

Course Notes for
CS 1501
Algorithm Implementation

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- These notes are intended for use by students in CS1501 at the University of Pittsburgh and no one else
- These notes are provided free of charge and may not be sold in any shape or form
- These notes are NOT a substitute for material covered during course lectures. If you miss a lecture, you should definitely obtain both these notes and notes written by a student who attended the lecture.
- Material from these notes is obtained from various sources, including, but not limited to, the following:
 - Algorithms in C++ by Robert Sedgewick
 - Algorithms, 4th Edition by Robert Sedgewick and Kevin Wayne
 - Introduction to Algorithms, by Cormen, Leiserson and Rivest
 - Various Java and C++ textbooks
 - Various online resources (see notes for specifics)



- CS 1501 Summer 2020 will be **delivered online**
 - ▶ Please review all course policies in the following location:
 - <https://people.cs.pitt.edu/~ramirez/cs1501/>
 - Click on the **Course Info** link
 - ▶ These policies will be discussed during the first synchronous lecture:
 - Monday, May 11, 2020, 10:20AM-11:15AM
 - ▶ After reading these policies, **if you have any questions / concerns / problems / or need any accommodation, please contact me ASAP**



- **Office Hours:**
 - Office hours will be held over Zoom
 - My office hours for the Summer Term will be:
 - Monday, 2:30PM-3:30PM
 - Tuesday, 10:00AM-12:00PM
 - Wednesday, 2:30PM-3:30PM
 - and by appointment
 - A **virtual waiting room** will be utilized and students will be seen one at a time
 - See email for Zoom link information
 - Your TA will also have office hours
 - See email for Zoom link information



- **Text:**

- ▶ **Algorithms Fourth Edition** by Sedgewick and Wayne [Pearson]

- See:

- <http://www.pearsonhighered.com/educator/product/Algorithms/9780321573513.page>

- An **e-copy should be available via inclusive access** from Pitt on Canvas

- **You will automatically be billed for this e-copy**

- If you would prefer to get it on your own, you can **opt out of this** and will not have to pay.

- This is done via Canvas in the e-book environment or by emailing the University Store: theuniversitystore@pitt.edu



- Definitions:

- ▶ **Offline Problem:**

- We provide the computer with some input and after some time receive some acceptable output
 - No hard deadline for *when* the problem should be solved (as opposed to a real-time problem)

- ▶ **Algorithm**

- A step-by-step procedure for solving a problem or accomplishing some end

- ▶ **Program**

- an algorithm expressed in a language the computer can understand

- ▶ An algorithm solves a problem if it produces an acceptable output on **EVERY** input



- Goals of this course:
 - 1) To learn **how to convert (nontrivial) algorithms into programs**
 - ▶ Often what seems like a fairly simple algorithm is not so simple when converted into a program
 - ▶ Other algorithms are complex to begin with, and the conversion must be carefully considered
 - ▶ Many issues must be dealt with during the **implementation process**
 - ▶ We will get a few ideas in our interactive lecture



- 2) To see and understand **differences in algorithms and how they affect the run-times** of the associated programs
- Many problems can be solved in more than one way
 - Different solutions can be compared using many factors
 - One important factor is program **run-time**
 - > Sometimes a better run-time makes one algorithm more desirable than another
 - > Sometimes a better run-time makes a problem solution feasible where it was not feasible before
 - However, there are **other factors** for comparing algorithms
 - > We will discuss some in the interactive lecture



- Asymptotic Analysis

- ▶ First we should determine what we want to analyze
 - Usually it is time (as in run-time) but not always
 - What other **resource** may we want to keep track of when a program is run?
- ▶ For time we determine a **key instruction** (or instructions) which drives the overall run-time
- ▶ Determine a **function** that models the overall run-time behavior of the algorithm
 - Ex: $F(N) = 2N^2 + 6N + 100$



$$F(N) = N^2$$

- ▶ But we will ignore / remove
 - lower order terms
 - multiplicative constants
 - Why?
 - Discuss and see notes below this slide
- ▶ Use some standard measure for categorization
 - Do we know any measures?
 - Big O
 - You should be familiar with this from CS 0445
 - However, there are other ways to categorize asymptotic run-times



- **Big O** gives us an **upper bound** on the asymptotic performance
 - Think: $F(N) \leq 2N^2 + 6N + 100 \Rightarrow O(N^2)$
- **Big Omega**
 - Gives us a **lower bound** on the asymptotic performance
 - Think: $F(N) \geq 2N^2 + 6N + 100 \Rightarrow \Omega(N^2)$
- **Theta**
 - **Upper and lower bound** on the asymptotic performance – **exact bound**
 - Think: $F(N) = 2N^2 + 6N + 100 \Rightarrow \Theta(N^2)$
- Why three different measures?
 - When analyzing algorithms we cannot always be precise
 - Sometimes we must make assumptions / simplifications



Algorithm Analysis

- These make it difficult to determine an exact (Theta) bound
- We also may have different goals
 - Big Omega tells us "we cannot do better than this"
 - Big O tells us "we can definitely do this well or better"
 - Theta tells us "we can do exactly this well"
 - > All of these are expressed within a constant factor
- ▶ Ex: Comparison based sorting
 - It has been proven that any comparison based sorting algorithm is $\Omega(N \lg N) \rightarrow (n \log(n))$
 - This tells us that any new algorithm that we develop to sort which compares values will require at least this much time



- Some sorting algorithms achieve this lower bound (ex: MergeSort always, QuickSort in average case)
 - We can say MergeSort is $\Theta(N \lg N)$
- Others do not

► Generally speaking

- Big-O tends to be easier to determine than Big-Omega
 - And also Theta since Theta requires both lower and upper bounds
 - This is why Big-O is commonly used
- However, if we can be precise enough to give a Theta bound it is better



- **Tilde approximations and order of growth**
 - Text introduces an alternative notation
 - Read text Section 1.4 for rationale for it
 - Practically speaking, tilde is like Theta but without dropping the multiplicative constants
 - Ex: $f(N) = 6N^2 + 18N - 50$
 - $\Theta(N^2)$ (or order of growth N^2)
 - $\sim 6N^2$
 - The constants can be useful for real analysis, but are somewhat implementation dependent
 - As discussed in previous slides



- So is algorithm analysis really important?
 - ▶ Yes! Different algorithms can have considerably different run-times
 - Sometimes the differences are subtle but sometimes they are extreme
 - Let's look at a table of growth rates on the next slide
 - Note how drastic some of the differences are for large problem sizes
 - Note especially the last two columns
 - > Solutions with exponential and hyper-exponential run-times are not practical for even relatively small values of N



Algorithm Analysis

| $\lg_2 N$ | N | $N \lg N$ | N^2 | 2^N | $N!$ |
|-----------|------|-----------|-------------|------------|-------------|
| 2 | 4 | 8 | 16 | 16 | 24 |
| 3 | 8 | 24 | 64 | 256 | 40320 |
| 4 | 16 | 64 | 256 | 64K | > 20T |
| 5 | 32 | 160 | 1024 | 4G | > 10^{35} |
| ... | ... | ... | ... | ... | ... |
| 10 | 1024 | 10240 | 1M | 2^{1024} | PUNT |
| 20 | 1M | 20M | 1T | 2^{1M} | PUNT |
| 30 | 1G | 30G | > 10^{18} | 2^{1G} | PUNT |



- ▶ Consider 2 choices for a programmer
 - 1) Implement an algorithm, then run it to find out how long it takes
 - 2) Determine the asymptotic run-time of the algorithm, then, based on the result, decide whether or not it is worthwhile to implement the algorithm
- Which choice would you prefer?
- Discuss
- The previous few slides should be (mostly) review of CS 0445 material



- Example:

- ▶ ThreeSum example from text

- Given a set of arbitrary integers (could be negative)
 - Find out how many distinct triples sum to exactly zero
 - Ex: $A = 10\ 40\ -20\ 25\ -10\ -15\ 30\ 5\ -35$
 - One answer is 25, -10, -15, another is -20, -10, 30
 - Simple solution:
 - Triple for loops (keeping track of indices)
 - > For each i from 0 to $N-1$ in the list
 - > For each j from $i+1$ to $N-1$ in the list
 - > For each k from $j+1$ to $N-1$ in the list
 - Count solution if $A[i] + A[j] + A[k] = 0$
 - > See ThreeSum.java
 - Works but clearly is $\Theta(N^3)$



- Better solution:
 - Sort the numbers first (using a fast sort)
 - > How long?
 - > $\Theta(N \lg N)$
 - Now we add two numbers in our initial two loops and
 - > If we can **find in the array a number with the opposite value of that sum**, we will have a solution
 - > We can do this using **binary search**
 - > For each i from 0 to $N-1$
 - > For each j from $i+1$ to $N-1$
 - > Let $\text{sum} = A[i] + A[j]$
 - > Let $k = \text{binarySearch}(A, -\text{sum})$
 - > if $(k > j)$ answer is i, j, k
 - Ex: $A[i] = -20$ $A[j] = -10$
 - > We can do a binary search for the value 30. If we find it, we have a solution, otherwise we do not



Algorithm Analysis

- We are still using two nested for loops to get $A[i]$ and $A[j]$, but we eliminate the third loop and replace it with a binary search
 - > Improve the run-time from $\Theta(N^3)$ to $\Theta(N^2 \lg N)$
 - > BUT - we must add the time to sort, which we did not have to do before
 - > We must be careful not to ignore extra computation that we may introduce
 - > But we know a fast sort takes $N \lg N$ so it does not add to our asymptotic run-time
 - > This gives us $\Theta(N \lg N + N^2 \lg N) \Rightarrow \Theta(N^2 \lg N)$
- See ThreeSumFast.java in text
 - > Note: Both versions assume no duplicates. How would duplicates complicate the solution?
 - > See comment below in notes

