

① a) Let $f(x) = \frac{1}{2} x^T A x + b^T x$, where A is a symmetric matrix and $b \in \mathbb{R}^n$ is a vector. What is $\nabla f(x)$

sol: $\nabla f(x) = \frac{1}{2} \times 2 A x + b \quad \left[\because \frac{\partial}{\partial x} x^T A x = 2 A x + \frac{\partial}{\partial x} b^T x = b \right]$
 $\nabla f(x) = A x + b$

b) Let $f(x) = g(h(x))$ where $g: \mathbb{R} \rightarrow \mathbb{R}$ is differentiable and $h: \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable. What is $\nabla f(x)$

$$\begin{aligned} \nabla f(x) &= g'(h(x)) \frac{\partial}{\partial x} h(x) \\ &= g'(h(x)) \nabla h(x) \end{aligned}$$

c) Let $f(x) = \frac{1}{2} x^T A x + b^T x$, where A is symmetric and $b \in \mathbb{R}^n$ is a vector. What is $\nabla^2 f(x)$

$$\nabla f(x) = \frac{1}{2} \times 2 A x + b$$

$$\nabla^2 f(x) = A$$

d) Let $f(x) = g(a^T x)$ where $g: \mathbb{R} \rightarrow \mathbb{R}$ is continuously differentiable & $a \in \mathbb{R}^n$ is a vector. What are $\nabla f(x)$ and $\nabla^2 f(x)$

$$\nabla f(x) = g'(a^T x) \cdot \frac{\partial}{\partial x} (a^T x)$$

$$= g'(a^T x) \cdot a$$

$$\nabla^2 f(x) = g''(a^T x) \cdot a \cdot \frac{\partial}{\partial x} a^T x$$

$$= g''(a^T x) a \cdot a^T$$

② a) Let $z \in \mathbb{R}^n$ be an n length vector. show that $A = zz^T$ is positive definite.

$$A = zz^T$$

A matrix A is positive definite if $x^T A x > 0$
 $\forall x \neq 0$ and $A^T = A$.

$$\begin{aligned} \text{So } x^T A x &= x^T z z^T x \\ &= (x^T z) (z^T x) \quad [\because z^T = z \text{ \& } x = x^T] \\ &= (x^T z)^2 \geq 0 \end{aligned}$$

$\therefore A = zz^T$ is positive semidefinite.

b) BAB^T

$$x^T A x > 0$$

substituting BAB^T in A we get

$$x^T B A B^T x$$

$$= (x^T B) A (B^T x)$$

$$= (x B^T)^T A (B^T x) \quad [\because B \in \mathbb{R}^{m \times n}]$$

$$= z^T A z \geq 0 \quad [z = B^T x \in \mathbb{R}^n \text{ \& since } A \text{ is PSD}]$$

$\therefore BAB^T$ is a positive semidefinite matrix.

③ The Avg empirical loss for logistic regression is given as

$$J(\theta) = \frac{1}{n} \sum_{i=1}^n \log(1 + e^{-y^{(i)} \theta^T x^{(i)}}) = -\frac{1}{n} \sum_{i=1}^n \log(h_\theta(y^{(i)} x^{(i)}))$$

where $y^{(i)} \in \{-1, 1\}$, $h_\theta(x) = g(\theta^T x)$ & $g(z) = \frac{1}{1 + e^{-z}}$

show that $z^T H z \geq 0$

Sol: Hessian H is the second derivative of the function.

$$\nabla J(\theta) = \frac{1}{n} \sum_{i=1}^n \frac{\frac{\partial}{\partial \theta} (1 + e^{-y^{(i)} \theta^T x^{(i)}})}{1 + e^{-y^{(i)} \theta^T x^{(i)}}}$$

$$= \frac{1}{n} \sum_{i=1}^n \frac{e^{-y^{(i)} \theta^T x^{(i)}} \cdot \frac{\partial}{\partial \theta} (-y^{(i)} \theta^T x^{(i)})}{1 + e^{-y^{(i)} \theta^T x^{(i)}}}$$

$$= -\frac{1}{n} \sum_{i=1}^n \frac{e^{-y^{(i)} \theta^T x^{(i)}}}{1 + e^{-y^{(i)} \theta^T x^{(i)}}} \cdot (-y^{(i)} x^{(i)})$$

$$= -\frac{1}{n} \sum_{i=1}^n \frac{1}{1 + e^{y^{(i)} \theta^T x^{(i)}}} \cdot (-y^{(i)} x^{(i)})$$

$$= +\frac{1}{n} \sum_{i=1}^n g(-y^{(i)} \theta^T x^{(i)}) (y^{(i)} x^{(i)})$$

$$= \frac{1}{n} \sum_{i=1}^n h_\theta(-y^{(i)} x^{(i)}) (y^{(i)} x^{(i)})$$

$$\nabla^2 J(\theta) = \frac{1}{n} \sum_{i=1}^n \frac{\partial}{\partial \theta} (h_\theta(-y^{(i)} x^{(i)})) (y^{(i)} x^{(i)})$$

$$= \frac{1}{n} \sum_{i=1}^n y^{(i)} x^{(i)} (h_\theta(-y^{(i)} x^{(i)})) (1 - h_\theta(-y^{(i)} x^{(i)})) \frac{\partial}{\partial \theta} (-y^{(i)} x^{(i)})$$

$$= -\frac{1}{n} \sum_{i=1}^n (y^{(i)} x^{(i)})^2 (h_\theta(-y^{(i)} x^{(i)})) (1 - h_\theta(-y^{(i)} x^{(i)}))$$

$$= -\frac{1}{n} \sum_{i=1}^n (x^{(i)})^2 (1 - h_\theta(x^{(i)})) (h_\theta(x^{(i)}))$$

$$H = \frac{1}{n} \sum_{i=1}^n h(x^{(i)}) (1 - h(x^{(i)})) x^{(i)} x^{(i)T} \left[\begin{array}{l} \therefore y^{(i)} \in \{-1, 1\} \\ \& h(-x) = h(x) \end{array} \right]$$

For any vector z , it holds that $z^T H z \geq 0$ that is to show that H is positive semidefinite

$$z^T H z$$

$$= z^T \left[\frac{1}{n} \sum_{i=1}^n h(x^{(i)}) (1 - h(x^{(i)})) x^{(i)} x^{(i)T} \right] z$$

$$= \frac{1}{n} \sum_{i=1}^n h(x^{(i)}) (1 - h(x^{(i)})) z^T x^{(i)} x^{(i)T} z$$

$$= \frac{1}{n} \sum_{i=1}^n h(x^{(i)}) (1 - h(x^{(i)})) (z^T x^{(i)})^2 \geq 0$$

Since $h(x^{(i)}) (1 - h(x^{(i)}))$ is positive & $(z^T x^{(i)})^2 \geq 0$

We can say that H is positive semidefinite.

③ b) The coefficients θ that were obtained after the fit are

$$\theta[0] = -2.6205115971801973$$

$$\theta[1] = 0.7603715358976769$$

$$\theta[2] = 1.1719467415671432$$