

# HOMEWORK 1

## 1 Probability Theory

### Question 1.1

1. Define the random variables and the values they can take on, both with symbols and numerically

$C$ : City where I am located

$C \in \mathcal{C} = \{\text{Amsterdam, Rotterdam}\}$

$W$ : Weather in the city

$W \in \mathcal{W} = \{\text{Rainy, Dry}\}$

$$\mathbb{P}(W = \text{Rainy} | C = \text{Amsterdam}) = 0.5$$

$$\mathbb{P}(C = \text{Amsterdam}) = 0.8$$

$$\mathbb{P}(W = \text{Rainy} | C = \text{Rotterdam}) = 0.75$$

$$\mathbb{P}(C = \text{Rotterdam}) = 0.2$$

2. What is the probability that it does not rain when you are in Rotterdam?

$$\mathbb{P}(W = \text{Dry} | C = \text{Rotterdam}) = 1 - \mathbb{P}(W = \text{Rainy} | C = \text{Rotterdam}) = 1 - 0.75 = 0.25$$

3. What is the probability that it rains where you are?

$$\begin{aligned}\mathbb{P}(W = \text{Rainy}) &= \mathbb{P}(W = \text{Rainy} | C = \text{Rotterdam}) \mathbb{P}(C = \text{Rotterdam}) \\ &\quad + \mathbb{P}(W = \text{Rainy} | C = \text{Amsterdam}) \mathbb{P}(C = \text{Amsterdam}) \\ &= 0.75 \cdot 0.2 + 0.5 \cdot 0.8 = 0.55\end{aligned}$$

4. [...] What is the probability that you are in Amsterdam?

$$\begin{aligned}\mathbb{P}(C = \text{Amsterdam} | W = \text{Rainy}) &= \frac{\mathbb{P}(W = \text{Rainy} | C = \text{Amsterdam}) \mathbb{P}(C = \text{Amsterdam})}{\mathbb{P}(W = \text{Rainy})} \\ &= \frac{0.5 \cdot 0.8}{0.55} = 0.72\end{aligned}$$

### Question 1.2

1. What is  $p(\text{cancer})$  and  $p(\text{not cancer})$ ?

$$p(\text{cancer}) = \frac{500}{500000} = 0.001$$

$$p(\text{not cancer}) = 1 - 0.001 = 0.999$$

2. If a patient takes the blood test and it returns positive, what is the probability the patient has cancer?

$$p(\text{positive} | \text{cancer}) = 0.99 \quad p(\text{positive} | \text{not cancer}) = 0.05$$

$$\begin{aligned}p(\text{positive}) &= p(\text{positive} | \text{cancer}) p(\text{cancer}) + p(\text{positive} | \text{not cancer}) p(\text{not cancer}) \\ &= 0.99 \cdot 0.001 + 0.05 \cdot 0.999 = 0.05094\end{aligned}$$

$$p(\text{cancer} | \text{positive}) = \frac{p(\text{positive} | \text{cancer}) p(\text{cancer})}{p(\text{positive})} = \frac{0.99 \cdot 0.001}{0.05094} = 0.0194$$

3. What are some of the assumptions we are implicitly making when answering this question?

There is no error in the classification used to estimate the a priori distribution of cancer. Besides, the trials of the diagnose test were performed on a representative sample of the population, i.e., the sample was not biased towards either the group with or without cancer, for example due to reasons like age or medical pre-conditions.

### Question 1.3

1. Write down the general expression for a posterior distribution, using  $\theta$  for the parameter,  $\mathcal{D}$  for the data. Indicate the *prior*, *likelihood*, *evidence*, and *posterior*.

$$p(\theta|\mathcal{D}) = \frac{p(\mathcal{D}|\theta)p(\theta)}{p(\mathcal{D})}$$

2. Write the posterior for this particular example. You do not need an analytic solution.

$$p(\mu | \{x_1, \dots, x_N\}, \sigma^2) = \frac{\left[ \prod_{i=1}^N \mathcal{N}(x_i | \mu, \sigma^2) \right] \mathcal{N}(\mu | \mu_0, \sigma_0^2)}{\int_{\mathbb{R}} \left[ \prod_{i=1}^N \mathcal{N}(x_i | \mu', \sigma^2) \right] \mathcal{N}(\mu' | \mu_0, \sigma_0^2) d\mu'}$$

## 2 Basic Linear Algebra and Derivatives

### Question 2.1

1. Compute  $\mathbf{A}\mathbf{b}$

$$\mathbf{A}\mathbf{b} = \begin{bmatrix} 3 & 5 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 9 \\ 5 \end{bmatrix} = \begin{bmatrix} 27 + 25 \\ 18 + 15 \end{bmatrix} = \begin{bmatrix} 52 \\ 33 \end{bmatrix}$$

2. Compute  $\mathbf{b}^T \mathbf{A}$

$$\mathbf{b}^T \mathbf{A} = \begin{bmatrix} 9 & 5 \end{bmatrix} \begin{bmatrix} 3 & 5 \\ 2 & 3 \end{bmatrix} = \begin{bmatrix} 27 + 10 & 45 + 15 \end{bmatrix} = \begin{bmatrix} 37 & 60 \end{bmatrix}$$

3. What is the vector  $\mathbf{c}$  for which  $\mathbf{A}\mathbf{c} = \mathbf{b}$ ?

$$\mathbf{c} = \begin{bmatrix} -2 \\ 3 \end{bmatrix} \text{ since } \mathbf{A}\mathbf{c} = \begin{bmatrix} 3 & 5 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} -2 \\ 3 \end{bmatrix} = \begin{bmatrix} -6 + 15 \\ -4 + 9 \end{bmatrix} = \begin{bmatrix} 9 \\ 5 \end{bmatrix} = \mathbf{b}$$

4. What is  $\mathbf{A}^{-1}$ ?

$$\mathbf{A}^{-1} = \frac{1}{3 \cdot 3 - 2 \cdot 5} \begin{bmatrix} 3 & -5 \\ -2 & 3 \end{bmatrix} = \begin{bmatrix} -3 & 5 \\ 2 & -3 \end{bmatrix}$$

5. Verify that  $\mathbf{A}^{-1}\mathbf{b} = \mathbf{c}$ . Show that this must be the case.

$$\mathbf{A}^{-1}\mathbf{b} = \begin{bmatrix} -3 & 5 \\ 2 & -3 \end{bmatrix} \begin{bmatrix} 9 \\ 5 \end{bmatrix} = \begin{bmatrix} -27 + 25 \\ 18 - 15 \end{bmatrix} = \begin{bmatrix} -2 \\ 3 \end{bmatrix} = \mathbf{c}$$

Since  $\mathbf{A}\mathbf{c} = \mathbf{b}$  and  $\mathbf{A}$  is invertible, we have  $\mathbf{A}^{-1}\mathbf{A}\mathbf{c} = \mathbf{A}^{-1}\mathbf{b} \rightarrow \mathbf{c} = \mathbf{A}^{-1}\mathbf{b}$ , as required.

## Question 2.2

Find the gradient of the following functions

1.  $f(x) = x^2 + 2x + 3 \rightarrow \nabla f = 2x + 2$
2.  $f(x) = (2x^3 + 1)^2 \rightarrow \nabla f = 2(2x^3 + 1) \cdot 3 \cdot 2 \cdot x^2 = 12x^2(2x^3 + 1)$

Find the partial derivative of the following functions with respect to  $x$ ,  $y$ ,  $z$

1.  $f(x, y, z) = (x + 2y)^2 \sin(xy)$

$$\frac{\partial f}{\partial x} = 2(x + 2y) \sin(xy) + (x + 2y)^2 y \cos(xy) = (x + 2y)(2 \sin(xy) + y(x + 2y) \cos(xy))$$

$$\frac{\partial f}{\partial y} = 4(x + 2y) \sin(xy) + (x + 2y)^2 x \cos(xy) = (x + 2y)(4 \sin(xy) + x(x + 2y) \cos(xy))$$

$$\frac{\partial f}{\partial z} = 0$$

2.  $f(x, y, z) = 2 \log(x + y^2 - z)$

$$\frac{\partial f}{\partial x} = \frac{2}{x + y^2 - z}$$

$$\frac{\partial f}{\partial y} = \frac{4y}{x + y^2 - z}$$

$$\frac{\partial f}{\partial z} = \frac{-2}{x + y^2 - z}$$

3.  $f(x, y, z) = \exp(x(\cos(y + z)))$

$$\frac{\partial f}{\partial x} = \cos(y + z) \exp(x(\cos(y + z)))$$

$$\frac{\partial f}{\partial y} = -x \sin(x + y) \exp(x(\cos(y + z))) = \frac{\partial f}{\partial z}$$

## Question 2.3

1. Expand the expression and gather terms.

$$\alpha = (\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu}) + (\boldsymbol{\mu} - \boldsymbol{\mu}_0)^T \mathbf{S}^{-1} (\boldsymbol{\mu} - \boldsymbol{\mu}_0) =$$

$$(\mathbf{x}^T - \boldsymbol{\mu}^T) \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu}) + (\boldsymbol{\mu}^T - \boldsymbol{\mu}_0^T) \mathbf{S}^{-1} (\boldsymbol{\mu} - \boldsymbol{\mu}_0) =$$

$$(\mathbf{x}^T \boldsymbol{\Sigma}^{-1} - \boldsymbol{\mu}^T \boldsymbol{\Sigma}^{-1}) (\mathbf{x} - \boldsymbol{\mu}) + (\boldsymbol{\mu}^T \mathbf{S}^{-1} - \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) (\boldsymbol{\mu} - \boldsymbol{\mu}_0) =$$

$$\mathbf{x}^T \boldsymbol{\Sigma}^{-1} \mathbf{x} - \mathbf{x}^T \boldsymbol{\Sigma}^{-1} \boldsymbol{\mu} - \boldsymbol{\mu}^T \boldsymbol{\Sigma}^{-1} \mathbf{x} + \boldsymbol{\mu}^T \boldsymbol{\Sigma}^{-1} \boldsymbol{\mu} + \boldsymbol{\mu}^T \mathbf{S}^{-1} \boldsymbol{\mu} - \boldsymbol{\mu}^T \mathbf{S}^{-1} \boldsymbol{\mu}_0 - \boldsymbol{\mu}_0^T \mathbf{S}^{-1} \boldsymbol{\mu} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1} \boldsymbol{\mu}_0 =$$

$$\mathbf{x}^T \boldsymbol{\Sigma}^{-1} \mathbf{x} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1} \boldsymbol{\mu}_0 - \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} \mathbf{x} + \mathbf{S}^{-1} \boldsymbol{\mu}_0) - (\mathbf{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) \boldsymbol{\mu} + \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1}) \boldsymbol{\mu}$$

2. Collect all the terms that *depend* on  $\boldsymbol{\mu}$  and those that *do not*.

$$\alpha = \mathbf{x}^T \boldsymbol{\Sigma}^{-1} \mathbf{x} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1} \boldsymbol{\mu}_0 - \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} \mathbf{x} + \mathbf{S}^{-1} \boldsymbol{\mu}_0) - (\mathbf{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) \boldsymbol{\mu} + \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1}) \boldsymbol{\mu}$$

3. Take the derivative with respect to  $\boldsymbol{\mu}$ , set to 0, and solve for  $\boldsymbol{\mu}$ .

$$\begin{aligned} \frac{\partial \alpha}{\partial \boldsymbol{\mu}} &= -\frac{\partial}{\partial \boldsymbol{\mu}} \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} \mathbf{x} + \mathbf{S}^{-1} \boldsymbol{\mu}_0) - \frac{\partial}{\partial \boldsymbol{\mu}} (\mathbf{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) \boldsymbol{\mu} + \frac{\partial}{\partial \boldsymbol{\mu}} \boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1}) \boldsymbol{\mu} \\ &= -(\mathbf{x}^T (\boldsymbol{\Sigma}^{-1})^T + \boldsymbol{\mu}_0^T (\mathbf{S}^{-1})^T) - (\mathbf{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) + \boldsymbol{\mu}^T [(\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1}) + (\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1})^T] \\ &= -2(\mathbf{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \mathbf{S}^{-1}) + 2\boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} + \mathbf{S}^{-1}) = 0 \end{aligned}$$

$$2\boldsymbol{\mu}^T (\boldsymbol{\Sigma}^{-1} + \boldsymbol{S}^{-1}) = 2 (\boldsymbol{x}^T \boldsymbol{\Sigma}^{-1} + \boldsymbol{\mu}_0^T \boldsymbol{S}^{-1})$$

$$(\boldsymbol{\Sigma}^{-1} + \boldsymbol{S}^{-1}) \boldsymbol{\mu} = \boldsymbol{\Sigma}^{-1} \boldsymbol{x} + \boldsymbol{S}^{-1} \boldsymbol{\mu}_0$$

Assuming  $\boldsymbol{\Sigma}$  and  $\boldsymbol{S}$  are positive definite matrices,  $\det(\boldsymbol{\Sigma})$ ,  $\det(\boldsymbol{S})$ ,  $\det(\boldsymbol{\Sigma}^{-1})$ , and  $\det(\boldsymbol{S}^{-1})$  are all strictly greater than 0. Then, by the Minkowski determinant theorem,

$$\det(\boldsymbol{\Sigma}^{-1} + \boldsymbol{S}^{-1}) \geq \det(\boldsymbol{\Sigma}^{-1}) + \det(\boldsymbol{S}^{-1}) > 0.$$

Thus,  $(\boldsymbol{\Sigma}^{-1} + \boldsymbol{S}^{-1})$  is also invertible and it is possible to solve for  $\boldsymbol{\mu}$ :

$$\boldsymbol{\mu} = (\boldsymbol{\Sigma}^{-1} + \boldsymbol{S}^{-1})^{-1} (\boldsymbol{\Sigma}^{-1} \boldsymbol{x} + \boldsymbol{S}^{-1} \boldsymbol{\mu}_0)$$