

Exercise 01 - Math Refresher

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Linear Algebra

Problem 1

$$f(x, y, z) = x^T A y + B x - y^T C z D - y^T E^T y + F$$

$$\mathbb{R}^M, \mathbb{R}^N, \mathbb{R}^{P \times Q} = \mathbb{R}^{1 \times M} \mathbb{R}^{M \times N} \mathbb{R}^N \quad \mathbb{R}^{1 \times M} \mathbb{R}^M = \mathbb{R}^{1 \times N} \mathbb{R}^{N \times P} \mathbb{R}^{P \times Q} = \mathbb{R}^{1 \times N} \mathbb{R}^{N \times N} \mathbb{R}^N = \mathbb{R}^1$$

$$A = \mathbb{R}^{M \times N}$$

$$D = \mathbb{R}^Q$$

$$B = \mathbb{R}^{1 \times M}$$

$$E = \mathbb{R}^{N \times N}$$

$$C = \mathbb{R}^{N \times P}$$

$$F = \mathbb{R}^1$$

Problem 2

$$\sum_i^N \sum_j^N x_i x_j M_{ij} =$$

$$\begin{aligned} x_1 x_1 M_{11} + x_1 x_2 M_{12} + x_1 x_3 M_{13} + \dots &= x_1 (x_1 M_{11} + x_2 M_{12} + x_3 M_{13}) \\ + x_2 x_1 M_{21} + x_2 x_2 M_{22} + x_2 x_3 M_{23} + \dots &= x_2 (x_1 M_{21} + x_2 M_{22} + x_3 M_{23}) \\ + x_3 x_1 M_{31} + x_3 x_2 M_{32} + x_3 x_3 M_{33} + \dots &= x_3 (x_1 M_{31} + x_2 M_{32} + x_3 M_{33}) \end{aligned}$$

$$\Rightarrow \sum_i^N x_i \sum_j^N x_j M_{ij} = \sum_i^N x_i \cdot (M \cdot x) = x^T M x$$

Problem 3

a) - $\text{rank}(A) = 4$

- $\det(A) \neq 0$

- $\ker(A) = \{0\}$

b) $|\det(A)| = \prod_i \lambda_i = -5 \cdot 0 \cdot 1 \cdot 1 \cdot 3 = 0 \parallel \text{Rank}(A) = \# \text{ Non } 0 \text{ EV}$

\Rightarrow No unique solution

Problem 4

$$BA = AB = I$$

$$\Rightarrow B = A^{-1}$$

$$\Rightarrow A \text{ invertierbar} \Rightarrow \det(A) \neq 0 \Rightarrow \forall i: \lambda_i \neq 0$$

Problem 5

$$x^T A x \geq 0$$

$$= x^T \lambda x \geq 0$$

$$= \lambda x^T x \geq 0$$

$$= \lambda \sum x_i^2 \geq 0$$

$$= \lambda \geq 0$$

Problem 6

$$B = A^T A \Rightarrow Bx = \lambda_B x = A^T A x = \lambda_A \lambda_A x = \lambda_A^2 x$$

$$\Rightarrow \lambda_B = \lambda_A^2$$

$$\Rightarrow \lambda_B \text{ immer} \geq 0$$

$$\Rightarrow B \text{ PSD}$$

Calculus

Problem 7

- a) Unique solution: global minimum $a > 0$
 Infinitely many solutions: many local minima $a = b = 0$
 No solution: No lower bound $a < 0$

b) $\min_{x \in \mathbb{R}} f(x) = f'(x) = 0$

$$f'(x) = ax + b = 0$$

$$x^* = \arg\min_{x \in \mathbb{R}} f(x) = \frac{-b}{a}$$

Missing

Problem 8

a) $g(x) = \frac{1}{2} x^T A x + b^T x + c = \frac{1}{2} (x_1 x_2 \dots x_n) A \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} + b^T \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} + c$

$$\nabla^2 g(x) = \begin{pmatrix} \frac{\partial^2 g}{\partial x_1 \partial x_1} & \dots & \frac{\partial^2 g}{\partial x_1 \partial x_n} \\ \vdots & & \vdots \\ \frac{\partial^2 g}{\partial x_n \partial x_1} & \dots & \frac{\partial^2 g}{\partial x_n \partial x_n} \end{pmatrix}$$

$$\begin{bmatrix} A_{11} & A_{12} & \dots & 0 \\ 0 & A_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & A_{nn} \end{bmatrix}$$

$$\nabla^2 g(x) = \begin{bmatrix} \frac{\partial^2 g}{\partial x_i \partial x_j} \end{bmatrix}$$

$$\frac{1}{2} \sum_i \sum_j A_{ij} x_i x_j + \sum_i (b^T)_i x_i + c = g(x)$$

$$g'(x) = \begin{cases} A_{ij} & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

Missing

$$\forall i: A_{ii} \neq 0 \Rightarrow \det(A) \neq 0 \quad \text{unique } \nabla^2$$

Missing

b) $g(x) = \frac{1}{2} x^T A x + b^T x + c = \frac{1}{2} \lambda \sum x_i^2 + \sum b_i^T x_i + c$

$$g''(x) = \lambda \Rightarrow \text{Curvature same in all dimensions}$$

$$\Rightarrow \text{Global minimum} \Rightarrow \text{Convex Problem}$$

Missing

$$\Rightarrow \text{Negative EV} \Rightarrow \text{Only saddle point}$$

c) $x^* = \arg\min_{x \in \mathbb{R}^n} g(x) \Rightarrow g'(x) = 0$

$$g'(x) = \frac{\partial}{\partial x} \left(\frac{1}{2} x^T A x + b^T x + c \right)$$

$$= \frac{1}{2} x^T A \frac{\partial x}{\partial x} + b^T \frac{\partial x}{\partial x} + c \cdot \frac{\partial 1}{\partial x}$$

$$= \frac{1}{2} x^T A + b^T + 0$$

$$= \frac{1}{2} x^T A + b^T = 0$$

$$x^T A = -2b^T$$

$$x^T = -2b^T A^{-1}$$

$$x = (-2b^T A^{-1})^T$$

Missing

formalisieren
mit $\&$ =

\Rightarrow Macht alle = untereinander