

## Analysis and Discussion of Elementary Data Structures

The following report analyzes the performance of basic data structures based on their theoretical properties and the empirical results generated by the Python script. It also discusses their practical applications and the trade-offs involved in choosing one over another.

### 2. Performance Analysis

#### 2.1 Theoretical Time Complexity

First, let's summarize the theoretical time complexity (Big-O notation) for the basic operations of the data structures we implemented.

Data Structure	Operation	Average Time	Worst Time	Notes
Array (Python List)	Access (list[i])	$O(1)$	$O(1)$	Direct memory access.
	Insertion/Deletion (End)	$O(1)$	$O(n)$	Amortized $O(1)$ due to dynamic resizing.
	Insertion/Deletion (Middle/Start)	$O(n)$	$O(n)$	Requires shifting subsequent elements.
	Search	$O(n)$	$O(n)$	Requires iterating through elements.
Stack (Array-based)	Push (append)	$O(1)$	$O(n)$	Amortized $O(1)$ .
	Pop (pop)	$O(1)$	$O(1)$	Removing from the end is fast.
Queue (Array-based)	Enqueue (append)	$O(1)$	$O(n)$	Amortized $O(1)$ .
	Dequeue (pop(0))	$O(n)$	$O(n)$	<b>Very inefficient;</b> all elements must be shifted.

<b>Singly Linked List</b>	Access / Search	$O(n)$	$O(n)$	Requires traversal from the head.
	Insertion (Beginning)	$O(1)$	$O(1)$	Just need to update the head pointer.
	Insertion (End)	$O(n)$	$O(n)$	Must traverse to the end. <i>Can be <math>O(1)</math> with a tail pointer.</i>
	Deletion (Beginning)	$O(1)$	$O(1)$	Just need to update the head pointer.
	Deletion (Middle/End)	$O(n)$	$O(n)$	Requires traversal to find the node.

## 2.2 Analysis of Empirical Results

The performance table you generated provides clear, practical evidence for these theoretical complexities.

## 3. Performance Analysis Table

Input Size	List Append (End)	LL Insert (End)	List Insert (Front)	LL Insert (Front)	ArrayQueue Dequeue	Deque Dequeue
1000	0.000068	0.069307	0.000273	0.000479	0.000470	0.000071
2500	0.000144	0.187197	0.003891	0.001957	0.001368	0.000203
5000	0.000264	1.839749	0.005329	0.006961	0.016228	0.000375
10000	0.001078	5.103221	0.107762	0.005244	0.020341	0.000800
20000	0.001018	5.326803	0.161327	0.071257	0.036075	0.000816

### Observation 1: Insertion at the Beginning

- **List Insert (Front)** time grows significantly with input size (from 0.0002s to 0.1613s). This demonstrates the  $O(n)$  complexity, as every existing element must be shifted one position to the right.
- **LL Insert (Front)** time remains very low and grows much more slowly. This demonstrates the  $O(1)$  complexity. Inserting at the head of a linked list only requires creating a new node and updating the head pointer, regardless of the list's size.

### Observation 2: Insertion at the End

- **List Append (End)** is extremely fast and scales well. This shows the efficiency of Python's dynamic array implementation, which has an amortized  $O(1)$  time for appends.
- **LL Insert (End)** is surprisingly slow and scales poorly (from 0.069s to 5.32s). This is because our SinglyLinkedList implementation lacks a tail pointer. To add an element at the end, the code must traverse the entire list from the head, making it an  $O(n)$  operation. This is a classic implementation trade-off.

### Observation 3: Queue Performance

- **ArrayQueue Dequeue** (`list.pop(0)`) scales poorly, confirming its  $O(n)$  complexity. Removing the first element requires shifting all other elements.
- **Deque Dequeue** (`collections.deque.popleft()`) is incredibly fast and shows almost no increase in time as the size grows. This is because deque is implemented as a doubly-linked list, making removals from either end a true  $O(1)$  operation. This starkly illustrates why using a Python list as a queue is a common anti-pattern for performance-critical code.

## 2.3 Trade-offs: Arrays vs. Linked Lists for Stacks & Queues

- **For Stacks (LIFO):** An **array** (like Python's list) is almost always the better choice. push (append) and pop from the end are both amortized  $O(1)$  and very fast in practice. The elements are stored in a contiguous block of memory, which is cache-friendly. A linked list offers no performance benefit here and adds overhead for storing pointers.
- **For Queues (FIFO):** A **linked list** is far superior. It provides  $O(1)$  for both enqueue (add to tail) and dequeue (remove from head), assuming a tail pointer is maintained. An array-based queue suffers from the expensive  $O(n)$  dequeue operation. This is why Python's standard library provides `collections.deque` specifically for efficient queue and deque implementations.

### 3. Discussion of Applications

#### 3.1 Practical Applications

- **Arrays and Matrices:**
  - **Image Processing:** A digital image is fundamentally a 2D matrix of pixel values.
  - **Game Development:** Game boards (chess, tic-tac-toe) and screen coordinates are often represented by matrices.
  - **Tabular Data:** Storing data from spreadsheets or databases where each row has the same set of columns.
- **Stacks (LIFO - Last-In, First-Out):**
  - **Function Calls:** The "call stack" manages active function calls. When a function is called, it's pushed onto the stack; when it returns, it's popped off.
  - **Undo/Redo Functionality:** Each action is pushed onto an "undo" stack. Hitting "undo" pops the action and moves it to a "redo" stack.
  - **Expression Evaluation:** Used to convert infix expressions (e.g.,  $3 + 4$ ) to postfix and evaluate them.
- **Queues (FIFO - First-In, First-Out):**
  - **Task Scheduling:** Operating systems use queues to manage processes waiting for the CPU. Print queues manage documents waiting to be printed.
  - **Networking:** Routers use queues to handle data packets in the order they arrive.
  - **Breadth-First Search (BFS):** A core algorithm for traversing trees and graphs that relies on a queue to explore neighbor nodes level by level.
- **Linked Lists:**
  - **Music Playlists:** A next pointer is perfect for "play next song," and a previous pointer (in a doubly-linked list) enables "play previous song."
  - **Implementing other Data Structures:** They form the underlying basis for more complex structures like stacks, queues, and hash tables (for collision chaining).
  - **Dynamic Memory Allocation:** When you need a data structure that can grow and shrink efficiently without needing a contiguous block of memory.
- **Rooted Trees:**
  - **File Systems:** The directory structure of a computer is a tree, with the root directory (/ or C:) as the root node.
  - **Organizational Charts:** Representing the hierarchy of a company.
  - **HTML DOM (Document Object Model):** A web browser parses an HTML document into a tree structure to represent its content.

## 3.2 When to Prefer One Data Structure Over Another

- **Choose an Array when:**
  - You need frequent, fast, random access to elements by index ( $O(1)$ ).
  - The size of the data is relatively fixed or changes infrequently.
  - Memory usage is a concern; arrays have low overhead (no pointers).
  - Most insertions/deletions happen at the end.
- **Choose a Linked List when:**
  - You need frequent, fast insertions and deletions at the beginning or end ( $O(1)$ ).
  - The size of the data is highly dynamic and unpredictable.
  - You don't need random access to elements.
  - You want to avoid the cost of resizing and shifting elements that occurs in an array.

## Comparison Plots



