

\mathbb{R}^n interlude

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Here are some concepts from vector calculus that will be useful to understand the methods we will study

$$\mathbb{R}^n = \{ (x_1, x_2, \dots, x_n)^T \mid x_1 \in \mathbb{R}, x_2 \in \mathbb{R} \dots x_n \in \mathbb{R} \}$$

= set of vectors with n real entries.

Example: $\mathbb{R}^2 \equiv \text{plane}$, $\underline{x} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} \in \mathbb{R}^2$

$\mathbb{R}^3 = \text{3d space}$ $\underline{x} = \begin{pmatrix} -1 \\ \frac{\pi}{2} \\ 2 \end{pmatrix} \in \mathbb{R}^3$

etc...

The notion of angle and length in \mathbb{R}^n is given by:
inner product (a.k.a. dot product, scalar prod.)

Def (inner prod in \mathbb{R}^n):

$$(\underline{x}, \underline{y}) = \underline{x}^T \underline{y} = \sum_{i=1}^n x_i y_i = \text{sum of prod of entries.}$$

Example: $\underline{x} = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$, $\underline{y} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$

$$(\underline{x}, \underline{y}) = 2 \times 2 + 2 \times 1 = 6$$

• Length using inner prod:

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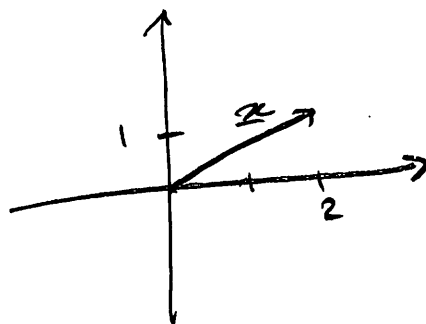
The length of a vector \underline{x} is denoted $\|\underline{x}\|$ and:

$$\|\underline{x}\| = (\underline{x}, \underline{x})^{\frac{1}{2}} = \left(\sum_{i=1}^n x_i \right)^{\frac{1}{2}}$$

Example:

$$\underline{x} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$\|\underline{x}\| = \sqrt{4+1} = \sqrt{5}$$



• Angle using inner prod:

$$(\underline{x}, \underline{y}) = \|\underline{x}\| \|\underline{y}\| \cos \theta, \quad \theta \equiv \text{angle between } \underline{x}, \underline{y}.$$

Given way of computing angles easily in \mathbb{R}^n .

Example: What is angle between $\underline{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ and $\underline{y} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$:

$$\underline{x} \cdot \underline{y} = 1 = \frac{\|\underline{x}\| \|\underline{y}\| \cos \theta}{\sqrt{2} \cdot 1}$$

$$\angle = 45^\circ = \frac{\pi}{4}$$

$$\Rightarrow \cos \theta = \frac{1}{\sqrt{2}} \Rightarrow \theta = \frac{\pi}{4}$$

Def (Orthogonality)

Two vectors are orthogonal (\perp) iff

$$(\underline{x}, \underline{y}) = 0$$

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Theorem (orthogonal basis)

Let $\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n$ be $n \perp$ vectors $\neq 0$,
(i.e. $(\underline{v}_i, \underline{v}_j) = 0$ if $i \neq j$)

Then any vector \underline{x} in \mathbb{R}^n can be expressed as a linear comb. of these vectors.

$$\underline{x} = \sum_{i=1}^n x_i \underline{v}_i = \sum_{i=1}^n \frac{(\underline{x}, \underline{v}_i)}{(\underline{v}_i, \underline{v}_i)} \underline{v}_i$$

Proof.

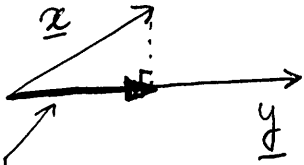
The coeff x_i are determined uniquely since

$$\begin{aligned} (\underline{x}, \underline{v}_j) &= \left(\sum_{i=1}^n x_i \underline{v}_i, \underline{v}_j \right) \\ &= \sum_{i=1}^n x_i (\underline{v}_i, \underline{v}_j) \\ &= x_j (\underline{v}_j, \underline{v}_j) \end{aligned}$$

$$\Rightarrow \left. \begin{array}{l} x_j = \frac{(\underline{x}, \underline{v}_j)}{(\underline{v}_j, \underline{v}_j)} \end{array} \right\} j=1 \dots n.$$

Projection

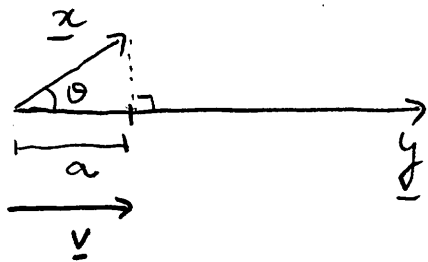
The projection of a vector \underline{x} onto another vector \underline{y} is: the vector \underline{v}



$$\underline{v} = \frac{(\underline{x}, \underline{y})}{(\underline{y}, \underline{y})} \underline{y}$$

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We can check this using our knowledge of projections from geometry and the notion of angle that is defined by the inner product:



length $a = \|\underline{x}\| \cos \theta$

∴ proj vector of \underline{x} onto \underline{y} is:

$$\underline{v} = a \times \frac{\underline{y}}{\|\underline{y}\|}$$

← vector of length 1 in the direction \underline{y} .

$$= \|\underline{x}\| \cos \theta \frac{\underline{y}}{\|\underline{y}\|}$$

$$= \frac{\|\underline{x}\| \|\underline{y}\| \cos \theta}{\|\underline{y}\|^2} \underline{y}$$

$$= \frac{(\underline{x}, \underline{y})}{(\underline{y}, \underline{y})} \underline{y}$$

Example of orthogonal basis theorem in \mathbb{R}^2 :

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$$\text{Let } \underline{v}_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \underline{v}_2 = \begin{pmatrix} -1 \\ 2 \end{pmatrix} \in \mathbb{R}^2.$$

Check that \underline{v}_1 and \underline{v}_2 are \perp and find the coeff α_i in:

$$\sum_{i=1}^2 \alpha_i \underline{v}_i = \alpha_1 \underline{v}_1 + \alpha_2 \underline{v}_2 = \begin{pmatrix} 2 \\ 2 \end{pmatrix} \equiv \underline{x}$$

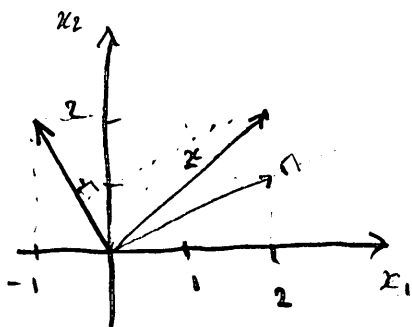
$$(\underline{v}_1, \underline{v}_2) = -2 + 2 = 0 \text{ so } \underline{v}_1 \text{ is } \perp \text{ to } \underline{v}_2.$$

$$\alpha_1 = \frac{(\underline{x}, \underline{v}_1)}{(\underline{v}_1, \underline{v}_1)} = \frac{6}{5}$$

$$\alpha_2 = \frac{(\underline{x}, \underline{v}_2)}{(\underline{v}_2, \underline{v}_2)} = \frac{2}{5}$$

Let's check that these coeff. work:

$$\frac{6}{5} \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \frac{2}{5} \begin{pmatrix} -1 \\ 2 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 12 & -2 \\ 6 & 4 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 10 \\ 10 \end{pmatrix} = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$



Any vector in \mathbb{R}^n is the sum of its projections along n \perp vectors.