

Python Notes for Data Structures & Algorithms (DSA)

Welcome back to DSA! Since you have a background in C++, C, Java, and JavaScript, you'll find Python's simplicity and rich standard library a huge advantage for DSA. These notes will highlight key Python features and best practices for problem-solving.

1. Core Python Features for DSA

Python's built-in data types and structures are incredibly powerful and often abstract away complexities you'd handle manually in C++ or Java.

1.1. Basic Data Types

- **Integers (int):** Arbitrary precision (no overflow issues like in C++/Java int).
a = 10
b = 10**100 # Very large integer, no problem
- **Floats (float):** Standard double-precision floating-point numbers.
- **Booleans (bool):** True or False.
- **Strings (str):** Immutable sequences of characters.
 - **Concatenation:** s1 + s2
 - **Slicing:** s[start:end:step] (e.g., s[::-1] for reverse)
 - **Common methods:** len(s), s.lower(), s.upper(), s.strip(), s.split(), s.join(list_of_strings)

1.2. Essential Built-in Data Structures

1.2.1. Lists (list) - Your Dynamic Array / Stack / Queue

Lists are Python's most versatile sequence type. They are dynamic arrays, meaning they can grow and shrink in size.

- **Creation:**
my_list = []
my_list = [1, 2, "hello", True]
- **Accessing Elements:**
print(my_list[0]) # 1
print(my_list[-1]) # True (last element)
- **Common Operations:**
 - **append(element):** Add to the end (Amortized O(1)).
 - **pop():** Remove and return last element (O(1)).

- `pop(index)`: Remove and return element at index ($O(N)$).
- `insert(index, element)`: Insert at index ($O(N)$).
- `remove(value)`: Remove first occurrence of value ($O(N)$).
- `len(my_list)`: Get length ($O(1)$).
- `sort()`: Sorts in-place ($O(N \log N)$). `sorted(list)` returns a new sorted list.
- `reverse()`: Reverses in-place ($O(N)$). `list[::-1]` creates a new reversed list.
- `in` operator: element in `my_list` ($O(N)$ for lists).
- **Slicing**: Powerful for sub-lists.
`sub_list = my_list[1:3]` # [2, "hello"]
`copy_list = my_list[:]` # Creates a shallow copy
- **Using Lists as Stacks**:
`stack = []`
`stack.append(1)` # Push
`stack.append(2)`
`top_element = stack.pop()` # Pop
- **Using Lists as Basic Queues (Inefficient for large queues)**:
`queue = []`
`queue.append(1)` # Enqueue
`queue.append(2)`
`front_element = queue.pop(0)` # Dequeue ($O(N)$) - avoid for large queues

1.2.2. Tuples (tuple) - Immutable Sequences

Tuples are similar to lists but are immutable. Once created, their elements cannot be changed.

- **Creation**:
`my_tuple = (1, 2, "a")`
`single_element_tuple = (5,)` # Comma is essential for single-element tuples
- **Use Cases**: Often used for fixed collections of items, function return values (e.g., returning (value, index)), and as dictionary keys (because they are hashable due to immutability).

1.2.3. Sets (set) - Unordered Collections of Unique Elements

Sets are useful for quickly checking membership, removing duplicates, and performing set operations.

- **Creation**:

```
my_set = {1, 2, 3, 2} # my_set will be {1, 2, 3}
empty_set = set()
```

- **Common Operations (mostly $O(1)$ on average):**

- `add(element)`
- `remove(element)` (raises `KeyError` if not found)
- `discard(element)` (no error if not found)
- `element in my_set` (fast membership test)
- `union()`, `intersection()`, `difference()`

1.2.4. Dictionaries (dict) - Your Hash Map / Hash Table

Dictionaries store key-value pairs and provide very fast lookups based on keys.

- **Creation:**

```
my_dict = {"name": "Alice", "age": 30}
empty_dict = {}
```

- **Accessing/Modifying:**

```
print(my_dict["name"]) # "Alice"
my_dict["city"] = "New York"
```

- **Common Operations (mostly $O(1)$ on average):**

- `my_dict[key]`: Access value (raises `KeyError` if key not found).
- `my_dict.get(key, default_value)`: Safely get value, returns `default_value` if key not found (default is `None`).
- `key in my_dict`: Check if key exists (fast membership test).
- `del my_dict[key]`: Delete a key-value pair.
- `my_dict.keys()`: Returns a view of keys.
- `my_dict.values()`: Returns a view of values.
- `my_dict.items()`: Returns a view of key-value pairs.

- **Iteration:**

```
for key, value in my_dict.items():
    print(f"{key}: {value}")
```

1.3. Control Flow

- **if/elif/else:** Standard conditional statements.

```
if x > 0:
    print("Positive")
elif x < 0:
```

```
    print("Negative")
else:
    print("Zero")
```

- **for loops:** Iterate over sequences (lists, strings, tuples, dictionaries).

```
for item in my_list:
    print(item)
for i in range(5): # 0, 1, 2, 3, 4
    print(i)
for i in range(len(my_list)): # Iterate with index
    print(my_list[i])
```

- **while loops:** Loop as long as a condition is true.

```
count = 0
while count < 5:
    print(count)
    count += 1
```

1.4. Functions

Define reusable blocks of code.

```
def greet(name):
    return f"Hello, {name}!"
```

```
message = greet("World")
print(message)
```

1.5. Classes (Basic Object-Oriented Programming)

Essential for implementing custom data structures like Linked Lists, Trees, Graphs, etc.

```
class Node:
    def __init__(self, value):
        self.value = value
        self.next = None # For linked lists
        self.left = None # For trees
        self.right = None # For trees

    def __str__(self): # Optional: for easy printing
```

```
return f"Node({self.value})"
```

Example usage:

```
node1 = Node(10)
```

```
node2 = Node(20)
```

```
node1.next = node2
```

```
print(node1) # Node(10)
```

```
print(node1.next) # Node(20)
```

1.6. Recursion

A function calling itself. Crucial for many tree and graph algorithms.

```
def factorial(n):
```

```
    if n == 0: # Base case
```

```
        return 1
```

```
    else: # Recursive step
```

```
        return n * factorial(n - 1)
```

```
print(factorial(5)) # 120
```

1.7. List Comprehensions

A concise way to create lists.

```
squares = [x**2 for x in range(10) if x % 2 == 0] # [0, 4, 16, 36, 64]
```

2. DSA Specifics in Python

2.1. Time and Space Complexity Analysis

Understanding Big O notation is critical. Python's built-in operations have specific complexities:

- **len(list) / len(dict) / len(set):** $O(1)$
- **List append() / pop() (from end):** Amortized $O(1)$
- **List insert() / pop(0) / remove():** $O(N)$ (because elements need to be shifted)
- **element in list:** $O(N)$ (linear search)
- **key in dict / element in set:** Average $O(1)$, Worst $O(N)$ (due to hash collisions, rare in practice)

- **Dictionary/Set insertion/deletion:** Average $O(1)$, Worst $O(N)$

2.2. Common DSA Patterns & Pythonic Ways

- **Efficient Queues:** For true $O(1)$ enqueue and dequeue from both ends, use `collections.deque`.

```
from collections import deque
```

```
q = deque()
q.append(1) # Add to right (enqueue)
q.append(2)
print(q.popleft()) # Remove from left (dequeue) -  $O(1)$ 
```

- **Heaps (Priority Queues):** Use the `heapq` module. Python's `heapq` implements a min-heap.

```
import heapq
```

```
min_heap = []
heapq.heappush(min_heap, 3)
heapq.heappush(min_heap, 1)
heapq.heappush(min_heap, 5)
print(heapq.heappop(min_heap)) # 1 (smallest element)
```

- **Graphs:** Typically represented using an **adjacency list** (dictionary where keys are nodes and values are lists/sets of neighbors).

```
graph = {
    'A': ['B', 'C'],
    'B': ['A', 'D'],
    'C': ['A', 'E'],
    'D': ['B'],
    'E': ['C']
}
```

Or using sets for faster neighbor lookup if order doesn't matter

```
graph_set = {
    'A': {'B', 'C'},
    'B': {'A', 'D'},
    'C': {'A', 'E'},
    'D': {'B'},
    'E': {'C'}
}
```

- **Trees:** Implemented using custom Node classes, as shown in the Classes section.

2.3. Input/Output

- **Reading input:**

```
s = input() # Reads a line as a string
```

```
n = int(input()) # Reads a line and converts to integer
```

```
nums = list(map(int, input().split())) # Reads space-separated integers into a list
```

- **Printing output:**

```
print("Hello")
```

```
print(f"The answer is {result}") # f-strings are great for formatting
```

3. Tips for DSA Practice in Python

1. **Leverage Built-ins:** Don't re-invent the wheel unless the problem specifically asks for it. Use list, dict, set, collections.deque, heapq where appropriate.
2. **Understand Underlying Implementations:** While Python abstracts details, knowing that list.pop(0) is $O(N)$ is crucial for choosing deque when a true queue is needed.
3. **Practice Regularly:** Consistency is key. Start with easy problems on platforms like LeetCode and gradually move to medium and hard ones.
4. **Focus on Logic First:** Get a working (even if inefficient) solution first. Then, think about optimization (time and space complexity).
5. **Use collections Module:** Besides deque, explore collections.Counter (for frequency counting) and collections.defaultdict (for dictionaries with default values).
6. **Read Solutions:** After attempting a problem, always review optimal solutions, even if you solved it. Learn new approaches and Pythonic tricks.
7. **Write Clean Code:** Python's readability helps here. Use meaningful variable names.

By focusing on these Pythonic approaches, you'll find your DSA practice much more efficient and enjoyable compared to the lower-level complexities of C++. Good luck!