

Statistics 151a, Spring 2015 (Linear Modelling - Theory and Applications) Homework Five

Due on April 27, 2015

14 April, 2015

1. Consider the linear model $y_i = \beta_0 + \beta_1 x_{i1} + \cdots + \beta_p x_{ip} + e_i$ for $i = 1, \dots, n$ where $e \sim N(0, \sigma^2 I_n)$.
 - (a) Let \hat{y}_i denote the i th fitted value and let $\hat{y}_{i(i)}$ denote the predicted response value for the i th subject without including the i th subject in the regression. Write the difference $\hat{y}_i - \hat{y}_{i(i)}$ in terms of the i th standardized residual and the i th leverage. **(2 points)**
 - (b) Calculate the distribution of $\hat{y}_i - \hat{y}_{i(i)}$. **(3 points)**
 - (c) Can you obtain an unbiased estimator for σ^2 that is independent of $\hat{y}_i - \hat{y}_{i(i)}$? If yes, specify such an unbiased estimator. If no, explain why. **(3 points)**
2. I fit a linear model to the usual data y_1, \dots, y_n and x_{ij} for $i = 1, \dots, n$ and $j = 1, \dots, p$. Let RSS denote the residual sum of squares and \hat{e} denote the vector of residuals.

I have been told that data on an explanatory variable has not been collected. More specifically, the right model here is apparently

$$y_i = \beta_0 + \beta_1 x_{i1} + \cdots + \beta_p x_{ip} + \gamma z_i + e_i$$

where e_1, \dots, e_n are uncorrelated mean zero errors with constant variance σ^2 . Here z_1, \dots, z_n denote the values of a variable that has not been observed unfortunately.

- (a) Is $RSS/(n - p - 1)$ an unbiased estimator of σ^2 ? If yes, explain with reason. If no, calculate the bias. **(4 points)**
 - (b) Is the sum of the residuals \hat{e}_i zero? Answer with reason. **(2 points)**
 - (c) What is the expected value of \hat{e} ? **(3 points)**
3. Consider the usual data on response y_1, \dots, y_n and explanatory variable data x_{ij} for $i = 1, \dots, n$ and $j = 1, \dots, p$. Suppose y_i can be modelled as a Poisson random variable with mean λ_i . Moreover suppose y_1, \dots, y_n can be assumed to be independent.
 - (a) Write down the form of the canonical GLM. **(2 points)**
 - (b) Write down the log-likelihood as a function of β . **(2 points)**
 4. I got the following linear model output for a dataset consisting of a response variable and three explanatory variables:

Call:

```
lm(formula = y ~ x1 + x2 + x3)
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Residuals:

	Min	1Q	Median	3Q	Max
	-4.4359	-1.3803	0.1258	1.4362	4.9530

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-6.94080	2.07498	-3.345	0.00111 **
x1	3.00762	0.07457	40.332	< 2e-16 ***
x2	-5.71932	0.17992	-31.789	< 2e-16 ***
x3	0.05069	0.17223	0.294	0.76904

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 1.827 on 116 degrees of freedom

Multiple R-squared: 0.9624, Adjusted R-squared: 0.9614

F-statistic: 989.2 on 3 and 116 DF, p-value: < 2.2e-16

The three plots in Figure 1 give the three partial regression or added variable plots for this regression.

- Can you identify which plot corresponds to which variable? Provide reasoning. (4 points).
- Consider the data in the first added variable plot. Suppose I fit a linear model to the y -variable based on the x -variable. What is the value of the RSS for this regression? (2 points)

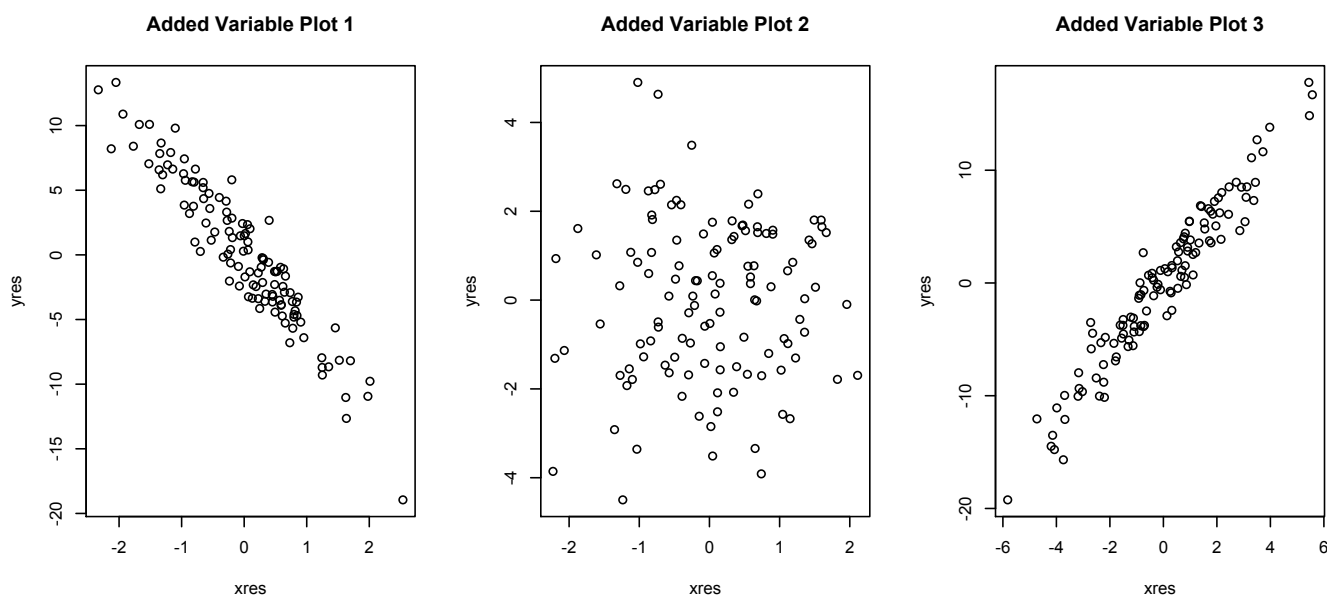


Figure 1: The three partial regression plots for the linear model shown above. One of these three plots corresponds to the first explanatory variable; one corresponds to the second explanatory variable and one to the third explanatory variable

- Determine whether each of following statements is true or false. Provide reasons in each case.
 - The magnitude of a predicted residual is never smaller than the magnitude of the corresponding residual. (1 point)
 - Leverage of the i th subject depends on the value of y_i . (1 point)
 - Model selection via AIC tends to produce smaller models than BIC. (1 point)
 - The GCV is a computationally simpler model selection technique than PRESS. (1 point)

- (e) The optimal model selected by Mallows' C_p criterion can have a C_p value that is more than $p + 1$.
(1 point)