

Dot product between two vectors

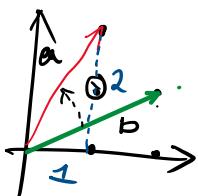
$$a = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, b = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} \Rightarrow a \cdot b = 1*4 + 2*5 + 3*6 = 32$$

- What does the dot product tell us about the vectors?

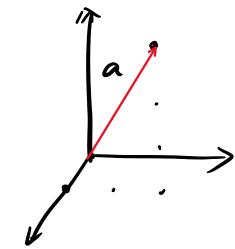
- Geometric interpretation of a vector

$$a = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$b = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$



$$a = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$



- Geometric length of a vector $\|a\|_2 = \sqrt{a_1^2 + a_2^2}$

$$(L_2 \text{ norm of } a) = \sqrt{1^2 + 2^2}$$

- Angle between vectors $-1 \leq \frac{a \cdot b}{\|a\| \|b\|} \leq 1$ (Cauchy-Schwarz inequality)



$$\cos(\angle a, b) = \frac{a \cdot b}{\|a\| \|b\|}$$

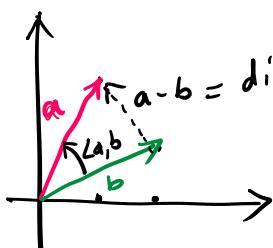
$\cos(0) = 1$
 $\cos(\pi) = -1$
 $\cos(\pi/2) = 0$

$$\boxed{a \cdot b = 0 \quad \angle a, b = \pi/2}$$

$$\boxed{a \cdot b = -ve \quad \angle a, b > \pi/2}$$

$$\boxed{a \cdot b = +ve \quad \angle a, b < \pi/2}$$

- Distance between vectors



How big is the difference vector $= \|a - b\|_2$

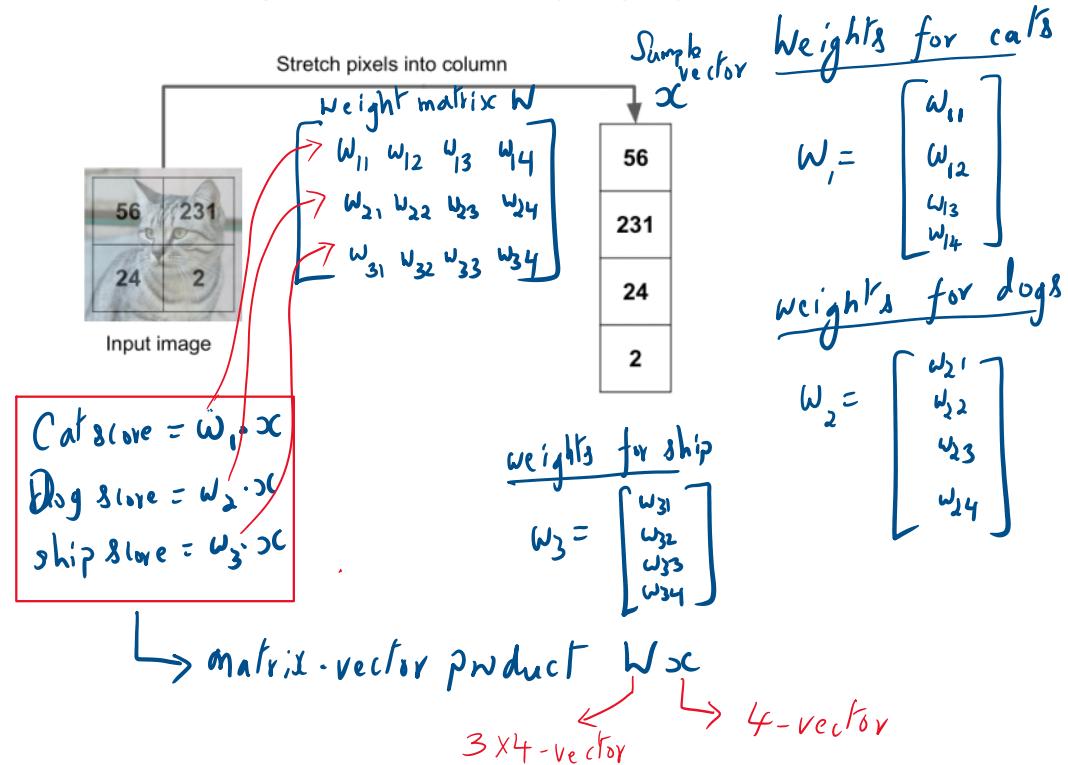
$$= \left\| \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\| = \sqrt{(-1)^2 + 1^2} = \sqrt{2}$$

• An important relationship: $a = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$

$$\|a\|^2 = a \cdot a$$

$$\left(\sqrt{a_1^2 + a_2^2} \right)^2 \Leftrightarrow a_1^2 + a_2^2$$

Example with an image with 4 pixels, and 3 classes (cat/dog/ship)



Why did we not do this?

$$w_1 \cdot x \quad ? \quad \begin{bmatrix} w_{11} \\ w_{12} \\ w_{13} \\ w_{14} \end{bmatrix} \quad \begin{bmatrix} w_{21} \\ w_{22} \\ w_{23} \\ w_{24} \end{bmatrix} \quad \begin{bmatrix} w_{31} \\ w_{32} \\ w_{33} \\ w_{34} \end{bmatrix} \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} w_1 \cdot x \\ w_2 \cdot x \\ w_3 \cdot x \end{bmatrix}$$

$$w_2 \cdot x$$

$$w_3 \cdot x$$

$$W \quad x$$

A matrix-vector product Wx = sequence of dot products
= dot products between the rows of matrix W (seen as vectors) and the vector x

Tensor-Vector Product

Matrix-Vector Product

Tensor-Vector product

E.g. $T = \begin{matrix} & \downarrow \\ (4, 3, 2) & \end{matrix}$ $x = \begin{matrix} & \downarrow \\ (2,) & \end{matrix}$

E.g. $A = \begin{matrix} & \downarrow \\ (3, 4) & \end{matrix}$ $x = \begin{matrix} & \downarrow \\ (4,) & \end{matrix}$

Linear combination of columns of a matrix A

$$\left[\begin{array}{cccc} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{array} \right] \left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_4 \end{array} \right] = \left[\begin{array}{c} a^{(1)} \cdot x \\ a^{(2)} \cdot x \\ a^{(3)} \cdot x \end{array} \right] = \left[\begin{array}{c} 0 \\ 0 \\ 0 \end{array} \right]$$

Annotations:

- 1st row of A
- 2nd row of A
- 3rd row of A
- 1st column of A
- 2nd column of A
- 3rd column of A
- 4th column of A

$$x_1 a_1 + x_2 a_2 + x_3 a_3 + x_4 a_4 = \left[\quad \right]$$

$x_1 a_1 + x_2 a_2 + x_3 a_3 + x_4 a_4 =$ linear combination of columns
of A using the elements of x
as multipliers

Matrix-vector product

E.g. $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, x = \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix} = \begin{bmatrix} 1 \times 7 + 2 \times 8 + 3 \times 9 \\ 4 \times 7 + 5 \times 8 + 6 \times 9 \end{bmatrix}$

Ax is defined as $= \begin{bmatrix} (1\text{st row of } A) \cdot x \\ (2\text{nd row of } A) \cdot x \end{bmatrix} = \begin{bmatrix} a^{(1)} \cdot x \\ a^{(2)} \cdot x \end{bmatrix}$

Tensor-Vector product

E.g. $T =$
 $\begin{matrix} \nwarrow 2 \times 3 \times 2 \\ \text{Timestamps} \end{matrix}$ $\begin{matrix} \searrow \text{patients} \\ \text{features} \end{matrix}$

$\rightarrow \text{Time stamp } 0$
 $\rightarrow \text{Time stamp } 1$

$$T = \begin{bmatrix} \begin{bmatrix} 76 & 120 \\ 74 & 124 \\ 78 & 136 \end{bmatrix} \\ \begin{bmatrix} 98 & 135 \\ 80 & 124 \\ 70 & 120 \end{bmatrix} \end{bmatrix}$$

$$1 \rightarrow c = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} 76 & 120 \\ 74 & 124 \\ 78 & 136 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 76 \\ 74 \\ 78 \end{bmatrix} \rightarrow \begin{bmatrix} 76 & 74 & 78 \\ 78 & 80 & 70 \end{bmatrix}$$

$$\begin{bmatrix} 78 & 136 \\ 80 & 124 \\ 70 & 120 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 78 \\ 80 \\ 70 \end{bmatrix}$$

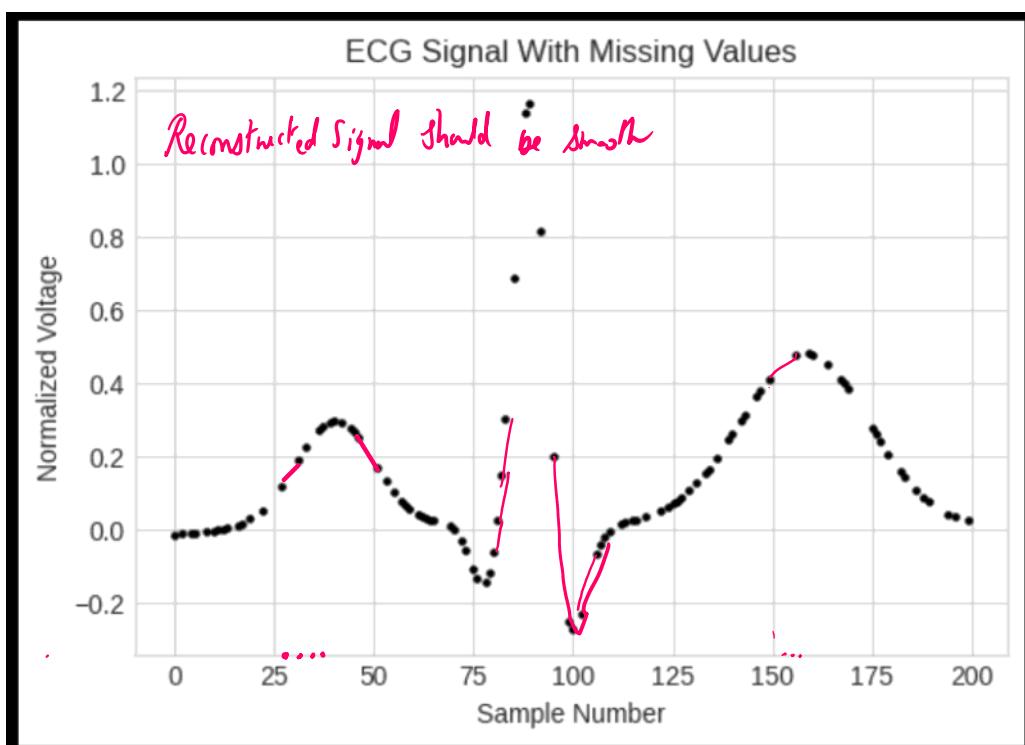
Matrix-Matrix Product

E.g. $A =$

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

1st row of A 1st column of B
2nd row of A 2nd column of B

$$= \begin{bmatrix} a^{(1)} \cdot b_1 & a^{(1)} \cdot b_2 \\ a^{(2)} \cdot b_1 & a^{(2)} \cdot b_2 \end{bmatrix}_{2 \times 2}$$



$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} \cdot x = \begin{bmatrix} x_1 \\ x_2 \\ v_1 \\ x_4 \\ v_2 \\ v_3 \end{bmatrix}$$

missing ECG values

$$\begin{bmatrix} x_1 \\ x_2 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ v_1 \\ x_4 \\ v_2 \\ v_3 \end{bmatrix}.$$

↑ Transposes

Filter matrix for known Ech values

Ech vector with unknown values turned to zeros

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ 0 \\ x_4 \\ 0 \\ 0 \end{bmatrix}$$

↓ Vector of known Ech values

$$\underbrace{\begin{bmatrix} 1 & 0 & 0 & ? \\ 0 & 1 & 0 & ? \\ 0 & 0 & 0 & ? \\ 0 & 0 & 1 & ? \\ 0 & 0 & 0 & ? \\ 0 & 0 & 0 & ? \end{bmatrix}}_{S_1} \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ x_4 \end{bmatrix}}_{x_{\text{known}}} + \underbrace{\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{S_2} \underbrace{\begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}}_{x_{\text{unknown}}} = \begin{bmatrix} x_1 \\ x_2 \\ v_1 \\ x_4 \\ v_2 \\ v_3 \end{bmatrix}$$

↓ Full Ech vector with missing values

$$\Rightarrow x = S_1 x_{\text{known}} + S_2 x_{\text{unknown}}$$

↓ Ech vector

Known part of the Ech vector

Unknown part of the Ech vector

$(y_1 - y_2) - (y_2 - y_3)$

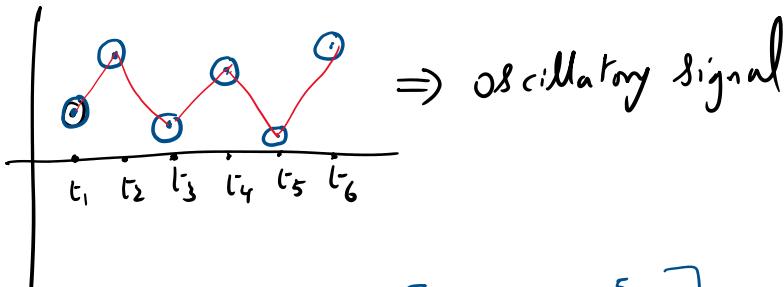
Axi. 10

Aside

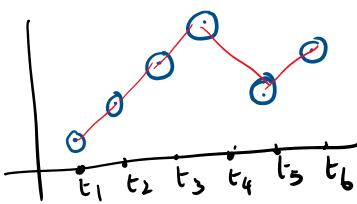
vector vector $(y_1 - y_2) - (y_2 - y_3)$

$$\underbrace{\begin{bmatrix} 1 & -2 & 1 & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & 0 & 0 \\ 0 & 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 0 & 1 & -2 & 1 \end{bmatrix}}_D \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \end{bmatrix} = \begin{bmatrix} y_1 - 2y_2 + y_3 \\ y_2 - 2y_3 + y_4 \\ y_3 - 2y_4 + y_5 \\ y_4 - 2y_5 + y_6 \end{bmatrix}$$

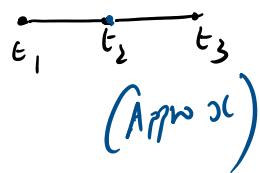
$$y = \begin{bmatrix} 15 \\ 30 \\ 10 \\ 25 \\ 5 \\ 30 \end{bmatrix} \Rightarrow \mathcal{D}y = \begin{bmatrix} 15 - 60 + 10 \\ 30 - 20 + 25 \\ 10 - 50 + 5 \\ 25 - 10 + 30 \end{bmatrix} = \begin{bmatrix} -35 \\ 35 \\ -35 \\ 45 \end{bmatrix}$$



$$y = \begin{bmatrix} 15 \\ 20 \\ 25 \\ 30 \\ 20 \\ 25 \end{bmatrix}, \mathcal{D}y = \begin{bmatrix} 15 - 40 + 25 \\ 20 - 50 + 30 \\ 25 - 60 + 20 \\ 30 - 40 + 25 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ -15 \\ 15 \end{bmatrix}$$



Discrete version of second derivative



$$\frac{y_2 - y_1}{t_2 - t_1} = \frac{\text{change in temperature}}{\text{change in time}} = \frac{^{\circ}\text{C}}{\text{Hr}}$$

Sensitivity of Temperature wrt. Time

$$y_3 - y_2 = \underline{^{\circ}\text{C}}$$

$$\text{Definir} \quad \frac{y_3 - y_2}{t_3 - t_2} = \frac{c}{Hr}$$

$$x = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \quad \boxed{\frac{\|x\|}{\sqrt{n}}} = \sqrt{\frac{1^2 + 1^2 + 1^2 + 1^2}{4}} = \sqrt{\frac{4}{4}} = 1$$

$$x = \begin{bmatrix} -1 \\ -1 \\ -1 \end{bmatrix}, \quad \frac{\|x\|}{\sqrt{n}} = 1$$

Reconstruction goal $\left\{ \begin{array}{l} \text{the reconstructed ECG signal should be} \\ \text{as smooth as possible} \end{array} \right.$

$\text{RMS} \left(D \left(S_1 x_{\text{known}} + S_2 x_{\text{unknown}} \right) \right)$ to be as small as possible

$$\text{small as possible} \Rightarrow \left\| \frac{D(S_1 x_{\text{known}} + S_2 x_{\text{unknown}})}{\sqrt{n}} \right\|_2$$

$\downarrow \text{constant}$

Should be as small as possible

$$\Rightarrow \text{Find the vector } x_{\text{unknown}} \text{ s.t. } \boxed{\left\| D(S_1 x_{\text{known}} + S_2 x_{\text{unknown}}) \right\|_2^2}$$

is as small as possible.

$b = 4\text{-vector}$ $A = 4 \times 3\text{-matrix}$ $x_{\text{unknown}} = 3\text{-vector}$

$$\Rightarrow \boxed{\begin{matrix} DS_1 & x_{\text{known}} \\ \downarrow & \downarrow \\ 4 \times 6 & 6 \times 3 \\ \underbrace{4 \times 3} & 3 \times 1 \end{matrix}} + \boxed{\begin{matrix} DS_2 & x_{\text{unknown}} \\ \downarrow & \downarrow \\ 4 \times 6 & 6 \times 3 \\ \underbrace{4 \times 3} & 3 \times 1 \end{matrix}} \Rightarrow \left\| Ab + Ax \right\|_2^2$$

All linear systems of equations $Ax = b$ are

Solvable

$$\boxed{\begin{bmatrix} -2 & 1 & 0 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 \end{bmatrix}}$$

\diagdown

$$\begin{array}{cccccc} & L & & & & \\ \left[\begin{array}{cccccc} 1 & -2 & 1 & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & 0 & 0 \\ 0 & 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & -2 \end{array} \right] & & & & & D \text{ matrix} \\ \cdot & & & & & \end{array}$$

Standard least squares problem

Find a vector x s.t. $\|Ax - b\|_2^2$ is minimized

↓ ↓
 Known matrix Known vector

In Python, we say `linalg.lstsq(A,b)` to get the solution for the least squares problem.

For the Econ data, recall that we wanted to minimize

$$\left\| \underbrace{DS_1}_{\text{known}} x_{\text{known}} + \underbrace{DS_2}_{\text{A}} \underbrace{x_{\text{unknown}}}_{\text{I}} \right\|_2^2$$

Match A and b in the expression above with $\|Ax - b\|_2^2$

Solving systems of equations

Consider the following model for opinion formation among n individuals, each of whom interact with a certain number of individuals in the group. The numerical value of the i th person's opinion is denoted as x_i . The value of x_i is influenced by the following:

- The i th person's self opinion denoted as s_i
- The opinions of the remaining individuals x_j , where $j = 1, 2, \dots, n$ and $j \neq i$.

Assuming that the i th person gives a weightage w_{ij} to the j th person's opinion, we can compute x_i as follows:

$$x_i = \frac{s_i + \sum_{j \neq i} w_{ij} x_j}{1 + \sum_{j \neq i} w_{ij}}, \quad i = 1, \dots, n.$$

1. From the equation above, what do you see is the weightage that a person gives to their own opinion?

2. The equation above can be written as $(A + I)x = s$, where A is an $n \times n$ -matrix and I represents the identity matrix. What are the elements of the matrix A , vectors x and s ?

Scenario 4 people in a network

Self-opinion vector = $s = \begin{bmatrix} s_1 \\ s_2 \\ s_3 \\ s_4 \end{bmatrix}$ Can be calculated based on individual history.

Opinion vector = $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$ That is built after interaction with others

Weight matrix = $A = \begin{bmatrix} w_{11} & w_{12} & w_{13} & w_{14} \\ w_{21} & w_{22} & w_{23} & w_{24} \\ w_{31} & w_{32} & w_{33} & w_{34} \\ w_{41} & w_{42} & w_{43} & w_{44} \end{bmatrix}$

1 → weightage that the 1st individual gives to the 2nd individual

Weights matrix = $W = \begin{bmatrix} w_{11} & w_{12} & w_{13} & w_{14} \\ w_{21} & w_{22} & w_{23} & w_{24} \\ w_{31} & w_{32} & w_{33} & w_{34} \\ w_{41} & w_{42} & w_{43} & w_{44} \end{bmatrix}$

gives to the i th individual's opinion

$$x_i = \frac{s_i + \sum_{j \neq i} w_{ij} x_j}{1 + \sum_{j \neq i} w_{ij}}, \quad \text{Suppose } i=1, \text{ the 1st individual}$$

$$\text{Opinion of 1st individual } x_1 = \frac{s_1 + w_{12} x_2 + w_{13} x_3 + w_{14} x_4}{w_{11} + w_{12} + w_{13} + w_{14}}$$

$$\Rightarrow x_1(w_{11}) + [x_1 w_{12} + x_1 w_{13} + x_1 w_{14}] = s_1 + w_{12} x_2 + w_{13} x_3 + w_{14} x_4$$

$$\Rightarrow x_1(1 + w_{12} + w_{13} + w_{14}) + (-w_{12}) x_2 + (-w_{13}) x_3 + (-w_{14}) x_4 = s_1$$

known known unknown known known known

$$\Rightarrow \begin{cases} w_{12}(x_1 - x_2) + w_{13}(x_1 - x_3) + w_{14}(x_1 - x_4) + 1 \cdot x_1 = s_1 \\ w_{21}(x_2 - x_1) + 1 \cdot x_2 + w_{23}(x_2 - x_3) + w_{24}(x_2 - x_4) = s_2 \\ w_{31}(x_3 - x_1) + w_{32}(x_3 - x_2) + 1 \cdot x_3 + w_{34}(x_3 - x_4) = s_3 \\ w_{41}(x_4 - x_1) + w_{42}(x_4 - x_2) + w_{43}(x_4 - x_3) + 1 \cdot x_4 = s_4 \end{cases}$$

Given the weights matrix W and the self-opinion vector s , we get
4 equations in 4 unknowns

Recap of the matrix-vector product

$$A = [[1 \ 2 \ -1 \ -1] \ [2 \ 4 \ -2 \ 3] \ [-1 \ 1 \ -2 \ 4]], \quad x = \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

$$Ax = \begin{bmatrix} a^{(1)}x \\ a^{(2)}x \\ a^{(3)}x \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \quad \left| \begin{array}{l} \text{Linear combination of the columns} \\ \text{of } A \text{ using the elements of } x \\ \text{as multipliers} \end{array} \right.$$

$$a_1 = \begin{bmatrix} 1 \\ 2 \\ -1 \\ -1 \end{bmatrix}, a_2 = \begin{bmatrix} 2 \\ 4 \\ -2 \\ 3 \end{bmatrix}, a_3 = \begin{bmatrix} -1 \\ 1 \\ -2 \\ 4 \end{bmatrix}, a_4 = \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

In general, if

L10

In general, if

$A: m \times n$ -matrix

$x: n$ -vector

Then $Ax = \begin{cases} \begin{bmatrix} a_1^{(1)} \\ a_1^{(2)} \\ \vdots \\ a_1^{(m)} \end{bmatrix} x \\ x_1 a_1 + x_2 a_2 + \dots + x_n a_n \end{cases}$ linear combination

Same answer

$$\begin{aligned} a_1 &= \begin{bmatrix} 2 \\ -1 \end{bmatrix}, a_2 = \begin{bmatrix} 4 \\ 1 \end{bmatrix}, a_3 = \begin{bmatrix} -2 \\ -2 \end{bmatrix}, a_4 = \begin{bmatrix} 3 \\ 4 \end{bmatrix} \\ \Rightarrow x_1 a_1 + x_2 a_2 + x_3 a_3 + x_4 a_4 &= -1 \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} + (1) \begin{bmatrix} 2 \\ 4 \\ 1 \end{bmatrix} + (-1) \begin{bmatrix} -1 \\ -2 \\ -2 \end{bmatrix} + (3) \begin{bmatrix} 3 \\ 4 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \end{aligned}$$

System of equations with

(1) no solution: $\begin{cases} x_1 + x_2 = 1 \\ x_1 + x_2 = 5 \end{cases}$ no solution

Solution exists means the system is consistent. otherwise the system is inconsistent

(2) unique solution: $\begin{cases} x_1 + x_2 = 1 \\ x_1 + 2x_2 = 1 \end{cases} \Rightarrow \begin{bmatrix} x_1 = 1 \\ x_2 = 0 \end{bmatrix}$ unique solution

(3) Infinitely many solutions: $\begin{cases} x_1 + x_2 = 1 \\ x_1 + 2x_2 = 1 \end{cases} \Rightarrow \begin{bmatrix} x_1 = 1 \\ x_2 = 0 \end{bmatrix}, \begin{bmatrix} x_1 = 0.5 \\ x_2 = 0.5 \end{bmatrix}, \begin{bmatrix} x_1 = 0.4 \\ x_2 = 0.6 \end{bmatrix}$ infinitely many solutions

Elementary row operations

1. Divide/multiply a row by a non zero constant.
2. Subtract a scalar multiple of one row from another row.
3. Exchange two rows.

$$\begin{cases} x_1 + x_2 = 1 \\ x_1 + 2x_2 = 1 \end{cases}$$



$$2(x_1 + x_2) = 2$$

$$x_1 + 2x_2 = 1$$

$$\begin{cases} 2x_1 + 2x_2 = 2 \\ x_1 + 2x_2 = 1 \end{cases}$$

$$\begin{cases} x_1 + x_2 = 1 \\ x_1 + 2x_2 = 1 \end{cases}$$



$$\begin{cases} x_1 + 2x_2 = 1 \\ x_1 + x_2 = 1 \end{cases}$$

$$Eq(1) = Eq(1) + 2 * Eq(2)$$

$$\begin{cases} 3x_1 + 5x_2 = 3 \\ x_1 + 2x_2 = 1 \end{cases}$$

Reduced row echelon form (RREF)

$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 0 & | & 2 \\ 0 & 0 & 0 & 1 & | & 0 \end{array} \right]$$

$$\left[\begin{array}{cccc|c} 1 & 0 & 1 & 0 & | & 1 \\ 0 & 1 & 0 & 1 & | & 0 \end{array} \right]$$

1st column
2nd column
3rd column
4th column

$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 2 & 2 & 0 & 1 & 6 \\ 0 & 1 & -1 & 1 & 3 \\ -1 & -2 & 1 & 1 & -1 \end{array} \right] \xrightarrow{\text{Original augmented matrix}} \left[\begin{array}{cccc|c} 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & -1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \xrightarrow{\text{rref}}$$

4n pivot
 ↓
 x_1, x_2, x_4
 as the pivot
 variables
 and x_3 is the
 free variable

$$\left\{ \begin{array}{l} x_1 + x_2 = 2 \\ 2x_1 + 2x_2 + x_4 = 6 \\ x_2 - x_3 + x_4 = 3 \\ -x_1 - 2x_2 + x_3 + x_4 = -1 \end{array} \right. \Rightarrow \left\{ \begin{array}{l} x_1 + x_3 = 1 \\ x_2 - x_3 = 1 \\ x_4 = 2 \\ \boxed{0=0} \end{array} \right. \begin{array}{l} \text{anything about } x_3 \text{ here?} \\ \text{Nothing!} \\ x_3 \text{ can be anything} \end{array}$$

$$\begin{aligned}
 \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} &= \begin{bmatrix} 1 - x_3 \\ 1 + x_3 \\ x_3 \\ 2 \end{bmatrix} \\
 \Rightarrow x &= \begin{bmatrix} 1 \\ 1 \\ 0 \\ 2 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} \quad \text{such that}
 \end{aligned}$$

$$\Rightarrow \left\{ \begin{array}{l} x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 2 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} \\ x_3 \in \mathbb{R} \end{array} \right\}$$

No solution case

$$\begin{aligned}
 x_1 + x_2 &= 1 \\
 x_1 + x_2 &= 5
 \end{aligned}$$

$$\Rightarrow \text{Augmented matrix} = \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 1 & 1 & 5 \end{array} \right]$$

$$\xrightarrow{\text{Row 2} = \text{Row 2} - \text{Row 1}} \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & 0 & 4 \end{array} \right]$$

Ininitely many solutions case

$$\begin{aligned}
 x_1 + x_2 &= 1 \\
 2x_1 + 2x_2 &= 2
 \end{aligned}$$

$$\Rightarrow \text{Augmented matrix} = \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 2 & 2 & 2 \end{array} \right]$$

$$\Rightarrow \text{Augmented matrix} = \left[\begin{array}{cccc|c} 1 & 2 & -1 & -1 & 1 \\ 2 & 4 & -2 & 3 & 3 \\ -1 & 1 & -2 & 4 & 2 \end{array} \right]$$

$$RREF = \left[\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right]$$

Example , solving $Ax = b$

$$\left[\begin{array}{cccc|c} 1 & 2 & -1 & -1 & 1 \\ 2 & 4 & -2 & 3 & 3 \\ -1 & 1 & -2 & 4 & 2 \end{array} \right]$$

Original augmented matrix \uparrow

RREF using SymPy \uparrow

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \underbrace{\begin{bmatrix} -x_3 - \frac{2}{5} \\ x_3 + \frac{4}{5} \\ x_3 \\ \frac{1}{5} \end{bmatrix}}_{\text{solution to } Ax=0} + \underbrace{\begin{bmatrix} -\frac{2}{5} \\ \frac{4}{5} \\ 0 \\ \frac{1}{5} \end{bmatrix}}_{\text{particular solution}}$$

Solution Solution vector x

$$A \left(x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

RHS vector b

$$A \left(x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} \right) + A \begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix}$$

$$= x_3 * A \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} + A \begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix}$$

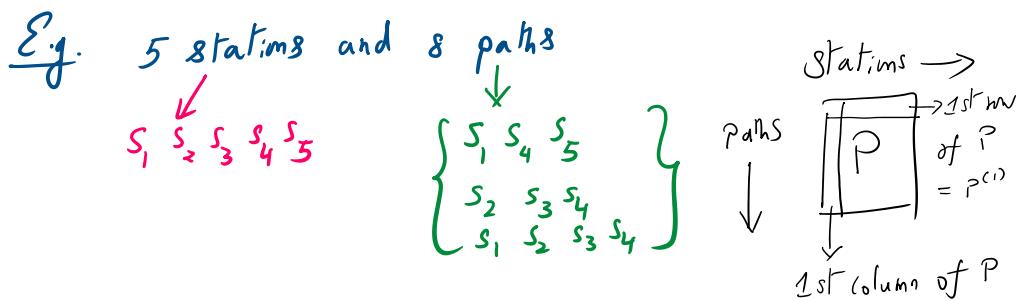
$x_3 *$ $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} +$

Algebra Sessional-2 Review

(i) Matrices : matrix-vector and matrix-matrix products

Questions will be scenario-based

Questions will be scenario-based



P_1 : 8-vector, e.g. $P_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$

$P^{(1)}$: 5-vector, e.g. $P^{(1)} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$

Station-1 is in the first path
Station-2 is in the second path
Station-1 is not in the third path

$P^{(1)}$ contains station-1
 $P^{(1)}$ does not contain station-2

E.g.

The 8×5 -matrix P

1-vector = $\begin{bmatrix} \vdots \\ \vdots \end{bmatrix}$, P_1 is a special matrix-vector product

$P = \begin{bmatrix} P & P^{(1)^T} \end{bmatrix} \Rightarrow P_1 = \begin{bmatrix} P & P^{(1)^T} \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$

$= \begin{bmatrix} 3 \\ 2 \\ 2 \\ 4 \\ 3 \\ 0 \\ 5 \\ 2 \end{bmatrix}$

No. of stations in Path-1
No. of stations in Path-2

E.g. What about the matrix-vector product P_1^T ?

5×8 -matrix 8 -vector

QUESTION

5x8-matrix \downarrow 8-vector

$(P^T \mathbf{1})_1$ \rightarrow 1st component of the vector
 \downarrow vector = How many paths in which station-1 shows up

$(P \mathbf{1})_3$ = 3rd component of the vector
 \downarrow vector = How many stations are in the 3rd path

Recall the unit vectors: $e_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$, $e_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$

$$Pe_1 = \begin{bmatrix} 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = 1 \times \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + 0 \times \dots$$

= Paths in which station-1 shows up $= P_1 = 1\text{st column of } P$

Recall example

$$P = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, \quad x = \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix}$$

$$Px = \left\{ \begin{array}{l} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{bmatrix} \cdot \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix} \\ 7 \begin{bmatrix} 1 \\ 4 \end{bmatrix} + 8 \begin{bmatrix} 2 \\ 5 \end{bmatrix} + 9 \begin{bmatrix} 3 \\ 6 \end{bmatrix} \end{array} \right. \begin{array}{l} \text{Dot-product version} \\ \text{linear combination version} \end{array}$$

$$e_1^T P e_3 = ?$$

\downarrow matrix-vector product $= P_3 = 3\text{rd column of matrix } P$
 $=$ Paths in which station-3

\downarrow
 matrix-vector product = P_3 = 3rd column of matrix P
 $= P_{:,3}$ in which stamp-3

$e_1^T P_3 = P_3^T e_1 = P_3 \cdot e_1 = e_1^T P_3 = 1st$ component of
 matrix-vector product Dot-product the vector P_3

$$P_3 \cdot e_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = 0$$

In general $\left\{ \begin{array}{l} Pe_i = P_i = i^{th} \text{ column of matrix } P \\ e_i^T Pe_j = (i,j)^{th} \text{ element of matrix } P \\ e_i^T P = p^{(i)} = i^{th} \text{ row of matrix } P \end{array} \right.$

Properties of matrix-vector product: $(AB)^T = B^T A^T$, $(A^T)^T = A$

$$e_i^T P = (P^T e_i)^T = e_i^T (P^T)^T = e_i^T P$$

$\downarrow \quad \downarrow \quad \downarrow \quad \downarrow$
 $A \quad B \quad B^T \quad A^T$

ith column of P^T = ith row of P

Build a matrix-vector product from a description

E.g. a signal sampled at 6 time steps

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} . \text{Samples at only the even time steps}$$

$$= S \underset{\substack{\text{Sampling matrix} \\ \downarrow}}{x} \underset{\substack{\text{Signal}}}{} = \begin{bmatrix} x_2 \\ x_4 \\ x_6 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} = \begin{bmatrix} x_2 \\ x_4 \\ x_6 \end{bmatrix}$$

What should be the matrix S such that we weight three successive timestamps equally to generate a new value.

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} \rightarrow \begin{array}{l} \frac{1}{3} \cdot x_1 + \frac{1}{3} \cdot x_2 + \frac{1}{3} \cdot x_3 \\ \frac{1}{3} \cdot x_2 + \frac{1}{3} \cdot x_3 + \frac{1}{3} \cdot x_4 \\ ? \\ ? \end{array}$$

$$\begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 & 0 & 0 \\ 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 & 0 \\ 0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 \\ 0 & 0 & 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} \Downarrow S \Downarrow x$$

Matrix-Matrix Product

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

(2x3) 5x2

Matrix-Vector product $\begin{bmatrix} * \\ * \end{bmatrix}$ 2x2

Matrix-Vector product $\begin{bmatrix} * \\ * \end{bmatrix}$

$$P = \begin{bmatrix} 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{bmatrix}, P^T = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 & 1 & 0 & 1 & 1 \end{bmatrix}$$

8 × 5 - matrix

5 × 8 - matrix

Which matrix-matrix product makes sense:

$\cancel{P^T P}$ or $\cancel{PP^T}$ or $\cancel{P^2 P}$
 $\downarrow \quad \downarrow \quad \downarrow$
 $5 \times 8 \quad 8 \times 5 \quad 8 \times 5 \quad 5 \times 8$
 $\downarrow \quad \downarrow \quad \downarrow \quad \downarrow$
 $5 \times 5 \quad 8 \times 8 \quad 8 \times 5 \quad 8 \times 5$
 $-matrix \quad -matrix \quad \downarrow \quad \downarrow$
inner dimensions do not match

$$[AB]_{i,j} = a^{(i)} \cdot b_j = a^{(i)T} b_j$$

$\downarrow \quad \downarrow$
*i*th row *j*th column
of *A* of *B*

$$[P^T P]_{ij} = (\text{i-th row of } P^T) \cdot (\text{j-th column of } P)$$

\downarrow
(i-th column of *P*) . (j-th column of *P*)

$$= P_i \cdot P_j = P_i^T P_j$$

$$\text{for e.g. } i=1, j=2 \Rightarrow [P^T P]_{1,2} = P_1 \cdot P_2$$

$$P = \begin{bmatrix} 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{bmatrix}$$

P_1, P_2

$P_1 \cdot P_2 = 1 = \text{no. of paths common}$
 $\text{to station-1 and station-2}$

How about $P_1 - P_2$? $P_1 - P_2 = \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \\ 0 \end{bmatrix}$

RREF

$$\left[\begin{array}{ccccc|c} 1 & 0 & 1 & 0 & -2/5 \\ 0 & 1 & -1 & 0 & 4/5 \end{array} \right]$$

Always do this first

Is the system
consistent
or ...

$$\left[\begin{array}{cccc|c} 1, & 0, & 1, & 0, & -2/5, \\ 0, & 1, & -1, & 0, & 4/5, \\ 0, & 0, & 0, & 1, & 1/5 \end{array} \right]$$

consistent
or
inconsistent?

$$\left[\begin{array}{cccc} x_1 & x_2 & x_3 & x_4 \\ 1 & 0, & 1, & 0, \\ 0, & 1, & -1, & 0, \\ 0, & 0, & 0, & 1 \end{array} \right] \quad \left[\begin{array}{c} -2/5, \\ 4/5, \\ 1/5 \end{array} \right]$$

x_1, x_2, x_4 = pivot variables
 x_3 = free variable

$$\begin{aligned} x_1 + x_3 &= -2/5 \\ x_2 - x_3 &= 4/5 \\ x_4 &= 1/5 \end{aligned} \Rightarrow$$

$$\begin{cases} x_1 = -2/5 - x_3 \\ x_2 = 4/5 + x_3 \\ x_3 = x_3 \\ x_4 = 1/5 \end{cases}$$

$$\Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \text{ where } x_3 \in \mathbb{R}$$

This system has infinitely many solutions, but here is an easy choice of x_3 which is equal to 0.

$$\left[\begin{array}{cccc} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{array} \right] \left(\begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

Coefficient matrix A

Solution vector x

Right hand side vector b

$$\left[\begin{array}{cccc} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{array} \right] \left(\begin{bmatrix} -2/5 \\ 4/5 \\ 0 \\ 1/5 \end{bmatrix} + x_3 \begin{bmatrix} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} \right)$$

Matrix-vector product

$$= \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

Matrix-vector product

$$= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\underbrace{\begin{bmatrix} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{bmatrix}}_A \quad x_3 \underbrace{\begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix}}_{\text{solution to } Ax=0} + \underbrace{\begin{bmatrix} -\frac{2}{5} \\ \frac{4}{5} \\ 0 \\ \frac{1}{5} \end{bmatrix}}_{\text{particular solution}}$$

Solution vector

and there are infinitely

many of them

$$= \underbrace{\begin{bmatrix} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{bmatrix}}_A \cdot x_3 \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} + \underbrace{\begin{bmatrix} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{bmatrix}}_A \begin{bmatrix} -\frac{2}{5} \\ \frac{4}{5} \\ 0 \\ \frac{1}{5} \end{bmatrix}$$

$$= x_3 \underbrace{\begin{bmatrix} 1 & 2 & -1 & -1 \\ 2 & 4 & -2 & 3 \\ -1 & 1 & -2 & 4 \end{bmatrix}}_A \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

$$= x_3 \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

Linear combination of vectors

$$a_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, a_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

One linear combination: $0 * \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 0 * \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Multiples

Another linear combination: $1 * \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 0 * \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

One more linear combination: $1 * \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 1 * \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$

One last linear combination: $? * \begin{bmatrix} 1 \\ 2 \end{bmatrix} + ? * \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Matrix-vector
product

$$\begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

R.H.S. vector is special,
 it is the zero vector

Augmented matrix = $\left[\begin{array}{cc|c} 1 & 2 & 0 \\ 2 & 1 & 0 \end{array} \right] \xrightarrow{\text{Elementary operations}} \left[\begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \end{array} \right]$

$\Rightarrow \boxed{x_1 = 0 \text{ and } x_2 = 0}$ RREF

The only multipliers that can result in a zero linear combination

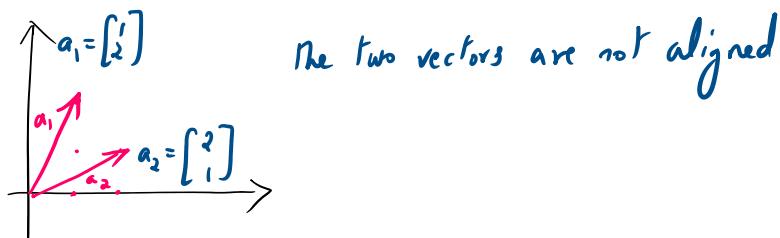
of the vectors $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ are zeros. In other words,

$Ax = 0$ is possible only when $x = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$, the zero vector

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad \begin{bmatrix} 2 \\ 1 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

a_1 and $a_2 \Rightarrow$ The vectors a_1 and a_2 (columns of matrix A) are said to be linearly independent.

Geometric perspective of linear independence



E.g. $a_1 = \begin{bmatrix} 72 \\ 76 \end{bmatrix}, a_2 = \begin{bmatrix} 144 \\ 152 \\ 112 \end{bmatrix}$

$$\underline{a_1} = \begin{bmatrix} 72 \\ 76 \\ 80 \end{bmatrix}, \quad a_2 = \begin{bmatrix} 144 \\ 152 \\ 160 \end{bmatrix}$$

Are a_1 and a_2 linearly independent?

$$Ax = 0 \Rightarrow \begin{bmatrix} 72 & 144 \\ 76 & 152 \\ 80 & 160 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Augmented matrix

$$\left[\begin{array}{cc|c} 72 & 144 & 0 \\ 76 & 152 & 0 \\ 80 & 160 & 0 \end{array} \right] \xrightarrow[\text{Elementary row operations}]{\text{RREF}} \left[\begin{array}{cc|c} 1 & 2 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right]$$

$$x_1 + 2x_2 = 0 \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

Choose the free variable $x_2 = 1 \Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$

$$\Rightarrow x_1 a_1 + x_2 a_2 = -2a_1 + 1 \cdot a_2 = 0$$

$$\Rightarrow \boxed{a_2 = 2a_1} \Rightarrow \boxed{BP = 2 * HR \text{ is what we understand}}$$

The vectors a_1 and a_2 are linearly dependent

Suppose we had HR , BP , and $TEMP$. We observe that

(BP) $(mmHg)$ $(^\circ C)$

$$\boxed{BP = 1.2 * HR + 1.05 * TEMP}$$

$$\boxed{Hemoglobin = \beta_0 + \beta_1 * HR + \beta_2 * BP + \beta_3 * TEMP}$$

Linear model

E.g.

$$\underbrace{\begin{bmatrix} 1 & 2 & 4 \\ 2 & 1 & 5 \end{bmatrix}}_A \cdot \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}}_x = \underbrace{\begin{bmatrix} 0 \\ 0 \end{bmatrix}}_{0 \text{ vector}} \Rightarrow x_1 a_1 + x_2 a_2 + x_3 a_3 = 0$$

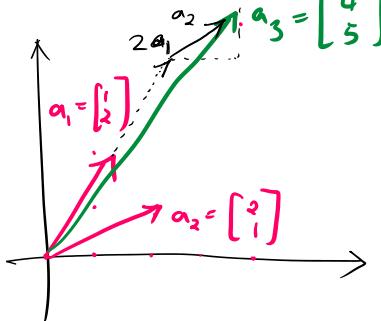
$$\left[\begin{array}{ccc|c} 1 & 0 & 2 & 0 \\ 0 & 1 & 1 & 0 \end{array} \right] = \text{RREF of the augmented matrix}$$

$$\begin{matrix} x_1 \\ \parallel \\ \text{pivot} \end{matrix} \quad \begin{matrix} x_2 \\ \parallel \\ \text{pivot} \end{matrix} \quad \begin{matrix} x_3 \\ \parallel \\ \text{free} \end{matrix}$$

$$x_1 + 2x_3 = 0 \Rightarrow x_1 = -2x_3$$

$$x_2 + x_3 = 0 \Rightarrow x_2 = -x_3$$

$$x_3 = x_3 \Rightarrow x_3 = x_3$$



$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -2x_3 \\ -x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ -1 \\ 1 \end{bmatrix}$$

Easy choice for $x_3 = 1 \Rightarrow x = \begin{bmatrix} -2 \\ -1 \\ 1 \end{bmatrix} \Rightarrow -2a_1 + (-1)a_2 + a_3 = 0 \Rightarrow a_3 = 2a_1 + a_2$

E.g. $A = \begin{bmatrix} 1 & 4 & 7 & 10 \\ 2 & 5 & 8 & 11 \\ 3 & 6 & 9 & 12 \end{bmatrix}$ are the columns linearly dependent or independent?

Solving $Ax = 0$ (RHS vector is the zero vector)

$$\text{RREF} = \left[\begin{array}{cccc|c} 1 & 0 & -1 & -2 & 0 \\ 0 & 1 & 2 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$\begin{matrix} x_1 & x_2 & x_3 & x_4 \\ \hline \text{Pivots} & \text{free} \end{matrix}$

$$x_1 - x_3 - 2x_4 = 0$$

$$x_2 + 2x_3 + 3x_4 = 0$$

$$\Rightarrow x_1 = x_3 + 2x_4$$

$$x_2 = -2x_3 - 3x_4$$

$$x_3 = x_3$$

$$x_4 = x_4$$

$$\Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 2 \\ -3 \\ 0 \\ 1 \end{bmatrix}$$

Easy choices for x_3, x_4 are $x_3=1$ and $x_4=1$

$$\Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 3 \\ -5 \\ 1 \\ 1 \end{bmatrix} \Rightarrow 3a_1 - 5a_2 + 1a_3 + 1a_4 = 0$$

\Rightarrow Columns of A are linearly dependent

For an $m \times n$ -matrix

(1) $m > n$ (more rows than columns)

No. of pivots = at most n

E.g. $m=5, n=2$, RREF =
$$\left[\begin{array}{cc|c} 1 & 0 & * \\ 0 & 1 & * \\ 0 & 0 & * \\ 0 & 0 & * \\ 0 & 0 & * \end{array} \right]$$

(2) $m < n$ (more columns than rows)

No. of pivots = at most m

E.g. $m=2, n=5$, RREF =
$$\left[\begin{array}{ccccc|c} 1 & 0 & * & * & * & * \\ 0 & 1 & * & * & * & * \end{array} \right]$$

(3) $m=n$ (same no. of rows and columns)

No. of pivots = at most $m =$ at most n

E.g. $m=5, n=5$, RREF =
$$\left[\begin{array}{ccccc} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right]$$

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 4 & 3 & 2 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} [1, 0, -1, -2, 0] \\ [0, 1, 2, 3, 0] \\ [0, 0, 0, 0, 0] \end{bmatrix}$$

Coefficient matrix

RREF

$$\boxed{\begin{aligned} x_1 - x_3 - 2x_4 &= 0, \\ x_2 + 2x_3 + 3x_4 &= 0, \end{aligned}} \Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_3 + 2x_4 \\ -2x_3 - 3x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 2 \\ -3 \\ 0 \\ 1 \end{bmatrix}$$

$$\boxed{\mathbf{a}_1 - 2\mathbf{a}_2 + \mathbf{a}_3 = \mathbf{0} \Rightarrow \mathbf{a}_3 = -\mathbf{a}_1 + 2\mathbf{a}_2.}$$

$$\boxed{2\mathbf{a}_1 - 3\mathbf{a}_2 + \mathbf{a}_4 = \mathbf{0} \Rightarrow \mathbf{a}_4 = -2\mathbf{a}_1 + 3\mathbf{a}_2.}$$

↓
The only columns in A that matter are \mathbf{a}_1 and \mathbf{a}_2

Columns corresponding to the free variables can be written as a linear combination of columns corresponding to the pivot variables

Column space of matrix A denoted as $C(A)$

The set of all possible vectors that can be generated using a linear combination of the columns of A

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 4 & 3 & 2 & 1 \end{bmatrix}$$

$\downarrow \quad \downarrow \quad \downarrow \quad \downarrow$
 $\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3 \quad \mathbf{a}_4$

$$\begin{aligned} C(A) &= \left\{ \alpha_1 \mathbf{a}_1 + \alpha_2 \mathbf{a}_2 + \alpha_3 \mathbf{a}_3 + \alpha_4 \mathbf{a}_4 \mid \alpha_1, \alpha_2, \alpha_3, \alpha_4 \in \mathbb{R} \right\} \\ &= \left\{ \alpha_1 \mathbf{a}_1 + \alpha_2 \mathbf{a}_2 \mid \alpha_1, \alpha_2 \in \mathbb{R} \right\} \end{aligned}$$

Why do we need to define a 'column space' of a matrix?

To solve $Ax = b$

$\downarrow \quad \downarrow \quad \downarrow$
 $m \times n \quad n \quad m$

If this system is consistent, here is a set

of values x_1, x_2, \dots, x_n s.t.

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + x_3 \mathbf{a}_3 + \dots + x_n \mathbf{a}_n = b$$

E.g. $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \text{ solve } Ax = b$

Augmented matrix =
$$\left[\begin{array}{cc|c} x_1 & x_2 & \\ 1 & 0 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{array} \right] \Rightarrow \begin{array}{l} x_1 = 1 \\ x_2 = 2 \end{array}$$

Now we have $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$, $b = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, solve $Ax = b$

Augmented matrix = $\left[\begin{array}{cc|c} 1 & 0 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 3 \end{array} \right]$

\rightarrow Inconsistent

Is $b \in C(A)^\perp$?

$$C(A) = \left\{ \alpha_1 a_1 + \alpha_2 a_2 \mid \alpha_1, \alpha_2 \in \mathbb{R} \right\}$$

$$= \left\{ \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ 0 \end{bmatrix} \mid \alpha_1, \alpha_2 \in \mathbb{R} \right\}$$

Can we find a 'Compromise' solution

- We want to solve $Ax = b$
- There is no s.t. $Ax = b$. Why? Because $b \notin C(A)$
- Let us find a compromise solution x s.t.

$\|Ax - b\|^2$ is as small as possible

$$\begin{aligned} &= \left\| \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\|^2 \\ &= \left\| \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\|^2 = \left\| \begin{bmatrix} x_1 - 1 \\ x_2 - 2 \\ -3 \end{bmatrix} \right\|^2 \end{aligned}$$

$$\Rightarrow x_1 = 1, x_2 = 2$$

The least squares solution to $Ax = b$ is obtained by
minimizing $\|Ax - b\|^2$ Gradient Calculation

$$x = A(A^T A)^{-1} A^T b$$

E.g. $\|2x - 4\|^2$, $x = 2$

$$\left\| \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} - \begin{bmatrix} 4 \\ 6 \end{bmatrix} \right\|^2, \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

- Projection of a vector onto the direction of another vector

E.g.

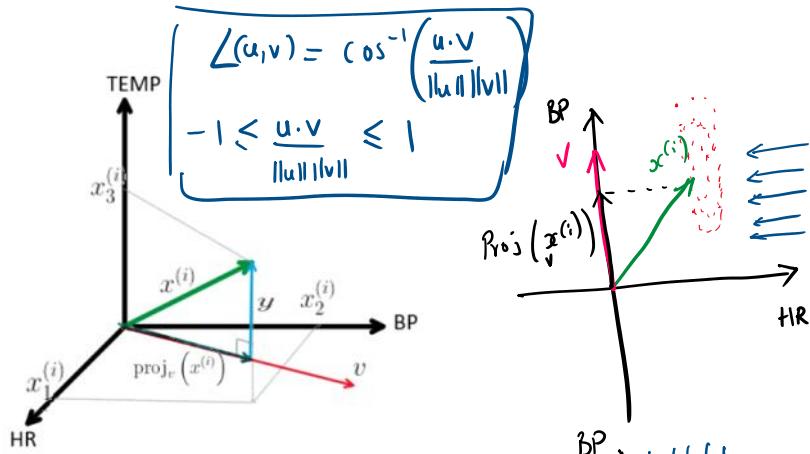
Data matrix

	HR	BP	Temp
Patient-1	76	126	38.0
Patient-2	74	120	38.0
Patient-3	72	118	37.5
Patient-4	78	136	37.0

$$x^{(i)} = \begin{bmatrix} 76 \\ 126 \\ 38 \end{bmatrix}$$

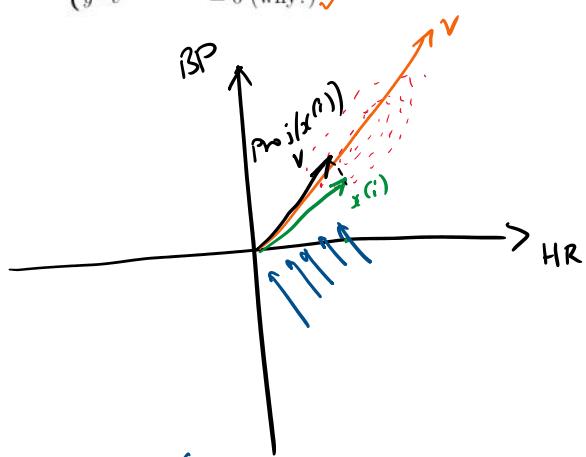
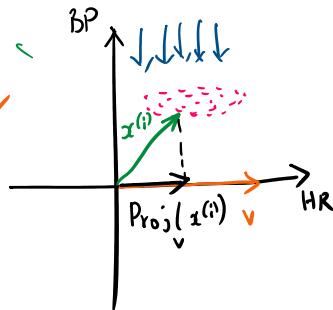
E.g.

$$v = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, 2v = 2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$



We can derive an expression for the projection as follows:

$$\begin{cases} \text{proj}_v(x^{(i)}) = cv, \text{ for some unknown constant } c \text{ (why?)} \\ y = x^{(i)} - \text{proj}_v(x^{(i)}) \text{ (why?)} \\ y \cdot v = 0 \text{ (why?)} \end{cases}$$



$y \cdot v = 0$ (y and v are orthogonal to each other)

$$\Rightarrow [x^{(i)} - \text{proj}_v(x^{(i)})] \cdot v = 0$$

$\Rightarrow c v \cdot v = 0$

$$\Rightarrow x^{(i)} \cdot v - \text{Proj}_v(x^{(i)}) \cdot v = 0$$

$$\Rightarrow x^{(i)} \cdot v - (c v) \cdot v = 0 \Rightarrow c(v \cdot v) = x^{(i)} \cdot v$$

$$\Rightarrow c = \frac{x^{(i)} \cdot v}{v \cdot v} \Rightarrow \text{Proj}_v(x^{(i)}) = cv = \left[\frac{x^{(i)} \cdot v}{v \cdot v} \right] v$$

$$v \cdot v = \|v\|^2, \quad v = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \Rightarrow v \cdot v = v_1^2 + v_2^2$$

$$\|v\|^2 = \left(\sqrt{v_1^2 + v_2^2} \right)^2 = v_1^2 + v_2^2$$

$$\text{Proj}_v(x^{(i)}) = \left[\frac{(x^{(i)} \cdot v)}{\|v\|} \right] \frac{v}{\|v\|} = \|v\|^2 = v \cdot v$$

$$\text{Proj}_v(x^{(i)}) = \left[\frac{x^{(i)} \cdot v}{\|v\|} \right] \frac{v}{\|v\|}$$

Magnitude Direction
of the projection of projection
(shadow length)

unit vector (direction)

$$\text{e.g. } v = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \begin{array}{l} \uparrow \\ \rightarrow \end{array}$$

$$\frac{v}{\|v\|} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}, \quad \left\| \frac{v}{\|v\|} \right\| = \sqrt{\left(\frac{1}{\sqrt{2}}\right)^2 + \left(\frac{1}{\sqrt{2}}\right)^2} = 1$$

Suppose the vector v is already a unit vector,

$$\text{Proj}_v(x^{(i)}) = (x^{(i)} \cdot v) v$$

Block matrix-vector operations

$$\begin{bmatrix} (x^{(1)})^T v \\ (x^{(2)})^T v \\ (x^{(3)})^T v \\ (x^{(4)})^T v \end{bmatrix} = \begin{bmatrix} x^{(1)\top} \\ x^{(2)\top} \\ x^{(3)\top} \\ x^{(4)\top} \end{bmatrix} v \downarrow \text{matrix} \quad \downarrow \text{vector} = X_v$$

$$\begin{bmatrix} 1 \cdot 5 \\ 2 \cdot 5 \\ 3 \cdot 5 \\ 4 \cdot 5 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

	HR	BP	Temp	Data matrix in terms of its rows	Data matrix in terms of its columns
Patient-1	76	126	38.0	$x = \begin{bmatrix} x^{(1)\top} \\ x^{(2)\top} \\ x^{(3)\top} \\ x^{(4)\top} \end{bmatrix}$	$= [x_1 \ x_2 \ x_3]$
Patient-2	74	120	38.0		
Patient-3	72	118	37.5		
Patient-4	78	136	37.0		

a = vector of heart rates

b = vector of BP

This tells us how HR and BP are correlated.

$\frac{a \cdot b}{\|a\| \|b\|}$ is guaranteed to be between -1 and 1

$$\frac{a \cdot b}{\|a\| \|b\|}$$

Correlation coeff. between HR and BP

is also guaranteed to be between -1 and 1

ALA Final Exam Review

- (1) Given a vector, calculate the following for that vector: average, standard deviation, RMS.
- (2) Given a pair of vectors, calculate the following between those vectors: angle, correlation, distance
- (3) Given a data matrix, interpret dot-products, distances, and angles between vectors in terms of the data (like the MAHE registrar problem on the sessional).
- (4) Interpret the dot product in terms of the data (like, what is the meaning of dot product with the ones-vector, with the unit vector etc.)
- (5) Interpret linear combination of vectors intuitively in terms of the data (like, which feature is weighted more.)
- (6) Understand scalar (shadow length) and vector projection, be able to calculate for simple vectors, and interpret the results (we had an assignment on this.)
- (7) Calculate matrix-vector and matrix-matrix products and explain, in terms of the data, what the results mean. Also, explain matrix-vector products like Ae_1 , $A1$ (like in the stations example on the sessional) in terms of the data.
- (8) Build a matrix-vector product that describes the scenario. For example, calculating the raw scores for an image, the ECG problem that we had as an assignment, the material-movement example that I gave as a challenge problem, and the projection of samples which we had as an assignment.
- (9) Intuitively explain what the rows and columns of a data matrix convey about the data.
- (10) Understand the different ways a matrix-vector product can be computed, and the intuitive meaning behind those operations and the results.
- (11) Understand how a matrix-matrix product is computed and interpret the results intuitively in terms of the data.
- (12) Understand the RREF of an augmented matrix, check if a system of equations is consistent or not; if consistent, check if there are free variables or only pivot variables; write the solution as a set of vectors; write down the null space of a matrix.

E.g. $a = \begin{bmatrix} -1 \\ 2 \\ -3 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$

$$\text{avg}(a) = \frac{(-1) + 2 + (-3)}{3} = -\frac{2}{3}$$

$$= \frac{1}{3}(a \cdot \underline{1}) = \frac{1}{3} a^T \underline{1}$$

$$\text{STD}(a) = \text{RMS}(a_m)$$

$$a_m = a - \left(\frac{1}{3} a^T \underline{1} \right) \underline{1}$$

\downarrow
ones vector
 $\text{avg}(a)$

$$\text{RMS}(a) = \sqrt{\frac{1}{3} ((-1)^2 + 2^2 + (-3)^2)}$$

Deviation vector

$$\text{std}(a) = \text{RMS}(a_m)$$

$$\Rightarrow \mathbf{a}_m = \begin{bmatrix} -1 \\ 2 \\ -3 \end{bmatrix} - \left(-\frac{2}{3} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 + \frac{2}{3} \\ 2 + \frac{2}{3} \\ -3 + \frac{2}{3} \end{bmatrix} \Rightarrow \text{Deviation vector}$$

$$S\text{td}(\mathbf{a}) = \text{RMS}(\mathbf{a}_m) = \sqrt{\frac{1}{3} \left(\left(-\frac{1}{3} \right)^2 + \left(\frac{8}{3} \right)^2 + \left(-\frac{7}{3} \right)^2 \right)}$$

Large standard deviation for a vector \Rightarrow Components of the vector will be spread around the average value

$$\mathbf{a} = \begin{bmatrix} 1.01 \\ 0.99 \\ 1.02 \\ 0.98 \end{bmatrix}, \quad \mathbf{a}_m \approx \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

E.g. $x^{(1)} = \begin{bmatrix} 1.6 \\ 4 \end{bmatrix}$, $x^{(2)} = \begin{bmatrix} 1.5 \\ 4 \end{bmatrix}$, $x^{(3)} = \begin{bmatrix} 1.6 \\ 2 \end{bmatrix}$

area in m² of Sq. ft -
no. of bedrooms

House-1 is more similar to house-2 or house-3?

Distance $\|x^{(1)} - x^{(3)}\|^2 = \left\| \begin{bmatrix} 1.6 \\ 4 \end{bmatrix} - \begin{bmatrix} 1.6 \\ 2 \end{bmatrix} \right\|^2 = \left\| \begin{bmatrix} 0 \\ 2 \end{bmatrix} \right\|^2 = \sqrt{0^2 + 2^2} = 2$

Squared distances $\|x^{(1)} - x^{(2)}\|^2 = \left\| \begin{bmatrix} 1.6 \\ 4 \end{bmatrix} - \begin{bmatrix} 1.5 \\ 4 \end{bmatrix} \right\|^2 = \left\| \begin{bmatrix} 0.1 \\ 0 \end{bmatrix} \right\|^2 = \sqrt{(0.1)^2 + 0^2} = 0.1$

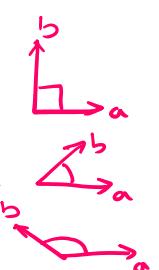
Standardize vectors before calculating distances.

Angle between vectors

$$\cos(\angle_{\mathbf{a}, \mathbf{b}}) = \frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{a}\| \|\mathbf{b}\|}$$

$$\mathbf{a} = \begin{bmatrix} -1 \\ 2 \\ 3 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \Rightarrow \cos(\angle_{\mathbf{a}, \mathbf{b}}) = \frac{-1 + 0 + 6}{\sqrt{(-1)^2 + 2^2 + 3^2} \times \sqrt{1^2 + 0^2 + 2^2}}$$

$$\mathbf{a} \cdot \mathbf{b} = \begin{cases} 0, & \angle_{\mathbf{a}, \mathbf{b}} = \pi/2 \\ +ve, & \angle_{\mathbf{a}, \mathbf{b}} < \pi/2 \\ -ve, & \angle_{\mathbf{a}, \mathbf{b}} > \pi/2 \end{cases}$$



$$0 \leq \angle_{\mathbf{a}, \mathbf{b}} \leq \pi$$

$$0^\circ \leq \angle_{\mathbf{a}, \mathbf{b}} \leq 180^\circ$$

between -1 and 1

$$\text{Correlation between } a \text{ and } b = \cos(\angle a_m \cdot b_m) = \frac{a_m \cdot b_m}{\|a_m\| \|b_m\|}$$

between -1 and 1

[10 points] [TLO 1.1, CO 1] The MAHE registrar has the complete list of courses taken by each graduating student in a program. This data is represented as a matrix X with m rows and n columns as follows:

$$X = \begin{array}{c|cccc} \diagdown & \text{Course} \\ \text{Student} & 1 & 2 & \dots & n \\ \hline 1 & 1 & 1 & \dots & 0 \\ 2 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ m & 1 & 0 & \dots & 1 \end{array}$$

$x^{(i)}$ = $\begin{bmatrix} 1 \\ \vdots \\ 0 \end{bmatrix}$

$x_{\cdot j}$ = $\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$

The entries of the data matrix are 1s and 0s representing whether a particular student has taken a particular course. For example, the red-highlighted entry 1 means that the 1st student has taken the 1st course and the blue-highlighted entry 0 means that the m th student has not taken the 2nd course. Recall that the i th student vector is represented as $x^{(i)}$ and the j th course vector is represented as $x_{\cdot j}$.

(a) The total number of students in the 2nd course is a dot product of two vectors.

$$\text{What are those vectors? } = x_2 \cdot \mathbf{1} = \text{Sum of elements of vector } x_2$$

$$x^{(4)} \cdot \mathbf{1} = \text{no. of courses taken by the 4th student}$$

$$x_k \cdot x_l = \text{dot product of } k\text{th course vector and } l\text{th course vector}$$

$$= \text{no. of students who have taken both courses } k \text{ and } l$$

$$x^{(k)} \cdot x^{(l)} = \text{no. of courses taken in common by student's } k \text{ and } l$$

$$\|x^{(5)} - x^{(6)}\| = \text{Difference between the 5th and the 6th student}$$

[15 points] [TLO 1.2, CO 1] In a multiple channel marketing campaign, potential customers are divided into m market segments, which are groups of customers with similar demographics, e.g., male, female, married, college educated women aged 25-29 etc. A company markets its products by purchasing advertising in a set of n channels, i.e., specific TV or radio shows, magazines, web sites, blogs, direct mail, and so on. The ability of each channel to deliver impressions or views by potential customers is characterized by the data matrix X , which has m rows and n columns representing m customer segments and n advertising channels. The entry in the i th row and j th column of the data matrix is the number of views of customers in segment i for each Rupee spent on channel j . The n -vector c will denote the company's purchases of advertising, in Rupees, in the n channels. Finally, we introduce the m -vector a , where a_i gives the profit in Rupees per impression in market segment i .

- What do the values in the first column of the data matrix indicate?
- What do the values in the third row of the data matrix indicate?
- Express the total amount of money the company spends on advertising using a dot product. $= \sum c_i = c \cdot \mathbf{1}$, $\mathbf{1}$ = ones vector $= \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$
- Consider the vector $c_1 x_1 + c_2 x_2 + \dots + c_n x_n$. This is called a linear combination of the vectors x_1, x_2, \dots, x_n . Interpret this vector in plain English; do this by starting with interpreting the under-highlighted terms below in plain English:

$c_i x_i \rightarrow \begin{array}{l} \text{Views on TV} \\ \text{Rupee} \end{array}$

$\sum c_i x_i = \sum \underbrace{c_i}_{\text{Total TV advertisement budget (Rupees)}} \times \underbrace{x_i}_{\text{Views on TV}}$

$c = \begin{bmatrix} 100000 \\ 50000 \\ 200000 \\ 500000 \end{bmatrix}$

$x = \begin{bmatrix} (x_1)_1 & (x_1)_2 & \dots & (x_1)_m \\ \vdots & \vdots & \ddots & \vdots \\ (x_n)_1 & (x_n)_2 & \dots & (x_n)_m \end{bmatrix}$

$c \cdot x = \begin{bmatrix} c_1 x_1 \\ c_2 x_2 \\ \vdots \\ c_n x_n \end{bmatrix} = \begin{bmatrix} 100000 & 50000 & \dots & 500000 \end{bmatrix}$

$c \cdot x = \text{Total no. of views on TV across all market segments}$

$\nearrow m=4$

	Tv	Radio	Magazine	Internet
Male	0.1			
Female	0.3			
Married	1.5	0	0.7	5.
College				
25-29				

$x = \begin{bmatrix} x_1 & x_2 & x_3 & x_4 \end{bmatrix}$

$0.1 = \text{no. of male views per Rupee spent on TV advertisement}$

$C = \begin{bmatrix} 100000 \\ 50000 \\ 200000 \\ 500000 \end{bmatrix} \rightarrow \begin{array}{l} \text{Total budget in Rupees for} \\ \text{TV advertisement} \end{array}$

- (e) Consider the dot product $a \cdot (c_1x_1 + c_2x_2 + \dots + c_nx_n)$. In two words, state what the resulting number indicates.

\Rightarrow Total no. of views on TV across all market segments

$$\begin{bmatrix} 200000 \\ 500000 \end{bmatrix} \quad \text{TV advertisement}$$

$$c_1x_1 + c_2x_2 + \dots + c_nx_n = \text{Total no. of views across all market segments}$$

$$a = \begin{bmatrix} 0.1 \text{ Rupees} \\ \text{view} \\ 0.2 \text{ R/V} \\ 0.05 \text{ R/V} \\ 0.2 \text{ R/V} \end{bmatrix} = \text{Earnings-per-view vector}$$

$$\underbrace{(c_1x_1 + c_2x_2 + \dots + c_nx_n)}_{\text{Total views vector}} \cdot \underbrace{a}_{\text{Profit-per-view vector}} = \text{Total profit (or) earnings}$$

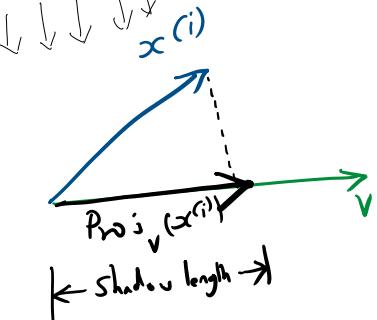
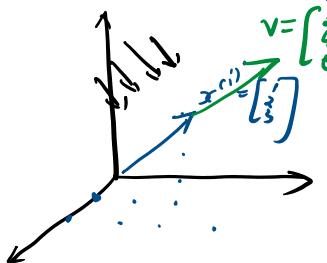
$$\text{proj}_v(x^{(i)}) = \left(\frac{x^{(i)} \cdot v}{v \cdot v} \right) v = \left(\frac{x^{(i)} \cdot v}{\|v\|^2} \right) v = \boxed{\left(\frac{x^{(i)} \cdot v}{\|v\|} \right) \frac{v}{\|v\|}}$$

shadow length direction

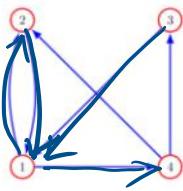
$$x^{(i)} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, v = \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix} \quad \text{Scalar projection}$$

$$\text{Scalar projection (shadow length)} = \frac{x^{(i)} \cdot v}{\|v\|} = \frac{\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix}}{\sqrt{2^2 + 4^2 + 6^2}} = \frac{2 + 8 + 18}{\sqrt{56}} = \frac{28}{\sqrt{56}} = \frac{28}{\sqrt{56}}$$

$$\text{Vector projection} = \left(\frac{x^{(i)} \cdot v}{\|v\|} \right) \frac{v}{\|v\|} = \frac{28}{\sqrt{56}} \times \frac{\begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix}}{\sqrt{56}} = \frac{28}{\sqrt{56}} \begin{bmatrix} 2 \\ 4 \\ 6 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = x^{(i)}$$



[10 points] [CO 2, BT 3] Suppose we have four stations that are connected by train services as shown in the following graph:



$$A = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{bmatrix}$$

The adjacency matrix associated with this graph has entries defined as

$$A_{ij} = \begin{cases} 1 & \text{if direct train service exists from station } j \text{ to station } i, \\ 0 & \text{otherwise.} \end{cases}$$

(a) Write the adjacency matrix A associated with this graph clearly showing its entries.

Sum of elements in each row - $A_1 = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix}$

$(A_1)_2 = \text{no. of stations from which we have a direct train to station-2}$

$$Ae_1 = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \text{stations which have a direct train from station-1}$$

Two stations from which we have a direct train to station-1

Sum of elements in each column - $A^T 1 = \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix} \rightarrow 1 \text{ station for which we have a direct train from station-3}$

$$A = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \quad Ax = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} x_5 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \text{down-shifted version of the vector } x$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ v_1 \\ x_4 \\ v_2 \\ v_3 \end{bmatrix}$$

$$\perp \frac{\begin{bmatrix} (x^{(1)})^T v \\ (x^{(2)})^T v \\ (x^{(3)})^T v \\ (x^{(4)})^T v \end{bmatrix}}{\|v\|} = \text{projection of all samples onto the direction of } v$$

$\|v\|=1$
because v is a direction vector

$$= X^T v \text{ (or) } Xv$$

$$= \begin{bmatrix} x^{(1)\top} \\ x^{(2)\top} \\ x^{(3)\top} \\ x^{(4)\top} \end{bmatrix} v = Xv$$

$$\begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ \dots \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ -1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \dots$$

$A * A$

$$\underbrace{\begin{bmatrix} AB \\ \vdots \end{bmatrix}}_{m \times 1} = \underbrace{\begin{bmatrix} Ab_1 & Ab_2 & \dots & Ab_p \end{bmatrix}}_{n \times p}$$

matrix-vector product