

MATRICES

⇒ A matrix is a rectangular array of numbers/symbols/expressions arranged in rows and columns.

Application of matrices:-

① Data representation

Suppose there is a dataset, with features

	math-score	physics-score	biology-score	
①	55	65	75	→
②	65	60	70	→
③	70	60	72	→
	↓	↓	↓	

columns

Rows

This can be represented as

$$\left[\begin{array}{ccc} 55 & 65 & 75 \\ 65 & 60 & 70 \\ 70 & 60 & 72 \end{array} \right] \xrightarrow{\text{combine}} \left[\begin{array}{ccc} 55 & 65 & 75 \\ 65 & 60 & 70 \\ 70 & 60 & 72 \end{array} \right]$$

A set of vectors matrix

matrix is represented as a set of vectors

② Representing images in computer vision

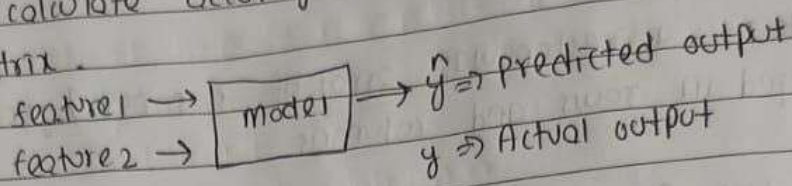
Suppose there is a 3x3 grayscale image with pixel values

$$\text{Image} = \begin{bmatrix} 0 & 128 & 120 \\ 123 & 122 & 115 \\ 0 & 118 & 100 \end{bmatrix} \xrightarrow{\text{can be represented as}} \begin{bmatrix} 0 & 128 & 120 \\ 123 & 122 & 115 \\ 0 & 118 & 100 \end{bmatrix}$$

3x3 grayscale image pixels 3x3 matrix

(iii) calculating accuracy of a model:-

To calculate accuracy of the model, we use confusion matrix.



We can find difference between predicted and actual outputs and create a confusion matrix to find accuracy of model.

$$\text{confusion matrix} = \begin{bmatrix} 50 & 10 \\ 5 & 35 \end{bmatrix}$$

50 \Rightarrow True positive

5 \Rightarrow False positive

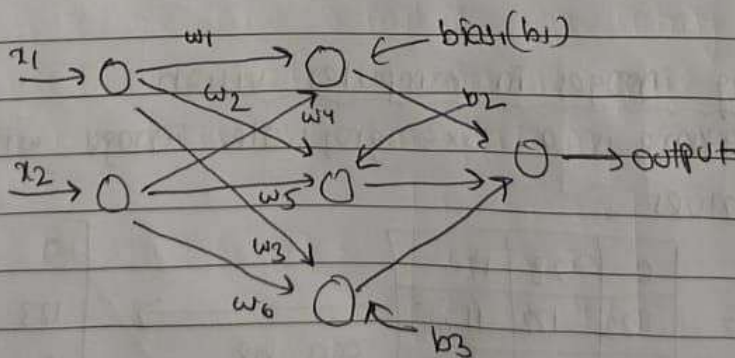
10 \Rightarrow False negative

35 \Rightarrow True negative

$$\text{Accuracy} = \frac{\text{True positive} + \text{True negative}}{\text{False positive} + \text{False negative} + \text{True positive} + \text{True negative}}$$

(iv) Neural networks:-

If forward propagation, we perform matrix operations



$$Z = W^T X + b = \begin{bmatrix} w_1 & w_2 & w_3 \\ w_4 & w_5 & w_6 \end{bmatrix}^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$z = \begin{bmatrix} w_1 & w_4 \\ w_2 & w_5 \\ w_3 & w_6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

So, we use matrix multiplication and matrix addition operations in neural networks

⑤ Linear regression:-

suppose, for a dataset,

no. of study hours	IQ	score
4	100	90
5	90	85
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no. of study hours } Independent
IQ } feature

score } dependent / output
feature

This is an example of regression

$x_1 \Rightarrow$ no. of study hours

$x_2 \Rightarrow$ IQ

The equation will be,

$$y = m_1 x_1 + m_2 x_2 + c$$

It can be presented as

$$\boxed{y = m^T x + c}, \text{ where } m = [m_1 \ m_2]$$

$$x = [x_1 \ x_2]$$

we use matrix multiplication and matrix addition operations in regression.

MATRIX OPERATIONS

⇒ matrix operations help to manipulate and analyze multidimensional data efficiently.

Some matrix operations are:-

- (i) matrix addition and subtraction
- (ii) scalar matrix multiplication
- (iii) matrix multiplication.

⇒ Matrix addition and subtraction:-

We add or subtract corresponding elements of 2 matrices of same dimensions.

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}_{3 \times 3} \quad B = \begin{bmatrix} 4 & 5 & 6 \\ 7 & 8 & 9 \\ 1 & 2 & 3 \end{bmatrix}_{3 \times 3}$$

$$A+B = \begin{bmatrix} 1+4 & 2+5 & 3+6 \\ 4+7 & 5+8 & 6+9 \\ 7+1 & 8+2 & 9+3 \end{bmatrix} = \begin{bmatrix} 5 & 7 & 9 \\ 11 & 13 & 15 \\ 8 & 10 & 12 \end{bmatrix}$$

⇒ Scalar multiplication:-

It involves multiplying every element of a matrix by scalar. Suppose we have a matrix representing product prices and if we want to increase price of all products by 1.05 times, we can perform scalar multiplication.

$$A(\text{original prices}) = \begin{bmatrix} 10 & 20 & 30 \\ 15 & 25 & 35 \\ 20 & 30 & 40 \end{bmatrix}$$

$$\text{Inflation rate} = 1.05$$

$$B(\text{adjusted}) = A \times \text{inflation rate} = 1.05 \begin{bmatrix} 10 & 20 & 30 \\ 15 & 25 & 35 \\ 20 & 30 & 40 \end{bmatrix} = \begin{bmatrix} 10.5 & 21 & 31.5 \\ 15.75 & 26.25 & 36.75 \\ 21 & 31.5 & 42 \end{bmatrix}$$

$$B(\text{adjusted price}) = \text{Inflation price} \cdot A(\text{original price})$$

$$= 1.05 \begin{bmatrix} 10 & 20 & 30 \\ 15 & 25 & 35 \\ 20 & 30 & 40 \end{bmatrix} = \begin{bmatrix} 10.5 & 21 & 31.5 \\ 15.75 & 26.25 & 36.75 \\ 21 & 31.5 & 42 \end{bmatrix}$$

If we want to increase/decrease all elements by a factor, then we can achieve it through ^{scalar} matrix multiplication.

→ matrix multiplication

It involves dot product of rows of the first matrix with columns of the second matrix.

$$A(m \times n) \times B(n \times p) = C(m \times p)$$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \quad B = \begin{bmatrix} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{bmatrix}$$

$$A \times B = A_{(2 \times 3)} \times B_{(3 \times 2)} = C_{(2 \times 2)}$$

$$C = A \times B = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{bmatrix} = \begin{bmatrix} 1 \cdot 7 + 2 \cdot 9 + 3 \cdot 11 & 1 \cdot 8 + 2 \cdot 10 + 3 \cdot 12 \\ 4 \cdot 7 + 5 \cdot 9 + 6 \cdot 11 & 4 \cdot 8 + 5 \cdot 10 + 6 \cdot 12 \end{bmatrix}$$

$$= \begin{bmatrix} 7+18+33 & 8+20+36 \\ 28+45+66 & 32+50+72 \end{bmatrix} = \begin{bmatrix} 58 & 64 \\ 139 & 154 \end{bmatrix}_{(2 \times 2)}$$