

# Cross-Review Summary

## 1. Overview of Algorithms

- **Kadane's Algorithm (Almat's implementation):**

- Purpose: Finds the maximum subarray sum in a sequence of integers.
- Approach: Uses dynamic programming, maintaining the maximum subarray ending at each index.
- Applications: Stock market analysis, signal processing, and optimization problems.

- **Boyer-Moore Majority Vote Algorithm (Elkham's implementation):**

- Purpose: Identifies the majority element in an array (appearing more than  $\lfloor n/2 \rfloor$  times).
- Approach: Two-phase process: (1) candidate selection via voting, (2) verification by counting occurrences.
- Applications: Data stream analysis, voting systems, distributed consensus.

## 2. Design and Implementation Comparison

- **Modularity:**

- Kadane's code was structured with clear handling of edge cases (empty and single-element arrays).
- Boyer-Moore was separated into candidate selection and verification, with metrics collection for benchmarking.

- **Metrics and Benchmarking:**

- Kadane's implementation focused on correctness with JUnit tests.
- Boyer-Moore integrated a `PerformanceTracker` to measure comparisons, array accesses, memory usage, and execution time.

- **Edge Cases:**

- Kadane: Explicitly handled arrays of size 0 and 1.
- Boyer–Moore: Covered empty arrays, all-equal arrays, and no-majority cases.

### 3. Theoretical Complexity

- **Kadane's Algorithm:**

- Time Complexity:  $O(n)$  (single pass).
- Space Complexity:  $O(1)$ .

- **Boyer–Moore Majority Vote:**

- Time Complexity:  $O(n)$  (two passes).
- Space Complexity:  $O(1)$ .

Both algorithms are optimal in linear time and constant space, though applied to different problem domains.

### 4. Empirical Results

- **Kadane's Algorithm:**

- Scales linearly with input size, confirming theoretical  $O(n)$ .
- Extremely fast in practice since it involves simple additions and comparisons.

- **Boyer–Moore Algorithm:**



- Benchmarks confirmed linear time across different input distributions.
- Constant factors (comparisons, array accesses) vary depending on distribution

(e.g., all-equal vs. random).



- Verification adds extra passes but does not impact asymptotic performance.

## 5. Strengths and Weaknesses

- **Kadane’s Algorithm:**

-  Efficient, simple, and widely applicable.
-  Limited to maximum subarray sum; not useful for majority element detection.

- **Boyer–Moore Majority Vote:**

-  Very memory-efficient; scales to very large arrays.
-  Only works for majority element; must verify candidate in a second pass.

## 6. Conclusion

Both algorithms are efficient  $O(n)$  solutions with constant space, but optimized for different problem types. Kadane’s focuses on **optimization of numeric sums**, while Boyer–Moore targets **frequency detection**. Together, they highlight how linear-time algorithms can solve distinct computational challenges effectively.