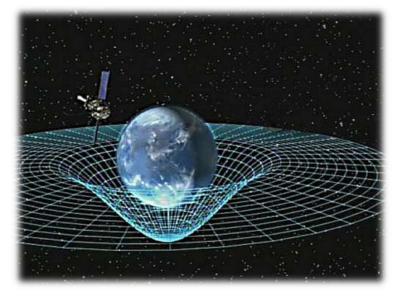
# Order of Growth

# In Physics, We consider

• Time

Space



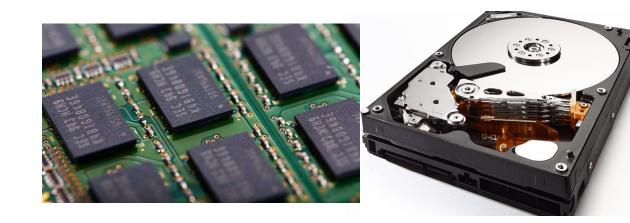


## In CS, we consider

- Time
  - how long it takes to run a program



- Space
  - how much memory do we need to run the program



#### Order of Growth Analogy

- Suppose you want to buy a Bluray movie from Amazon (~40GB)
- Two options:
  - Download
  - 2-day Prime Shipping
- Which is faster?



# The Infinity Saga Box Set





#### Order of Growth Analogy

- Buy the full set?
  - 23 movies
- Two options:
  - Download
  - 2-day Prime Shipping

Which is faster?



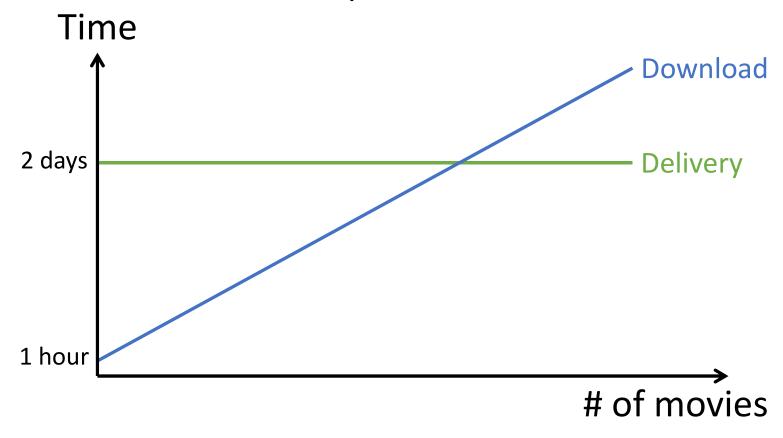
# Order of Growth Analogy

• Or even more movies?





## Download vs Delivery



#### **Ultimate Question**

- If the "volume" increased
- How much more resources, namely time and space, grow?

#### Will they grow in the same manner?

- From
  - factorial (10)
- To
  - factorial (20)
- To
  - factorial (100)
- To
  - factorial (10000)

- From
  - fib(10)
- To
  - fib(20)
- To
  - fib(100)
- To
  - fib(10000)

#### Order of Growth

- •is NOT...
  - The absolute time or space a program takes to run
- •is
  - the proportion of growth of the time/space of a program w.r.t. the growth of the input

#### Let's try it on something we know

```
def factorial(n):
    if n <= 1:
       return 1
    else:
        return n * factorial(n - 1)
def fib(n):
    if (n == 0):
       return 0
    elif (n == 1):
      return 1
    else:
        return fib (n - 1) + fib (n - 2)
```

#### Let's try it on something we know

```
nfact, nfib = 0,0
def factorial(n):
    global nfact
   nfact +=1
    if n <= 1:
        return 1
    else:
        return n * factorial(n - 1)
def fib(n):
    global nfib
    nfib +=1
    if (n == 0):
        return 0
    elif (n == 1):
        return 1
    else:
        return fib(n - 1) + fib(n - 2)
```

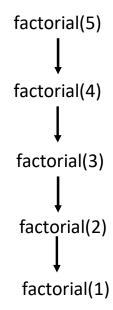
# Compare

>>> factorial(5)	>>> fib(5)
120	5
>>> nfact	>>> nfib
5	15
>>> nfact = 0	>>> nfib = 0
>>> factorial(10)	>>> fib(10)
3628800	55
>>> nfact	>>> nfib
10	177
>>> nfact = 0	>>> nfib = 0
>>> factorial(20)	>>> fib(20)
2432902008176640000	6765
>>> nfact	>>> nfib
20	21891

#### Order of Growth of Factorial

```
>>> factorial(5)
120
>>> nfact
5
>>> nfact = 0
>>> factorial(10)
3628800
>>> nfact
10
>>> nfact
10
>>> nfact = 0
>>> factorial(20)
2432902008176640000
>>> nfact
20
```

- Factorial is simple
  - If the input is n, then the function is called n times
  - Because each time n reduced by 1
- So the number of times of calling the function is proportional to n



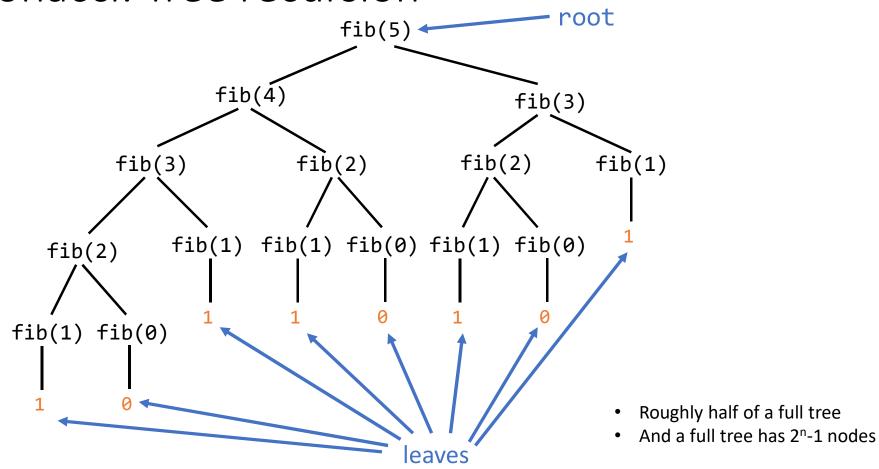
#### Fib

• More complicated

Why?

```
>>> fib(5)
5
>>> nfib1
5
>>> nfib = 0
>>> fib(10)
55
>>> nfib
177
>>> nfib = 0
>>> fib(20)
6765
>>> nfib
21891
```

#### Fibonacci: Tree recursion



### Compare

>>> factorial(5)	>>> fib(5)			
120	5			
>>> nfact	>>> nfib1			
5	5			
>>> nfact = 0	>>> nfib = 0			
>>> factorial(10)	>>> fib(10)			
3628800	55			
>>> nfact	>>> nfib			
10	177			
>>> nfact = 0	>>> nfib = 0			
>>> factorial(20)	>>> fib(20)			
2432902008176640000	6765			
>>> nfact	>>> nfib			
20	21891			

No of calls proportional to n

No of calls proportional to 2<sup>n</sup>

#### Searching in a list of n items

- Linear search
  - # comparisons proportional to n
  - (Because in average, the expected number of search is n/2)
- Binary search
  - # comparisons proportional to log n
  - Because, we divide the list into half for at most log n times

#### Sorting a list of n Items

- Selection/Bubble Sort
  - # comparisons proportional to n<sup>2</sup>
  - Because we looped n times, and each time you need to arrange 1 to n items
- Merge sort
  - # comparisons proportional to n log n
  - Because, we divide the list into half for at most log n times
  - And each time arrange n items

#### **Bubble Sort**

8	4	5	9	2	3	7	1	6	0
4	8	5	9	2	3	7	1	6	0
4	5	8	9	2	3	7	1	6	0
4	5	8	9	2	3	7	1	6	0
4	5	8	2	9	3	7	1	6	0
4	5	8	2	3	9	7	1	6	0
4	5	8	2	3	7	9	1	6	0
4	5	8	2	3	7	1	9	6	0
4	5	8	2	3	7	1	6	9	0
4	5	8	2	3	7	1	6	0	9

## **Bubble Sort**

4	5	8	2	3	/	1	6	Ü	9
4	5	8	2	3	7	1	6	0	9
4	5	8	2	3	7	1	6	0	9
4	5	2	8	3	7	1	6	0	9
4	5	2	3	8	7	1	6	0	9
4	5	2	3	7	8	1	6	0	9
4	5	2	3	7	1	8	6	0	9
4	5	2	3	7	1	6	8	0	9
4	5	2	3	7	1	6	0	8	9

#### **Bubble Sort**

```
def bubble(my_list):
    for i in range(len(my_list)-1):
        if my_list[i] > my_list[i+1]:
            if my_list[i] > my_list[i+1]:
                my_list[i], my_list[i+1] = my_list[i+1], my_list[i]

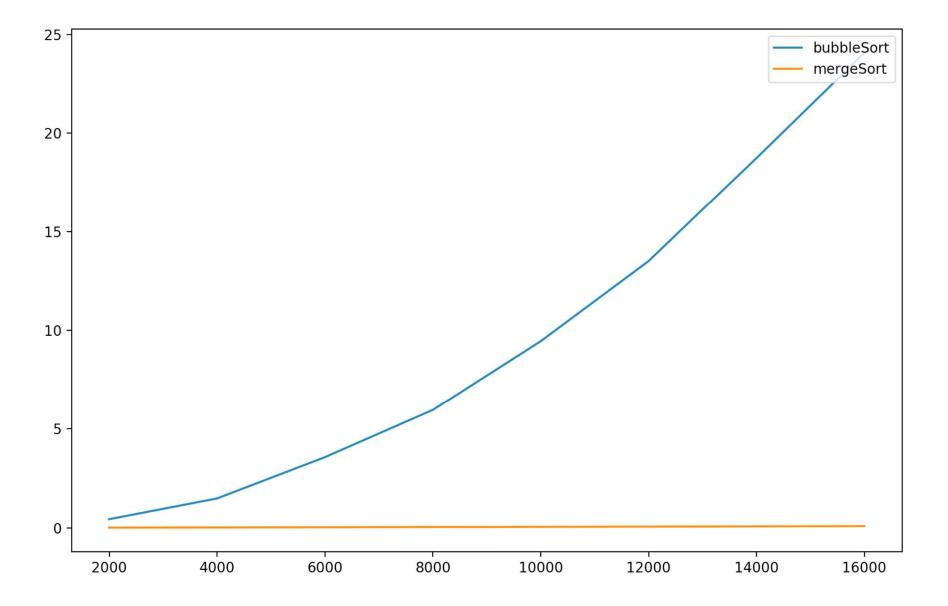
def bubblesort(my_list):
    for i in range(len(my_list)-1):
        bubble(my_list)

my_list_1 = [38,2,10,3,1]

pubblesort(my_list_1)

print(my_list_1)
```

```
from random import randint
from time import time
ln = [2000, 4000, 6000, 8000, 10000, 12000, 14000, 16000]
bstat = []
mstat = \square
for n in ln:
    rl = [randint(1,100000) for i in range(n)]
    st = time()
    bubbleSort(rl)
    btime = time()-st
    st = time()
    mergeSort(rl)
    mtime = time()-st
    print(f'For n = {n}, bubbleSort: {btime}s mergeSort: {mtime}s')
    bstat.append(btime)
    mstat.append(mtime)
```



# Algorithm

Anyone can give some algorithms

#### BogoSort

- BogoSort(L)
  - Repeat:
    - Choose a random permutation of the list L.
    - If L is sorted, return L.
- If you wait enough time, L is sorted?



#### Bogo Sort

Randomly shuffle the list till the list is sorted

```
import random
def is_not_sorted(shuffled_list):
    for i in range(len(shuffled_list)-1):
        if shuffled_list[i] > shuffled_list[i+1]:
            return True
    return False

def bogosort(my_list):
    while is_not_sorted(my_list):
        random.shuffle(my_list)

my_list = [38,2,10,3,1]
bogosort(my_list)
print(my_list)
```

Can we do better?

#### Hill-Climbing for Sorting

- Optimization algorithm
  - Require an evaluation function
- Which metric is better for evaluation?
  - Let *my\_list* be our list
  - Number of index pairs i, j such that such that  $my_list[i] > my_list[j]$
- Example: >>> my\_list = [38,2,10,3,1] >>> my\_list [38,2,10,3,1] >>> my\_list = [1, 2, 3, 10, 38] >>> my\_list = [1, 2, 3, 10, 38] >>> my\_list | Value(my\_list) = 0 + 0 + 0 + 0 = 0 [1, 2, 3, 10, 38]

#### Hill-Climbing for Sorting

- Repeat the following either till value of the list is zero or a predetermined number of times
  - Shuffle the list
  - Accept the shuffled list if its value is lower than that of current list

# Algorithm

Anyone can give some algorithm

But how fast is your algorithm?

#### How about

```
QuantumBogoSort(A[1..n])
```

- a) Choose a random permutation of the array A.
- b) If A is sorted, return A.
- c) If A is not sorted, destroy the universe.

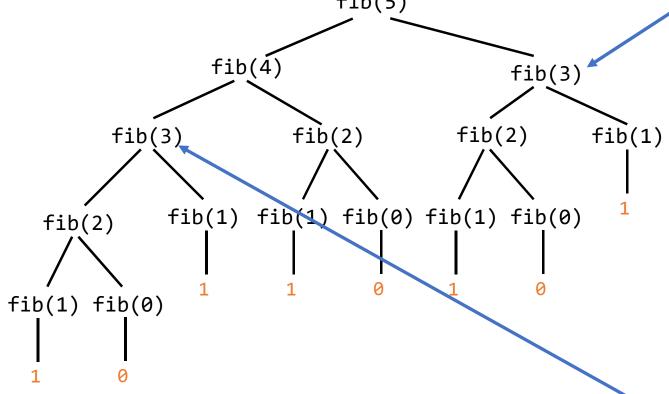
 Remember QuantumBogoSort when you learn about nondeterministic Turing Machines

# Improvement?

Let's try fib(n)

# Easy Way: Memoization fib(5)fib(4)

Instead of recomputing fib(3) here



We save the result when it was computed

#### Memoization

- Create a dictionary to remember the answer if fibm(n) is computed before
- If the ans was computed before, get the answer from the dictionary
- Otherwise, compute the ans and put it into the dictionary for later use

```
def fibm(n):
    if n in fibans.keys():
        return fibans[n]
    if (n == 0):
        ans = 0
    elif (n == 1):
        ans = 1
    else:
        ans = fibm(n - 1) + fibm(n - 2)
    fibans[n] = ans
    return ans
```

 $fibans = {}$ 

# Can you use Memoization to compute nCk?



#### Recursion Removal

- Store all the answers in an array
- Add a new fib(i)
  - as fib(i-1) + fib(i-2)

- Wait a min...
  - Do we need all the past numbers if we only need fib(n)?

```
def fibi(n):
    ans = [0,1,1]
    if n < 3:
        return ans[n]
    for i in range(3,n+1):
        ans.append(ans[i-1]+ans[i-2])
    return ans[n]</pre>
```

#### Recursion Removal 2

```
Add a new fib(i)

as fib(i-1) + fib(i-2)

And I only need to keep fib(i-1) and fib(i-2)

def fibi2(n):

if n < 3:
<ul>
return 1
fibminus1, fibminus2 = 1,1

for i in range(3,n+1):

fibminus2, fibminus1 = fibminus1, fibminus1 + fibminus2
```

#### Improvement

- For IT5001, you should know how to compute fib(n) with time proportional to n
  - The fastest algorithm to compute fib(n) with time proportional to log n
- To know more about this to improve human race and save the world
  - CS1231, CS2040, CS3230, etc
- What you learn today is called the Big O notation
  - O(n),  $O(\log n)$ ,  $O(n^2)$ ,  $O(n \log n)$ ,  $O(2^n)$ , etc