



K.R. MANGALAM UNIVERSITY
THE COMPLETE WORLD OF EDUCATION

Course: Data Structure

(Course Code : ENCS205)

UNIT-1: Foundations of Data Structures

School of Engineering & Technology
K.R. Mangalam University, Gurugram (Haryana)

SESSION 13:

Introduction to Sparse Matrices

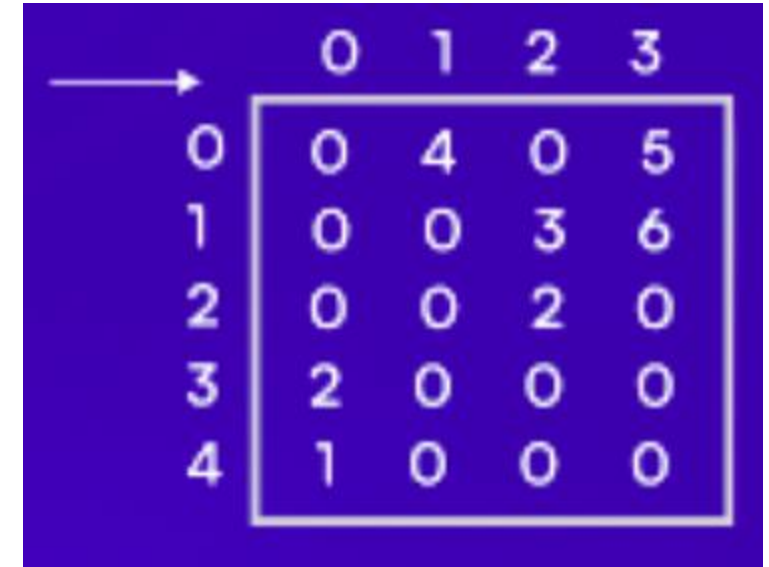
Session 4: Learning Objectives

By the end of this session, will be able to:

- Understand what a sparse matrix is and why it is used
- Identify different types of sparse matrices (lower, upper, tri-diagonal)
- Explain the need for efficient representation of sparse matrices
- Describe and compare array-based and linked list-based representations
- Analyze and Compare the time and space complexities of each method.

What Is a Sparse Matrix?

- A matrix where **most elements are zero**.
- Common in large-scale, real-world datasets—e.g., in machine learning, data science, and graph theory



	0	1	2	3
0	0	4	0	5
1	0	0	3	6
2	0	0	2	0
3	2	0	0	0
4	1	0	0	0

Why Use Special Representations?

- **Problems with Normal Storage:**

- Wastes memory (stores zeros unnecessarily)
- More processing time for traversals/operations

- **Advantages of Sparse Storage:**

- Saves memory
- Reduces time complexity for operations involving only non-zero elements

- **Example:**

- A 1000×1000 matrix = **1 million elements**
- If only 1000 are non-zero, normal storage wastes **99.9% memory**

Types of Sparse Matrices

- **Lower Triangular:** all entries above the main diagonal are zero
- **Upper Triangular:** all entries below the main diagonal are zero
- **Tri-diagonal:** non-zero values only on main, sub-, and super-diagonals

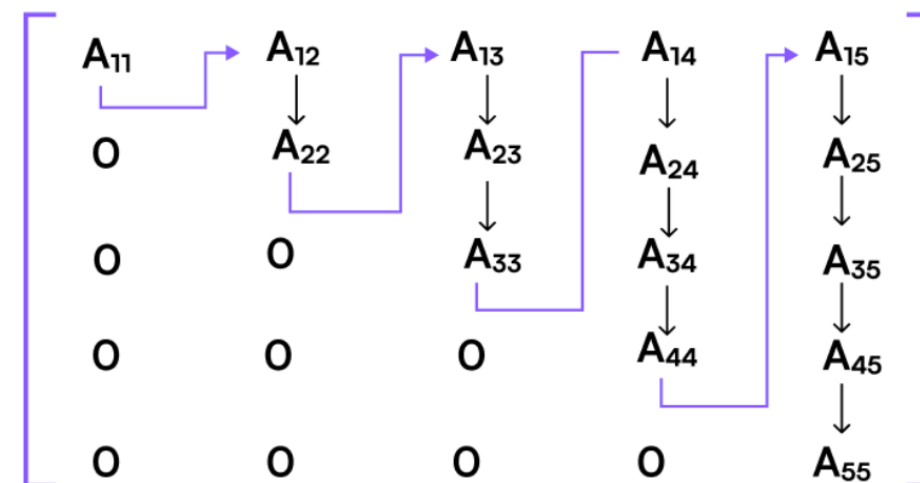
Lower Triangular Matrix

Representation of Lower Triangular Matrix $A[5,5]$

$$\begin{bmatrix} A_{11} & 0 & 0 & 0 & 0 \\ A_{21} & A_{22} & 0 & 0 & 0 \\ A_{31} & A_{32} & A_{33} & 0 & 0 \\ A_{41} & A_{42} & A_{43} & A_{44} & 0 \\ A_{51} \longrightarrow A_{52} \longrightarrow A_{53} \longrightarrow A_{54} \longrightarrow A_{55} \end{bmatrix}$$

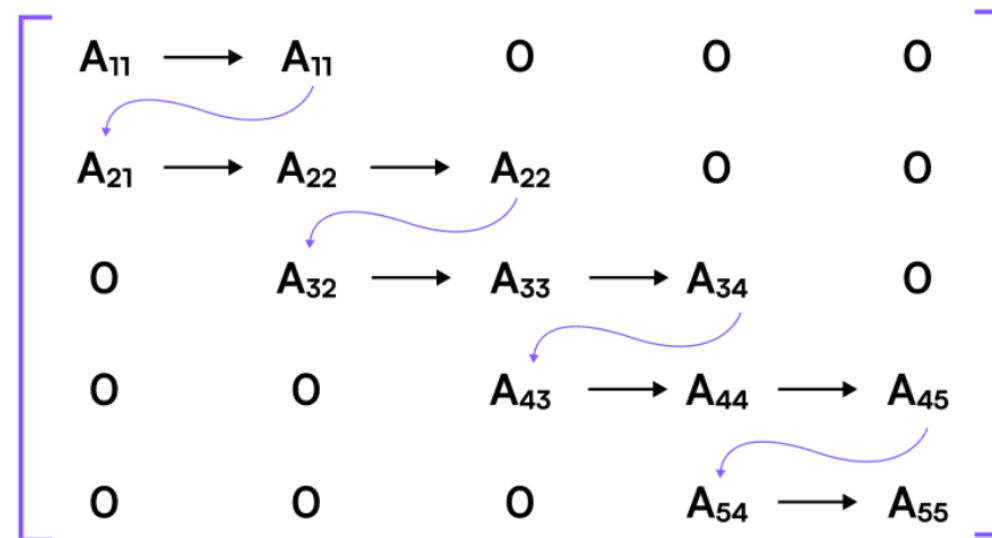
Upper Triangular Matrix

Representation of Upper Triangular Matrix
 $A[5,5]$



Tri-Diagonal Sparse Matrix

Representation of Tri-Diagonal Sparse Matrix
 $A[5,5]$



Representation Methods

1. Using Array (Compact Storage)

- Example C implementation stores non-zero entries as a compact $3 \times N$ array (row, col, value)

2. Using Linked List

- Use a list of nodes; each node stores value, row index, column index, and a pointer to the next node

Array (Triplet) Representation

- **Idea:** Store only non-zero elements and their positions
- **Format:** Three arrays or a 2D array with 3 columns
 - Row index
 - Column index
 - Value

• **Example:**
Original Matrix (5×4):

Sparse Matrix →

	0	1	2	3
0	0	4	0	5
1	0	0	3	6
2	0	0	2	0
3	2	0	0	0
4	1	0	0	0

Table Structure

Row	Column	Value
0	1	4
0	3	5
1	2	3
1	3	6
2	2	2
3	0	2
4	0	1
5	4	7

Array Representation — Implementation

Pseudocode:

function sparse_matrix_to_triplet(matrix, rows, cols):

1. triplet = []
2. for i in range(rows):
3. for j in range(cols):
4. if matrix[i][j] != 0:
5. triplet.append([i, j, matrix[i][j]])
6. return triplet

- **Space Complexity:** $O(k)$ where k = number of non-zero elements
- **Time Complexity:** $O(N \times M)$ for conversion

Linked List Representation

- **Idea:** Store non-zero elements as nodes in a linked list
- **Advantages over array:**
 - Dynamic size
 - Easier to insert/delete non-zero elements
- **Example:**
- The same matrix as before becomes:
 $(0,1,4) \rightarrow (0,3,5) \rightarrow (1,2,3) \rightarrow (1,3,6) \rightarrow (2,2,2) \rightarrow (3,0,6) \rightarrow (4,0,1)$

Linked List — Implementation

Pseudocode:

```
function sparse_matrix_to_linked_list(matrix, rows, cols):  
    head = None  
    for i in range(rows):  
        for j in range(cols):  
            if matrix[i][j] != 0:  
                newNode = Node(i, j, matrix[i][j])  
                insert_at_end(head, newNode)  
    return head
```

Complexity:

- Space: $O(k)$
- Conversion Time: $O(N \times M)$

Complexity Comparison

Representation	Time Complexity (Conversion)	Space Complexity	Pros	Cons
Array	$O(N \times M)$	$O(k)$	Simple, fast lookup	Fixed size, harder insert/delete
Linked List	$O(N \times M)$	$O(k)$ + pointer mem	Dynamic, easy modifications	Slower lookup

Summary of Session 14

- Sparse matrices save space and time when dealing with mostly zero elements.
- **Array Representation:** Good for quick access, compact storage.
- **Linked List Representation:** Flexible, better for frequent updates.
- Choosing the right representation depends on **access pattern** and **update frequency**.

Exercise

1. You have a 1000×1000 matrix with only 1500 non-zero elements.
 - a. Calculate the space complexity for storing it in normal 2D array form and in array (triplet) representation, assuming each integer takes 4 bytes.
 - b. Which saves more memory and by how much?

2. Consider the following triplet representation:

Row	Col	Value

0	2	5
0	3	8
1	0	3
2	1	6

Reconstruct the 3×4 matrix from this triplet form.

Q3. For a matrix with $n \times n$ rows, $m \times m$ columns, and $k \times k$ non-zero elements:

- a. What is the time complexity of converting it to array (triplet) form?
- b. What is the space complexity of the triplet form?

Q4. In linked list representation of a sparse matrix:

- a. If you need to find the value at position (i, j) , what is the worst-case time complexity? Why?
- b. Suggest one modification to improve lookup time.



Thank You



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