

Problem 1

a) For a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$, the gradient is

$$\nabla f(x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} = \begin{bmatrix} \frac{\partial f}{\partial x} \end{bmatrix}^T$$

The direction of the gradient vector can be interpreted as pointing in the direction of steepest ascend.

b) For a function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$, the Jacobian is

$$\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

c) $f(x): \mathbb{R}^n \rightarrow \mathbb{R} \Leftrightarrow x \in \mathbb{R}^n, f \in \mathbb{R}$

Size of ∇f : $n \times 1$

d) $f(x): \mathbb{R}^n \rightarrow \mathbb{R}^m \Leftrightarrow x \in \mathbb{R}^n, f \in \mathbb{R}^m$

Size of $\frac{\partial f}{\partial x}$: $m \times n$

Problem 2

$$f(x) = Ax = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{bmatrix}$$

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

a)

$$\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = A$$

This is the Jacobian

b) $\frac{\partial(f(x))}{\partial x} = A \frac{\partial x}{\partial x} = A$

move constants out

Problem 3

$$f(x, y) = x^T G y$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad 2 \times 1 \quad G = \begin{bmatrix} g_{11} & g_{12} & g_{13} \\ g_{21} & g_{22} & g_{23} \end{bmatrix} \quad 2 \times 3 \quad y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} \quad 3 \times 1$$

a) $f(x, y) : (2 \times 1)^T \cdot 2 \times 3 - 3 \times 1 = 1 \times 1 \Leftrightarrow \text{scalar}$

$\nabla_x f(x, y)$: gradient wrt. x

$$= \left[\frac{\partial f}{\partial x} \right]^T$$

No, $\nabla_x f(x, y) = \left[\frac{\partial f(x, y)}{\partial x} \right]^T$

$$b) f(x, y) = [x_1 \ x_2] G \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} g_{11} y_1 + g_{12} y_2 + g_{13} y_3 \\ g_{21} y_1 + g_{22} y_2 + g_{23} y_3 \\ y_3 \end{bmatrix}$$

$$= x_1 (g_{11} y_1 + g_{12} y_2 + g_{13} y_3) + x_2 (g_{21} y_1 + g_{22} y_2 + g_{23} y_3)$$

$$\nabla_x f(x, y) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} g_{11} y_1 + g_{12} y_2 + g_{13} y_3 \\ g_{21} y_1 + g_{22} y_2 + g_{23} y_3 \end{bmatrix} = \underline{\underline{G}} \underline{\underline{y}}$$

$$c) f(x, y) = y_1 (x_1 g_{11} + x_2 g_{21}) + y_2 (x_1 g_{12} + x_2 g_{22}) + y_3 (x_1 g_{13} + x_2 g_{23})$$

$$\nabla_y f(x, y) = \begin{bmatrix} \frac{\partial f}{\partial y_1} \\ \frac{\partial f}{\partial y_2} \\ \frac{\partial f}{\partial y_3} \end{bmatrix} = \begin{bmatrix} x_1 g_{11} + x_2 g_{21} \\ x_1 g_{12} + x_2 g_{22} \\ x_1 g_{13} + x_2 g_{23} \end{bmatrix} = \underline{\underline{x}}^T \underline{\underline{G}} = \underline{\underline{G}}^T \underline{\underline{x}}$$

$$d) f(x) = \underline{\underline{x}}^T \underline{\underline{H}} \underline{\underline{x}} \quad x \in \mathbb{R}^n \quad H \in \mathbb{R}^{n \times n}$$

$$\nabla f(x) = \nabla_{\underline{\underline{x}}} f(x) + \nabla_x f(x) = \underline{\underline{H}} \underline{\underline{x}} + \underline{\underline{H}}^T \underline{\underline{x}} = (\underline{\underline{H}} + \underline{\underline{H}}^T) \underline{\underline{x}}$$

If H is symmetric: $H = H^T$

$$\Rightarrow \nabla f(x, y) = \underline{\underline{2Hx}}$$

Problem 4

- $\lambda(\dots) \in \mathbb{R}$
- $G = G^T$ (symmetric)
- $x \in \mathbb{R}^n : n \times 1$

$$\begin{aligned} \mathcal{L}(x, \lambda, \mu) &= x^T G x + \lambda^T (Cx - d) + \mu^T (Ex - h) \\ &= x^T G x + \lambda^T C x - \lambda^T d + \mu^T E x - \mu^T h \end{aligned}$$

a) $\nabla_x \mathcal{L}(x, \lambda, \mu) = \underline{2Gx + \lambda^T C + \mu^T E}$

b) $\nabla_\lambda \mathcal{L}(x, \lambda, \mu) = \underline{Cx - d}$

c) $\nabla_\mu \mathcal{L}(x, \lambda, \mu) = \underline{Ex - h}$