

Mathematics Basics

BASIC DEFINITION

$$Ax = b$$

Range \longrightarrow has solution?

Kernel \longrightarrow how many?

SVD

($\sigma_i \geq 0$)

SVD of A is:

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

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

$$\Sigma = \begin{bmatrix} \sigma_1 & & \\ & \ddots & \\ & & \sigma_n \end{bmatrix}$$

NORMS

$$\langle u, v \rangle = \sum u_i v_i = v^T u$$

$$\|u\|_2 = \sqrt{\langle u, u \rangle}$$

p-norms, 1-norm, ...

$$\|u\|_1 = \sum |u_i|$$

SPECIAL EIGEN-DECOMPOSITION

$$A = X \Lambda X^{-1}$$

if A symmetric: $B = X \Lambda X^T$

$$Ax = A(\sum \langle x, v_i \rangle v_i)$$

$$= \sum \langle x, v_i \rangle A v_i$$

$$= \sum \lambda_i \langle x, v_i \rangle v_i$$

$$x^T A x = \sum \lambda_i \langle x, v_i \rangle^2 \leq \sum \lambda_{\max} \langle x, v_i \rangle^2 \leq \lambda_{\max} \sum \langle x, v_i \rangle^2 = \lambda_{\max} \|x\|^2$$

$$\sum \langle x, v_i \rangle^2 = \|x\|^2$$

APPROXIMATION

Taylor Approximation:

$$f(x + \varepsilon) \approx f(x) + \langle \nabla f(x), \varepsilon \rangle + \frac{1}{2} \varepsilon^T D^2 f(x) \varepsilon + o(\|\varepsilon\|^2)$$

$$\left[\frac{\partial^2 f}{\partial x_i \partial x_j} \right]_{i,j}$$

$$= f(x) + \langle \nabla f(x), \varepsilon \rangle + \frac{1}{2} \varepsilon^T \underbrace{D^2 f(x)}_{\text{Hessian}} \varepsilon$$

ε random

$$\textcircled{1} \begin{array}{c} \varepsilon \\ \uparrow \\ \varepsilon \\ \downarrow \\ \varepsilon \end{array}$$

$$\textcircled{2} \mathbb{E}[f(x + \varepsilon)] = ?$$

Expected value

$$\mathbb{E}[f(x + \varepsilon)] = \mathbb{E}[f(x) + \langle \nabla f(x), \varepsilon \rangle + \frac{1}{2} \varepsilon^T D^2 f(x) \varepsilon]$$

$$= f(x) + \mathbb{E}[\langle \nabla f(x), \varepsilon \rangle] + \frac{1}{2} \mathbb{E}[\varepsilon^T D^2 f(x) \varepsilon]$$

$$= f(x) + \langle \nabla f(x), \mathbb{E}[\varepsilon] \rangle + \frac{1}{2} \mathbb{E}[\varepsilon^T D^2 f(x) \varepsilon]$$

$$= f(x) + 0 + \frac{1}{2} \mathbb{E}[\varepsilon^T D^2 f(x) \varepsilon]$$

Assume that for all $y \in \mathbb{R}^n$, $D^2 f(x)$ has the property that $\lambda_{\max} \leq \bar{\lambda}$, $\lambda_{\min} \leq \underline{\lambda}$

$$\Delta \| \varepsilon \|^2 \leq \varepsilon^T D^2 f(x) \varepsilon \leq \bar{\lambda} \| \varepsilon \|^2$$

$$f(x) \leq f(x) + \underbrace{\Delta \mathbb{E}[\| \varepsilon \|^2]}_{=1} \leq \mathbb{E}[f(x + \varepsilon)] \leq f(x) + \underbrace{\bar{\lambda} \mathbb{E}[\| \varepsilon \|^2]}_{=1} \rightarrow 2^{\text{nd}} \text{ Moment (= variance)}$$

