2/7 Asymptotic Notation

(formal way to model growth of functions)

Three Notations:

1. Big-Oh: O()

2. Big-Omega: Ω()

3. Theta: thetasymbol()

Big-Oh

*Represents the upper bound of an algorithm's running time

- Formal Definition: The function f(n) = O(g(n)) if
 - revE (there exists) constants c and n0 such that..
 - $f(n) \le c * g(n)$
 - For all $n \ge n0$
 - Then it will be true
 - Usually we think n>=1 since we don't worry abt empty/neg sized arrays/data (works if nothing if specified)

Ex: math example

- f(n) = 10n+100
- A c*g(n) combo that is >= f(n) for n >= 1
 - 10n + 100 <= 110n for all n>=1
 - 10n+100 = O(n) [c = 110, g(n) = n]
 - Trick to quickly find function that satisfies:
 - Take summation of highest degree terms by looking at the f(n)
 - Eg: $4n + 7 \le 4n + 7n \rightarrow 4n + 7 \le 11n$ for all $n \ge 1$
 - 4n+7 = O(n) [c=11, g(n) = n]
- ** but we cant we can say that for f(n) = 10n+100
 - $10n+100 \le 110n^2$, for all n>=1 i.e $f(n) = O(n^2)$
 - Works for higher order like O(n^3)
 - Yes it can
- But we try to use a g(n) that as close as possible , b/c its not helpful to say i am shorter than an giraffe
 - We need a frame of reference that is close

Big-Omega

*represents the lower bound of na algorithms running time

- Formal Def: the function $f(n) = \Omega g(n)$ if
 - There exists constants c and n0
 - $c*g(n) \le f(n)$
 - For all n >- n0

Ex: math example

- f(n) = 10n + 100
- Can you think of a c*g(n) combo that is less than/ equal to f(n) for n >=1
 - Possible solution: n <= 10n+100, for all n >=1
 - $-10n + 100 = \Omega(n)$
 - Similar to O(n) the following are also true
 - 10n + 100 = Ω (rt n)
 - 10n + 100 = Ω (log n)
 - $-10n +100 = \Omega(1)$
 - But we want to get as close of a bound for it to be helpful, there is no point saying i am definitely bigger than a mouse

Theta

*represents BOTH upper and lower bounds of an algorithms running time Formal Definition:

- The function $f(n) = \theta g(n)$ if..
 - $C1 * g(n) \le f(n) \le c2 * g(n)$
 - For all $n \ge n0$

Ex: 10n+ 100

Alr know lower and upper bounds of this fn

Possible sol.

- n<= 10n +100 <= 110n, for all n >= 1
- C1 = 1, c2 = 110
- Then $10n + 100 = \theta(n)$
 - We can't say that $10n+100 = \theta(n^2)$ or .. $\theta(\log n)$ (for this def)

Visual Summary

Check slides

Ok to cross the cg(n), if its before the n0, for O(n), shows the upper bound

- Opposite of omega
- For θ , we're trying to trap f(n) between the two boundaries , if it isnt true before n0 it is fine, but after n0 everything is true
- N0 = 1, since its p much useless for us to do an empty array, which is special case

Practice:

$$f(n) = n^2 + 2n + 1$$
, $n^0 = 1$

Big-o: factor out the largest term?

- :
$$4n^2$$
, c = 4, g(n) = n^2 , \rightarrow **O(n^2)**

big-omega: 4, \rightarrow we want the closest one which is \rightarrow omega(n^2)

- Take the biggest term on its own.,
- n^2 <= n^2 + 2n +1, we want the closest one so, o(n) still works but we want the closest so omega(n^2)

theta :
$$1 <= n^2 + 2n + 1 <= 4n^2, \rightarrow$$

- Since $f(n) = O(n^2)$ and $f(n) = omega(n^2)$, we can conclude that $f(n) = theta(n^2)$

Ex2:

$$f(n) = n^2 \log n + 2n$$
; $(n^0 = 1)$

big -O: just take the highest order term

- O(n): $n^2 \log n + 2n \le n^2 \log n + 2n^2 \log n$
- O(n^2 logn)

big-Omega:

- omega(n^2 logn)

Theta:

- theta(n): since both big=0 and big-omega are n^2logn, we can conclude that
- Theta is n^2 logn

Ex3: practice finding out O(n), Omega(n), Theta

$$f(n) = n!, n0 = 1$$

O:
$$n!= n^*(n-1)^*(n-2) \rightarrow = n^*n^*n^*n^*n \rightarrow f(n) = O(n^2)$$

- O(n^n)

Omega:
$$1*1*1*1*1 \le n*(n-1)*(n-2)... \to f(n) = Omega(1)$$

- Omega(1), ** just because its a constant it is omega(1), not b/c it is 1

Theta:

- Since upper(big o) and lower(big omega) bounds are different we cannot define theta for this f(n)
- We can only say $f(n) = O(n^n)$ and f(n) = Omega(1)

Desmos might help on hw*

*Asymptotic Notation in Code

- We can say that the order of growth for the linear function(defined above) is O(n)

Best, worst, and average cases

- Linear searching (goes thru every value)
 - **Best case** if key is in first value (1 comparison)
 - In any size, consider it all, as array grows, comparisons don't
 - In asymptotic notation, the best case time complexity is O(1),
 Omega(1), Theta(1), can use all three do not confuse best and worse case with upper and lower bound
 - Worst Case if key is last or not in the array
 - As size of array grow, the # of comparisons grow
 - In asymptotic notation, worst case time complexity O(n)... works for ohm and they also since is the same
 - Average case if key is somewhere in the middle
 - Comparisons do increase as size of array increase
 - Avg comparisons (1+2+3...n)/2 = n(n+1)/n
 - Is also O(n).... though we probably would expect it to be n/2, but in big o , you drop constants