

# Machine Learning and Pattern Recognition, Tutorial Sheet Number 1

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**The students have been given answers to these and there should be no need to discuss these in class unless time permits and there is demand**

1. Probability. Lois knows that on average radio station RANDOM-FM plays 1 out of 4 of her requests. If she makes 3 requests, what is the probability that at least one request is played?
2. Let  $A$  and  $\mathbf{v}$  be defined as

$$A = \begin{pmatrix} 1 & 1 & 3 \\ 1 & 1 & -3 \\ 3 & -3 & -3 \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix}$$

Calculate  $A\mathbf{v}$ . Is  $\mathbf{v}$  an eigenvector of  $A$ , and if so what is the corresponding eigenvalue?

3. A random vector  $\mathbf{x}$  has zero mean and a diagonal covariance

$$E(\mathbf{x}\mathbf{x}^T) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

where  $E$  stands for expectation (or mean average) of a random variable. If  $\mathbf{y} = A^T\mathbf{x}$  (using  $A$  from Q1) what is the covariance of the resulting random vector  $\mathbf{y}$ :  $E(\mathbf{y}\mathbf{y}^T)$ ? You may use the fact that expectation is linear:  $E(R\mathbf{x}\mathbf{x}^T S) = RE(\mathbf{x}\mathbf{x}^T)S$ . This shows how covariances change under linear transformations.

4. Find the partial derivatives of the function  $f(x, y, z) = (x + 2y)^2 \sin(xy)$ .
5. Let  $x$  be a Gaussian random variable with mean  $\mu$  and variance  $\sigma^2$ . What is the expected value of  $2x^2$ . Show what form the distribution of  $2(x - \mu)^2$  takes. Hint: the distribution of  $x^2$  for a standard normal ( $N(0,1)$ ) is chi-squared distributed.

**The following ones should be discussed in the tutorial**

6. You have data consisting of records of the form  $(x_1, x_2, x_3, y)$  where  $x_1, x_2$  and  $x_3$  are features, and  $y$  is a class label. Each element takes the value 0 or 1.

Define clearly and precisely the Naive Bayes classification method for obtaining the probability of  $y$  given the values of the other attributes. Show that the maximum likelihood estimate of the parameter  $p(x_i = 0|y = 1)$  is proportional to the number of times attribute  $i$  is 0 for class 1 data. Convince yourself that you could use a similar method to show that  $p(y = 1)$  is proportional to the number of times class 1 occurs in the data.

The training data has the form

$$(1, 0, 1, 1), (0, 1, 0, 1), (0, 1, 1, 0), (0, 0, 0, 0), (1, 1, 1, 1), (0, 0, 0, 0), (1, 0, 1, 0)$$

- Using Naive Bayes on this data, what is the probability that  $y = 1$  given  $x_1 = 1$ ,  $x_2 = 1$  and  $x_3 = 0$ ?
  - Using Naive Bayes on this data, what is the probability that  $y = 0$  given  $x_1 = 1$ ,  $x_2 = 1$  and  $x_3 = 0$ ?
  - Using Naive Bayes on this data, what is the highest posterior classification for  $y$  given  $x_1 = 1$ ?
7. If  $\mathbf{a}$  and  $\mathbf{b}$  are  $D \times 1$  column vectors and  $M$  is a  $D \times D$  symmetric matrix, show that  $\mathbf{a}^T M \mathbf{b} = \mathbf{b}^T M \mathbf{a}$ .

8. Consider the Gaussian distribution

$$p(\mathbf{x}) \propto \exp \left\{ -\frac{1}{2} (\mathbf{x}^T A \mathbf{x} - 2 \mathbf{x}^T \mathbf{b}) \right\}$$

where  $A$  is a symmetric matrix. Show that the mean and covariance are given by

$$\text{mean}(\mathbf{x}) = A^{-1} \mathbf{b}. \quad \text{cov}(\mathbf{x}) = A^{-1}.$$

9. Consider a bivariate Gaussian  $p(x_1, x_2) = \mathcal{N}(x|\boldsymbol{\mu}, \Sigma)$  with

$$\boldsymbol{\mu} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \quad \Sigma = \begin{pmatrix} \sigma_1^2 & \sigma_{12} \\ \sigma_{12} & \sigma_2^2 \end{pmatrix}$$

- a. What is  $p(x_1)$ ?
  - b. Suppose  $\boldsymbol{\mu} = 0$ .  $p(x_1|x_2)$  is a Gaussian. Write down its mean and variance (using the formulae given in the Gaussian lecture).
  - c. Contrast your answers from parts a. and b. Sketch the conditional mean as a function of  $x_2$ .
  - d. **Bonus question.** Derive the form of  $p(x_1|x_2)$  from first principles. HINT:  $p(x_1|x_2) \propto p(x_1, x_2)$  when  $x_2$  is viewed as fixed.
10. (If there is time.) Suppose that  $X$  and  $Y$  are independent and each is uniformly distributed on  $(0, 1)$ . Let  $U = X + Y$  and  $V = X - Y$ .
- Sketch the range (or region of non zero probability) of  $(X, Y)$  and the range of  $(U, V)$ .
  - Find the density function of  $(U, V)$ .
  - Find the density function of  $U$ .
  - Find the density function of  $V$ .