PSTAT 126 - Assignment 4 Fall 2022

Due: Tuesday, October 25 at 11:59 pm on Canvas

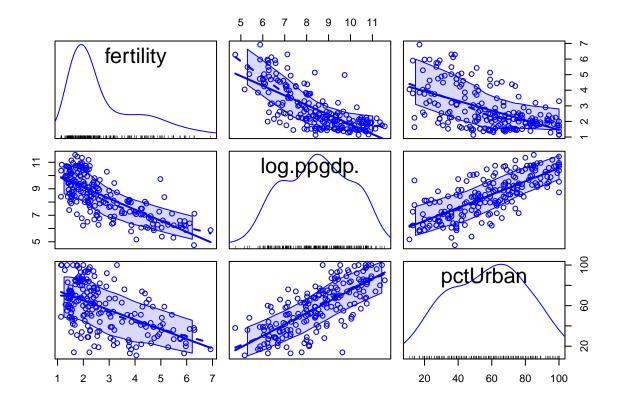
Note: Submit both your Rmd and generated pdf file to Canvas. Use the same indentation level as Solution markers to write your solutions. Improper indentation will break your document.

library(alr4)
library(ggplot2)
data(UN11)

- 1. This problem uses the data set UN11 from the alr4 package.
 - (a) Examine the figure generated by using scatterplotMatrix function for attributes (fertility, log(ppgdp), pctUrban), and comment on the marginal relationships.

 Solution:

scatterplotMatrix(~fertility+log(ppgdp)+pctUrban, data= UN11)



Fertility and log(ppgdp) have a negative correlation, Fertility and pctUrban have a negative correlation, log(ppgdp) and pctUrban have a positive correlation.

(b) Fit the two simple regressions 'fertility' \$\sim\$ log('ppgdp') and 'fertility' \$\sim\$ 'pctUrban', a **Solution**: fit1 <- lm(fertility ~ log(ppgdp), data = UN11)</pre> fit2 <- lm(fertility ~ pctUrban,data = UN11)</pre> summary(fit1) ## ## lm(formula = fertility ~ log(ppgdp), data = UN11) ## ## Residuals: ## Min 1Q Median 3Q Max ## -2.16313 -0.64507 -0.06586 0.62479 3.00517 ## ## Coefficients: Estimate Std. Error t value Pr(>|t|) ## 0.36529 ## (Intercept) 8.00967 21.93 <2e-16 *** 0.04245 -14.61 ## log(ppgdp) -0.62009 <2e-16 *** ## ---## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1 ## Residual standard error: 0.9305 on 197 degrees of freedom ## Multiple R-squared: 0.52, Adjusted R-squared: 0.5175 ## F-statistic: 213.4 on 1 and 197 DF, p-value: < 2.2e-16 summary(fit2) ## ## lm(formula = fertility ~ pctUrban, data = UN11) ## Residuals: Min 1Q Median 30 Max ## -2.4932 -0.7795 -0.1475 0.6517 ## Coefficients: Estimate Std. Error t value Pr(>|t|) 0.213681 21.339 ## (Intercept) 4.559823 <2e-16 *** ## pctUrban -0.031045 0.003421 -9.076 <2e-16 *** ## ---## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1 ## Residual standard error: 1.128 on 197 degrees of freedom

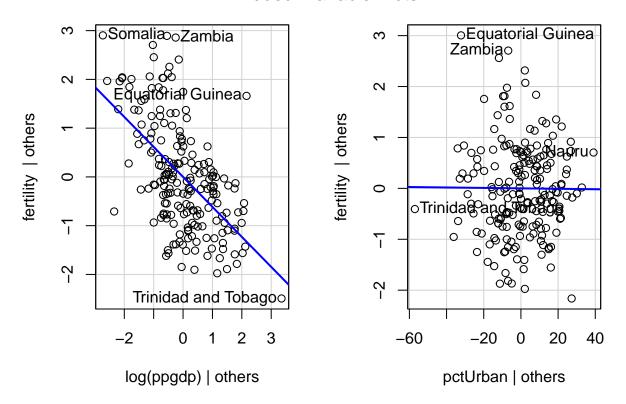
Observing the above summaries, Yes, the slopes are significantly different than zero at any conventional level.

Multiple R-squared: 0.2948, Adjusted R-squared: 0.2913
F-statistic: 82.37 on 1 and 197 DF, p-value: < 2.2e-16</pre>

(c) Obtain the added-variable plots for both predictors. Based on the added-variable plots, does log('p. **Solution**:

```
fit <- lm(fertility ~ log(ppgdp)+pctUrban,data = UN11)
avPlots(fit)</pre>
```

Added-Variable Plots



The av plot for log(ppgdp) after adjusting for pctUrban is useful as it maintains a steep slope, while pctUrban after adjusting for log(ppgdp) has a neutral slope and is not useful.

2. Consider a multiple linear regression model with two continuous predictors:

$$Y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \varepsilon_i, \qquad \varepsilon_i \stackrel{iid}{\sim} \mathcal{N}(0, \sigma^2).$$

(a) Suppose that x_{i1} and x_{i2} are exactly related in that $x_{i1} = 2.2x_{i2}$ for all i. For example, x_{i2} could be weight in kilograms and x_{i1} weight in pounds for the i-th individual. Describe the appearance of the added variable plot for x_{i2} after adjusting for x_{i1} .

Solution:

Since xi2 is a linear function of xi1 and are exactly related, the residuals are zero and therefor the avPlot will look like a vertical line.

b) Suppose that \$x_{i1}\$ and \$x_{i2}\$ are not perfectly correlated, but that \$Y_i = 3x_{i1}\$, i.e. \$Y_ **Solution**: Since xi1 and xi2 are not perfectly correlated, the avPlot for xi2 would look like a straight line (no-slope or little-to-no-slope)

c) (**Bonus**): Simulate some data for each of the situations in parts a) and b) and create an added-v **Solution**: