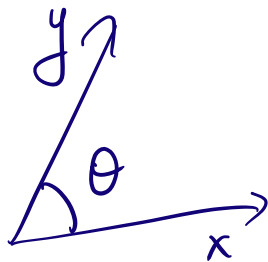


Lecture 22

- Vector projections
- Span
- Bases
- Gram-Schmidt Process

$$x^T y = x \cdot y = \|x\| \|y\| \cos \theta$$



Vector correlation:

$$\cos(\theta) = \frac{x^T y}{\|x\| \|y\|} = r$$

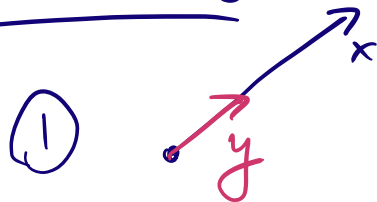
vector correlation

$$x^T y = \sum_{i=1}^N x_i y_i$$

$$\begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = x_1 y_1 + x_2 y_2 + x_3 y_3$$

$$r = \frac{\sum_i x_i y_i}{\sqrt{\sum_i x_i^2} \sqrt{\sum_i y_i^2}}$$

extreme cases

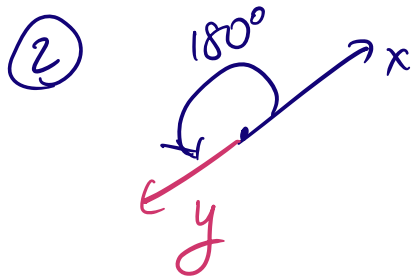


$$\vec{y} = c \cdot \vec{x}$$

$$c > 0$$

$$\theta = \angle(\vec{x}, \vec{y}) = 0^\circ$$

$$r = \cos(0^\circ) = 1$$

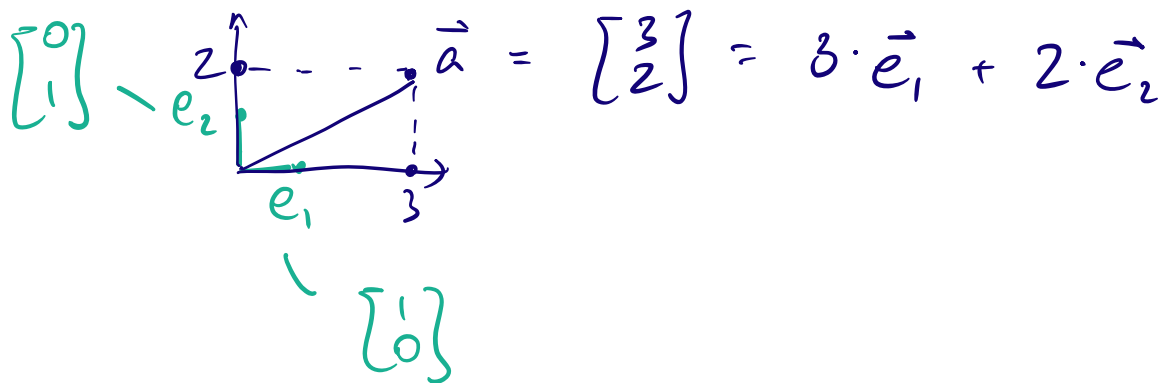


$$\vec{y} = c \cdot \vec{x}$$

$$c < 0$$

$$\theta = \angle(x, y) = 180^\circ$$

$$r = \cos(180^\circ) = -1$$

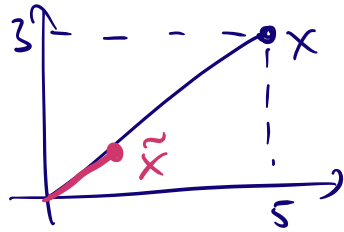


Any vector x with a norm of 1
is a unit vector

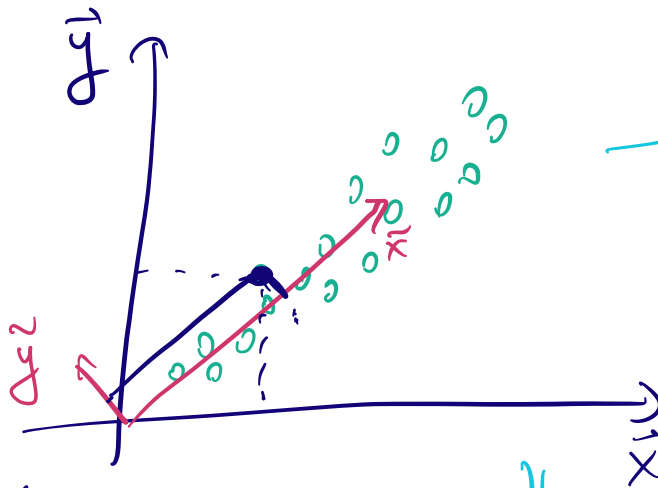
$$\vec{x} = \begin{bmatrix} 5 \\ 3 \end{bmatrix} \Rightarrow \tilde{x} = \frac{\vec{x}}{\|\vec{x}\|}$$

$$\|\tilde{x}\| = 1$$

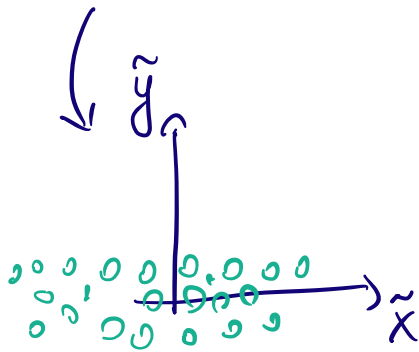
$$\|\tilde{x}\| = \left\| \frac{\vec{x}}{\|\vec{x}\|} \right\| = \frac{\|\vec{x}\|}{\|\vec{x}\|} = 1$$



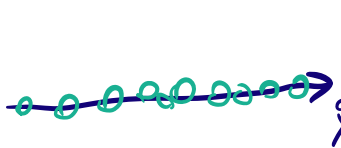
Why project?



→ mean subtraction

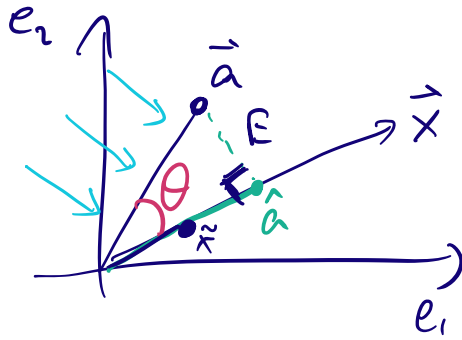


① uncorrelated data for feature extraction



② Dimensionality Reduction

Projection



$$\hat{a} = \text{proj}_x \vec{a}$$

\equiv projection of \vec{a} onto \vec{x}

$$\cos(\theta) = \frac{x^T a}{\|x\| \|a\|}$$

$$\|\hat{a}\| = \|a\| \cdot \cos(\theta) = \|a\| \frac{x^T a}{\|x\| \|a\|}$$

$$\tilde{x} = \frac{\vec{x}}{\|\vec{x}\|}$$

$$\Rightarrow \hat{a} = \|\hat{a}\| \tilde{x} = \cancel{\|a\|} \frac{x^T a}{\|x\| \cancel{\|a\|}} \cdot \frac{x}{\|x\|}$$

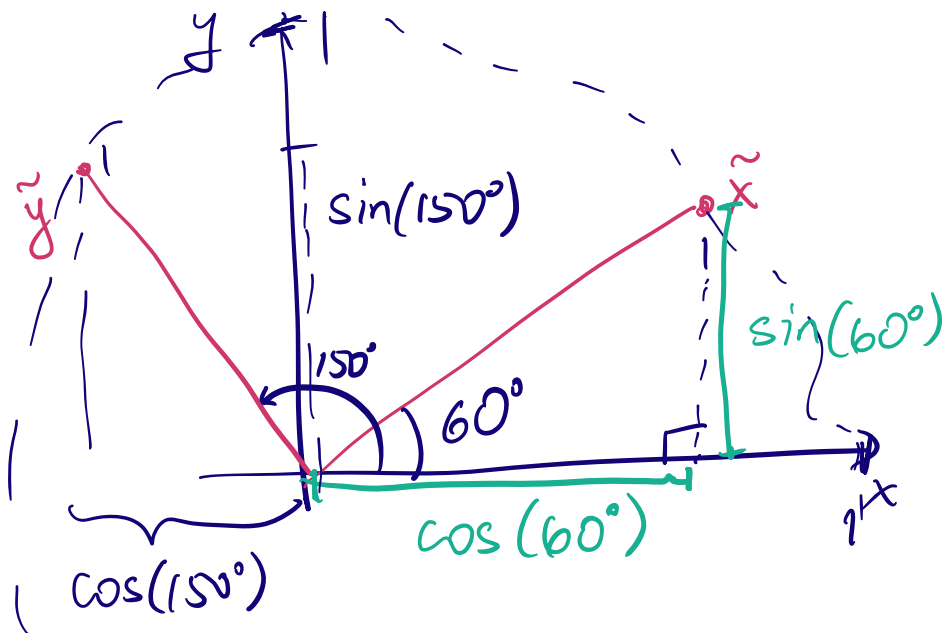
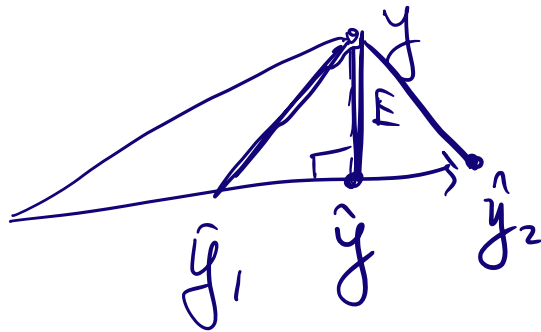
$$\hat{a} = \text{proj}_x \vec{a} = \frac{x^T a}{\|x\|} \cdot \frac{x}{\|x\|}$$

$$= \underbrace{\tilde{x}^T a}_{\text{displacement along the direction } \tilde{x}} \cdot \tilde{x}$$

displacement along the direction \tilde{x}

$$\|E\| = \|y\| \cdot \sin \theta$$

The perpendicular direction of projection is the one that minimizes the error of projection



Rotation

$$\tilde{x} = \begin{bmatrix} \cos(60^\circ) \\ \sin(60^\circ) \end{bmatrix} = \begin{bmatrix} 1/2 \\ \sqrt{3}/2 \end{bmatrix} \Rightarrow \|\tilde{x}\| = 1$$

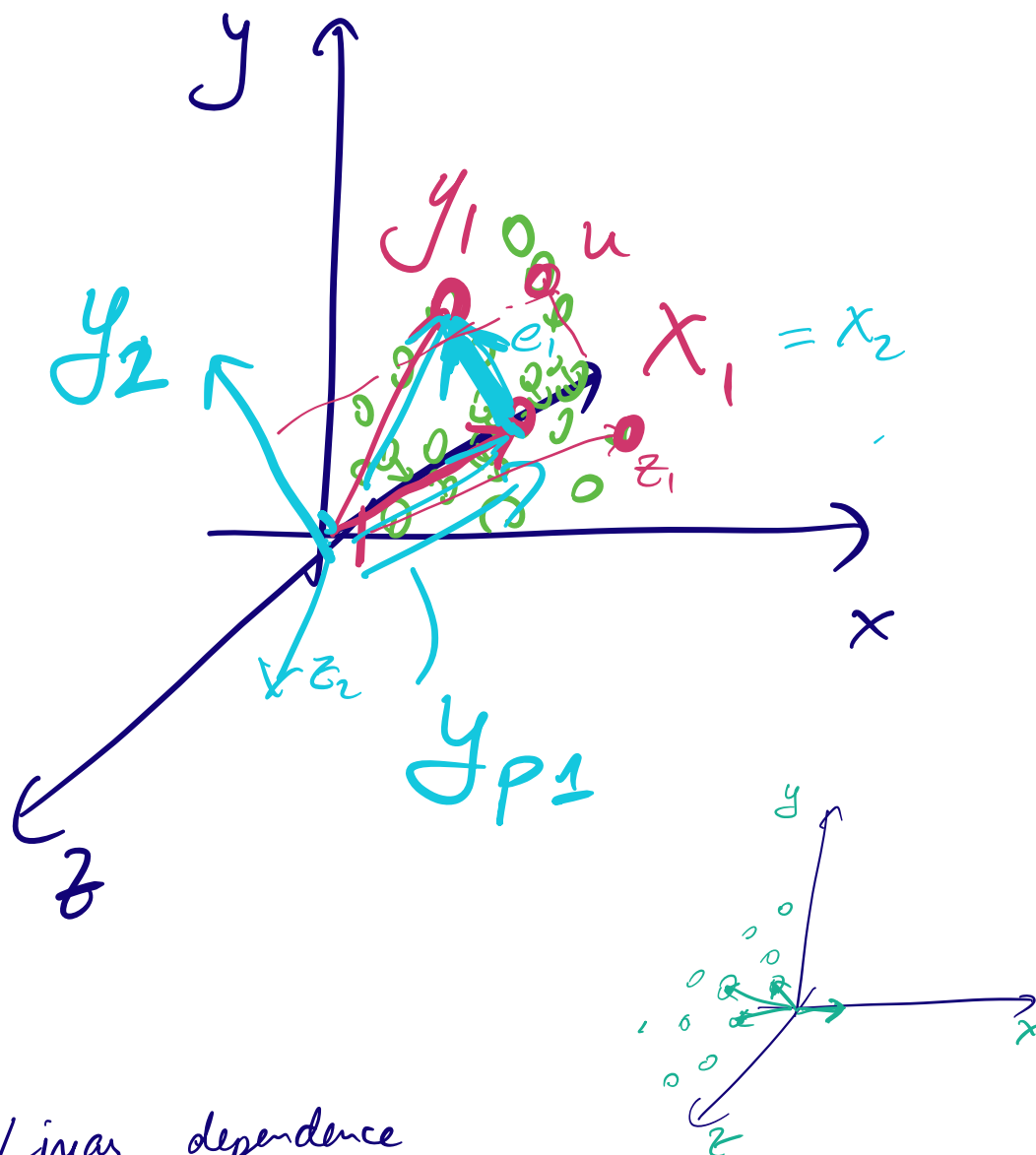
$$\tilde{y} = \begin{bmatrix} \cos(150^\circ) \\ \sin(150^\circ) \end{bmatrix} = \begin{bmatrix} -\sqrt{3}/2 \\ 1/2 \end{bmatrix} \Rightarrow \|\tilde{y}\| = 1$$

If $\|\tilde{x}\| = 1$, $\|\tilde{y}\| = 1$ and $\tilde{x}^T \tilde{y} = 0$,

then \tilde{x} & \tilde{y} are said to be

Orthonormal

Gram-Schmidt Process



Linear dependence

a_0, \dots, a_{k-1} are linearly dependent

if \exists non-zero constants $\beta_0, \dots, \beta_{k-1}$
 there exists \uparrow Such that:

$$\beta_0 a_0 + \beta_1 a_1 + \dots + \beta_{k-1} a_{k-1} = 0$$

$$a_0 = - \left(\frac{\beta_1}{\beta_0} a_1 + \dots + \frac{\beta_{k-1}}{\beta_0} a_{k-1} \right)$$

$$\mathbb{R}^3 : \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Assume: that they are linearly dependent:

$$\beta_0 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \beta_1 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + \beta_2 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} \beta_0 &\neq 0 \\ \beta_1 &\neq 0 \\ \beta_2 &\neq 0 \end{aligned}$$

$$\text{However } \begin{matrix} \rightarrow \\ \rightarrow \\ \rightarrow \end{matrix} \begin{bmatrix} \beta_0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ \beta_1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} \Rightarrow \begin{matrix} \rightarrow \beta_0 = 0 \\ \rightarrow \beta_1 = 0 \\ \rightarrow \beta_2 = 0 \end{matrix} &\Rightarrow \text{Goes against our} \\ &\text{assumptions for linear} \\ &\text{dependence.} \end{aligned}$$

\Rightarrow Thus, these 3 vectors are in fact
linearly independent

Gram-Schmidt Process

⇒ produces a set of orthonormal vectors.

Orthogonal vectors are linearly independent

Proof:

Assume that 2 orthogonal vectors \vec{u}, \vec{v} are not linearly independent.

$$\vec{u}^T \vec{v} = \vec{u} \cdot \vec{v} = 0$$

→ $\exists \{\beta_0, \beta_1\} \neq \{0, 0\}$ s.t.:

$$\beta_0 \cdot \vec{u} + \beta_1 \cdot \vec{v} = \vec{0}$$

Left-multiply by \vec{u}^T on both sides:

$$\beta_0 \cdot \vec{u}^T \vec{u} + \beta_1 \underbrace{\vec{u}^T \vec{v}}_{=0} = \underbrace{\vec{u}^T \vec{0}}_0$$

$$\Rightarrow \beta_0 \cdot \vec{u}^T \vec{u} = 0$$

$$\Rightarrow \beta_0 \cdot \|\vec{u}\|^2 = 0$$

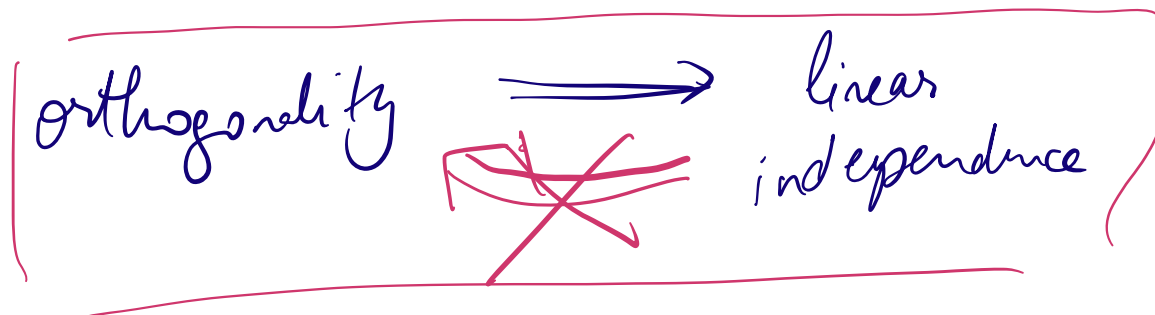
Either $\beta_0 = 0$ or $\|\vec{u}\|^2 = 0$

• However, the norm of $\|\vec{u}\|^2$ will only be 0 iff $\vec{u} = \vec{0}$

So, $\beta_0 = 0 \Rightarrow$ However, $\beta_0 \neq 0$

Thus, \vec{u} & \vec{v} are linearly independent.

So, orthogonal vectors are linearly independent.



$$u = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad v = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$

Assume that they are linearly dependent.

$$\exists \{\beta_1, \beta_2\} \neq \{0, 0\} :$$

$$\beta_1 \vec{u} + \beta_2 \vec{v} = \vec{0}$$

$$\Leftrightarrow \beta_1 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + \beta_2 \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Leftrightarrow \begin{matrix} \rightarrow \\ \rightarrow \\ \rightarrow \end{matrix} \begin{bmatrix} \beta_1 \\ \beta_1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ \beta_2 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Leftrightarrow \left. \begin{matrix} \beta_1 = 0 \\ \beta_1 + \beta_2 = 0 \\ \beta_2 = 0 \end{matrix} \right\} \begin{matrix} \text{violates the condition} \\ \beta_1 \neq 0, \beta_2 \neq 0 \\ \Rightarrow \vec{u} \text{ \& \; } \vec{v} \text{ are linearly } \underline{\text{independent}} \end{matrix}$$

\vec{u} & \vec{v} are linearly independent ✓

But are they orthogonal?

$$\vec{u}^T \vec{v} = \begin{bmatrix} 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = 1 \cdot 0 + 1 \cdot 1 + 0 \cdot 1 \\ = 1 \neq 0$$

\therefore They are not orthogonal

Identity matrix

$$I_k = \begin{bmatrix} 1 & & 0 \\ & 1 & \\ 0 & & \ddots \end{bmatrix} \left. \vphantom{\begin{bmatrix} 1 & & 0 \\ & 1 & \\ 0 & & \ddots \end{bmatrix}} \right\} k \text{ rows}$$

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$