

# Review Assignment 1

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## 1 Probability theory review

Solve each of the following problems and show all the steps of your working.

1. Show that the covariance matrix is always symmetric and positive semidefinite.

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$$\begin{aligned}\sum_{ij} &= \text{cov}(x_i, x_j) = E[x_i x_j] - E[x_i]E[x_j] \\ &= E[x_j x_i] - E[x_j]E[x_i]\end{aligned}$$

$E$  is symmetric

2.  $X \in \mathbb{R}^n$  and  $Y \in \mathbb{R}^m$  are independent random variables. Their expectations and covariances are  $E[X] = 0$ ,  $\text{Cov}[X] = I$ ,  $E[Y] = \mu$  and  $\text{Cov}[Y] = \sigma I$ , where  $I$  is the identity matrix of the appropriate size and  $\sigma$  is a scalar. What is the expectation and covariance of the random variable  $Z = AX + Y$ , where  $A \in \mathbb{R}^{m \times n}$ ?

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3. Thomas and Viktor are friends. It is Friday night and Thomas does not have a phone. Viktor knows that there is a  $2/3$  probability that Thomas goes to the party to downtown. There are 5 pubs in downtown and there is an equal probability of Thomas going to any of them if he goes to the party. Viktor already looked for Thomas in 4 of the bars. What is the probability of Viktor ending Thomas in the last bar?

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4. Derive the mean for the Beta Distribution, which is defined as

$$\text{Beta}(x|a, b) = \frac{1}{B(a, b)} a^{-1} (1-x) b^{-1} \quad (1)$$

where  $B(a, b), \Gamma(a)$  are Beta and Gamma functions respectively:

$$B(a, b) \triangleq \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} \quad (2)$$

and

$$\Gamma(x) \triangleq \int_0^\infty u^{x-1} e^{-u} du \quad (3)$$

*Hint: Use integration by parts.*

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5. Let  $A \in \mathbb{R}^{n \times n}$  be a positive definite square matrix,  $b \in \mathbb{R}^n$ , and  $c$  be a scalar. Prove that

$$\int_{x \in \mathbb{R}^n} e^{-\frac{1}{2}x^T A x - x^T b - c} dx = \frac{(2\pi)^{n/2} |A|^{-1/2}}{e^{c - \frac{1}{2}b^T A^{-1}b}}$$

*Hint: Use the fact that the integral of the Gaussian probability density function of a random variable with mean  $\mu$  and covariance  $\Sigma$  is 1.*

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6. From the definition of conditional probability of multiple random variables, show that

$$f(x_1, x_2, \dots, x_n) = f(x_1) \prod_{i=2}^n f(x_i | x_1, \dots, x_{i-1})$$

where  $x_1, \dots, x_n$  are random variables and  $f$  is a probability density function of its arguments.

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