Bayesian Statistics III/IV (MATH3361/4071)

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## **Exercise Sheet: Bayesian Statistics**

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### Part I

# Matrix & vector calculus

The exercises about Matrix & vector calculus are optional and can be skipped.

**Exercise 1.**  $(\star)$ Let A, B be  $K \times K$  invertible matrices. Show that

$$(A+B)^{-1} = A^{-1}(A^{-1} + B^{-1})^{-1}B^{-1}$$

**Exercise 2.**  $(\star\star)$ [Woodbury matrix identity] Verify that

$$(A + UCV)^{-1} = A^{-1} - A^{-1}U (C^{-1} + VA^{-1}U)^{-1}VA^{-1}$$

if A and C are non-singular.

Exercise 3.  $(\star\star)$ [Sherman-Morrison formula] Let A be a  $K\times K$  invertible matrix and u and v two  $K\times 1$  column vectors. Verify that

$$(A + uv^{\top})^{-1} = A^{-1} - \frac{1}{1 + v^{\top} A^{-1} u} A^{-1} uv^{\top} A^{-1}$$

if  $1 + v^{\top} A^{-1} u \neq 0$ , and if A is non-singular.

**Exercise 4.**  $(\star\star\star)$ [Block partition matrix inversion] Let A be  $K\times K$  invertible matrix, and let  $B=A^{-1}$  its inverse.

Consider Partition

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}; B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

Namely,  $B_{11} = \left[A^{-1}\right]_{11}$  is the upper corner of the  $A^{-1}$ , etc...

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$$A_{11}^{-1} = B_{11} = B_{12}B_{22}^{-1}B_{21}$$
$$A_{11}^{-1}A_{12} = -B_{12}B_{22}^{-1}$$

Hint: Start by noticing that

$$AB = I \iff \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \iff \begin{cases} A_{11}B_{11} + A_{12}B_{21} & = I \\ A_{11}B_{12} + A_{12}B_{22} & = 0 \end{cases}$$

### Part II

## Random variables

Exercise 5. (\*)Let  $y \in \mathcal{Y} \subseteq \mathbb{R}$  be a univariate random variable with CDF  $F_y(\cdot)$ . Consider a bijective function  $h: \mathcal{Y} \to \mathcal{Z}$  with z = h(y), and  $h^{-1}$  its inverse. The PDF of z is

$$F_z(z) = \begin{cases} F_Y(h^{-1}(z)) & \text{if } h \not \\ \\ 1 - F_Y(h^{-1}(z)) & \text{if } h \searrow \end{cases}$$

Exercise 6.  $(\star)$ Let  $y \in \mathcal{Y} \subseteq \mathbb{R}$  be a univariate random variable with PDF  $f_y(\cdot)$ . Consider a bijective function  $h: \mathcal{Y} \to \mathcal{Z} \subseteq \mathbb{R}$  and let  $h^{-1}$  be the inverse function of h. Consider a univariate random variable such that z = h(y).

The PDF of z is

$$f_z(z) = f_y(y) |\det(\frac{dy}{dz})| = f_y(h^{-1}(z)) |\det(\frac{d}{dz}h^{-1}(z))|$$

**Exercise 7.** (\*)Let  $y \sim \operatorname{Ex}(\lambda)$  r.v. with Exponential distribution with rate parameter  $\lambda > 0$ , and  $f_{\operatorname{Ex}(\lambda)}(y) = \lambda \exp(-\lambda y) \mathbb{1}(y \geq 0)$ . Let  $z = 1 - \exp(-\lambda y)$ . Calculate the PDF of z, and recognize its distribution.

#### **Exercise 8.** $(\star)$ Prove the following properties

1. Let matrix  $A \in \mathbb{R}^{q \times d}$ ,  $c \in \mathbb{R}^q$ , and z = c + Ay then

$$E(z) = E(c + Ay) = c + AE(y)$$

2. Let random variables  $z \in \mathcal{Z}$  and  $y \in \mathcal{Y}$ , and let functions  $\psi_1$  and  $\psi_2$  defined on  $\mathcal{Z}$  and  $\mathcal{Y}$ , then

$$E(\psi_1(z) + \psi_2(y)) = E(\psi_1(z)) + E(\psi_2(y))$$

3. If random variables  $z \in \mathcal{Z}$  and  $y \in \mathcal{Y}$  are independent then

$$E(\psi_1(z)\psi_2(y)) = E(\psi_1(z))E(\psi_2(y))$$

for any functions  $\psi_1$  and  $\psi_2$  defined on  $\mathcal{Z}$  and  $\mathcal{Y}$ .

#### **Exercise 9.** $(\star)$ Prove the following properties of the covariance matrix

- 1.  $Cov(z, y) = E(zy^{\top}) E(z) (E(y))^{\top}$
- 2.  $Cov(z, y) = (Cov(y, z))^{\top}$
- 3.  $\operatorname{Cov}_{\pi}(c_1 + A_1 z, c_2 + A_2 y) = A_1 \operatorname{Cov}_{\pi}(x, y) A_2^{\top}$ , for fixed matrices  $A_1, A_2$ , and vectors  $c_1, c_2$  with suitable dimensions.

4. If z and y are independent random vectors then Cov(z, y) = 0

**Exercise 10.** (\*)Prove that the (i, j)-th element of the covariance matrix between vector z and y is the covariance between their elements  $z_i$  and  $y_j$ :

$$[Cov(z, y)]_{i,j} = Cov(z_i, y_j)$$

**Exercise 11.**  $(\star)$ Prove the following properties of Var(Y) for a random vector  $y \in \mathcal{Y} \subseteq \mathbb{R}^d$ 

- 1.  $Var(y) = E(yy^{\top}) E(y) (E(y))^{\top}$
- 2.  $Var(c + Ay) = AVar(y)A^{T}$ , for fixed matrix A, and vectors c with suitable dimensions.
- 3.  $Var(y) \ge 0$ ; (semi-positive definite)

**Exercise 12.**  $(\star)$ Prove the following properties of characteristic functions

- 1.  $\varphi_{A+Bx}(t) = e^{it^T A} \varphi_x(B^T t)$  if  $A \in \mathbb{R}^d$  and  $B \in \mathbb{R}^{k \times d}$  are constants
- 2.  $\varphi_{x+y}(t) = \varphi_x(t)\varphi_y(t)$  if and only if x and y are independent
  - 3. if  $M_x(t) = \mathrm{E}(e^{t^T x})$  is the moment generating function, then  $M_x(t) = \varphi_x(-it)$

Exercise 13. (\*)Show that if  $X \sim \operatorname{Ex}(\lambda)$  then  $\varphi_X(t) = \frac{\lambda}{\lambda - it}$ .

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- 1. Find  $\varphi_X(t)$  if  $X \sim \text{Br}(p)$ .
- 2. Find  $\varphi_Y(t)$  if  $Y \sim \text{Bin}(n, p)$

Exercise 15.  $(\star\star\star\star)$ Prove the following statement related to the Bayesian theorem:

Assume a probability space  $(\Omega, \mathscr{F}, P)$ . Let a random variable  $y : \Omega \to \mathcal{Y}$  with distribution  $F(\cdot)$ . Consider a partition  $y = (x, \theta)$  with  $x \in \mathcal{X}$  and  $\theta \in \Theta$ . Then the probability density function (PDF), or the probability mass function (PMF) of  $\theta | x$  is

$$f(\theta|x) = \frac{f(x|\theta)f(\theta)}{\int f(x|\theta)dF(\theta)}$$
(1)

Hint Consider cases where x is discrete and continuous. In the later case use the mean value theorem:

$$\int_A f(x)g(x)\mathrm{d}x = f(\xi)\int_A g(x)\mathrm{d}x$$

where  $\xi \in A$  if A is connected, and  $g(x) \ge 0$  for  $x \in A$ .

**Exercise 16.**  $(\star)$ Prove that:

1. if 
$$Z \sim N(0, I)$$
 then  $\varphi_Z(t) = \exp(-\frac{1}{2}t^Tt)$ , where  $Z \in \mathbb{R}^d$ 

2. if  $X \sim \mathrm{N}(\mu, \Sigma)$  then  $\varphi_X(t) = \exp(it^T \mu - \frac{1}{2} t^T \Sigma t)$  , where  $X \in \mathbb{R}^d$ 

**Hint:** Assume as known that if  $Z \sim N(0,1)$  then  $\varphi_Z(t) = \exp(-\frac{1}{2}t^2)$ , where  $Z \in \mathbb{R}$ 

Exercise 17.  $(\star)$ Show the following properties of the Characteristic Function

- 1.  $\varphi_x(0) = 1$  and  $|\varphi_x(t)| \le 1$  for all  $t \in \mathbb{R}^d$
- 2.  $\varphi_{A+Bx}(t)=e^{it^TA}\varphi_x(B^Tt)$  if  $A\in\mathbb{R}^d$  and  $B\in\mathbb{R}^{k\times d}$  are constants
- 3. x and y are independent then  $\varphi_{x+y}(t) = \varphi_x(t)\varphi_y(t)$  (we do not proov the other way around)
- 4. if  $M_x(t) = \mathrm{E}(e^{t^T x})$  is the moment generating function, then  $M_x(t) = \varphi_x(-it)$