Mathematical Concepts 1

Solution 1:

Gradient

- (a) The gradient $\nabla f(\mathbf{x}) = (2x_1 + x_2, x_2 + x_1)^{\top}$ is continuous $\Rightarrow f \in \mathcal{C}^1$.
- (b) The direction of greatest increase is given by the gradient, i.e., $\nabla f(1,1) = (3,2)^{\top}$.
- (c) Let $\mathbf{v} \in \mathbb{R}^2$ be a direction with fixed length $\|\mathbf{v}\|_2 = r > 0$. The directional derivative $D_{\mathbf{v}}f(\mathbf{x}) = \nabla f(\mathbf{x})^{\top}\mathbf{v} = \|\nabla f(\mathbf{x})\|_2 \|\mathbf{v}\|_2 \cos(\theta) = \|\nabla f(\mathbf{x})\|_2 r \cos(\theta)$. This becomes minimal if $\theta = \pi$, i.e., if \mathbf{v} points in the opposite direction of $\nabla f \Rightarrow \mathbf{v} = -\nabla f(\mathbf{x})$ if $r = \|\nabla f(\mathbf{x})\|_2$. Here, the direction of greatest decrease is given by $-\nabla f(1, 1) = (-3, -2)^{\top}$.
- (d) $D_{\mathbf{v}}f(\mathbf{x}) = \nabla f(1,1)^{\top}\mathbf{v} \stackrel{!}{=} 0 \Rightarrow (3,2) \cdot \mathbf{v} = 0 \iff \mathbf{v} = \alpha \cdot (-2,3)^{\top} \text{ with } \alpha \in \mathbb{R} \text{ and } \alpha \neq 0.$
- (e) When we differentiate both sides of the equation $f(\tilde{\mathbf{x}}(t)) = f(1,1)$ w.r.t. t we arrive at $\frac{\partial f(\tilde{\mathbf{x}}(t))}{\partial t} = 0$. Via the chain rule it follows that $\frac{\partial f}{\partial \tilde{\mathbf{x}}} = 0$.
- (f) The gradient is orthogonal to the tangent line of the level curves.

Solution 2:

Convexity

(a) Let $x, y \in \mathbb{R}$ and $t \in [0, 1]$ then it holds that

$$(f+g)(x+t(y-x)) = f(x+t(y-x)) + g(x+t(y-x))$$

$$\leq f(x) + t(f(y) - f(x)) + g(x) + t(g(y) - g(x))$$

$$= f(x) + g(x) + t(f(y) + g(y) - (f(x) + g(x)))$$

$$= (f+g)(x) + t((f+g)(y) - (f+g)(x)).$$
(f, g are convex)

(b) Let $x, y \in \mathbb{R}$ and $t \in [0, 1]$ then it holds that

$$(g \circ f)(x + t(y - x)) = g(f(x + t(y - x)))$$

$$\leq g(f(x) + t(f(y) - f(x))) \qquad (g \text{ is non-decreasing, } f \text{ is convex})$$

$$\leq g(f(x)) + t(g(f(y)) - g(f(x)))) \qquad (g \text{ is non-decreasing, } f \text{ is convex})$$

$$= (g \circ f)(x) + t((g \circ f)(y) - (g \circ f)(x)).$$

Solution 3:

Convexity

Consider the bivariate function $f: \mathbb{R}^2 \to \mathbb{R}, (x_1, x_2) \mapsto \exp(\pi \cdot x_1) - \sin(\pi \cdot x_2) + \pi \cdot x_1 \cdot x_2$

(a)
$$\nabla f(\mathbf{x}) = \pi \cdot (\exp(\pi x_1) + x_2, -\cos(\pi x_2) + x_1)^{\top}$$

(b)
$$\nabla^2 f(\mathbf{x}) = \pi \cdot \begin{pmatrix} \pi \exp(\pi x_1) & 1 \\ 1 & \pi \sin(\pi x_2) \end{pmatrix}$$

(c)
$$T_{1,\mathbf{a}}(\mathbf{x}) = f(\mathbf{a}) + \nabla f(\mathbf{a})^{\top}(\mathbf{x} - \mathbf{a}) = 1 + \pi \cdot (2,1) \cdot (x_1, x_2 - 1)^{\top} = 1 - \pi + 2\pi x_1 + \pi x_2$$

(d)

$$\begin{split} T_{2,\mathbf{a}}(\mathbf{x}) &= T_{1,\mathbf{a}}(\mathbf{x}) + \frac{1}{2}(\mathbf{x} - \mathbf{a})^{\top} \nabla^2 f(\mathbf{a})(\mathbf{x} - \mathbf{a}) \\ &= T_{1,\mathbf{a}}(\mathbf{x}) + \frac{1}{2} \mathbf{x}^{\top} \nabla^2 f(\mathbf{a}) \mathbf{x} + \mathbf{x}^{\top} \nabla^2 f(\mathbf{a}) \mathbf{a} + \frac{1}{2} \mathbf{a}^{\top} \nabla^2 f(\mathbf{a}) \mathbf{a} \end{split}$$

With $\nabla^2 f(\mathbf{a}) = \begin{pmatrix} \pi^2 & \pi \\ \pi & 0 \end{pmatrix}$ we get that

$$T_{2,\mathbf{a}}(\mathbf{x}) = T_{1,\mathbf{a}}(\mathbf{x}) + 0.5\pi^2 x_1^2 + \pi x_1 x_2 - \pi x_1 + 0.$$

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(e) $T_{2,\mathbf{a}}(\mathbf{x})$ is multivariate polynomial of degree 2 which means its Hessian is constant and we can directly see that $\mathbf{H} := \nabla^2 T_{2,\mathbf{a}}(\mathbf{x}) = \nabla^2 f(\mathbf{a})$. For the eigenvalues of the Hessian it must hold that

$$\det(\mathbf{H} - \lambda \mathbf{I}) = 0$$

$$\iff \det\begin{pmatrix} \pi^2 - \lambda & \pi \\ \pi & -\lambda \end{pmatrix} = 0$$

$$\iff (\pi^2 - \lambda) \cdot (-\lambda) - \pi^2 = 0$$

$$\iff \lambda^2 - \pi^2 \lambda - \pi^2 = 0.$$

From which it follows that $\lambda_{1,2} = \frac{\pi^2 \pm \sqrt{\pi^4 + 4\pi^2}}{2} \Rightarrow \lambda_1 \approx 10.785, \lambda_2 \approx -0.915$. Since $\lambda_2 < 0$ $T_{2,\mathbf{a}}$ is not convex.