

CSCI 136:
Data Structures
and
Advanced Programming

Lecture 9

Asymptotic analysis

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Announcements

- Lab 3: what's the deal with loops?
- Lab 1: feedback sent
- Lab 1: if feedback has mistakes...



Outline

1. Quiz
2. Study tip
3. Asymptotic analysis

Quiz

Code Review

Study tip #3: vocabulary

Every course is a "foreign language"

To learn effectively, study the vocab.

Maintain a "glossary."



Asymptotic analysis

How do we know if an algorithm is faster than another?



Why can't we just measure "wall time"?

Why can't we just measure "wall time"?

- Other things are happening at the same time
- Total running time usually varies by input
- Different computers may produce different results!

Let's just count instructions, then

- What do we count?
 - Count all computational steps?
 - What is a "step"?
 - What about steps inside loops?

Stepping back...

- How accurate do we need to be?
 - If one algorithm takes 64 steps and another 128 steps, do we need to know the precise number?

What we do

Instead of precisely counting steps, we usually develop an **approximation** of a program's **time** or **space complexity**.

This approximation **ignores tiny details** and focuses on the big picture: *how do time and space requirements grow as a function of the size of the input?*

Example

```
// pre: array length n > 0
public static int findPosOfMax(int[] arr) {
    int maxPos = 0
    for (int i = 1; i < arr.length; i++)
        if (arr[maxPos] < arr[i]) maxPos = i;
    return maxPos;
}
```

- Can we count steps exactly? Do we even want to?
 - **if** complicates counting
- Idea: **overcount**: assume **if** block always runs
 - in the worst case, it does
- Overcounting gives **upper bound** on run time
- Can also **undercount** for **lower bound**

Overcounting Example

```
// pre: array length n > 0
public static int findPosOfMax(int[] arr) {
    int maxPos = 0 ..... line 1 cost:  $c_1$ 
    for (int i = 1; i < arr.length; i++) ..... line 2 cost:  $nc_2$ 
        if (arr[maxPos] < arr[i]) ..... line 3 cost:  $nc_3$ 
            maxPos = i; ..... line 4 cost:  $nc_4$ 
    return maxPos; ..... line 5 cost:  $c_5$ 
}
```

$$\begin{aligned} \text{Total cost: } & c_1 + nc_2 + nc_3 + nc_4 + c_5 \\ & = c_1 + n(c_2 + c_3 + c_4) + c_5 \\ & = n(c_2 + c_3 + c_4) + c_1 + c_5 \\ & = O(n) \\ & \text{(as you shall see)} \end{aligned}$$

Focus is on order of magnitude

We can do this analysis for the **best**, **average**, and **worst** cases. We often focus on the worst case.

Big-O notation

Let **f** and **g** be real-valued functions that are defined on the same set of real numbers. Then **f** is of order **g**, written **f(n) is O(g(n))**, if and only if there exists a positive real number **c** and a real number **n₀** such that for all **n** in in the common domain of **f** and **g**,

$$|f(n)| \leq c \times |g(n)|, \text{ whenever } n > n_0.$$

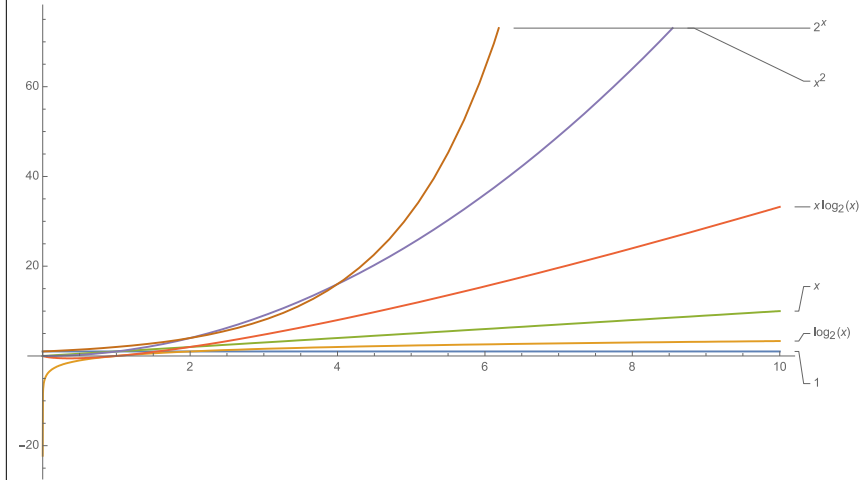
We read this as: "**f(n) is O(g(n))**"
as "**f** of **n** is big-oh of **g** of **n**."

Function growth

Consider the following functions, for $x \geq 1$

- $f(x) = 1$
- $g(x) = \log_2(x)$ // Reminder: if $x=2^n$, $\log_2(x) = n$
- $h(x) = x$
- $m(x) = x \log_2(x)$
- $n(x) = x^2$
- $p(x) = x^3$
- $r(x) = 2^x$

Function growth



Function growth & Big-O

- Rule of thumb: **ignore multiplicative constants**
- Examples:
 - Treat n and $n/2$ as same order of magnitude
 - $n^2/1000$, $2n^2$, and $1000n^2$ are "pretty much" just n^2
 - $a_0n^k + a_1n^{k-1} + a_2n^{k-2} + \dots + a_k$ is roughly n^k
- The key is to find the **most significant** or **dominant term**
- Ex: $\lim_{x \rightarrow \infty} (3x^4 - 10x^3 - 1)/x^4 = 3$ (Why?)
 - So $3x^4 - 10x^3 - 1$ grows "like" x^4

Recap & Next Class

Today we learned:

Intro to asymptotic analysis

Next class:

Big-O notation

Induction