



# LINEAR ALGEBRA FOR MACHINE LEARNING

# WHY LINEAR ALGEBRA IS IMPORTANT FOR ML?

It's a language of Machine learning.

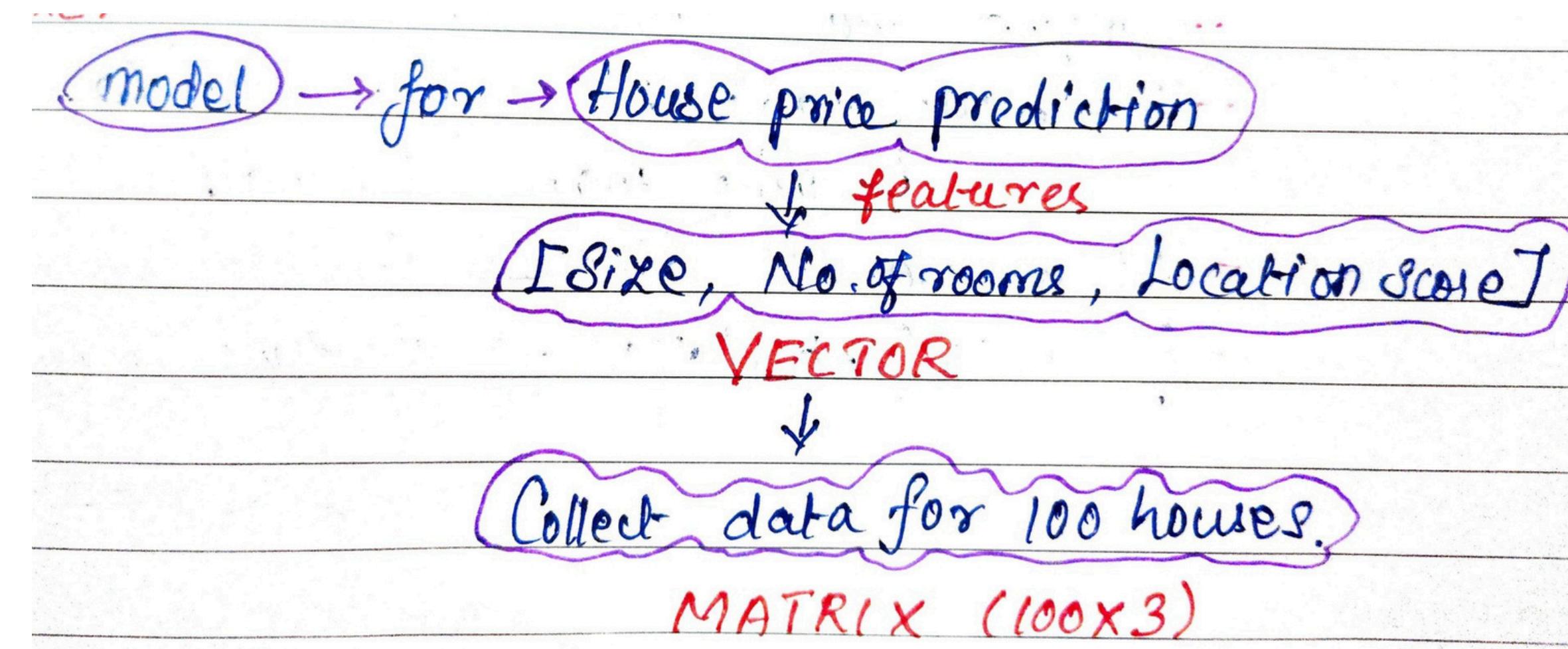
We Work with:

Datasets → Stored as Matrices

Features of data → represented as a vector

Model calculations → done using matrix operations

# EXAMPLE



# WHAT IS “WEIGHTS”

A number that tells the model how important a feature is when making a prediction

Size (sq)	No. of rooms	Location score	→ [1000, 2, 8]
1000	2	8	Input Vector

Now, the model applies weights to each of these features:

weights : [0.5, 10000, 20000]

Next step is :

$$= (1000 \times 0.5) + (2 \times 10000) + (8 \times 20000)$$

$$= 500 + 20000 + 160000$$

$$= £ 180,500 - \text{predicted price}$$

each weight tells how much that feature contributes to the final prediction.

real-world Analogy :

final exam mark = (Assignment  $\times 20\%$ ) +  
(Quiz  $\times 30\%$ ) +  
(project  $\times 50\%$ )

20%, 30%, 50% - weights .

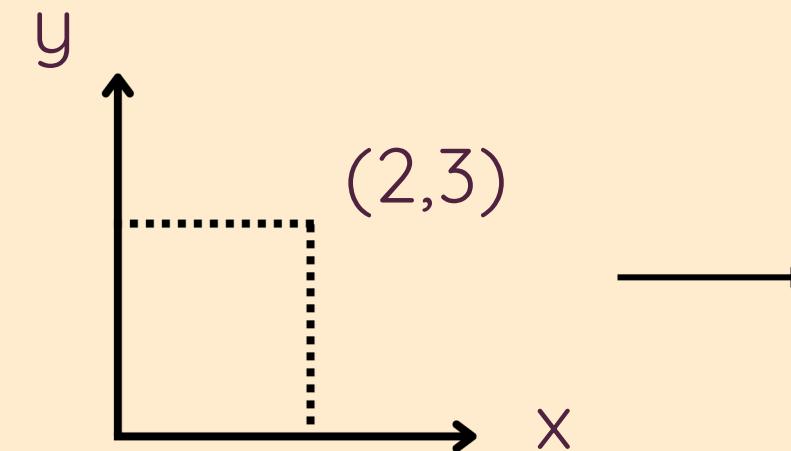
# IN MACHINE LEARNING:

- Weights are learned automatically during training.
- The model adjusts weights so that predictions become more accurate.
- If a weight is small → the feature is less important.
- If a weight is large → the feature is very important.

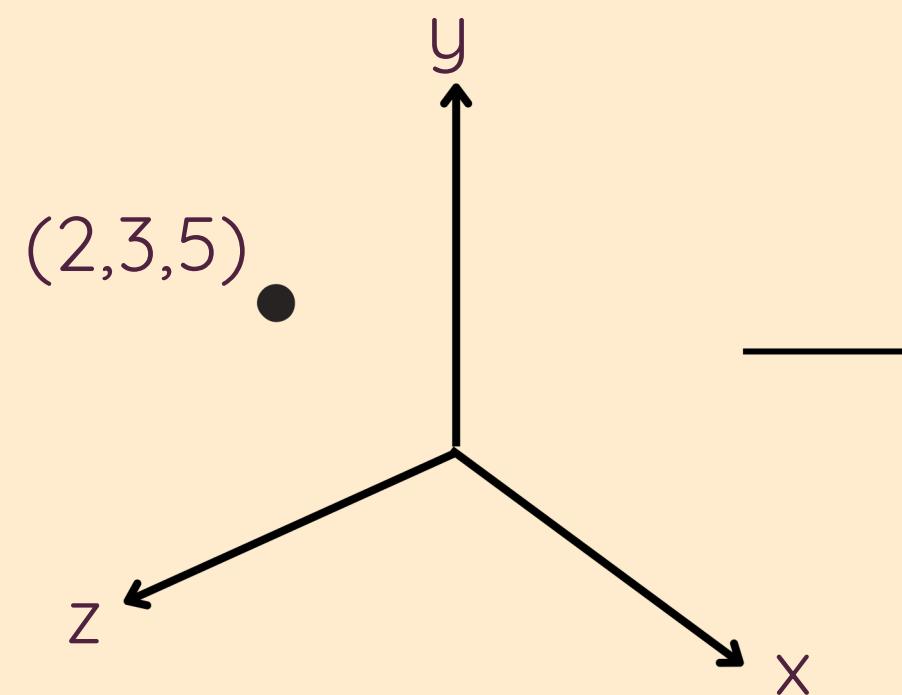
Data = matrix  
feature = vector  
Calculations = matrix operations  
ML = built on linear algebra



# HIGH DIMENSIONS INTERPRETATION

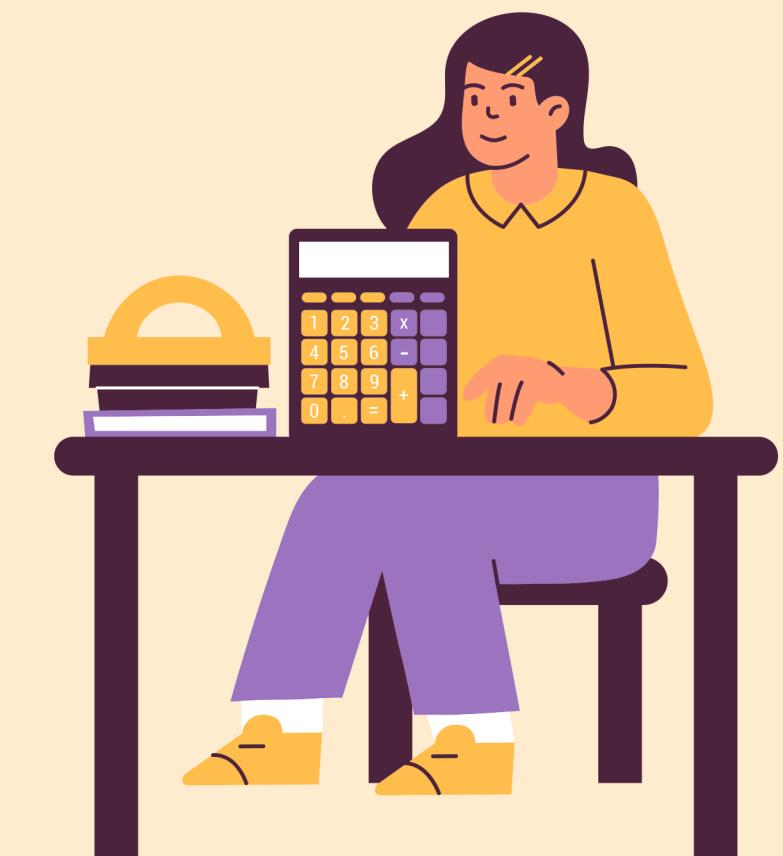


$(2,3) \rightarrow$  2 dimension vector



$(2,3,5) \rightarrow$  3 dimension vector

$(2,3,6,7,8,1) \rightarrow$  5 dimension vector



# SCALARS

Single numerical values.

ML: It is used to adjust things like weights in a model or the learning rate during training.

# OPERATIONS IN LINEAR ALGEBRA

## 1. ADDITION & SUBTRACTION

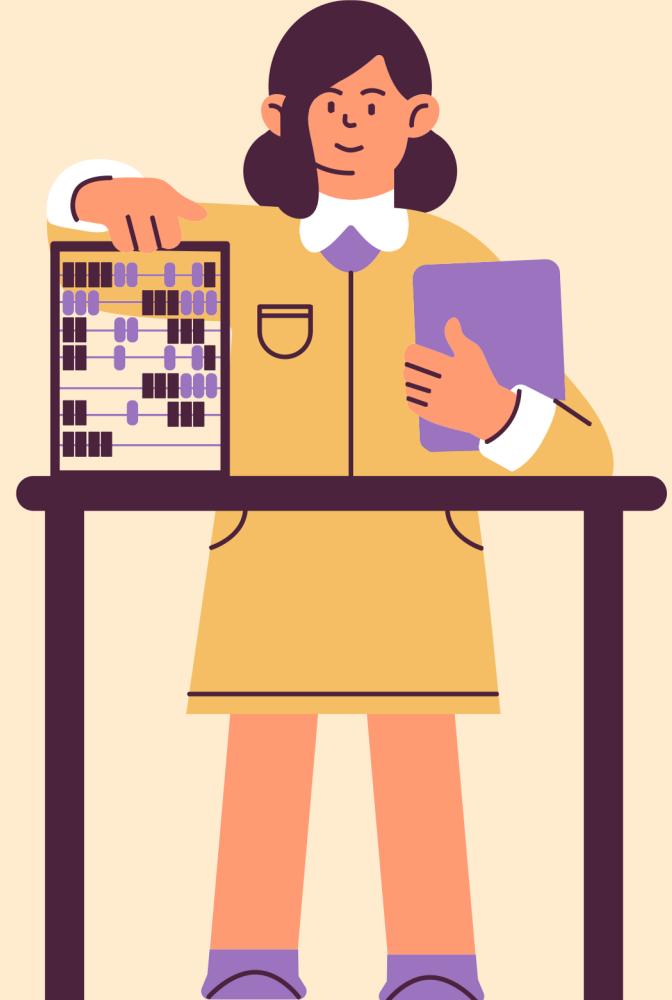
$$u = \begin{bmatrix} 2 \\ -1 \\ 4 \end{bmatrix}, v = \begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix}$$

addition  $\Rightarrow u + v = \begin{bmatrix} 2 \\ -1 \\ 4 \end{bmatrix} + \begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix}$

$$\therefore \begin{bmatrix} 2+3 \\ -1+0 \\ 4+2 \end{bmatrix} = \begin{bmatrix} 5 \\ -1 \\ 6 \end{bmatrix}$$

Subtraction  $\Rightarrow u - v = \begin{bmatrix} 2 \\ -1 \\ 4 \end{bmatrix} - \begin{bmatrix} 3 \\ 0 \\ 2 \end{bmatrix}$

$$= \begin{bmatrix} 2-3 \\ -1-0 \\ 4-2 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix}$$



# OPERATIONS IN LINEAR ALGEBRA

## 1. SCALAR MULTIPLICATION

Scalar  $k = 3$       Vector  $v = \begin{bmatrix} 2 \\ -1 \\ 4 \end{bmatrix}$

$$k \cdot v = 3 \cdot \begin{bmatrix} 2 \\ -1 \\ 4 \end{bmatrix} = \begin{bmatrix} 3 \cdot 2 \\ 3 \cdot -1 \\ 3 \cdot 4 \end{bmatrix} = \begin{bmatrix} 6 \\ -3 \\ 12 \end{bmatrix}$$



# OPERATIONS IN LINEAR ALGEBRA

## 1. DOT PRODUCT & CROSS PRODUCT

Dot product (Scalar product) → row vector

$$u = [u_1, u_2, u_3] \quad v = [v_1, v_2, v_3]$$

$$\text{dot product} \rightarrow u \cdot v = u_1 v_1 + u_2 v_2 + u_3 v_3$$

Cross product (Vector product)

$$u \times v = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} \times \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$

$$= \begin{bmatrix} u_2 v_3 - u_3 v_2 \\ u_3 v_1 - u_1 v_3 \\ u_1 v_2 - u_2 v_1 \end{bmatrix}$$



# TRANSPOSE

- flip the matrix over its diagonal.

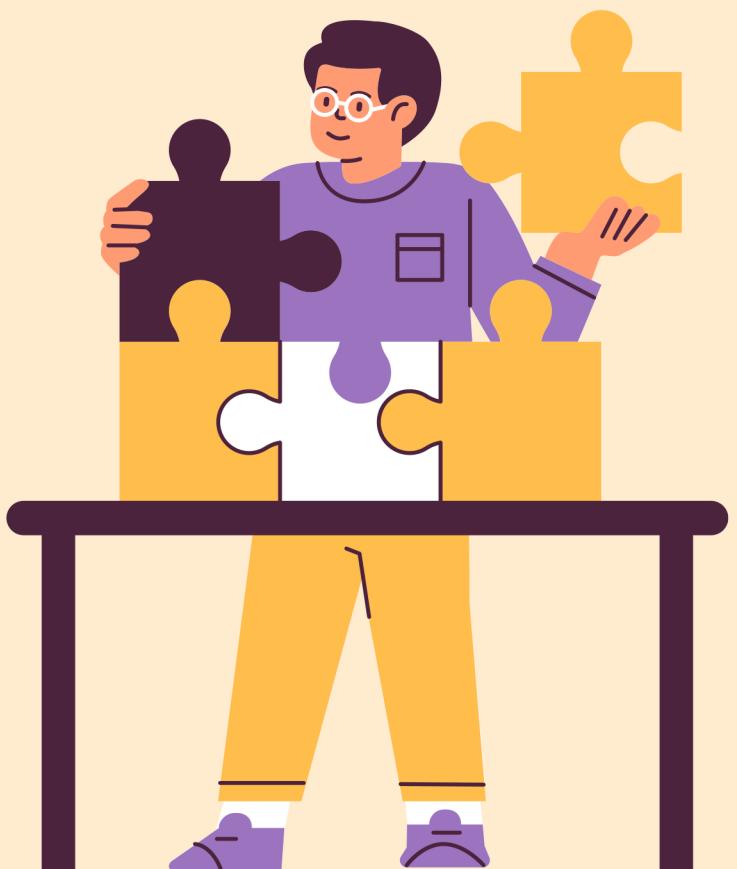
Rows become Columns and Columns become rows.

eg:

$$A = \begin{bmatrix} [1, 2, 3], \\ [4, 5, 6] \end{bmatrix} \rightarrow 2 \times 3$$

rows  
columns

$$A^T = \begin{bmatrix} [1, 4], \\ [2, 5], \\ [3, 6] \end{bmatrix} \rightarrow 3 \times 2$$



Thank  
you

