



Lecture 16 : Derivatives of Inverse and Singular Values

$$\frac{dA^{-1}}{dt} = -A^{-1} \frac{dA}{dt} A^{-1} \quad \frac{dx}{dt} = y^T(t) \frac{dA}{dt} x(t)$$

$$\frac{d}{dt} A^2 = ? \quad \frac{d\sigma}{dt} = ?$$

NOT $2A \frac{dA}{dt}$

$$\frac{(A + \Delta A)^2 - A^2}{\Delta t} = \frac{A(\Delta A) + (\Delta A)A + (\Delta A)^2}{\Delta t}$$

$$\frac{dA^2}{dt} = A \frac{dA}{dt} + \frac{dA}{dt} A$$

Derivative of $\frac{d\sigma}{dt}$

$$\sigma(t) = u^T A v$$

$$(\sigma v)^T v = (A u)^T v$$

$$\begin{aligned} \frac{d\sigma}{dt} &= \frac{du^T}{dt} A v + u^T \frac{dA}{dt} v + u^T A \frac{dv}{dt} \\ &= \sigma \frac{du^T}{dt} u + u^T \frac{dA}{dt} v + \sigma u^T \frac{dv}{dt} \rightarrow 0 \\ &= u^T \frac{dA}{dt} v \end{aligned}$$

$$Av = \sigma u$$

$$u^T u = 1$$

$$v^T v = 1$$

$$\frac{d}{dt} u^T u + u^T \frac{du}{dt} = \frac{d}{dt} (1) = 0$$

$$x^T y = y^T x$$

$$S \quad \lambda_1 \geq \lambda_2 \geq \dots$$

$$S + \underbrace{\Theta u u^T}_{\text{positive change}} \quad \mu_1 \geq \mu_2 \geq \dots$$

$$\mu_1 \geq \underbrace{\lambda_1}_{\text{old}} \geq \underbrace{\mu_2}_{\text{new}} \geq \lambda_2 \geq \dots$$

Weyl's inequality for symmetric S, T

$$\lambda_{i+j-1}(S+T) \leq \lambda_i(S) + \lambda_j(T)$$

let $j=1$: $\lambda_i(S+T) \leq \lambda_i(S) + \lambda_1(T)$
 \downarrow
 λ_{\max}

let $j=2$: $\lambda_{i+1}(S+T) \leq \lambda_i(S) + \lambda_2(T)$

Nuclear Norm $\|A\|_n = \sigma_1 + \dots + \sigma_r$
 (like $\|v\|_1$)

Completion (Netflix)

$$A = \begin{bmatrix} 3 & 2 & ? \\ 1 & ? & ? \\ 2 & 4 & 6 \end{bmatrix}$$

Problem : put in missing data

Idea : Numbers that minimize $\|A\|_n$ are a good choice.

$\|v\|_1$ from $\|v\|_0 = \# \text{ nonzeros in } v$
 $\hookrightarrow \sum |v_i|$

$\|A\|_n$ from $\|A\|_0 = \text{rank}(A)$