My Illinois Compass 2q





Welcome Take Test: Quiz 5

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Test Information

Description

Instructions

Multiple Attempts This test allows 3 attempts. This is attempt number 2.

Force Completion This test can be saved and resumed later.

¥ Question Completion Status:

QUESTION 1

2 points Saved

Which of the following criteria/statistics can be used to select the order of polynomial regression? Check all that apply.

- A. AIC
- ☑ B. F-test
- C. RSS
- $\hfill \hfill \hfill$

QUESTION 2

2 points Saved

Suppose g is a cubic spline defined on [a, b]. Which of the following statements are true? Circle all that apply.

- A. g is a continuous function.
- B. The first derivative of g is continuous.
- C. The third derivative of g is continuous.
- D. The second derivative of g is continuous.

QUESTION 3

2 points

Saved

A robot needs to follow a path that passes consecutively through six points (x_i, y_i) where without loss of generality, assume x_i 's are arranged in an increasing order and they are unique. To find a smooth path you would recommend which of the following? Circle all that apply.

- A. Fit a linear regression model based on the 6 data points.
- B. Fit a natural cubic spline function with knots at the six points (x1 to x6).
- C. Fit a cubic polynomial function of x based on the 6 data points.
- D. Fit a cubic spline function with two knots {z1, z2}, where z1 is the average of (x1, x2, x3) and z2 is the average of (x4, x5, x6).

QUESTION 4

4 points

Saved

A cubic spline function g is defined between 0 and 2 as follows

- $g(x) = 1 + 2 x x^3 \text{ if } 0 \le x < 1;$
- $g(x) = a + b(x-1) + c(x-1)^2 + d(x-1)^3$, if $1 \le x \le 2$.

Find the value of a, b, c, and d. If the value is not unique, write "NA" in the box.

- a= 2
- b= -1
- C= -3
- d= NA

Click Save and Submit to save and submit. Click Save All Answers to save all answers.

Save All Answers

Save and

Round your answer to the 2nd digits after the decimal point.	
Use the poly() function to fit a cubic polynomial regression to predict "nox" using "dis".	
What's the residual sum of squares? 1.93 (A number between 1.50 and 2.10)	
• What's the predicted "nox" when dis=6? 0.44 (A number between 0.30 and 0.60)	
• Is the p-value for the cubic term less than 5%? Yes (Fill in "Yes" or "No")	
Next use the poly() function to fit a fourth-degree polynomial regression model.	
 What's the residual sum of squares? 1.93 (A number between 1.50 to 2.10) What's the predicted "nox" when dis=6? 0.44 (A number between 0.30 and 0.60) Is the p-value for the highest polynomial term less than 5%? No (Fill in "Yes" or "No") 	
QUESTION 6	2 points Save
This question uses the variables "dis" and "nox" from the Boston data. We use the following R command to fit a cubic spline model to predict "nox" using "dis" library(MASS) attach(Boston) myfit1 = lm(nox ~ bs(dis, df=3), data=Boston)	
Which of the following R command would return the same model as "myfit1"? Circle all that apply.	
✓ A. Im(nox ~ bs(dis, df= 4, intercept=TRUE), data=Boston)	
B. Im(nox ~ bs(dis, knots=quantile(dis, prob=c(0.25, 0.5, 0.75)), data=Boston)	
C. Im(nox ~ bs(dis, knots=median(dis)), data=Boston)✓ D. Im(nox ~ poly(dis, 3), data=Boston)	
E. Im(nox ~ bs(dis, df= 5, intercept=TRUE), data=Boston)	
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E. Im(nox ~ bs(dis, df= 5, intercept=TRUE), data=Boston) QUESTION 7	2 points Save
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QUESTION 7 Suppose we use the following R command to fit a cubic spline model to predict "nox" using "dis"	2 points Save
Suppose we use the following R command to fit a cubic spline model to predict "nox" using "dis" myfit2 = Im(nox ~ bs(dis, df=4), data=Boston) Which of the following R command would return the same model as "myfit2"? Circle all that apply. A. Im(nox ~ bs(dis, df= 5, intercept=TRUE), data=Boston)	2 points Save
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Save All Answers