

Exercise SHEET 4

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Problem 1

- a) Let's consider the case where x_1 and x_2 are jointly Gaussian with

$$E[(x_1, x_2)] = (\bar{x}_1, \bar{x}_2) \quad \text{and}$$

$$\text{Var}[(x_1, x_2)] = \begin{pmatrix} \sigma_1^2 & c \\ c & \sigma_2^2 \end{pmatrix} \quad \text{where } c = \text{covariance}$$

$$\therefore c = \text{Cov}[x_1, x_2] = E[x_1 x_2] - E[x_1]E[x_2]$$

Now, considering the square of the L^2 -Wasserstein distance for the sake of simplicity:

$$(W_2(\pi x_1, \pi x_2))^2 = \inf_c E[(x_1 - x_2)^2]$$

$$= \inf_c E[x_1^2 - 2x_1 x_2 + x_2^2] \quad \text{--- ①}$$

$$\text{We know, } \text{Var}(x) = E[x^2] - (E[x])^2$$

$$\therefore E[x^2] = (E[x])^2 + \text{Var}(x) = \bar{x}_1^2 + \sigma_1^2$$

$$\begin{aligned} \therefore E[x_1^2] &= \bar{x}_1^2 + \sigma_1^2 \\ \text{and } E[x_2^2] &= \bar{x}_2^2 + \sigma_2^2 \end{aligned} \quad \left| \begin{array}{l} c = E[x_1 x_2] - E[x_1]E[x_2] \\ \therefore E[x_1 x_2] = c + \bar{x}_1 \bar{x}_2 \end{array} \right.$$

$$\text{From ①} \Rightarrow (W_2(\pi x_1, \pi x_2))^2 = \inf_c (\bar{x}_1^2 + \sigma_1^2 - \cancel{2c} + \bar{x}_2^2 + \sigma_2^2 - \cancel{2c})$$

$$= \inf_c (\bar{x}_1 - \bar{x}_2)^2 + \sigma_1^2 + \sigma_2^2 - 2c$$

The maximum eligible value of c is given by the Cauchy-Schwarz inequality:

$$(\text{Cov}[x_1, x_2])^2 \leq \text{Var}[x_1] \text{Var}[x_2] \Rightarrow c \leq \sigma_1 \sigma_2$$

Therefore,

$$\begin{aligned} (W_2(\pi x_1, \pi x_2))^2 &= (\bar{x}_1 - \bar{x}_2)^2 + \sigma_1^2 + \sigma_2^2 - 2\sigma_1 \sigma_2 \\ &= (\bar{x}_1 - \bar{x}_2)^2 + (\sigma_1 - \sigma_2)^2 \end{aligned}$$

(b) The Kullback-Leibler (KL) divergence of πx_1 from πx_2 is given by:

$$D_{KL}(\pi x_1 \parallel \pi x_2) := \int_{\mathcal{R}} \log \left(\frac{\pi x_1(x)}{\pi x_2(x)} \right) \pi x_1(x) dx$$

$$= \int_{\mathcal{R}} [\log \pi x_1(x) - \log \pi x_2(x)] \pi x_1(x) dx \quad \text{--- (1)}$$

For Gaussian distribution, PDF

$$\pi X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left(-\frac{(x-\bar{x})^2}{2\sigma^2} \right)$$

Substituting the formula for the KL divergence gives:

$$D_{KL}(\pi x_1 \parallel \pi x_2) = \int_{\mathcal{R}}$$

Taking logarithm on both sides, we get,

$$\begin{aligned} \log \pi X(x) &= \log \left(\frac{1}{\sqrt{2\pi\sigma^2}} \exp \left(-\frac{(x-\bar{x})^2}{2\sigma^2} \right) \right) \\ &= \log \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) + \log \left(\exp \left(-\frac{(x-\bar{x})^2}{2\sigma^2} \right) \right) \\ &= -\frac{1}{2} \log(2\pi\sigma^2) - \frac{(x-\bar{x})^2}{2\sigma^2} \end{aligned}$$

$$\begin{aligned} \text{So, (1)} \Rightarrow \int_{\mathcal{R}} \left[-\frac{1}{2} \log(2\pi\sigma_1^2) - \frac{(x_1 - \bar{x}_1)^2}{2\sigma_1^2} + \frac{1}{2} \log(2\pi\sigma_2^2) + \frac{(x_1 - \bar{x}_2)^2}{2\sigma_2^2} \right] \pi x_1(x) dx \\ = \mathbb{E} \left[\frac{1}{2} \log \frac{\sigma_2^2}{\sigma_1^2} - \frac{1}{2\sigma_1^2} (x_1 - \bar{x}_1)^2 + \frac{1}{2\sigma_2^2} (x_1 - \bar{x}_2)^2 \right] \end{aligned}$$

Here, the first term is constant, and the second term is the variance of X_1 , that is,

$$\frac{1}{2\sigma_1^2} \mathbb{E}[(X_1 - \bar{x}_1)^2] = \frac{1}{2\sigma_1^2} \cdot \sigma_1^2 = \frac{1}{2}$$

For the third term,

$$\begin{aligned} \mathbb{E}[(X_1 - \bar{x}_2)^2] &= \mathbb{E}[X_1^2 - 2\bar{x}_2 X_1 + \bar{x}_2^2] \\ &= (\sigma_1^2 + \bar{x}_1^2) - 2\bar{x}_1 \bar{x}_2 + \bar{x}_2^2 \\ &= \sigma_1^2 + (\bar{x}_1 - \bar{x}_2)^2 \end{aligned}$$

Collecting all three terms, we conclude that,

$$\begin{aligned} D_{KL}(\pi x_1 \parallel \pi x_2) &= \frac{1}{2} \left[\log \frac{\sigma_2^2}{\sigma_1^2} + \frac{\sigma_1^2 + (\bar{x}_1 - \bar{x}_2)^2}{\sigma_2^2} - 1 \right] \\ &= \frac{1}{2} \left[\log \frac{\sigma_2^2}{\sigma_1^2} + \frac{\sigma_1^2}{\sigma_2^2} + \frac{(\bar{x}_1 - \bar{x}_2)^2}{\sigma_2^2} - 1 \right] \end{aligned}$$

Note that this value becomes zero,

$$\text{if } \bar{x}_1 = \bar{x}_2 \text{ and } \sigma_1^2 = \sigma_2^2$$

Problem 2(a):

Given the sets of,

$$X_1 = \{a_1 = 1, a_2 = 2, a_3 = 3\}$$

$$X_2 = \{b_1 = 1.5, b_2 = 2, b_3 = -1\}$$

and,

$$p(a_i) = p(b_i) = 1/3, \text{ where } i = 1, 2, 3.$$

$$T = [x_{ij}]_{i,j=1}^3 \in \mathbb{R}^{3 \times 3} \quad (x_{ij} \geq 0).$$

$$J(T) = \sum_{i,j=1}^3 x_{ij} |b_i - a_j|^2.$$

To solve this problem, we assume that we have three water bottles that filled with $1/3$ unit of water. From the baseline the bottle heights are a_1, a_2, a_3 . Now, we target to re distribute the water into another sets of water bottle at height b_1, b_2, b_3 . The required energy required to move a unit from one bottle to another is proportion of square of height difference. To find a strategy that minimizes energy consumption,

if D_{ij} energy consumption per unit from a_j to b_i , D_{ij} can be expressed by following matrix

$$[D_{ij}] = \begin{pmatrix} (a_1 - b_1)^2 & (a_1 - b_2)^2 & (a_1 - b_3)^2 \\ (a_2 - b_1)^2 & (a_2 - b_2)^2 & (a_2 - b_3)^2 \\ (a_3 - b_1)^2 & (a_3 - b_2)^2 & (a_3 - b_3)^2 \end{pmatrix}$$

$$= \begin{pmatrix} 0.25 & 1 & 4 \\ 0.25 & 0 & 9 \\ 2.25 & 1 & 16 \end{pmatrix}$$

value

$$\begin{bmatrix} a_1 = 1, b_1 = 1.5 \\ a_2 = 2, b_2 = 2 \\ a_3 = 3, b_3 = -1 \end{bmatrix}$$

In the third column, the cost from anywhere to b_3 is higher in every scenarios. So, from here we can start rearrange-trick. We want to fill b_3 as much as from a_1 , and if it is not possible, then we may use water from a_2 and so on.

In the Question, we have exactly $1/3$ unit in a_1 and exactly $1/3$ will be transferred into b_3 , so,

$$[T_{ij}] = \begin{pmatrix} 0 & 0 & 1/3 \\ t_{21} & t_{22} & 0 \\ t_{31} & t_{32} & 0 \end{pmatrix}$$

Now, we repeat for the rest of bottles. Now cost to fill b_1 dominates one required to fill b_2 , we optimize b_1 first by Assign $1/3$ from a_2 to b_1 , namely $t_{21} = 1/3$. So, the resulting optimal strategy is,

$$[T_{ij}^*] = \begin{pmatrix} 0 & 0 & 1/3 \\ 1/3 & 0 & 0 \\ 0 & 1/3 & 0 \end{pmatrix}$$

If we change the order of indices to visualize the strategy. By re-arranging, we have,

$$[T_{ij}] = \begin{pmatrix} 1/3 & 0 & 0 \\ 0 & 1/3 & 0 \\ 0 & 0 & 1/3 \end{pmatrix}$$

$[T_{ij}]$	b_3	b_1	b_2
a_1	$1/3$	0	0
a_2	0	$1/3$	0
a_3	0	0	$1/3$

⇒ Re-arrange order $[T_{ij}]$

$[D_{ij}]$	b_3	b_1	b_2
a_1	4	0.25	1
a_2	9	0.25	0
a_3	16	2.25	1

⇒ Rearrange order $[D_{ij}]$

Problem 2b

The cost function is defined as

$$J(T) = \sum_{i,j=1}^3 t_{ij} |b_i - a_j|^2$$

We need to calculate $J(T)$, where the elements of the optimal strategy matrix

$$[T_{ij}^*] = \begin{bmatrix} 0 & 0 & 1/3 \\ 1/3 & 0 & 0 \\ 0 & 1/3 & 0 \end{bmatrix} \quad \text{from 2a}$$

$$\text{and } |b_i - a_j|^2 = D_{ij}$$

$$[D_{ij}] = \begin{bmatrix} 0.25 & 1 & 4 \\ 0.25 & 0 & 9 \\ 2.25 & 1 & 16 \end{bmatrix} \quad \text{from 2a}$$

$$\text{So, } J(T^*) = \sum_{i,j=1}^3 [T_{ij}^*] [D_{ij}]$$

$$= \sum \begin{bmatrix} 0 & 0 & 0.33 \\ 0.33 & 0 & 0 \\ 0 & 0.33 & 0 \end{bmatrix} \times$$

$$\begin{bmatrix} 0.25 & 1 & 4 \\ 0.25 & 0 & 9 \\ 2.25 & 1 & 16 \end{bmatrix}$$

$$= \sum \begin{bmatrix} 2.25 \times 0.33 & 0.33 & 0.33 \times 16 \\ 0.33 \times 0.25 & 0.33 & 0.33 \times 4 \\ 0.33 \times 0.25 & 0 & 0.33 \times 9 \end{bmatrix}$$

$$= \sum \begin{bmatrix} 0.74 & 0.33 & 5.28 \\ 0.08 & 0.33 & 1.32 \\ 0.08 & 0 & 2.97 \end{bmatrix}$$

$$= \begin{pmatrix} 0.74 + 0.33 + 5.28 + 0.08 + 0.33 \\ 0 + 1.32 + 0.08 + 2.97 \end{pmatrix}$$

$$= 11.13$$

$$\begin{bmatrix} 1 & 1 & 22.0 \\ 0 & 0 & 22.0 \\ 0 & 1 & 22.0 \end{bmatrix} = \begin{bmatrix} \mu_0 \end{bmatrix}$$

$$\begin{bmatrix} \mu_0 \end{bmatrix} \begin{bmatrix} \mu_0^* \end{bmatrix} = \begin{bmatrix} \mu \end{bmatrix}$$

$$\begin{bmatrix} 22.0 & 0 & 0 \\ 0 & 0 & 22.0 \\ 0 & 22.0 & 0 \end{bmatrix} =$$