

## 第一题

$$f(x) = \frac{1}{1+e^{-x}} \rightarrow f'(x) = \frac{e^{-x}}{(1+e^{-x})^2}$$

## 第二题

$$\frac{df}{dx} = \frac{df}{dz} \cdot \frac{dz}{dy} \cdot \frac{dy}{dx}$$

$$\text{其中, } \frac{df}{dz} = -\frac{1}{2}e^{-\frac{1}{2}z}, \frac{dz}{dy} = y^T(S^{-1} + (S^{-1})^T), \frac{dy}{dx} = I \in \mathbb{R}^{D \times D}$$

$$\therefore \frac{df}{dx} = -\frac{1}{2}e^{-\frac{1}{2}z} y^T(S^{-1} + (S^{-1})^T) = -\frac{1}{2}e^{-\frac{1}{2}((x-\mu)^T S^{-1}(x-\mu))} (x-\mu)^T (S^{-1} + (S^{-1})^T)$$

偏导数维度:

$$\frac{df}{dz} \in \mathbb{R}^{1 \times 1}, \frac{dz}{dy} \in \mathbb{R}^{1 \times D}, \frac{dy}{dx} \in \mathbb{R}^{D \times D}, \frac{df}{dx} \in \mathbb{R}^{1 \times D}$$

## 第三题

$$\text{令 } x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_D \end{pmatrix}, \text{ 则 } xx^T = \begin{pmatrix} x_1^2 & x_1x_2 & \cdots & x_1x_D \\ x_2x_1 & x_2^2 & \cdots & x_2x_D \\ \vdots & \vdots & \ddots & \vdots \\ x_Dx_1 & x_Dx_2 & \cdots & x_D^2 \end{pmatrix}$$

$$\text{因此, } f(x) = \text{tr}(xx^T + \sigma^2 I) = \sum_{i=1}^D (x_i^2 + \sigma^2)$$

$$\therefore \frac{df}{dx} = \begin{pmatrix} 2x_1 & 2x_2 & \cdots & 2x_D \end{pmatrix}$$

## 第四题

设  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  在点  $\mathbf{x}$  可微

取任意单位向量  $\mathbf{u}$ , 沿  $\mathbf{u}$  方向的方向导数  $D_{\mathbf{u}}f(\mathbf{x}) = \nabla f(\mathbf{x}) \cdot \mathbf{u}$

$\therefore \nabla f(\mathbf{x}) \cdot \mathbf{u} \leq \|\nabla f(\mathbf{x})\| \cdot \|\mathbf{u}\| = \|\nabla f(\mathbf{x})\|$  (当且仅当  $\mathbf{u}$  与  $\nabla f(\mathbf{x})$  共线且同向, 即  $\mathbf{u} = \frac{\nabla f(\mathbf{x})}{\|\nabla f(\mathbf{x})\|}$ , 取等号)

$\therefore$  在所有单位方向中, 方向导数的最大值为  $\|\nabla f(\mathbf{x})\|_2$ , 达到该最大值的方向就是梯度方向, 故函数上升最快的方向为  $\nabla f(\mathbf{x})$ ; 同理, 下降最快的方向为  $-\nabla f(\mathbf{x})$

$\therefore$  函数变化最快的方向是梯度方向, 证明完毕

补充: 若  $\nabla f(\mathbf{x}) = \mathbf{0}$ , 则对任意  $\mathbf{u}$  有  $D_{\mathbf{u}}f(\mathbf{x}) = 0$ , 此时一阶变化对方向不敏感

## 第五题

$$\therefore f(x) = x^T A x + b^T x + c$$

$$\therefore \nabla f = x^T (A + A^T) + b^T$$

$$\therefore \text{Hessian} = \nabla^2 f = A + A^T$$

## 第六题

$$f(x) = \langle Ax, x \rangle = (Ax)^T x = x^T A^T x = 4x_1^2 + 11x_1x_2 + 7x_2^2$$

$$\text{令 } x_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\therefore f(x_0) = 4 \cdot (1)^2 + 11 \cdot 1 \cdot 1 + 7 \cdot (1)^2 = 22$$

$$\therefore \nabla f = \begin{pmatrix} 8x_1 + 11x_2 & 11x_2 + 14x_1 \end{pmatrix}$$

$$\therefore \nabla f(x_0) = \begin{pmatrix} 19 & 25 \end{pmatrix}$$

$$\therefore Hessian = \begin{pmatrix} 8 & 11 \\ 11 & 14 \end{pmatrix}$$

$$\therefore \nabla^2 f(x_0) = \begin{pmatrix} 8 & 11 \\ 11 & 14 \end{pmatrix}$$

$$\begin{aligned} \therefore f(x) &= f(x_0) + \nabla f(x_0)(x - x_0) + \frac{1}{2}(x - x_0)^T \nabla^2 f(x_0)(x - x_0) \\ &= 22 + 19(x_1 - 1) + 25(x_2 - 1) + 4(x_1 - 1)^2 + 11(x_1 - 1)(x_2 - 1) + 7(x_2 - 1)^2 \end{aligned}$$

## 第七题

$$f(x) = e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

$$\therefore f'(x) = e^{-\frac{1}{2\sigma^2}(x-\mu)^2} \cdot \left(-\frac{1}{\sigma^2}\right) \cdot 2(x-\mu) = -\frac{1}{\sigma^2}(x-\mu)e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$