

# LECTURE 14

No participation slips until further notice. (We're not an evening class, but some students commute and it makes no sense to come for fewer classes.) Stay safe!

Recall Two types of elem. orthogonal matrices (rigid transformations)

- Givens rotations: (rotation matrices)

$$G = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & c & s \\ & & -s & c \end{bmatrix}$$

(i.e under a permutation of bases,  
this is just id in all components except  
the two spanning a particular plane; then  
do plane rotation there.)

- Householder trans (reflection matrices)

$$H = I - 2 \frac{vv^T}{v^Tv} = I - 2 \cdot \frac{v}{\|v\|} \cdot \frac{v^T}{\|v\|} = I - 2uu^T$$

$\Rightarrow I - 2P$  is an orthogonal matrix (HW04)

↗

## Projection with orthonormal basis

Suppose you have a subspace given as the span of an orthonormal basis:  $\text{span } \{q_1, \dots, q_n\} \subset \mathbb{R}^m$ .

Let  $Q = [q_1 \dots q_n]$

$\underbrace{\left[ \begin{array}{c} Q \\ \hline n \end{array} \right]}_{\text{Then } QQ^T \text{ is an orthogonal projector too!}}$

$$\left[ \begin{array}{c} \uparrow \\ q_1 \\ \uparrow \\ \vdots \\ q_n \\ \downarrow \end{array} \right] \left[ \begin{array}{c} \xrightarrow{q_1} \\ \xrightarrow{Q^T} \\ \xrightarrow{q_n} \end{array} \right] \left[ \begin{array}{c} v \\ \vdots \\ v_m \end{array} \right] = [q_1 \dots q_n] \left[ \begin{array}{c} q_1^T \\ \vdots \\ q_n^T \end{array} \right] v$$

$$= \left( \sum_{i=1}^n q_i q_i^T \right) v$$

{\underbrace{\quad}\_{\text{sum of rank 1}}}  
 $m \times m$  matrices!

The complementary projector is also orthogonal (clear from defn!)

## Projection with arbitrary basis (related to something familiar...)

Now let  $\{a_1, \dots, a_n\}$  be an arbitrary basis for our subspace we want to project onto.

Organize as  $A = [a_1 \dots a_n] \in \mathbb{R}^{m \times n}$  (recall  $n \leq m$ )

\* Note that  $A$  is full (column) rank.

Let  $v \in \mathbb{R}^m$ . Then  $v = v_1 + v_2$  where  $v_1 \in \text{range}(A)$ ,  $v_2 \in \text{null}(A^\top)$

$\Rightarrow v_1 = Ax$  for some  $x \in \mathbb{R}^n$ . (domain of  $A$  is  $\mathbb{R}^n$ )

" $\forall i, a_i^\top v_2 = v - v_1 = v - Ax$ " (want orthogonal projection)

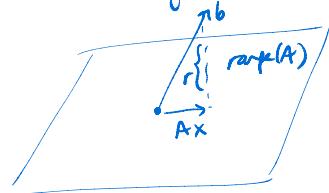
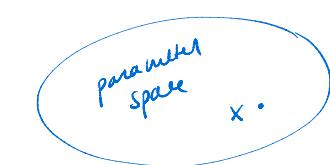
iff  $a_i^\top (v - Ax) = 0$  for all  $i$

iff  $A^\top (v - Ax) = 0$ .  $\rightsquigarrow \textcircled{?}$

iff  $A^\top A x = A^\top v$  the normal eqns!

recall we had wanted to project onto the range of a particular matrix.  
This is exactly what we were doing.

Q. What is the actual projector in the normal eqns? (Where are we projecting?  
A: onto  $\text{range}(A)$ )



$$A^\top A x = A^\top b$$

Wanted to solve for  $x$  (the parameters of our model)  
and I claimed it was the image of  $b$  (where under projection).

So solve: (recall  $A$  full rank  $\Rightarrow A^\top A$  invertible)

$$x = (A^\top A)^{-1} A^\top b \quad \text{We want to project } b \text{ onto } \text{range } A.$$

$$\Rightarrow P_A(b) = Ax = A(A^\top A)^{-1} A^\top b \Rightarrow \boxed{P_A = A(A^\top A)^{-1} A^\top}$$

Aside Pseudo inverses. Let  $A \in \mathbb{R}^{m \times n}$ .  $\mathbb{R}^n \xrightarrow{A} \mathbb{R}^m$

If  $A$  (as above) isn't square, then there is no " $A^{-1}$ ".

However, if we ask for a matrix  $B$  such that

$$B \cdot A = I_n \text{ and } A \cdot B = I_m$$

- $BA = I_n$  since domain( $A$ ) =  $\mathbb{R}^n$
- $AB = I_m$  since codomain( $A$ ) =  $\mathbb{R}^m$

} the dream.

\* also clear here that  $A$  must be full column rank!  
Otherwise not an injective function  $\Rightarrow$  can't possibly compose with  $B$  to be injective.

So assume  $A$  is full column rank. (as in homal equations)

$\Rightarrow A^T A$  is invertible. Let  $B = (A^T A)^{-1} A^T$

- Then:
- $BA = (A^T A)^{-1} A^T \cdot A = I_n$  indeed!
  - $AB = A(A^T A)^{-1} A^T \dots$  well, this is the best we got.

if  $A$  were square, and thus invertible, we would have access to " $A^{-1}$ " and this would really be  $I_m$ .

defn The pseudo inverse  $A^+$  of a full rank matrix  $A \in \mathbb{R}^{m \times n}$

$$(i.e. n \leq m) \text{ is } A^+ = (A^T A)^{-1} A^T \in \mathbb{R}^{n \times m}$$

Compare with

$$\Rightarrow P_A(b) = Ax = A(A^T A)^{-1} A^T b \Rightarrow \boxed{P_A = A(A^T A)^{-1} A^T}$$

$$\text{i.e. } P_A = A^+ A .$$

Rmk On the subspace  $\text{range}(A) \subset \mathbb{R}^m$ ,  $P_A$  really is the identity:

$$\text{If } v = Ax, \text{ then } P_A v = v \quad (\text{recall: } P_A v - v = P_A x - x = 0)$$

$$\text{And so } P_A|_{\text{range}(A)} = \text{Id}_{\text{range}(A)} \text{ dim range } A = n; P_A|_{\text{range}(A)} \sim I_n .$$