### Linear Algebra

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### **Transpose**

Interchange row and columns.

Uses: To make matrix operation compatible.

$$\begin{pmatrix} 3 & 2 & 1 \\ 4 & 7 & 5 \\ 5 & 8 & 9 \end{pmatrix}^{T} = \begin{pmatrix} 3 & 4 & 5 \\ 2 & 7 & 8 \\ 1 & 5 & 9 \end{pmatrix}$$

## Why do we go for Transposition?



2	3		3	4	5		18	26	34
3	4	X	4	6	8	=	25	36	47
5	6						39	56	73

### **Transpose**

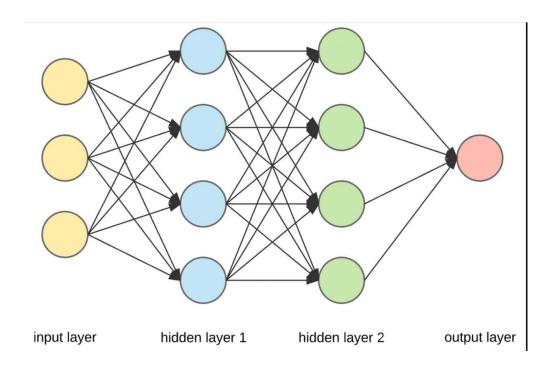


#### **Example**

```
In [1]: 1 import numpy as np
          2 arr =np.array([[2,3,5],[3,4,6]])
          3 brr =np.array([[3,4,5],[4,6,8]])
In [2]:
        1 np.matmul(arr.T,brr)
Out[2]: array([[18, 26, 34],
               [25, 36, 47],
               [39, 56, 73]])
In [3]:
        1 np.matmul(arr,brr.T)
Out[3]: array([[43, 66],
               [55, 84]])
In [ ]:
```

#### **Uses of Transpose**

Evaluation of weight matrix and aggregate value:



## Using Transpose To Develop a Model for Neural Networks

$$A^{[1]} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad W^{[1]} = \begin{bmatrix} w_{11} & w_{21} & w_{31} & w_{41} \\ w_{12} & w_{22} & w_{32} & w_{42} \\ w_{13} & w_{23} & w_{33} & w_{43} \\ w_{13} & w_{23} & w_{33} & w_{43} \end{bmatrix} \quad b^{[1]} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$
(3x1)

Note: A and W cannot be multiplied directly.

$$z^{[L]} = W^{[L]T*}A^{[L-1]} + b^{[L]}$$

# Using Transpose To Develop a Model for Neural Networks : Change **W** to W<sup>T</sup>

$$A^{[1]} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad W^{[1]T} = \begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \\ w_{41} & w_{42} & w_{43} \end{bmatrix} \quad b^{[1]} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

$$z^{[L]} = W^{[L]T*}A^{[L-1]} + b^{[L]}$$

# Using Transpose To Develop a Model for Neural Networks : **Plug it into expression.**

$$W^{[1]T} = \begin{pmatrix} w_{11} & w_{12} & w_{13} & & & & & & & \\ w_{21} & w_{22} & w_{23} & & & & & & \\ w_{31} & w_{32} & w_{33} & & & & & & \\ w_{41} & w_{42} & w_{43} & & & & & & \\ \end{pmatrix} + \begin{pmatrix} b_1 \\ b_2 \\ x_3 \\ b_4 \end{pmatrix}$$

Vector(4x1) Vector(4x3) Vector(3x1) Vector(4x1)

 $z^{[L]} = W^{[L]T*}A^{[L-1]} + b^{[L]}$  (Successfully computed !!)

## Example

```
In [21]:
           1 W = np.random.rand(3,4)
           2 X = np.random.rand(3,1)
             b= np.random.rand(4,1)
             c= np.random.rand(3,1)
           5
           6
             W,X,b
Out[21]:
         (array([[0.899255], 0.52149056, 0.75988558, 0.46614012],
                  [0.24637199, 0.49076638, 0.83288382, 0.51380146],
                  [0.38120651, 0.16910539, 0.93104312, 0.54162068]])
          array([[0.60389517],
                  [0.45503992],
                  [0.60113909]]),
          array([[0.91962447],
                  [0.39762239],
                  [0.68937966],
                  [0.69974358]]))
In [22]:
           1 # Here W is 4X3 weight matrix.
            #Step1 : take transpose of W.
             Y = np.matmul(W.T,X)+b
Out[22]: array([[1.80394745],
                 [1.03752217],
                [2.0869527],
                [1.54063289]])
```

#### Sanity Check (For each layer)

- Number of outputs = Number of neurons.
- 4 = 4
- Number of Inputs = Number of Columns in W<sup>T</sup>.
- 3 = 3
- Number of outputs = Number of bias vector element.
- 4 = 4
- Number of rows in W<sup>T</sup> = Number of neurons.
- 4 = 4

#### Inverse of Matrices

**Invertible Matrices?** 

Simple telltale signs for invertible matrices !!

- Matrix is full rank.
- The determinant of is not zero.
- The matrix has non-zero singular values.

#### Inverse of Matrices

 The inverse of a square matrix, sometimes called a reciprocal matrix, is a matrix such that

$$AA^{-1} = I$$
 (Identity Matrix)

the matrix inverse is

$$\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} \begin{bmatrix} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} & \begin{vmatrix} a_{13} & a_{12} \\ a_{33} & a_{32} \end{vmatrix} & \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \\ \begin{vmatrix} a_{23} & a_{21} \\ a_{33} & a_{31} \end{vmatrix} & \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} & \begin{vmatrix} a_{13} & a_{11} \\ a_{23} & a_{21} \end{vmatrix} \\ \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} & \begin{vmatrix} a_{12} & a_{11} \\ a_{32} & a_{31} \end{vmatrix} & \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{21} & a_{22} \end{vmatrix} \\ a_{21} & a_{22} \end{vmatrix}$$

#### Inverse of Matrices

- $(AB)^{-1} = B^{-1}A^{-1}$
- $AA^{-1} = I$  (Important)
- $A^T = A^{-1}$ ,  $A^TA = I$  (Important :symmetric matrices)
- Inverse of Diagonal Matrices

$$D^{k} = \begin{bmatrix} d_1^{k} & 0 & \cdots & 0 \\ 0 & d_2^{k} & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & d_n^{k} \end{bmatrix}$$

$$D^{k} = \begin{bmatrix} d_{1}^{k} & 0 & \cdots & 0 \\ 0 & d_{2}^{k} & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & d_{n}^{k} \end{bmatrix} \qquad D^{-1} = \begin{bmatrix} 1/d_{1} & 0 & \cdots & 0 \\ 0 & 1/d_{2} & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & 1/d_{n} \end{bmatrix}$$

#### What is a Matrix?

- It is a transformation.
- Inverse will exist if a non singular matrix A has a unique matrix  $A^{-1}$  such that  $AA^{-1} = I$ .
- Relation between A , $A^{-1} \sim x$  ,f(x).
- This concept is exploited in *change of basis transformation.*

#### Inverse of Matrices: Uses

#### Change of Basis:

- vector X has basis u<sub>1</sub>,u<sub>2</sub>...u<sub>n</sub>.
- vector Y has basis v<sub>1</sub>,v<sub>2</sub>...v<sub>m</sub>.
- Transformation A such that,  $X = \sum x_i u_i$ ,  $Y = \sum y_j v_j$

- vector X has basis t<sub>1</sub>,t<sub>2</sub>...t<sub>n</sub>.
- vector Y has basis w<sub>1</sub>,w<sub>2</sub>...w<sub>m</sub>.
- Transformation A' such that,  $X = \sum x'_i t_i$ ,  $Y = \sum y'_i w_i$

$$A'X' = Y'$$

#### Inverse of Matrices: Uses

- Change the each basis component to target component
- $t_i = \sum t_{ji} v_i$ ,  $w_i = \sum w_{ji} u_j$
- Basis for  $B_t = [t_1, t_2, ...]$ ,  $B_w = [w_1, w_2, ...]$  A'X' = Y' $AB_t X' = B_w Y'$

$$B^{-1}_{w}AB_{t} = A'$$

## Example

A vector is rotated through α degrees. Find transformation matrix. (NND Matrix representation page 6.6)

- X be a vector, AX be the transformed vector.
- If we rotate s<sub>2</sub> α degrees anticlockwise , transformation of s2.
- If we rotate s<sub>1</sub> α degrees anticlockwise, transformation of s1.

$$AS_1 = s_1 \cos(\alpha) + s_2 \sin(\alpha)$$

$$AS_2 = s_1 \cos(90 + \alpha) + s_1 \sin(90 + \alpha)$$

$$AS_3 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_1 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_2 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_3 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_4 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

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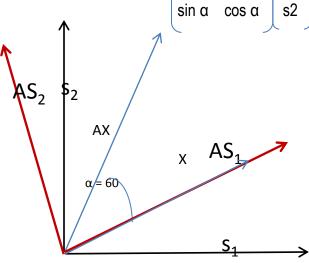
$$S_4 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_5 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_7 = s_1 \cos(90 + \alpha) + s_2 \sin(90 + \alpha)$$

$$S_7 = s_1 \cos(90 + \alpha)$$

Here Matrix is rotation transformation.



#### Inverse of Matrices: Uses

**SVD** 

## Eigen Values and Vectors

## Examples

## Eigen Values and Vectors: Uses

Learning Rates

## Example

## **Practical Examples**