

OPTIMIZATION. HOMEWORK 1

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- (1) Let $f_1(x_1, x_2) = x_1^2 - x_2^2$, $f_2(x_1, x_2) = 2x_1x_2$. Represent the level sets associated with $f_1(x_1, x_2) = 12$ and $f_2(x_1, x_2) = 16$ on the same figure using Python. Indicate on the figure, the points $\mathbf{x} = [x_1, x_2]^T$ for which $f(\mathbf{x}) = [f_1(x_1, x_2), f_2(x_1, x_2)]^T = [12, 16]^T$.
- (2) Consider the function $f(\mathbf{x}) = (\mathbf{a}^T \mathbf{x})(\mathbf{b}^T \mathbf{x})$, where \mathbf{a} , \mathbf{b} , and \mathbf{x} are n -dimensional vectors.
 - Compute the gradient $\nabla f(\mathbf{x})$ and the Hessian $\nabla^2 f(\mathbf{x})$.
- (3) Compute the gradient of

$$f(\theta) \stackrel{def}{=} \frac{1}{2} \sum_{i=1}^n [g(\mathbf{x}_i) - g(\mathbf{A}\mathbf{x}_i + \mathbf{b})]^2$$

with respect to θ , where $\theta = [a_{11}, a_{12}, a_{21}, a_{22}, b_1, b_2]^T$, $\mathbf{x}_i \in \mathbb{R}^2$, $\mathbf{A} \in \mathbb{R}^{2 \times 2}$, $\mathbf{b} \in \mathbb{R}^2$ are defined as follows

$$\begin{aligned} \mathbf{A} &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \\ \mathbf{b} &= [b_1, b_2]^T \end{aligned}$$

and $g : \mathbb{R}^2 \rightarrow \mathbb{R} \in \mathcal{C}^1$.

- (4) Let $f(r, \theta)$ be $\mathbb{R}^2 \rightarrow \mathbb{R}$ with $r = \sqrt{x^2 + y^2}$ and $\theta = \arctan \frac{y}{x}$. Compute $\frac{\partial f}{\partial x}$ and $\frac{\partial f}{\partial y}$
- (5) The directional derivative $\frac{\partial f}{\partial \mathbf{v}}(x_0, y_0, z_0)$ of a differentiable function f are $\frac{3}{\sqrt{2}}$, $\frac{1}{\sqrt{2}}$ and $-\frac{1}{\sqrt{2}}$ in the directions of vectors $[0, \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}]^T$, $[\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}]^T$ and $[\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0]^T$. Compute $\nabla f(x_0, y_0, z_0)$.
- (6) Show that the level curves of the function $f(x, y) = x^2 + y^2$ are orthogonal to the level curves of $g(x, y) = \frac{y}{x}$ for all (x, y) .
- (7) Let f, g, h be differentiable functions, with $f : \mathbb{R}^n \rightarrow \mathbb{R}^3$, $g : \mathbb{R}^n \rightarrow \mathbb{R}^3$ and $h : \mathbb{R}^n \rightarrow \mathbb{R}$

$$h(\mathbf{x}) = f(\mathbf{x})^T g(\mathbf{x})$$

show that

$$Dh(\mathbf{x}) = f(\mathbf{x})^T Dg(\mathbf{x}) + g(\mathbf{x})^T Df(\mathbf{x})$$

(8) Consider the **induced matrix norm**

$$\|\mathbf{A}\|_p = \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|\mathbf{Ax}\|_p}{\|\mathbf{x}\|_p}$$

where $\|\cdot\|_p$ is the ℓ_p norm, ie

$$\|\mathbf{x}\|_p = (|x_1|^p + |x_2|^p + \cdots + |x_n|^p)^{1/p}$$

Show that

$$\|\mathbf{AB}\|_p \leq \|\mathbf{A}\|_p \|\mathbf{B}\|_p$$