Exercises 3.15

Exercise 3.15.1

Let $p,\ p_i,\ q,\ q_i$ be density functions on $\mathbb{R}$ and $\alpha \in \mathbb{R}$. Show that the cross-entropy satificies the following properties:

a. $S(p_1 + p_2, q) = S(p_1, q) + S(p_2, q)$;

b. $S(\alpha p, q) = \alpha S(p, q) = S(p, q^\alpha)$;

c. $S(p, q_1 q_2) = S(p, q_1) + S(p, q_2)$.

Exercise 3.15.2

Show that the cross entropy satifies the following inequality

$$S(p, q) \geq 1 - \int p(x)q(x)dx$$

Exercise 3.15.3

Let p a fixed density. Show that the symetric relative entropy

$$D_{KL}(p\|q) + D_{KL}(q\|p)$$

reaches its minimum for $p = q$, and the minimum is equal to zero.

Exercise 3.15.4

Consider two exponential densities, $p_1 = \xi^1 e^{\xi^1 x}$ and $p_2 = \xi^2 e^{\xi^2 x}$, $x \geq 0$.

a. Show that $D_{KL}(p_1\|p_2) = \dfrac{\xi^2}{\xi^1} - \ln \xi^2 \xi^1 - 1$.

b. Verify $D_{KL}(p_1\|p_2) \neq D_{KL}(p_2\|p_2)$.

c. Show that the triangle inequality doesn't hold for three arbitrary densities.

Exercise 3.15.5

Let X be a discrete random variable. Show the inequality

$$H(X) \geq 0.$$

Exercise 3.15.7

We asume the target variable Z is $\mathcal{E}$-mesurable. What is mean squared error function in this case?

Exercise 3.15.8

Asume that a neural network has an input-output function $f_{w,b}$ linear in w and b. Show that the cost function (3.3.1) reaches its minimum for a unique pair $(w^\star, b^\star)$, which can be computed explicitly.

Exercise 3.15.9

Show that the Shannon entropy can be retrived from the Reyni entropy as

$$H(p) = \lim_{\alpha \to 1} H_\alpha(x).$$

Exercise 3.15.10

Let $\phi_\sigma(x) = \dfrac{1}{\sqrt{2\pi\sigma}} e^{-\frac{t^2}{2\sigma^2}}$. Consider the convolution operation $(f * g)(x) := \displaystyle\int f(t)g(x-t)dt$.

a. Show that $\phi_\sigma * \phi_\sigma = \phi_{\sigma\sqrt{2}}$;

b. Find $\phi_\sigma * \phi_{\sigma'}$ in the case $\sigma \neq \sigma'$.

Exercise 3.15.11

Consider two probability densitie, $p(x)$ and $q(x)$. The Cauchy-Schwartz divergence is defined by

$$D_{CS}(p, q) := -\ln\Big(\frac{\int p(x)q(x)dx}{\sqrt{\int p(x)^2 dx}\sqrt{\int q(x)^2 dx}}\Big)$$

.

Show the following:

a. $D_{CS}(p, q) = 0$ if and only if $p = q$;

b. $D_{CS}(p, q) \geq 0$;

c. $D_{CS}(p, q) = D_{CS}(q, p)$;

d. $D_{CS}(p, q) = -\ln \displaystyle\int pq dx - \frac{1}{2}H_2(p) - \frac{1}{2}H_2(q)$, where $H_2(\cdot)$ denotes the quadratic Reyni entropy.

Exercise 3.15.12

a. Show that for any function $f \in L^1[0, 1]$ we have the inequality $\|\tanh(f)\|_1 \leq \|f\|_1$.

b. Show that for any function $f \in L^2[0, 1]$ we have the inequality $\|\tanh\|_2 \leq \|f\|_2$.

Exercise 3.15.13

Consider two distributions on the sample space $\mathcal{X} = \{x_1, x_2\}$ given by

$$p = \begin{pmatrix} x_1 & x_2 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}, \; q = \begin{pmatrix} x_1 & x_2 \\ \frac{1}{2} & \frac{2}{3} \end{pmatrix}$$

Consider the function $\phi : \mathcal{X} \to \mathbb{R}^2$ defined by $\phi(x_1) = (0, 1)$ $\phi(x_2) = (1, 0)$. Find the maximum mean discrepancy between p and q.

SOLUTIONS

3.15.1 (a)

The claim follows from the linearity of the integral operator. In symbols we have:

$$\begin{aligned} S(p_1 + p_2, q) &= -\int_{\mathbb{R}} (p_1(x) + p_2(x)) \ln q(x) dx = -\int_{\mathbb{R}} p_1(x) \ln q(x) dx - \int_{\mathbb{R}} p_2(x) \ln q(x) dx \\ &= S(p_1, q) + S(p_2, q). \end{aligned}$$

□

3.15.1 (b)

From the linearity of the integral operator, and the property $c\ \ln(x) = \ln(x^c)$ we have:

$$\begin{aligned} S(\alpha p, q) &= -\int_{\mathbb{R}} \alpha p(x) \ln q(x) dx = -\alpha \int_{\mathbb{R}} p(x) \ln q(x) dx = \alpha S(p, q) \\ &= -\int_{\mathbb{R}} \alpha p(x) \ln q(x) dx = -\int_{\mathbb{R}} p(x) \ln q(x)^{\alpha} dx = S(p, q^{\alpha}). \end{aligned}$$

□

3.15.1 (c)

Using the addition identity for the logarithms we get:

$$\begin{aligned} S(p, q_1 q_2) &= -\int_{\mathbb{R}} p(x) \ln q_1(x) q_2(x) dx = -\int_{\mathbb{R}} p(x) \ln q_1(x) dx - \int_{\mathbb{R}} p(x) \ln q_1(x) \\ &= S(p, q_1) + S(p, q_2). \end{aligned}$$

□

0.1 3.15.2

By the inequality $\ln(x) \leq x - 1,\ \forall x \in \mathbb{R}^+$, and the definition of cross-entropy follows:

$$\begin{aligned} S(p, q) &= -\int_{\mathbb{R}} p(x) \ln q(x) dx \geq -\int_{\mathbb{R}} p(x)(q(x) - 1) dx \\ &\geq -\int_{\mathbb{R}} -p(x) dx - \int_{\mathbb{R}} p(x) q(x) dx = 1 - \int_{\mathbb{R}} p(x) q(x) dx. \end{aligned}$$

□