## Projection theorem Normal Equations ROB 501

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- Last time: Pre-projection theorem & Gram Schmidt (GS) process
  - Will start with some useful results related to GS and orthogonal subspaces
- Projection theorem
- Normal equations

## Orthogonal complement

Consider an inner product space  $(X, F, <\cdot, >)$ Def: Suppose SCX is a subset. Then,

 $S^{\perp} := \{ x \in X \mid x \perp S \}$  is the orthogonal complement

Exercise: (1)  $S^{\perp}$  is a subspace of X. (2)  $S^{\perp} = (span(S))^{\perp}$ 

(2) 
$$S^{\perp} = (span(s))^{\perp}$$

$$\mathcal{T}^{\perp} = \{0\}$$

Proposition: Let (X, F=R or C, <.,.) be an inner product space and finite-dimensional. Let MCX be a subspace. Then, X = M & M I direct sum" Yx €X, Jm, EM, Jm2 EM+ s.t. x=m,+m2 Note: MAM = {0}  $x \in M$  and  $x \in M^{\perp} = > \langle x, x \rangle = 0 \Leftrightarrow x = 0$ Proof: Let k=dim M and n=dim (X). . If k=n, then  $M^{\perp}=\{0\}$  (since M=X if  $\dim M=\dim X$ ), and we are done. · Assume 15k<n and let {y',...,yk} be a basis for M. Complete this basis to be basis & y',..., y' y the for X. Apply G.S. to produce v & v',..., vn 3 s.t. orthogonal vectors M = span & y', --, yk' = span & v', -, vk } and { v1, --, v2} 1 { vE+1, --, vn}.

 $X = \sup_{M} \frac{1}{1} = \sup_{M} \frac{1}{1} = \sup_{M} \frac{1}{1}$ i.e.  $\forall x \in X$ , we have  $x = \underset{M}{\text{and}} \frac{1}{1} = \sup_{M} \frac{1}{1}$   $\lim_{M \to \infty} \frac{1}{1} = \sup_{M} \frac{1}{1} = \sup_{M \to \infty} \frac{1}{1} = \sup_{M \to \infty}$ 

## Projection theorem

Projection Theorem: Let (X, R, <-,.>) be a finite-dim inner product space, MCX a subspace and x E X. Then, I a unique x EM s.t. & = argmin 11x-mll and x - & IM. (Revall pre-projection theorem. Bp: "FXEM s.t. & = argmin 11x-mil") => 2 is unique. ② p ←> ∃ 201/s.t. x-x L M.)

 $\frac{Proof}{}$ : Only thing left to show is existence of  $\hat{x}$  s.t.  $x-\hat{x} \perp M$  (the rest is shown as part of preprojection theorem)

Since X is fin. dimensional, this implies X-MBML

: Im EM and mEML s.t. x=m+m Then  $x-m=\widetilde{m}\in M^{\perp}$ , hence  $\hat{x}=m$ .

(X, R, <.,.), finite dim, M = spen & y1,...,yk} (sobspace) where {y',...,y'} are lin. indep...

Given x EX, seek an explicit formula for x.  $\hat{x} = argmin ||x-m||$ 

We know x-x 1 M = span & y', ---, y = 3 x-x 1 yi , 1 < i < k  $\langle x - \hat{x}, \hat{y}^i \rangle = 0$   $|\xi| \leqslant k$ 

 $\langle x, y^i \rangle - \langle \hat{x}, y^i \rangle = 0 , \quad 1 \leq i \leq \infty$ 

 $-(\langle x, y^i \rangle = \langle \hat{x}, y^i \rangle), \quad 1 \leq i \leq k$ Recall & EM (=)  $\hat{x} = \alpha_1 y^1 + ... + \alpha_k y^k$ 

 $\alpha_1 < y^1, y^i > + \dots + \alpha_k < y^k, y^i > = < x, y^i > 1 \le i \le k$ Lequations, one per i. We can write in matrix form:

Note that  $[G(y', y')]_{ij} := \langle y', y' \rangle$ 

where  $G^T = G$   $F = \mathbb{R}$ (because  $\langle y^i, y^i \rangle = \langle y^i, y^i \rangle$ )

$$G^T x = \beta$$

GT X = B | "Normal equations"

Cosolve for a, then get  $\hat{x} = \alpha_1 y^1 + ... + \alpha_k y^k$ 

Def: The Gram Matrix G and

the Gram determinant g(y',-,yk)=det6

Proposition: { y', --, y'} lin. indep. (y',..., y k) ≠ 0 (Gram motrix is invertible) (in which case, we can uniquely solve for  $x = 6^{-1}B.$ Proof: From the projection theorem and Dniqueness (details will be posted in an handout at the end of this lecture) Recall Regression: Ax=b, AEIRnxm b EIRn, We were solving for Q:= argmin || Ad-b||<sup>2</sup> (we claimed \argmin RRM A = [A, | A2 | --- | Am]  $A \alpha = \alpha_1 A_1 + \alpha_2 A_2 + \dots + \alpha_m A_m$  $rank(A) = M \leq M$ (overdeter mined)

Let's formulate using Normal Equations  $\chi = R^{\prime\prime}, \quad \mathcal{F} = R, \quad \langle x, y \rangle = x^{T}y$ M = span & A,,..., Am & seek REM s.t. IIR-bllg is minimized My yi's are Ai's. & = argmin 11 b -m1/2 mEM By normal equations: Since REM, R= X, A, + X2 A2+-- IXMAM GT. X = B  $G_{ij} = G_{i,j}^T = \langle A_i, A_j \rangle$ ,  $\beta_i = \langle b, A_i \rangle$  $\begin{bmatrix} A^T & A \end{bmatrix}_{ij} = \begin{bmatrix} \begin{bmatrix} A_1^T & \\ A_2^T \\ \vdots & A_m^T \end{bmatrix} \begin{bmatrix} A_1 & ... & ... & A_m \end{bmatrix} = A_1^T A_j = \langle A_i, A_j \rangle$  $G = G^T = A^T A$  scalar  $\beta_i = \langle b, A_i \rangle = b^T A_i = (b^T A_i)^T = A_i^T b$ 

$$B = \begin{bmatrix} A^T b \\ A^T b \end{bmatrix} = A^T b$$

$$C^T x = B \iff A^T A x = A^T b$$

$$= X = (A^T A)^T A^T b$$

$$A^T A \text{ is invertible because rank (A)=m}$$

$$A^T A x = A^T b$$

$$A = b$$

$$x = A^T b$$

$$A \in \mathbb{R}$$

$$A = A^T b$$

$$A \in \mathbb{R}$$

$$A = A^T b$$

$$A \in \mathbb{R}$$

$$A = A^T b$$

$$A = A^$$

## null (A) CIRM

$$B_i = b^T A_i$$

$$B = \begin{bmatrix} b^T A_1 \\ \vdots \\ b^T A_m \end{bmatrix} = \begin{bmatrix} A_1 & b \\ \vdots \\ A_m & b \end{bmatrix}$$

**Prop.**  $g(y^1, y^2, \dots, y^k) \neq 0 \Leftrightarrow \{y^1, \dots, y^k\}$  is linearly independent.

**Proof:**  $g(y^1, y^2, \dots, y^k) = 0 \leftrightarrow \exists \alpha \neq 0$  such that  $G^{\top} \alpha = 0$ .

From our construction of the normal equations,  $G^{\top}\alpha = 0$  if, and only if

$$\langle \alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k, y^i \rangle = 0 \quad i = 1, 2, \dots, k.$$

This is equivalent to

$$(\alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k) \perp y^i = 0 \ i = 1, 2, \dots, k$$

which is equivalent to

$$(\alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k) \perp \operatorname{span}\{y^1, \dots, y^k\} =: M$$

and thus

$$(\alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k) \in M^{\perp}.$$

Because  $\alpha_1 y^1 + \alpha_2 y^2 + \cdots + \alpha_k y^k \in M$ , we have that

$$(\alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k) \in M \cap M^{\perp}$$

and therefore

$$\alpha_1 y^1 + \alpha_2 y^2 + \dots + \alpha_k y^k = 0.$$

By the linear independence of  $\{y^1, \dots, y^k\}$ , we deduce that

$$\alpha_1 = \alpha_2 = \cdots = \alpha_k = 0. \square$$