

# Linear Algebra

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\* Please take a look at the  
typed lecture notes

## 4.4.2. Gram-Schmidt Process:

From  
last lecture

### Algorithm 4.4.10.

Given  $n$  linearly independent vectors  $a_1, \dots, a_n$  (they are a basis for  $C(A)$ ) the Gram-Schmidt Process computes an orthonormal basis  $q_1, \dots, q_n$  of  $C(A)$  in the following way

$$q_1 = \frac{a_1}{\|a_1\|}$$

for  $k=2, \dots, n$  do

$$q_k = \frac{a_k - \text{Proj}_{\text{Span}(q_1, \dots, q_{k-1})}(a_k)}{\|a_k - \text{Proj}_{\text{Span}(\dots)}(a_k)\|}$$

$$\text{Span}(q_1, \dots, q_{k-1}) = \text{Span}(a_1, \dots, a_{k-1})$$

Let  $A$   $m \times n$  with ind columns

Let  $Q$   $m \times n$  whose columns are the output of Gram-Schmidt on  $a_1, \dots, a_n$

$$C(Q) = C(A)$$

$$R := Q^T A$$

$R$  is  $n \times n$

$$R_{ij} = 0 \text{ if } i > j.$$

$$R = \begin{bmatrix} \times & & & \\ 0 & \times & & \\ & 0 & \times & \\ & & & \ddots \\ & & & & 0 & \dots \end{bmatrix}$$

upper triangular

$$R_{ij} = q_i^T a_j$$

$$q_i \perp \text{Span}(a_1, \dots, a_{i-1})$$

$$q_i \perp a_j \quad \checkmark$$

$$\Leftrightarrow R = Q^T A$$

$$QR = QQ^T A \\ = A$$

$QQ^T$  is the proj  
on  $C(Q)$

but  $C(A) = C(Q)$

$QQ^T$  is proj on  $C(A)$ .

(Def: 4.4.12)

QR-decomposition.

$A$   $m \times n$  with ind columns.

$$A = QR$$

$$Q^T Q = I$$

$R$  is upper triangular.

Fact 4.4.13.

$A$   $m \times n$  with ind columns.

$$A = QR$$

$$C(A) = C(Q)$$

Proj on  $C(A)$  is the same as proj on  $C(Q)$

$$\text{Proj}_{C(A)}(b) = QQ^T b$$

→ Least Squares of  $Ax = b$

is the solution to  $AA^T \hat{x} = A^T b$

$$A = QR$$

$$\hookrightarrow (QR)^T (QR) \hat{x} = (QR)^T b$$

$$R^T \underbrace{Q^T Q}_{=I} R \hat{x} = R^T Q^T b$$

$$R^T R \hat{x} = R^T Q^T b$$

$$R \hat{x} = Q^T b$$

(since  $R^T$  is inv)

can be solved "backsubstitution"

since  $R$  is up. triangular

what is  $N(R)$ ?

if  $x \in N(R)$

$$QRx = 0$$

$$Ax = 0$$

$x \in N(A)$

$$\Rightarrow x = 0$$

so  $N(R) = \{0\}$

$R$  is inv

$R^T$  is inv

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Guiding Question: How do we "invert" a matrix that's not invertible?

Goal: Given a matrix  $A$   $m \times n$  we want to define / construct  $A^+$  pseudo-inverse of  $A$ .

$$A : x \longmapsto Ax = b$$

$\in \mathbb{R}^n$                        $\in \mathbb{R}^m$

Goal  $A^+ : b \rightarrow x$  ???

why not possible:  
(i) maybe for some  $b \in \mathbb{R}^m$  there is no  $x$  s.t.  $Ax=b$   
(ii) for some  $b \in \mathbb{R}^m$  there may be many  $x \in \mathbb{R}^n$  such that  $Ax=b$   
which one do we pick?  
(iii) should be of the form  $\hat{x} = A^+ b$   
minim.

Def 4.5.1

$A \in \mathbb{R}^{m \times n}$

ind columns  $\Leftrightarrow \text{rank}(A) = n$

$\Leftrightarrow$  full column rank

$[A]$

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

$A^+$  takes  $b$  to the least squares sol. of  $Ax=b$

Prop:

$$A^+ A = I$$

$$[A^+] [A] = [I]$$

$$[A] [A^+] \neq I$$

Proof:

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





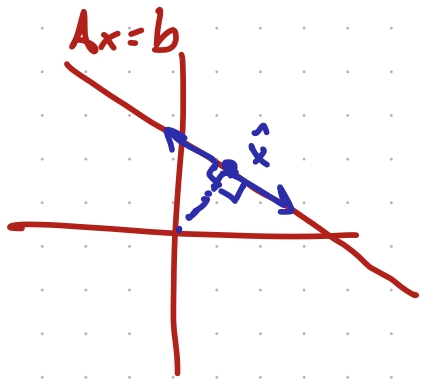
$$A \quad m \times n$$

$$\text{rank}(A) = m$$

For any  $b \in \mathbb{R}^m$

$Ax = b$  has a solution  $x$ .

Any  $x \in N(A)$  is a solution to  $Ax = 0$ .



$$\begin{bmatrix} \phantom{x} \\ \phantom{x} \\ \phantom{x} \end{bmatrix} = 0$$

natural:  $\hat{x}$  to be the solution of

$$\begin{aligned} \min \|x\|^2 \\ \text{s.t. } Ax = b \end{aligned} \quad (10)$$

both  $x_1$  and  $x_2$  are solutions of  $Ax = b$

$$A(x_1 - x_2) = Ax_1 - Ax_2 = b - b = 0 \Rightarrow x_1 - x_2 \in N(A)$$

If  $y \in N(A)$  and  $Ax = b$  then  $A(x+y) = Ax + Ay = b$

given  $x_0$  s.t.  $Ax_0 = b$  all solutions of

$Ax = b$  are of the form  $x_0 + y$  for  $y \in N(A)$ .

(Careful proof on the notes)

$\hat{x}$  unique sol of (10) is  $\hat{x}$  s.t.

$$Ax = b \quad \text{and} \quad x \in N(A)^\perp = C(A^T)$$

Prop: for  $A$  full row rank

$\hat{x} = A^+ b$  is the unique sol of (10).

Proof:  $\rightarrow A \hat{x} = A A^+ b = b \quad \checkmark$

need to show

$$\hat{x} \in N(A)^\perp \Leftrightarrow \hat{x} \in C(A^T)$$

$$\hat{x} = A^+ b = \underline{A^T (A A^T)^{-1}} b \in C(A^T) \quad \checkmark$$

What if  $A$  is neither full row or column rank?

idea:

$\hat{x} = A^+ b$  to be sol of  $\min \|x\|$   
s.t.  $A^T A x = A^T b$