

Quiz_ Quiz 3_ Linear Regression

Friday, September 30, 2022 2:22 AM

Quiz_ Quiz
3_ Linear...

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Quiz: Quiz 3: Linear Regression

Question 1

1 pts

Which one is not a term of degree 2 polynomial features?

☐ $x_1 x_2$

☐ $2x_1 x_2$

☐ x_1^2

☒ $x_1 x_2^2$

Question 2

1 pts

Which of the following statements is not true regarding linear regression?

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☐ It figures out the best fitting line/plane that describes two or more variables.

☐ It quantifies a relationship between two continuous variables.

☐ Random disturbance in the relationship between the independent variables and the dependent variable) is the same across all values of the independent variables.

☒ It can predict the outcome of a binary variable with continuous variables. F

Question 3

1 pts

Let $\mathbf{X} \in \mathbb{R}^{n \times d}$, $\mathbf{y}, \mathbf{w} \in \mathbb{R}^d$.
Let $f(\mathbf{w}) := \frac{1}{2} \|\mathbf{X}\mathbf{w} - \mathbf{y}\|_2^2 + \frac{\lambda}{2} \|\mathbf{w}\|_2^2$

What is $\frac{\partial f}{\partial^2 \mathbf{w}} f(\mathbf{w})$?

☐ $\mathbf{X}^\top \mathbf{X}$

☐ $2\mathbf{X}^\top \mathbf{X} + 2\lambda I$

☒ $\mathbf{X}^\top \mathbf{X} + \lambda I$

☐ $2\mathbf{X}^\top \mathbf{X}$

$$\frac{\partial f}{\partial \mathbf{w}} f(\mathbf{w}) = (\mathbf{X}\mathbf{w} - \mathbf{y})\mathbf{w} + \lambda \mathbf{w}$$
$$\frac{\partial^2 f}{\partial \mathbf{w}^2} f(\mathbf{w}) = \mathbf{X}^\top \mathbf{X} + \lambda I$$

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Question 4

1 pts

When can we say that a model generalizes?

☐ When generalization gap is small

☐ When the training loss is low

☒ When the test loss is low and training loss is low

☐ When the test lost is higher than the train loss

Question 5

1 pts

Suppose \mathbf{X} is invertible and let $\mathbf{U}\mathbf{\Sigma}\mathbf{V}^\top$ be its singular value decomposition. Which of the following is the pseudoinverse of \mathbf{X} ?

☐ $\mathbf{X}^\dagger = \mathbf{V}^\top \mathbf{\Sigma}^{-1} \mathbf{U}^\top$

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☒ $\mathbf{X}^\dagger = \mathbf{V}\mathbf{\Sigma}^{-1}\mathbf{U}^\top$

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☐ $\mathbf{X}^\dagger = \mathbf{V}^\top \mathbf{\Sigma}^{-1} \mathbf{U}$

Question 6

1 pts

Which of the following is not true about overfitting?

☐ Testing loss is much higher than training loss.

☒ Both training and testing loss are high.

☐ The model is fitted too well to the data.

☐ The model is too complex.

Question 7

1 pts

Given a dataset $\mathbf{D} = \{(0, 1), (4, 5), (5, 6)\}$, where the first number in each tuple is the input x_i and the second number is the label y_i . Consider the regression model $\hat{y}_i = b$ (i.e.: the

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☒ 4

☐ 1

☐ 0

☐ 5

predicted label is always the same, but b is a model parameter and can be optimized). Suppose the loss function $L(b) = \sum_{i=1}^N (y_i - \hat{y}_i)^2$. What is the optimal parameter ?

$$\vec{w} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$$

Question 8

1 pts

Given a dataset $\mathbf{D} = (x_1, y_1), (x_2, y_2), \dots, (x_N, y_N)$ and assume both x and y are scalars, we apply a univariate linear regression model $\hat{y}_i = wx_i + b$. Let w and b^* denote the optimal model parameters, b under minimization of the square loss $L(w, b) = \sum_{i=1}^N (y_i - \hat{y}_i)^2$. Which of the followings is false?

☐ The loss function $L(w, b)$ is convex in both w and b T

☐ $b^* = \frac{1}{N} \sum_{i=1}^N (y_i - w^* x_i)$

☒ $w^* = \frac{\sum_{i=1}^N y_i (x_i - \bar{x})}{\sum_{i=1}^N x_i^2 - \bar{x}^2}$ where $\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$

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☐ The optimization problem that should be solved is $L(w, b) = \sum_{i=1}^N (y_i - wx_i - b)^2$ T

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