CSE 332 Summer 2024 Lecture 2: Algorithm Analysis pt.1

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Warm up:

- I have a pile of string
- I have one end of the string in-hand
- I need to find the other end in the pile
- How can I do this efficiently?

Algorithm Ideas



• Ideas:

Algorithm Running Times

- How do we express running time?
- Units of "time"
- How to express efficiency?



End-of-Yarn Finding

1. Set aside the already-obtained "beginning"



3. Separate the pile of yarn into 2 piles, note which connects to ______ the beginning (call it pile A, the other pile B)

R

Repeat on pile with end

4. Count the number of strands crossing the piles

5. If the count is even, pile A contains the end, else pile B does

Why Do Resource Analysis?

- Allows us to compare *algorithms,* not implementations
 - Using observations *necessarily* couples the algorithm with its implementation
 - My implementation on my computer takes more time than your implementation on your computer. Do you have a better algorithm or computer?
- We can predict an algorithm's running time before implementing
- Understand where the bottlenecks are in our algorithm

Goals for Algorithm Analysis

- Identify a *function* which maps the algorithm's input size to a measure of resources used
 - Input of the function: **sizes** of the input
 - Number of characters in a string, number of items in a list, number of pixels in an image
 - Output of the function: **counts** of resources used
 - Number of times the algorithm:
 - Adds two numbers together
 - does a > or < comparison
 - Number of bytes of memory needed
- Important note: Make sure you know the "units" of your domain (input size) and range (resource used)!

Analysis Process From 123/143

- Count the number of "primitive operations"
 - +, -, compare, arr[i], arr.length, etc
- Write that count as an expression using n (the input size)
- Put that expression into a "bucket" by ignoring constants and "nondominant" terms, then put a O() around it.
 - $4n^2 + 8n 10$ ends up as $O(n^2)$
 - $\frac{1}{2}n + 80$ ends up as O(n)
 - n(n+1) ends up as $O(n^2)$

Worst Case Analysis (in general)

- If an algorithm has a worst case resource complexity of f(n)
 - Among all possible size-n inputs, the "worst" one will use f(n) "resources"
 - f(n) gives the maximum count needed from among all inputs of size n

Worst Case Analysis (in general)

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 - Among all possible size-n inputs, the "worst" one will use f(n) "resources"
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Worst Case Running Time Analysis

- If an algorithm has a worst case **running time** of f(n)
 - Among all possible size-n inputs, the "worst" one will do f(n) "operations"
 - f(n) gives the maximum count of **operations** needed from among all inputs of size n

Worst Case Space Analysis

- If an algorithm has a worst case space complexity of f(n)
 - Among all possible size-n inputs, the "worst" one will need f(n) "memory units" (usually bits)
 - f(n) gives the maximum number of bits needed from among all inputs of size n

myFunction(List n){

b = 55 + 5;

b = c + 100;

c = b / 3;

Worst Case Running Time - Example

```
for (i = 0; i < n.size(); i++) {
```

```
b++;
```

```
}
```

```
if (b % 2 == 0) {
```

```
c++;
```

```
}
else {
```

```
for (i = 0; i < n.size(); i++) {
```

C++;

return c;

```
Questions to ask:
```

- What are the units of the input size?
- What are the operations we're counting?
- For each line:
 - How many times will it run?
 - How long does it take to run?
 - Does this change with the input size?

Worst Case Running Time – Example 2

```
beAnnoying(List n){
```

```
List m = [];
```

```
for (i=0; i < n.size(); i++){
```

m.add(n[i]);

```
for (j=0; j< n.size(); j++){
```

```
print ("Hi, I'm annoying");
```

Questions to ask:

- What are the units of the input size?
- What are the operations we're counting?
- For each line:
 - How many times will it run?
 - How long does it take to run?
 - Does this change with the input size?

```
return;
```

Worst Case Running Time – General Guide

- Add together the time of consecutive statements
- Loops: Sum up the time required through each iteration of the loop
 - If each takes the same time, then [time per loop × number of iterations]
- Conditionals: Sum together the time to check the condition and time of the slowest branch
- Function Calls: Time of the function's body
- Recursion: Solve a recurrence relation

Defining your running time function

- Worst-case complexity:
 - max number of steps algorithm takes on "most challenging" input
- Best-case complexity:
 - min number of steps algorithm takes on "easiest" input
- Average/expected complexity:
 - avg number of steps algorithm takes on random inputs (context-dependent)
- Amortized complexity:
 - max total number of steps algorithm takes on M "most challenging" consecutive inputs, divided by M (i.e., divide the max total sum by M).

Amortized Complexity Example - ArrayList

```
public void add(T value){
    if(data.length == size)
        resize();
    data[size] = value;
    size++;
}
private void resize(){
    T[] oldData = data;
    data = (T[]) new Object[data.length*2];
    for(int i = 0; i < oldData.length; i++)</pre>
        data[i] = oldData[i];
```

}

- What is the worst case running time of add?
 - Input size:
 - Operations counted:
 - 0(?)

Amortized Complexity Example - ArrayList

```
public void add(T value){
    if(data.length == size)
        resize();
                             Every time we resize, we earn
    data[size] = value;
                             data.length more adds
    size++;
                             guaranteed to not resize!
}
private void resize(){
    T[] oldData = data;
    data = (T[]) new Object[data.length*2];
    for(int i = 0; i < oldData.length; i++)</pre>
        data[i] = oldData[i];
```

• Amortized Analysis Idea:

- Suppose we have a program that in total does *n* adds.
- How much time was spent "on average" across all *n*?
- Let c be the initial size of data
 - The first c adds take: c + c = 2c
 - The next 2c adds: 2c + 2c = 4c
 - The next 4c adds: 4c + 4c = 8c

• Overall:
$$\frac{14c}{7c} = 2c$$