

Kernel trick:

$$J = \|Ax - b\|_2^2 + \lambda \|x - x^d\|_2^2$$

where $x^d \in \mathbb{R}^n$
is given vector.

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

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

$$x^d \in \mathbb{R}^n$$

} given and

$m < n$.

$$\lambda > 0$$

(A is a wide matrix)

$$A = \begin{bmatrix} & \end{bmatrix}_{m \times n}$$

$$\tilde{A} \in \mathbb{R}^{(m+n) \times n}$$

corresponding to J

$$\tilde{A} = \underbrace{\begin{bmatrix} A \\ \sqrt{\lambda} I \end{bmatrix}}_{n \text{ columns}} \begin{matrix} \} m \text{ rows} \\ \} n \text{ rows} \end{matrix}$$

Note that \tilde{A} always
has full column
rank.

$$\hat{x} = (A^T A + \lambda I)^{-1} (A^T b + \lambda x^d)$$

$$= (A^T A + \lambda I)^{-1} (A^T b + (\lambda I + A^T A) x^d - A^T A x^d)$$

$$= (A^T A + \lambda I)^{-1} A^T (b - A x^d) + x^d$$

claim: $(A^T A + \lambda I)^T A^T = A^T (A A^T + \lambda I)^{-1}$

observe: $A^T (A A^T + \lambda I) = (A^T A + \lambda I) A^T$

Multiply by $(A^T A + \lambda I)^T$ on left & $(A A^T + \lambda I)^{-1}$ on the right of the above eqⁿ:

$$\hat{x} = A^T (A A^T + \lambda I)^{-1} (b - A x^d) + x^d$$

$$\begin{aligned} A &= \begin{bmatrix} & \end{bmatrix} \quad \text{small} \times \text{large} \\ A^T A &= \begin{bmatrix} \end{bmatrix} \quad \text{large} \times \text{large} \\ A A^T &= \begin{bmatrix} \end{bmatrix} \quad \text{small} \times \text{small} \end{aligned}$$

Digression:

No inverse required.

$A \in \mathbb{R}^{n \times n}$ invertible & $b \in \mathbb{R}^n$ given.
Solve $Ax = b$

Obtain QR decomposition of A .

$A = QR$ where

$Q \in \mathbb{R}^{n \times n}$
is orthogonal

$$(Q^T Q = Q Q^T = I)$$

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

is an upper triangular matrix.

$$Ax = b$$

$$\Rightarrow Q^T Ax = Q^T b$$

$$\Rightarrow Rx = Q^T b$$