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State Finished

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Time taken 13 mins 57 secs

Grade 10.00 out of 10.00 (100%)

Question **1**

Correct

Mark 1.00 out of 1.00

Consider the ML example below for prediction of sales based on advertising

Year	Sales (Million Euro)	Advertising (Million Euro)
1	651	23
2	762	26
3	856	30
4	1,063	34
5	1,190	43
6	1,298	48
7	1,421	52
8	1,440	57
9	1,518	58

In this example, Advertising is the

- ☐ Response
- ☒ Regressor
- ☐ Regression coefficient
- ☐ Model error



Your answer is correct.

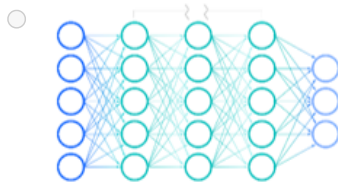
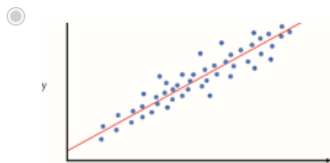
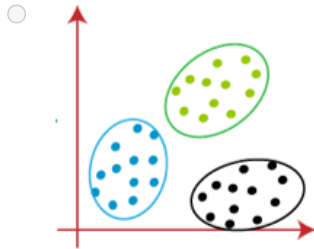
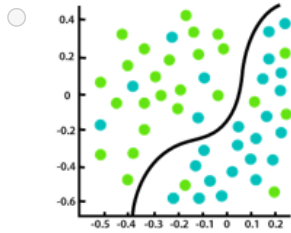
The correct answer is: Regressor

Question 2

Correct

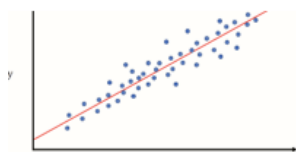
Mark 1.00 out of 1.00

Which figure below represents linear regression



Your answer is correct.

The correct answer is:



Question **3**

Correct

Mark 1.00 out of 1.00

Consider the linear regression model below

$$y(k) = h_0 + h_1x_1(k) + \cdots + h_nx_n(k) + \epsilon(k)$$

The quantities h_i are

- ☐ Regressor
- ☐ Response
- ☐ Model error
- ☒ Regression coefficient



Your answer is correct.

The correct answer is: Regression coefficient

Question 4

Correct

Mark 1.00 out of 1.00

The learning model for the linear regression problem described in class is

- ☐
$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(M) \end{bmatrix}}_{\bar{\mathbf{y}}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}(1) \\ \bar{\mathbf{x}}(2) \\ \vdots \\ \bar{\mathbf{x}}(M) \end{bmatrix}}_{\bar{\mathbf{X}}} \bar{\mathbf{h}} + \underbrace{\begin{bmatrix} \epsilon(1) \\ \epsilon(2) \\ \vdots \\ \epsilon(M) \end{bmatrix}}_{\bar{\boldsymbol{\epsilon}}}$$
- ☒
$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(M) \end{bmatrix}}_{\bar{\mathbf{y}}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}^T(1) \\ \bar{\mathbf{x}}^T(2) \\ \vdots \\ \bar{\mathbf{x}}^T(M) \end{bmatrix}}_{\bar{\mathbf{X}}} \bar{\mathbf{h}} + \underbrace{\begin{bmatrix} \epsilon(1) \\ \epsilon(2) \\ \vdots \\ \epsilon(M) \end{bmatrix}}_{\bar{\boldsymbol{\epsilon}}}$$
- ☐
$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(M) \end{bmatrix}}_{\bar{\mathbf{y}}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}(1) \\ \bar{\mathbf{x}}(2) \\ \vdots \\ \bar{\mathbf{x}}(M) \end{bmatrix}}_{\bar{\mathbf{X}}} \bar{\mathbf{h}}^T + \underbrace{\begin{bmatrix} \epsilon(1) \\ \epsilon(2) \\ \vdots \\ \epsilon(M) \end{bmatrix}}_{\bar{\boldsymbol{\epsilon}}}$$
- ☐
$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(M) \end{bmatrix}}_{\bar{\mathbf{y}}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}^T(1) \\ \bar{\mathbf{x}}^T(2) \\ \vdots \\ \bar{\mathbf{x}}^T(M) \end{bmatrix}}_{\bar{\mathbf{X}}} \bar{\mathbf{h}}^T + \underbrace{\begin{bmatrix} \epsilon(1) \\ \epsilon(2) \\ \vdots \\ \epsilon(M) \end{bmatrix}}_{\bar{\boldsymbol{\epsilon}}}$$



Your answer is correct.

The correct answer is:

$$\underbrace{\begin{bmatrix} y(1) \\ y(2) \\ \vdots \\ y(M) \end{bmatrix}}_{\bar{\mathbf{y}}} = \underbrace{\begin{bmatrix} \bar{\mathbf{x}}^T(1) \\ \bar{\mathbf{x}}^T(2) \\ \vdots \\ \bar{\mathbf{x}}^T(M) \end{bmatrix}}_{\bar{\mathbf{X}}} \bar{\mathbf{h}} + \underbrace{\begin{bmatrix} \epsilon(1) \\ \epsilon(2) \\ \vdots \\ \epsilon(M) \end{bmatrix}}_{\bar{\boldsymbol{\epsilon}}}$$

Question 5

Correct

Mark 1.00 out of 1.00

The regression coefficient vector from the training data is determined as

- ☐ $\bar{\mathbf{h}} = \mathbf{X}^T (\mathbf{X}^T \mathbf{X})^{-1} \bar{\mathbf{y}}$
- ☐ $\bar{\mathbf{h}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X} \bar{\mathbf{y}}$
- ☐ $\bar{\mathbf{h}} = (\mathbf{X} \mathbf{X}^T)^{-1} \mathbf{X}^T \bar{\mathbf{y}}$
- ☒ $\bar{\mathbf{h}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \bar{\mathbf{y}}$



Your answer is correct.

The correct answer is:

$$\bar{\mathbf{h}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \bar{\mathbf{y}}$$

Question **6**

Correct

Mark 1.00 out of 1.00

Consider the linear regression problem with the design matrix \mathbf{X} and response vector $\bar{\mathbf{y}}$ given below

$$\mathbf{X} = \begin{bmatrix} 1 & -1 \\ -1 & 1 \\ 1 & 1 \\ -1 & -1 \end{bmatrix}, \bar{\mathbf{y}} = \begin{bmatrix} -2 \\ 2 \\ -3 \\ -1 \end{bmatrix}$$

The vector of regression coefficients is

- ☐ $\frac{1}{2} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$
- ☐ $\frac{1}{2} \begin{bmatrix} 2 \\ -3 \end{bmatrix}$
- ☒ $\frac{1}{2} \begin{bmatrix} -3 \\ 1 \end{bmatrix}$
- ☐ $\frac{1}{2} \begin{bmatrix} -1 \\ 2 \end{bmatrix}$



Your answer is correct.

The correct answer is:

$$\frac{1}{2} \begin{bmatrix} -3 \\ 1 \end{bmatrix}$$

Question **7**

Correct

Mark 1.00 out of 1.00

Logistic regression can be used in which of the following applications

- ☒ Disease detection
- ☐ Stock price forecasting
- ☐ Predicting the price of a home
- ☐ Clustering of users based on shopping information



Your answer is correct.

The correct answer is: Disease detection

Question 8

Correct

Mark 1.00 out of 1.00

As $z \rightarrow -\infty$, $z \rightarrow \infty$, the logistic function approaches the limits

- ☐ 1,0
- ☒ 0,1
- ☐ 0, ∞
- ☐ ∞ ,0



Your answer is correct.

The correct answer is:

0,1

Question 9

Correct

Mark 1.00 out of 1.00

The log-likelihood of the regression parameter $\bar{\mathbf{h}}$ in logistic regression can be written as

- ☐ $\sum_{k=1}^M (1 - y(k)) \ln g(\bar{\mathbf{x}}(k)) + y(k) \ln (1 - g(\bar{\mathbf{x}}(k)))$
- ☐ $\prod_{k=1}^M (g(\bar{\mathbf{x}}(k)))^{y(k)} (1 - g(\bar{\mathbf{x}}(k)))^{1-y(k)}$
- ☐ $\prod_{k=1}^M (g(\bar{\mathbf{x}}(k)))^{1-y(k)} (1 - g(\bar{\mathbf{x}}(k)))^{y(k)}$
- ☒ $\sum_{k=1}^M y(k) \ln g(\bar{\mathbf{x}}(k)) + (1 - y(k)) \ln (1 - g(\bar{\mathbf{x}}(k)))$



Your answer is correct.

The correct answer is:

$$\sum_{k=1}^M y(k) \ln g(\bar{\mathbf{x}}(k)) + (1 - y(k)) \ln (1 - g(\bar{\mathbf{x}}(k)))$$

Question **10**

Correct

Mark 1.00 out of 1.00

The threshold function $g(\bar{\mathbf{x}})$ for the perceptron learning algorithm is given as

- ☐ -1 for $\bar{\mathbf{h}}^T \bar{\mathbf{x}} \leq 0$ and 0 otherwise
- ☒ 1 for $\bar{\mathbf{h}}^T \bar{\mathbf{x}} \geq 0$ and 0 otherwise
- ☐ $\frac{e^{-\bar{\mathbf{x}}^T \bar{\mathbf{h}}}}{1 + e^{-\bar{\mathbf{x}}^T \bar{\mathbf{h}}}}$
- ☐ $\frac{1}{1 + e^{-\bar{\mathbf{x}}^T \bar{\mathbf{h}}}}$



Your answer is correct.

The correct answer is:

1 for $\bar{\mathbf{h}}^T \bar{\mathbf{x}} \geq 0$ and 0 otherwise