



**Deep Learning**

**Assignment- Week 2**

**TYPE OF QUESTION: MCQ/MSQ**

**Number of questions: 10**

**Total mark: 10 X 2 = 20**

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**QUESTION 1:**

Suppose if you are solving an n-class problem, how many discriminant function you will need for solving?

- a.  $n-1$
- b.  $n$
- c.  $n+1$
- d.  $n-2$

**Correct Answer: b**

**Detailed Solution: For n class problem we need n number of discriminant function.**

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**QUESTION 2:**

If we choose the discriminant function  $g_i(x)$  as a function of posterior probability. i.e.  $g_i(x) = f(p(w_i/x))$ . Then which of following cannot be the function  $f()$ ?

- a.  $f(x) = a^x, \text{ where } a > 1$
- b.  $f(x) = a^{-x}, \text{ where } a > 1$
- c.  $f(x) = 2x + 3$
- d.  $f(x) = \exp(x)$

**Correct Answer: b**

**Detailed Solution:**

**The function  $f()$  should be a monotonic increasing function.**

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**QUESTION 3:**

What will be the nature of decision surface when the covariance matrices of different classes are identical but otherwise arbitrary? (Given all the classes has equal class probabilities)

- a. Always orthogonal to two surfaces
- b. Generally not orthogonal to two surfaces
- c. Bisector of the line joining two mean, but not always orthogonal to two surface.
- d. Arbitrary

**Correct Answer: c**

**Detailed Solution:**

**Options are self-explanatory.**

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**QUESTION 4:**

The mean and variance of all the samples of two different normally distributed class  $\omega_1$  and  $\omega_2$  are given

$$\mu_1 = \begin{bmatrix} 3 \\ 6 \end{bmatrix}; \Sigma_1 = \begin{bmatrix} 1/2 & 0 \\ 0 & 2 \end{bmatrix} \text{ and } \mu_2 = \begin{bmatrix} 3 \\ -2 \end{bmatrix}; \Sigma_2 = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

What will be the value expression of decision boundary between these two classes if both the class has equal class probability 0.5? For the input sample  $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  consider  $g_i(x) = x^t - \frac{1}{2}\Sigma_i^{-1}x + \Sigma_i^{-1}\mu_i x - \frac{1}{2}\mu_i^t \Sigma_i^{-1} \mu_i - \frac{1}{2}\ln|\Sigma_i| + \ln|P(\omega_i)|$

- a.  $x_2 = 3.514 - 1.12x_1 + 0.187x_1^2$
- b.  $x_1 = 3.514 - 1.12x_2 + 0.187x_2^2$
- c.  $x_1 = 0.514 - 1.12x_2 + 0.187x_2^2$
- d.  $x_2 = 0.514 - 1.12x_2 + 0.187x_2^2$

**Correct Answer: a**

**Detailed Solution:**

**This is the most general case of discriminant function for normal density. The inverse matrices are**



$$\Sigma_1^{-1} = \begin{bmatrix} 2 & 0 \\ 0 & 1/2 \end{bmatrix}, \text{ and } \Sigma_2^{-1} = \begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix}$$

Setting  $g_1(x) = g_2(x)$  we get the decision boundary as  $x_2 = 3.514 - 1.12x_1 + 0.187x_1^2$

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### **QUESTION 5:**

For a two class problem, the linear discriminant function is given by  $g(x) = a'y$ . What is the updating rule for finding the weight vector  $a$ . Here  $y$  is augmented feature vector.

- Adding the sum of all augmented feature vector which are misclassified multiplied by the learning rate to the current weigh vector.
- Subtracting the sum of all augmented feature vector which are misclassified multiplied by the learning rate from the current weigh vector.
- Adding the sum of the all augmented feature vector belonging to the positive class multiplied by the learning rate to the current weigh vector.
- Subtracting the sum of all augmented feature vector belonging to the negative class multiplied by the learning rate from the current weigh vector.

**Correct Answer: a**

**Detailed Solution:**

$$a(k+1) = a(k) + \eta \sum y$$

**For derivation refer to video lectures.**

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### **QUESTION 6:**

For minimum distance classifier which of the following must be satisfied?

- All the classes should have identical covariance matrix and diagonal matrix.
- All the classes should have identical covariance matrix but otherwise arbitrary.
- All the classes should have equal class probability.
- None of above.

**Correct Answer: c**

**Detailed Solution: Options are self-explanatory.**



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**QUESTION 7:**

Which of the following is the updating rule of gradient descent algorithm? Here  $\nabla$  is gradient operator and  $\eta$  is learning rate.

- a.  $a_{n+1} = a_n - \eta \nabla F(a_n)$
- b.  $a_{n+1} = a_n + \eta \nabla F(a_n)$
- c.  $a_{n+1} = a_n - \eta \nabla F(a_{n-1})$
- d.  $a_{n+1} = a_n + \eta \nabla F(a_{n-1})$

**Correct Answer: a**

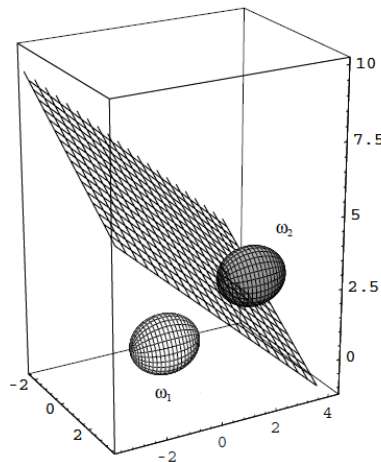
**Detailed Solution:**

Gradient descent is an optimization algorithm used to minimize some function by iteratively moving in the direction of steepest descent as defined by the negative of the gradient.

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**QUESTION 8:**

The decision surface between two normally distributed class  $\omega_1$  and  $\omega_2$  is shown on the figure. Can you comment which of the following is true?



- a.  $p(\omega_1) = p(\omega_2)$
- b.  $p(\omega_2) > p(\omega_1)$
- c.  $p(\omega_1) > p(\omega_2)$
- d. None of the above.

**Correct Answer: c**

**Detailed Solution:**

**If the prior probabilities are not equal, the optimal boundary hyperplane is shifted away from the more likely mean.**

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**QUESTION 9:**



In k-nearest neighbour's algorithm (k-NN), how we classify an unknown object?

- a. Assigning the label which is most frequent among the  $k$  nearest training samples.
- b. Assigning the unknown object to the class of its nearest neighbour among training sample.
- c. Assigning the label which is most frequent among the all training samples except the  $k$  farthest neighbor.
- d. None of this.

**Correct Answer: a**

**Detailed Solution:**

**Options are self-explanatory.**

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**QUESTION 10:**

What is the direction of weight vector w.r.t. decision surface for linear classifier?

- a. Parallel
- b. Normal
- c. At an inclination of 45
- d. Arbitrary

**Correct Answer: b**

**Detailed Solution:**

**Options are self-explanatory.**

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