## Imperial College London

# Coursework 1

### IMPERIAL COLLEGE LONDON

DEPARTMENT OF COMPUTING

# **Mathematics for Machine Learning**

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Date: October 15, 2019

#### 1 Statistics and Probabilities

#### Question 1

Given the dataset  $\mathcal{D}$ :

$$\mathcal{D} = \left\{ \begin{bmatrix} 1\\2\\3 \end{bmatrix}, \begin{bmatrix} -1\\0\\0 \end{bmatrix}, \begin{bmatrix} -4\\4\\2 \end{bmatrix} \right\} \tag{1}$$

The sample mean  $\overline{\mathcal{D}}$  is given by a column vector whose  $j^{th}$  element is equal to the mean of the  $j^{th}$  element in each of the vectors in the dataset, where N is the number of samples in the dataset  $\mathcal{D}$ :

$$\overline{\mathcal{D}}_j = \frac{1}{N} \sum_{i=1}^N \mathcal{D}_{ij}$$
 (2)

$$\overline{\mathcal{D}} = \frac{1}{N} \sum_{i=1}^{N} \overline{\mathcal{D}_i}$$
 (3)

For N = 3, the sample mean  $\overline{\mathcal{D}}$  is therefore:

$$\overline{\mathcal{D}} = \frac{1}{3} \begin{bmatrix} 1 - 1 - 4 \\ 2 + 0 + 4 \\ 3 + 0 + 2 \end{bmatrix} = \begin{bmatrix} \frac{-4}{3} \\ 2 \\ \frac{5}{3} \end{bmatrix}$$
 (4)

The sample covariance matrix using 1/N (biased estimate)  $Cov(\mathcal{D})$  is computed as followswhere  $\overline{\mathcal{D}}$  is the sample mean column matrix and  $\mathcal{D}_i$  is the  $i^{th}$  sample in the dataset  $\mathcal{D}$ :

$$Cov(\mathcal{D}) = \frac{1}{N} \sum_{i=1}^{N} (\mathcal{D}_{i} - \overline{\mathcal{D}}) (\mathcal{D}_{i} - \overline{\mathcal{D}})^{T}$$
 (5)

 $Cov(\mathcal{D})$  is a KxK matrix where K is the number of entries in each sample column vector:

$$Cov(\mathcal{D}) = \frac{1}{3} \begin{bmatrix} \frac{1}{9} \begin{bmatrix} 49 & 0 & 28\\ 0 & 0 & 0\\ 28 & 0 & 16 \end{bmatrix} + \frac{1}{9} \begin{bmatrix} 1 & -6 & -5\\ -6 & 36 & 30\\ -5 & 30 & 25 \end{bmatrix} + \frac{1}{9} \begin{bmatrix} 64 & -48 & -8\\ -48 & 36 & -6\\ -8 & -6 & 1 \end{bmatrix}$$
 (6)

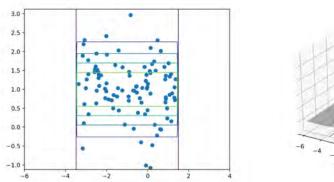
$$= \begin{bmatrix} \frac{38}{9} & -2 & \frac{5}{9} \\ -2 & \frac{8}{3} & \frac{4}{3} \\ \frac{5}{9} & \frac{4}{3} & \frac{14}{9} \end{bmatrix} \tag{7}$$

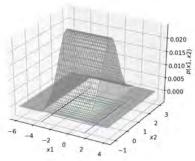
#### Question 2

Below shows two sets of data, each with 100 data points with mean

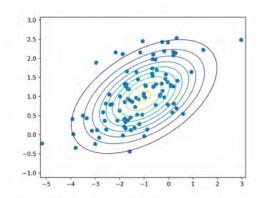
$$\mu = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

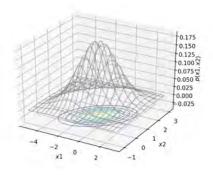
and marginal variances  $\sigma_1^2 = 2$  and  $\sigma_2^2 = 0.5$ .





**Figure 1:** Dataset 1, A joint distribution of a uniform distribution with mean = -1 and variance = 2 and a Gaussian distribution with mean = 1 and variance = 0.5





**Figure 2:** Dataset 2, A bivariate Gaussian distribution with mean = [-1,1] and marginal variances= [2,0.5]

Both of the datasets have the same mean,  $\mu$  and marginal variances  $\sigma_1^2$  and  $\sigma_2^2$ . However, the shape of the two distributions are clearly different as shown in Figure 1 and 2. The shapes of the distributions have been ensured to be different. The two entries of data points in dataset 1 have been generated from a different one-dimensional distribution, where  $x_1$  is chosen from a normal distribution:

$$\mathcal{U}(-\frac{\sqrt{24}}{2} - 1, \frac{\sqrt{24}}{2} - 1) \tag{8}$$

and  $x_2$  is chosen from a one-dimensional Gaussian distribution:

$$\mathcal{N}(1,0.5) \tag{9}$$

While dataset 2 is generated from a bivariate Gaussian distribution with mean  $\mu$  and the covariance matrix

$$Cov(D) = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 2 \end{bmatrix}$$
 (10)

The shape of the distribution is tilted due to the covariance between the univariate Gaussian distribution that forms the bivariate Gaussian distribution.

#### Question 3

The probability of successful compilation is described by the Bernoulli distribution:

$$P(x|\mu) = \mu^{x} (1-\mu)^{1-x} \tag{11}$$

The likelihood is given by a binomial distribution with mean  $\mu$  and the number of trials N = 20:

$$P(x|\mu, 20) = {20 \choose x} \mu^x (1-\mu)^{20-x}$$
 (12)

The conjugate prior beta distribution with parameters  $\alpha = 2$  and  $\beta = 2$  is given by:

$$P(\mu) = \mu^{\alpha - 1} (1 - \mu)^{\beta - 1} \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)}$$
(13)

(14)

The joint distribution is given by:

$$P(x|\mu, 20)P(\mu) = {20 \choose x} \mu^{x} (1-\mu)^{20-x} \mu^{\alpha-1} (1-\mu)^{\beta-1} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}$$
(15)

$$= {20 \choose x} \mu^{x+\alpha-1} (1-\mu)^{20-x+\beta-1} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}$$
 (16)

The evidence P(x) is the marginal distribution of x:

$$P(x) = \int_0^1 {20 \choose x} \mu^{x+\alpha-1} (1-\mu)^{20-x+\beta-1} \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} d\mu$$
 (17)

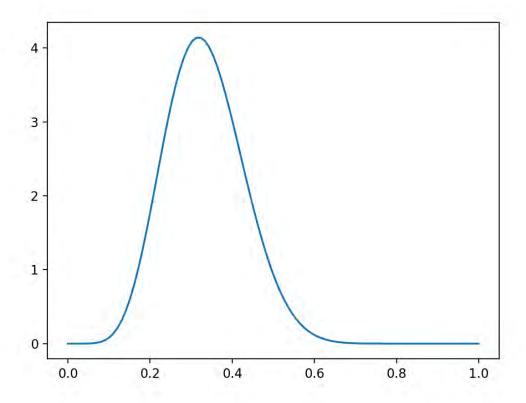
$$= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} {20 \choose x} \frac{\Gamma(\alpha+x)\Gamma(\beta+20-x)}{\Gamma(\alpha+\beta+20)}$$
(18)

For successes x = 6,  $\alpha = 2$  and  $\beta = 2$ , the posteria distribution on  $\mu$  is therefore:

$$\frac{P(x|\mu,20)P(\mu)}{P(x)} = \frac{\Gamma(\alpha+\beta+20)}{\Gamma(\alpha+x)\Gamma(\beta+20-x)} \mu^{x+\alpha-1} (1-\mu)^{20-x+\beta-1}$$
(19)

$$=\frac{\Gamma(24)}{\Gamma(8)\Gamma(16)}\mu^7(1-\mu)^{15}$$
 (20)

$$=3922511\mu^{7}(1-\mu)^{15} \tag{21}$$



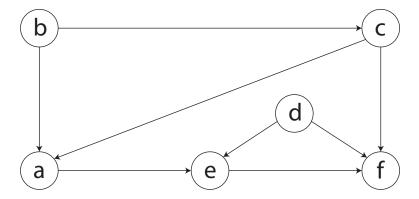
**Figure 3:** The plot of the posterior distribution on  $\mu$ 

Both the prior and the posterior are beta distributions, as the beta distribution is a conjugate prior of the Bernoulli distribution and the binomial distribution. The hyperparameters  $\alpha$  and  $\beta$  have however changed from  $\alpha = 2$  and  $\beta = 2$  to  $\alpha' = \alpha + x = 8$  and  $\beta' = \beta + N - x = 16$ , where  $\alpha'$  and  $\beta'$  are the new hyperparameters of the posterior beta distribution, N = 20 is the number of trials and x = 6 is the number of observed success.

# 2 Graphical Models

# Question 1

$$p(a, b, c, d, e, f) = p(a|b, c)p(c|b)p(d)p(e|d, a)p(f|c, d, e)$$
(22)



**Figure 4:** The graphical model of p(a, b, c, d, e, f)

#### **Question 2**

- (a) The statement  $a \perp \!\!\! \perp g | k$  is true, all paths from a to g are blocked by e as e is not in the set  $\{k\}$ .
- (b) The statement  $d \perp \!\!\! \perp h | i$  is false, one unblocked path is  $d \to j \to h$ .
- (c) The statement  $g \perp \perp a | c$  is true, all paths from g to a are blocked by e as e is not in the set  $\{c\}$ .
- (d) The statement  $e \perp j \mid k$  is false, one unblocked path is  $e \rightarrow d \rightarrow j$ .
- (e) The statement  $v \perp \!\!\! \perp e \mid j$  is false, one unblocked path is  $b \rightarrow d \rightarrow e$ .
- (f) The statement  $j \perp \!\!\! \perp c | \{k, g\}$  is false, one unblocked path is  $j \rightarrow h \rightarrow i \rightarrow c$ .
- (g) The statement  $a \perp \!\!\! \perp k$  is false, one unblocked path is  $a \rightarrow b \rightarrow d \rightarrow k$ .
- (h) The statement  $a \perp \!\!\! \perp k | e$  is false, one unblocked path is  $a \rightarrow b \rightarrow d \rightarrow k$ .
- (i) The statement  $h \perp \perp d | \{j, e\}$  is false, one unblocked path is  $h \rightarrow i \rightarrow d$ .
- (j) The statement  $b \perp \!\!\! \perp h|e$  is false, one unblocked path is  $b \to j \to h$ , as e is a descendant of j which is in the set  $\{e\}$ .
- (k) The statement  $h \perp \!\!\! \perp c | d$  is false, one unblocked path is  $h \rightarrow i \rightarrow c$ .
- (1) The statement  $a \perp \perp f | k$  is false, one unblocked path is  $a \rightarrow b \rightarrow j \rightarrow f$ , as k is a descendant of j which is in the set  $\{k\}$ .
- (m) The statement  $i \perp \!\!\! \perp a | j$  is false, one unblocked path is  $i \to h \to j \to b \to a$ , as j is in the set  $\{j\}$ .
- (n) The statement  $g \perp \!\!\!\perp h | e$  is false, one unblocked path is  $g \rightarrow e \rightarrow d \rightarrow j \rightarrow h$ , as e is in the set  $\{e\}$ .
- (o) The statement  $g \perp \!\!\! \perp f | d$  is true, all paths from g to f are blocked by e as e is not in the set  $\{d\}$ .
- (p) The statement  $g \perp \!\!\! \perp h$  is true, all paths from g to h are blocked by e as e is not in the set  $\emptyset$ .
- (q) The statement  $a \perp \!\!\! \perp i | g$  is true, all paths from a to i are blocked by j and d as neither has descendants in the set  $\{g\}$ .
- (r) The statement  $b \perp \!\!\! \perp h$  is true, all paths from b to h are blocked by j and d as neither has descendants in the set  $\emptyset$ .
- (s) The statement  $b \perp \!\!\! \perp h | g$  is true, all paths from b to h are blocked by j and d as neither has descendants in the set  $\{g\}$ .
- (t) The statement  $i \perp \!\!\! \perp a$  is true, all paths from i to a are blocked by j and d as neither has descendants in the set  $\emptyset$ .