

R/K: 5

Part A:

- 1 Each student's performance across subjects can be represented as a vector:

$$S_i = (\text{Math, Physics, chemistry, biology})$$

Thus, every student is a 5-d vector $\in \mathbb{R}^5$

→ Each component corresponds to one subject score.

→ The vector captures overall academic performance.

2 Norm 1 Norm 2 of vector

→ Norm: 1 (Manhattan norm) represents the total academic score across all subjects:

$$\|S_i\|_1 = |x_1| + |x_2| + \dots + |x_5|$$

→ Norm: 2 (Euclidean norm) represents magnitude of a student's performance vector:

$$\|S_i\|_2 = \sqrt{x_1^2 + x_2^2 + \dots + x_5^2}$$

It reflects the overall strength of a student's performance.

A norm is a function that measures the magnitude or length of vector

3 Dot Product & angle between 2 student vector

→ The dot product measures similarity between two students' performance:

$$S_i \cdot S_j$$

→ The angle between vectors indicates how similar or different 2 students are:

Small angle: similar performance pattern

Large angle: different strengths & weakness

→ This is useful for comparing student performance profiles.

→ Angle betⁿ 2 vectors = $\cos \theta = \frac{x \cdot y}{\|x\| \|y\|}$

$$\frac{x \cdot y}{\|x\| \|y\|}$$

4 Cross Product:-

→ When considering only 3 subjects:

Maths, Physics, Chemistry

→ The cross product produces a vector orthogonal to both student vectors in 3D space

→ It highlights independent variation between two students' performance in those subjects.

5 Projection of one vector in another:

→ vector projector represents how much one student's performance aligns with another's:

→ Projection of student A onto student B shows:

The portion of A's performance explained by B's performance pattern

→ This is useful for performance benchmarking.

Part B.

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6 Matrix addition & multiplication:

→ The dataset can be represented as:

$$X = \begin{bmatrix} \text{student 1} \\ \text{student 2} \\ \vdots \\ \text{student n} \end{bmatrix}$$

Rows: students

Column: subject

→ It allows:

Addition: Combining datasets

Multiplication: Transformations & similarity analysis

Transpose & inverse

→ Transpose - It swaps rows & columns, useful for covariance computation

Eg: $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$, $A^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$

→ Inverse:- If it exists it helps in solving linear systems.

→ Determinant:- It indicates whether a matrix is invertible & whether the data has linear dependency

Eg: $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$, $|A| = (1 \times 4) - (2 \times 3)$

→ $|A| = -2$

Part C:

8 → Exp line, plane & hyperplane:

→ Line (1D):- It defines performance in one subject.

→ Plane (2D):- It defines performance in 2 subjects.

→ Hyperplane:- full dataset with 5 subject

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9. Show how dimensionality increases.
from 2D \rightarrow 3D

\rightarrow 2 subjects \rightarrow 2D plane

3 subjects \rightarrow 3D plane

5 subjects \rightarrow 5D hyperplane

\rightarrow higher dimensions allows richer representation but reduce visual interpretability, motivating dimensionality reduction.