

Algorithm Analysis?

EMBRY-RIDDLE
Aeronautical University

- What is algorithm analysis?
 - Algorithm analysis is a means of comparing algorithms to one another
 - Given any problem, many sequences of steps may result in identical results
 - How do you select which is the “best”?

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Analysis Criteria include:

- Correctness
- Simplicity
- Time consumed
- Space used

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Scientific Method and Analysis

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- Often difficult to formulate a precise measure of an algorithm's performance
 - Varies based upon system inputs
 - Algorithms can be too complex to determine exact time or space performance
- Scientific method can be employed:
 - **Observe** – observe the system
 - **Hypothesize** – develop a model to approximate the system
 - **Predict** – Predict output given inputs
 - **Verify** – evaluate performance versus predictions
 - **Validate** – re-run until hypothesis and observations agree
- Hypothesis must be **falsifiable**
- Experimental design should permit repeated steps

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Scientific Method and Analysis

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In this chapter, the authors are demonstrating this process with the **ThreeSum** example:

- Observations are made
- Models and experiments are used to validate model
- Propositions are made as the authors provide examples of propositions made, analysis and each the text.



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Observation

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- Observation is not enough to, but observing run-time can help us develop our model
- e.g. Stopwatch / Wall clock time
 - Measurement of the runtime is one approach to observe an algorithms timing performance
 - What are some of the reasons that wall clock time can be problematic?
 - Cannot use results to produce generalized predictions of runtime
 - Conditions such as CPU, system load, compilation methods, available RAM, etc. can influence the runtime of software
 - More rigorous experimentation with statistical analysis needed to



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Analysis of Experimental Data

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- Given our metric, e.g. run-time, our experiment would need to vary the size of the problem being solved
 - e.g. size of the data set being processed, or some other run-time parameter
- Plotting timing results
 - Standard Plot
 - Logarithmic Plot
- Textbook provides an example of this technique called the *Double Rule*.

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Modeling Complexity

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- Donald Knuth observed:
 - Cost of executing each statement
 - How long does the op take?
 - Frequency of execution of each statement
 - How often is a particular operation made?
- Moving away from real-time we focus on approximation:
 - Assume that each read/write/access operation costs roughly one (1) time unit.
 - Each problem is of some size N , which is defined at runtime
 - E.g. an array's length would be of size N for an array processing problem.
- We seek to create as accurate model of the runtime prediction $f(N)$ as possible, where $f(N)$ is the number of operations performed as a function of problem size N .

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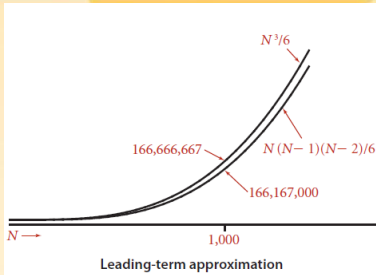
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Approximation of our model

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function	tilde approximation	order of growth
$N^3/6 - N^2/2 + N/3$	$\sim N^3/6$	N^3
$N^2/2 - N/2$	$\sim N^2/2$	N^2
$\lg N + 1$	$\sim \lg N$	$\lg N$
3	~ 3	1

Typical tilde approximations

Tilde Approximation:

from textbook:

Definition. We write $\sim f(N)$ to represent any function that, when divided by $f(N)$, approaches 1 as N grows, and we write $g(N) \sim f(N)$ to indicate that $g(N)/f(N)$ approaches 1 as N grows.

order of growth	
description	function
constant	1
logarithmic	$\log N$
linear	N
linearithmic	$N \log N$
quadratic	N^2
cubic	N^3
exponential	2^N

Com
order

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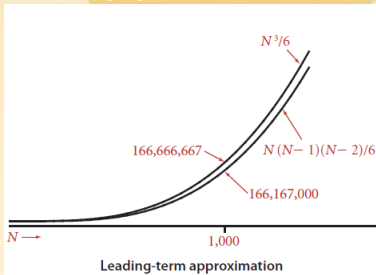
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Alternatively, we can represent our mathematical models with arithmetic approximation methods such as:

- Big-O
- Big-Omega
- Big-Theta

This approximations are similar, but different to tilde, and will be discussed later in this module.

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Com
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Example

```
public class ThreeSum
{
    public static int count(int[] a)
    {
        int N = a.length;
        int cnt = 0;
        for (int i = 0; i < N; i++)
        {
            for (int j = i+1; j < N; j++)
            {
                for (int k = j+1; k < N; k++)
                {
                    if (a[i] + a[j] + a[k] == 0)
                    {
                        cnt++;
                    }
                }
            }
        }
        return cnt;
    }

    public static void main(String[] args)
    {
        int[] a = In.readInts(args[0]);
        StdOut.println(count(a));
    }
}
```

blocks of statements

frequencies of execution

inner loop

Anatomy of a program's statement execution frequencies

statement block	time in seconds	frequency	total time
E	t_0	x (depends on input)	$t_0 x$
D	t_1	$N^3/6 - N^2/2 + N/3$	$t_1(N^3/6 - N^2/2 + N/3)$
C	t_2	$N^2/2 - N/2$	$t_2(N^2/2 - N/2)$
B	t_3	N	$t_3 N$
A	t_4	1	t_4

$$\begin{aligned} \text{grand total} &= (t_1/6) N^3 \\ &+ (t_2/2 - t_1/2) N^2 \\ &+ (t_1/3 - t_2/2 + t_3) N \\ &+ t_4 + t_0 x \end{aligned}$$

tilde approximation $\sim (t_1/6) N^3$ (assuming x is small)

order of growth N^3

Analyzing the running time of a program



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Example

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            for (int j = i+1; j < N; j++)
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                for (int k = j+1; k < N; k++)
                {
                    if (a[i] + a[j] + a[k] == 0)
                    {
                        cnt++;
                    }
                }
            }
        }
        return cnt;
    }

    public static void main(String[] args)
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tilde approximation $\sim (t_1/6) N^3$ (assuming x is small)

order of growth N^3

Analyzing the running time of a program (example)

We will need to evaluate our analysis experimentally to confirm, but this seems like a reasonable estimate.

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