Advanced Data Structure and Algorithms fo Problem Solving

Lecture # 04

Vineet Gandhi

Center for Visual Information Technology (CVIT),
IIIT Hyderabad



Recursion

- Recursion is a mathematical technique that evaluates a function by calling the same function repeatedly on smaller inputs.
- Most programming languages support such a style of programming.
 - Often very elegant to study.
- Helps in problem solving too.

Recursion

 Q: How many twists does it take to screw in a light bulb?

A: Is it already screwed in? Then zero. If not, then twist it once, ask me again, and add 1 to my answer.

Recursion

- Relates to mathematical induction
- Divide and Conquer algorithms

Lets start with an examples

A mathematical view of computer science

Lets start with an examples

- A mathematical view of computer science
- |factorial(n) = n* factorial(n-1)

Three Laws of Recursion

- Recursion must have a base case
- A recursive algorithm must change its state and move toward the base case.
- A recursive algorithm must call itself, recursively.

$$H(n=1)$$
 return 1

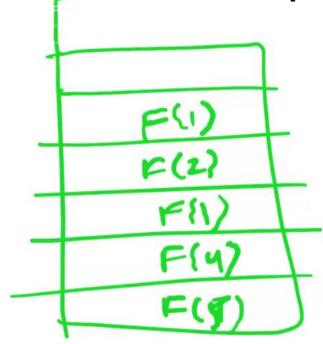
All return $1 \times f(x)$
 $1 \times f(x)$

test (5) 4(h==1) rdom 1 - F(51) 5xF(4) R elu return nxfact(n-1) - F14) YXF(3) R 7 F(1) 3 x F(1) R F(I)

$$H(n=1)$$
 ration | $feat(s)$

All rations $n \times feat(n=1)$ $rden$ $state$
 $T(n) = T(n-1) + 3$ $respectively = 1$ $respectivel$

Space and time complexity



Inductive Reasoning

- A recursive algorithm must have a base case.
- If I call factorial(n) with n=1, I am done
- If I call factorial(n) with n>1, it makes a recursive call with a smaller value of n; must eventually reach n=1.

Recursion and Induction

Recursion with multiple base cases

$$fib(n) = \begin{cases} fib(n-1) + fib(n-1) + fib(n-1) \\ n \end{cases} + fib(n) \end{cases}$$

$$fib(n) \begin{cases} fib(n) \end{cases}$$

$$fib(n) \begin{cases} fib(n-1) + fib(n-2) \end{cases} = fid(n-1) + fid(n-2) \end{cases} = fid(n-1)$$

Fibonacci deeper look

Time and space complexity

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

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

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

$$T(n) = 2[2 \cdot T(n-4) + C] + C$$

$$T(n) = 2^{K} T(n-2K) + (2^{K}-1) C$$

This is lower bound as we have replaced n-1 with n-2. To find upper bound, replace n-2 with n-1 (2^n)

```
lecture4 - recursion - vi fib_rec.c - 80×24
#include<stdio.h>
int fib(int n){
if(n<=1) return n;
else return fib(n-1) + fib(n-2);
int main(){
         int a;
         printf("Enter a number:");
         scanf("%d",&a);
         printf("%d\n",fib(a));
         return 0;
   INSERT --
```

```
#include<stdio.h>
int fib(int n){
                                                                     Ŧ
        if(n \le 1)
                return n;
        int i, sum=0, f0=0, f1=1;
        for(i=2;i<=n;i++){
                 sum = f0+f1;
                f0 = f1;
                f1 = sum;
        }
        return sum;
}
int main(){
        int a;
        printf("enter a number:");
        scanf("%d",&a);
        printf("%d \n",fib(a));
        return 0;
"fib_iter.c" 21L, 265C
```

Recursion with memoization

```
Fib In
( it m (21) when h;
 if Fn is in money run Fn
 ~ ~ = = 15 (m-1) + Fis (m=2)
    Sam Fy in many
```

Exponentiation

$$\mathcal{H}^{n} = \begin{cases} \mathbf{x}^{n/2} & \mathbf{x$$