

Tuesday Oct 27

## § 5.3 Orthogonal transformations

Matrix transposes

projection formulas and matrices

- Finish Monday notes : pages 2 - 4

Surprising applications for Chapter 5!

Wednesday - § 5.4 - in class

Friday - § 5.5 - in LCB 115  
computer lab.  
(project 2)

Then play with transpose algebra :

$$(A^T)^T = A$$

$$(AB)^T = B^T A^T \quad (\text{notice order, like for inverse})$$

check this:

$$(A^{-1})^T = (A^T)^{-1}$$

check this:

Def :  $A = [a_{ij}]$  is symmetric iff  $A^T = A$  ( $a_{ij} = a_{ji} \forall i, j$ )

$A = [a_{ij}]$  is antisymmetric iff  $A^T = -A$  ( $a_{ij} = -a_{ji} \forall i, j$ )

$$\begin{bmatrix} 1 & 4 & 5 \\ 4 & 2 & 6 \\ 5 & 6 & 3 \end{bmatrix}$$

$$\begin{bmatrix} 0 & -1 & 2 \\ 1 & 0 & -3 \\ -2 & 3 & 0 \end{bmatrix}$$

If  $A_{m \times n}$  the  $A^T A = A_{n \times m}^T A_{m \times n}$  is  $n \times n$  symmetric  
 $A A^T = A_{m \times n} A_{n \times m}^T$  is  $m \times m$  symmetric

check this:

example  $A = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ . Compute  $A^T A$

$$A A^T$$

(2)

Matrix for projection onto a subspace  $V \subset \mathbb{R}^n$  (uses transpose)

Let  $\{\tilde{u}_1, \tilde{u}_2, \dots, \tilde{u}_k\}$  be an orthonormal basis for  $V$

Find the matrix for  $T(\vec{x}) = \text{proj}_V \vec{x}$

Not so hard, actually:

$$\text{proj}_V \vec{x} = (\vec{x} \cdot \tilde{u}_1) \tilde{u}_1 + (\vec{x} \cdot \tilde{u}_2) \tilde{u}_2 + \dots + (\vec{x} \cdot \tilde{u}_k) \tilde{u}_k$$

$$= \begin{bmatrix} \tilde{u}_1 & \tilde{u}_2 & \dots & \tilde{u}_k \end{bmatrix} \begin{bmatrix} \vec{x} \cdot \tilde{u}_1 \\ \vec{x} \cdot \tilde{u}_2 \\ \vdots \\ \vec{x} \cdot \tilde{u}_k \end{bmatrix}$$

$$= \begin{bmatrix} \tilde{u}_1 & \tilde{u}_2 & \dots & \tilde{u}_k \end{bmatrix} \begin{bmatrix} \tilde{u}_1^T \\ \tilde{u}_2^T \\ \vdots \\ \tilde{u}_k^T \end{bmatrix} \begin{bmatrix} \vec{x} \end{bmatrix} = (A A^T) \vec{x}$$

$$\text{for } A = \begin{bmatrix} \tilde{u}_1 & \tilde{u}_2 & \dots & \tilde{u}_k \end{bmatrix}$$

Example 1: Let  $V = \left\{ \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \text{ s.t. } -x_1 + 2x_2 + x_3 = 0 \right\}$ . Find the matrix for  $T(\vec{x}) = \text{proj}_V \vec{x}$

Check:

b) Chapter 3-4 way:

The matrix  $B$  for  $T$   
with respect to the basis

$$E = \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\} \text{ is}$$

$$B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ why?}$$

$$S_{E \leftrightarrow E} = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} \text{ why?}$$

then find  $A A^T$ , see if  
answer agrees with similar  
matrix way, on the right.

$$S_B S$$

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

$$= \begin{bmatrix} \frac{5}{6} & \frac{1}{3} & \frac{1}{6} \\ \frac{1}{3} & \frac{1}{3} & -\frac{1}{3} \\ \frac{1}{6} & -\frac{1}{3} & \frac{5}{6} \end{bmatrix}$$

honest!

## § 5.4 Least squares approximate solutions to $A\vec{x} = \vec{b}$

In experiments you must often find a solution to

$$A\vec{x} = \vec{b}$$

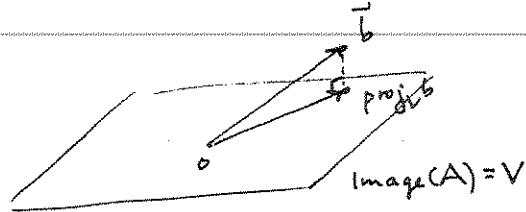
where no solution exactly exists  $\rightarrow$  because e.g.  $\vec{x}$  is a collection of data points.  
(we'll see lots of real examples tomorrow.)

The problem for an inconsistent system is that  $\vec{b}$  is not in the image of  $A$ .

The solution(s) to

\*  $A\vec{x} = \text{proj}_{\text{Im}(A)} \vec{b}$

get  $A\vec{x}$  as close to  $\vec{b}$  as possible.  
they are called "least squares" sol'n's.



Example 2: Find the least-squares sol'n to

$\text{Image}(C) = \text{the plane } -x_1 + 2x_2 + x_3 = 0,$   
see example 1.  $\begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix}$  is not on this plane!

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} = \vec{b}$$

"C"

Using  $\mathcal{U} = \left\{ \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} \right\}$  basis for  $\text{Image}(C)$ ,

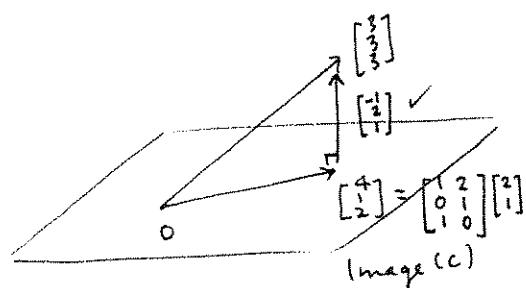
$$\text{proj}_V \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} = (-) \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + (-) \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

$$= \begin{bmatrix} 4 \\ 1 \\ 2 \end{bmatrix} \quad \text{closest pt. to } \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} \text{ on image}(C)$$

$$\begin{array}{r|rr|r} 1 & 2 & 4 \\ 0 & 1 & 1 \\ 1 & 0 & 2 \end{array}$$

$$\begin{aligned} x_1 &= 2 \\ x_2 &= 1 \end{aligned} \quad \text{consistent!}$$

$$\boxed{\vec{x} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}}$$



There's actually a smart way to find least squares solutions that doesn't require an orthonormal basis:

$$A\vec{x} = \text{proj}_{\text{Im}(A)} \vec{b}$$

$$\Leftrightarrow \vec{b} - A\vec{x} \in (\text{Im}(A))^\perp$$

$\Leftrightarrow \vec{b} - A\vec{x}$  is  $\perp$  to a collection of vectors spanning  $\text{Im}(A)$   
e.g. the columns of  $A$ !

$$\Leftrightarrow A^T(\vec{b} - A\vec{x}) = \vec{0} \quad \text{the rows of } A^T \text{ are the columns of } A!$$

$$\Leftrightarrow A^T\vec{b} - A^TA\vec{x} = \vec{0}$$

$$\Leftrightarrow A^TA\vec{x} = A^T\vec{b}$$

this system will always be consistent, since least square's sol'n's always exist. If columns of  $A$  are independent they are a basis for  $\text{Im}(A)$ , so sol'n is unique.

Example 3 : redo example 2 w/o orthonormal basis

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} \quad \text{no sol'n.}$$

$$\begin{bmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 \\ 2 & 1 & 0 \\ 3 & 3 & 3 \end{bmatrix} \quad \text{least square's sol'n.}$$

multiply out &  
solve this  $2 \times 2$  system.  
answer should agree!

projection matrices without o.n.basis  
& lots of least squares examples,  
applications, tomorrow. we have a  
look at 65.4.