

# HOMEWORK 1

1. 证明《统计学习方法》习题 1.2:

通过经验风险最小化推导极大似然估计。证明模型是条件概率分布，当损失函数是对数损失函数时，经验风险最小化等价于极大似然估计。

2. 请证明下述 Hoeffding 引理:

**Lemma 1.** *Let  $X$  be a random variable with  $E(X) = 0$  and  $P(X \in [a, b]) = 1$ . Then it holds*

$$E\{\exp(sX)\} \leq \exp\{s^2(b-a)^2/8\}.$$

3. 请列举一个实际中有监督学习的应用，请说明（1）问题背景、（2）因变量和自变量分别是什么，以及（3）通过机器学习建模如何解决该实际问题。

4. Please read the background and then prove the following results.

Background:

Let  $\mathbf{y} = \Psi(\mathbf{x})$ , where  $\mathbf{y}$  is an  $m \times 1$  vector, and  $\mathbf{x}$  is an  $n \times 1$  vector. Denote

$$\frac{\partial \mathbf{y}}{\partial \mathbf{x}^\top} = \begin{bmatrix} \frac{\partial y_1}{\partial x_1} & \frac{\partial y_1}{\partial x_2} & \cdots & \frac{\partial y_1}{\partial x_n} \\ \frac{\partial y_2}{\partial x_1} & \frac{\partial y_2}{\partial x_2} & \cdots & \frac{\partial y_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial y_m}{\partial x_1} & \frac{\partial y_m}{\partial x_2} & \cdots & \frac{\partial y_m}{\partial x_n} \end{bmatrix}.$$

Prove the results:

(a) Let  $\mathbf{y} = \mathbf{Ax}$ , where  $\mathbf{y}$  is  $m \times 1$ ,  $\mathbf{x}$  is  $n \times 1$ ,  $\mathbf{A}$  is  $m \times n$ , and  $\mathbf{A}$  does not depend on  $\mathbf{x}$ , then

$$\frac{\partial \mathbf{y}}{\partial \mathbf{x}^\top} = \mathbf{A}.$$

(b) Let the scalar  $\alpha$  be defined by  $\alpha = \mathbf{y}^\top \mathbf{Ax}$ , where  $\mathbf{y}$  is  $m \times 1$ ,  $\mathbf{x}$  is  $n \times 1$ ,  $\mathbf{A}$  is  $m \times n$ , and  $\mathbf{A}$  is independent of  $\mathbf{x}$  and  $\mathbf{y}$ , then

$$\frac{\partial \alpha}{\partial \mathbf{x}^\top} = \mathbf{x}^\top \mathbf{A}^\top \left( \frac{\partial \mathbf{y}}{\partial \mathbf{x}^\top} \right) + \mathbf{y}^\top \mathbf{A}.$$

(c) For the special case in which the scalar  $\alpha$  is given by the quadratic form  $\alpha = \mathbf{x}^\top \mathbf{A} \mathbf{x}$  where  $\mathbf{x}$  is  $n \times 1$ ,  $\mathbf{A}$  is  $n \times n$ , and  $\mathbf{A}$  does not depend on  $\mathbf{x}$ , then

$$\frac{\partial \alpha}{\partial \mathbf{x}^\top} = \mathbf{x}^\top (\mathbf{A} + \mathbf{A}^\top).$$

(d) Let the scalar  $\alpha$  be defined by  $\alpha = \mathbf{y}^\top \mathbf{A} \mathbf{x}$ , where  $\mathbf{y}$  is  $m \times 1$ ,  $\mathbf{x}$  is  $n \times 1$ ,  $\mathbf{A}$  is  $m \times n$ , and both  $\mathbf{y}$  and  $\mathbf{x}$  are functions of the vector  $\mathbf{z}$ , where  $\mathbf{z}$  is a  $q \times 1$  vector and  $\mathbf{A}$  does not depend on  $\mathbf{z}$ . Then

$$\frac{\partial \alpha}{\partial \mathbf{z}^\top} = \mathbf{x}^\top \mathbf{A}^\top \left( \frac{\partial \mathbf{y}}{\partial \mathbf{z}^\top} \right) + \mathbf{y}^\top \mathbf{A} \left( \frac{\partial \mathbf{x}}{\partial \mathbf{z}^\top} \right).$$

(e) Let  $\mathbf{A}$  be a nonsingular,  $m \times m$  matrix whose elements are functions of the scalar parameter  $\alpha$ . Then

$$\frac{\partial \mathbf{A}^{-1}}{\partial \alpha} = -\mathbf{A}^{-1} \frac{\partial \mathbf{A}}{\partial \alpha} \mathbf{A}^{-1}.$$

5. Please write  $\hat{\mathbf{a}}$  as the solution of the minimization problem:

$$\min_{\mathbf{a}} \|\mathbf{X}\mathbf{a} - \mathbf{y}\|_2,$$

where  $\mathbf{X}$  is a  $n \times p$  matrix,  $\mathbf{y}$  is a  $n \times 1$  vector and  $\mathbf{a}$  is a  $p \times 1$  vector.  $\mathbf{X}^\top \mathbf{X}$  is nonsingular.

提交时间：9 月 18 日 20:00 之前。请预留一定的时间，迟交作业扣 3 分，作业抄袭 0 分。作业作答使用中英文皆可。