

pseudo inverse of A $n \times m$

= the matrix A^+ so that $Ax = b$ has "optimal" solution $x = A^+b$
smallest least-squares

- if A invertible, $A^+ = A^{-1}$
- if A has full column rank, $A^+ = (A^T A)^{-1} A^T$
- if $A = U \Sigma V^T$ (SVD), then

$$(AB)^{-1} = B^{-1}A^{-1}$$

$$A^+ = V \Sigma^+ U^T$$

Why?

$$Ax = b \Leftrightarrow U \Sigma V^T x = b$$

$x = Vy$
 x, y same norm

$$\Sigma y = U^T b$$

optimal sol: $y = \Sigma^+ U^T b$
 $x = Vy = \underbrace{V \Sigma^+ U^T}_{A^+} b$

EG $A = \begin{bmatrix} 1 & -1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$

Option 1: A full column rank

$$\Rightarrow A^+ = (A^T A)^{-1} A^T = \frac{1}{3} \begin{bmatrix} 1 & 1 & 2 \\ -1 & 2 & 1 \end{bmatrix}$$

Option 2: $A = U \Sigma V^T$

$$\text{with } U = \begin{bmatrix} -2/\sqrt{6} & 0 & -1/\sqrt{3} \\ 1/\sqrt{6} & 1/\sqrt{2} & -1/\sqrt{3} \\ -1/\sqrt{6} & 1/\sqrt{2} & 1/\sqrt{3} \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sqrt{3} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \quad V = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\Rightarrow A^+ = V \Sigma^+ U^T$$

$$\text{with } \Sigma^+ = \begin{bmatrix} 1/\sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$