Programs: Algorithms + Data Structures

- **Algorithm:** a finite set of instructions that, if followed, accomplishes a general, well-specified task.
 - o **It has:** Input/Output
 - o It should be: Clear, finite and correct
 - o Good algorithms are efficient and easy to implement
 - o **Example:** Euclid algorithm for GCD calculation
- Classifying algorithms by:
 - o **Problem domain:** numeric, text processing, sorting, searching, networks, machine learning, etc.
 - o **Design Strategy:** divide and conquer, greedy, dynamic programming, backtracking, etc.
 - o Complexity: constant, linear, quadratic, cubic, exponential, etc.
 - o **Implementation Dimensions:** sequential, parallel, recursive, iterative, etc.
- **Data type:** Set of values and operations on those values.
 - o **Primitive data types:** integer, float, Boolean, character, string*, pointer*, etc.
 - o Complex data types (class): employee, list, stack, etc.
- Data Structures: Arrangement of data for being able to store and retrieve information.
 - Choosing the right one depends on:
 - Is it filled completely at the beginning? Are there insertions/deletions/lookups/updates?
 - Will the items be processed in order? Do we need random access?
 - o **Example:** List, BST, Hash Table, etc.
- Efficient software minimizes coding/debugging/running/system integration time and memory.
- **Pseudocode:** A high-level description of an algorithm
 - o It is **more structured** than English and **less detailed** than a program.
- **Algorithm analysis:** analyzing how the resource requirements of an algorithm will scale when increasing the input size.
 - We asymptotically analyze the worst case time complexity T(n) of algorithms using RAM model.
 - We do **theoretical analysis** because experimental analysis takes much time for coding and testing and depends on the hardware/software.
 - \circ **Theoretical analysis**: characterizing running time as a function of the input size T(n)
 - It can be done on a pseudocode independently from the hardware/software environment
 - o To measure T(n) we need to know **execution time** and **frequency count** for every instruction
 - T(n) [approximately] = **Execution time * Frequency count**
 - Execution time is measures using **RAM model**:
 - Each simple operation (+, *, -, =, if, call) takes exactly one-step.
 - Loops and subroutines are not considered simple operations.
 - Each memory access takes exactly one time stamp.
 - \circ Example: Insertion sort algorithm $T(n) = a.n^2 + b.n + c$ (where a,b,c are constants)
- **Big Oh Analysis:** O(n) = T(n) ignoring constant factors and lower-order terms
 - \circ f(n) = O(g(n)) means c . g(n) is an upper bound on f(n).
 - $\exists c, n_0 > 0 \mid f(n) < c.g(n)$ $\forall n > n_0$
 - o $f(n) = \Omega(g(n))$ means c.g(n) is a lower bound on f(n).
 - $\exists c, n_0 > 0 \mid f(n) > c.g(n)$ $\forall n > n_0$
 - o $f(n) = \Theta(g(n))$ means $c_1.g(n)$ is an upper bound on f(n) and $c_2.g(n)$ is a lower bound on f(n)
 - $\exists c_1, c_2, n_0 > 0 \mid c_1.g(n) < f(n) < c_2.g(n)$ $\forall n > n_0$
 - This means that g(n) provides a nice, tight bound on f(n).
- **Space Complexity:** Determining how much space an algorithm requires by analyzing its storage requirements as a function of the input size.

Elementary data structure:

- ADT vs DS
- List
 - o Properties:
 - o Methods:
 - o ArrayList
 - o LinkedList
- Stack
 - o Properties:
 - o Methods:
 - o Array Implementation
 - **o** Linked-List Implementation
- Queue
 - o **Properties:**
 - o Methods:
 - o Array Implementation
 - o Linked-List Implementation
- Map/Dictionary ADT
 - o **Properties:**
 - o Methods:
 - o Implementations
 - TreeMap
 - HashMap
 - Hash function:
 - o Ideal hash function
 - Hash code:
 - o Memory address
 - o Integer cast
 - o Component sum
 - o Polynomial accumulation
 - Hash table:
 - Compression function:
 - **Our Contract of Street Contract**
 - o Multiply, add and take the modulus
 - Collisions:
 - Collision handling strategies:
 - Separate chaining
 - Open addressing
 - Double hashing
 - Analysis of collision handling strategies
 - Load factor

Algorithmic Strategies:

- Commonly used algorithmic strategies:
 - o Brute Force:
 - Examples
 - o Divide-and-conquer
 - o Dynamic programming
 - o Greedy algorithms
- Recursion:
- Recurrence relations
 - o Master Theorem
 - Case1:
 - Case2:
 - Case3:
- Maximum subarray problem:
 - o Using brute force
 - Optimized brute force
 - o Divide-and-conquer
 - o Dynamic programming

• Sorting algorithms:

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• Binary Search Trees:

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• Binary heap

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- Graphs:
 - o Representation:
 - o Traversal algorithms:
 - BFS
 - DFS
 - **o** Common problems:
 - Connected components
 - Minimum spanning tree
 - Kruskal
 - Prim
 - Shortest distance
 - Dijkstra
 - Bellman-Ford
 - Floyd Warshall
 - Max Flow/Min cut
 - Check document in Materials folder.