

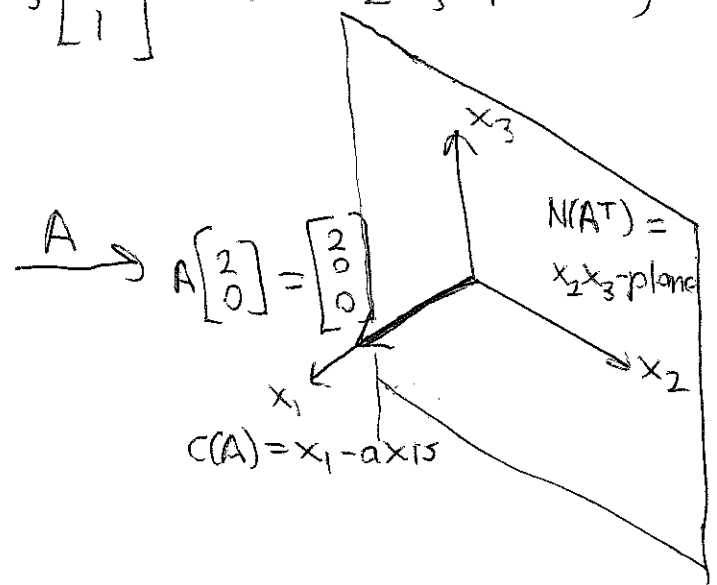
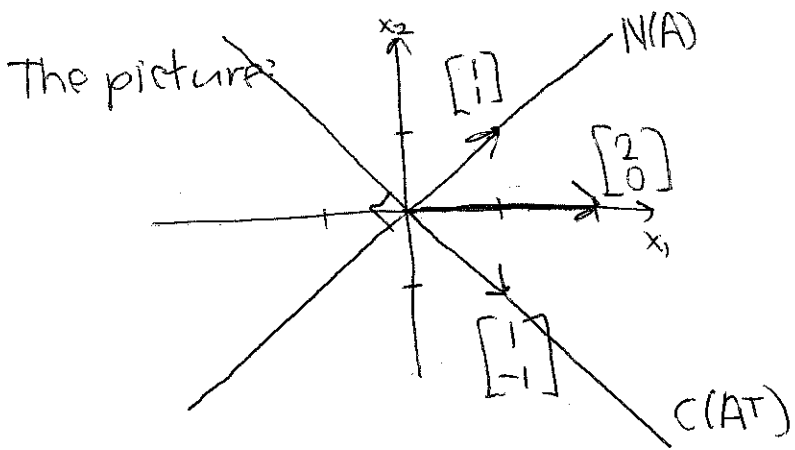
$$C(A) = \text{span} \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix} \right) = \text{span} \left( \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right)$$

$$N(A) : \begin{bmatrix} 1 & -1 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \text{ means } x_1 = x_2 \rightsquigarrow N(A) = \text{all } x_2 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$$C(A^T) = \text{span} \left( \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right) = \text{span} \left( \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \right) \leftarrow \text{perpendicular!}$$

$$N(A^T) = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \text{ means } x_1 = 0 \rightsquigarrow$$

$$N(A^T) = \text{all } \begin{bmatrix} 0 \\ x_2 \\ x_3 \end{bmatrix} = \text{all } x_2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad (= x_2 x_3\text{-plane})$$



What are  $\vec{x}_r$  and  $\vec{x}_n$ ? Need to write

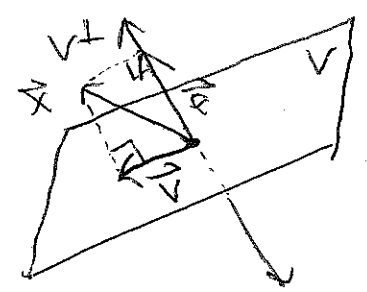
$$\begin{bmatrix} 2 \\ 0 \end{bmatrix} = c \underbrace{\begin{bmatrix} 1 \\ -1 \end{bmatrix}}_{\vec{x}_r} + d \underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}}_{\vec{x}_n} \quad \text{Looks like } c=d=1 \text{ will work.}$$

## Section 4.2 Projections

Big linear algebra problem: Figure out how to write  $\vec{x} = \vec{x}_r + \vec{x}_n$

To say another way: Figure out how to project  $\vec{x}$  onto a subspace (such as  $N(A)$ )

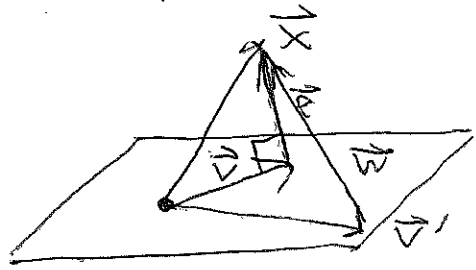
This means: Write  $\vec{x} = \underbrace{\vec{v}}_{\text{in } V} + \underbrace{\vec{e}}_{\text{in } V^\perp}$



Why call it  $\vec{e}$ ? We are trying to approximate  $\vec{x}$  by a vector in  $V$ .  
 $\vec{v}$  is the approximation  
 $\vec{e}$  is the error of the approximation.

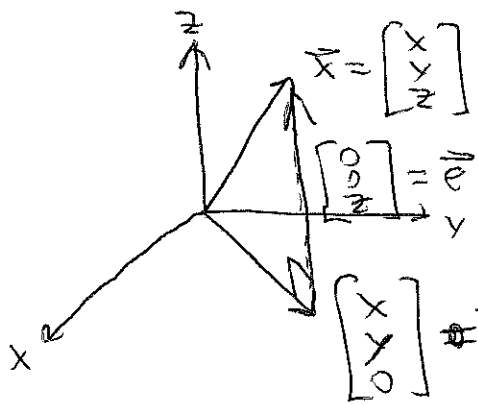
If we did it differently:

We could also write  $\vec{x} = \vec{v}' + \vec{w}$   
where  $\vec{w}$  is not in  $V^\perp$ , but then  
 $\|\vec{w}\| > \|\vec{e}\|$ , worse error.



So what we want to do is orthogonal projection (minimize error).

Very simple example: Project  $\vec{x}$  in  $\mathbb{R}^3$  onto the  $xy$ -plane.



Error vector  $\vec{e}$  is just  $\begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix}$ , which is  
 $\perp$  to  $\begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$  in the  $xy$ -plane.  
Size of the error is  $\left\| \begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix} \right\| = |z|$ .

Turns out, there is a matrix that does this projection!

$$\vec{x} = \vec{v} + \vec{e}$$
$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix}$$

To project  $\vec{x}$  onto  $V^\perp$  instead (z-axis) use  $I - P$ :

$$\begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$
$$\vec{e} = (I - P) \vec{x}$$

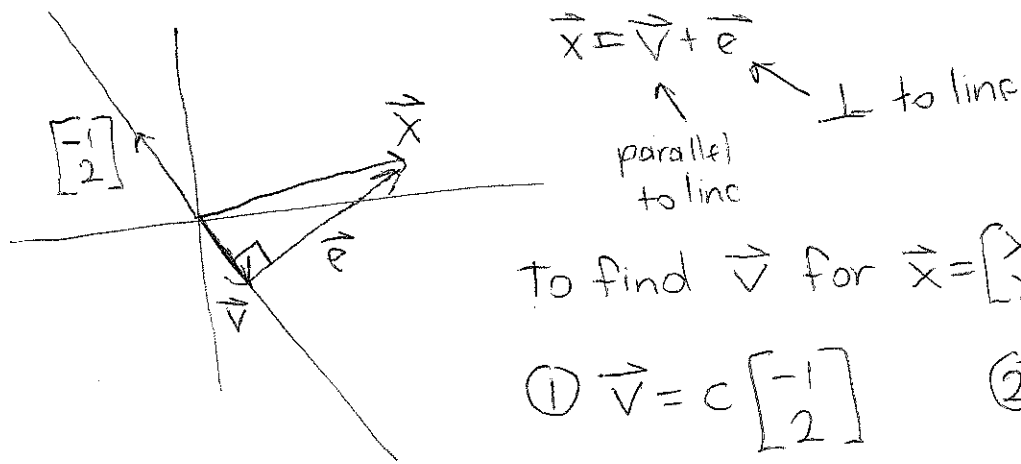
$$\begin{bmatrix} x \\ y \\ 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$
$$\vec{v} = P \vec{x}$$

$P$  and  $I - P$  have two special properties:

Any matrix with these two properties projects vectors onto some subspace!

- ① Symmetric:  $P = P^T$
- ②  $P^2 = P$  (projecting twice is the same as projecting once)

Now let's find projection matrix for projecting onto a line in  $\mathbb{R}^2$ : (104)



To find  $\vec{v}$  for  $\vec{x} = \begin{bmatrix} x \\ y \end{bmatrix}$ , we know 2 things:

①  $\vec{v} = c \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

②  $\vec{p} = \vec{x} - \vec{v}$  is  $\perp$  to  $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$

$$c \left( \begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix} \right) = \begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \quad \left( \vec{x} - c \begin{bmatrix} -1 \\ 2 \end{bmatrix} \right) \cdot \begin{bmatrix} -1 \\ 2 \end{bmatrix} = 0$$

So  $\vec{v} = c \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \frac{\begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{\begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix}} \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

Just a number,  $c$ .

Rewrite by putting number  $\begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$  to the right of  $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$ :

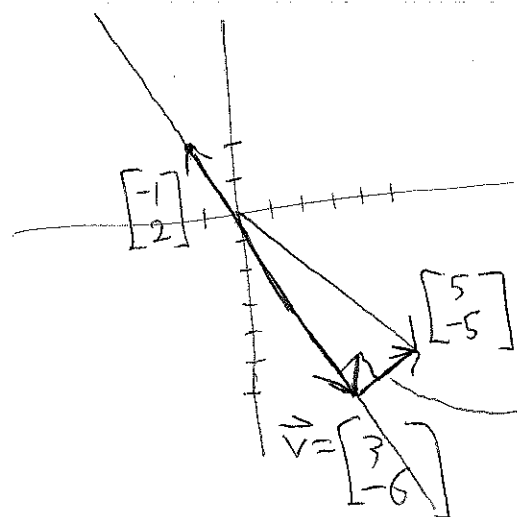
$$\vec{v} = \frac{\begin{bmatrix} -1 \\ 2 \end{bmatrix} \begin{bmatrix} -1 & 2 \end{bmatrix}}{\begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix}} \begin{bmatrix} x \\ y \end{bmatrix}$$

← This is a  $2 \times 2$  matrix,  $P$ .

$$P = \frac{1}{1+4} \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix} = \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix}$$

Example: Project  $\vec{x} = \begin{bmatrix} 5 \\ -5 \end{bmatrix}$  onto this line:

$$\vec{v} = P \begin{bmatrix} 5 \\ -5 \end{bmatrix} = \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix} \begin{bmatrix} 5 \\ -5 \end{bmatrix} = \begin{bmatrix} 3 \\ -6 \end{bmatrix}$$



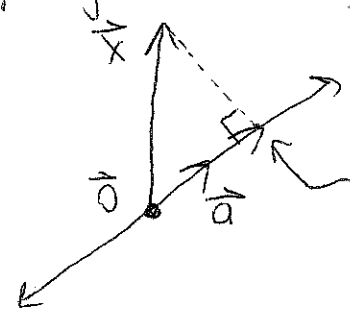
$$\vec{p} = \begin{bmatrix} 5 \\ -5 \end{bmatrix} - \begin{bmatrix} 3 \\ -6 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

Size of error:  $\left\| \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\| = \sqrt{5}$

Check P is a projection matrix: (1) Symmetric,  $P = P^T$  ✓

$$(2) P^2 = \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix} \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix} = \begin{bmatrix} \frac{1+4}{25} & \frac{-2-8}{25} \\ \frac{-2-8}{25} & \frac{4+16}{25} \end{bmatrix} = \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix} = P \quad \checkmark$$

Now project onto any line in  $\mathbb{R}^n$ :



Line =  $\text{span}(\vec{a})$

$\vec{v} = c\vec{a}$ , and  $\vec{e} = \vec{x} - c\vec{a}$  is  $\perp$  to  $\vec{a}$ :

$$\vec{a}^T(\vec{x} - c\vec{a}) = 0 \rightarrow \vec{a}^T \vec{x} = c(\vec{a}^T \vec{a})$$

$$\rightarrow c = \frac{\vec{a}^T \vec{x}}{\vec{a}^T \vec{a}}$$

So to project  $\vec{x}$  onto  $\text{span}(\vec{a})$ :

$$\vec{v} = \frac{\vec{a}^T \vec{x}}{\vec{a}^T \vec{a}} \vec{a}$$

Look at  $(\underbrace{\vec{a}^T \vec{x}}_{\text{scalar}}) \underbrace{\vec{a}}_{\text{vector}} = (a_1 x_1 + a_2 x_2 + \dots + a_n x_n) \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} a_1(a_1 x_1 + \dots + a_n x_n) \\ a_2(a_1 x_1 + \dots + a_n x_n) \\ \vdots \\ a_n(a_1 x_1 + \dots + a_n x_n) \end{bmatrix}$

$$= \begin{bmatrix} a_1 a_1 & a_1 a_2 & \dots & a_1 a_n \\ a_2 a_1 & a_2 a_2 & \dots & a_2 a_n \\ \vdots & \vdots & \ddots & \vdots \\ a_n a_1 & a_n a_2 & \dots & a_n a_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

~~Now we can write this as a matrix multiplication~~  $\leftarrow n \times n \text{ matrix}$

$$\text{So } \vec{v} = \frac{\vec{a} \vec{a}^T}{\vec{a}^T \vec{a}} \vec{x}$$

The "outer product,"  $\vec{a} \vec{a}^T$

scalar

This means the projection matrix is:  $P = \frac{\vec{a}\vec{a}^T}{\vec{a}^T\vec{a}}$

$P$  has the 2 properties of a projection matrix:

① Symmetric:  $P^T = \frac{1}{\vec{a}^T\vec{a}} (\vec{a}\vec{a}^T)^T = P$  ✓  
 $\uparrow$   
 $(\vec{a}^T)^T \vec{a}^T = \vec{a}\vec{a}^T$

②  $P^2 = \left(\frac{1}{\vec{a}^T\vec{a}}\right)^2 \vec{a}\vec{a}^T\vec{a}\vec{a}^T = \frac{\vec{a}\vec{a}^T}{\vec{a}^T\vec{a}} = P$  ✓  
 scalar, cancels with one in the denominator

Example Projection onto the line  $\text{span}\left(\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}\right)$  in  $\mathbb{R}^3$ :

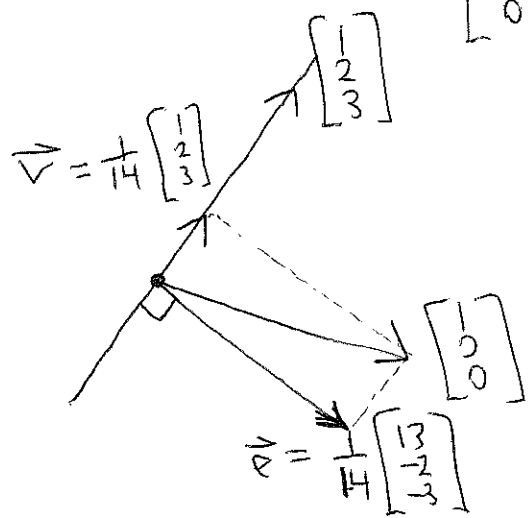
$$P = \frac{\vec{a}\vec{a}^T}{\vec{a}^T\vec{a}} = \frac{\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}}{\begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}} = \frac{1}{14} \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix}$$

↖ All rows are multiples of each other  
 →  $\text{rank}(P) = 1$   
 →  $\dim(C(P)) = 1$   
 →  $C(P)$  is a line

It's the span of  $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

Let's project  $\vec{x} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$  onto this line:  $P \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \frac{1}{14} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$  ← This is  $\vec{v}$ .

Error:  $\vec{e} = \vec{x} - P\vec{x} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \frac{1}{14} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \frac{1}{14} \begin{bmatrix} 13 \\ -2 \\ -3 \end{bmatrix}$  ← This vector is  $\perp$  to  $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ .



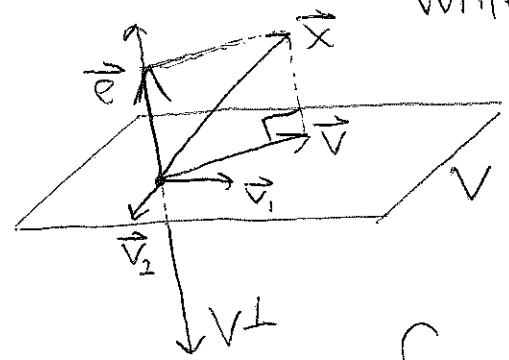
Notice that  $\vec{e} = \vec{x} - P\vec{x}$   
 $= (\mathbf{I} - P)\vec{x}$

$\mathbf{I} - P$  projects  $\vec{x}$  onto the orthogonal complement,  $\vec{v} \perp$ .

Now what if  $V$  is bigger than a line?

Suppose  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  is a basis for a subspace  $V$  in  $\mathbb{R}^n$ :

Project  $\vec{x}$  onto  $V$ :



Write  $\vec{x} = \vec{v} + \vec{e}$   
in  $V$       in  $V^\perp$

Know 2 things:

- (1)  $\vec{v} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_m \vec{v}_m$
- (2)  $\vec{x} - \vec{v}$  is  $\perp$  to all of  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m$ .

From (1): 
$$\vec{v} = \underbrace{\begin{bmatrix} \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_m \end{bmatrix}}_A \underbrace{\begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_m \end{bmatrix}}_{\vec{c}} = A \vec{c}$$

From (2): 
$$\begin{aligned} \vec{v}_1^T (\vec{x} - A \vec{c}) &= 0 \\ \vec{v}_2^T (\vec{x} - A \vec{c}) &= 0 \\ &\vdots \\ \vec{v}_m^T (\vec{x} - A \vec{c}) &= 0 \end{aligned} \quad \rightsquigarrow \quad \begin{matrix} \swarrow A^T \\ \begin{bmatrix} \vec{v}_1^T \\ \vec{v}_2^T \\ \vdots \\ \vec{v}_m^T \end{bmatrix} (\vec{x} - A \vec{c}) = \vec{0} \end{matrix}$$

So  $A^T \vec{x} = \underbrace{(A^T A)}_{\substack{\uparrow m \times n \\ \uparrow n \times m}} \vec{c}$

We want to solve for  $\vec{c}$ . But is  $A^T A$  invertible?

→ It's  $m \times m$ , so yes if it's columns are independent

→ Yes if  $N(A^T A) = \{\vec{0}\}$ .

Need to show  $\vec{x} = \vec{0}$ .

Let's prove  $N(A^T A) = \{\vec{0}\}$ : Suppose  $\vec{x}$  is in the nullspace:

Then  $A^T A \vec{x} = \vec{0}$ . This means  $\vec{x}^T A^T A \vec{x} = 0$  also

$\downarrow$   
 $(A \vec{x})^T (A \vec{x}) = 0$

$A \vec{x}$  is  $\perp$  to itself! Only possible if  $A \vec{x} = \vec{0}$ , so  $\vec{x}$  is in  $N(A)$ .

But  $A$  has independent columns (basis for  $V$ ), so its null space is  $\{\vec{0}\}$ .

So  $\vec{x}$  has to be  $\vec{0}$ . ✓

We showed:  $\vec{x}$  in  $N(A^T A)$  means  $\vec{x} = \vec{0}$ , that is,  $N(A^T A) = \{\vec{0}\}$ , (108)  
 $\rightarrow A^T A$  is an invertible  $m \times m$  matrix.

Now go back to  $(A^T A) \vec{c} = A^T \vec{x}$  (called the "normal equation" for  $\vec{c}$ )

$\vec{c} = (A^T A)^{-1} A^T \vec{x}$ . Now remember that the projection vector was  $\vec{v} = A \vec{c}$ .

$$\text{So } \vec{v} = A \vec{c} = [A(A^T A)^{-1} A^T] \vec{x}$$

$\hookleftarrow$  This is the projection matrix  $P$ !

So if you want to project  $\vec{x}$  onto  $V$ :

(1) Find a basis for  $V = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$

(2) Form  $A = [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_m]$

(3) Project matrix is  $P = A(A^T A)^{-1} A^T$

(4) Project  $\vec{x}$ :  $\vec{v} = P \vec{x}$

Example  $V =$  plane in  $\mathbb{R}^3$  with equation  $x + 2y + 3z = 0$

(1) Find basis:  $V =$  all  $\begin{bmatrix} x \\ y \\ z \end{bmatrix}$  with  $x = -2y - 3z$

$$= \text{all } \begin{bmatrix} -2y - 3z \\ y \\ z \end{bmatrix} = \text{all } y \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} + z \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \quad \begin{array}{l} \text{linearly independent} \\ \text{spanning set, so a} \\ \text{basis for } V \end{array}$$

(2)  $A = \begin{bmatrix} -2 & -3 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$  Notice:  $A$  is not invertible, but  $A^T A$  is!

$$(3) A^T A = \begin{bmatrix} -2 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix} \begin{bmatrix} -2 & -3 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 5 & 6 \\ 6 & 10 \end{bmatrix}, (A^T A)^{-1} = \frac{1}{50 - 36} \begin{bmatrix} 10 & -6 \\ -6 & 5 \end{bmatrix}$$

$\hookleftarrow 1/14$

$$P = A(A^T A)^{-1} A^T = \begin{bmatrix} -2 & -3 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \frac{1}{14} \begin{bmatrix} 10 & -6 \\ -6 & 5 \end{bmatrix} \begin{bmatrix} -2 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix}$$

$$= \frac{1}{14} \begin{bmatrix} -2 & -3 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -2 & 10 & -6 \\ -3 & -6 & 5 \end{bmatrix} = \frac{1}{14} \begin{bmatrix} 13 & -2 & -3 \\ -2 & 10 & -6 \\ -3 & -6 & 5 \end{bmatrix}$$

Should look familiar!

④ Let's project  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$  onto  $V$ :  $\vec{v} = P \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \frac{1}{14} \begin{bmatrix} 13 \\ -2 \\ -3 \end{bmatrix}$

Here,  $V$  is the orthogonal complement of the line from earlier example:

So our new  $P$  is actually  $I - (\text{old } P)$   
 ↗ for the line.

Question: What is the rank of our new  $P$ ?

$$\text{rank}(P) = \dim \text{ of } C(P) \leftarrow \begin{array}{l} \text{all vectors like } P\vec{x} \\ \text{= all vectors in } V \end{array}$$

$$= \dim V = 2$$

In general: rank of a projection matrix = dimension of the subspace you are projecting onto.