

CSCI 4470 Algorithms

Part II Sorting and Order Statistics

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Sorting in Linear Time

Introduction

- Linear-time sorting algorithms are efficient for large datasets.
- They don't make comparisons but use other methods.
 - Counting Sort , Radix Sort , and Bucket Sort Algorithms

Lower Bounds for Sorting

- **Comparison sorts:** Any algorithm that sorts by comparing elements.
- **Lower bound of comparison sorts:** *Proves that any comparison sort algorithm requires $\Omega(n \log n)$ time in the WORST CASE.*

Decision Tree Model

- A tool used to prove the lower bound of comparison sorts.
- *Every comparison-based sorting algorithm has a decision tree*

Properties of Decision Trees

- **Nodes:** Represent decisions or comparisons.
- **Edges:** Represent the outcome of a decision.
- **Leaves:** Represent final outcomes or sorted permutations.

Height and Leaves

- **Height h :** The maximum number of comparisons needed in the worst case.
- **Number of Leaves:** Given n elements, the total number of leaves is $n!$ (n factorial), representing all possible sorted permutations.
- **Inequality:** To have all possible sorted permutations, $2^h \geq n!$.

Height Boundaries

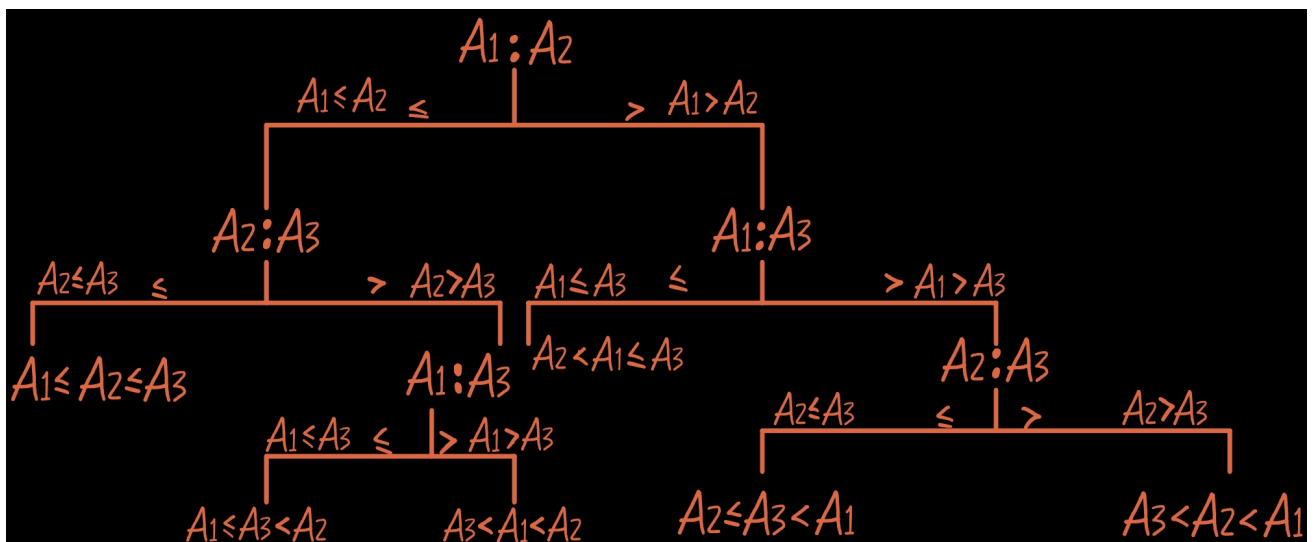
- The height h must satisfy $2^h \geq n!$ to ensure the tree has sufficient leaves.
- Using Stirling's approximation: $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$, the height is bounded by $h \geq n \log(n)$.
- **Logarithmic Boundaries:**
 - $h \geq \log(n!)$
 - $h \geq n \log(n)$

Decision Tree Model for Comparison Sort

Example of Decision Tree Given the permutation of 3 input values $\{A_1, A_2, A_3\}$, for an input of n elements, how many leaves do we have?

- **In a decision tree for n elements, the number of leaves, representing all possible sorted permutations, is $n!$ (n factorial).**
- By the example, with 3 elements $\{A_1, A_2, A_3\}$, there are $3! = 6$ leaves, each corresponding to a unique permutation.

- **Decision Tree for**



$\{A_1, A_2, A_3\}$:

- Visualizes the comparisons between the three elements to determine all possible sorted permutations.

- **Left Subtree:** Represents “less than or equal to” comparisons.
- **Right Subtree:** Represents “greater than” comparisons.
- **First Level:** We start by comparing A_1 and A_2 .
- **Second Level:** Depending on the result of the first comparison, we have two paths:
 - **Left Subtree:** If A_1 is less than or equal to A_2 , we compare A_2 and A_3 and A_1 and A_3 .
 - **Right Subtree:** If A_1 is greater than A_2 , we compare A_1 with A_3 and A_2 with A_3 .
- **Third Level:** Here, we reach the leaf nodes, which represent the sorted permutations of the input set.
- The sorted permutations corresponding to each leaf (L1 to L8):
- **L1:** $A_1 \leq A_2 \leq A_3$, **L2:** $A_1 \leq A_3 < A_2$, **L3:** $A_3 < A_1 \leq A_2$
- **L4:** $A_1 \leq A_2 < A_3$, **L5:** $A_2 < A_1 \leq A_3$, **L6:** $A_2 < A_3 < A_1$
- **L7:** $A_3 < A_2 < A_1$, **L8:** $A_2 < A_1 < A_3$

The height h of the decision tree:

- the height of the decision tree is based on the number of leaves, which is equal to the number of possible permutations ($n!$). The formula is:
 - $h \approx \log_2(n!)$
- By the example with 3 elements ($\{A_1, A_2, A_3\}$), we apply the formula:
 - $h \approx \log_2(3!) = \log_2(6) \approx 2.585$
- round up to get an integer value for the height:
 - $h = 3$
- ***This indicates that we will need up to 3 comparisons in the worst case.***

Given 3 input values $\{A_1, A_2, A_3\}$, we want to find out how many leaves we have in the decision tree.

1. Number of Leaves:

- In a decision tree for n elements, the number of leaves is $n!$.
- For $\{A_1, A_2, A_3\}$, there are $3! = 6$ leaves.
- Each leaf corresponds to a unique permutation of the input values.

2. Decision Tree Structure:

- **First Level:**
 - Compare A_1 and A_2 .
- **Second Level:**
 - If $A_1 \leq A_2$, compare A_2 and A_3 and A_1 and A_3 .
 - If $A_1 > A_2$, compare A_1 with A_3 and A_2 with A_3 .
- **Third Level:**
 - Reach the leaf nodes, representing the sorted permutations of the input set.

3. Sorted Permutations:

- Each leaf (L1 to L8) corresponds to a sorted permutation of $\{A_1, A_2, A_3\}$.

4. Height of the Decision Tree:

- Formula: $h \approx \log_2(n!)$.
- For $\{A_1, A_2, A_3\}$: $h \approx \log_2(3!) = \log_2(6) \approx 2.585$.
- Round up: $h = 3$.
- Indicates up to 3 comparisons in the worst case.

Determine Min Worst Case of The Decision Tree

1. Height and Leaves:

- The decision tree for a given comparison sort algorithm with input size = n has $n!$ leaves.
- Each leaf represents a distinct ordering of a_1, a_2, \dots, a_n .
- h is the height of the tree.
- $2^h \geq n!$.

2. Inequality:

- $n! \geq \sqrt{2\pi n}$.
- $h \geq \log(n!)$.
- $h \geq n \log(n)$.

3. Maximum Number of Leaves in the Tree:

- If a binary tree has height h , the maximum number of leaves in the tree is 2^h .

Counting Sort Algorithm

- **Working:** Sorts integers within a range. It counts the occurrence of each element to sort them.
- **Time complexity:** $O(n + k)$, where n is the number of elements and k is the range of input values.
- **Definition:** Sorts integers within a specific range by counting occurrences.

Characteristics

- Non-comparison based.
- Assumes input of integers within a small range.
- Utilizes an auxiliary array for determining the position of each element.

Counting Sort($A[1 \dots n]$, $B[1 \dots n]$, k)

1. let $C[0 \dots k]$ be a new array
2. for $i = 0$ to k // initialize C
3. $C[i] = 0$
4. for $j = 1$ to $A.length$ //count keys
5. $C[A[j]] = C[A[j]] + 1$
6. for $i = 1$ to k // accumulate C (determine rank of each key)
7. $C[i] = C[i] + C[i - 1]$
8. for $j = A.length$ downto 1 // place keys in B
9. $B[C[A[j]]] = A[j]$
10. $C[A[j]] = C[A[j]] - 1$

Algorithm Steps

Counting Sort($A[1..n]$, $B[1 \dots n]$, k)

1. Let $C[0 \dots k]$ be a new array
2. Initialize C with zeros
3. Count the occurrence of each element in A and store in C
4. Update C to store the cumulative count
5. Build the sorted array B using C and A

Example of Counting Sort

Given: $A = \{3, 2, 4, 3, 0, 4, 3, 1, 2\}$

Step 1:, Find Max and Min for the range:

- Max is 4, Min is 0

Question: What are the contents of array C at different stages?

Step 2: After initialization of the Counting Array C (lines 1-2): $C = \{0, 0, 0, 0, 0\}$,

- The size of the Counting Array is $k + 1$, where k is 4, $C[k + 1] = C[5]$

Step 3: After counting occurrences (lines 3-4): $C = \{1, 1, 2, 3, 2\}$

- $C[0] = 1, C[1] = 1, C[2] = 2, C[3] = 3, C[4] = 2$
- The occurrences of the elements of A

Step 4: After cumulative count (line 5): $C = \{1, 2, 4, 7, 9\}$

- **After Cumulative Count (After line 7):**
 - $C[0] = 1$ (remains the same)
 - $C[1] = C[1] + C[0] = 1 + 1 = 2$
 - $C[2] = C[2] + C[1] = 2 + 2 = 4$
 - $C[3] = C[3] + C[2] = 3 + 4 = 7$
 - $C[4] = C[4] + C[3] = 2 + 7 = 9$

Step 5: Resulting sorted array $B = \{0, 1, 2, 2, 3, 3, 3, 4, 4\}$

Step 5.1: Given the cumulative count array $C = \{1, 2, 4, 7, 9\}$ and the original array $A = \{3, 2, 4, 3, 0, 4, 3, 1, 2\}$,

Step 5.2 (Line 8-10): We iterate over array A from the end to the start.

Step 5.3: For each element $A[j]$ in array A , we do the following:

- Find the cumulative count $C[A[j]]$
- Place $A[j]$ at index $C[A[j]] - 1$ in array B
- Decrease the cumulative count $C[A[j]]$ by 1

Step 5.4:

- **For** $A[8] = 2$ (the last element in the array):
 - $C[2] = 4$, so we place 2 at index $4 - 1 = 3$ in array B : $B[3] = 2$ (0-indexed)
 - We then decrease $C[2]$ to 3: $C = \{1, 2, 3, 7, 9\}$
- **For** $A[7] = 1$:
 - $C[1] = 2$, so we place 1 at index $2 - 1 = 1$ in array B : $B[1] = 1$ (0-indexed)
 - We then decrease $C[1]$ to 1: $C = \{1, 1, 3, 7, 9\}$
- **For** $A[6] = 3$:
 - $C[3] = 7$, so we place 3 at index $7 - 1 = 6$ in array B : $B[6] = 3$ (0-indexed)
 - We then decrease $C[3]$ to 6: $C = \{1, 1, 3, 6, 9\}$

We continue this process for all elements in array A , updating array B and array C as we go. After processing all elements, we will obtain the sorted array B :

$$B = \{0, 1, 2, 2, 3, 3, 3, 4, 4\}$$

Analysis of Counting Sort

- **Time Complexity:** $T(n) = O(n + k)$. If $k = O(n)$, then $T(n) = O(n)$. **
- **Stability:** Counting sort is stable if the auxiliary array is used correctly.
- **In-Place:** Not in-place due to the auxiliary array.
- **Comparisons:** No element comparisons are made.
- Counting sort is stable and works in $O(n + k)$ where k is the range of input.

Reasonable Values for k 合理的 k 值

- Depends on n and memory capacity.
- Avoid large k values to prevent performance issues.
- Suitable for small bit-size numbers (like 4 or 8 bits) but not for larger bit sizes (like 16 or 32 bits).

Application

- The Counting sort is a subroutine in the Radix sort algorithm, a linear-time sorting algorithm.

Example of the Counting Sort 02

Example 02, $A = \{2, 5, 3, 0, 2, 3, 0, 3\}$

Step 1: Identify the Range

- Find the maximum and minimum values in the Input Array.
 - Min: 0, Max: 5

Step 2: Create a Count Array

- The size of Array C is $k + 1 = 5 + 1 = 6$

- The Counting Array $C = \{0, 0, 0, 0, 0, 0\}$
- After processing the first element (2) 處理第一個元素 (2) 後 :
 - $C = \{0, 0, 1, 0, 0, 0\}$

Input Array	2	5	3	0	2	3	0	3
Index[i]	A[0]	A[1]	A[2]	A[3]	A[4]	A[5]	A[6]	A[7]
Counting Array C	2	0	2	3	0	1		
Index[] of C $5 + 1 = 6$	C[0]	C[1]	C[2]	C[3]	C[4]	C[5]	x	x
Cumulated Array C	2	2	4	7	7	8		
Index of C	C[0]	C[1]	C[2]	C[3]	C[4]	C[5]	x	x

Step 3: Calculate Cumulative Counts 計算累積計數

- $C[0]$
 - Remains the same, $C[0] = 2$ 保持不變, $C[0] = 2$
- $C[1]$
 - $C[1] = C[1] + C[0]$
 - $C[1] = 0 + 2 = 2$
- $C[2]$
 - $C[2] = C[2] + C[1]$
 - $C[2] = 2 + 2 = 4$
- $C[3]$
 - $C[3] = C[3] + C[2]$
 - $C[3] = 3 + 4 = 7$
- $C[4]$
 - $C[4] = C[4] + C[3]$
 - $C[4] = 0 + 7 = 7$
- $C[5]$
 - $C[5] = C[5] + C[4]$
 - $C[5] = 1 + 7 = 8$

$$C = \{2, 2, 4, 7, 7, 8\}$$

Step 4: Build the Sorted Array

- Create an output array of the same size as the input array.
- Traverse the input array from end to start, and use the cumulative count array to find the correct position for each element in the output array.

Initial State

A = [2, 5, 3, 0, 2, 3, 0, 3]
C = [2, 2, 4, 7, 7, 8]
B = [-1, -1, -1, -1, -1, -1, -1, -1]

Input Array	2	5	3	0	2	3	0	3
Index[i]	A[0]	A[1]	A[2]	A[3]	A[4]	A[5]	A[6]	A[7]
Cumulated Array C	2	2	4	7	7	8		
Index of C	C[0]	C[1]	C[2]	C[3]	C[4]	C[5]	x	x
Sorted Array B								
Index[i]	B[0]	B[1]	B[2]	B[3]	B[4]	B[5]	B[6]	B[7]

Step 1: Consider $A[7] = 3$

- Find $C[3] = 7$
- Set $B[7 - 1] = B[6] = 3$
- Decrease $C[3]$ by 1, $C[3] = 6$

B = [-1, -1, -1, -1, -1, -1, 3, -1]
C = [2, 2, 4, 6, 7, 8]

Step 2: Consider $A[6] = 0$

- Find $C[0] = 2$
- Set $B[2 - 1] = B[1] = 0$
- Decrease $C[0]$ by 1, $C[0] = 1$

B = [-1, 0, -1, -1, -1, -1, 3, -1]
C = [1, 2, 4, 6, 7, 8]

Step 3: Consider $A[5] = 3$

- Find $C[3] = 6$
- Set $B[6 - 1] = B[5] = 3$
- Decrease $C[3]$ by 1, $C[3] = 5$

B = [-1, 0, -1, -1, -1, 3, 3, -1]

C = [1, 2, 4, 5, 7, 8]

Step 4: Consider $A[4] = 2$

- Find $C[2] = 4$
- Set $B[4 - 1] = B[3] = 2$
- Decrease $C[2]$ by 1, $C[2] = 3$

B = [-1, 0, -1, 2, -1, 3, 3, -1]

C = [1, 2, 3, 5, 7, 8]

Counting Sort Algorithm_

- **Best Case**
 - **Scenario:** When all elements are the same.
 - **Time Complexity:** $O(n + k)$, where n is the number of elements and k is the range of input values.
- **Worst Case**
 - **Scenario:** When the elements are evenly distributed across a large range.
 - **Time Complexity:** $O(n + k)$. The worst case is the same as the best case because Counting Sort has a linear time complexity.
- **Average Case**
 - **Scenario:** In general scenarios where the input has a moderate range and distribution.
 - **Time Complexity:** $O(n + k)$. The average case is also linear, making Counting Sort efficient for small to moderate ranges of k .
- In all cases, the space complexity remains $O(n + k)$ because of the auxiliary arrays used in the algorithm.

Radix Sort

- **Working:** Sorts numbers digit by digit from least significant digit (LSD) to most significant digit (MSD) or vice versa.
- **Steps:**
 - i. Determine the maximum number to know the number of digits.
 - ii. Sort numbers using Counting Sort for each digit.
- **Stable Algorithm:** Maintains the relative order of records with equal keys.
- **Non-comparison based algorithm:** Does not use comparison operators to sort.
- **Time complexity:** $O(d(n + k))$, where d is the number of digits, n is the number of elements, and k is the range of input values.
- Processes each digit of the number.

- Starts from the least significant digit and moves to the most significant.
- Uses a stable sort (often counting sort) to sort each digit.

Analysis of Radix Sort

- Radix sort works in $O(nk)$ for n numbers of k digits.

RADIX-SORT(A,D)

1. for $i = 1$ to d

2. // use a stable sort to sort array A on digit i

- Use induction on length of input values (i.e., number of digits / bits) to prove this algorithm is correct
- This is left as an exercise

- **Example:** For numbers 170, 45, 75, 90, 802, 24, 2, 66, sort based on units place, then tens place, then hundreds place.

initial	Step 1	Step 2	Step 3
-,-,-	-, -, ↓	-, ↓, -	↓, -, -
170	170	2	2
45	90	802	24
75	802	24	45
90	2	45	66
802	24	66	75
24	45	170	90
2	75	75	170
66	66	90	802

- **Example 02,** Array A{329, 457, 657, 839, 436, 720, 355}, to use Radix Sort

initial	Step 1	Step 2	Step 3
-,-,-	-, -, ↓	-, ↓, -	↓, -, -
329	720	720	329
457	355	329	355

initial	Step 1	Step 2	Step 3
657	436	436	436
839	457	839	457
436	657	355	657
720	329	457	720
355	839	657	839

Complexity Analysis

- **Time Complexity:** $O(nk)$, where n is the number of elements and k is the number of digits in the maximum number.
- **Space Complexity:** $O(n + k)$, as it uses Counting Sort as a subroutine which requires additional space.
- **Stability:** Radix Sort is stable, meaning numbers with the same value appear in the output array in the same order as they do in the input array.

Note

- Radix Sort is efficient when k is not significantly larger than n .
- It's a non-comparative sorting algorithm.
- Best suited for sorting integers or strings.

Was used by the card-sorting machines to read punch cards

- How IBM made its money initially

The key is to sort digit by digit

- Start with least significant digit (or bit)
- Starting with the most significant digit requires extra storage.
- Sorting method used to sort each digit must be stable

1. First sort units digit (or bit)
2. Then sort tens digit (or twos bit)
3. Continue to most significant

Radix-Sort(A,d)

1. for $i = 1$ to d
2. use a stable sort to sort array A on digit i

- Use induction on length of input values (i.e., number of digits / bits) to prove this algorithm is correct

Bucket Sort

- **Working:** Distributes elements into buckets and then sorts each bucket individually.
- **Time complexity:** $O(n)$ average time under uniform distribution, but $O(n^2)$ in the worst case.
- Divides the interval of input numbers into equal-sized buckets.
- Distributes the numbers into buckets.
- Sorts each bucket and then gathers numbers back.

Analysis of Algorithms:

- Bucket sort assumes that input is uniformly distributed to achieve linear time.
- Worst Case is all numbers in one bucket Cost is $O(n^2)$

Limitations

- Counting sort is not suitable for sorting strings of varying length.
- Radix and bucket sort are not comparison-based and hence might not be suitable for all types of data.

Bucket Example:

- *A[i] are the value of each element,
and n is the A.length which is 10 in the example*

Index(A)	A	$\lfloor A[i] \times n \rfloor$	B	Sorted
A[0]	.38	$.38 \times 10 = \lfloor 3.8 \rfloor = 3$	0	
A[1]	.93	$.93 \times 10 = \lfloor 9.3 \rfloor = 9$	1	.11
A[2]	.77	$.77 \times 10 = \lfloor 7.7 \rfloor = 7$	2	.29
A[3]	.11	$.11 \times 10 = \lfloor 1.1 \rfloor = 1$	3	.38 \rightarrow .39
A[4]	.95	$.95 \times 10 = \lfloor 9.5 \rfloor = 9$	4	.43
A[5]	.29	$.29 \times 10 = \lfloor 2.9 \rfloor = 2$	5	.95
A[6]	.72	$.72 \times 10 = \lfloor 7.2 \rfloor = 7$	6	

Index(A)	A	$\lfloor A[i] \times n \rfloor$	B	Sorted
A[7]	.43	$.43 \times 10 = \lfloor 4.3 \rfloor = 4$	7	$.77 \rightarrow .72$
A[8]	.39	$.39 \times 10 = \lfloor 3.9 \rfloor = 3$	8	
A[9]	.99	$.99 \times 10 = \lfloor 9.9 \rfloor = 9$	9	$.93 \rightarrow .95 \rightarrow .99$

Analysis of the Average Run Time of Bucket Sort Algorithm

BUCKET-SORT (A)

```

1. n = A.length
2. let B[0 ... n-1] be a new array
3. for i = 0 to n-1
4.   make B[i] an empty list
5. for i = 1 to n
6.   insert A[i] into list B[ $\lfloor n \cdot A[i] \rfloor$ ]
7. for i = 0 to n-1
8.   sort list B[i] with insertion sort
// Cost is just O(c) constant
9. concatenate the lists B[0], B[1],
..., B[n-1] together in order

```

All lines except line 8 (sort bucket) takes $O(n)$ time in the worst case

To analyze the average run time, we need to find $E[T(n)]$, which is given by:

$$E[T(n)] = \Theta(n) + O\left(\sum_{i=0}^{n-1} E[n_i^2]\right)$$

Where n_i is the number of elements in bucket $B[i]$.

Step 1: Define n_i

Define n_i as the number of elements in bucket i . It can be expressed in terms of X_{ij} as:

$$n_i = \sum_{j=1}^n X_{ij}$$

This equation means that we sum up all the X_{ij} for a given i , where X_{ij} is 1 if $A[j]$ falls in bucket i and 0 otherwise.

Step 2: Express $E[n_i^2]$

We aim to find $E[n_i^2]$, which can be expressed as:

$$E[n_i^2] = E\left[\left(\sum_{j=1}^n X_{ij}\right)^2\right]$$

Step 3: Expand the Square

Expanding the square gives:

$$E[n_i^2] = E \left[\sum_{j=1}^n \sum_{k=1}^n X_{ij} X_{ik} \right]$$

Step 4: Separate the Expectation

We can separate the expectation over the sum:

$$E[n_i^2] = \sum_{j=1}^n \sum_{k=1}^n E[X_{ij} X_{ik}]$$

Step 5: Find $E[X_{ij} X_{ik}]$

Now we need to find $E[X_{ij} X_{ik}]$ for two cases: when $j = k$ and when $j \neq k$.

1. When $j = k$:

$$E[X_{ij} X_{ik}] = E[X_{ij}^2]$$

But since X_{ij} is a Bernoulli variable (it can only take values 0 or 1), $X_{ij}^2 = X_{ij}$. So,

$$E[X_{ij}^2] = E[X_{ij}] = \frac{1}{n}$$

2. When $j \neq k$:

The events are independent, so

$$E[X_{ij} X_{ik}] = E[X_{ij}] E[X_{ik}] = \frac{1}{n} \cdot \frac{1}{n} = \frac{1}{n^2}$$

Step 6: Substitute the Values Back

Now we substitute these values back into the equation for $E[n_i^2]$:

$$E[n_i^2] = \sum_{j=1}^n \sum_{k=1}^n E[X_{ij} X_{ik}]$$

This sum can be broken down into two parts: when $j = k$ and when $j \neq k$.

So we have:

$$E[n_i^2] = \sum_{j=1}^n E[X_{ij}^2] + \sum_{j=1}^n \sum_{k \neq j} E[X_{ij} X_{ik}]$$

Now substituting the values we found in step 5:

$$E[n_i^2] = \sum_{j=1}^n \frac{1}{n} + \sum_{j=1}^n \sum_{k \neq j} \frac{1}{n^2}$$

Step 7: Calculate the Sums

Now we calculate each sum separately:

1. $\sum_{j=1}^n \frac{1}{n} = 1$
2. $\sum_{j=1}^n \sum_{k \neq j} \frac{1}{n^2} = n(n-1) \frac{1}{n^2}$

Step 8: Final Calculation

Now we add these two results together to find $E[n_i^2]$:

$$E[n_i^2] = 1 + n(n-1) \frac{1}{n^2} = 1 + (n-1) \frac{1}{n} = 2 - \frac{1}{n}$$

Now we have found $E[n_i^2]$ as $2 - \frac{1}{n}$.

$$E[n_i^2] = 1 + n(n-1) \frac{1}{n^2} = 1 + (n-1) \frac{1}{n} = 2 - \frac{1}{n}$$

Let n_i = number of elements in bucket $B[i]$

We are looking to find the expected time $E[T(n)]$, which is given by:

$$E[T(n)] = \Theta(n) + O\left(\sum_{i=0}^{n-1} E[n_i^2]\right)$$

Substituting the values we found for $E[X_{ij}X_{ik}]$ when $j = k$ and $j \neq k$, we find:

$$E[n_i^2] = n \left(\frac{1}{n}\right) + n(n-1) \left(\frac{1}{n^2}\right) = 2 - \frac{1}{n}$$

So, we can now find $E[T(n)]$ using the formula we derived earlier:

$$T(n) = \Theta(n) + O\left(n \left(2 - \frac{1}{n}\right)\right) = \Theta(n) + O(n) = \Theta(n)$$

This shows that the average running time of the bucket sort algorithm is linear.

Conclusion

This analysis shows that the average running time of the bucket sort algorithm is linear, $\Theta(n)$.