$$

$$\text{so } 2n^2+1 = \theta(n^2) \text{ with } c_1 = 1, c_2 = 3 \text{ and } n_0 = 1$$



Now Try to solve below Questions

- I. Is $5n^3 \in Q(n^4)$??
- II. How about $3^{2n} \in Q(3^n)$??





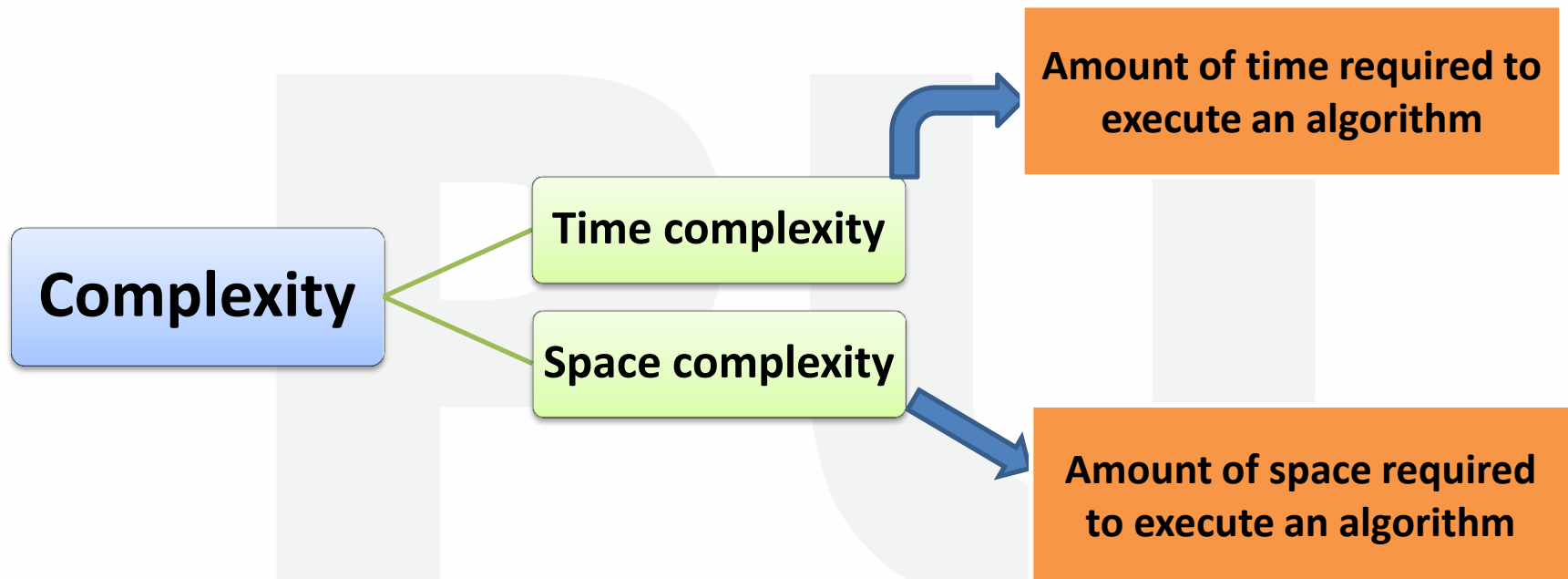
Performance measurements of Algorithm

- Measure of the amount of time and/or space required by an algorithm for an input of a given size (n).
- **What effects run time of an algorithm?**
 - I. computer used, hardware platform
 - II. representation of abstract data types (ADT's)
 - III. efficiency of compiler
 - IV. competence of implementer (programming skills)
 - V. complexity of underlying algorithm
 - VI. size of the input

Out of these above (V) and (VI) are generally most significant



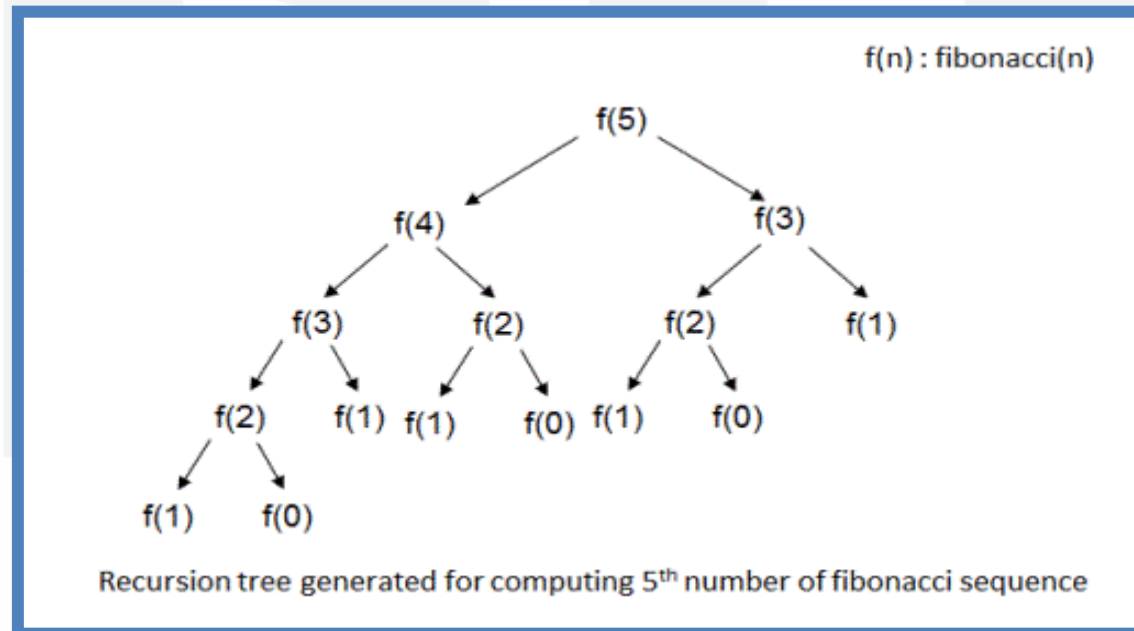
Time complexity and Space complexity



Analysis of recursive algorithms through recurrence

Recurrence relations:

- Refer as an equation that recursively defines a sequence where the next term is a function of the previous terms
- Can easily describe the runtime of recursive algorithms.



Let's create recurrence relation

Example:

```
f(n)
{
    if (n == 0)
        return 1;
    else
        return f(n - 1);
}
```



Recurrence relation

- The base case(termination condition) is reached when $n == 0$. The method performs one comparison. Thus, the number of operations when $n == 0$, $T(0)$, is some constant c .
- When $n \neq 0$, the method calls itself,
- using ONE recursive call, with a parameter $n - 1$.



Recurrence relation

Therefore the recurrence relation is:

Recursive Case: $T(n)=T(n-1)$, for $n>0$

Base Case: $T(n)=c$, for $n=0$ (Here $T(n)$ is running time and n is input size)

Recurrence relation for factorial is $T(n)=T(n-1)$, for $n>0$
 $T(n)=c$, for $n=0$



Exercise

Find out recurrence relation for following algorithm.

```
1.  int fib(int n)
    {
    if (n <= 1)
        return n;
    else
        return fib(n-1) + fib(n-2);
    }
```



Exercise

Find out recurrence relation for following algorithm.

```
ii.) int A(int n) {  
    if (n == 1)  
        return 2;  
    else  
        return A (n / 2) + A( n / 2) + 5;  
}
```





Solving Recurrence Relation

- I. Substitution Method
- II. Iterative Method
- III. Recurrence Tree Method
- IV. Master's Method



Solving Recurrence Relations- Substitution Method

I. $T(n)=T(n-1)+1$, for $n>0$

$T(n)=1$, for $n=0$

Solution:

$T(n)=T(n-1)+1$ 1)

find $T(n-1)$ and put value in equation 1)

$T(n-1)=T(n-2)+1$ 2)

$T(n-2)=T(n-3)+1$ 3)

put the value of equation 2) and 3) in 1)

$T(n)=T(n-3)+3$



Solving Recurrence Relations- Substitution Method

continue for k times .now equation look like,

$$T(n)=T(n-k)+k$$

to reach base case , $n-k=0$ so $k=n$

$$T(n)=T(n-n)+n$$

$$T(n)=1+n$$

$$\text{so ,} T(n)=O(n)$$



Solving Recurrence Relations- Substitution Method

II. $T(n) = T(n-1) + n$, for $n > 1$

$T(n) = 1$, for $n = 1$

Solution:

$$T(n) = T(n-1) + n \quad \dots\dots\dots 1)$$

$$T(n-1) = T(n-2) + (n-1)$$

$$T(n-2) = T(n-3) + (n-2)$$

Put the value in equation 1)

$$T(n) = T(n-2) + (n-1) + n$$

$$T(n) = T(n-3) + (n-2) + (n-1) + n \quad \dots\dots\dots 2)$$



Solving Recurrence Relations- Substitution Method

continue for k times .now equation look like,

$$T(n) = T(n-(k+1)) + (n-k) + \dots + (n-2) + (n-1) + n \quad \dots \dots \dots 3)$$

To reach base case , $n-(k+1)=1$ so $k=n-2$

Put the value in equation 2)

$$T(n) = T(n-(n-2+1)) + (n-(n-2)) + \dots + (n-2) + (n-1) + n$$

$$T(n) = T(1) + 2 + \dots + (n-2) + (n-1) + n$$

$$T(n) = 1 + 2 + \dots + (n-2) + (n-1) + n$$

$$T(n) = (n(n+1))/2$$

$$\text{So } T(n) = O(n^2)$$



Exercise

Find the complexity of the below recurrence:

$$T(n) = \begin{cases} 3T(n - 1), & \text{if } n > 0, \\ 1, & \text{otherwise} \end{cases}$$



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