





The first claim of this course...



"Bad programmers worry about the <u>code</u>. Good programmers worry about <u>data structures</u> and their relationships."

(by Linus Trovals, creator of Linux kernel)

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Some Terminology



- Algorithm
 - □ A high-level <u>description</u> of a piece-by-piece or stepby-step process.
- Abstract Data Type (ADT)
 - □ <u>Formal</u> (<u>Mathematical</u>) <u>description/model</u> of components of the data and the set of operations that are allowed.
- Data structure
 - $\hfill \Box$ A specific $\underline{\text{organization}}$ of data for implementing an ADT
- Implementation
 - □ A specific <u>implementation</u> of a data structure in a specific programming language.



The Definition of an Algorithm



Brief and concise definition:

An **algorithm** is a clearly specified set of simple instructions to be followed to solve a problem.

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The Definition of an Algorithm



A more <u>complete</u> definition:

An **algorithm** is a well-ordered collection of clear and effectively computable operations that when executed produces a result and halts in a finite amount of time*.

*Schneider, M. and J. Gersting (1995), *An Invitation to Computer Science*, West Publishing Company, New York, NY, p. 9.





The Definition of an Algorithm

- Some relevant points of this definition:
 - ☐ An algorithm is a **well-ordered** collection...
 - ☐ An algorithm has **clear** operations...
 - ☐ An algorithm has **effectively computable** operations...
 - ☐ An algorithm **produces a result**...
 - □ An algorithm halt in a finite amount of time

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Algorithm Efficiency



- How efficient is an algorithm?
 - □ **Efficiency** covers lots of resources, including:
 - CPU (running time) usage
 - Memory (internal) usage
 - Disk (external memory) usage
 - Network usage
 - □ All are important but we will mostly talk about the **time complexity** (**CPU** usage).



Algorithm CPU Efficiency



- Performance vs Complexity?
 - □ Performance:
 - How much time is used when a program is executed.
 - Please note that this depends on the computer, operating system, compiler, etc.
 - □ Complexity:
 - What happens as the size of the problem being solved gets larger?
 - We want to evaluate the "quality" of the algorithms independent of their implementation.

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Algorithm Analysis



- In this module we shall discuss:
 - ☐ How to **estimate** the **running time** required for an **algorithm**.
 - □ How to reduce the running time of an algorithm (from days or years to fractions of a second).
 - ☐ The results of careless use of **recursion**.
 - □ Some examples of **efficient algorithms**.



Algorithm Analysis



Very Important!

- When we are trying to find the complexity of an algorithm, we are not interested in the exact number of operations that are necessary to solve the problem.
- □ We are interested in the relation of the number of operations to the problem size
- □ We are usually interested in the **worst case** the maximum number of operations that are necessary to be performed for a given problem size.

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Algorithm Analysis



- Most of the problems we will study have an **input size**.
- Example:
 - □ For the problem of sorting, the size of the input is the number of elements to be sorted.
- The CPU time and space (internal memory) used by an algorithm will in general be a function of this input size.
- For a given input size n we often express the time T to execute the algorithm as a function of n, written as T(n).
- We will always assume <u>T(n) is a non-negative value</u> <u>function</u>.





Algorithm Analysis

Example:

- Consider a simple algorithm to solve the problem of finding the largest (max) value in an array of n integers.
- The algorithm looks at each integer in turn, saving the position of the largest value seen so far.
- The **total time** to run this algorithm is approximately **cn**, because we must make **n** comparisons, with each comparison costing **c** time.
- Then for this algorithm **T(n)** = **cn**

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Algorithm Analysis



Example:

- □ An algorithm with running time $T(n) = n^2$ requires 1024 x 1024 = 1,048,576 time steps for an input of size n = 1024.
- □ An algorithm with running time T(n) = nlogn requires
 1024 x 10 = 10,240 time steps for an input of size
 n = 1024.
- □ It's easy to verify that $n^2 > 10$ nlogn when n > 58.





You would be **much better** off **changing algorithms** instead of **buying a computer ten times faster**.

