

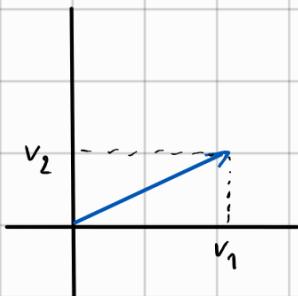
inner/dot product $\underline{u}, \underline{v} \in \mathbb{R}^n$

$$\underline{u} \bullet \underline{v} = \underline{u}^T \underline{v} = \begin{bmatrix} u_1 & \dots & u_m \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_m \end{bmatrix} = u_1 v_1 + \dots + u_m v_m$$

$$i \begin{bmatrix} A \\ \vdots \\ B \end{bmatrix} \begin{bmatrix} j \\ | \\ \vdots \\ c \end{bmatrix} = i \begin{bmatrix} \dots \\ c \end{bmatrix}$$

$c_{ij} = \text{row}_i(A)^T \bullet \text{col}_j(B)$

Length of a vector



$$\underline{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$\text{length} = \sqrt{v_1^2 + v_2^2}$$

$$= \sqrt{\underline{v} \bullet \underline{v}}$$

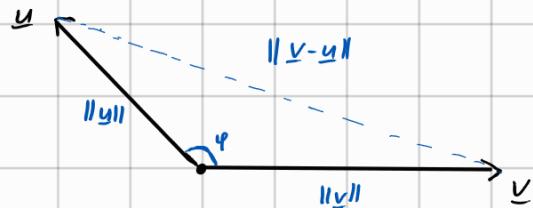
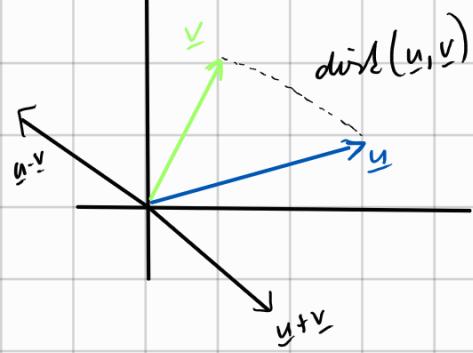
in \mathbb{R}^3 :

$$\underline{v} = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$
$$\text{length} = \sqrt{v_1^2 + v_2^2 + v_3^2}$$
$$= \sqrt{\underline{v} \bullet \underline{v}}$$

length/norm of a vector \underline{v} is defined by

$$\|\underline{v}\| = \sqrt{\underline{v} \bullet \underline{v}}$$

Distance between 2 vectors



law of cosine

$$\begin{aligned}
 \|\underline{v} - \underline{u}\|^2 &= \|\underline{u^2}\| + \|\underline{v^2}\| - 2 \cdot \|\underline{u}\| \cdot \|\underline{v}\| \cdot \cos \varphi \\
 (\underline{v} - \underline{u}) \cdot (\underline{v} - \underline{u}) &= \\
 \underline{v} \cdot (\underline{v} - \underline{u}) - \underline{u} \cdot (\underline{v} - \underline{u}) &= \\
 \underline{v} \cdot \underline{v} - \cancel{\underline{v} \cdot \underline{u}} - \cancel{\underline{u} \cdot \underline{v}} + \underline{u} \cdot \underline{u} &= \\
 -2 \cancel{\underline{u} \cdot \underline{v}} + \cancel{\|\underline{v}\|^2} + \cancel{\|\underline{u}\|^2} &
 \end{aligned}$$

$$\text{so, } \underline{u} \cdot \underline{v} = \|\underline{u}\| \|\underline{v}\| \cos \varphi$$

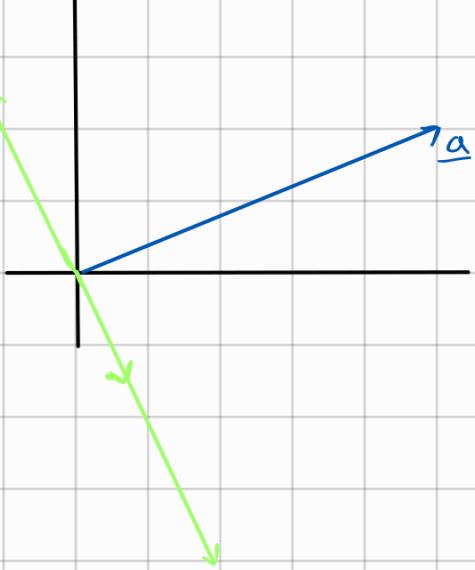
(note $\cos \varphi = 0$ iff
 $\varphi = \frac{\pi}{2} + 2k\pi \quad k \in \mathbb{R}$
 $= \frac{3\pi}{2} + 2k\pi$)

so, if $\underline{u} \neq \underline{0}$ and $\underline{v} \neq \underline{0}$, then $\underline{u} \cdot \underline{v} = 0 \Leftrightarrow \varphi \in \{\frac{\pi}{2}, \frac{3\pi}{2}\}$
 $\Leftrightarrow \underline{u} \text{ and } \underline{v} \text{ are perpendicular}$

Two vectors are orthogonal ($\underline{u} \perp \underline{v}$) if $\underline{u} \cdot \underline{v} = 0$

(Ex) $2x_1 + x_2 = 0$ $\underline{a} \cdot \underline{x} = 0$ with $\underline{a} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$

solution set $2x_1 + x_2 = 0 \Rightarrow x_1 = -\frac{1}{2}x_2$
 $\Rightarrow \underline{x} = x_2 \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix} = x_2 \begin{bmatrix} -1 \\ 2 \end{bmatrix}$



sol red: $x_2 \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

indeed all vectors orthogonal
to \underline{a}

In general: a hyperplane through the origin (homo. eq.)

$$a_1x_1 + \dots + a_nx_n = 0 \quad \text{or} \quad \underline{a} \cdot \underline{x} = 0$$

↓

all vectors $\underline{x} \in \mathbb{R}^m$
that are orthogonal
to \underline{a}

- * in \mathbb{R}^2 : you get a line
- * in \mathbb{R}^3 : you get a plane
- * in \mathbb{R}^n : you get a hyperplane (subspace of $\dim n-1$)

Recall the Null space of a matrix $A_{m \times n}$

$$\text{Null}(A) = \{ \underline{x} \in \mathbb{R}^m : A\underline{x} = \underline{0} \}$$

$$\begin{bmatrix} -r_1 - \\ -r_2 - \\ \vdots \\ -r_m - \end{bmatrix} \begin{bmatrix} | \\ \underline{x} \\ | \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\underline{r}_1^T \cdot \underline{x} = 0$$

for every $\underline{x} \in \text{Null}(A)$ is

11.1 and 11.2 up

$$\left\{ \begin{array}{l} \underline{r_2} \cdot \underline{x} = 0 \\ \vdots \\ \underline{r_n} \cdot \underline{x} = 0 \end{array} \right.$$

orthogonal to each of the rows
of A

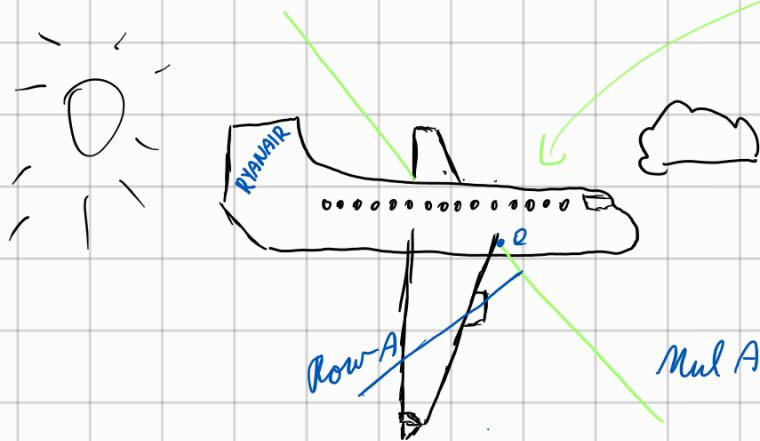
$$(\underline{c_1} \underline{r_1}^T + \underline{c_2} \underline{r_2}^T + \dots + \underline{c_m} \underline{r_m}^T) \underline{x} =$$

$$c_1 (\underline{r_1}^T \underline{x}) + c_2 (\underline{r_2}^T \underline{x}) + \dots + c_m (\underline{r_m}^T \underline{x}) = 0 + 0 + \dots + 0$$

So every vector in $\text{Nul}(A)$ is orthogonal to every vector in $\text{Row}(A)$

$$\Rightarrow \text{Nul}(A) \perp \text{Row}(A)$$

plane in \mathbb{R}^3



$$\text{Nul}(A) \perp \text{Row}(A)$$

$$\begin{matrix} \downarrow & \downarrow \\ \text{in } \mathbb{R}^m & \text{in } \mathbb{R}^m \\ \dim = \# \text{ free var} & \dim = \# \text{ pivot} \end{matrix}$$

$$\sum \dim = m$$

and since $\text{Col}(A) = \text{Row}(A^T)$ we also have
 $\text{Nul}(A^T) \perp \text{Col}(A)$

W : subspace of \mathbb{R}^m

W^\perp (" W perpendicular") : orthogonal complement of

all vectors in \mathbb{R}^m shall be orthogonal to W

$$(\text{Row}(A))^\perp = \text{Null}(A) \quad \text{and} \quad (\text{Col}(A))^\perp = \text{Null}(A^T)$$

W^\perp is also a subspace of \mathbb{R}^m

In general $\dim(W) + \dim(W^\perp) = m$

$\{\underline{v}_1, \dots, \underline{v}_m\}$ is an orthogonal set of $\underline{v}_i \cdot \underline{v}_j = 0$ for all $i \neq j$

Theorem: If $S = \{\underline{v}_1, \dots, \underline{v}_m\}$ is an orthogonal set and $0 \notin S$ then S is lin. indep. and thus S forms a basis for $\text{Span}\{\underline{v}_1, \dots, \underline{v}_m\}$

Orthogonal basis for a subspace W of \mathbb{R}^m : it is a basis of W and orthogonal set

$\{\underline{u}_1, \dots, \underline{u}_k\}$ orth. basis for W . Let $y \in W$.

$$y = c_1 \underline{u}_1 + \dots + c_k \underline{u}_k$$

What are the weights c_1, \dots, c_k ?

$$y \cdot \underline{u}_1 = (c_1 \underline{u}_1 + \dots + c_k \underline{u}_k) \cdot \underline{u}_1 = c_1 \underline{u}_1 \cdot \underline{u}_1 + \cancel{c_2 \underline{u}_2 \cdot \underline{u}_1} + \cancel{\dots} + \cancel{c_k \underline{u}_k \cdot \underline{u}_1}$$

$$= c_1 \underline{u}_1 \cdot \underline{u}_1$$

$$y \cdot \underline{u}_1$$

$$y \cdot \underline{u}_2$$

$$\Rightarrow c_1 = \frac{1}{\underline{u}_1 \cdot \underline{u}_1} \quad c_2 = \frac{1}{\underline{u}_2 \cdot \underline{u}_2} \quad \text{etc.}$$

So, it's easy to find the weights " \parallel "
 (non-orthogonal basis: solving an SLE)

$\{\underline{v}_1, \dots, \underline{v}_k\}$ is an orthonormal set if it is orthogonal unit vectors
 \hookrightarrow vectors of length 1

(Ex) $\left\{ \begin{bmatrix} 2/\sqrt{5} \\ -1/\sqrt{5} \end{bmatrix}, \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix} \right\}$ forms an orthonormal basis for \mathbb{R}^2

$\left\{ \begin{bmatrix} 2 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\}$ forms an orthogonal basis for \mathbb{R}^2

$\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ forms an orthonormal basis for \mathbb{R}^2

$\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\}$ forms a basis for \mathbb{R}^2

Now to see whether $\{\underline{v}_1, \dots, \underline{v}_m\}$ is orthogonal or orthonormal?

$$A = \begin{bmatrix} | & | & | \\ \underline{v}_1 & \underline{v}_2 & \cdots & \underline{v}_k \\ | & | & | \end{bmatrix}$$

Compute $A^T A = \begin{bmatrix} \underline{v}_1^T \\ \underline{v}_2^T \\ \vdots \end{bmatrix} \begin{bmatrix} | & | & | \\ \underline{v}_1 & \underline{v}_2 & \cdots & \underline{v}_k \\ | & | & | \end{bmatrix} = \begin{bmatrix} \underline{v}_1 \cdot \underline{v}_1 \\ \underline{v}_2 \cdot \underline{v}_2 \\ \vdots \\ \underline{v}_k \cdot \underline{v}_k \end{bmatrix}$

$$\left[\begin{array}{c} \underline{v_1} \\ \vdots \\ \underline{v_k} \end{array} \right] \left[\begin{array}{ccc} \top & \top & \top \end{array} \right] = \left[\begin{array}{ccc} \top & \top & \top \end{array} \right]$$

$\{\underline{v}_1, \dots, \underline{v}_k\}$ is orthogonal $\Leftrightarrow A^T A$ is diagonal

$\{\underline{v}_1, \dots, \underline{v}_k\}$ is orthonormal $\Leftrightarrow A^T A$ is identity matrix

A square A is an orthogonal matrix

$$\Leftrightarrow A^T A = I_n \Leftrightarrow A^{-1} = A^T$$

Watch out: an orthogonal matrix has orthonormal cols

Properties: $T: \underline{x} \mapsto A\underline{x}$, where A is orthogonal:

$$* |A\underline{x}| = \sqrt{(\underline{A}\underline{x}) \cdot (\underline{A}\underline{x})} = \sqrt{(\underline{A}\underline{x})^T (\underline{A}\underline{x})} = \\ = \sqrt{\underline{x}^T A^T A \underline{x}} = \sqrt{\underline{x}^T I_n \underline{x}} = \sqrt{\underline{x}^T \underline{x}}$$

