Question_2

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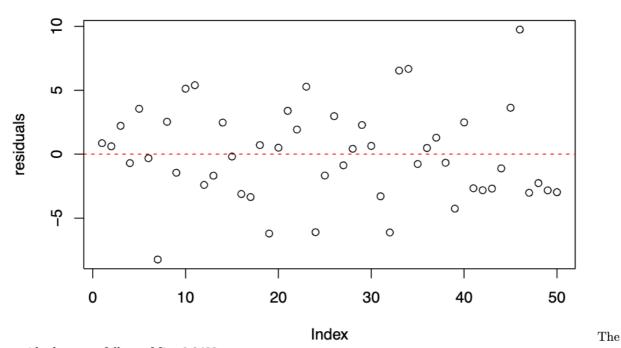
Question 2

First we will fit the Linear regression model.

```
library(faraway)
library(MASS)
# Load the dataset
data(savings)
# View the structure of the dataset
str(savings)
## 'data.frame':
                50 obs. of 5 variables:
## $ sr : num 11.43 12.07 13.17 5.75 12.88 ...
## $ pop15: num 29.4 23.3 23.8 41.9 42.2 ...
## $ pop75: num 2.87 4.41 4.43 1.67 0.83 2.85 1.34 0.67 1.06 1.14 ...
## $ dpi : num 2330 1508 2108 189 728 ...
## $ ddpi : num 2.87 3.93 3.82 0.22 4.56 2.43 2.67 6.51 3.08 2.8 ...
# (1) Fit a linear regression model
model <- lm(sr ~ pop15 + pop75 + dpi + ddpi, data = savings)</pre>
model
##
## Call:
## lm(formula = sr ~ pop15 + pop75 + dpi + ddpi, data = savings)
## Coefficients:
## (Intercept)
                                                            ddpi
                     pop15
                                  pop75
                                                 dpi
## 28.5660865
               -0.4611931 -1.6914977
                                         -0.0003369
                                                       0.4096949
summary(model)
##
## Call:
## lm(formula = sr ~ pop15 + pop75 + dpi + ddpi, data = savings)
## Residuals:
              1Q Median
                             3Q Max
##
      Min
```

```
## -8.2422 -2.6857 -0.2488 2.4280 9.7509
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 28.5660865 7.3545161
                                      3.884 0.000334 ***
                                     -3.189 0.002603 **
## pop15
               -0.4611931 0.1446422
## pop75
               -1.6914977 1.0835989
                                     -1.561 0.125530
## dpi
               -0.0003369 0.0009311
                                     -0.362 0.719173
## ddpi
               0.4096949 0.1961971
                                      2.088 0.042471 *
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 3.803 on 45 degrees of freedom
## Multiple R-squared: 0.3385, Adjusted R-squared: 0.2797
## F-statistic: 5.756 on 4 and 45 DF, p-value: 0.0007904
#Examining the residuals
residuals <- residuals(model)</pre>
plot(residuals)
abline(h = 0, col = "red", lty = 2)
title("Residual Plot")
```

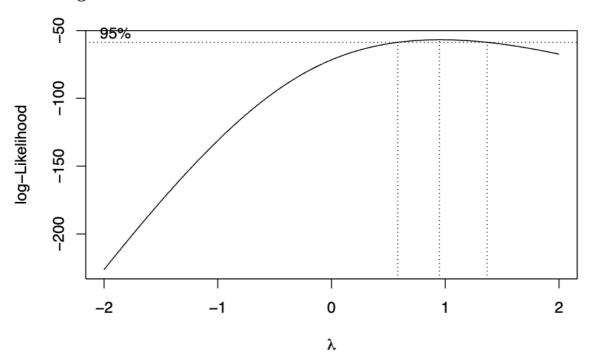
Residual Plot



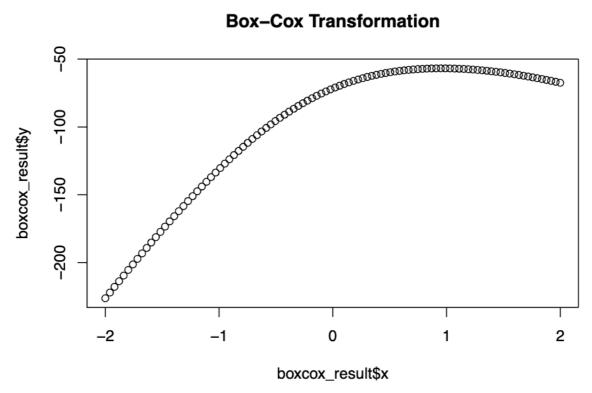
residuals are as follows: Min -8.2422

1Q -2.6857 Median -0.2488 3Q 2.4280 Max 9.7509

Including Plots



Optimal lambda: 0.9494949



The optimal lambda comes out to be: 0.9494949

##Part(c): The main purpose of the Box-Cox plot is to help us identify the value of lambda that maximizes the log-likelihood. The optimal lambda corresponds to the peak of the plot, and it indicates the best power transformation for stabilizing the variance and improving the fit of the linear regression model. The plot provides a visual representation of how the log-likelihood changes across different lambda values. Examining the shape of the curve can give you insights into the nature of the transformation and its impact on the model's goodness of fit.