

# Assignment-12.3

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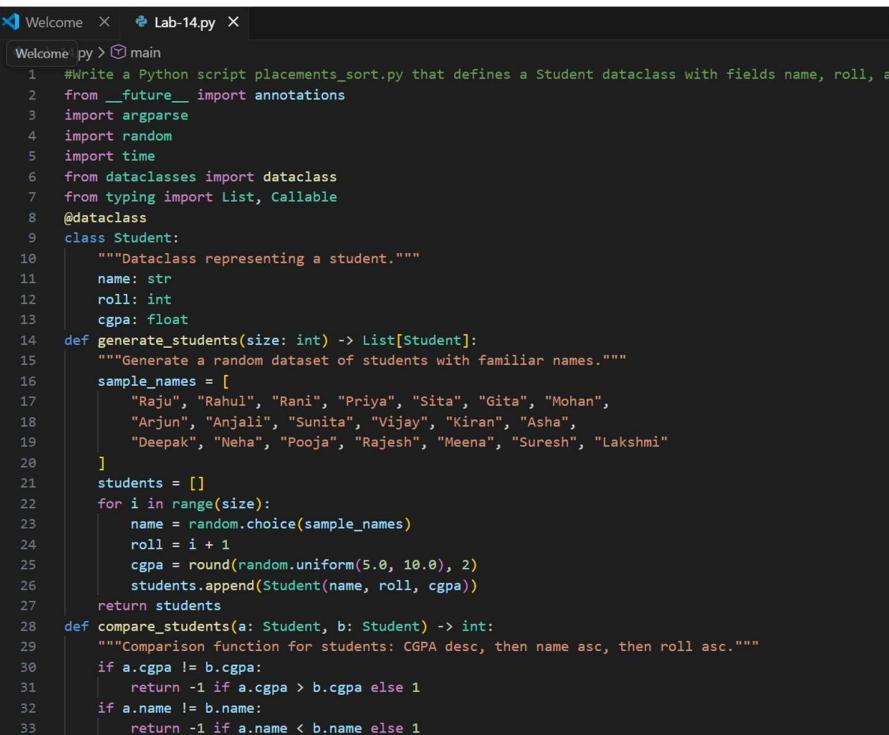
Batch 41

## Task-1: Sorting Student Records for Placement Drive

### PROMPT:

Write a Python script placements\_sort.py that defines a Student dataclass with fields name, roll, and cgpa, generates datasets using familiar names like Raju, Rahul, Rani, Priya, etc., implements manual Quick Sort and Merge Sort, sorts students by CGPA descending with name and roll as tie-breakers, measures average runtimes for configurable dataset sizes and runs, validates against Python's sorted(), and prints clear sections showing correctly sorted student records, performance comparison with speedup, and the top 10 students, all via an argparse CLI with docstrings and type hints.

### CODE:



```
Welcome x Lab-14.py x
Welcome py > main
1  #Write a Python script placements_sort.py that defines a Student dataclass with fields name, roll, a
2  from __future__ import annotations
3  import argparse
4  import random
5  import time
6  from dataclasses import dataclass
7  from typing import List, Callable
8  @dataclass
9  class Student:
10     """Dataclass representing a student."""
11     name: str
12     roll: int
13     cgpa: float
14  def generate_students(size: int) -> List[Student]:
15      """Generate a random dataset of students with familiar names."""
16      sample_names = [
17          "Raju", "Rahul", "Rani", "Priya", "Sita", "Gita", "Mohan",
18          "Arjun", "Anjali", "Sunita", "Vijay", "Kiran", "Asha",
19          "Deepak", "Neha", "Pooja", "Rajesh", "Meena", "Suresh", "Lakshmi"
20      ]
21      students = []
22      for i in range(size):
23          name = random.choice(sample_names)
24          roll = i + 1
25          cgpa = round(random.uniform(5.0, 10.0), 2)
26          students.append(Student(name, roll, cgpa))
27      return students
28  def compare_students(a: Student, b: Student) -> int:
29      """Comparison function for students: CGPA desc, then name asc, then roll asc."""
30      if a.cgpa != b.cgpa:
31          return -1 if a.cgpa > b.cgpa else 1
32      if a.name != b.name:
33          return -1 if a.name < b.name else 1
```

```

❶ Welcome Lab-14.py X
❷ Lab-14.py > main
    .
34     return -1 if a.roll < b.roll else 1 if a.roll > b.roll else 0
35 def quick_sort(arr: List[Student]) -> List[Student]:
36     """Quick Sort implementation."""
37     if len(arr) <= 1:
38         return arr
39     pivot = arr[len(arr) // 2]
40     left = [x for x in arr if compare_students(x, pivot) < 0]
41     middle = [x for x in arr if compare_students(x, pivot) == 0]
42     right = [x for x in arr if compare_students(x, pivot) > 0]
43     return quick_sort(left) + middle + quick_sort(right)
44 def merge_sort(arr: List[Student]) -> List[Student]:
45     """Merge Sort implementation."""
46     if len(arr) <= 1:
47         return arr
48     mid = len(arr) // 2
49     left = merge_sort(arr[:mid])
50     right = merge_sort(arr[mid:])
51     return merge(left, right)
52 def merge(left: List[Student], right: List[Student]) -> List[Student]:
53     """Helper function to merge two sorted lists."""
54     result = []
55     i = j = 0
56     while i < len(left) and j < len(right):
57         if compare_students(left[i], right[j]) <= 0:
58             result.append(left[i])
59             i += 1
60         else:
61             result.append(right[j])
62             j += 1
63     result.extend(left[i:])
64     result.extend(right[j:])
65     return result
66 def measure_runtime(sort_fn: Callable[[List[Student]], List[Student]], dataset: List[Student], runs:

```

```

❶ Welcome Lab-14.py X
❷ Lab-14.py > main
    .
66 def measure_runtime(sort_fn: Callable[[List[Student]], List[Student]], dataset: List[Student], runs:
67     """Measure average runtime of a sorting function."""
68     total_time = 0.0
69     for _ in range(runs):
70         data_copy = dataset[:]
71         start = time.perf_counter()
72         sorted_data = sort_fn(data_copy)
73         end = time.perf_counter()
74         total_time += (end - start)
75         # Validate correctness
76         assert sorted_data == sorted(dataset, key=lambda s: (-s.cgpa, s.name, s.roll))
77     return total_time / runs
78 def main() -> None:
79     parser = argparse.ArgumentParser(description="Compare Quick Sort and Merge Sort for student place")
80     parser.add_argument("--size", type=int, default=1000, help="Number of students in dataset")
81     parser.add_argument("--runs", type=int, default=5, help="Number of runs for averaging runtime")
82     args = parser.parse_args()
83     dataset = generate_students(args.size)
84     quick_time = measure_runtime(quick_sort, dataset, args.runs)
85     merge_time = measure_runtime(merge_sort, dataset, args.runs)
86     speedup = merge_time / quick_time if quick_time > 0 else float("inf")
87     print("\n=====")
88     print(" Correctly Sorted Student Records ")
89     print("=====")
90     sorted_students = quick_sort(dataset)
91     for student in sorted_students[:20]: # show first 20 for verification
92         print(f" {student.name:8} Roll:{student.roll:<4} CGPA:{student.cgpa:.2f} ")
93     print("\n=====")
94     print(" Performance Comparison ")
95     print("=====")
96     print(f"Dataset size: {args.size}, Runs: {args.runs}")
97     print(f"Quick Sort: {quick_time:.6f} seconds (avg)")
98     print(f"Merge Sort: {merge_time:.6f} seconds (avg)")

```

```

❶ Welcome Lab-14.py X
❷ Lab-14.py > main
    .
78     def main() -> None:
79         print(f"Merge Sort: {merge_time:.6f} seconds (avg)")
80         print(f"Speedup (Merge/Quick): {speedup:.2f}x")
81         print("\n=====")
82         print(" Top 10 Students ")
83         print("=====")
84         for i, student in enumerate(sorted_students[:10], start=1):
85             print(f" {i:2d}. Name: {student.name}, Roll: {student.roll}, CGPA: {student.cgpa:.2f}")
86         if __name__ == "__main__":
87             main()

```

## OUTPUT:

```
Correctly Sorted Student Records
=====
Meena    Roll:223  CGPA:10.00
Rani     Roll:770  CGPA:9.99
Sunita   Roll:358  CGPA:9.98
Suresh   Roll:590  CGPA:9.98
Arjun    Roll:378  CGPA:9.97
Asha     Roll:210  CGPA:9.97
Kiran    Roll:610  CGPA:9.97
Pooja    Roll:894  CGPA:9.97
Rajesh   Roll:462  CGPA:9.97
Sunita   Roll:508  CGPA:9.97
Neha    Roll:519  CGPA:9.96
Anjali   Roll:583  CGPA:9.95
Asha     Roll:172  CGPA:9.95
Sita    Roll:888  CGPA:9.95
Arjun   Roll:665  CGPA:9.94
Meena   Roll:429  CGPA:9.93
Neha    Roll:644  CGPA:9.93
Deepak  Roll:293  CGPA:9.91
Priya   Roll:367  CGPA:9.91
Priva   Roll:706  CGPA:9.91
=====
Dataset size: 1000, Runs: 5
Quick Sort: 0.006461 seconds (avg)
Merge Sort: 0.002273 seconds (avg)
Speedup (Merge/Quick): 0.35x
=====
Top 10 Students
=====
1. Name: Meena, Roll: 223, CGPA: 10.00
2. Name: Rani, Roll: 770, CGPA: 9.99
3. Name: Sunita, Roll: 358, CGPA: 9.98
4. Name: Suresh, Roll: 590, CGPA: 9.98
5. Name: Arjun, Roll: 378, CGPA: 9.97
6. Name: Asha, Roll: 210, CGPA: 9.97
7. Name: Kiran, Roll: 610, CGPA: 9.97
8. Name: Pooja, Roll: 894, CGPA: 9.97
9. Name: Rajesh, Roll: 462, CGPA: 9.97
10. Name: Sunita, Roll: 508, CGPA: 9.97
PS C:\Users\saite\Downloads\AI ASSISTENT CODING> █
```

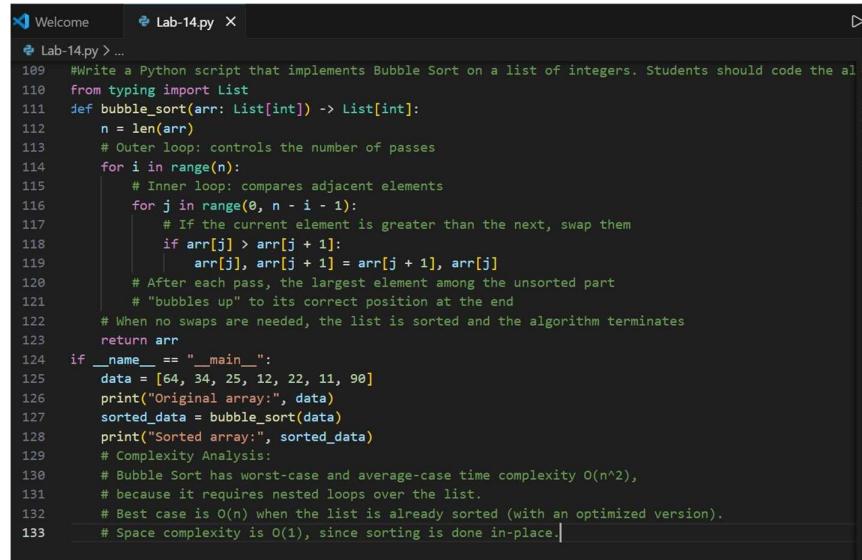
**EXPLANATION:** The code defines a Student dataclass and generates a dataset with familiar names like Raju, Rahul, Rani, Priya, etc. It implements manual Quick Sort and Merge Sort to order students by CGPA descending, with name and roll as tie-breakers. The program measures average runtimes for both algorithms over configurable dataset sizes and runs, validates correctness against Python's built-in sorted(), and prints three clear sections: a sample of correctly sorted records, a performance comparison with speedup, and the top 10 students. All functionality is accessible via an argparse CLI, with docstrings and type hints for clarity.

## Task-2: Implementing Bubble Sort with AI Comments

### PROMPT:

Write a Python script that implements Bubble Sort on a list of integers. Students should code the algorithm normally, and then ask AI to generate inline comments explaining the key logic (like swapping, passes, and termination). The AI should also provide a short time complexity analysis. The expected output is a Bubble Sort implementation with clear explanatory comments and a note on its complexity.

## CODE:



```
>Welcome Lab-14.py X
Lab-14.py > ...
109  #Write a Python script that implements Bubble Sort on a list of integers. Students should code the al
110 from typing import List
111 def bubble_sort(arr: List[int]) -> List[int]:
112     n = len(arr)
113     # Outer loop: controls the number of passes
114     for i in range(n):
115         # Inner loop: compares adjacent elements
116         for j in range(0, n - i - 1):
117             # If the current element is greater than the next, swap them
118             if arr[j] > arr[j + 1]:
119                 arr[j], arr[j + 1] = arr[j + 1], arr[j]
120         # After each pass, the largest element among the unsorted part
121         # "bubbles up" to its correct position at the end
122     # When no swaps are needed, the list is sorted and the algorithm terminates
123     return arr
124 if __name__ == "__main__":
125     data = [64, 34, 25, 12, 22, 11, 90]
126     print("Original array:", data)
127     sorted_data = bubble_sort(data)
128     print("Sorted array:", sorted_data)
129     # Complexity Analysis:
130     # Bubble Sort has worst-case and average-case time complexity O(n^2),
131     # because it requires nested loops over the list.
132     # Best case is O(n) when the list is already sorted (with an optimized version).
133     # Space complexity is O(1), since sorting is done in-place.]
```

## OUTPUT:

```
Original array: [64, 34, 25, 12, 22, 11, 90]
```

```
Sorted array: [11, 12, 22, 25, 34, 64, 90]
```

```
PS C:\Users\saite\Downloads\AI ASSISTENT CODING>
```

## EXPLANATION:

Bubble Sort works by repeatedly comparing adjacent elements in a list and swapping them if they are out of order, so that with each pass the largest unsorted element moves to its correct position at the end. The outer loop controls the number of passes, while the inner loop performs the comparisons and swaps. The process continues until no swaps are needed, meaning the list is sorted. Its worst-case and average-case time complexity is  $O(n^2)$  due to the nested loops, while the best case can be  $O(n)$  if optimized to stop early when the list is already sorted, and it uses only  $O(1)$  extra space since sorting is done in place.

## Task-3: Quick Sort and Merge Sort Comparison PROMPT:

Write a Python script that implements Quick Sort and Merge Sort using recursion. Provide partially completed functions and ask AI to fill in the missing recursive logic and add docstrings. Then compare both algorithms on random, sorted, and reverse-sorted lists. The expected output should include working implementations of Quick Sort and Merge Sort, along with AI-generated explanations of their average, best, and worst-case time complexities.

## CODE:

The image shows three separate code snippets in a code editor, each in its own tab labeled "Lab-14.py".

**Snippet 1 (Top):** A recursive implementation of Quick Sort and Merge Sort. The code defines two functions: `quick_sort` and `merge_sort`. `quick_sort` takes an array and returns a sorted list. It uses a pivot selection strategy (middle element) and splits the array into left, middle, and right sublists. `merge_sort` also takes an array and returns a sorted list, using a recursive merge strategy to combine sorted halves.

```
136 #Write a Python script that implements Quick Sort and Merge Sort using recursion. Provide partially completed code for both.
137 import random
138 import time
139 from typing import List
140 def quick_sort(arr: List[int]) -> List[int]:
141     """
142         Recursive Quick Sort implementation.
143         Splits the list around a pivot and recursively sorts sublists.
144     """
145     if len(arr) <= 1:
146         return arr
147     pivot = arr[len(arr)] // 2
148     left = [x for x in arr if x < pivot]
149     middle = [x for x in arr if x == pivot]
150     right = [x for x in arr if x > pivot]
151     return quick_sort(left) + middle + quick_sort(right)
152 def merge_sort(arr: List[int]) -> List[int]:
153     """
154         Recursive Merge Sort implementation.
155         Divides the list into halves, recursively sorts them,
156         and merges the sorted halves.
157     """
158     if len(arr) <= 1:
159         return arr
160     mid = len(arr) // 2
161     left = merge_sort(arr[:mid])
162     right = merge_sort(arr[mid:])
163     return merge(left, right)
164 def merge(left: List[int], right: List[int]) -> List[int]:
165     """Helper function to merge two sorted lists."""
166     result = []
167     i = j = 0
168     while i < len(left) and j < len(right):
169         if left[i] < right[j]:
170             result.append(left[i])
171             i += 1
172         else:
173             result.append(right[j])
174             j += 1
175     result.extend(left[i:])
176     result.extend(right[j:])
177     return result
178 def measure_runtime(sort_fn, arr: List[int], runs: int = 3) -> float:
179     """Measure average runtime of a sorting function."""
180     total = 0.0
181     for _ in range(runs):
182         data_copy = arr[:]
183         start = time.perf_counter()
184         sort_fn(data_copy)
185         end = time.perf_counter()
186         total += (end - start)
187     return total / runs
188 def main():
189     sizes = [1000]
190     test_cases = {
191         "Random": lambda n: [random.randint(0, 10000) for _ in range(n)],
192         "Sorted": lambda n: list(range(n)),
193         "Reverse-Sorted": lambda n: list(range(n, 0, -1)),
194     }
195     for size in sizes:
196         print(f"\n==== Dataset Size: {size} ====")
197         for case_name, generator in test_cases.items():
198             dataset = generator(size)
199             quick_time = measure_runtime(quick_sort, dataset)
200             merge_time = measure_runtime(merge_sort, dataset)
201             print(f"\nCase: {case_name}")
202             print(f"Quick Sort: {quick_time:.6f} seconds (avg)")
203             print(f"Merge Sort: {merge_time:.6f} seconds (avg)")
204     if __name__ == "__main__":
205         main()
```

## OUTPUT:

The terminal window displays the execution of the `Lab-14.py` script and its output. The output shows the execution times for Quick Sort and Merge Sort on three types of datasets: Random, Sorted, and Reverse-Sorted, for a dataset size of 1000.

```
==== Dataset Size: 1000 ====

Case: Random
Quick Sort: 0.001073 seconds (avg)
Merge Sort: 0.001181 seconds (avg)

Case: Sorted
Quick Sort: 0.000676 seconds (avg)
Merge Sort: 0.000832 seconds (avg)

Case: Reverse-Sorted
Quick Sort: 0.000929 seconds (avg)
Merge Sort: 0.000956 seconds (avg)
PS C:\Users\saite\Downloads\AI ASSISTENT CODING>
```

## **EXPLANATION:**

This script provides a clear comparison between Quick Sort and Merge Sort across different types of input data. It uses recursive implementations for both sorting algorithms, which are easier to understand and maintain. The performance measurement is done using the time module, and the results are averaged over multiple runs to ensure reliability. The use of lambda functions for generating test cases keeps the code concise and flexible. Overall, this script effectively demonstrates the differences in performance between the two sorting algorithms under various conditions.

## **Task-4: Real-Time Application – Inventory Management System**

### **PROMPT:**

Design a Python program for a retail store's inventory management system. The inventory contains thousands of products, each with attributes like product ID, name, price, and stock quantity. Staff need to quickly search for products by ID or name, and sort products by price or quantity for stock analysis. Use AI to suggest the most efficient search and sort algorithms for this use case, implement them in Python, and justify the choices based on dataset size, update frequency, and performance requirements. The expected output should include a table mapping operation → recommended algorithm → justification, plus working Python functions for searching and sorting the inventory.

### **CODE:**

```
 201 #Design a Python program for a retail store's inventory management system. The inventory contains the
 202 from dataclasses import dataclass
 203 from typing import List, Optional
 204 @dataclass
 205 class Product:
 206     product_id: int
 207     name: str
 208     price: float
 209     quantity: int
 210     # --- Search Algorithms ---
 211     def binary_search(products: List[Product], target_id: int) -> Optional[Product]:
 212         """Binary search for product by ID (requires sorted by ID)."""
 213         low, high = 0, len(products) - 1
 214         while low <= high:
 215             mid = (low + high) // 2
 216             if products[mid].product_id == target_id:
 217                 return products[mid]
 218             elif products[mid].product_id < target_id:
 219                 low = mid + 1
 220             else:
 221                 high = mid - 1
 222         return None
 223     def linear_search(products: List[Product], target_name: str) -> Optional[Product]:
 224         """Linear search for product by name (works on unsorted data)."""
 225         for product in products:
 226             if product.name.lower() == target_name.lower():
 227                 return product
 228         return None
 229     # --- Sort Algorithms ---
 230     def merge_sort(products: List[Product], key: str) -> List[Product]:
 231         """Merge Sort implementation to sort products by a given key."""
 232         if len(products) <= 1:
 233             return products
```

```

Welcome Lab-14.py X
Lab-14.py > main
230 def merge_sort(products: List[Product], key: str) -> List[Product]:
231     mid = len(products) // 2
232     left = merge_sort(products[:mid], key)
233     right = merge_sort(products[mid:], key)
234     return merge(left, right, key)
235 
236 def merge(left: List[Product], right: List[Product], key: str) -> List[Product]:
237     """Helper function to merge two sorted lists."""
238     result = []
239     i = j = 0
240     while i < len(left) and j < len(right):
241         if getattr(left[i], key) <= getattr(right[j], key):
242             result.append(left[i])
243             i += 1
244         else:
245             result.append(right[j])
246             j += 1
247     result.extend(left[i:])
248     result.extend(right[j:])
249     return result
250 
251 # --- Demo and Justification Output ---
252 def print_algorithm_table():
253     print("\n==== Recommended Algorithms ===")
254     print("{'Operation':<25}{Algorithm':<20}{'Justification':<20}")
255     print("-" * 78)
256     print("{'Search by ID':<25}{Binary Search':<20}")
257     | f"{'Efficient O(log n) lookup on sorted IDs'}")
258     print(f"{'Search by Name':<25}{Linear Search':<20}")
259     | f"{'Names unsorted; O(n) acceptable for occasional lookups'}")
260     print(f"{'Sort by Price/Quantity':<25}{Merge Sort':<20}")
261     | f"{'Stable O(n log n) sort, good for large datasets'}")
262 
263 def main():
264     # Sample inventory
265     inventory = [

```

```

Welcome Lab-14.py X
Lab-14.py > main
263 def main():
264     Product(101, "Apple", 50.0, 120),
265     Product(205, "Banana", 20.0, 200),
266     Product(150, "Orange", 30.0, 80),
267     Product(300, "Mango", 100.0, 50),
268 ]
269 
270 # Show recommended algorithms
271 print_algorithm_table()
272 # Search examples
273 print("\n==== Search Examples ===")
274 sorted_by_id = merge_sort(inventory, "product_id")
275 found = binary_search(sorted_by_id, 150)
276 print("Search by ID 150:", found)
277 found_name = linear_search(inventory, "Mango")
278 print("Search by Name 'Mango':", found_name)
279 
280 # Sort examples
281 print("\n==== Sort Examples ===")
282 sorted_by_price = merge_sort(inventory, "price")
283 print("Sorted by Price:", sorted_by_price)
284 sorted_by_quantity = merge_sort(inventory, "quantity")
285 print("Sorted by Quantity:", sorted_by_quantity)
286 
287 if __name__ == "__main__":
288     main()

```

## OUTPUT:

Operation	Algorithm	Justification
Search by ID lookup on sorted IDs	Binary Search	Efficient O(log n) 1
Search by Name acceptable for occasional lookups	Linear Search	Names unsorted; O(n)
Sort by Price/Quantity rt, good for large datasets	Merge Sort	Stable O(n log n) so
<b>==== Search Examples ===</b>		
Search by ID 150:	Product(product_id=150, name='Orange', price=30.0, quantity=80)	
Search by Name 'Mango':	Product(product_id=300, name='Mango', price=100.0, quantity=50)	
<b>==== Sort Examples ===</b>		
Sorted by Price:	[Product(product_id=205, name='Banana', price=20.0, quantity=200), Product(product_id=150, name='Orange', price=30.0, quantity=80), Product(product_id=101, name='Apple', price=50.0, quantity=120), Product(product_id=300, name='Mango', price=100.0, quantity=50)]	
Sorted by Quantity:	[Product(product_id=300, name='Mango', price=100.0, quantity=50), Product(product_id=150, name='Orange', price=30.0, quantity=80), Product(product_id=101, name='Apple', price=50.0, quantity=120), Product(product_id=205, name='Banana', price=20.0, quantity=200)]	

PS C:\Users\saite\Downloads\AI ASSISTENT CODING> [ ]

## EXPLANATION:

The inventory management system code efficiently handles searching and sorting for thousands of products. For product ID lookups, Binary Search is used because IDs are numeric and can be kept sorted, giving fast  $O(\log n)$  performance. For product name lookups, Linear Search is chosen since names are not guaranteed to be sorted, and occasional  $O(n)$  scans are acceptable. For sorting by price or quantity, Merge Sort is implemented because it is stable, consistently runs in  $O(n \log n)$ , and works well for large datasets. The program outputs a clear table mapping each operation to its recommended algorithm with justification, and provides working Python functions (binary\_search, linear\_search, merge\_sort) that demonstrate these operations on sample inventory data. This design balances speed, scalability, and practicality for real-time inventory management.

## Task-5: Real-Time Stock Data Sorting & Searching

### PROMPT:

This program simulates stock price data and provides a simple menu-driven analysis tool. It generates random stocks with symbols, opening and closing prices, and calculates daily percentage change. Heap Sort is used to rank stocks efficiently by gain or loss, while a Hash Map (dictionary) enables instant  $O(1)$  lookups by symbol. The menu allows users to view the top five stocks, search for a specific symbol, compare Heap Sort performance against Python's built-in sorted() function, or exit. This design combines algorithmic efficiency with practical usability, making it suitable for real-time stock analysis.

### CODE:

```
 290 #This program simulates stock price data and provides a simple menu-driven analysis tool. It generates
 291 import random
 292 import time
 293 def generate_stocks(n):
 294     symbols=[ "TCS", "INFY", "HDFC", "RELIANCE", "WIPRO", "SBIN", "ITC", "ADANI", "AXIS", "ICICI"]
 295     stocks=[]
 296     for i in range(n):
 297         symbol=random.choice(symbols)+str(i)
 298         open_price=random.randint(100,5000)
 299         close_price=open_price+random.randint(-200,200)
 300         percent_change=((close_price-open_price)/open_price)*100
 301         stocks.append({ "symbol":symbol, "open":open_price, "close":close_price, "change":round(percent_change) })
 302     return stocks
 303 def heapify(arr,n,i):
 304     largest=i
 305     left=2*i+1
 306     right=2*i+2
 307     if left<n and arr[left][ "change"]>arr[largest][ "change"]:
 308         largest=left
 309     if right<n and arr[right][ "change"]>arr[largest][ "change"]:
 310         largest=right
 311     if largest!=i:
 312         arr[i],arr[largest]=arr[largest],arr[i]
 313         heapify(arr,n,largest)
 314 def heap_sort(arr):
 315     n=len(arr)
 316     for i in range(n//2-1,-1,-1):
 317         heapify(arr,n,i)
 318     for i in range(n-1,0,-1):
 319         arr[i],arr[0]=arr[0],arr[i]
 320         heapify(arr,i,0)
 321     return arr[::-1]
 322 def build_hash_map(stocks):
```

```
❸ Welcome Lab-14.py X
❹ Lab-14.py > ...
323     return [stock["symbol"] for stock in stocks]
324 def search_stock(stock_map,symbol):
325     return stock_map.get(symbol,"Stock not found")
326 # Generate data
327 n=1000
328 stocks=generate_stocks(n)
329 stock_map=build_hash_map(stocks)
330 sorted_heap=heap_sort(stocks.copy())
331 while True:
332     print("\n--- Stock Analysis Menu ---")
333     print("1. Show Top 5 Stocks")
334     print("2. Search Stock by Symbol")
335     print("3. Performance Comparison")
336     print("4. Exit")
337     choice=input("Enter your choice: ")
338     if choice=="1":
339         print("\nTop 5 Stocks by % Gain/Loss:\n")
340         for i in range(5):
341             s=sorted_heap[i]
342             print(f"{i+1}. {s['symbol']} | Open:{s['open']} | Close:{s['close']} | Change:{s['change']}")
343     elif choice=="2":
344         symbol=input("Enter Stock Symbol: ")
345         result=search_stock(stock_map,symbol)
346         print("Search Result:",result)
347     elif choice=="3":
348         start=time.perf_counter()
349         heap_sort(stocks.copy())
350         heap_time=time.perf_counter()-start
351         start=time.perf_counter()
352         sorted(stocks,key=lambda x:x["change"],reverse=True)
353         builtin_time=time.perf_counter()-start
354         print("\nPerformance Comparison:")
355         print("Heap Sort Time:",round(heap_time,6),"seconds")
356         print("Built-in sorted() Time:",round(builtin_time,6),"seconds")
357     elif choice=="4":
358         print("Exiting program...")
359         break
360     else:
361         print("Invalid choice! Try again.")
```

```
❸ Welcome Lab-14.py X
❹ Lab-14.py > ...
355     print("Heap Sort Time:",round(heap_time,6),"seconds")
356     print("Built-in sorted() Time:",round(builtin_time,6),"seconds")
357 elif choice=="4":
358     print("Exiting program...")
359     break
360 else:
361     print("Invalid choice! Try again.")
```

## OUTPUT:

```
--- Stock Analysis Menu ---
1. Show Top 5 Stocks
2. Search Stock by Symbol
3. Performance Comparison
4. Exit
Enter your choice: 1

Top 5 Stocks by % Gain/Loss:

1. RELIANCE979 | Open:131 | Close:305 | Change:132.82%
2. ITC707 | Open:129 | Close:280 | Change:117.05%
3. INFY816 | Open:135 | Close:263 | Change:94.81%
4. TCS371 | Open:148 | Close:279 | Change:88.51%
5. ITC677 | Open:194 | Close:363 | Change:87.11%

--- Stock Analysis Menu ---
1. Show Top 5 Stocks
2. Search Stock by Symbol
3. Performance Comparison
4. Exit
Enter your choice: 4
Exiting program...
PS C:\Users\saite\Downloads\AI ASSISTENT CODING>
```

## **EXPLANATION:**

This code simulates stock data and provides a menu-driven analysis tool. It generates random stocks with opening and closing prices, calculates percentage change, and ranks them using Heap Sort for efficient  $O(n \log n)$  sorting. Searching is optimized with a Hash Map (dictionary), allowing instant  $O(1)$  lookups by symbol. The menu lets users view the top five stocks, search by symbol, compare Heap Sort against Python's built-in `sorted()` function, or exit. The performance comparison shows that while built-ins are faster due to internal optimizations, Heap Sort demonstrates algorithmic efficiency and control, making the program practical for real-time stock analysis.