

# SECTION VI:

## ORTHOGONAL SETS and BASES

We have defined a pair of vectors being orthogonal.

More generally, we would like to define a set of arbitrary size being orthogonal.

## DEFINITION 6.29

The set of  $n$ -vectors  $S$  is **orthogonal** if each pair of vectors in  $S$  is orthogonal; that is,

$$\vec{x} \cdot \vec{y} = 0$$

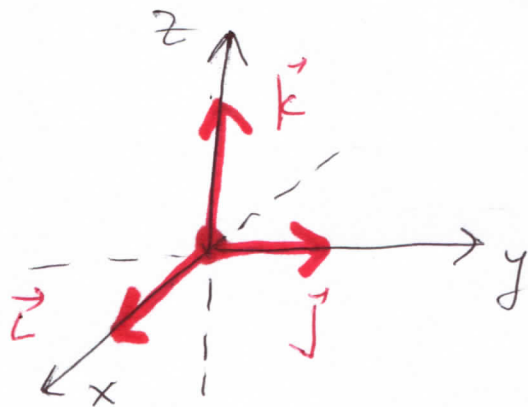
whenever  $\vec{x}$  and  $\vec{y}$  are two different vectors in  $S$ .

## Examples 6.30

$$(a) \{ \vec{i} \equiv (1, 0, 0), \vec{j} \equiv (0, 1, 0), \vec{k} \equiv (0, 0, 1) \}$$

is orthogonal, since

$$\vec{i} \cdot \vec{j} = 0 = \vec{i} \cdot \vec{k} = \vec{j} \cdot \vec{k}$$



$\vec{i}, \vec{j}, \vec{k}$  are popular in physics;  
note that, for any real  $a, b, c$ ,

$$(a, b, c) = a\vec{i} + b\vec{j} + c\vec{k}$$

$$(b) \left\{ \vec{a} = \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \vec{b} = \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \vec{c} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \vec{d} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 7 \\ 1 \end{bmatrix} \right.$$

is orthogonal, since  $0 = \vec{a} \cdot \vec{b} = \vec{a} \cdot \vec{c}$

= ... (6 pairs of dot products)  
to check

So long as we avoid the trivial vector  $\vec{0} \equiv (0, 0, 0, \dots)$ , orthogonality is stronger than linear independence.

## THEOREM 6.31

If  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\} \equiv S$  is an orthogonal set of nontrivial vectors, then  $S$  is linearly independent.

**Proof:** Suppose a linear combination from  $S$  equals  $\vec{0}$ ; that is,

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$$(*) \quad (c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_m \vec{v}_m) = \vec{0},$$

for some real numbers

$$c_1, c_2, \dots, c_m.$$

Recall that linear independence is equivalent to  $(*)$  occurring only when  $c_1 = 0 = c_2 = c_3 = \dots = c_m$ ; that is, only the trivial linear combination equals  $\vec{0}$ .

$(*)$  implies that

$$\begin{aligned} 0 &= \vec{v}_1 \cdot \vec{0} = \vec{v}_1 \cdot (c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots) \\ &= c_1 (\vec{v}_1 \cdot \vec{v}_1) + c_2 (\vec{v}_1 \cdot \vec{v}_2) + \dots \\ &= c_1 \|\vec{v}_1\|^2, \text{ by orthogonality.} \end{aligned}$$

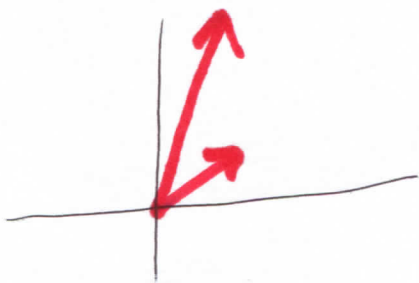
since  $\vec{v}_1$  is nontrivial,

$\|\vec{v}_1\|^2 \neq 0$ , thus  $c_1 = 0$ .

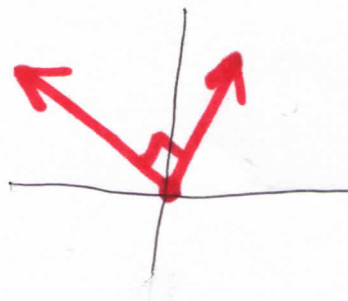
The same argument,  
with  $\vec{v}_j$ ,  $j = 2, 3, \dots, m$  replacing  
 $\vec{v}_1$ , shows that

$c_j = 0$  for  $j = 1, 2, 3, \dots, m$ ,  
as desired.

Here is the picture in  $\mathbb{R}^2$ :



linear  
independence



orthogonality

For a pair of vectors in  $\mathbb{R}^2$ , linear independence means they are not parallel, while orthogonality means they are perpendicular.

### DEFINITION 6.32

If  $W$  is a subspace with  $\dim(W) = p$ , then any set of orthogonal set of  $p$  nontrivial vectors in  $W$  is called an

**orthogonal basis**

for  $W$ .

Note that Theorems 4.56 and 6.31 imply that an orthogonal basis is a basis in the sense of Chapter IV.

Theorems 6.31 and 4.56 show that an orthogonal set is better than a linearly independent set and an orthogonal basis is better than a basis.

Is this superiority substantive? Or is it "so what"?



The remainder of this section will describe three substantial advantages to using orthogonal sets of vectors.

### 6.33. ORTHOGONAL ADVANTAGE (1).

(extended Pythagorean Theorem)

If  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  is an orthogonal set of vectors, then

$$\|\vec{v}_1 + \vec{v}_2 + \dots + \vec{v}_m\|^2 = \|\vec{v}_1\|^2 + \|\vec{v}_2\|^2 + \dots + \|\vec{v}_m\|^2.$$

# Examples 6.34

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(a) See Examples 6.30(b).

For any real  $c_1, c_2, c_3, c_4$ ,

by 6.33,

$$\left\| c_1 \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} + c_4 \begin{bmatrix} 0 \\ 0 \\ 0 \\ 7 \\ 1 \end{bmatrix} \right\|^2 =$$

$$c_1^2 \|(1, 2, 0, 0, 0)\|^2 + c_2^2 \|(-2, 1, 0, 0, 0)\|^2 \\ + c_3^2 \|(0, 0, 1, 0, 0)\|^2 + c_4^2 \|(0, 0, 0, 7, 1)\|^2 =$$

$$5c_1^2 + 5c_2^2 + c_3^2 + 50c_4^2.$$

$$(b) \|(1, -1) + (0, 1)\|^2 = \|(1, 0)\|^2 = 1, \text{ but}$$

$$\|(1, -1)\|^2 + \|(0, 1)\|^2 = 2 + 1 = 3;$$

6.33 NEEDS orthogonality.

Given a basis  $\{\vec{v}_1, \vec{v}_2, \dots\}$   
 for a vector space  $V$ , we  
 know that any  $\vec{x}$  in  $V$  is a  
 linear combination of the basis:

$$(*) \quad (c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots) = \vec{x},$$

for some real numbers

$$c_1, c_2, c_3, \dots$$

Determining  $c_1, c_2, \dots$  is, in general,  
 a very big deal, as we have  
 all learned from bitter experience,

(\*) is the vector form of a  
 linear system in the variables

$c_1, c_2, \dots$ , to be solved (normally) by Gauss-Jordan elimination.

## Examples 6.35

(a) Write  $(1, 2, 3, 4)$  as a linear combination of

$$\left\{ \begin{array}{l} (5, 6, 7, 8), (9, 0, 1, 2), (3, 4, 5, 6), \\ (-7, -8, -9, 0) \end{array} \right\} ??$$

We would seek real numbers

$c_1, c_2, c_3, c_4$  so that

$$c_1 \begin{bmatrix} 5 \\ 6 \\ 7 \\ 8 \end{bmatrix} + c_2 \begin{bmatrix} 9 \\ 0 \\ 1 \\ 2 \end{bmatrix} + c_3 \begin{bmatrix} 3 \\ 4 \\ 5 \\ 6 \end{bmatrix} + c_4 \begin{bmatrix} -7 \\ -8 \\ -9 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$$

This would mean Gauss-Jordan elimination, applied to the augmented matrix

$$\left[ \begin{array}{cccc|c} 5 & 9 & 3 & -7 & 1 \\ 6 & 0 & 4 & -8 & 2 \\ 7 & 1 & 5 & -9 & 3 \\ 8 & 2 & 6 & 0 & 4 \end{array} \right]$$

(b) Write  $(1, 2, 3, 4)$  as a linear combination of

$$\left\{ \begin{array}{l} (1, 0, 0, 0), (0, 1, 0, 0), (0, 0, 1, 0), (0, 0, 0, 1) \end{array} \right\} ??$$

$$\left\{ \begin{array}{l} \equiv \vec{e}_1 \\ \equiv \vec{e}_2 \\ \equiv \vec{e}_3 \\ \equiv \vec{e}_4 \end{array} \right.$$

This is such a nice basis (we called it the "standard basis")

that we might guess  
the answer:

$$(1, 2, 3, 4) = 1\vec{e}_1 + 2\vec{e}_2 + 3\vec{e}_3 + 4\vec{e}_4.$$

Notice that

$$1 = (1, 2, 3, 4) \cdot \vec{e}_1$$

$$2 = (1, 2, 3, 4) \cdot \vec{e}_2$$

$$3 = (1, 2, 3, 4) \cdot \vec{e}_3$$

$$4 = (1, 2, 3, 4) \cdot \vec{e}_4$$

This technique will work for any orthogonal basis, except that we need unit vectors. Thus,

if  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m$  is an orthogonal basis, we will have

$$\begin{aligned} \vec{x} &= \left( \vec{x} \cdot \frac{\vec{v}_1}{\|\vec{v}_1\|} \right) \left( \frac{\vec{v}_1}{\|\vec{v}_1\|} \right) + \left( \vec{x} \cdot \frac{\vec{v}_2}{\|\vec{v}_2\|} \right) \left( \frac{\vec{v}_2}{\|\vec{v}_2\|} \right) \\ &+ \dots \\ &= \left( \frac{\vec{x} \cdot \vec{v}_1}{\|\vec{v}_1\|^2} \right) \vec{v}_1 + \left( \frac{\vec{x} \cdot \vec{v}_2}{\|\vec{v}_2\|^2} \right) \vec{v}_2 + \dots \end{aligned}$$

### 6.36 ORTHOGONAL ADVANTAGE (2)

If  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  is an orthogonal basis for  $W$ , then, for any  $\vec{x}$  in  $W$ ,

$$\vec{x} = \left( \frac{\vec{x} \cdot \vec{v}_1}{\|\vec{v}_1\|^2} \right) \vec{v}_1 + \left( \frac{\vec{x} \cdot \vec{v}_2}{\|\vec{v}_2\|^2} \right) \vec{v}_2 + \dots + \left( \frac{\vec{x} \cdot \vec{v}_m}{\|\vec{v}_m\|^2} \right) \vec{v}_m$$

### Example 6.37 Write

$(5, 0, -3)$  as a linear combination of  $\{(1, 1, -2), (0, 2, 1), (5, -1, 2)\}$ .

CHECK first that the set of vectors is orthogonal; since there are three vectors and  $\dim(\mathbb{R}^3) = 3$ , the set of vectors is an orthogonal basis for  $\mathbb{R}^3$ .



We need dot products:

$$(5, 0, -3) \cdot (1, 1, -2) = 11$$

$$\|(1, 1, -2)\|^2 = 6$$

$$(5, 0, -3) \cdot (0, 2, 1) = -3$$

$$\|(0, 2, 1)\|^2 = 5$$

$$(5, 0, -3) \cdot (5, -1, 2) = 19$$

$$\|(5, -1, 2)\|^2 = 30$$

→

$$(5, 0, -3) = \frac{11}{6}(1, 1, -2) - \frac{3}{5}(0, 2, 1) + \frac{19}{30}(5, -1, 2)$$

## REMARKS 6.38

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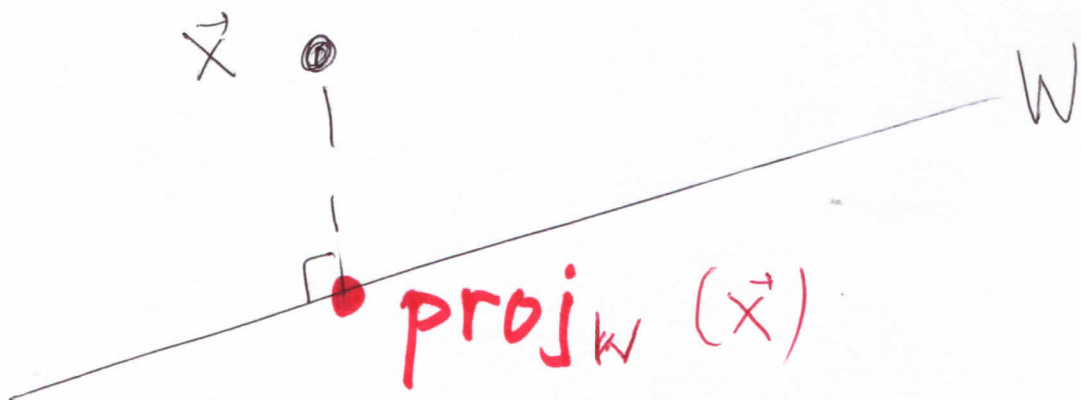
We are putting off the proof of 6.36, since it will be seen to be a special case of Orthogonal Advantage (3).

Notice that 6.36 represents a vector  $\vec{x}$  as a sum of one-dimensional projections of  $\vec{x}$  onto each member of the orthogonal basis

$$\vec{x} = \text{proj}_{\vec{v}_1}(\vec{x}) + \text{proj}_{\vec{v}_2}(\vec{x}) + \dots$$

(see 6.16 - 6.18)

For  $W$  a subspace  
of  $\mathbb{R}^n$  and  $\vec{x}$  in  $\mathbb{R}^n$ , we  
have defined the projection  
of  $\vec{x}$  onto  $W$ , denoted  
 $\text{proj}_W(\vec{x})$ , in 6.13.



For  $W \equiv \text{span}(\{\vec{b}\})$ ,

$$\text{proj}_W(\vec{x}) \equiv \text{proj}_{\vec{b}}(\vec{x}) = \left( \frac{\vec{x} \cdot \vec{b}}{\|\vec{b}\|^2} \right) \vec{b}$$

(see Theorem 6.19)

## Example 6.39

For a multi-dimensional example, take

$$W \equiv \left\{ \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix} \mid x_1, x_2 \text{ real} \right\},$$

the Cartesian plane as a subspace of  $\mathbb{R}^3$ .

Projection onto  $W$  is the oppressive voice of authority, declaring petulantly "No floating!" The  $x_3$  component, representing height

above the ground, is deleted, sending you crashing to the ground.

Specifically, let's spoil the fun of

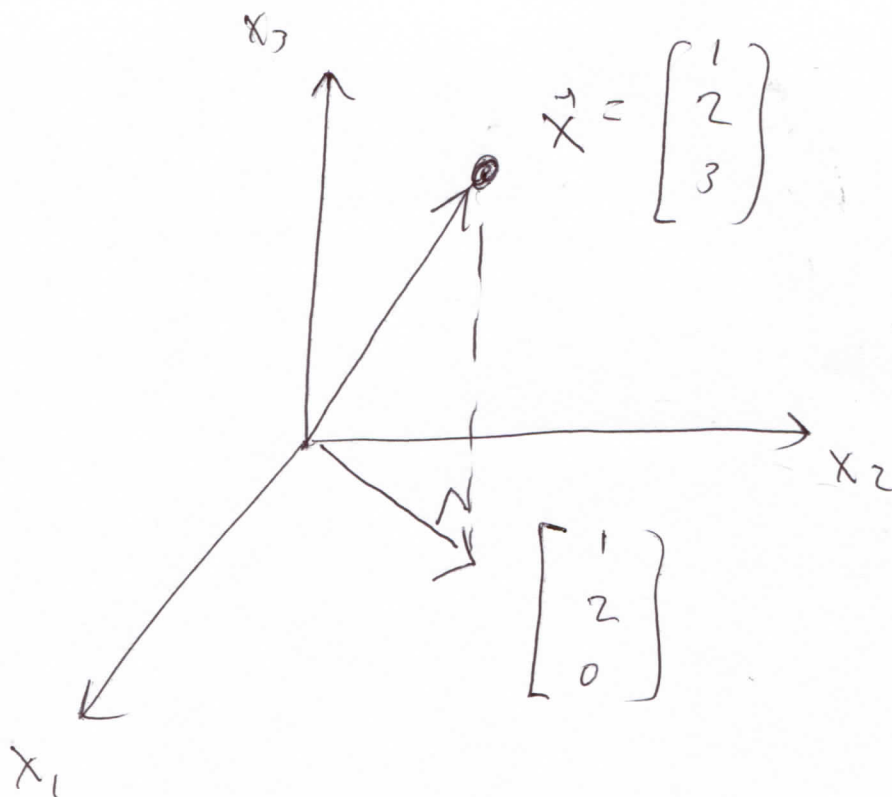
$$\vec{x} \equiv \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

by projecting it onto  $W$ .

The simplest way to get to

$W$  is to remove the  $x_3$  coordinate:

$$P_W(\vec{x}) = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} ?$$



PROOF that  $P_W(\vec{x}) = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$  :

$$\left( \vec{x} - \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 0 \\ 3 \end{bmatrix} \perp \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix},$$

for any real  $x_1, x_2$ ; that is,

$$\left( \vec{x} - \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} \right) \perp W,$$

END OF PROOF

NOTE that

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\} \text{ is an}$$

orthogonal basis for  $W$ , and

$$P_W(\vec{x}) = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix} =$$

$$\left( \frac{\vec{x} \cdot (1, 0, 0)}{\|(1, 0, 0)\|^2} \right) \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \left( \frac{\vec{x} \cdot (0, 1, 0)}{\|(0, 1, 0)\|^2} \right) \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}.$$

This example is meant to motivate the following.

## 6.40 ORTHOGONAL ADVANTAGE (3)

If  $W$  is a subspace of  $\mathbb{R}^n$  and  $\{\vec{w}_1, \vec{w}_2, \dots\}$  is an orthogonal basis for  $W$ , then, for any  $\vec{x}$  in  $\mathbb{R}^n$ ,

$$\begin{aligned} P_W(\vec{x}) &= \text{proj}_{\vec{w}_1}(\vec{x}) + \text{proj}_{\vec{w}_2}(\vec{x}) + \dots \\ &= \left( \frac{\vec{x} \cdot \vec{w}_1}{\|\vec{w}_1\|^2} \right) \vec{w}_1 + \left( \frac{\vec{x} \cdot \vec{w}_2}{\|\vec{w}_2\|^2} \right) \vec{w}_2 + \dots \end{aligned}$$

**Proof:**  $P_W(\vec{x})$  is in  $W$ , so

$$P_W(\vec{x}) = c_1 \vec{w}_1 + c_2 \vec{w}_2 + c_3 \vec{w}_3 + \dots$$

for some real  $c_1, c_2, c_3, c_4, \dots$



By definition of  $P_W(\vec{x})$ , p. 457

$$0 = (\vec{x} - P_W(\vec{x})) \cdot \vec{w}_i,$$

for all  $\vec{w}_i$  in  $W$ ; in particular,

$$0 = (\vec{x} - P_W(\vec{x})) \cdot \vec{w}_1 =$$

$$(\vec{x} - (c_1 \vec{w}_1 + c_2 \vec{w}_2 + \dots)) \cdot \vec{w}_1 =$$

$$(\vec{x} \cdot \vec{w}_1) - (c_1 (\vec{w}_1 \cdot \vec{w}_1) + c_2 (\vec{w}_2 \cdot \vec{w}_1) + \dots)$$

$$= (\vec{x} \cdot \vec{w}_1) - c_1 \|\vec{w}_1\|^2, \text{ by}$$

orthogonality, so that

$$(\vec{x} \cdot \vec{w}_1) = c_1 \|\vec{w}_1\|^2 \rightarrow c_1 = \frac{(\vec{x} \cdot \vec{w}_1)}{\|\vec{w}_1\|^2}.$$

For  $j = 2, 3, \dots$ , the same argument, with  $\vec{w}_j$  replacing  $\vec{w}_1$ , shows that

$$c_j = \left( \frac{\vec{x} \cdot \vec{w}_j}{\|\vec{w}_j\|^2} \right),$$

so that

$$P_W(\vec{x}) = \left( \frac{\vec{x} \cdot \vec{w}_1}{\|\vec{w}_1\|^2} \right) \vec{w}_1 + \left( \frac{\vec{x} \cdot \vec{w}_2}{\|\vec{w}_2\|^2} \right) \vec{w}_2 + \dots,$$

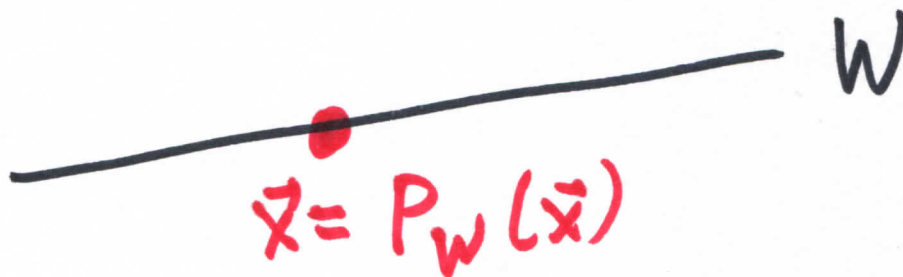
as desired.

## REMARKS 6.41

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The reader should notice similarity between 6.40 and 6.36. This is because 6.36 is the special case of 6.40 when  $\vec{x}$  is itself in  $W$ , which is equivalent to

$$\vec{x} = P_W(\vec{x}).$$



In Example 6.39,

$\vec{x}$  being in  $W$  means you are already on the ground.

The correct answer ~~then~~ to the authority figure of Example 6.39 saying "Drop and give me twenty" is "I can't drop any further, I'm already on the ground."

**Example 6.42** Suppose

$$W = \text{span} \{ (0, 1, 1, 0), (1, 0, 0, 1), (-1, 1, -1, 1) \}$$

Get  $P_W((-2, 0, 5, 3))$

CHECK that

$$\{\vec{w}_1 \equiv (0, 1, 1, 0), \vec{w}_2 \equiv (1, 0, 0, 1), \vec{w}_3 \equiv (-1, 1, -1, 1)\}$$

is orthogonal. Then

$\{\vec{w}_1, \vec{w}_2, \vec{w}_3\}$  is an orthogonal basis for  $W \equiv \text{span}\{\vec{w}_1, \vec{w}_2, \vec{w}_3\}$ ,

thus, with  $\vec{x} \equiv (-2, 0, 5, 3)$ ,

$$P_W(\vec{x}) =$$

$$\left( \frac{\vec{x} \cdot \vec{w}_1}{\|\vec{w}_1\|^2} \right) \vec{w}_1 + \left( \frac{\vec{x} \cdot \vec{w}_2}{\|\vec{w}_2\|^2} \right) \vec{w}_2 + \left( \frac{\vec{x} \cdot \vec{w}_3}{\|\vec{w}_3\|^2} \right) \vec{w}_3$$

$$= \frac{5}{2} \vec{w}_1 + \frac{1}{2} \vec{w}_2$$

We will conclude this section with a definition that encompasses two desirable properties.

### DEFINITION 6.43

An orthogonal set of unit vectors is called an **orthonormal set**.

### Example and Remark

6.44 (a)  $\left\{ \frac{1}{\sqrt{2}}(1,1), \frac{1}{\sqrt{2}}(1,-1) \right\}$

is an orthonormal set.

(b) Notice how tidy  
the formulas in 6.36 and  
6.40 are when  $\{\vec{w}_1, \vec{w}_2, \dots\}$   
is an orthonormal basis  
for  $W$ :

$$P_W(\vec{x}) = (\vec{x} \cdot \vec{w}_1) \vec{w}_1 + (\vec{x} \cdot \vec{w}_2) \vec{w}_2 \\ + \dots$$

# SECTION IV D:

## GRAM-SCHMIDT

## ORTHOGONALIZATION

In the last section we argued the superiority of orthogonal sets over linearly independent sets. This section offers constructive consolation: a particular way to change a linearly independent set into an orthogonal set, without changing span.



# GRAM-SCHMIDT ORTHOGONALIZATION

## 6.44

Suppose  $\{\vec{w}_1, \vec{w}_2, \dots, \vec{w}_m\}$  is a linearly independent set.

Construct a set of vectors  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  as follows.

$$\vec{v}_1 \equiv \vec{w}_1$$

$$\vec{v}_2 \equiv \vec{w}_2 - P_{\text{span}(\vec{v}_1)}(\vec{w}_2)$$

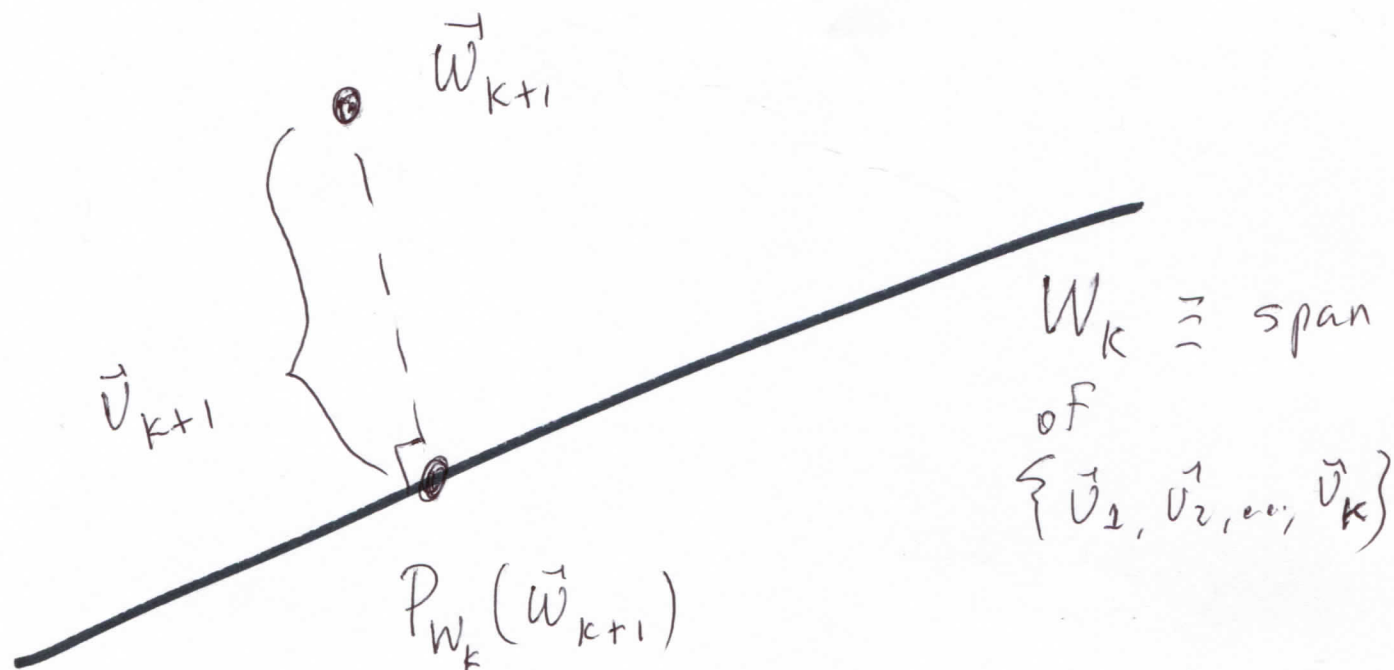
$$\vec{v}_3 \equiv \vec{w}_3 - P_{\text{span}(\vec{v}_1, \vec{v}_2)}(\vec{w}_3)$$

•

•

•

$$\vec{v}_m \equiv \vec{w}_m - P_{\text{span}(\vec{v}_1, \vec{v}_2, \dots, \vec{v}_{m-1})}(\vec{w}_m)$$



$$(1 \leq k \leq m-1)$$

(see 6.40  
for  $P_{W_k}$ )

Then  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  is orthogonal  
and

$$\text{span}(\vec{v}_1, \vec{v}_2, \dots, \vec{v}_j) = \text{span}(\vec{w}_1, \vec{w}_2, \dots, \vec{w}_j)$$

for  $1 \leq j \leq m$ .

# Examples 6.45

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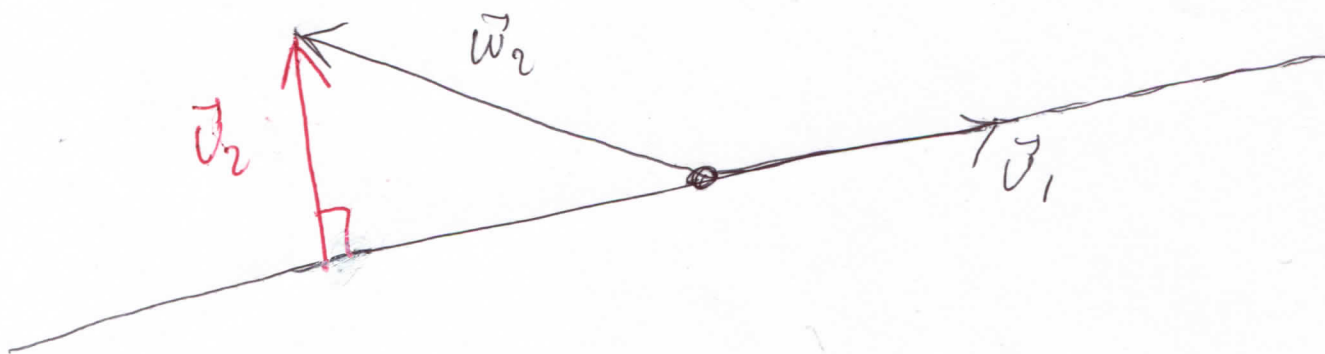
In each of the following,  
apply Gram-Schmidt.

$$(a) \left\{ \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} \right\}$$

$\vec{w}_1$                        $\vec{w}_2$

$$\vec{v}_1 = \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix}$$

$$\vec{v}_2 = \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} - \text{proj}_{(1, -3, 0)} \left( \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} \right)$$



$$= \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} - \left( \frac{(-1, 1, 2) \cdot (1, -3, 0)}{\|(1, -3, 0)\|^2} \right) \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} - \left( \frac{-4}{10} \right) \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} + \frac{2}{5} \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix}$$

$$= \frac{1}{5} \begin{bmatrix} -5 \\ 5 \\ 10 \end{bmatrix} + \begin{bmatrix} 2 \\ -6 \\ 0 \end{bmatrix} = \frac{1}{5} \begin{bmatrix} -3 \\ -1 \\ 10 \end{bmatrix};$$

$$\{ \vec{v}_1, \vec{v}_2 \} = \left\{ \begin{bmatrix} 1 \\ -3 \\ 0 \end{bmatrix}, \frac{1}{5} \begin{bmatrix} -3 \\ -1 \\ 10 \end{bmatrix} \right\}$$

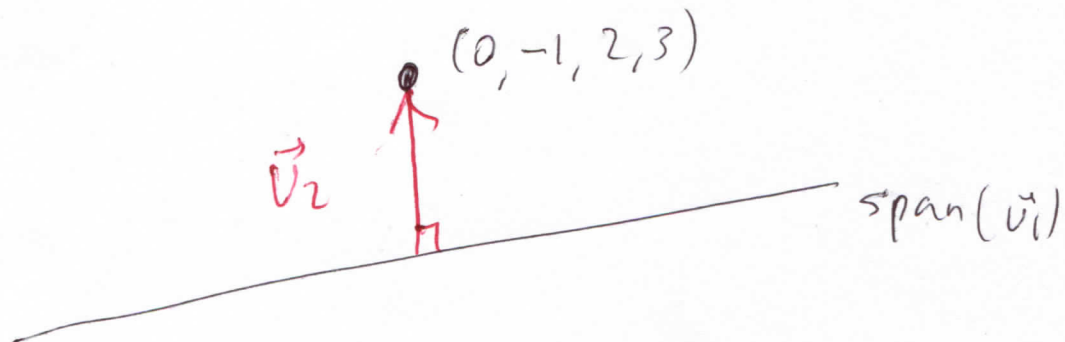
NOTE that  $\vec{v}_1 \cdot \vec{v}_2 = 0$ .

$$(b) \{ (1, 0, 1, 1), (0, -1, 2, 3), (2, 1, 0, 0) \}$$

$$\vec{v}_1 \equiv (1, 0, 1, 1) \leftarrow (\text{SAVE})$$

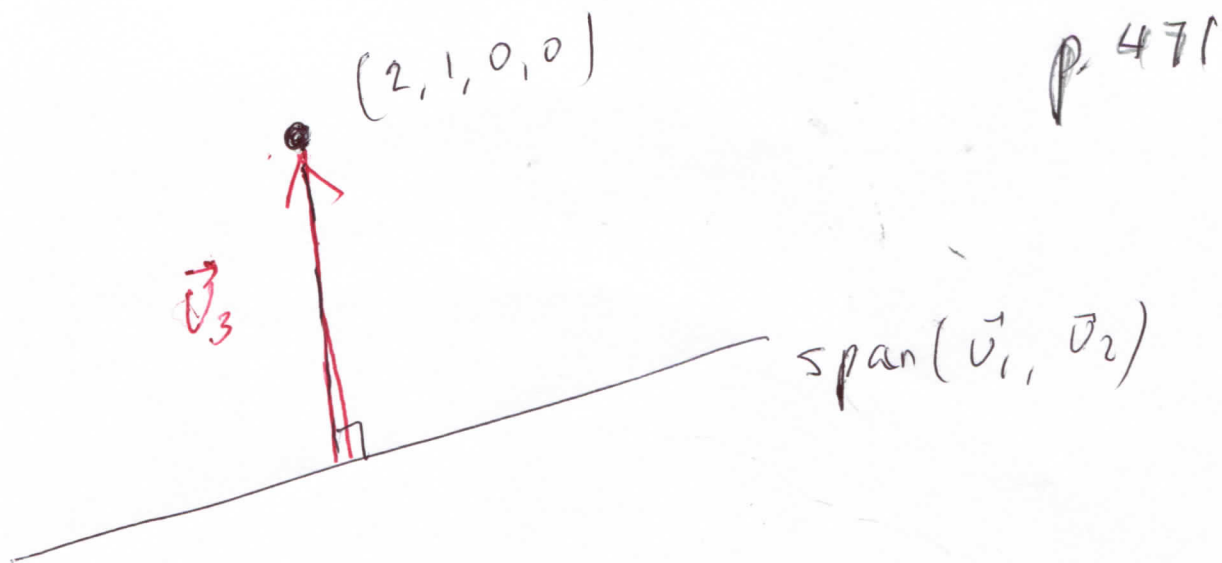
$$\begin{aligned} \vec{v}_2 &\equiv (0, -1, 2, 3) - \text{proj}_{(1, 0, 1, 1)}(0, -1, 2, 3) \\ &= (0, -1, 2, 3) - \left( \frac{(0, -1, 2, 3) \cdot (1, 0, 1, 1)}{\|(1, 0, 1, 1)\|^2} \right) (1, 0, 1, 1) \\ &= (0, -1, 2, 3) - \left( \frac{5}{3} \right) (1, 0, 1, 1) = \\ &= \frac{1}{3} \left[ (0, -3, 6, 9) - (5, 0, 5, 5) \right] = \end{aligned}$$

$$\frac{1}{3}(-5, -3, 1, 4) = \vec{v}_2 \leftarrow (\text{SAVE})$$



$$\begin{aligned}
\vec{v}_3 &= (2, 1, 0, 0) - P_{\text{span}(\vec{v}_1, \vec{v}_2)} (2, 1, 0, 0) = \\
&= (2, 1, 0, 0) - \left[ \text{proj}_{\vec{v}_1} (2, 1, 0, 0) + \text{proj}_{\vec{v}_2} (2, 1, 0, 0) \right] \\
&= (2, 1, 0, 0) - \left[ \left( \frac{(2, 1, 0, 0) \odot (1, 0, 1, 1)}{\|(1, 0, 1, 1)\|^2} \right) (1, 0, 1, 1) \right. \\
&\quad \left. + \left( \frac{(2, 1, 0, 0) \odot \frac{1}{3}(-5, -3, 1, 4)}{\|\frac{1}{3}(-5, -3, 1, 4)\|^2} \right) \frac{1}{3}(-5, -3, 1, 4) \right] \\
&= (2, 1, 0, 0) - \left[ \frac{2}{3}(1, 0, 1, 1) + \left(\frac{-13}{51}\right)(-5, -3, 1, 4) \right] \\
&= \frac{1}{51} \left[ (102, 251, 0, 0) - (34, 0, 34, 34) \right. \\
&\quad \left. + (65, 39, -13, -52) \right] = \\
&= \frac{1}{51} (3, 12, -21, 18) = \frac{1}{17} (1, 4, -7, 6)
\end{aligned}$$

$\vec{v}_3$



Collecting the red boxes:

$$\{ \vec{v}_1, \vec{v}_2, \vec{v}_3 \} =$$

$$\left\{ (1, 0, 1, 1), \frac{1}{3}(-5, -3, 1, 4), \frac{1}{17}(1, 4, -7, 6) \right\}$$

NOTE that  $0 = (1, 0, 1, 1) \odot (-5, -3, 1, 4)$   
 $= (1, 0, 1, 1) \odot (1, 4, -7, 6) =$   
 $(-5, -3, 1, 4) \odot (1, 4, -7, 6)$