

More on Arrays and Structures

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What We Have Done

- Why Correctness Proof?
- Sequential and Binary Search
- Space Complexity
- Time Complexity
- Asymptotic Notations
- Practical Complexity
- Arrays
- Dense Array versus Sparse Array
- Concrete versus Abstract Data Type
- Reading Assignment:
Some Examples of Space/Time Complexity, Dynamic 1-D Array

Arrays: from Implementation to Abstraction

C Implementation View

(One-dimensional) array is **a block of consecutive memory** that

- holds a list of N elements
- allows users to retrieve the k -th element
- allows users to store to the k -th location

An Abstract View

Abstract (one-dimensional) array

- holds a list of N elements
- allows users to retrieve the k -th element
- allows users to store to the k -th location

different implementations:
different space/time complexity

Dense Array versus Sparse Array

one abstract array, two possible implementations

```
1 int dense[10] = {1, 3, 0, 0, 0, 0, 0, 0, 0, 2};  
2 int sparse[3][2] = {{0, 1}, {1, 3}, {9, 2}};
```

- dense array: store everything (consecutively), needs 10 positions
 - space: $O(N)$ for a length- N array
 - retrieving: $O(1)$
 - storing: $O(1)$
 - creating: $O(1)$
- sparse array: store only non-zero (index, element) pairs, needs 3 pairs
 - space: $O(E)$ for E elements, better than $O(N)$ if E small
 - retrieving: $O(\log E)$ if index ordered (HOW?)
 - storing: ???
 - creating: ???

note: often use **array** to mean dense array only

Concrete Data Type (Sec. 1.4)

array consists of ...

- objects: a set of (*index*, *element*) pairs (== a list of elements)
- actions: retrieve, store, create which sets/gets the objects

- concrete data type: the **actual outcome** of the type
 - object representation + action implementation
 - for actual coding, per-platform optimization, etc.

(dense 1-D) array in C

- object representation: a block of consecutive memory, with a chunk representing each *element* for each *index*
- action implementation: `[·]` for retrieving and storing, `malloc` for creating, etc.

Abstract Data Type (Sec. 1.4)

array consists of ...

- objects: a set of (*index*, *element*) pairs (== a list of elements)
- actions: retrieve, store, create which sets/gets the objects
- abstract data type: the **pseudo essence (contract)** of the type
 - object **specification** + action **specification**
 - for illustration, high-level analysis, etc.

abstract 1-D array

- object specification: (*index*, *element*) pairs with $index \in \{0, \dots, N - 1\}$
- action specification:
 - retrieve(*index*) returns the *element* associated with *index*;
 - store(*index*, *element*) sets *element* to be associated with *index*;
 - create(*N*) creates the objects, etc.

(sometimes with time/space constraints)

will usually look at abstract data type first before going concrete

2-D Array (Subsec. 2.2.2): by 1-D Array



abstract rectangular 2-D array

- object specification: $(index, element)$ pairs with $index \in \{(0, 0), (0, 1), \dots, (N - 1, M - 1)\}$
- action specification:
`retrieve(index)`; `store(index, element)`; `create(N, M)`, etc.

2-D array by 1-D array in C

- object representation: a block of consecutive memory of size $N * M$, with a chunk representing each *element* for each *index*
- action implementation:

2-D Array (Subsec. 2.2.2): by 1-D Array

```
1 #define N (100)
2 #define M (200)
3 int* twodim = (int*)malloc(sizeof(int)*N*M);
4
5 int get(int* arr, int n, int m)
6     { return arr[n*M + m]; }
```


2-D Array: by 1-D Array with Constant Folding

abstract rectangular 2-D array

- object specification: $(index, element)$ pairs with $index \in \{(0, 0), (0, 1), \dots, (N - 1, M - 1)\}$
- action specification:
retrieve($index$); store($index, element$); create(N, M), etc.

2-D array by 1-D array with constant folding in C

- object representation: a block of consecutive memory of size $N * M$, with a chunk representing each *element* for each *index*
- action implementation:

2-D Array: by 1-D Array with Constant Folding

```
1 #define N (100)
2 #define M (200)
3 int twodim[N][M];
4
5 int get(int arr[][M], int n, int m)
6     { return arr[n][m];}
```

2-D Array: by Array of Arrays



abstract rectangular 2-D array

- object specification: $(index, element)$ pairs with $index \in \{(0, 0), (0, 1), \dots, (N - 1, M - 1)\}$
- action specification:
retrieve($index$); store($index, element$); create(N, M), etc.

2-D array by array of arrays in C

- object representation: N blocks of consecutive memory of size M
- action implementation:

2-D Array: by Array of Arrays

```
1 #define N (100)
2 #define M (200)
3 int** twodim = (int**)malloc(sizeof(int*)*N);
4 for(int n=0;n<N;n++)
5     twodim[n] = (int*)malloc(sizeof(int)*M);
6 int get(int** arr, int n, int m)
7     { return arr[n][m];}
```

Comparison of Three Implementations

```
1 int* twodim = (int*)malloc(sizeof(int)*N*M);
2 int twodim[N][M];
3 int** twodim = (int**)malloc(sizeof(int*)*N);
```

	1	2	3
space	$N * M$ integers	$N * M$ int.	$N * M$ int. + N pointers
type	<code>int*</code>	<code>int*[M]</code>	<code>int**</code>
create	constant	constant	$O(N)$
retrieve	arithmetic+dereference	arith.+deref.	deref.+deref.

method 2 for static allocating; method 1 or 3 for dynamic allocating (your choice)

A Tale between Two Programs

```
1 int rowsum(){
2   int i, j;
3   int res = 0;
4   for(i=0;i<MAXROW;i++)
5     for(j=0;j<MAXCOL;j++)
6       res += array[i][j];
7 }
```

0	1	2
3	4	5

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0	1	2	3	4	5
---	---	---	---	---	---

9

```
1 int colsum(){
2   int i, j;
3   int res = 0;
4   for(j=0;j<MAXCOL;j++)
5     for(i=0;i<MAXROW;i++)
6       res += array[i][j];
7 }
```

0	2	4
1	3	5

1

0	2	4	1	3	5
---	---	---	---	---	---

Reading Assignment

be sure to go ask the TAs or me if you are still confused

Polynomials (Sec. 2.4)

```
1 typedef struct {
2     int degree;
3     double* coef;
4 } densepoly;
5 typedef struct {
6     int nTerms;
7     int* expo; /* expo[i] ordered */
8     double* coef;
9 } sparsepoly;
```

- `densepoly` versus `sparsepoly`: like dense array versus sparse
- a simple polynomial adding algorithm
 - allocate the resulting poly
 - fill in the values
- trivial for `densepoly` (HW1), slightly harder for `sparsepoly`

Adding Sparse Polynomials

```
1 typedef struct {
2     int nTerms;
3     int* expo; /* expo[i] ordered */
4     double* coef;
5 } sparsepoly;
```

add $x^{100} + 2x^3 + 3$ and $4x^4 + 5x^3 + 6x + 7$

① → ③ → ② → ③ → ④ →

- allocate the resulting poly

x^{100} x^4 x 1

- fill in the values

Reading Assignment

be sure to go ask the TAs or me if you are still confused

Sparse Matrix (Sec. 2.5)

$$\begin{bmatrix} 15 & 0 & 0 & 22 & 0 & -15 \\ 0 & 11 & 3 & 0 & 0 & 0 \\ 0 & 0 & 0 & -6 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 91 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 28 & 0 & 0 & 0 \end{bmatrix}$$

Specialty

a rectangular 2-D array that contains many common elements (0) that we may not want to repeatedly store

Matrix Abstract Data Type

matrix consists of ...

- objects: a set of (*row*, *column*, *value*) triples where *value* is numerical
 - actions: create, transpose, add, multiply, (retrieve, store)
-
- dense implementation: as 2D dense arrays
 - array of array implementation:
 - “(dense 1D) of (sparse 1D)”
 - “(sparse 1D) of (sparse 1D)”
 - ordered triples implementation: our next topic

Ordered Triples Implementation

$$\begin{bmatrix} 15 & 0 & 0 & 22 & 0 & -15 \\ 0 & 11 & 3 & 0 & 0 & 0 \\ 0 & 0 & 0 & -6 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 91 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 28 & 0 & 0 & 0 \end{bmatrix}$$

Ordered(-by-row-then-by-col) Triples

6	6	8
<i>row</i>	<i>col</i>	<i>value</i>
0	0	15
0	3	22
0	5	-15
1	1	11
1	2	3
2	3	-6
4	0	91
5	2	28

• space complexity ?

$O(E)$

• time complexity for retrieve?

$O(\log E)$

simple exercise: compare to unordered triple implementation

Transposing A Sparse Matrix

$$\begin{bmatrix} 15 & 0 & 0 & 22 & 0 & -15 \\ 0 & 11 & 3 & 0 & 0 & 0 \\ 0 & 0 & 0 & -6 & 0 & 0 \end{bmatrix}^T = \begin{bmatrix} 15 & 0 & 0 \\ 0 & 11 & 0 \\ 0 & 3 & 0 \\ 22 & 0 & -6 \\ 0 & 0 & 0 \\ -15 & 0 & 0 \end{bmatrix}$$

Ordered(-by-row-then-by-col) Triples

6 3	3 6	6		6	3	6	
row	col	value					
0	0	15		0	0	15)
0	3	22		1	1	11)
0	5	-15		2	1	3)
1	1	11		3	0	22)
1	2	3		3	2	-6)
2	3	-6		5	0	-15)

Transpose by Scanning Each *col* (Prog. 2.8)

```
set up a sparse matrix res of attributes (col, row, elements)
for j ← 0 to col - 1 do
  for e ← 0 to elements do
    if e.col == j
      append (e.col, e.row, e.value) to res
    end if
  end for
end for
```

- space complexity: $\Theta(\textit{elements})$ for *res*, constant for others
- time complexity: $\Theta(\textit{col} * \textit{elements})$ (so $O(\textit{col} * \textit{elements})$)

Netflix competition:

row = 17700, *col* = 480189, *elements* = 100480507
col * *elements* $\approx 5 \cdot 10^{13}$ (> 5 hr on 2.8 GHz CPU)

Transpose by Scanning Once (Prog. 2.9) I

- to save time on transposing, we want to scan only once

set up a sparse matrix *res* of attributes (*col*, *row*, *elements*)

```
for  $j \leftarrow 0$  to  $col - 1$  do
```

```
  for  $e \leftarrow 0$  to  $elements$  do
```

```
    if  $e.col == j$ 
```

```
      append ( $e.col$ ,  $e.row$ ,  $e.value$ ) to the ( $e.col$ )-th pile
```

```
    end if
```

```
  end for
```

```
end for
```


Transpose by Scanning Once (Prog. 2.9) II

- where's the j -th pile? **pre-compute pile size and starting locations**

set up two arrays *pilesizes* and *pilestarts*, each of length *col*

for $e \leftarrow 0$ to *elements* **do**

if $e.col == j$

$pilesizes[j] \leftarrow pilesizes[j] + 1$

end if

end for

for $j \leftarrow 0$ to $col - 1$ **do**

$pilestarts[j] \leftarrow pilestarts[j - 1] + pilesizes[j]$

end for

Transpose by Scanning Once (Prog. 2.9) III

- space complexity: $\Theta(\text{elements})$ for *res*, $\Theta(\text{col})$ for the helping arrays
- time complexity: $\Theta(\cancel{\text{col}})$ for pre-computing, $\Theta(\text{elements})$ for scanning
ele + col

Netflix competition:

row = 17700, *col* = 480189, *elements* = 100480507
 $\text{col} * 2 + \text{elements} * 10^8$ (0.04 * *constant* sec. on 2.8 GHz CPU)

Recap: Trade-off

Concrete Implementation of the SparseMatrix Data Structure

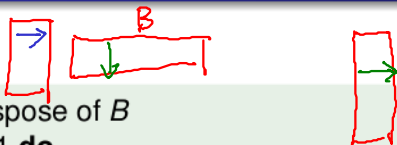
- unordered triples: simpler transpose $O(\text{elements})$, time-consuming retrieval $O(\text{elements})$
- ordered triples: harder transpose $O(\text{col} + \text{elements})$, efficient retrieval $O(\log \text{row} * \log \text{col})$

The Transpose Algorithm

- scanning each column: smaller space $O(\text{elements})$, time-consuming algorithm $O(\text{col} * \text{elements})$
- scanning once: bigger space $O(\text{col} + \text{elements})$, efficient algorithm $O(\text{col} + \text{elements})$

Good Programmer (a.k.a. you):
understand the trade-off clearly and make wise choices!

Matrix Multiplication (Subsec. 2.5.4)



do a temporary transpose of B

for $i \leftarrow 0$ to $rowA - 1$ **do**

for $j \leftarrow 0$ to $colB - 1$ **do**

 compute $C[i, j]$ by multiplying $A[i, :]$ and $B[:, j]^T$

end for

end for

- multiplying $A[i, :]$ and $B[:, j]^T$: similar to (sparse) polynomial adding —keep it in your toolbox
- time complexity:
 - each multiplying takes $O(\#A_i + \#B_j)$
 - total (careful counting): $O(colB * \#A + rowA * \#B)$

Multi-Dim Array

Reading Assignment

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