

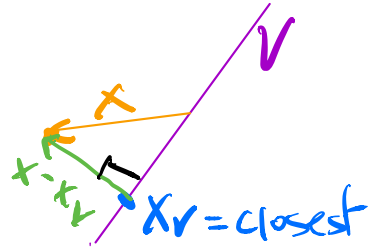
Orthogonal Projections

Question: If V is a subspace & x is a vector, what's the **closest** vector to x in V ?

Closest: the vector $x_v \in V$ minimizing $\|x - x_v\|$

Observe: x_v is the closest vector

$$\Leftrightarrow x - x_v \perp V \Leftrightarrow x - x_v \in V^\perp.$$



Def: The **orthogonal projection** of x onto V is the closest vector x_v to x in V . It's characterized by:
 $x_{v^\perp} = x - x_v \in V^\perp.$

The **orthogonal decomposition** of x relative to V is

$$x = \underbrace{x_v}_{\text{closest vector in } V} + \underbrace{x_{v^\perp}}_{\text{closest vector in } V^\perp}$$

NB: x_{v^\perp} is the projection of x onto V^\perp :
 $x - x_{v^\perp} = x_v \in V = (V^\perp)^\perp$

How to compute?

Step 0: suppose $V = \text{Col}(A)$.

Want $x_v \in V = \text{Col}(A) \Rightarrow x_v = A\hat{x}$ for $\hat{x} \in \mathbb{R}^n$

$V^\perp = \text{Nul}(A^T)$ want $x - x_v \in V^\perp$

$$\Rightarrow A^T(x - x_v) = 0 \Rightarrow A^T x = A^T x_v = A^T A \hat{x}$$

"a linear combo of cols of A "

Solve $A^T A \hat{x} = A^T x \rightsquigarrow x_v = A \hat{x}$

Eg: Project $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ onto $\text{Col} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$.

$$A^T A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 & 2 \\ 2 & 2 \end{pmatrix}$$

$$A^T x = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Solve $A^T A \hat{x} = A^T x$:

$$\left(\begin{array}{cc|c} 2 & 2 & 1 \\ 2 & 2 & 1 \end{array} \right) \xrightarrow{\text{RREF}} \left(\begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 1/2 \end{array} \right) \Rightarrow \hat{x} = \begin{pmatrix} 0 \\ 1/2 \end{pmatrix}$$

$$\Rightarrow x_v = A \hat{x} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1/2 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 1/2 \\ 0 \end{pmatrix}$$

$$x_{v^\perp} = x - x_v = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1/2 \\ 1/2 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 \\ -1/2 \\ 0 \end{pmatrix}$$

orthogonal decomposition: $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 1/2 \\ 0 \end{pmatrix} + \begin{pmatrix} 1/2 \\ -1/2 \\ 0 \end{pmatrix}$

Distance of $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ from $\text{Col}(A)$: $\|x_{v^\perp}\| = \left\| \begin{pmatrix} 1/2 \\ -1/2 \\ 0 \end{pmatrix} \right\| = \sqrt{\frac{1}{4} + \frac{1}{4}} = \frac{1}{\sqrt{2}}$.

Def: The normal equation of $Ax = b$ is $A^T A x = A^T b$

Procedure: To find the orthogonal decomposition of x relative to $\text{Col}(A) = V$:

(1) Solve the normal equation $A^T A \hat{x} = A^T x$ for \hat{x}

(2) $x_v = A \hat{x}$ $x_{v^\perp} = x - x_v$

Q: How to project onto $\text{Nul}(A)$?

Eg: Project $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ onto $\text{Nul}\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$.

$$V^\perp = \text{Row}\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \text{Col}\begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$$

$$\Rightarrow X_{V^\perp} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{pmatrix} \Rightarrow X_V = X - X_{V^\perp} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ -\frac{\sqrt{2}}{2} \\ 0 \end{pmatrix}.$$

Moral: Decide **first** if it's easier to project onto V or V^\perp , then compute the easier one!

What if $V = \text{Span}\{v\}$ is a line? $V = \text{Col}(A)$ $A = v$
 $A^T A \hat{x} = A^T x \Leftrightarrow v^T v \hat{x} = v^T x \Leftrightarrow (v \cdot v) \hat{x} = v \cdot x$
 $\Rightarrow \hat{x} = \frac{v \cdot x}{v \cdot v} \Rightarrow X_V = A \hat{x} = \frac{v \cdot x}{v \cdot v} v$

Projection onto a Line: If $V = \text{Span}\{v\}$, $v \neq 0$, then

$$X_V = \frac{v \cdot x}{v \cdot v} v$$

Eg: Project $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ onto $\text{Span}\left\{\begin{pmatrix} 1 \\ 1 \end{pmatrix}\right\}$

$$X_V = \frac{\begin{pmatrix} 1 \\ 0 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Properties of Projections:

$$(1) x_v = x \iff x \in V$$

$$(2) x_v = 0 \iff x \in V^\perp$$

orthogonal decomposition

$$x = x_v + 0$$

$$x = 0 + x_{v^\perp}$$

Project Matrices

HW#6.17

Fact: A has full column rank $\iff A^T A$ is invertible

Suppose A has full col rank, $V = \text{Col}(A)$.

To compute x_v :

$$A^T A \hat{x} = A^T x \iff \hat{x} = (A^T A)^{-1} A^T x$$

$$\implies x_v = A \hat{x} = A (A^T A)^{-1} A^T x$$

If $P = A (A^T A)^{-1} A^T$ then $x_v = P x$.

↑ cols of A are a basis for V

Eg: $V = \text{Col} \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \rightsquigarrow A^T A = \begin{pmatrix} 3 & 2 \\ 2 & 2 \end{pmatrix} \rightarrow (A^T A)^{-1} = \frac{1}{2} \begin{pmatrix} 2 & -2 \\ -2 & 3 \end{pmatrix}$

$$P = A (A^T A)^{-1} A^T = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{2} \begin{pmatrix} 2 & -2 \\ -2 & 3 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \leftarrow \begin{matrix} \text{don't compute} \\ \text{by hand} \end{matrix}$$

Check: $x_v = P x$
 $x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$P \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 1/2 \\ 0 \end{pmatrix}$$

Def: Let V be a subspace. The projection matrix onto V is the matrix P_V such that $P_V x = x_v$ for all x .

Procedure: To compute P_V :

(1) Find a **basis** for V .

(2) Let A = the matrix with those vectors.

A has full col rank, $V = \text{Col}(A)$

$$(3) P_V = A(A^T A)^{-1} A^T$$

Eg: $V = \text{Col} \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix} \xrightarrow{\text{use}} A = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}$

What if $V = \text{Span}\{v\}$ is a line? $V = \text{Col}(A)$ $A = v$
 $A^T A = v^T v = v \cdot v$

$$\Rightarrow P_V = v \left(\frac{1}{v \cdot v} \right) v^T = \frac{1}{v \cdot v} v v^T \quad \text{"outer product"}$$

Eg: $V = \text{Span}\{(1)\}$ $v = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$$P_V = \frac{1}{(1) \cdot (1)} (1) (1 \ 1) = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$

Check: $x_v = P_V x$ $x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$P_V x = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \checkmark$$

Projection Matrix for a Line

If $V = \text{Span}\{v\}$ for $v \neq 0$ then

$$P_V = \frac{1}{v \cdot v} v v^T$$

Properties of Projection Matrices

HW #6.8

$$(1) P^2 = P$$

$$(2) P^T = P$$

$$(3) \underline{I_n = P_V + P_{V^\perp}}$$

Moral: Decide **first** if it's easier to compute P_V or P_{V^\perp} , then compute the easier one!