# My Solutions for Exercises of Deep Learning Fundamentals by Bishop

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- 1 The Deep Learning Revolution
- 1.1 No Exercises

#### **Probabilities** $\mathbf{2}$

### Exercise 2.1

Bayes rule

$$P[C=1|T=1] = \frac{P[T=1|C=1] * P[C=1]}{P[T=1|C=1] * P[C=1] + P[T=1|C=0] * P[C=0]}$$

$$P[C=1|T=1] = \frac{0.90 * 0.001}{0.90 * 0.001 + 0.03 * 0.999} = 0.0292$$
(2)

$$P[C=1|T=1] = \frac{0.90*0.001}{0.90*0.001 + 0.03*0.999} = 0.0292$$
 (2)

Given that the test result was positive, there is a 2.92% chance that you have cancer.

#### 2.1 Exercise 2.2

Not attempted

### Exercise 2.3

$$p(\mathbf{y}) = \int p_{\mathbf{u}, \mathbf{v}}(\mathbf{u}, \mathbf{y} - \mathbf{u}) d\mathbf{u}$$

$$= \int p_{\mathbf{u}}(\mathbf{u}) p_{\mathbf{v}}(\mathbf{y} - \mathbf{u}) d\mathbf{u}$$
(3)

$$= \int p_{\mathbf{u}}(\mathbf{u}) p_{\mathbf{v}}(\mathbf{y} - \mathbf{u}) d\mathbf{u} \tag{4}$$

#### 2.2 Exercise 2.4

Not attempted

### Exercise 2.5

Exponential:

$$p(x|\lambda) = \lambda e^{-\lambda x} \tag{5}$$

Laplace:

$$p(x|\mu,\gamma) = \frac{1}{2\gamma} e^{-\frac{|x-\mu|}{\gamma}} \tag{6}$$

Verifying that the exponential distribution is normalized:

$$p(x|\lambda) = \lambda e^{-\lambda x} \tag{7}$$

$$\int_0^\infty \lambda e^{-\lambda x} = -e^{-\lambda x} \Big|_0^\infty = \frac{1}{e^\infty} + \frac{1}{e^0}$$
 (8)

$$=1 \tag{9}$$

Verifying the laplace distribution:

$$p(x|\mu,\gamma) = \frac{1}{2\gamma} e^{-\frac{|x-\mu|}{\gamma}} \tag{10}$$

$$\begin{cases} \frac{1}{2\gamma} e^{-\frac{x-\mu}{\gamma}} & \text{if } x \ge \mu\\ \frac{1}{2\gamma} e^{-\frac{-x+\mu}{\gamma}} & \text{if } x < \mu \end{cases}$$
 (11)

$$\int_{\mu}^{\infty} \frac{1}{2\gamma} e^{-\frac{x-\mu}{\gamma}} = \frac{1}{2} e^{-\frac{x-\mu}{\gamma}} \Big|_{\mu}^{\infty} = -\frac{1}{2} (e^{-\infty} - e^0)$$
 (12)

$$=\frac{1}{2}\tag{13}$$

$$= \frac{1}{2}$$

$$\int_{-\infty}^{\mu} \frac{1}{2\gamma} e^{-\frac{-x+\mu}{\gamma}} = \frac{1}{2} e^{-\frac{-x+\mu}{\gamma}} \Big|_{-\infty}^{\mu} = \frac{1}{2} (e^0 - e^{-\infty})$$
(13)

$$=\frac{1}{2}\tag{15}$$

$$\frac{1}{2} + \frac{1}{2} = 1\tag{16}$$

#### 2.3 Exercise 2.6

Not attempted

#### 2.4 Exercise 2.7

$$P(x|D) = \frac{1}{N} \sum_{n=1}^{N} \delta(x - x_n)$$
 (17)

$$E[f] = \int p(x)f(x)dx \tag{18}$$

$$E[f] = \int \frac{1}{N} \sum_{n=1}^{N} \delta(x - x_n) f(x) dx$$
(20)

$$E[f] = \frac{1}{N} \sum_{n=1}^{N} \int_{x_n - \varepsilon}^{x_n + \varepsilon} \delta(x - x_n) f(x) dx$$
 (21)

$$E[f] = \frac{1}{N} \sum_{n=1}^{N} f(x_n) \int_{x_n - \varepsilon}^{x_n + \varepsilon} \delta(x - x_n) dx$$
 (22)

$$E[f] = \frac{1}{N} \sum_{n=1}^{N} f(x_n)$$
 (23)

#### 2.5Exercise 2.8

Not attempted

### Exercise 2.9

$$cov[X, y] = E_{x,y}[xy] - E[x]E[y]$$

$$(24)$$

If x and y are independent, the joint distribution is equal to the product of the marginals. p(x,y) = p(x)p(y). If  $E_{x,y}[xy] = E[x]E[y]$ , then the covariance will be zero.

#### 2.7 Exercise 2.10

### 2.8 Exercise 2.11

Proving  $E[x] = E_y[E_x[x|y]]$ :

$$E_x[x|y] = \int p(x|y)xdx \tag{25}$$

$$E[x] = E_y[\int p(x|y)xdx] \tag{27}$$

$$E[x] = \int E_y[p(x|y)]xdx \tag{28}$$

$$E[x] = \int \int p(x|y)xp(y)dxdy \tag{29}$$

$$E[x] = \int \int \frac{p(x,y)}{p(y)} x p(y) dx dy$$
(30)

$$E[x] = \int \int p(x,y)x dx dy \tag{31}$$

$$E[x] = E[x] \tag{32}$$

### 2.9 Exercise 2.12

Not attempted

### 2.10 Exercise 2.13

$$E[x] = \int_{-\infty}^{\infty} x \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{1}{2\sigma^2}(x-\mu)^2} dx$$
 (33)

Change of variables 
$$z = \frac{x - \mu}{\sigma}, \sigma dz = dx$$
 (34)

$$E[x] = \int_{-\infty}^{\infty} \sigma \frac{\sigma z + \mu}{\sqrt{2\pi\sigma^2}} e^{\frac{1}{2}z^2} dz$$
 (35)

$$E[x] = \frac{\sigma}{\sqrt{2\pi}} \int_{-\infty}^{\infty} z e^{\frac{1}{2}z^2} + \frac{\mu}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\frac{1}{2}z^2}$$
 (36)

$$E[x] = \frac{\sigma}{\sqrt{2\pi}} * 0 + \frac{\mu}{\sqrt{2\pi}} * \sqrt{2\pi}$$
 (37)

$$E[x] = \mu \tag{38}$$

### 2.11 Exercise 2.14

### 2.12 Exercise 2.15

Solving for  $\mu_{ml}$ :

$$\log p(x|\mu, \sigma^2) = \frac{-1}{2\sigma^2} \sum_{n=1}^{N} (x_n - \mu)^2 - \frac{N}{2} \log \sigma^2 - \frac{N}{2} \log 2\pi$$
 (39)

$$\frac{d}{d\mu}log p(x|\mu,\sigma^2) = \frac{1}{2\sigma^2} \sum_{n=1}^{N} 2(x_n - \mu)$$

$$\tag{40}$$

$$0 = \frac{1}{\sigma^2} \sum_{n=1}^{N} (x_n - \mu) \tag{41}$$

$$0 = \sum_{n=1}^{N} x_n - \sum_{n=1}^{N} \mu \tag{42}$$

$$N\mu = \sum_{n=1}^{N} x_n \tag{43}$$

$$\mu_{ml} = \frac{1}{n} \sum_{n=1}^{N} x_n \tag{44}$$

Solving for  $\sigma_{ml}$ :

$$\log p(x|\mu, \sigma^2) = \frac{-1}{2\sigma^2} \sum_{n=1}^{N} (x_n - \mu)^2 - \frac{N}{2} \log \sigma^2 - \frac{N}{2} \log 2\pi$$
 (45)

$$\frac{d}{d\sigma^2}logp(x|\mu,\sigma^2) = \frac{1}{2\sigma^4} \sum_{n=1}^{N} (x_n - \mu)^2 - \frac{N}{2\sigma^2}$$
(46)

$$\frac{N}{2\sigma^2} = \frac{1}{2\sigma^4} \sum_{n=1}^{N} (x_n - \mu)^2 \tag{47}$$

$$\sigma_{ml}^2 = \frac{1}{N} \sum_{m=1}^{N} (x_n - \mu_{ml})^2 \tag{48}$$

### 2.13 Exercise 2.16

not attempted

### 2.14 Exercise 2.17

Finding expectation of  $\hat{\sigma}^2$ 

$$E[\hat{\sigma}^2] = E[\frac{1}{N} \sum_{n=1}^{N} (x_n - \mu)^2]$$
(49)

$$= \frac{1}{N} \sum_{n=1}^{N} E[x_n^2 - 2x_n\mu + \mu^2]$$
 (50)

$$= \frac{1}{N} \sum_{n=1}^{N} E[x_n^2] - E[2x_n\mu] + E[\mu^2]$$
 (51)

$$= \frac{1}{N} \sum_{n=1}^{N} E[x_n^2] - 2E[x_n]E[x_n] + E[\mu^2]$$
 (52)

$$= \frac{1}{N} \sum_{n=1}^{N} E[x_n^2] - E[x_n]^2$$
 (53)

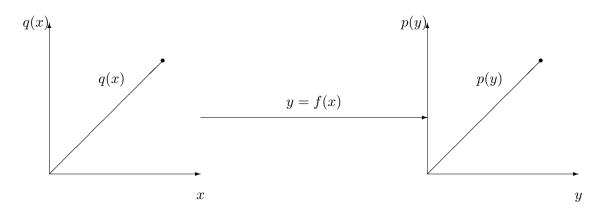
$$=\frac{1}{N}\sum_{n=1}^{N}\mu^2 + \sigma^2 - \mu^2 \tag{54}$$

$$=\sigma^2\tag{55}$$

### 2.15 Exercise 2.18

Not attempted

### 2.16 Exercise 2.19



### 2.17 Exercise 2.20

#### 2.18 Exercise 2.21

Showing  $h(p^2) = 2h(p)$ :

$$h(p) = h(p(x_1)) + h(p(x_2)) + \dots + h(p(x_n))$$
(56)

$$h(p^2) = h(x_1^2) + h(x_2^2) + \dots + h(x_n^2)$$
(57)

$$\therefore h(x) = -\log_2 p(x),\tag{58}$$

$$h(p^2) = 2h(x_1) + 2h(x_2) + \dots + 2h(x_n)$$
(59)

$$h(p^2) = 2h(p) \tag{60}$$

This can be applied to any exponent which inclues any choice of n integer or  $\frac{n}{m}$  positive rational number.

$$h(p^x) = xh(p) \ \forall \ Q^+ \tag{61}$$

$$\therefore \qquad (62)$$

$$h(p) \propto lnp \tag{62}$$

#### 2.19 Exercise 2.22

Not attempted

#### 2.20Exercise 2.23

Not attempted

#### 2.21Exercise 2.24

Not attempted

#### 2.22 Exercise 2.25

Not attempted

#### Exercise 2.26 2.23

Not attempted

#### 2.24Exercise 2.27

Not attempted

#### 2.25 Exercise 2.28

Not attempted

#### 2.26 Exercise 2.29

### 2.27 Exercise 2.30

Not attempted

### 2.28 Exercise 2.31

Not attempted

### 2.29 Exercise 2.32

Not attempted

### 2.30 Exercise 2.33

Not attempted

### 2.31 Exercise 2.34

Not attempted

### 2.32 Exercise 2.35

Not attempted

### 2.33 Exercise 2.36

Not attempted

### 2.34 Exercise 2.37

Not attempted

### 2.35 Exercise 2.38

Not attempted

### 2.36 Exercise 2.39

Not attempted

### 2.37 Exercise 2.40

Not attempted

### 2.38 Exercise 2.41