



# Advance Algorithm Analysis (COMP 502 A)

Sharoon Nasim

[sharoonnasim@fccollege.edu.pk](mailto:sharoonnasim@fccollege.edu.pk)

**Office Hours:**

**Office: S-426 D**

Pre-requisite: Design and Analysis of Algorithms

## **Course Description:**

This course is an introductory graduate-level and advanced undergraduate course on design and analysis of algorithms. Techniques such as divide-and-conquer, dynamic programming, randomized algorithms, and graph algorithms will be taught.

# Course Material

## **Textbooks:**

- Introduction to Algorithms by T. Cormen, C. Lieserson et al.
- Algorithms 2006 (DPV) S. Dasgupta, .et al
- Algorithm Design by J. Kleinberg and E. Tardos

## **Lectures**

Will be uploaded on Moodle.

## Lectures and Examinations:

- One weekly lecture of 150 minutes duration
- Two in-class midterm examinations
- Comprehensive final examination
- Home works

## Course Assessment (Tentative):

Attendance	5%
Class Activities/ Participation	5%
Assignments/ Homework	10 %
Quizzes	15 %
Midterm – 1	15 %
Midterm - 2	20%
Final Exam	30 %
<b>Total</b>	<b>100.00%</b>

# Assessment



What is the difference between dynamic Programming and Divide and Conquer?



What is Recursion?



What approach do you use to crack a password?



Best sorting algorithm and its complexity?



What is the Edit distance problem and which approach is used to solve this problem?



Difference between P and NP class?

# Algorithm

An algorithm is a

- sequence of computational steps
  - to solve any problem
- that transforms input into output.

It should be Correct & Terminate in finite amount of time

# Basic Goal for an Algorithm

- always correct
- always terminates
- This class: performance
  - Performance often draws the line between what is possible and what is impossible.

# Focus of this Course

- Understanding Problem
- Design Algorithm (Different Approaches)  
OR Understand the existing Algorithm
- Analysis of Algorithm (Complexity)



# Approach

- Brute Force Algorithm
- Divide and Conquer
- Dynamic Programming Algorithm
- Greedy Algorithm
- Backtracking Algorithm

# Class Activity

Think of Logic to guess a number between (1 – 100)

At every guess, you will know that your guess is lower or above.

In how many attempts you can guess any number?



Timer

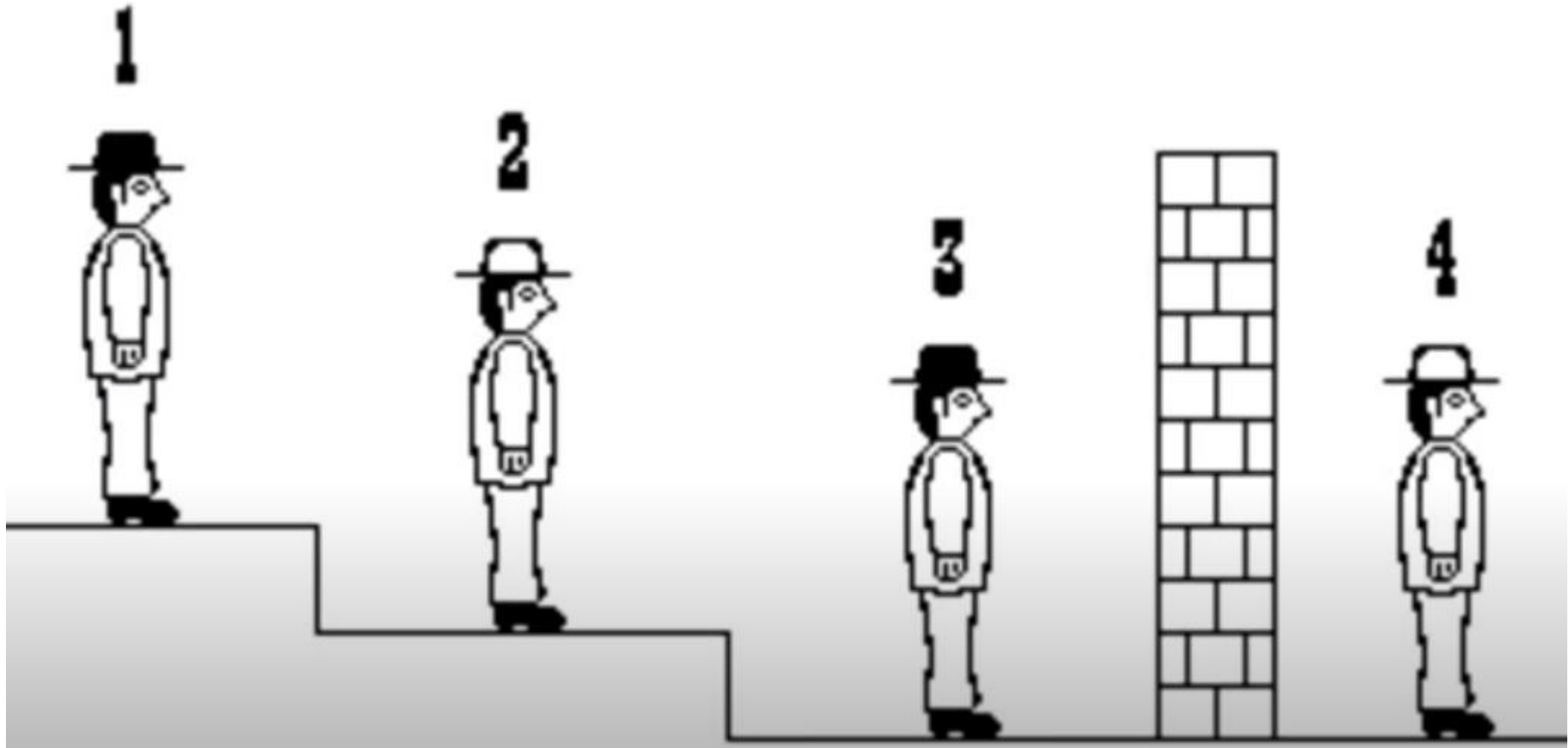


- 4 bottles of milk, one bottle is **poisonous**
- For testing, feed the milk samples to Rat
- **Poison** takes effect at **10 hr**
- Within **24 hr**, you must consume 3 bottles... else you die

## PUZZLE

What strategy would ensure that you drink 3 bottles of milk within **24 hours** ?

Who will shout first and tells his hat colour?



# Hat Problem

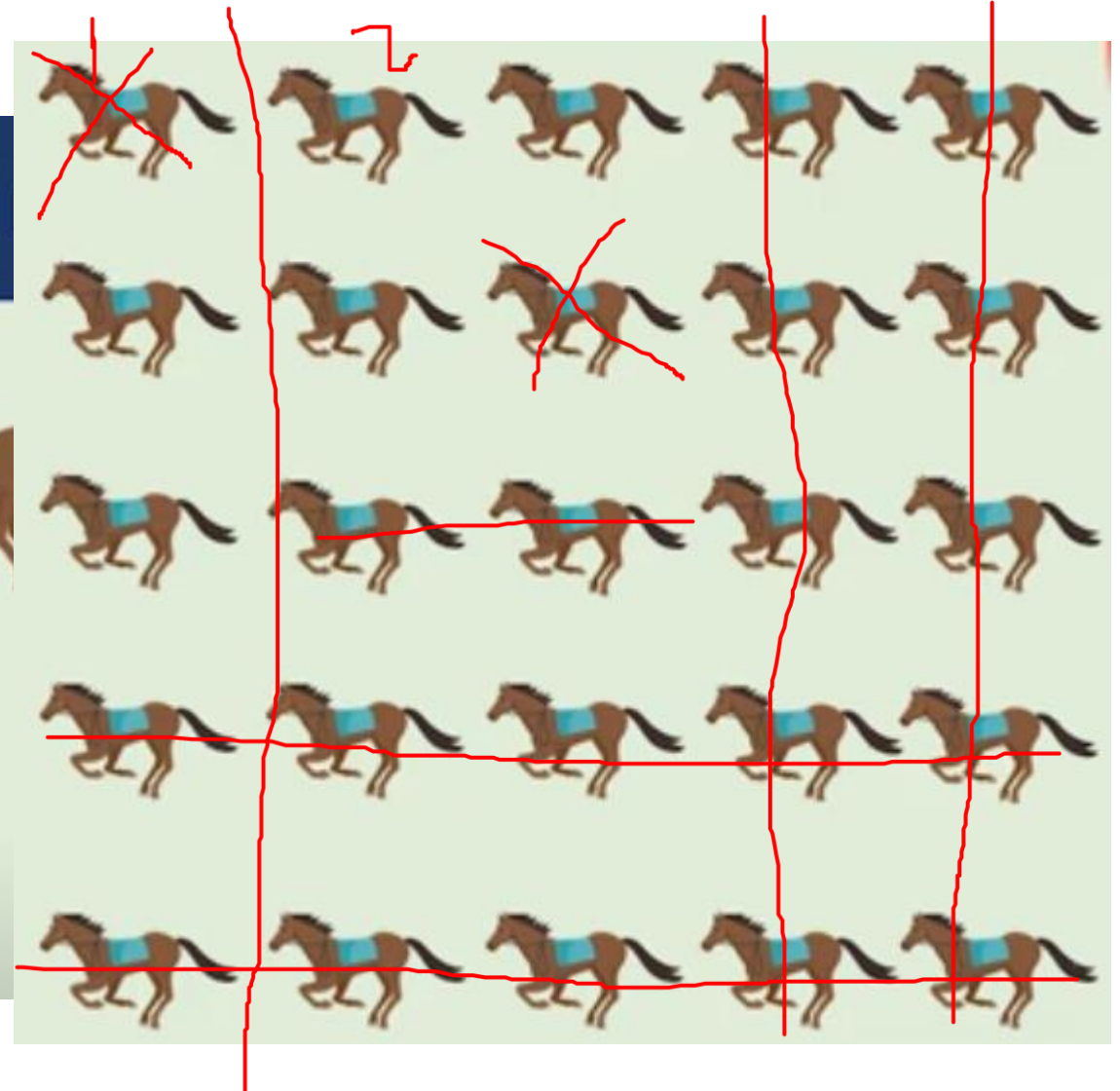


You have total number 25 horses.

**How many races you will**

**Conditions are:**

1. You can run only 5 horses together.
2. There is no clock or timer available.



$$O(n^{\frac{1}{2}})$$

$$O(\log \log n)$$

$$O(\log n)$$

$$O(1)$$

$$O(n^2)$$

$$O(n^n)$$

$$O(n)$$

$$O(2^n)$$

$$O(2^{2^n})$$

$$O(n^3)$$

$$\begin{aligned}
& O(1) < O(\log \log n) < O(\log n) \\
& < O(n^{\frac{1}{2}}) < O(n) < O(n^2) < O(n^3) \\
& < O(2^n) < O(2^{2^n}) < O(n^n)
\end{aligned}$$



Fibonacci Series:

0,1,1,2,3,5,8,13,21,34,55,.....

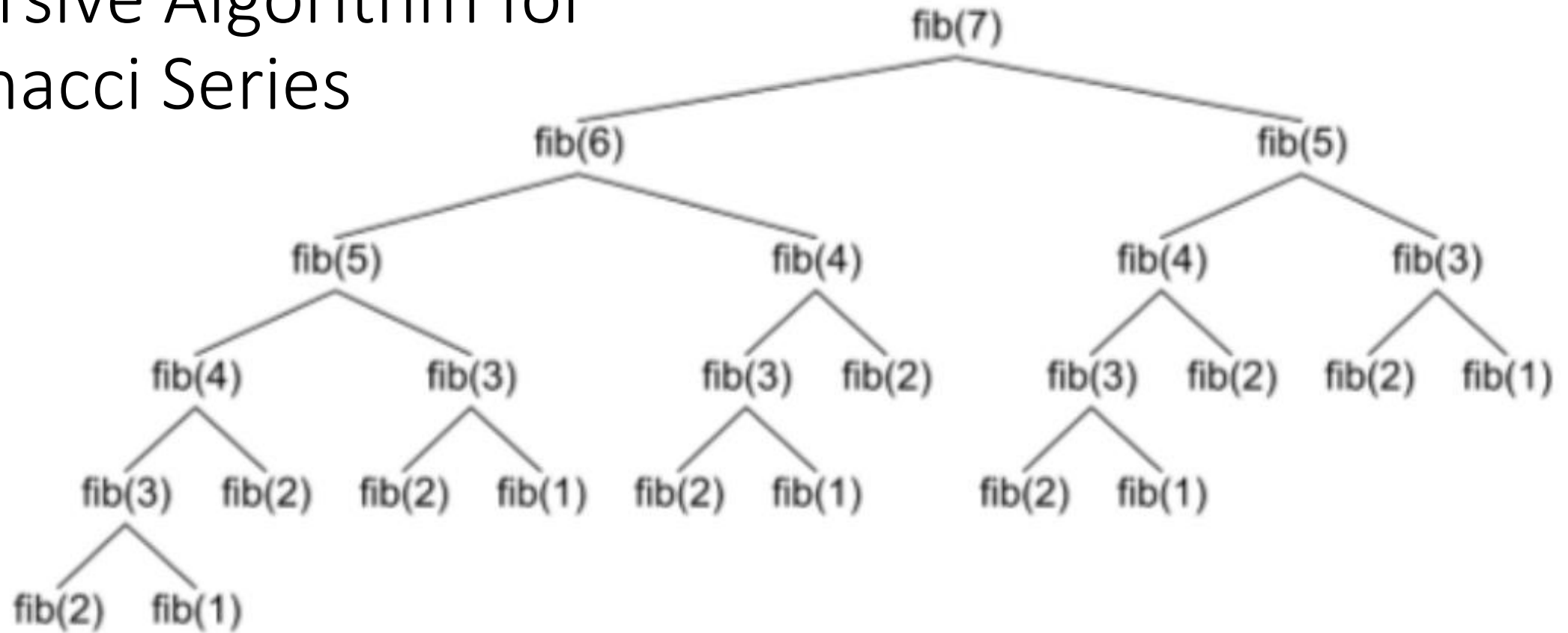
**Each number is the sum of the  
previous two number**

# Iterative Approach

```
def fib(n):  
    a,b = 0,1  
    for i in range(n):  
        a,b = b,a+b  
    return a
```

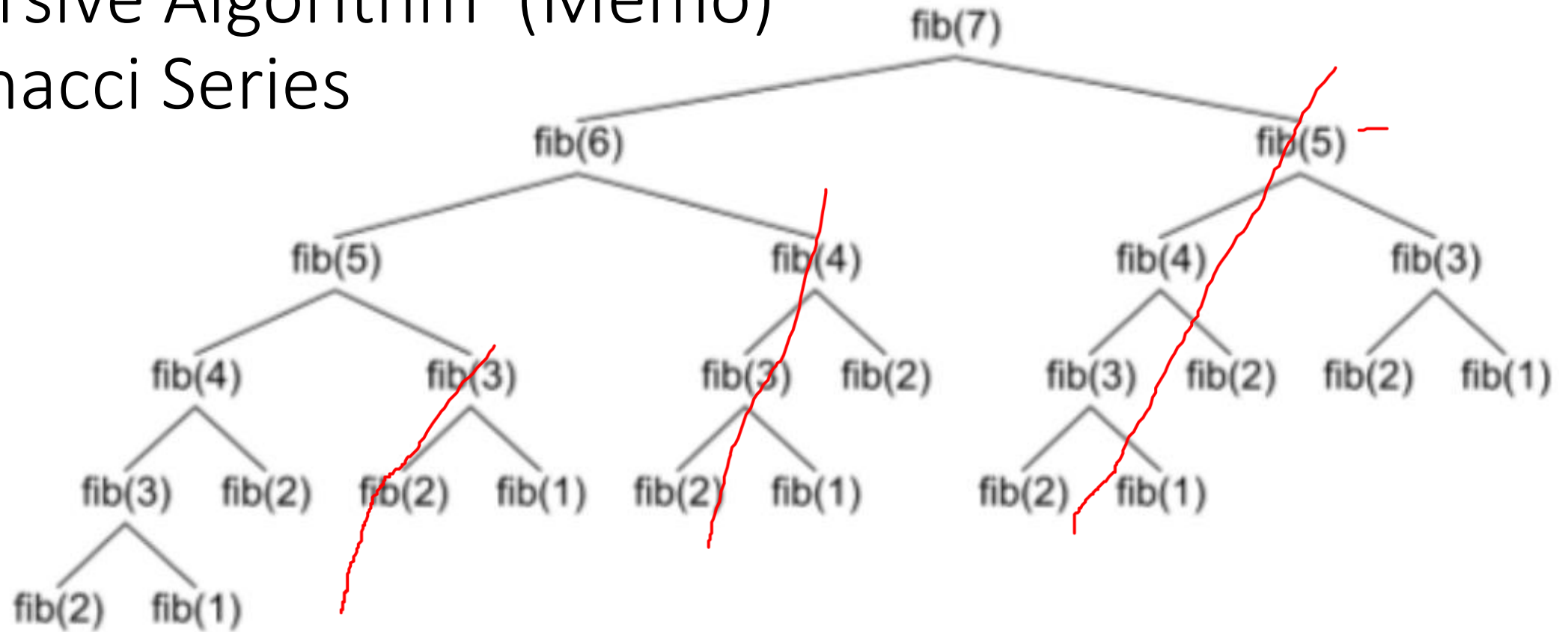
$O(n)$

# Recursive Algorithm for Fibonacci Series



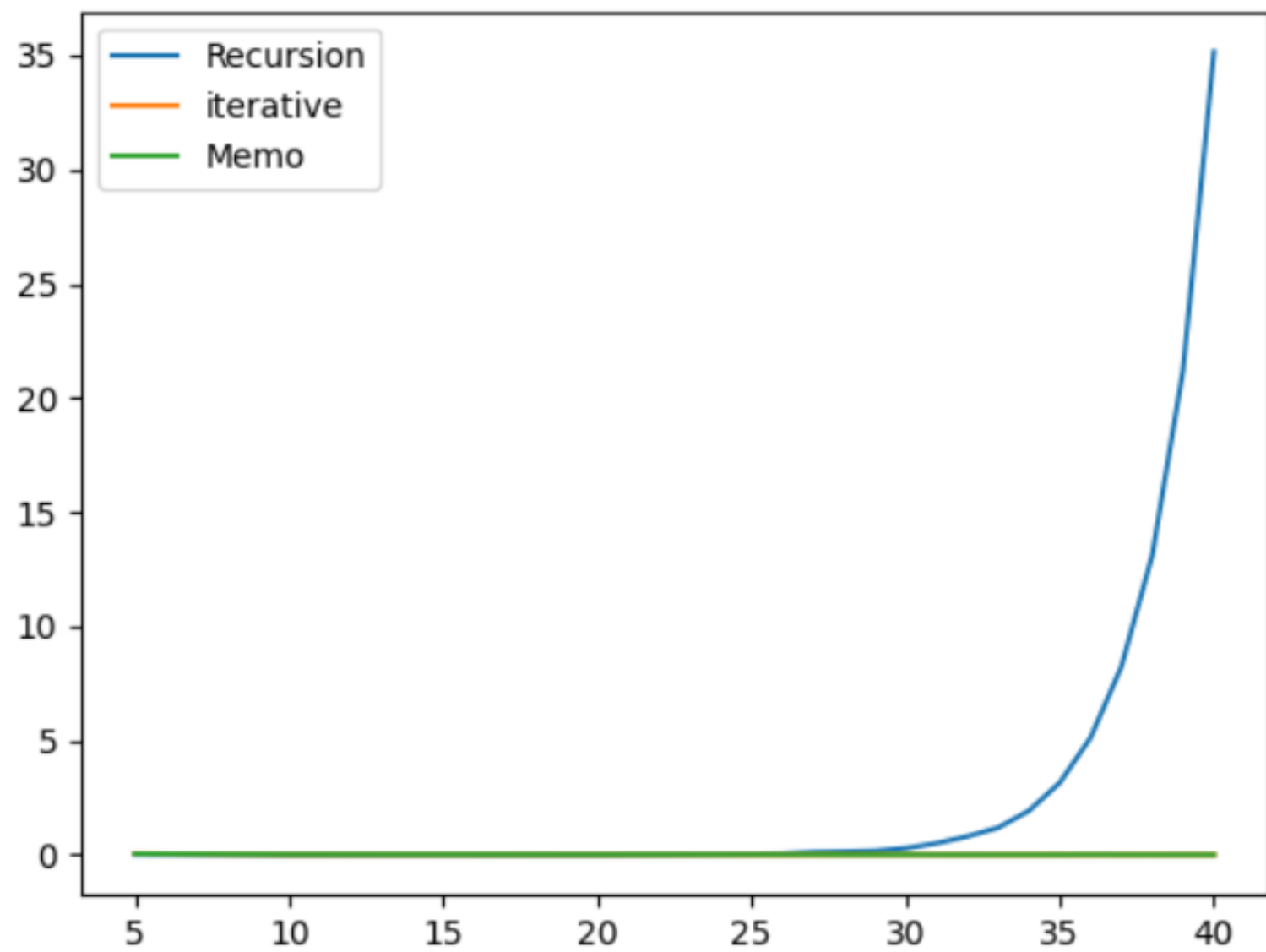
```
def recursive_fib(n):  
    if n <= 1:  
        return n  
    else:  
        return(recursive_fib(n-1) + recursive_fib(n-2))
```

# Recursive Algorithm (Memo) Fibonacci Series



```
def dp_fib(n):  
    if n in memo.keys():  
        return memo[n]  
    else:  
        memo[n] = dp_fib(n-1) + dp_fib(n-2)  
        return(memo[n])
```

0	1						
Fib(0)	Fib(1)	Fib(2)	Fib(3)	Fib(4)	Fib(5)	Fib(6)	



$f = \text{Big O}(g)$

can be thought of as

$f \leq g$

$f = \text{Big Omega}(g)$

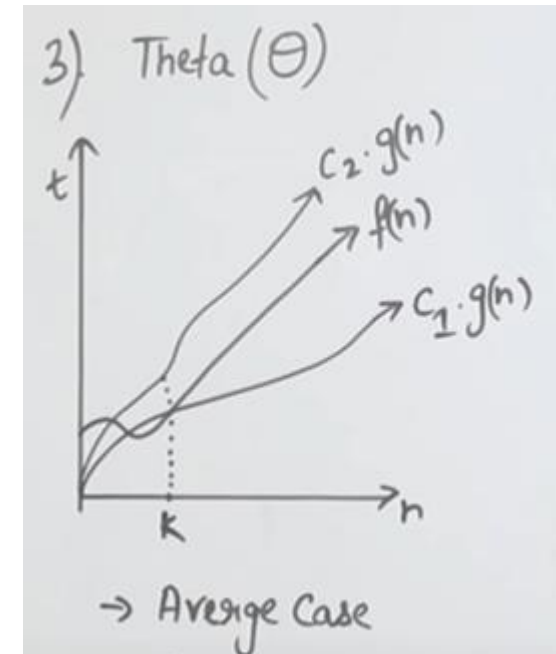
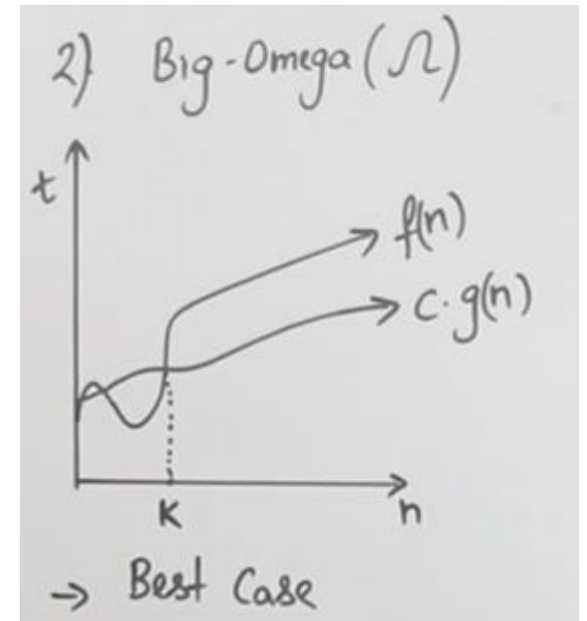
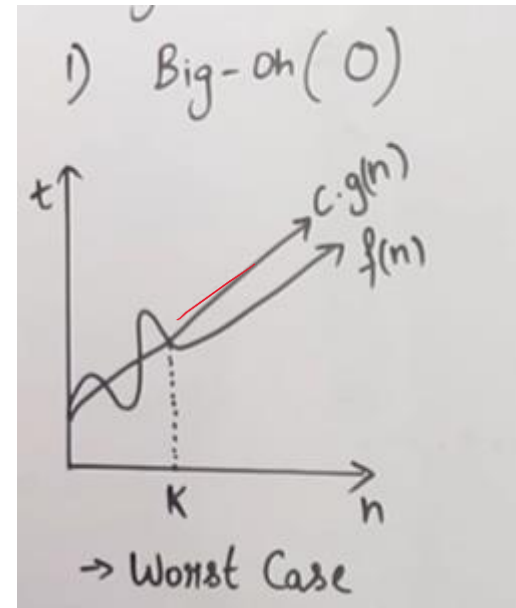
can be thought of as

$f \geq g$

$f = \text{Big Theta}(g)$

can be thought of as

$f = g$



# Posterior and Priori Analysis

**Posteriori analysis** is a relative analysis. It depends on the compiler's language and the hardware type. time algorithm takes to execute on the system.

**Priori analysis** is an absolute analysis. It is independent of the compiler's language and the hardware types. Asymptotic Notations are used to estimate run time instead of running it on the machine.

# Tractability vs Intractability

Tractable Problem: a problem that is solvable by a polynomial-time algorithm. The upper bound is polynomial.

P Class

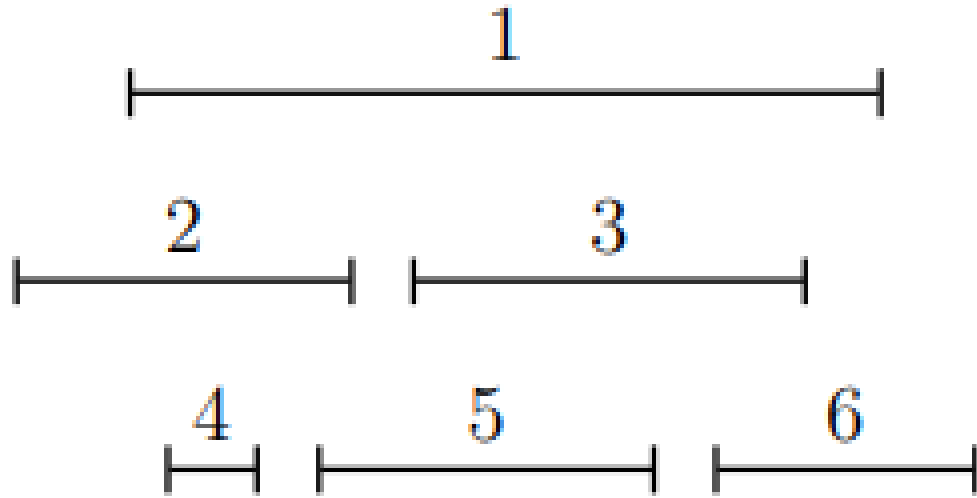
Intractable Problem: a problem that cannot be solved by a polynomial-time algorithm. The lower bound is exponential.

NP Class



	10	50	100	300	1000
$5n$	50	250	500	1500	5000
$n \times \log n$	33	282	665	2469	9966
$n^2$	100	2500	10000	90000	1 million (7 digits)
$n^3$	1000	125000	1 million (7 digits)	27 million (8 digits)	1 billion (10 digits)
$2^n$	1024	a 16-digit number	a 31-digit number	a 91-digit number	a 302-digit number
$n!$	3.6 million (7 digits)	a 65-digit number	a 161-digit number	a 623-digit number	unimaginably large
$n^n$	10 billion (11 digits)	an 85-digit number	a 201-digit number	a 744-digit number	unimaginably large

# Interval Scheduling



Goal: Select a compatible subset of requests of maximum size.

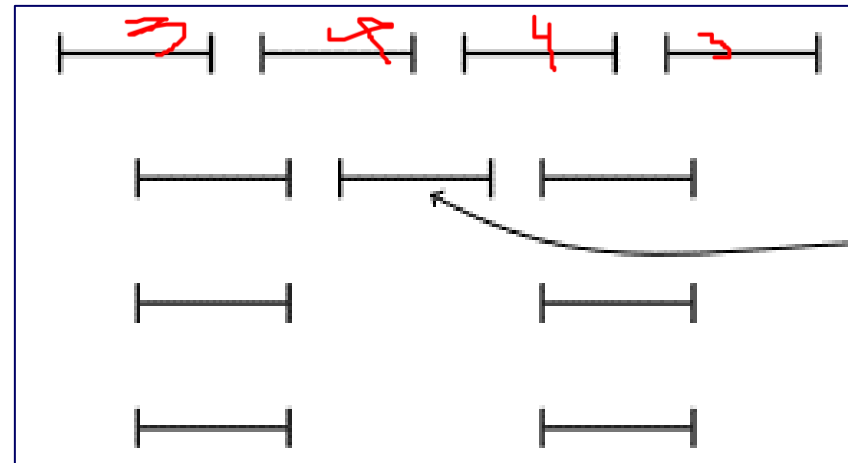
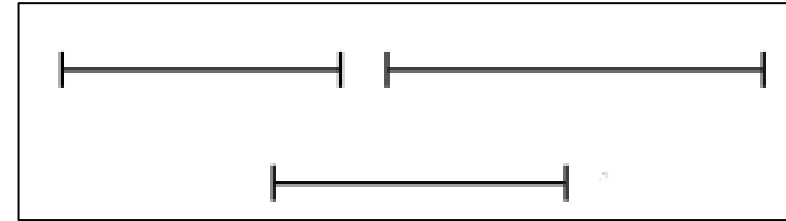
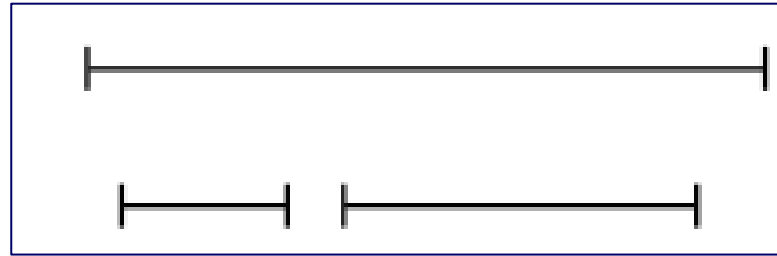
Claim: We can solve this using a greedy algorithm. A greedy algorithm is a myopic algorithm that processes the input one piece at a time with no apparent look ahead.

## Greedy Interval Scheduling

1. Use a simple rule to select a request  $i$ .
2. Reject all requests incompatible with  $i$ .
3. Repeat until all requests are processed

# Possible rules?

1. Select request that starts earliest, i.e., minimum  $s(i)$ .
2. Select request that is smallest, i.e., minimum  $f(i) - s(i)$ .
3. For each request, find number of incompatibles, and select request with minimum such number
4. Select request with earliest finish time, i.e., minimum  $f(i)$



**Claim:** Given list of intervals  $L$ , greedy algorithm with earliest finish time produces  $k^*$  intervals, where  $k^*$  is optimal.

Proof. Induction on  $k^*$ .

Base case:  $k^* = 1$  : this case is easy, any interval works.

Inductive step:

Suppose claim holds for  $k^*$  and we are given a list of intervals whose optimal schedule has  $k^* + 1$  intervals, namely

$$S^*[1, 2, \dots, k^* + 1] = \langle s(j_1), f(j_1) \rangle, \dots, \langle s(j_{k^*+1}), f(j_{k^*+1}) \rangle$$

Say for some generic  $k$ , the greedy algorithm gives a list of intervals

$$S[1, 2, \dots, k] = \langle s(i_1), f(i_1) \rangle, \dots, \langle s(i_k), f(i_k) \rangle$$

By construction, we know that  $f(i_1) \leq f(j_1)$ , since the greedy algorithm picks the earliest finish time.

Now we can create a schedule

$$S^{**} = \langle \underbrace{s(i_1), f(i_1)} \rangle, \langle s(j_2), f(j_2) \rangle, \dots, \langle s(j_{k^*+1}), f(j_{k^*+1}) \rangle$$

since the interval  $\langle s(i_1), f(i_1) \rangle$  does not overlap with the interval  $\langle s(j_2), f(j_2) \rangle$  and all intervals that come after that. Note that since the length of  $S^{**}$  is  $k^* + 1$ , this schedule is also optimal.

$L'$  = list of intervals that are compatible

$L'$  is the set of intervals with  $s(i) \geq f(i_1)$ .

Since  $S^{**}$  is optimal for  $L$ ,

$S^{**}[2, 3, \dots, k^* + 1]$  is optimal for  $L'$ , which implies that the optimal schedule for  $L'$  has  $k^*$  size.

We now see by our initial inductive hypothesis that running the greedy algorithm on  $L'$  should produce a schedule of size  $k^*$ . Hence, by our construction, running the greedy algorithm on  $L'$  gives us  $S[2, \dots, k]$ .

This means  $k - 1 = k^*$  or  $k = k^* + 1$ ,

which implies that  $S[1, \dots, k]$  is indeed optimal, and we are done.

# Weighted Interval Scheduling

Each request  $i$  has weight  $w(i)$ . Schedule subset of requests that are non-overlapping with maximum weight.

Greedy Approach 

# Weighted Interval Scheduling – Dynamic Programming

$$WIS(1 \dots n) = \max \{ W_1 + WIS(R), WIS(2 \dots n) \}$$

# R > remaining compatible requests

Sort all interval according to start time in ascending order.



	Intervals	W	Intervals (sorted)	W
1	1-3	5	1-3	5
2	2-5	6	2-5	6
3	4-6	7	4-6	7
4	6-7	4	5-8	11
5	5-8	11	6-7	4
6	7-9	2	7-9	2

$$WIS(1...n) = \max \{ W1 + WIS(R), WIS(2 \dots n) \}$$

$wis(1-6) = \max(5 + wis(2-6), wis(2-6))$   
 $wis(2-6) = \max(6 + wis(3-6), wis(3-6))$   
 $wis(3-6) = \max(7 + wis(5-6), wis(4-6))$   
 $wis(4-6) = \max(11 + wis(6), wis(5-6))$   
 $wis(5-6) = \max(4 + wis(6), wis(6))$   
 $wis(6) = 2$

(Handwritten notes in green ink on the left side of the equations, with arrows pointing to specific terms in the recursive calls: 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65, 66, 67, 68, 69, 70, 71, 72, 73, 74, 75, 76, 77, 78, 79, 80, 81, 82, 83, 84, 85, 86, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 100)