

# Linear Algebra and Calculus Exercises: Part I

CSCI 567 Machine Learning

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**MULTIPLE-CHOICE QUESTIONS:** one or more correct choices for each question.

## 1 Linear Algebra

**Q1** Which identities are NOT correct for real-valued matrices  $A$ ,  $B$ , and  $C$ ? Assume that inverses exist and multiplications are legal.

- (a)  $(AB)^{-1} = B^{-1}A^{-1}$
- (b)  $(I + A)^{-1} = I - A$
- (c)  $\text{tr}(AB) = \text{tr}(BA)$
- (d)  $(AB)^\top = A^\top B^\top$

**Q2** Suppose  $\mathbf{x}_1, \dots, \mathbf{x}_N$  are all  $D$ -dimensional vectors, and  $X \in \mathbb{R}^{N \times D}$  is a matrix where the  $n$ -th row is  $\mathbf{x}_n^\top$ . Then which of the following identities are correct?

- (a)  $X^\top X = \sum_{n=1}^N \mathbf{x}_n \mathbf{x}_n^\top$
- (b)  $X^\top X = \sum_{n=1}^N \mathbf{x}_n^\top \mathbf{x}_n$
- (c)  $XX^\top = \sum_{n=1}^N \mathbf{x}_n \mathbf{x}_n^\top$
- (d)  $XX^\top = \sum_{n=1}^N \mathbf{x}_n^\top \mathbf{x}_n$

## 2 Calculus

**Q1** Suppose  $\mathbf{a} \in \mathbb{R}^{n \times 1}$  is an arbitrary vector. Which one of the following functions is NOT convex:

- (a)  $f(\mathbf{x}) = \sum_{i=1}^n |x_i|$
- (b)  $f(\mathbf{x}) = \sum_{i=1}^n a_i x_i$
- (c)  $f(\mathbf{x}) = \min_{i \in \{1, \dots, n\}} a_i x_i$
- (d)  $f(\mathbf{x}) = \sum_{i=1}^n \exp(x_i)$

**Q2** Which of the following are correct chain rules ( $g, g_1, \dots, g_d$  are functions from  $\mathbb{R}$  to  $\mathbb{R}$ )?

- (a) For a composite function  $f(g(w))$ ,  $\frac{\partial f}{\partial w} = \frac{\partial f}{\partial g} \frac{\partial g}{\partial w}$ .
- (b) For a composite function  $f(g(w))$ ,  $\frac{\partial f}{\partial w} = \frac{\partial f}{\partial g} + \frac{\partial g}{\partial w}$ .
- (c) For a composite function  $f(g_1(w), \dots, g_d(w))$ ,  $\frac{\partial f}{\partial w} = \left( \frac{\partial f}{\partial g_1} \frac{\partial g_1}{\partial w}, \dots, \frac{\partial f}{\partial g_d} \frac{\partial g_d}{\partial w} \right)$ .
- (d) For a composite function  $f(g_1(w), \dots, g_d(w))$ ,  $\frac{\partial f}{\partial w} = \sum_{i=1}^d \frac{\partial f}{\partial g_i} \frac{\partial g_i}{\partial w}$ .

**Q3** A function  $f : \mathbb{R}^{n \times 1} \rightarrow \mathbb{R}$  is defined as  $f(\mathbf{x}) = \mathbf{x}^\top \mathbf{A} \mathbf{x} + \mathbf{b}^\top \mathbf{x}$  for some  $\mathbf{b} \in \mathbb{R}^{n \times 1}$  and  $\mathbf{A} \in \mathbb{R}^{n \times n}$ . What is the derivative  $\frac{\partial f}{\partial \mathbf{x}}$  (also called the gradient  $\nabla f(\mathbf{x})$ )?

- (a)  $(\mathbf{A} + \mathbf{A}^\top) \mathbf{x} + \mathbf{b}$
- (b)  $2\mathbf{A}^\top \mathbf{x} + \mathbf{b}$
- (c)  $2\mathbf{A} \mathbf{x} + \mathbf{b}$
- (d)  $2\mathbf{A} \mathbf{x} + \mathbf{x}$

**Q4** A function  $f : \mathbb{R}^{n \times 1} \rightarrow \mathbb{R}$  is defined as  $f(\mathbf{w}) = \ln(1 + e^{-\mathbf{w}^\top \mathbf{x}})$  for some  $\mathbf{x} \in \mathbb{R}^{n \times 1}$ . What is the derivative  $\frac{\partial f}{\partial \mathbf{w}}$ ?

- (a)  $-\frac{\mathbf{w}}{1+e^{\mathbf{w}^\top \mathbf{x}}}$
- (b)  $-\frac{\mathbf{x}}{1+e^{\mathbf{w}^\top \mathbf{x}}}$
- (c)  $-\frac{\mathbf{w}}{1+e^{-\mathbf{w}^\top \mathbf{x}}}$
- (d)  $-\frac{\mathbf{x}}{1+e^{-\mathbf{w}^\top \mathbf{x}}}$

**Q5** For a differential function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , which of the following statements are correct?

- (a) If  $\mathbf{x}^*$  is a minimizer of  $f$ , then  $\nabla f(\mathbf{x}^*) = \mathbf{0}$ .
- (b) If  $\mathbf{x}^*$  is a maximizer of  $f$ , then  $\nabla f(\mathbf{x}^*) = \mathbf{0}$ .
- (c) If  $\nabla f(\mathbf{x}^*) = \mathbf{0}$ , then  $\mathbf{x}^*$  is a minimizer of  $f$ .
- (d) If  $\nabla f(\mathbf{x}^*) = \mathbf{0}$ , then  $\mathbf{x}^*$  is a maximizer of  $f$ .