# **CSE 373**

#### **OCTOBER 4<sup>TH</sup> – ALGORITHM ANALYSIS**

## **TODAY'S LECTURE**

- Algorithm Analysis
  - Asymptotic analysis
  - bigO notation

#### **PROJECT 1**

- Checkpoint 1 due at 11:30 pm
- Submit only the files listed in the deliverables section
- If you submit as a group, make sure all files have both team names
- Helpful if you could add a comment on your canvas submission indicating your partner

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  - Analysis is for algorithms
  - Runtime, memory and correctness
  - Best case, average case, worst case
  - Over groups of inputs, not just one

Principles of analysis

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  - Determining performance behavior
  - How does an algorithm react to new data or changes?
  - Independent of language or implementation

- Example: find()
  - Sorted v Unsorted
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- What is the worst case?
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- How long does this take to run?

#### Consider the algorithm

}

```
public int binarySearch(int[] data, int toFind){
int low = 0; int high = data.length-1;
while(low <= high){
    int mid = (low+high)/2;
    if(toFind>mid) low = mid+1; continue;
    else if(toFind<mid) high = mid-1; continue;
    else return mid;
}
return -1;</pre>
```

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  - At the kth iteration?

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- How many iterations then? Solve for k.

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  - N can be things other than powers of two
  - Ceiling and floor rounding



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- If this isn't exact, is it still correct?
- Yes. We care about asymptotic growth.
  - How a the runtime of an algorithm grows with big data
- To incorporate this perspective, we use bigO notation

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- For example, if an algorithm A is
   O(log n), that means some logarithmic function upper bounds A.

- Formally, a function f(n) is O(g(n)) if there exists a c and n<sub>o</sub> such that:
- For all  $n \ge n_0$ , f(n) < c\*g(n)
- To prove a function is O(g(n)), simply find the c and n<sub>0</sub>

- Example: is  $5n^3 + 2n in O(n^4)$ ?
- Can we find a c,  $n_0$  such that:
- $5n^3$  +  $2n \leq c * n^4$  for all  $n \geq n_0$

- This is an upper bound, so if
- $5n^3 + 2n$  is in O( $n^4$ ), then
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- $ls 5n^3 + 2n in O(n3)?$
- Yes, let c be 7 and n > 1

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- For all  $n \ge n_0$ , f(n) > c\*g(n)

 If a function f(n) is in O(g(n)) and Ω(g(n)), then g(n) is a tight bound on f(n), we call this big theta.

- If a function f(n) is in O(g(n)) and Ω(g(n)), then g(n) is a tight bound on f(n), we call this big theta.
- Formally, iff f(n) is in O(g(n)) and  $\Omega(g(n))$ , then f(n) is in  $\theta(g(n))$
- Note that the two will have different c and n<sub>0</sub>

- What does this help us with?
  - Sort algorithms into families

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  - Sort algorithms into families
    - O(1): constant
    - O(log n): logarithmic
    - O(n) : linear
    - O(n<sup>2</sup>): quadratic
    - O(n<sup>k</sup>): polynomial
    - O(k<sup>n</sup>): exponential

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  - Remember that log<sub>a</sub> k and log<sub>b</sub> k differ by a constant factor?

- What does this help us with?
  - The constant multiple c lets us organize similar algorithms together.
  - Remember that log<sub>a</sub> k and log<sub>b</sub> k differ by a constant factor?
  - That makes all logs in the same family

## **NEXT CLASS**

- Recurrence Relations
  - How to analyze recursively defined functions
- Analyzing the naïve dictionary implementations