# Statistics 2 - Problem Sheet 1

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## Question - 1.

#### Question 1 a)

If **Y** and **X** are random vectors st  $\mathbf{Y} = C\mathbf{X}$  where C is a matrix of fixed coefficients, show that if  $\Sigma_X$  and  $\Sigma_Y$  are the covariance matrices for **X** and **Y** respectively then

$$\Sigma_y = C\Sigma_X C^T$$

#### Answer 1 a)

$$\mu_{y} = \mathbb{E}(Y)$$

$$= \mathbb{E}(CX)$$

$$= C\mathbb{E}(X)$$

$$= C\mu_{X}$$

$$\Sigma_{y} = \mathbb{E}[(Y - \mu_{Y})(Y - \mu_{Y})^{T}]$$

$$= \mathbb{E}[(CX - C\mu_{X})(CX - C\mu_{X})^{T}]$$

$$= \mathbb{E}[C(X - \mu_{X})(X - \mu_{X})^{T}C^{T}]$$

$$= C\mathbb{E}[(X - \mu_{X})(X - \mu_{X})^{T}]C^{T}$$

$$= C\Sigma_{X}C^{T}$$

#### Question 1 b)

Consider a multivariate normal random vector  $\mathbf{X} \sim \text{Normal}(\boldsymbol{\mu}_X, \Sigma_X)$  and suppose that the covariance matrix can be decomposed  $\Sigma_X = CC^T$  (THis can always be done for a full rank covariance matrix using a Choleski decomposition). Show that  $\Sigma_X^{-1} = C^{-T}C^{-1}$  and that  $\mathbf{Y} = C^{-1}(\mathbf{X} - \boldsymbol{\mu}_X) \sim \text{Normal}(\mathbf{0}, I)$ .

#### Answer 1 b)

$$\Sigma_{X}\Sigma_{X}^{-1} = I$$

$$\Longrightarrow CC^{T}\Sigma_{X}^{-1} = 1$$

$$\Longrightarrow C^{T}V_{X}^{-1} = C^{-1}$$

$$\Longrightarrow V_{X}^{-1} = C^{-1}C^{-1}$$

$$Y = C^{-1}(X - \mu_{X})$$

$$\mathbb{P}(Y = x) = \mathbb{P}(C^{-1}(X - \mu_{X}) = x)$$

$$TODO$$

#### Question 1 c)

Assuming that  $\mathbf{X} \sim \text{Normal}(\boldsymbol{\mu}_X, \Sigma_X)$  show that

$$(\mathbf{X} - \boldsymbol{\mu}_X)^T \Sigma_X^{-1} (\mathbf{X} - \boldsymbol{\mu}_X) = \mathbf{Y}^T \mathbf{Y} \text{ where } \mathbf{Y} \sim \text{Normal}(\mathbf{0}, I)$$

Answer 1 c)

TODO

### Question 1 d)

If  $Z_i \stackrel{\text{iid}}{\sim} \text{Normal}(0,1)$  random variables then

$$\sum_{i=1}^{n} Z_i^2 \sim \chi_n^2$$

What is the distribution of

$$(\mathbf{X} - \boldsymbol{\mu}_X)^T \Sigma_X^{-1} (\mathbf{X} - \boldsymbol{\mu}_X)$$

if  $\mathbf{X} \sim \text{Normal}(\boldsymbol{\mu}_X, \boldsymbol{\Sigma}_X)$ ?

#### Answer 1 d)

TODO

## Question - 2.

#### Question 2 a)

Define 
$$\mathbf{y} = \begin{pmatrix} 1 \\ -3 \end{pmatrix}$$
 and  $B = \begin{pmatrix} -1 & 2 & -1 \\ 2 & -3 & 0 \end{pmatrix}$ .

Find  $B^T \mathbf{y}$ 

$$\begin{pmatrix} -1 & 2 & -1 \\ 2 & -3 & 0 \end{pmatrix}^T \begin{pmatrix} 1 \\ -3 \end{pmatrix} = \begin{pmatrix} -1 & 2 \\ 2 & -3 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ -3 \end{pmatrix} = \begin{pmatrix} -1-7 \\ 2+9 \\ -1+0 \end{pmatrix} = \begin{pmatrix} -7 \\ 11 \\ -1 \end{pmatrix}$$

#### Answer 2 a)

#### Question 2 b)

Let A be a full rank  $3 \times 3$  matrix and B be a full rank  $5 \times 3$  matrix.

State the dimensions of the following, if they exist. For those that do not exist, state why in a single sentence.

- i)  $A^{-1}B^T$  $A^{-1} \in \mathbb{R}(3 \times 3), B^T \in \mathbb{R}(3 \times 5) \implies A^{-1}B^T \in \mathbb{R}(3 \times 5)$
- ii)  $A^{-1}B$  $A^{-1} \in \mathbb{R}(3 \times 3), B \in \mathbb{R}(5 \times 3) \implies$  these matrices are incompatible for multiplication.
- iii)  $B^{-1}A$  $B^{-1} \in \mathbb{R}(3 \times 5), A \in \mathbb{R}(3 \times 3) \implies$  these matrices are incompatible for multiplication.
- iv) BA $B \in \mathbb{R}(5 \times 3), A \in \mathbb{R}(3 \times 3) \implies BA \in \mathbb{R}(5 \times 3)$
- v)  $B^{-1}A^T$  $B^{-1} \in \mathbb{R}(3 \times 5), \ A^T \in \mathbb{R}(3 \times 3) \implies$  these matrices are incompatible for multiplication.
- vi)  $BA^{-1}$  $B \in \mathbb{R}(5 \times 3), A^{-1} \in \mathbb{R}(3 \times 3) \implies BA^{-1} \in \mathbb{R}(5 \times 3).$
- vii)  $(BA)^{-1}$ We know that  $BA \in \mathbb{R}(5 \times 3) \implies (BA)^{-1} \in \mathbb{R}(3 \times 5)$ .
- viii)  $B^T A$  $B^T \in \mathbb{R}(3 \times 5), A \in \mathbb{R}(3 \times 3) \implies$  these matrices are incompatible for multiplication.

- ix) B + A $B \in \mathbb{R}(3 \times 5)$ ,  $A \in \mathbb{R}(3 \times 3)$ . Matrices must have the exact same dimensions in order to be added together, thus this in an illegal equation.
- x)  $B + A^T$  $B \in \mathbb{R}(3 \times 5)$ ,  $A \in \mathbb{R}(3 \times 3)$ . Matrices must have the exact same dimensions in order to be added together, thus this in an illegal equation.

# Question - 3.

The Exponential Distribution is often a reasonable model of the times between random events. Suppose then, that  $x_1, \ldots, x_n$  are observations of times between hardware faults on a computer network, and it is reasonable to treat the faults as independent. To plan for fault tolerance the network managers need a reasonable model for the fault occurrence rate. The pdf of an Exponential Distribution is

$$f(x) = \mathbb{1}\{x \ge 0\}\lambda e^{-\lambda x}$$

where  $\lambda$  is a positive parameter.

The variance of an exponential random variable is  $\lambda^{-2}$ .

**Question 3 a)** - Let  $X \sim \text{Exponential}(\lambda)$ . Find  $\mathbb{E}(X)$ .

Answer 3 a)

$$\mathbb{E}(X) = \int_{-\infty}^{\infty} f_X(t)dt$$

$$= \int_{-\infty}^{\infty} \mathbb{1}\{t \ge 0\} \lambda e^{-\lambda t} dt$$

$$= \int_{0}^{\infty} \lambda e^{-\lambda t} dt$$

$$= \left[-\frac{1}{\lambda} e^{-\lambda t}\right]_{0}^{\infty}$$

$$= [0] - \left[-\frac{1}{\lambda}\right]$$

$$= \frac{1}{\lambda}$$

**Question 3 b)** - Hence, suggest an estimator,  $\hat{\lambda}$ , for  $\lambda$ .

Answer 3 b)

$$\hat{\lambda} = 1/\bar{x} \text{ where } \bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$

Question 3 c) - What is the variance of  $\hat{\lambda}^{-1}$ ? Answer 3 c)

$$\operatorname{var}\left(\hat{\lambda}^{-1}\right) = \operatorname{var}(\bar{X})$$

$$= \operatorname{var}\left(\frac{1}{n}\sum_{i=1}^{n}X_{i}\right)$$

$$= \frac{n}{n^{2}}\operatorname{var}(X_{1})$$

$$= \frac{1}{n\lambda^{2}}$$

### Question 3 d)

Let 
$$\bar{x} = \frac{1}{n} \sum x_i$$
.

Find a first order Taylor expansion of  $\hat{\lambda}$  about  $\mathbb{E}(\bar{x})$ , considering  $\hat{\lambda}$  as a function of  $\bar{x}$ .

### Answer 3 d)

TODO

#### Question 3 e)

Hence find an approximation for the variance of  $\hat{\lambda}_i$  in terms of n and  $\bar{x}$ . This use of Taylor expansions to computer approximate variances via linearisation is known as the  $\Delta$ -method in statistics.

#### Answer 3 e)

TODO

## Question - 4.

Consider again the setup from the previous question, but now taking a Bayesian approach. This means taht we need to augment our model with a prior distribution for the parameter,  $\lambda \sim \Gamma(\alpha, \theta)$ . So the prior pdf of  $\lambda$  is

$$f(\lambda) = \frac{\lambda^{\alpha - 1} e^{-\lambda/\theta}}{\theta^{\alpha} \Gamma(\alpha)}$$

with  $\mathbb{E}(\lambda) = \alpha \theta$  and  $var(\lambda) = \alpha \theta^2$ .

**Question 4 a)** - Write down the *pdf* for the joint distribution of the data  $x_1, x_2, \ldots$  given  $\lambda$ .

#### Answer 4 a)

$$f_n(\mathbf{x}; \lambda) = \prod_{i=1}^n f(x_i; \lambda)$$
$$= \prod_{i=1}^n \mathbb{1}\{x_i \ge 0\} \lambda e^{-\lambda x_i}$$
$$= \mathbb{1}\{\text{all } x > 0\} \lambda^n e^{-\lambda n\bar{x}}$$

**Question 4 b)** - By considering the joint distribution of  $\lambda$  and  $\mathbf{x}$ , indentify the posterior distribution of  $\lambda$  given  $\mathbf{x}$ .

#### Answer 4 b)

$$f(\lambda; \mathbf{x}) \propto f(\mathbf{x}; \lambda) f(\lambda)$$

$$= \lambda^n e^{-\lambda n \bar{x}} \cdot \frac{\lambda^{\alpha - 1} e^{-\lambda / \theta}}{\theta^{\alpha} \Gamma(\alpha)}$$

$$\propto \lambda^{n + \alpha - 1} e^{-\lambda (n \bar{x} + 1 / \theta)}$$

$$= \lambda^{n + \alpha - 1} e^{-\lambda \frac{n \bar{x} \theta + 1}{\theta}}$$

$$\sim \Gamma\left(n + \alpha, \frac{\theta}{n \bar{x} \theta + 1}\right)$$

**Question 4 c)** - What are the posterior expection and variance of  $\lambda$ ?

Answer 4 c)

$$\mathbb{E}[\lambda; \mathbf{x}] = \frac{\theta(n + \alpha - 1)}{n\bar{x}\theta + 1} \quad \text{var}(\lambda; \mathbf{x}) = \frac{\theta^2(n + \alpha - 1)}{(n\bar{x}\theta + 1)^2}$$

## Question 4 d)

Consider the situation in which  $n \to \infty$ .

What happens to the Bayesian and frequentist inferences aboute  $\lambda$  in this case?

Answer 4 d)

$$\mathbb{E}[\lambda; \mathbf{x}] \stackrel{n \to \infty}{\longrightarrow} \frac{1}{\bar{x}} = \hat{\lambda}_{\text{Frequentist}}$$