

Assignment 6: Medians and Order Statistics and Elementary Data Structures

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## Introduction

This assignment focuses on the implementation and analysis of algorithms related to order statistics, more specifically determining the k-th smallest element in an unsorted array. Two approaches are explored: Deterministic Selection using the Median of Medians method and Randomized Selection using Quickselect. Algorithms like this are useful for when the objective is to locate a specific ranked element without sorting the entire dataset, which is not as efficient for larger inputs.

Along with this, the assignment should demonstrate the implementations of elementary data structures including arrays, queues, stacks, and singly linked lists. These structures serve as core building blocks in computer science and are essential for managing data efficiently in real-world applications. (Cormen et al, 2022)

### Part 1: Selection Algorithms

#### Deterministic Selection (Median of Medians)

The deterministic selection algorithm ensures linear time complexity  $O(n)$  in the worst case. It does this by dividing the input array into groups of five elements, while finding the median of each group, then looping and selecting the median of those medians to use as the pivot.

The key implementation used in this assignment was this function below:

```
22
23  def deterministic_select(arr: List[int], k: int) -> int:
24      """Median-of-medians select. Returns k -th smallest (1-based)."""
25
```

This function confirms the input, then calls upon the internal looped function select() that then performs the Median of Medians logic. As stated above, the array then divides into groups

of five; medians are extracted, and a pivot is selected to partition the array into lower high values.

## Randomized Selection (Quickselect)

Randomized Selection picks a pivot by chance rather than a more structured one. Even though it's worst-case complexity can degrade to  $O(n^2)$ , instead its expected performance is  $O(n)$ , and usually performs better than the deterministic method in practice.

The main function implemented is:

```
51
52  def randomized_select(arr: List[int], k: int) -> int:
53      """Randomized Quickselect. Returns k -th smallest (1-based)."""
54
```

The function above used the helper function quickselect() and selects a pivot index at random using random.randint(). The partitioned array and recursion continued on until the k-th smallest position can be obtained.

## Experimental Results

Several timed experiments were conducted using different input sizes and distributions.

The function run\_selection\_experiments() created results for:

- Sorted arrays
- Random arrays
- Reverse sorted arrays

Timing results then showed that the randomized selection consistently performed faster than deterministic selection, demonstrating its efficiency for practical use cases even though it lacks worst-case guarantees.

## Part 2: Elementary Data Structures

The following array and matrices operations were implemented:

- Insertion using array\_insert()
- Direct access using array\_access()
- Deletion using array\_delete()
- Matrix access and modification using matrix\_get() and matrix\_set()

Example:

```
130
131  def array_insert(a: List[Any], index: int, value: Any) -> None:
132      """Inserts values at index in array a (list). O(n -index)."""
133      a.insert(index, value)
```

## Stack Implementation

I used a stack implementing using Python's list structure. This stack followed Last-In and First-Out (LIFO) based principles. The core operations include:

- Pop()
- Push()
- Peek()
- Is\_empty()

Example:

```
152
153 class Stack:
154     """Stack using Python list as array."""
155
156     def __init__(self):
157         self.data: List[Any] = []
158
159     def push(self, item: Any) -> None:
160         self.data.append(item)
161
162     def pop(self) -> Any:
163         if not self.data:
164             raise IndexError("pop from empty stack")
165         return self.data.pop()
166
167     def peek(self) -> Any:
168         if not self.data:
169             raise IndexError("peek from empty stack")
170         return self.data[-1]
171
172     def is_empty(self) -> bool:
173         return len(self.data) == 0
```

## Queue Implementation

First-In First-Out (FIFO) behavior is used in the queue. This allows it to use the list-based storage with enqueue and dequeue operations.

Example:

```
175
176     class Queue:
177         """
178             Simple queue using list.
179             Note: dequeue using pop(0) is O(n), which you can mention in analysis.
180         """
181
182         def __init__(self):
183             self.data: List[Any] = []
184
185         def enqueue(self, item: Any) -> None:
186             self.data.append(item)
187
188         def dequeue(self) -> Any:
189             if not self.data:
190                 raise IndexError("dequeue from empty queue")
191             return self.data.pop(0)
192
193         def peek(self) -> Any:
194             if not self.data:
195                 raise IndexError("peek from empty queue")
196             return self.data[0]
197
198         def is_empty(self) -> bool:
199             return len(self.data) == 0
```

## Singly Linked List

When it comes to singly linked lists, it was implemented inorder to demonstrate pointer-based data structure behavior. This includes:

- Insert\_at\_head
- Insert\_at\_tail
- Delete\_value
- Traverse

Example:

```

208 class SinglyLinkedList:
209     def __init__(self):
210         self.head: Optional[ListNode] = None
211
212     def insert_at_head(self, value: Any) -> None:
213         self.head = ListNode(value, self.head)
214
215     def insert_at_tail(self, value: Any) -> None:
216         new_node = ListNode(value)
217         if self.head is None:
218             self.head = new_node
219             return
220         curr = self.head
221         while curr.next:
222             curr = curr.next
223         curr.next = new_node
224
225     def delete_value(self, value: Any) -> bool:
226         curr = self.head
227         prev = None
228         while curr:
229             if curr.value == value:
230                 if prev is None:
231                     self.head = curr.next
232                 else:
233                     prev.next = curr.next
234                 return True
235                 prev = curr
236                 curr = curr.next
237             return False
238
239     def traverse(self) -> List[Any]:
240         result = []
241         curr = self.head
242         while curr:
243             result.append(curr.value)
244             curr = curr.next
245         return result
246

```

## Discussion

As shown above, the experiments results show that randomized selection is continually faster for the tested input sizes. However, deterministic selection allows theoretical reliability by guaranteeing linear time complexity. This showcases the trade-offs between predictability and performance, which is considered an important part of algorithm design. These data structures

illustrate how different storage models affect performance efficiency. Array-based stacks and the queues are basic but can help introduce overhead, particularly with dequeue operations that require more memory due to node references.

## **Conclusion**

This assignment helped me to successfully understand and demonstrate both order statistics algorithms and fundamental data structures. The deterministic and randomized selection algorithms help showcase different approaches to solving the same problem with a variety of efficiency trade-offs. At the same time, the implementation of stacks, queues, and linked lists help reinforce continuous understanding of core data management concepts essential when it comes to software development.

## **References**

Cormen, T. H., Leiserson, C. E., Rivest, R. L., & Stein, C. (2022). Introduction to Algorithms (4th ed.). MIT Press.