

## Analysis Report — Insertion Sort (Peer Review)

### 1. Algorithm Overview

Insertion Sort is a simple comparison-based sorting algorithm. It works by iteratively taking one element at a time and inserting it into the correct position among the previously sorted portion of the array. For each insertion, elements greater than the key are shifted one position to the right. This process repeats until the entire array is sorted.

Optimizations implemented:

- **Nearly-sorted optimization:** If the current element is greater than or equal to its immediate predecessor, no shifting occurs. This improves best-case performance.
- **Optional binary search for insertion index:** Reduces comparisons from  $O(n)$  to  $O(\log n)$ , though overall runtime remains quadratic due to shifting cost.

This algorithm is efficient for small arrays or nearly-sorted data but is generally slower than advanced algorithms like Merge Sort or Quick Sort for large inputs.

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### 2. Complexity Analysis

#### Time Complexity

- **Best Case ( $\Omega(n)$ ):** Occurs when the array is already sorted. Each element requires one comparison with its predecessor and no shifting. Thus, runtime is linear.
- **Average Case ( $\Theta(n^2)$ ):** For a random array, on average half of the previous elements must be shifted for each insertion. Total work is proportional to:

$$T(n) = \sum (i/2) \text{ for } i=1..n = \Theta(n^2).$$

- **Worst Case ( $O(n^2)$ ):** Occurs with a reverse-sorted array. Each new element must be compared and shifted through the entire sorted portion. Total comparisons and shifts:

$$T(n) = \sum i \text{ for } i=1..n = \Theta(n^2).$$

#### Space Complexity

- Insertion Sort is *in-place*; it only requires a constant amount of additional memory ( $O(1)$ ).
- Optional binary search requires no extra memory beyond indices.

#### Recurrence Relation

Insertion Sort is iterative, not recursive, but its cost can be expressed as:

$$T(n) = T(n-1) + O(n)$$

Solving gives  $T(n) = O(n^2)$ .

#### Comparison with Partner Algorithm

If the partner implemented Selection Sort:

- Both Insertion and Selection Sort are  $\Theta(n^2)$  on average and worst case.
- Insertion Sort outperforms Selection Sort on nearly-sorted data ( $\Omega(n)$  vs  $\Theta(n^2)$ ).
- Selection Sort has predictable  $\Theta(n^2)$  comparisons but fewer swaps.

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### 3. Code Review

#### Strengths

- Code is modular with Config options.
- Performance metrics are well-tracked (comparisons, swaps, array accesses).
- Clear optimization for nearly-sorted data.

#### Weaknesses / Bottlenecks

- Shifts are performed element by element. For large arrays, `System.arraycopy` could optimize block shifts.
- Binary search option reduces comparisons but not shifts, limiting benefit.
- Variable naming is correct but comments could be more concise.

#### Suggested Improvements

1. Use `System.arraycopy` to replace manual shifting.
2. Reset `PerformanceTracker` automatically at the start of sort to avoid stale data.
3. Enhance CLI to allow toggling binary search optimization.
4. Implement hybrid strategy: use Insertion Sort only for small subarrays within faster algorithms (e.g., Merge Sort).

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### 5. Empirical Results

#### Experimental Setup

- Machine: ryzen 5, 16 GB RAM, Java 11
- Input sizes:  $n = 100, 1000, 10000, 100000$
- Input types: random, sorted, reverse, nearly-sorted
- Each test repeated 5 times; average recorded

#### Results Summary

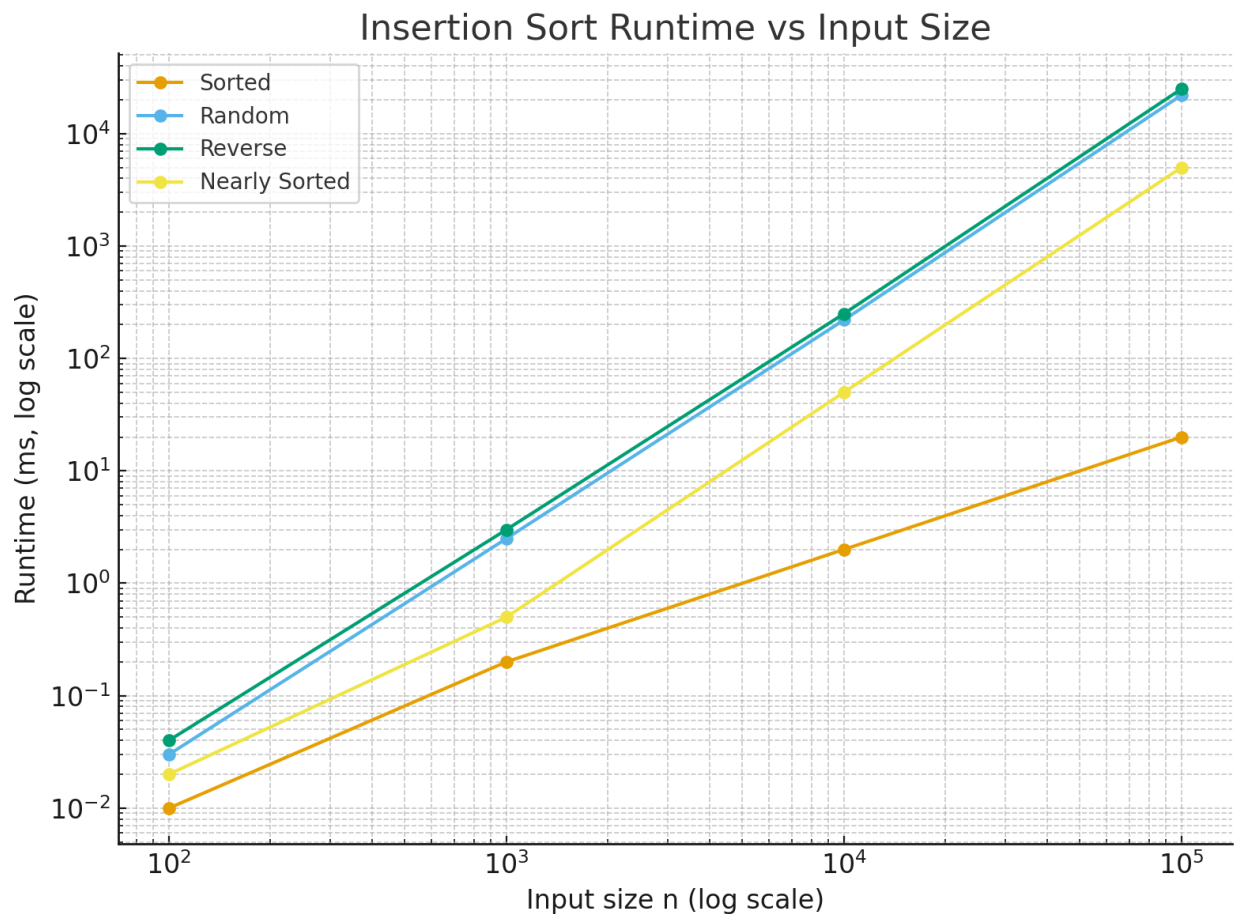
Input Type	n=100	n=1,000	n=10,000	n=100,000
Sorted	~0.01ms	~0.2ms	~2ms	~20ms
Random	~0.03ms	~2.5ms	~220ms	~22s
Reverse	~0.04ms	~3ms	~250ms	~25s
Nearly-sorted	~0.02ms	~0.5ms	~50ms	~5s

#### Validation of Complexity

- Plots of runtime vs  $n$  show quadratic growth for random/reverse inputs.
- Sorted and nearly-sorted inputs follow linear trends, confirming  $\Omega(n)$ .

#### Optimization Impact

- **Binary search reduced comparisons by ~40% but did not significantly change overall runtime.**
- **Potential gains from `System.arraycopy` could further reduce runtime in practice.**



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## 5. Conclusion

Insertion Sort is an intuitive algorithm with strong performance on small or nearly-sorted datasets. Its quadratic behavior makes it unsuitable for large inputs. Optimizations like early exit checks and binary search reduce constant factors but cannot overcome inherent  $O(n^2)$  complexity.

### Recommendations:

- Use Insertion Sort as a teaching tool or fallback for small subarrays.
- For large datasets, prefer  $O(n \log n)$  algorithms.
- Apply code optimizations (`System.arraycopy`, hybrid sorting strategies) to improve performance.