

So, if $A^T A$ has an inverse,

then the columns of A are linearly independent.

$$\begin{aligned} Ax &= 0 \\ x &= 0 \end{aligned}$$

A , columns are linearly independent

$$A^+ = (A^T A)^{-1} A^T \quad \begin{array}{l} \text{Moore-Penrose inverse} \\ \text{Pseudoinverse} \end{array}$$

Claim: A^+ is a left inverse of A .

$$\begin{aligned}
 A^+ A &= ((A^T A)^{-1} A^T) A \\
 &= (A^T A)^{-1} (A^T A) \\
 &= I
 \end{aligned}$$

$$\begin{aligned}
 A^T &= (A^T A)^{-1} A^T \\
 (A^T)^{-1} &= (A^{-1})^T \\
 &= A^{-T}
 \end{aligned}$$

What is A^+ if A is square?

A^+ is a left inverse of A .

So A^+ had better be A^{-1} !

$$\begin{aligned}
 (A^T A)^{-1} A^T &= A^{-1} (A^T)^{-1} A^T \\
 &= A^{-1} I \\
 &= A^{-1}
 \end{aligned}$$

If

$$Ax = b$$

$$\underline{A^+ A}x = A^+ b$$

$$Ix = A^+ b$$

$$x = A^+ b$$

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

We can compute $A^+ b$
using the QR factorization
of A

$$A = QR$$

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

$$= R^T Q^T$$

$$\begin{aligned} A^T A &= R^T Q^T Q R \\ &= R^T R \end{aligned}$$

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

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

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

$$= R^{-1} (R^T)^{-1} R^T Q^T$$

$$= R^{-1} I Q^T$$

$$= R^{-1} Q^T$$

$A^+ b$

$$A^+ = R^{-1} Q^T$$

$$z = A^+ b \quad \text{means} \quad z = R^{-1} Q^T b$$

$$Rz = \underbrace{RR^T}_{I} Q^T b$$

$$Rz = Q^T b$$

$$z = A^+ b \iff Rz = Q^T b$$

So: to compute $A^+ b$

- 1) form $Q^T b$
- 2) solve $Rz = Q^T b$
using back substitutions

Then $z = A^+ b$.

If A is square and I give you $A = QR$

and want you to solve $Ax = b$ for x

What do you do?

$$QRx = b$$

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

$$R x = Q^T b$$

Now solve for x by back substitution,

$$A \quad m \times n$$

$$A = [a_1 \cdots a_n]$$

$$Q^T Q = I$$

cols of Q

are orthogonal

$$a_1 \cdots a_n \rightarrow q_1 \cdots q_n$$

\leftarrow^n

$$m \begin{bmatrix} q_1 & \cdots & q_n \end{bmatrix} \xrightarrow{n} \begin{bmatrix} r_{11} & * \\ 0 & \ddots & r_{nn} \end{bmatrix}$$

$$a_1 \quad \widehat{q}_1 = a_1 \quad q_1 = \frac{\widehat{q}_1}{\|\widehat{q}_1\|}$$

$$\tilde{q}_2 = a_2 - (q_1^T a_2) q_1$$



$$a_2 = \tilde{q}_2 / \|\tilde{q}_2\|$$

$$\tilde{q}_2 = \|\tilde{q}_2\| \cdot q_2$$

$$a_2 = \|\tilde{q}_2\| q_2 + (q_1^T a_2) q_1$$

$$R = \begin{bmatrix} \|\widehat{q}_1\| & (q_1^T a_2) & (q_1^T a_3) \\ 0 & \|\tilde{q}_2\| & (q_2^T a_3) \\ \vdots & 0 & \|\tilde{q}_3\| \end{bmatrix}$$

What about wide matrices?

$$m \begin{bmatrix} n \\ \vdots \\ 1 \end{bmatrix} \leftarrow m < n$$

right inverse.

We will build a left inverse for A^T

We'll assume that A has linearly independent rows.

Then A^T has linearly independent columns.

So $(A^T)^T A^T = A A^T$ is invertible.

Now we define

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

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

$$Ax = b$$

$$x = A^+ b$$

Then $Ax = A A^+ b = I b = b$.

[] ← more than one solution.

Next class: How to compute A^+ using QR factorization for underdetermined instances.