CSD101: Introduction to computing and programming (ICP)

Multi-dimensional arrays

- We used arrays to represent sequences. But arrays are a very general way to store multi-dimensional data. In particular 2D arrays are very useful in practice.
- 2D or two dimensional data is so common in applications that we have full packages to support them - for example spreadsheets.
- Just as 1D arrays represented sequences or vectors 2D arrays represent mathematical objects called matrices.
- A matrix is a mathematical object that can be visualized as a 2D table of values - typically floating point numbers - and is naturally represented by a 2D array.

An m × n matrix (or array) has m rows and n columns. An example 3 × 4 matrix A is shown below:

$$A = \begin{bmatrix} 3.1 & 2.8 & 7.5 & 4.2 \\ 2.2 & 1.7 & 4.8 & 5.9 \\ 9.0 & 8.4 & 6.3 & 5.5 \end{bmatrix}$$

The corresponding array representation in C is: float a[3][4]={{3.1,2.8,7.5,4.2},{2.2,1.7,4.8,5.9},{9.0,8.4,6.3,5.4}

Example of a matrix and its array representation II

- Elements in matrices and arrays are referred to by row, column indices. For example, the value 4.8 can be accessed by a[1][2] remember indices start at 0 in C. The matrix notation will normally write it as a subscript A_{2,3} matrix index count starts from 1 (by convention). Matrices are usually denoted by capital letters.
- Arrays are stored in row major form (i.e. row-wise). So, in passing arguments only the first argument can remain unspecified. The number of columns has to be specified.

Matrix operations I

- As for other math objects we can define operations on matrices. Let A_{m×n}, B_{m×n}, C_{n×p} be three matrices with the dimensions as shown and let α ∈ ℝ be a scalar (i.e. real number). Each element in a matrix is denoted by the corresponding small case letter. So, a_{i,j} stands for the element in location (i, j) in matrix A.
- Multiplication by a scalar.
 α × A = [αa_{i,j}], for 1 ≤ i ≤ m, 1 ≤ j ≤ n. That is every element of A is multiplied by α.
- Addition, subtraction.

 $A \pm B = [a_{i,j} \pm b_{i,j}]$, for $1 \le i \le m$, $1 \le j \le n$. The corresponding elements are added or subtracted to get the result. Note *A*, *B* must be compatible (i.e. have the same dimensions).

Multiplication. C_{m×p} = A_{m×n} × B_{n×p}. The element c_{i,j} = ∑_{k=1}ⁿ (a_{i,k} × b_{k,j}). The product element is obtained by multiplying the ith row of A with the jth column of B and adding the corresponding products. So, the multiplicands A and B must be compatible for multiplication to be possible - that is - number of columns in A must equal number of rows in B.

Matrix operations III: transpose I

Transpose of an m×n matrix A is the n×m matrix A^T which has the rows and columns of A inter changed. For example, if A is:

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix}$$

then A^{T} is given by (the element references are to A):

$$A^{T} = \begin{bmatrix} a_{11} & a_{21} \\ a_{12} & a_{22} \\ a_{13} & a_{33} \end{bmatrix}$$

Determinant of a matrix I

- The determinant of an n × n square matrix A, denoted det(A) is a scalar value computed from the entries of the matrix. The determinant describes some properties of the linear transformation represented by the matrix A.
- The determinant of A can be calculated by an inductive calculation as follows called the Laplace expansion.
 - If A is a 1 × 1 matrix that is just a value a then det(A) = a.
 If A is a 2D matrix

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

then $det(A) = a \times d - b \times c$.

Determinant of a matrix II

- 3 If A is an $n \times n$ matrix then: $det(A) = \sum_{j=1..n}^{n} a_{1j}C_{1j}$ where C_{ij} is (ij) cofactor of A. Cofactor $C_{ij} = (-1)^{i+j}M_{ij}$, where M_{ij} is called the (ij) minor of A and is defined as $M_{ij} = det(A')$ where A' is obtained from A by deleting the i^{th} row and j^{th} column from A giving an $(n-1) \times (n-1)$ matrix.
- Using the above inductive definition we can recursively compute det(A).
- An example. Let A be:

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$det(A) = a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13}$$

= $a_{11}M_{11} - a_{12}M_{12} + a_{13}M_{13}$
= $a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31})$
+ $a_{13}(a_{21}a_{32} - a_{22}a_{31})$

Inverse of a matrix

- An **identity** matrix I_n is an $n \times n$ square matrix where the right diagonal has only 1s and all the rest are 0s.
- The inverse of an n×n matrix A written A⁻¹ is defined by the equation: AA⁻¹ = A⁻¹A = I_n provided det(A) ≠ 0. If det(A) = 0 then A is said to be singular and the inverse does not exist.
- A definition of *A*⁻¹ is:

$$A^{-1} = \frac{1}{det(A)} \operatorname{Adjoint}(A)$$

where Adjoint($A = C^{T}$), C is the cofactor matrix of A - that is each element a_{ij} is replaced by C_{ij} to obtain C.

- A calculation using the above formula is extremely inefficient.
- Gaussian elimination can be used to calculate det(A) and A⁻¹ efficiently.

A set of *n* linear equations in *n* unknowns x₁, x₂,..., x_n can be represented using an n × n coefficient matrix A and n × 1 column vector B representing the right hand sides of the equations. For example, the system of n linear equations:

•••

$$a_{11}x_1 + a_{12}x_2 + \ldots + a_{1n}x_n = b_1$$

$$a_{n1}x_1 + a_{n2}x_2 + \ldots + a_{nn}x_n = b_n$$

is represented by the $n \times n$ matrix $A = [a_{ij}], i, j = 1..n$ and the right handside by the $n \times 1$ column vector $B = [b_j], j = 1..n$.

 To solve the system of equations we do two steps i) elimination - converts A to an upper triangular matrix ii) backsubstitution which gives us the values for each unknown variable starting from x_n and moving to x₁.

Row reduction

Row reduction is done using elementary row operations:

- Swap any two rows.
- Multiply a row by a non-zero constant.
- Add a scalar multiple of one row to another.
- After row reduction is complete the orginal coefficient matrix becomes an upper triangular matrix.
- This can be easily used to find the values of the unknowns by the process of back substitution.
- Each elementary row operation is actually equivalent to pre-multiplying matrix A by a specific matrix corresponding to the elementary row operation. That is at the heart of why it works.

Example

Example: $x_1 x_2 x_3$ 2 | -| 8 -3 -| 2 -| -1 - 3 -2 | 2 -3 -2 | 2 -3 $x_1 x_1 , x_3 = x_3 + \frac{a_{31}}{a_{11}} x_1$ $x_1 , x_3 = x_3 + \frac{a_{31}}{a_{11}} x_1$ $x_1 , x_3 = x_3 + \frac{a_{31}}{a_{11}} x_1$ $x_1 , x_3 = x_3 + \frac{a_{31}}{a_{11}} x_1$ 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 8 2 | -| 82 | 5

$$\begin{array}{c} \mathbf{v}_{3} = \mathbf{v}_{3} + \frac{-\alpha_{32}}{\alpha_{21}} \cdot \mathbf{v}_{3} \\ \Rightarrow \\ 2 & 1 - 1 & 8 & \text{back substitute} \\ 0 & \frac{1}{2} & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & -1 & 1 \\ \end{array}$$

$$\begin{array}{c} \mathbf{v}_{3} = -1 \\ \mathbf{v}_{4} = \left(8 + \mathbf{v}_{3} - \mathbf{v}_{4}\right) \frac{1}{2} \end{array}$$