

Last time: · normalization of vectors $v \neq 0 \mapsto \frac{v}{\|v\|} \rightarrow$ unit vector

· orthogonal vectors : $u \perp v \Leftrightarrow u \cdot v = 0.$

· Prop: An orthogonal set of nonzero vectors $\{v_1, \dots, v_k\}$
is automatically lin ind.

Pf idea: eg. say $k=3$. $av_1 + bv_2 + cv_3 = 0$

to show, say, $b=0$, take the inner prod with v_2 .

$$\underbrace{av_1 \cdot v_2}_0 + \underbrace{b v_2 \cdot v_2}_{v_2 \cdot v_2 \neq 0} + \underbrace{c v_3 \cdot v_2}_0 = 0 \Rightarrow b=0.$$

Today: · orthogonal and orthonormal bases · orthogonal projections

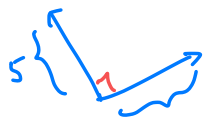
1. Orthogonal basis

Def. An orthogonal basis of \mathbb{R}^n is a basis of \mathbb{R}^n which is an orthogonal set.

An orthonormal basis of \mathbb{R}^n is an orthogonal basis consisting of unit vectors.

Eg. $n=2$. the standard basis $B_1 = \{e_1, e_2\} = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$

len: 5 \leftarrow is an orthonormal basis of \mathbb{R}^2 .



\downarrow normalize

the set $B_2 = \left\{ \begin{bmatrix} 4 \\ 3 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \end{bmatrix} \right\}$ is orthogonal (last lecture).
it has 2 elts, so B_2 is a B_2 is lin ind. basis
and an orthogonal basis. However, it is not orthonormal.



normalizing the elts in B_2 yields an orthonormal basis $B_2' = \left\{ \begin{bmatrix} 4/5 \\ 3/5 \end{bmatrix}, \begin{bmatrix} -3/5 \\ 4/5 \end{bmatrix} \right\}$.

Why are orthogonal bases interesting?

Recall: Given a basis $B = \{v_1, \dots, v_n\}$ of \mathbb{R}^n and any $v \in \mathbb{R}^n$, we can ask for the coordinate vector $[v]_B = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$. Finding $[v]_B$ is usually done by solving the equation $c_1 v_1 + c_2 v_2 + \dots + c_n v_n = v$.

For an orthogonal basis $C = \{v_1, \dots, v_n\}$, we can find the c_i 's and hence $[v]_C$ more directly:

Thm. Let $C = \{v_1, \dots, v_n\}$ be an orthogonal basis of \mathbb{R}^n . Then for all $v \in \mathbb{R}^n$, we have $v = c_1 v_1 + \dots + c_n v_n$

where $c_i = \frac{\langle v, v_i \rangle}{\langle v_i, v_i \rangle}$ } two inner prod.
comp. + one division.
↓

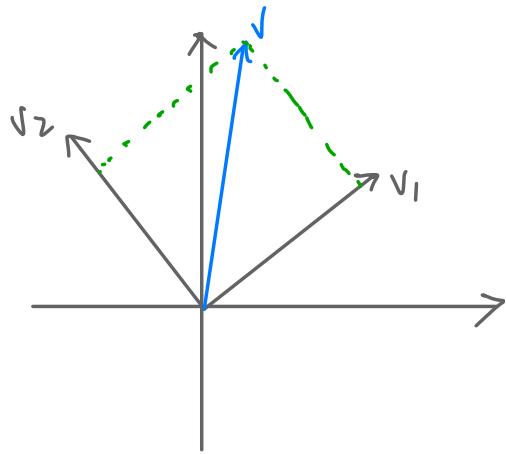
In particular, if C is orthonormal, then $c_i = \langle v, v_i \rangle$. → straight forward.

Pf.: Suppose $v = c_1 v_1 + \dots + c_n v_n$. Let $1 \leq i \leq n$. "Dotting with" v_i on both sides gives

$$\langle v, v_i \rangle = \underbrace{c_1 \langle v_1, v_i \rangle}_0 + \underbrace{c_2 \langle v_2, v_i \rangle}_0 + \dots + \underbrace{c_i \langle v_i, v_i \rangle}_{\text{only nonzero ip.}} + \dots + \underbrace{c_n \langle v_n, v_i \rangle}_0$$

So $\langle v, v_i \rangle = c_i \langle v_i, v_i \rangle$, therefore $c_i = \frac{\langle v, v_i \rangle}{\langle v_i, v_i \rangle}$. \square

Eg. $n=2$, $C = \left\{ \begin{bmatrix} 4 \\ 3 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \end{bmatrix} \right\}$, $v = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$. To find $[v]_C$, we can



(old way), solve $c_1 \begin{bmatrix} 4 \\ 3 \end{bmatrix} + c_2 \begin{bmatrix} -3 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 \\ 6 \end{bmatrix}$. EX. check that $c_1 v_1 + c_2 v_2 = v$.

(new way, since C is orthogonal):

$$c_1 = \frac{\langle \begin{bmatrix} 1 \\ 6 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \end{bmatrix} \rangle}{\langle \begin{bmatrix} 4 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 3 \end{bmatrix} \rangle} = \frac{22}{25}, \quad c_2 = \frac{\langle \begin{bmatrix} 1 \\ 6 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \end{bmatrix} \rangle}{\langle \begin{bmatrix} -3 \\ 4 \end{bmatrix}, \begin{bmatrix} -3 \\ 4 \end{bmatrix} \rangle} = \frac{21}{25}.$$

$[v]_C = \begin{bmatrix} 22/25 \\ 21/25 \end{bmatrix}$

2. Orthogonal projection

Let $\{v_1, \dots, v_n\}$ be an orthogonal basis of \mathbb{R}^n .

The decomp. $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$

has geometric significance. For any $1 \leq i \leq n$, we call the

part/summand $c_i v_i = \frac{\langle v, v_i \rangle}{\langle v_i, v_i \rangle} \cdot v_i$ the projection of v onto

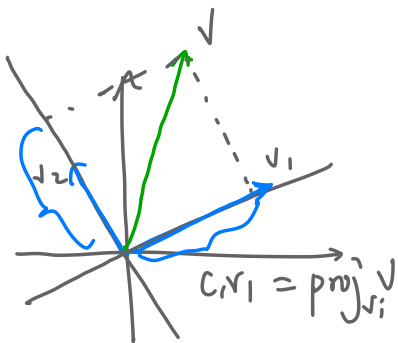
(the line spanned by) v_i , we denote $c_i v_i$ by $\text{proj}_{v_i} v$.

Note: $\forall i, \text{proj}_{v_i} v \in \text{Span}\{v_i\}$.

$\forall i, v - \text{proj}_{v_i} v \perp v_i$.

Ex:

$$c_2 v_2 = \text{proj}_{v_2} v$$



More generally ...

Def. (projection onto subspace) Let $C = \{v_1, v_2, \dots, v_n\}$ be an orthogonal basis of \mathbb{R}^n . Let $y \in \mathbb{R}^n$. Consider the decomp

$$y = c_1 v_1 + \dots + c_n v_n \quad \text{where} \quad c_n = \frac{\langle y, v_i \rangle}{\langle v_i, v_i \rangle}.$$

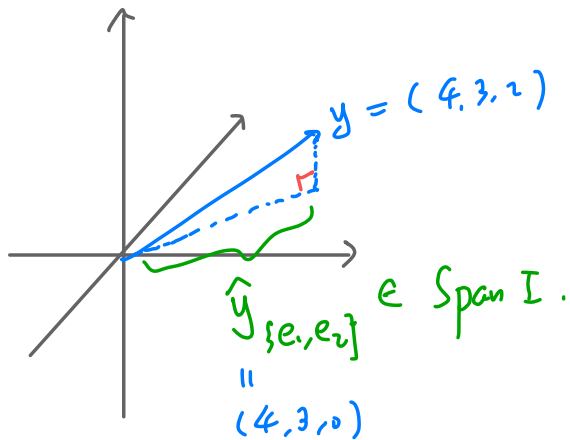
For any subset I of C , the linear combination

$$\hat{y}_I := \sum_{i: v_i \in I} c_i v_i$$

is called the orthogonal projection of y onto the span of I .

(On the previous page, we just took I to be a singleton set $\{v_i\}$.)

Ex. $n=3$. $C = \{e_1, e_2, e_3\}$ $I = \{e_1, e_2\}$. $\rightarrow \text{Span } I = \text{the } xy\text{-plane.}$



$$y = \begin{bmatrix} 4 \\ 3 \\ 2 \end{bmatrix} = \underbrace{4e_1 + 3e_2}_{\hat{y}_{\{e_1, e_2\}}} + \underbrace{2e_3}_{y - \hat{y}_I}$$

Thm. Let C, I, y, \hat{y}_I be as in the previous definition, then

(a). $y - \hat{y}_I \perp \hat{y}_I$

(b). $\forall w \in \text{Span } I$, we have $\|y - w\| \geq \|y - \hat{y}_I\|$.

(Thus, \hat{y}_I is the closest point to y in $\text{Span } I$.)

Ex. Let $C = \left\{ \begin{bmatrix} 1 \\ 6 \end{bmatrix}, \begin{bmatrix} -12 \\ 2 \end{bmatrix} \right\} \subseteq \mathbb{R}^2$.

(1) Verify that C is an orthogonal basis of \mathbb{R}^2 :

→ check $\langle u_1, u_2 \rangle = 0$, note that $|C| = 2$.

(2) Find $[v]_C$ for $v = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$. → use $c_i = \frac{\langle v, u_i \rangle}{\langle u_i, u_i \rangle}$.

(3). Find the closest point to v on the line $\text{Span}(u_1)$.

→ get c_1, c_2 for $v = c_1 u_1 + c_2 u_2$.

the desired point is $c_1 u_1$.

(4). Find the shortest distance between v and a point $u \in \text{Span}(u_1)$.

→ compute $\|v - c_1 u_1\|$.