

# ALGORITHM ANALYSIS / BIG-OH

# WHY ANALYZE ALGORITHMS?

- CPU time (effort) is a valuable resource
- There can be multiple different algorithms and implementations to solve a problem
- Want to make sure we choose an efficient approach
  - We'll focus on efficiency with respect to operations performed by CPU
  - Can also choose / analyze with respect to others such as space on computer used
  - Often tradeoff between these

# ALGORITHM

- Not the actual code (aka "implementation")
- The steps you take to solve the problem
- Analysis should happen:
  - at the algorithm stage
  - before you start coding

# COMPLEXITY

- How does the time depend on the problem size?
  - Use the phrase "time", but we don't really care about exact execution time
  - Look at growth function that says how time behaves
  - Actually only care about **asymptotic complexity**
    - Long term behavior as  $n$  increases
    - Based on dominant term
- Need to defined problem size ( $n$ )
  - Ex: # elements in array

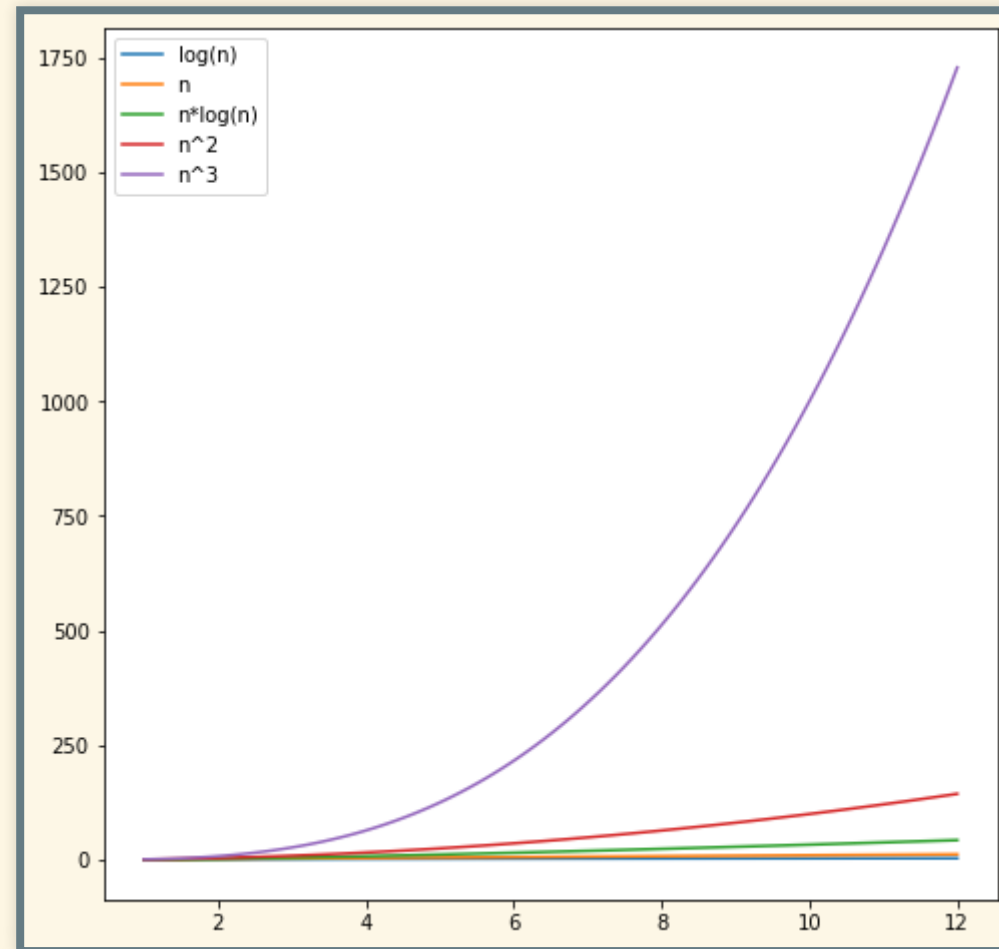
# COMPLEXITY

- Aka **order** of algorithm
- Big-Oh notation
  - $O(n^2)$  -> behaves like  $n^2$  as  $n$  increased
- Big-Oh formally:
  - $f(n)$  is  $O(g(n))$  if there exists  $c$  and  $m$  s.t.  
 $f(n) < c g(n)$  for all  $n > m$
- Basically, upper bound to growth function

# EXAMPLES

- $2n^3 + 12n^2 - n + 4$
- $4n^2 + 15n$
- $n + n \cdot \log(n) + 6$
- $\log(n) + 12$

# COMPLEXITY VISUALIZATIONS



# COMPLEXITY VISUALIZATIONS

