

# Homework 1

**Problem 1** Consider a 2 dimensional random Gaussian vector,

$$\mathbf{x} \equiv \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma}),$$

where  $\boldsymbol{\Sigma} = \begin{bmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix}$ , and  $-1 < \rho < 1$ .

1. (10') Derive the joint entropy  $H(\mathbf{x})$
2. (10') Derive the mutual information  $I(x_1; x_2)$
3. (10') If  $\rho$  can be varied, when is the joint entropy maximized? what is the mutual information between  $x_1$  and  $x_2$  then?

**Problem 2** We talked about cross-entropy in the class. That is,

$$H(p, q) = -\mathbb{E}_{z \sim p}[\log q(z)].$$

In a  $C$ -class image classification problem, we have  $N$  samples. Denote the  $i$ -th sample as  $(\mathbf{x}_i, y_i)$ , where  $\mathbf{x}_i$  is image and  $y_i \in \{0, \dots, C-1\}$  is its class label. Suppose our model predicts class probability as  $\hat{p}_i(y)$  where  $y \in \{0, \dots, C-1\}$ .

1. (10') Show that (an empirical estimate of) cross-entropy is

$$\frac{1}{N} \sum_{i=1}^N -\log \hat{p}_i(y_i)$$

2. Let us denote the above cross-entropy loss as  $\ell$ . We often report  $\ell$  as an indicator of model quality. A model with lower  $\ell$  is more accurate. However, in some applications, we also report a related metric, called *perplexity* (PPL), defined as

$$PPL = e^\ell.$$

Show that:

- 1) (5') A perfect model has  $PPL = 1$ .
- 2) (5') Any reasonable model should have  $PPL < C$ .

**Problem 3** Install numpy, matplotlib and pytorch in your laptop/desktop. Attach your code for all sub-problems below.

1. (10') Plot the following function

$$f(\mathbf{x}) = \frac{1}{2} \left\| \begin{bmatrix} 2 & 1 \\ -1 & 3 \\ 0 & -2 \end{bmatrix} \mathbf{x} - \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\|_2^2,$$

on mesh grid defined upon  $-1 \leq x_1, x_2 \leq 1$ .

2. (10') Use Pytorch's autograd ability to get the gradient at  $\mathbf{x} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$
3. (30') Write a program (using Pytorch's autograd) to search for the minimizer of  $f(\mathbf{x})$ . Call it  $\mathbf{x}^*$ . Also report the corresponding  $f(\mathbf{x}^*)$