

Arrays

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Basic Data Structures

■ Homogeneous/Heterogeneous array

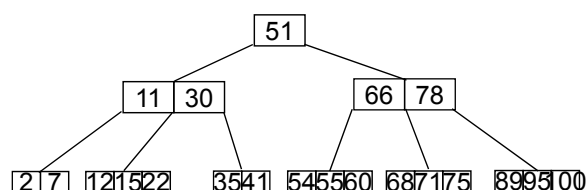


■ List

■ Stack

■ Queue

■ Tree



Definition of Array

- A data structure representing a linear list
 - Elements could be the same or different data types
- Examples:
 - Days of the week: {Sunday, Monday, ..., Saturday}
 - Deck of cards: {Ace, 2, 3, ..., King}
 - Phone Book: {(James, 31212), (Claire, 31213), ..., (Tony, #99999)}

Common Operations

- ADT $\text{array}[n] = \{a_0, a_1, \dots, a_{n-1}\}$
 - Find the length, n , of the array.
 - Read the array from left to right (or reverse).
 - Retrieve the i^{th} element, $0 \leq i < n$.
 - Store a new element into i^{th} position, $0 \leq i < n$.
 - Insert / delete the element at position i , $0 \leq i < n$.

Array Representations

■ Sequential mapping

- Element a_i is stored in the location i of the array
- The most commonly used
- Efficient random access

■ Non sequential mapping

- Carry out insertion and deletion efficiently
- E.g. Linked Lists in chapter 4

Building an ADT for Polynomials

$$p(x) = a_0x^{e_0} + a_1x^{e_1} + \dots + a_nx^{e_n} = \sum_{i=0}^n a_ix^{e_i}$$

- Each $a_ix^{e_i}$ is called a term with coefficient a_i
- The **degree** of $p(x)$ is the largest exponent from among the non-zero terms
- Example:
- Ex. $p(x) = x^5 + 4x^3 + 2x^2 + 1$
 - Has 4 terms with coefficients 1, 4, 2 and 1
 - The degree of $p(x)$ is 5
- Array representation
 - Store (a_i, e_i) as $(\text{array}[n-i], i)$ pair and n is the degree

Polynomial Operations

$$a(x) = \sum a_i x^i \text{ and } b(x) = \sum b_i x^i$$

■ Polynomial addition

- $a(x) + b(x) = \sum (a_i + b_i) x^i$

■ Polynomial multiplication

- $a(x) \cdot b(x) = \sum (a_i x^i \cdot \sum (b_j x^j))$

■ Examples

- $a(x) = x^5 + 4x^3 + 2x^2 + 1$ (degree = 5)
- $b(x) = 3x^6 + 4x^3 + x$ (degree = 6)
- $a(x) + b(x) = 3x^6 + x^5 + 8x^3 + 2x^2 + x + 1$ (degree = 6)

Polynomial : ADT

```
class Polynomial {
public:
    // Construct p(x) = 0
    Polynomial(void);
    // Destructor
    ~Polynomial(void);
    // Return the sum of *this and poly
    Polynomial Add(Polynomial poly);
    // Return multiplication of *this and poly
    Polynomial Mult(Polynomial poly);
    // Return the evaluation result
    float Eval(float f);
private:
    // Array representation
    ...
};
```

Polynomial: 1st Representation

```
// in class Polynomial
public:
    // degree  $\leq$  MaxDegree
    int degree;
    // coefficient array
    float coef[MaxDegree+1];
```

Usage:

```
Polynomial a;
a.degree = n;
a.coef[i] = an-i
```

- Coefficients are stored in order of decreasing exponents
- Advantages:
- Disadvantages:
 -

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Polynomial: 2nd Representation

```
class Term {
    friend Polynomial;
    float coef;
    int exp;
};
```

```
// in class Polynomial
private:
    // array of nonzero terms
    Term* termArray;
    int capacity; // size of termArray
    int terms; // number of nonzero terms
```

- Store only nonzero terms
 - Each nonzero term holds an exponent and its corresponding coefficient
- Advantages:
 -
- Disadvantages:
 -

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Polynomial Addition: Codes

```
Polynomial Polynomial::Add(Polynomial b)
{ // Return sum of polynomial *this and b
  Polynomial c;
  int aPos = 0, bPos = 0;
  while((aPos < terms) && (bPos < b.terms))
    if(termArray[aPos].exp == b.termArray[bPos].exp){
      float t = termArray[aPos].coef + b.termArray[bPos].coef;
      If(t) c.NewTerm(t, termArray[aPos].exp);
      aPos++; bPos++;}
    else if(termArray[aPos].exp < b.termArray[bPos].exp){
      c.NewTerm(b.termArray[bPos].coef, b.termArray[bPos].exp);
      bPos++;}
    else{
      c.NewTerm(termArray[aPos].coef, termArray[aPos].exp);
      aPos++;}
  // add in remaining terms of *this
  for(; aPos < terms; aPos++)
    c.NewTerm(termArray[aPos].coef, termArray[aPos].exp);
  // add in remaining terms of b
  for(; bPos < b.terms; bPos++)
    c.NewTerm(b.termArray[bPos].coef, b.termArray[bPos].exp);
  return c;}
```

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Time Complexity of Analysis

- Inside the while loop: every statement has $O(1)$ time
- How many times the “while loop” is executed in the **worst case** ?
 - Let $a(x)$ have m terms, and $b(x)$ have n terms.
 - In each iteration, we access **next element** in $a(x)$ or $b(x)$, or **both**.
 - Worst case: $m + n$.
eg. It happens when
 $A(x) = 7x^5 + x^3 + x$; $B(x) = x^6 + 2x^4 + 6x^2 + 3$
Access remaining terms in $A(x)$: $O(m)$
Access remaining terms in $B(x)$: $O(n)$
- Hence, total running time =

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Matrix

- A matrix $A_{m \times n}$ (read A is a m by n matrix) consists of
 - m rows
 - n columns
- Stored as a two dimensional array, $a[m][n]$
 - element at i^{th} row and j^{th} column could be accessed by $a[i][j]$

| col 0 | col 1 | col 2 | |
|-------|-------|-------|-------|
| -27 | 3 | 4 | row 0 |
| 6 | 82 | -2 | row 1 |
| 109 | -64 | 11 | row 2 |
| 12 | 8 | 9 | row 3 |
| 48 | 27 | 47 | row 4 |

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Matrix Operations

- Transpose
 - $C_{n \times m} = A_{m \times n}^t$
 - $c[i][j] = a[j][i]$
- Addition
 - $C_{m \times n} = A_{m \times n} + B_{m \times n}$
 - $c[i][j] = a[i][j] + b[i][j]$
- Multiplication
 - $C_{m \times p} = A_{m \times n} + B_{n \times p}$
 - $c[i][j] = \sum_{k=0}^{n-1} a[i][k] \times b[k][j]$

For more information, check the videos on the course webpage
Or, click <https://youtu.be/kYB8IZa5AuE>

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Matrix : ADT

```
class Matrix{
public:
    // Construct
    Matrix(int r, int c);
    // Return the transpose of (*this) matrix
    Matrix Transpose(void);
    // Return sum of *this and b
    Matrix Add(Matrix b);
    // Return the multiplication of *this and b
    Matrix Multiply(Matrix b);
private:
    // Array representation
    int **a, rows, cols;
};
```

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Transpose : Codes

```
Matrix Matrix::Transpose(void) {
    Matrix c(cols, rows);
    for (i=0; i<rows; i++)           // O(rows)
        for (j=0; j<cols; j++)       // O(cols)
            c[j][i]=a[i][j];
    return c;
}
```

■ Time complexity

■

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Add: Codes

```
Matrix Matrix::Add(Matrix b){
    Matrix c(rows, cols);
    for (i=0; i<rows; i++)          // O(rows)
        for (j=0; j<cols; j++)      // O(cols)
            c[i][j]=a[i][j]+b[i][j];
    return c;
}
```

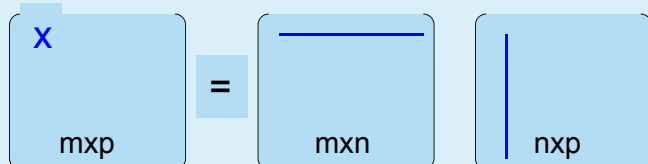
■ Time complexity

■

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Multiply: Codes

```
Matrix Matrix::Multiply(Matrix b){
    Matrix c(rows, b.cols);
    for (i=0; i<rows; i++) {          // O(rows)
        for (j=0; j<b.cols; j++) {    // O(b.cols)
            sum=0;
            for (k=0; k<cols; k++)      // O(cols)
                sum += a[i][k]*b[k][j];
            c[i][j]=sum;
        }
    }
    return c;
}
```



■ Time complexity

■

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Sparse Matrix

a[6][6] =

| | | | | | |
|----|----|----|----|---|-----|
| 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 |

- A matrix has many **zero** elements
 - E.g., a large matrix $A_{5000 \times 5000}$ which has only 100 nonzero elements
- 2D array representation is inefficient
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Sparse Matrix : ADT

```
class SparseMatrix{
public:
    // Construct, t is the capacity of nonzero terms
    SparseMatrix(int r, int c, int t);
    // Return the transpose of (*this) matrix
    SparseMatrix Transpose(void);
    // Return sum of *this and b
    SparseMatrix Add(SparseMatrix b);
    // Return the multiplication of *this and b
    SparseMatrix Multiply(SparseMatrix b);
private:
    // Sparse representation
    int rows, cols, terms, capacity;
    MatrixTerm *smArray;
};
```

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

Trivial Transpose

- $c[i][j] = a[j][i]$

| A | row | col | value | | A ^T | row | col | value |
|------------|-----|-----|-------|----------------|----------------|-----|-----|-------|
| smArray[0] | 0 | 0 | 15 | Transpose → | smArray[0] | 0 | 0 | 15 |
| smArray[1] | 0 | 3 | 22 | | smArray[1] | 3 | 0 | 22 |
| smArray[2] | 0 | 5 | -15 | | smArray[2] | 5 | 0 | -15 |
| smArray[3] | 1 | 1 | 11 | | smArray[3] | 1 | 1 | 11 |
| smArray[4] | 1 | 2 | 3 | | smArray[4] | 2 | 1 | 3 |
| smArray[5] | 2 | 3 | -6 | | smArray[5] | 3 | 2 | -6 |
| smArray[6] | 4 | 0 | 91 | | smArray[6] | 0 | 4 | 91 |
| smArray[7] | 5 | 2 | 28 | | smArray[7] | 2 | 5 | 28 |

- Problem:

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Smart Transpose

- Because the row and column are swapped, we trace the nonzero terms in a **column-major** order.

```
For(all elements in column j)
  Store a(i,j,value) as aT(j,i,value)
```

| A | row | col | value | | A ^T | row | col | value |
|------------|-----|-----|-------|--|----------------|-----|-----|-------|
| smArray[0] | 0 | 0 | 15 | | smArray[0] | 0 | 0 | 15 |
| smArray[1] | 0 | 3 | 22 | | smArray[1] | 0 | 4 | 91 |
| smArray[2] | 0 | 5 | -15 | | smArray[2] | 1 | 1 | 11 |
| smArray[3] | 1 | 1 | 11 | | smArray[3] | 2 | 1 | 3 |
| smArray[4] | 1 | 2 | 3 | | smArray[4] | 2 | 5 | 28 |
| smArray[5] | 2 | 3 | -6 | | smArray[5] | 3 | 0 | 22 |
| smArray[6] | 4 | 0 | 91 | | smArray[6] | 3 | 2 | -6 |
| smArray[7] | 5 | 2 | 28 | | smArray[7] | 5 | 0 | -15 |

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Smart Transpose : Codes

```
SparseMatrix SparseMatrix::Transpose()
{ // Return the transpose of (*this) matrix
  // b.smArray has the same number of nonzero terms
  SparseMatrix b(cols, rows, terms);
  if (terms > 0) // has nonzero terms
  {
    int currentB = 0;
    for(int c=0; c<cols; c++) // O(cols)
      for(int i=0; i<terms; i++) // O(terms)
        if(smArray[i].col == c)
        {
          b.smArray[currentB].row = c;
          b.smArray[currentB].col = smArray[i].row;
          b.smArray[currentB++].value = smArray[i].value;
        }
  }
  return b;
}
```

- Time complexity
- It can be faster!

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Fast Transpose

- We need to examine all terms only once!
- Use additional space to store
 - rowSize[i]: # of nonzero terms in i^{th} row of A^T
 - rowStart[i]: location of nonzero term in i^{th} row of A^T
 - For $i > 0$, $\text{rowStart}[i] = \text{rowStart}[i-1] + \text{rowSize}[i-1]$
- Copy element from A to A^T one by one.
- Time complexity:

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Fast Transpose

| | | | | | |
|----|----|----|----|---|-----|
| 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 |

Count the # of nonzero terms in each row of A

Calculate $rowstart[i] = rowSize[i-1] + rowStart[i-1]$

| A | row | col | value | A ^T | rowSize | rowStart | A ^T | row | col | value |
|------------|-----|-----|-------|----------------|---------|----------|----------------|-----|-----|-------|
| smArray[0] | 0 | 0 | 15 | [0] | 2 | 0 | smArray[0] | | | |
| smArray[1] | 0 | 3 | 22 | [1] | 1 | 2 | smArray[1] | | | |
| smArray[2] | 0 | 5 | -15 | [2] | 2 | 3 | smArray[2] | | | |
| smArray[3] | 1 | 1 | 11 | [3] | 2 | 5 | smArray[3] | | | |
| smArray[4] | 1 | 2 | 3 | [4] | 0 | 7 | smArray[4] | | | |
| smArray[5] | 2 | 3 | -6 | [5] | 1 | 7 | smArray[5] | | | |
| smArray[6] | 4 | 0 | 91 | | | | smArray[6] | | | |
| smArray[7] | 5 | 2 | 28 | | | | smArray[7] | | | |

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Fast Transpose

Copy element from A to A^T one by one

| A | row | col | value | A ^T | rowSize | rowStart | A ^T | row | col | value |
|------------|-----|-----|-------|----------------|---------|----------|----------------|-----|-----|-------|
| smArray[0] | 0 | 0 | 15 | [0] | 2 | 1 | smArray[0] | 0 | 0 | 15 |
| smArray[1] | 0 | 3 | 22 | [1] | 1 | 2 | smArray[1] | | | |
| smArray[2] | 0 | 5 | -15 | [2] | 2 | 3 | smArray[2] | | | |
| smArray[3] | 1 | 1 | 11 | [3] | 2 | 5 | smArray[3] | | | |
| smArray[4] | 1 | 2 | 3 | [4] | 0 | 7 | smArray[4] | | | |
| smArray[5] | 2 | 3 | -6 | [5] | 1 | 7 | smArray[5] | | | |
| smArray[6] | 4 | 0 | 91 | | | | smArray[6] | | | |
| smArray[7] | 5 | 2 | 28 | | | | smArray[7] | | | |

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Fast Transpose : Codes

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

    for(int i=0; i<terms; i++)
    { // copy terms 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]++;} // Increase the start pos by 1
    delete [] rowSize;
    delete [] rowStart;}
  return b;}
```

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Running Time Comparison

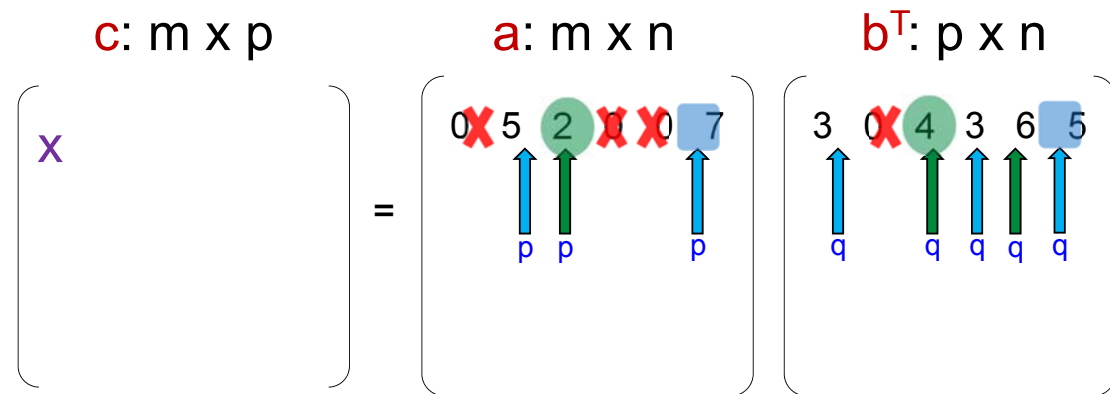
| Trivial Transpose | Smart Transpose | Fast Transpose |
|------------------------------------|-------------------------------------|---------------------------------|
| $O(\text{rows} \cdot \text{cols})$ | $O(\text{cols} \cdot \text{terms})$ | $O(\text{cols} + \text{terms})$ |

- For a dense matrix (terms = rows·cols)
 - Fast equals to trivial: $O(\text{rows} \cdot \text{cols})$
 - Smart is slowest: $O(\text{rows} \cdot \text{cols}^2)$
- For a sparse matrix (terms \ll rows·cols)
 - Fast transpose is faster than trivial and smart ones

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Sparse Matrix Multiplication

- Use approach similar to “**Polynomial Addition**” to compute the X !



Please refer textbook
for codes!

$$x = (2)(4) + (7)(5) = 43$$

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Time Complexity

```

SparseMatrix SparseMatrix::Multiply(SparseMatrix b)
{ // Compute the transpose of b
  SparseMatrix bT = b.FastTranspose(); // O(b.terms + b.cols)

  for ith row in smArray                // O(rows)
    for jth row in bT.smArray            // O(b.cols)
      Perform "Polynomial Addition"      // O(Terms[i] + b.Terms[j])
}

```

- Complexity:
 - $O(\text{rows} \cdot b.\text{cols} \cdot (\text{Term}[i] + b.\text{Terms}[j]))$
 - $\text{rows} \cdot \text{Term}[i] = a.\text{terms}$ and $b.\text{cols} \cdot b.\text{Terms}[j] = b.\text{terms}$
 - $O(\text{rows} \cdot b.\text{terms} + b.\text{cols} \cdot a.\text{terms})$

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