

What is efficiency in
programming?

Why efficiency is important?

Types of efficiency

Space and Time Efficiency

Our focus - Time

Techniques to measure time efficiency

Techniques

1. Measuring **time** to execute
2. **Counting** operations involved
3. Abstract notion of **order of growth**

1. Measuring Time

Problems with this approach

Different time for different algorithm	✓
Time varies if implementation changes	✗
Different machines different time	✗
Does not work for extremely small input	✗
Time varies for different inputs, but can't establish a relationship	✗

2. Counting Operations

COUNTING OPERATIONS

- assume these steps take **constant time**:
 - mathematical operations
 - comparisons
 - assignments
 - accessing objects in memory
- then count the number of operations executed as function of size of input

```
def c_to_f(c):  
    return c*9.0/5 + 32  
  
def mysum(x):  
    total = 0  
    for i in range(x+1):  
        total += i  
    return total
```

mysum → 1+3x ops

Annotations for mysum(x):
- The assignment 'total = 0' is labeled '1 op'.
- The loop 'for i in range(x+1):' is labeled 'loop x times'.
- The assignment 'total += i' is labeled '1 op'.
- The final return statement 'return total' is labeled '2 ops'.
- The overall complexity is labeled '3 ops'.

Problems with this approach

Different time for different algorithm	✓
Time varies if implementation changes	✗
Different machines different time	✓
No clear definition of which operation to count	✗
Time varies for different inputs, but can't establish a relationship	✓

What do we want

1. We want to evaluate the algorithm
2. We want to evaluate scalability
3. We want to evaluate in terms of input size

DIFFERENT INPUTS CHANGE HOW THE PROGRAM RUNS

- a function that searches for an element in a list

```
def search_for_elmt(L, e):  
    for i in L:  
        if i == e:  
            return True  
    return False
```

- when e is **first element** in the list → BEST CASE
- when e is **not in list** → WORST CASE
- when **look through about half** of the elements in list → AVERAGE CASE

3. Orders of Growth

ORDERS OF GROWTH

Goals:

- want to evaluate program's efficiency when **input is very big**
- want to express the **growth of program's run time** as input size grows
- want to put an **upper bound** on growth – as tight as possible
- do not need to be precise: “**order of**” not “**exact**” growth
- we will look at **largest factors** in run time (which section of the program will take the longest to run?)
- **thus, generally we want tight upper bound on growth, as function of size of input, in worst case**

EXACT STEPS vs O()

```
def fact_iter(n):
    """assumes n an int >= 0"""
    answer = 1
    while n > 1:
        answer *= n
        n -= 1
    return answer
```

- computes factorial
- number of steps: $1 + 5n + 1$
- worst case asymptotic complexity: $O(n)$
 - ignore additive constants
 - ignore multiplicative constants

So the idea is simple

$$n^2 + 2n + 2$$

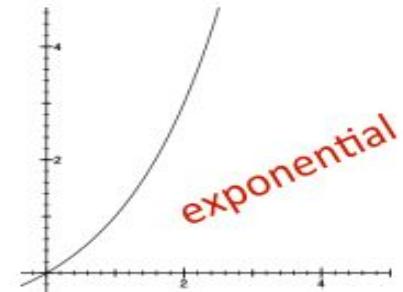
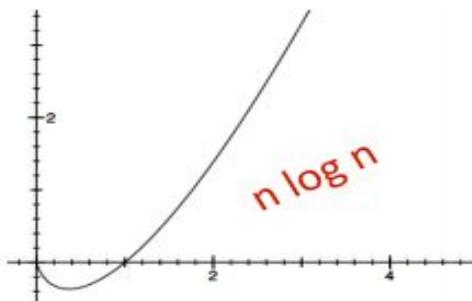
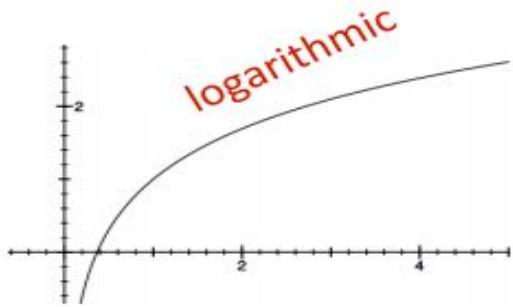
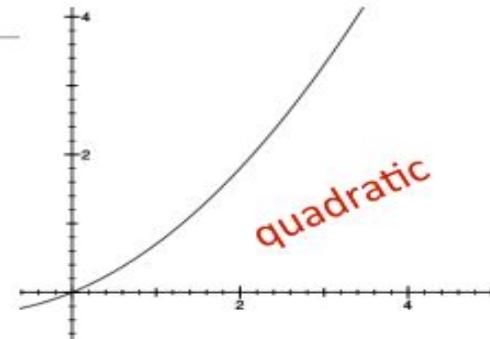
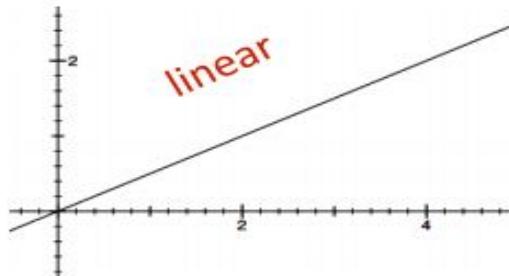
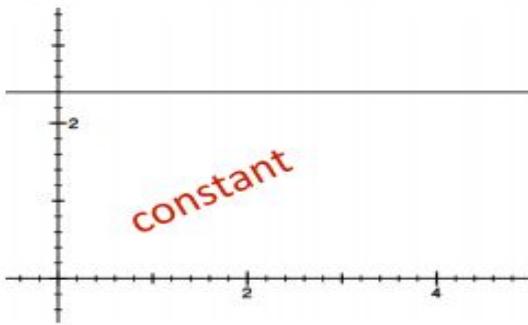
$$n^2 + 100000n + 3^{1000}$$

$$\log(n) + n + 4$$

$$0.0001 * n * \log(n) + 300n$$

$$2n^{30} + 3^n$$

TYPES OF ORDERS OF GROWTH



Law of addition

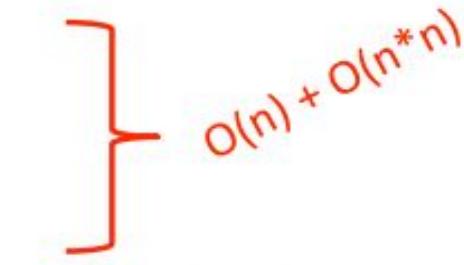
Law of Addition for $O()$:

- used with **sequential** statements
- $O(f(n)) + O(g(n))$ is $O(f(n) + g(n))$
- for example,

```
for i in range(n):  
    print('a')  
for j in range(n*n):  
    print('b')
```

$O(n)$
 $O(n^2)$

$O(n) + O(n^2)$



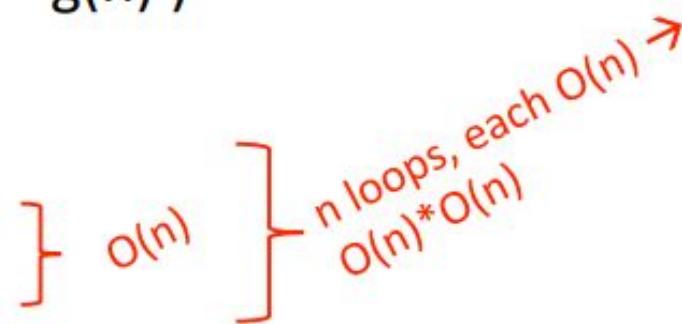
is $O(n) + O(n^2) = O(n+n^2) = O(n^2)$ because of dominant term

Law of multiplication

Law of Multiplication for O():

- used with **nested** statements/loops
- $O(f(n)) * O(g(n))$ is $O(f(n) * g(n))$
- for example,

```
for i in range(n):  
    for j in range(n):  
        print('a')
```



is $O(n)*O(n) = O(n*n) = O(n^2)$ because the outer loop goes n times and the inner loop goes n times for every outer loop iter.

Complexity Growth

CLASS	n=10	= 100	= 1000	= 1000000
O(1)	1	1		1
O(log n)	1	2		3
O(n)	10	100		1000
O(n log n)	10	200		3000
O(n^2)	100	10000		1000000000000
O(2^n)	1024	12676506 00228229 40149670 3205376	1071508607186267320948425049060 0018105614048117055336074437503 8837035105112493612249319837881 5695858127594672917553146825187 1452856923140435984577574698574 8039345677748242309854210746050 6237114187795418215304647498358 1941267398767559165543946077062 9145711964776865421676604298316 52624386837205668069376	Good luck!!