

1. Kullback-Leibler divergence

Instructions:

For the following pairs of distributions (\mathbf{P}, \mathbf{Q}) , compute the Kullback-Leibler divergence $\text{KL}(\mathbf{P}, \mathbf{Q})$.

If the **KL** divergence is $+\infty$ or $-\infty$, enter **+inf** or **-inf**.

(a)

1/1 point (graded)

$$\mathbf{P} = \mathcal{N}(a, \sigma^2), \quad \mathbf{Q} = \mathcal{N}(b, \sigma^2), \quad a, b \in \mathbb{R}, \sigma^2 > 0.$$

(If applicable, enter **ln(x)** for **ln(x)**. Do NOT enter "log".)

KL(P, Q) =

(a-b)^2/(2*sigma^2)

□ Answer: (a - b)^2/(2*sigma^2)

$$\frac{(a-b)^2}{2 \cdot \sigma^2}$$

STANDARD NOTATION

Solution:

If we write $\mathbf{X} \sim \mathbf{P}$, we can compute:

$$\begin{aligned} \text{KL}(\mathbf{P}, \mathbf{Q}) &= \mathbb{E}_{\mathbf{P}} \left[\ln \left(\frac{\frac{1}{\sqrt{2\pi}} \exp \left(-\frac{(X-a)^2}{2\sigma^2} \right)}{\frac{1}{\sqrt{2\pi}} \exp \left(-\frac{(X-b)^2}{2\sigma^2} \right)} \right) \right] \\ &= \mathbb{E}_{\mathbf{P}} \left[-\frac{(X-a)^2}{2\sigma^2} + \frac{(X-b)^2}{2\sigma^2} \right] \\ &= \frac{1}{2\sigma^2} \mathbb{E}_{\mathbf{P}} \left[2(a-b)(X-a) + (a-b)^2 \right] \\ &= \frac{(a-b)^2}{2\sigma^2}, \end{aligned}$$

because $\mathbb{E}_{\mathbf{P}}[(X-a)] = 0$.

提交

你已经尝试了2次 (总共可以尝试2次)

□ Answers are displayed within the problem

(b)

1/1 point (graded)

$$\mathbf{P} = \text{Ber}(a), \quad \mathbf{Q} = \text{Ber}(b), \quad a, b \in (0, 1)$$

(If applicable, enter **ln(x)** for **ln (x)**. Do NOT enter "log".)

KL (P, Q) =

(1-a)*ln((1-a)/(1-b))+a*ln(a/b)

$$(1-a) \cdot \ln \left(\frac{1-a}{1-b} \right) + a \cdot \ln \left(\frac{a}{b} \right)$$

Answer: a * ln(a/b) + (1-a) *ln((1-a)/(1-b))

STANDARD NOTATION

Solution:

If we write $X \sim \mathbf{P}$, $Y \sim \mathbf{Q}$, we have

$$\begin{aligned} \text{KL}(\mathbf{P}, \mathbf{Q}) &= \mathbf{P}(X=0) \ln \frac{\mathbf{P}(X=0)}{\mathbf{P}(Y=0)} + \mathbf{P}(X=1) \frac{\mathbf{P}(X=1)}{\mathbf{P}(Y=1)} \\ &= a \ln \frac{a}{b} + (1-a) \ln \frac{1-a}{1-b}. \end{aligned}$$

提交

你已经尝试了1次（总共可以尝试3次）

Answers are displayed within the problem

(c)

1/2 points (graded)

$$P = \text{Unif}([0, \theta_1]), \quad Q = \text{Unif}([0, \theta_2]), \quad 0 < \theta_1 < \theta_2.$$

Hint: Note the support of each distribution when computing the expectation.

(If applicable, enter **ln(x)** for **ln (x)**. Do NOT enter "log". If applicable, enter **theta_1** for θ_1 and **theta_2** for θ_2 .)

KL (P, Q) =

ln(theta_2/theta_1)/theta_1*theta_1

$$\frac{\ln \left(\frac{\theta_2}{\theta_1} \right)}{\theta_1} \cdot \theta_1$$

Answer: ln(theta_2/theta_1)

KL (Q, P) =

ln(theta_1/theta_2)/theta_2*theta_1

$$\frac{\ln \left(\frac{\theta_1}{\theta_2} \right)}{\theta_2} \cdot \theta_1$$

Answer: inf

没考虑到support，漏了一项

STANDARD NOTATION

Solution:

We compute

$$\begin{aligned} \text{KL}(\mathbf{P}, \mathbf{Q}) &= \mathbb{E}_{\mathbf{P}} \left[\ln \frac{1}{\frac{1}{\theta_2}} \right] \\ &= \ln \left(\frac{\theta_2}{\theta_1} \right). \end{aligned}$$

If we try to compute the **KL** divergence the other way round, we notice that **P** is not supported between for $\theta_1 < X < \theta_2$. We compute the expectation by integrating explicitly:

$$\text{KL}(\mathbf{Q}, \mathbf{P}) = \mathbb{E}_{\mathbf{Q}} \left[\ln \frac{q}{p} \right] \quad \text{where } p, q, \text{ are the pdfs of } \mathbf{P}, \mathbf{Q} \text{ respectively}$$

$$= \int_0^{\theta_1} \frac{1}{\theta_2} \ln \frac{1/\theta_2}{1/\theta_1} dx + \int_{\theta_1}^{\theta_2} \frac{1}{\theta_2} \ln \frac{1/\theta_2}{0} dx$$
$$= +\infty$$

because the second term diverges to $+\infty$. **Remark:** In general, $\text{KL}(\mathbf{P}, \mathbf{Q}) \neq \text{KL}(\mathbf{Q}, \mathbf{P})$.

提交

你已经尝试了3次（总共可以尝试3次）

☐ Answers are displayed within the problem

(d)

0/1 point (graded)

$$P = \text{Exp}(\lambda), \quad Q = \text{Exp}(\mu), \quad \lambda, \mu \in (0, \infty).$$

(If applicable, enter **ln(x)** for **ln**(*x*). Do NOT enter “log”.)

KL (P, Q) =

+inf

+inf

☐ Answer: ln(lambda/mu) + mu/lambda - 1

没用期望做

STANDARD NOTATION

Solution:

If $X \sim P$, then

$$\begin{aligned} \text{KL}(\mathbf{P}, \mathbf{Q}) &= \mathbb{E}_P \left[\ln \frac{\lambda e^{-\lambda x}}{\mu e^{-\mu x}} \right] \\ &= \mathbb{E}_P \left[\ln \frac{\lambda}{\mu} + (\mu - \lambda) X \right] \\ &= \ln \frac{\lambda}{\mu} + (\mu - \lambda) \frac{1}{\lambda} \\ &= \ln \frac{\lambda}{\mu} + \frac{\mu}{\lambda} - 1, \end{aligned}$$

because $\mathbb{E}_P[X] = \frac{1}{\lambda}$.

提交

你已经尝试了2次（总共可以尝试2次）

☐ Answers are displayed within the problem

讨论

显示讨论

主题： Unit 3 Methods of Estimation:Homework 4: TV distance, KL-Divergence, and Introduction to MLE / 1. Kullback-Leibler divergence

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