
Chapter 2

Arrays

Data Objects

- Data object: A collection of instances is a data object.
 - Order of the instances doesn't matter.
 - E.g., {1,2,3} and {3,2,1} are the same data object.
 - E.g., Integer = {0, +1, -1, +2, -2, +3, -3, ...}
 - E.g., daysOfWeek = {S,M,T,W,Th,F,Sa}

Data Objects

- Instances may or may not be related
- myDataObject = {man, chairs, apple, 100.005, @green, Jackie, 3}

Data Structures

- Data Structure =
data object +
relationships that exist among instances
and/or elements that comprise an instance
- Among instances of integer
 $369 < 370$

Data Structures

- Among elements that comprise an instance

369

3 is more significant than 6

3 is immediately to the left of 6

9 is immediately to the right of 6

Data Structures

- Relationships may exist by specifying **operations** on one or more instances.
 - Examples of operations: add, subtract, predecessor, multiply, ..., etc.

Linear (or Ordered) Lists

- One form of the simplest data structures
- Instances are of the form

$$(e_0, e_1, e_2, \dots, e_{n-1})$$

where e_i denotes a list element and $n-1 \geq 0$ is finite

- List size is n

Linear Lists

- $L = (e_0, e_1, e_2, e_3, \dots, e_{n-1})$

Relationships:

e_0 is the zero'th (or front) element

e_{n-1} is the last element

e_i immediately precedes e_{i+1}

Linear List Examples

Students in MyClass =

(Jackie, Abs, Jason, Hank, Mary, ..., Judy)

Exams in MyClass =

(exam1, exam2, exam3)

Days of Week = (S, M, T, W, Th, F, Sa)

Months = (Jan, Feb, Mar, Apr, ..., Nov, Dec)

Linear List Operations — Length()

- Determine number of elements in list
- $L = (a, b, c, d, e)$
- Length = 5

Linear List Operations — Retrieve(theIndex)

- Retrieve element with given index

➤ $L = (a, b, c, d, e)$

➤ $\text{Retrieve}(0) = a$

➤ $\text{Retrieve}(2) = c$

➤ $\text{Retrieve}(-1) = \text{error}$

Linear List Operations —IndexOf(theElement)

- Determine the index of an element
- E.g., $L = (a, b, d, b, a)$
 - $IndexOf(d) = 2$
 - $IndexOf(a) = 0$
 - $IndexOf(z) = -1$ (if the index of a nonexistent element is defined to be -1)

Linear List Operations — Delete(theIndex)

- Delete an element with given index and return its value
- E.g., $L = (a,b,c,d,e,f,g)$
 - *Delete*(2) returns c
 - ◆ and L becomes (a,b,d,e,f,g)
 - Indices of d,e,f , and g decreased by 1
 - *Delete*(-1) => error

Linear List Operations — Insert(theIndex, theElement)

- Insert an element so that the new element has a specified index
- $L = (a, b, c, d, e, f, g)$
 - $Insert(0, h) \Rightarrow L = (h, a, b, c, d, e, f, g)$
 - Index of a, b, c, d, e, f, g increased by 1
- $L = (a, b, c, d, e, f, g)$
 - $Insert(2, h) \Rightarrow L = (a, b, h, c, d, e, f, g)$
 - Index of c, d, e, f, g increased by 1
 - $Insert(10, h) \Rightarrow$ error

Data Structure Specification

- Abstract Data Type
 - I.e., implementations are language-independent
- Use **classes** for implementations of data structures in C++

Linear List as a Abstract Data Type

AbstractDataType *LinearList*

{

instances

ordered finite collections of elements

operations

IsEmpty(): return true iff the list is empty, false otherwise

Length(): return the list size (i.e., number of elements in the list)

Retrieve(index): return the *index*-th element of the list

IndexOf(x): return the index of the first occurrence of *x* in
the list, return -1 if *x* is not in the list

Delete(index): remove and return the *index*-th element,
elements with higher index have their index reduced by 1

Insert(theIndex, x): insert *x* as the *theIndex*-th element, elements
with *index* \geq *theIndex* have their index increased by 1

}

Linear List as a C++ Class

- To specify a general linear list as a C++ class, we need to use a template class.
- Will study C++ templates later.
- For now we restrict ourselves to linear lists whose elements are integers.

Linear List as a C++ Class

- `class LinearListOfIntegers`

```
{  
    bool IsEmpty() const;  
    int length() const;  
    int Retrieve(int index) const;  
    int IndexOf(int theElement) const;  
    int Delete(int index);  
    void Insert(int index, int theElement);  
}
```

- “Lists” can be implemented by a type of data structures — array. So let’s talk about “array” now!

Array

- Array:
 - a set of **index** and **value**
- Data structure:
 - For each index, there is a value associated with that index.
- Can be implemented by consecutive memory
- Example: `int a[5]`
 - `a[0]`, ..., `a[4]` each contains an integer

Operations on Ordered List

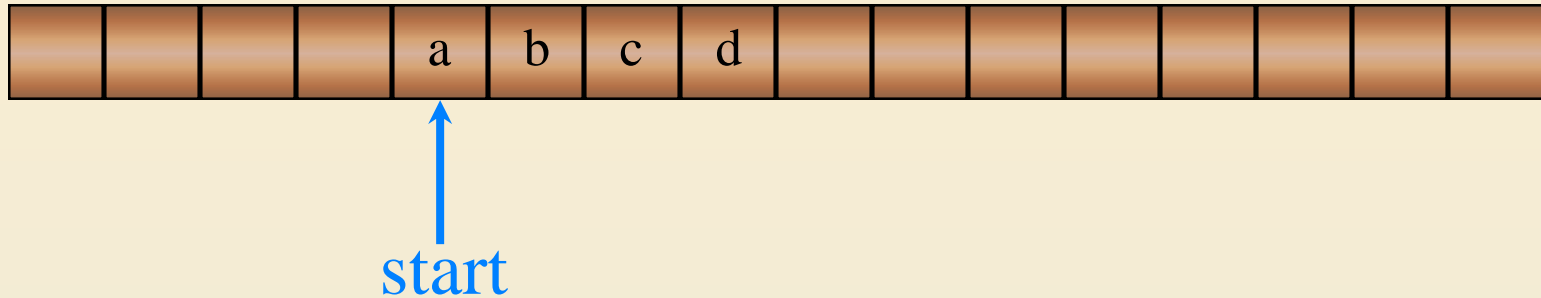
- (1) Find the length, n , of the list.
- (2) Read the items from left to right (or right to left).
- (3) Retrieve the i -th element.
- (4) Replace the old value by a new one at the i -th position.
- (5) **Insert a new element** at the position i , causing elements numbered $i, i+1, \dots, n$ to be numbered $i+1, i+2, \dots, n+1$
- (6) **Delete the element** at position i , causing elements numbered $i+1, \dots, n$ to be numbered $i, i+1, \dots, n-1$

Implementation on Ordered List by Array

- Implementing ordered list by array
 - Sequential mapping
 - Operations (1)~(4): OK
 - But operations 5 and 6 need data movement
 - this may be costly
- This overhead motivates us to consider non-sequential mapping of order lists in Chapter 4
 - Linked list (we will talk about this later in this course)

Example of Linear List – 1D Array Representation

Memory



- 1-dimensional array $x = [a, b, c, d]$
- Map into contiguous memory locations
- $\text{Location}(x[i]) = \text{start} + i$

2D Arrays

- The elements of a 2-dimensional array **a** declared as:

```
int [][]a = new int[3][4];
```

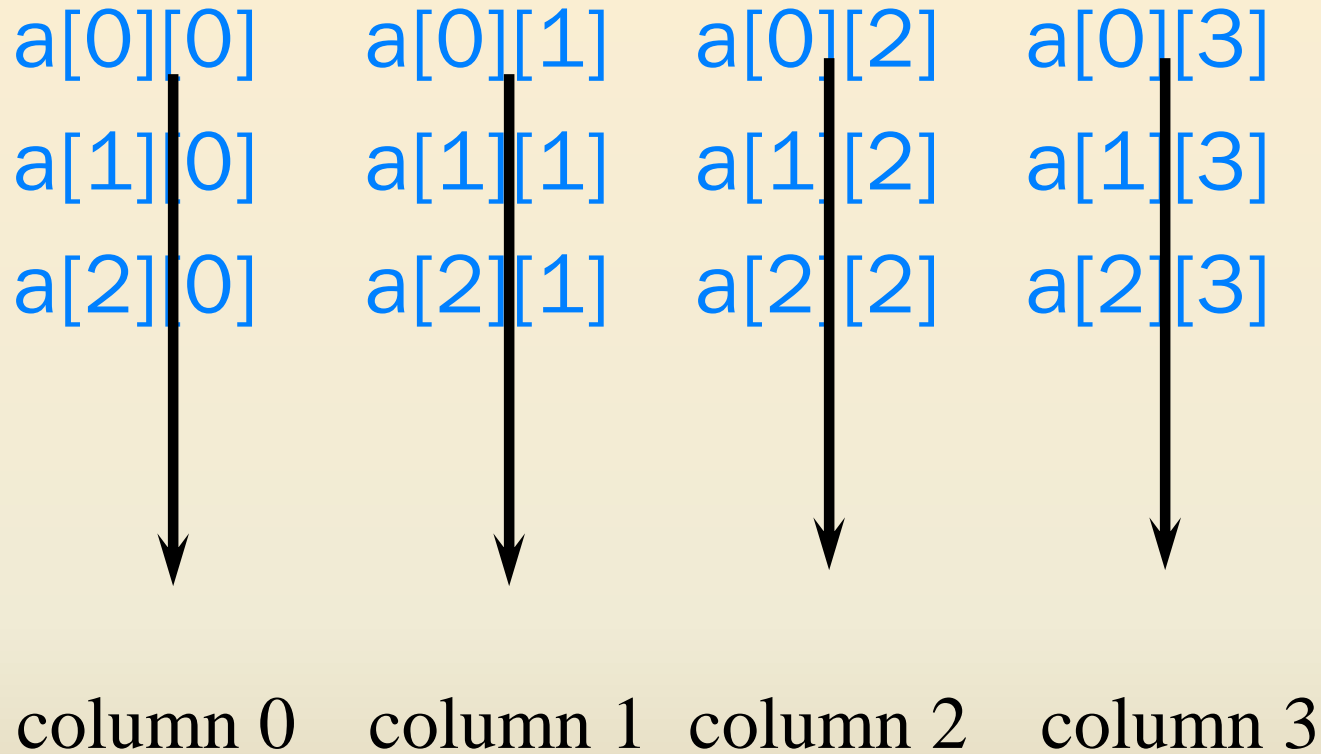
- may be shown as a table

a[0][0]	a[0][1]	a[0][2]	a[0][3]
a[1][0]	a[1][1]	a[1][2]	a[1][3]
a[2][0]	a[2][1]	a[2][2]	a[2][3]

Rows of a 2D Array

`a[0][0]` `a[0][1]` `a[0][2]` `a[0][3]` → row 0
`a[1][0]` `a[1][1]` `a[1][2]` `a[1][3]` → row 1
`a[2][0]` `a[2][1]` `a[2][2]` `a[2][3]` → row 2

Columns of a 2D Array



Row-Major Mapping

- Example of 3 x 4 array:

a b c d

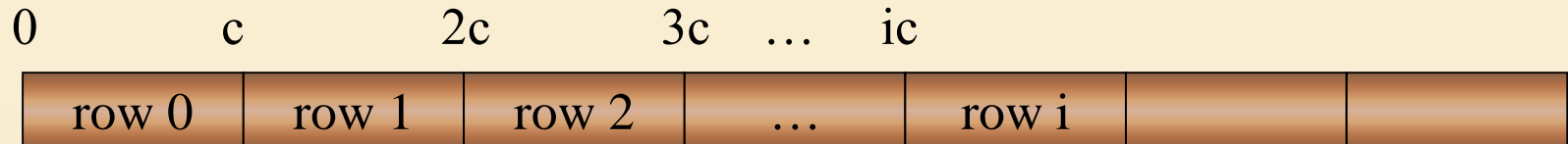
e f g h

i j k l

- To store 2D arrays by contiguous memory locations, use 1D array as follows:
 - Convert into 1D array "y" by collecting elements by rows.
 - Within a row elements are collected from left to right.
 - Rows are collected from top to bottom.
 - We get $y[] = \{a, b, c, d, e, f, g, h, i, j, k, l\}$

row 0	row 1	row 2	...	row i		
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Locating Element $x[i][j]$



- Assume x has r rows and c columns
- Each row has c elements
- i rows to the left of row i
- So ic elements to the left of $x[i][0]$
- So $x[i][j]$ is mapped to position
 - $ic + j$ of the 1D array

Column-Major Mapping

a b c d

e f g h

i j k l

- Convert into 1D array “y” by collecting elements by columns.
- Within a column elements are collected from top to bottom.
- Columns are collected from left to right.
- We get $y = \{a, e, i, b, f, j, c, g, k, d, h, l\}$
- We'll talk more about multi-dimensional arrays later if time permitted.

Matrix

- An entity consists of a table of values
 - Has rows and columns, but numbering begins at 1 rather than 0.
 - E.g.

a	b	c	d	row 1
e	f	g	h	row 2
i	j	k	l	row 3

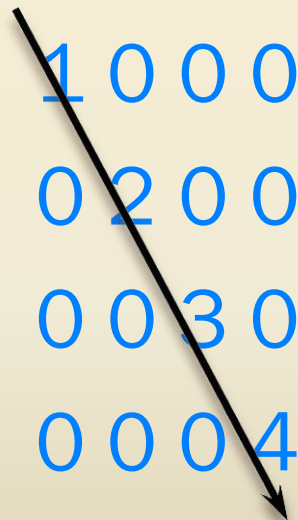
- Use notation $x(i,j)$ rather than $x[i][j]$.
- May use a 2D array to represent a matrix.

Shortcomings of Using a 2D Array for a Matrix

- Indexes are offset by 1
- C++ arrays do not support matrix operations such as **add**, **transpose**, **multiply**, etc.
 - Suppose that x and y are 2D arrays. Can't do $x + y$, $x - y$, $x * y$ immediately.
- Need to develop a class **Matrix** for OO features to support all matrix operations.
- Problems get worse when a matrix is sparse
 - Talk about this later, but let's first see some sparse matrix examples

Example – Diagonal Matrix

- An $n \times n$ matrix in which all nonzero terms are on the diagonal


$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix}$$

Example – Diagonal Matrix

- $x(i,j)$ is on diagonal iff $i = j$
- Number of diagonal elements in an $n \times n$ matrix is n
- Non-diagonal elements are zero
- Store diagonal only v.s. n^2 for the whole

Example – Lower Triangular Matrix

- An $n \times n$ matrix in which all nonzero terms are either on or below the diagonal.

1 0 0 0

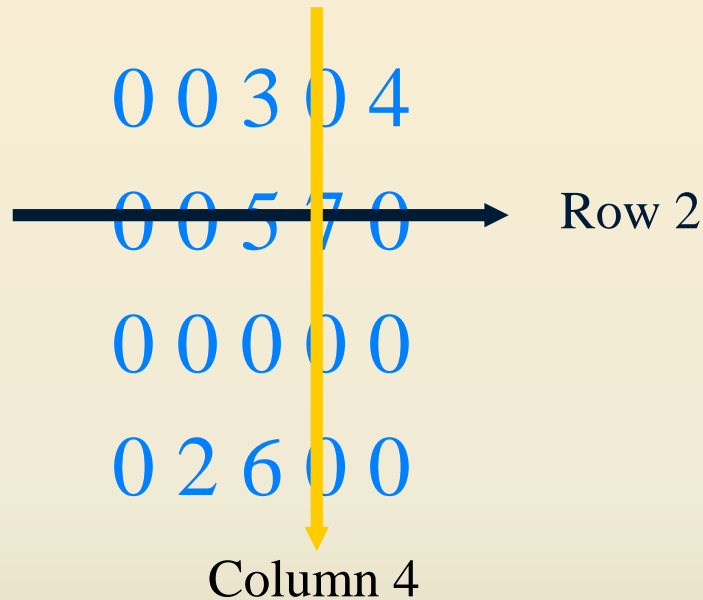
2 3 0 0

4 5 6 0

7 8 9 10

- $x(i,j)$ is part of lower triangle iff $i \geq j$.
- Number of elements in lower triangle is $1 + 2 + \dots + n = n(n+1)/2$.
- Store only the lower triangle

Example – Sparse Matrix



- 4 x 5 matrix
- 20 elements
- 6 nonzero elements

Sparse Matrices

- This example is a sparse matrix
- Sparse matrix →
 $(\text{\#nonzero elements})/(\text{\# all elements})$ is small.
- Why sparse matrices?
 - Because they are frequently seen in large scale computations
 - Regular matrices are easier to deal with
 - But for various algorithms to work on, sparse matrices usually require special data structures

Sparse Matrices

- Examples:
 - Diagonal matrix
 - Only elements along diagonal may be nonzero
 - $n \times n$ matrix \rightarrow ratio is $n/n^2 = 1/n$
 - Lower triangular matrix
 - Only elements on or below diagonal may be nonzero
 - Ratio is ~ 0.5
- These are **structured** sparse matrices
 - Nonzero elements are in a well-defined portion of the matrix.

Sparse Matrices

- A general $m \times n$ matrix consists of m rows and n columns of numbers.
 - Space complexity is $O(mn)$.
- Natural to store a matrix in a two dimensional array, but...
 - it wastes a lot of memory — most of 0's don't need to be stored.
 - Computational cost could be huge — most of computations involving 0's could be completely avoided.
 - **So** we'll study special data structures for sparse matrices to make computation and storage requirement more efficient.
 - Before this, let's talk more about real-world examples for sparse matrix.

Unstructured Sparse Matrices – Another Example

- Biological molecular interactions:
 - Biological molecules are numbered from 1 through n
 - $\text{Matrix}(i,j) = 1 \rightarrow$ molecules i and j interact
 - $\text{Matrix}(i,j) = 0 \rightarrow$ molecules i and j don't interact
 - Typically, n is 1000~10000, so lots of 0's are in the matrix because a molecule may interact with only a few other molecules
 - Important problem in Biology and Medicine, such as drug discovery

Unstructured Sparse Matrices – More Examples

- Web page matrix
 - web pages are numbered 1 through n
 - $\text{web}(i, j)$ = number of links from page i to page j
- Web analysis
 - authority page ... page that has many links to it
 - Other pages may have only few links
 - See IEEE Computer, August 1999, p. 60, for a paper that describes Web analysis based on such a matrix. Operations such as matrix transposition and multiplication are used.


Web Page Matrix

- Assume $n = 1$ billion (billion = 10^9) webpages
- $n \times n$ array of ints $\Rightarrow 4 * 10^{18}$ bytes ($4 * 10^9$ GB) (if an integer uses 4 bytes)
- Each page links to 10 (say) other pages on average
- On average there are 10 nonzero entries per row
- Space needed for nonzero elements is only approximately 10 billion x 4 bytes = 40 billion bytes (40 GB)

Sparse Matrix Example in Textbook's Fig. 2.2

15	0	0	22	0	-15
0	11	3	0	0	0
0	0	0	-6	0	0
0	0	0	0	0	0
91	0	0	0	0	0
0	0	28	0	0	0

Sparse Matrix Representation

							row	col	value	
15	0	0	22	0	-15		a[0]	0	0	15
0	11	3	0	0	0		a[1]	0	3	22
0	0	0	-6	0	0		a[2]	0	5	-15
0	0	0	0	0	0		a[3]	1	1	11
0	0	0	0	0	0		a[4]	1	2	3
91	0	0	0	0	0		a[5]	2	3	-6
0	0	28	0	0	0		a[6]	4	0	91
						a[7]	5	2	28	

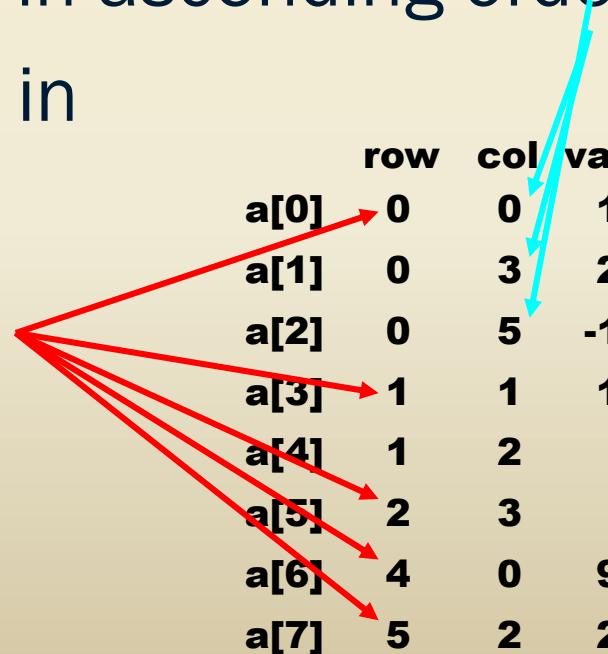
Need to store $6*6 = 36$ elements

Only need to store 8 elements now

Sparse Matrix Representation

- Ideally, use triplet <row, column, value>
- Store triplets row by row
- For all triplets within a row, their column indices are in ascending order
- Rows are in

ascending
order, too



	row	col	value
a[0]	0	0	15
a[1]	0	3	22
a[2]	0	5	-15
a[3]	1	1	11
a[4]	1	2	3
a[5]	2	3	-6
a[6]	4	0	91
a[7]	5	2	28

Sparse Matrix: Abstract Data Type

class SparseMatrix

// objects: A set of triplets, **<row, column, value>**, where row and
//column are integers and form a unique combinations

Public: SparseMatrix(int r, int c, int t);

// The **constructor** function creates a SparseMatrix with **r** rows, **c**
// columns, and a capacity of **t** nonzero terms

SparseMatrix Transpose();

// Returns the SparseMatrix obtained by interchanging the row
// and column value of every triplet in *this

Sparse Matrix Abstract Data Type

SparseMatrix Add(SparseMatrix b);

// If the dimensions of a (*this) and b are the same, then
// the matrix produced by adding corresponding items,
// i.e., those with identical row and column values is returned
// **else** error.

SparseMatrix Multiply(SparseMatrix b);

// If number of columns in a (*this) equals number of rows
// in b, then the matrix d produced by multiplying a by b
// according to $d[i][j] = \sum (a[i][k] * b[k][j])$, where
// $d[i][j]$ is the (i, j)-th element, is returned
// **else** error.

Matrix Transpose – Some More Codes

- `class MatrixTerm {`
`friend class SparseMatrix`
`private:`
 `int row, col, value;`
`};`

- **In class SparseMatrix, define:**
`private:`
 `int rows, cols, terms, capacity;`
 `MatrixTerm *smArray;`

(where terms: # non-zeros; capacity: size of smArray)

- We need smArray to represent a sparse matrix in a compact form using sets of the triplets.
- How to put these altogether is your homework. Refer to the homework assignment.

Matrix Transpose

row col value

a[0] 0 0 15

a[1] 0 3 22

a[2] 0 5 -15

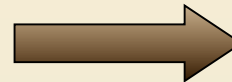
a[3] 1 1 11

a[4] 1 2 3

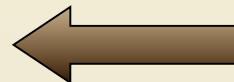
a[5] 2 3 -6

a[6] 4 0 91

a[7] 5 2 28



transpose



row col value

b[0] 0 0 15

b[1] 0 4 91

b[2] 1 1 11

b[3] 2 1 3

b[4] 2 5 28

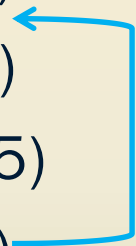
b[5] 3 0 22

b[6] 3 2 -6

b[7] 5 0 -15

a and b are
smArray in the
previous page

Matrix Transpose

- Naturally, one will do
 - For each **row** i ,
take element $\langle i, j, \text{value} \rangle$ and store it in element $\langle j, i, \text{value} \rangle$ of the transpose.
- E.g.
 - $(0, 0, 15) \implies (0, 0, 15)$
 - $(0, 3, 22) \implies (3, 0, 22)$
 - $(0, 5, -15) \implies (5, 0, -15)$
 - $(1, 1, 11) \implies (1, 1, 11)$
- But the method above has some problems because
 $(1, 1, 11)$ must be moved above $(3, 0, 22)$ (see p. 43)

Matrix Transpose

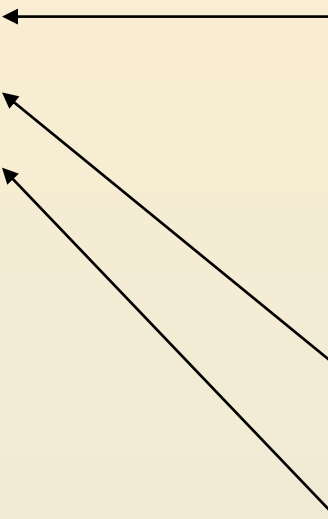
- So using rows to transpose may encounter some unforeseen problems like above.
- So the textbook suggests the following new method:
 - Instead of working on rows, just work on columns to make sure after transposing the rows (the columns in the original matrix) are in ascending order.
 - I.e.: “For all elements in **column** j , place element $\langle i, j, \text{value} \rangle$ in element $\langle j, i, \text{value} \rangle$ ”

Matrix Transpose – The Transpose Method

- I.e., “find all elements in column 0 and store them in row 0, and
 - find all elements in column 1 and store them in row 1, etc.”
 - Since the rows are already in ascending order, this can locate elements in the correct column order
- See example next
- This is the “**Transpose**” method in the textbook

The Transpose Method

	row	col	value		row	col	value	
a[0]	0	0	15	←	b[0]	0	0	15
a[1]	0	3	22		b[1]	0	4	91
a[2]	0	5	-15		b[2]	1	1	11
a[3]					b[3]	2	1	3
a[4]					b[4]	2	5	28
a[5]					b[5]	3	0	22
a[6]					b[6]	3	2	-6
a[7]					b[7]	5	0	-15



- Goal: Transpose b to a (not **transposing a to b**, because the textbook transposes b to a in a later example)
- Iteration 0: scan the array b and process the entries with col=0

The Transpose Method

	row	col	value		row	col	value
a[0]	0	0	15		b[0]	0	15
a[1]	0	3	22		b[1]	0	91
a[2]	0	5	-15		b[2]	1	11
a[3]	1	1	11	←	b[3]	2	3
a[4]	1	2	3	←	b[4]	2	28
a[5]					b[5]	3	22
a[6]					b[6]	3	-6
a[7]					b[7]	5	-15

- Iteration 1: scan the array b and process the entries with col=1
- Iteration 2: scan the array b and process the entries with col=2, ..., etc.

Matrix Transpose – using 2-Dimensional Array Representation

- So for each column, scan “#terms” times
 - Each iteration scans “#terms” times
- Since there are #columns, time complexity = $O(\text{cols} \times \text{terms})$

The Transpose Method – the Code

```
SparseMatrix SparseMatrix::Transpose()  
// Return the transpose of a (*this) {  
    SparseMatrix b(cols, rows, terms); // capacity of b.smArray is terms  
    if (terms > 0)  
    {  
        // nonzero matrix  
        int currentB = 0;  
        for (int c=0; c<cols; c++) // transpose by columns  
            for (int i = 0; i < terms; i++) // find and move terms in column c  
                if (smArray[i].col == c) {  
                    b.smArray[currentB].row=c;  
                    b.smArray[currentB].col=smArray[i].row;  
                    b.smArray[currentB++].value = smArray[i].value;  
                }  
    }  
    // end of if (terms > 0)  
    return b;  
} // end of transpose
```

Time complexity =
 $O(\text{cols} * \text{terms})$

Matrix Transpose – using 2-Dimensional Array Representation

- Traditional way for matrix transpose using 2-dimensional array representation

```
for (int j=0; j< cols; j++)  
    for (int i=0; i< rows; i++)  
        b[j][i]=a[i][j];
```

- Time complexity = $O(\text{columns} \times \text{rows})$

Compare the “Transpose” Method with 2-Dimensional Array Representation

- $O(\text{columns} \times \text{terms})$ vs. $O(\text{columns} \times \text{rows})$

2D array
representation

Transpose in
textbook

- If the matrix is non-sparse,
 - #terms \rightarrow columns \times rows,
 - Time complexity $\rightarrow O(\text{columns}^2 \times \text{rows})$
 - This is even worse than $O(\text{columns} \times \text{rows})$

- So which method is better depends on if the matrix is sparse.

Matrix Transpose – The FastTranspose Method

- **A clever solution – FastTranspose:**
 - Key: Find out which row of the transpose matrix to put the non-zero element of the original matrix.
 - How to do this?
 - Determine # elements in each column of the original matrix → store this as “RowSize”
 - I.e., **how many rows** in the transpose will be needed to store all the elements from the same column of the original matrix
 - For each column, determine its **starting row position** in the transpose → RowStart

Matrix Transpose – The FastTranspose Method

- The key: compute **RowStart** so that we know where non-zero elements should be transposed to the new matrix.
- Thus, this method requires more space than Transpose because it have to store the two variables
 - **RowSize**
 - **RowStart**

Matrix Transpose – The FastTranspose Method

Initialization

	row	col	value		row	col	value
a[0]				b[0]	0	0	15
a[1]				b[1]	0	4	91
a[2]				b[2]	1	1	11
a[3]				b[3]	2	1	3
a[4]				b[4]	2	5	28
a[5]				b[5]	3	0	22
a[6]				b[6]	3	2	-6
a[7]				b[7]	5	0	-15

For col 1, array b contains 2 elements

For col 0, array b contains 3 elements

Index [0] [1] [2] [3] [4] [5]
RowSize = 3 2 1 0 1 1
RowStart = 0 3 5 6 6 7

For col 2, array b contains 1 element

- Step 1: Calculate **RowSize** by scanning array b's col
- Step 2: Calculate: $\text{RowStart}[i] = \text{RowStart}[i-1] + \text{RowSize}[i-1]$, where $\text{RowStart}[0]$ is set to 0.

Matrix Transpose – The FastTranspose Method

- So in the initialization step, first compute
 - RowSize, then compute
 - RowStart

Matrix Transpose – The FastTranspose Method

- Transpose begins

row col value							
a[0]	0	0	15				
a[1]							
a[2]							
a[3]							
a[4]							
a[5]							
a[6]							
a[7]							
				b[0]	0	0	15
				b[1]	0	4	91
				b[2]	1	1	11
				b[3]	2	1	3
				b[4]	2	5	28
				b[5]	3	0	22
				b[6]	3	2	-6
				b[7]	5	0	-15
Index		[0]	[1]	[2]	[3]	[4]	[5]
RowSize	=	3	2	1	0	1	1
RowStart	=	0	3	5	6	6	7

RowStart[0] needs to increase by 1, so the next element from b's col 0 will be placed at the right location of a

RowStart[0]++

Matrix Transpose – The FastTranspose Method



	row	col	value		row	col	value
a[0]	0	0	15		b[0]	0	15
a[1]					b[1]	0	4 91
a[2]					b[2]	1	1 11
a[3]					b[3]	2	1 3
a[4]					b[4]	2	5 28
a[5]					b[5]	3	0 22
a[6]					b[6]	3	2 -6
a[7]					b[7]	5	0 -15

Index	[0]	[1]	[2]	[3]	[4]	[5]
RowSize =	3	2	1	0	1	1
RowStart =	1	3	5	6	6	7

RowStart[0] becomes 1 now

Matrix Transpose – The FastTranspose Method

- Transpose continues

	row	col	value			row	col	value	
a[0]	0	0	15			b[0]	0	0	15
a[1]						b[1]	0	4	91
a[2]						b[2]	1	1	11
a[3]						b[3]	2	1	3
a[4]						b[4]	2	5	28
a[5]						b[5]	3	0	22
a[6]	4	0	91			b[6]	3	2	-6
a[7]						b[7]	5	0	-15

Index	[0]	[1]	[2]	[3]	[4]	[5]
RowSize =	3	2	1	0	1	1
RowStart =	1	3	5	6	6	7

RowStart[4]++

RowStart[4] needs to increase by 1, so the next element from b's col 4 will be placed at a's right location

Matrix Transpose – The FastTranspose Method

- Transpose finishes

	row	col	value			row	col	value
a[0]	0	0	15		b[0]	0	0	15
a[1]	0	3	22		b[1]	0	4	91
a[2]	0	5	-15		b[2]	1	1	11
a[3]	1	1	11	←	b[3]	2	1	3
a[4]	1	2	3		b[4]	2	5	28
a[5]	2	3	-6		b[5]	3	0	22
a[6]	4	0	91		b[6]	3	2	-6
a[7]	5	2	28		b[7]	5	0	-15

(not easy to grasp, so spend time trying to understand the whole process)

Summary – The FastTranspose Method

- So for FastTranspose
 - In initialization we compute
 - RowSize, then compute
 - RowStart
- Then do the transpose

FastTranspose – the Code

```
SparseMatrix SparseMatrix::FastTranspose( )  
{// Return the transpose of *this in O(terms + cols) time  
  SparseMatrix b(cols , rows , terms);  
  if (terms > 0)  
  {// non-zero matrix  
    int *rowSize = new int[cols];  
    int *rowStart = new int[cols];  
    // compute rowSize[i] = number of terms in row i  
    ...  
    for (int i = 0 ; i < terms ; i ++ ) rowSize[smArray[i].col]++;  
  
    // rowStart[i] = starting position of row i  
    rowStart[0] = 0;  
    for (int i = 1 ; i < cols ; i ++ ) rowStart[i] = rowStart[i-1] + rowSize[i-1];
```

O(terms)

O(columns)

FastTranspose – the Code (cont.)

```
// start the transpose process
for (int i = 0 ; i < terms ; i++)
{
    // copy from *this to b
    int j = rowStart[smArray[i].col];
    b.smArray[j].row = smArray[i].col;
    b.smArray[j].col = smArray[i].row;
    b.smArray[j].value = smArray[i].value;
    rowStart[smArray[i].col]++;
} // end of for
delete [] rowSize;
delete [] rowStart;
} // end of if
return b;
}
```

$O(\text{terms})$

Overall complexity =
 $O(\text{columns} + \text{terms})$

Representation of Multi-dimensional Arrays by 1-dimensional Arrays

- Multidimensional arrays are usually implemented by 1 dimensional array via either **row major order** or **column major order**.

Array Applications — String

- Usually string is represented as a character array.
- General string operations include comparison, string concatenation, copy, insertion, string matching, printing, etc.

H	e	l	l	o		W	o	r	l	d	\0
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Array Applications — More

- Examples:
 - Stacks
 - Queues
 - String matching
 - Binary string representations for Genetic Algorithms
 - ...
- All these are important subjects in mid/high-level algorithm course, but we shall cover only some of them in this course.

Homework Assignment

- Go to Elearning to get the homework assignment
- Note the deadline.