

STA 108 Final Project: An analysis of crime prevalence using 5 socioeconomic variables for 4 geographic regions of the United States.

Import and format data:

```
library(MASS)
data_demo <- read.table("Demographic.txt")
names(data_demo) = c('id','county','state','land_area','total_population','percent_population_18to34','')
```

Question 1:

Generate Regression Models for each geographic region:

```
model_demo <- lm(crime ~ per_cap_income + unemployment + below_poverty + bachelor + highschool, data=data_demo)

regional_model_generator = function(data_demo,region){
  loader <- data_demo[data_demo$geo_region == region,]
  y <- as.numeric(loader$crime)
  x1 <- as.numeric(loader$per_cap_income)
  x2 <- as.numeric(loader$unemployment)
  x3 <- as.numeric(loader$below_poverty)
  x4 <- as.numeric(loader$bachelor)
  x5 <- as.numeric(loader$highschool)
  output = lm(y ~ x1+x2+x3+x4+x5, data=data_demo)
  return(output)
}

loader1 = regional_model_generator(data_demo= data_demo,region =1)
loader2 = regional_model_generator(data_demo= data_demo,region =2)
loader3 = regional_model_generator(data_demo= data_demo,region =3)
loader4 = regional_model_generator(data_demo= data_demo,region =4)

loader1_data = subset(data_demo, geo_region ==1)
loader2_data = subset(data_demo, geo_region ==2)
loader3_data = subset(data_demo, geo_region ==3)
loader4_data = subset(data_demo, geo_region ==4)
```

Exploratory Data Analysis:

The initial regression model for the demographic dataset has the coefficients $Y = -28798.161 + 4.729 - 815.696 + 3982.371 + 254.546 - 858.541$. These negative and positive values for each coefficient responds to the real world relationship between these variables, with the 4.729 coefficient for per capita income indicating a positive relationship between serious crimes and amount of per capita income. Likewise, the negative coefficients for unemployment and prevalence of high school degrees indicate a negative relationship between serious crimes and these factors. A negative relationship in this situation indicates that a increase in the unemployment rate and decrease in the prevalence of high school degrees would lead to an increase in serious crimes.

The residual plots for each region all display fairly similar results, with the median centered around 0 and all are quite right skewed due to a few extreme outliers.

```
summary(loader1)
```

```
##
## Call:
## lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data = data_demo)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -127848  -16268   -3728   13978  479383
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  47040.612 153904.343   0.306   0.761
## x1              3.670     2.336   1.571   0.119
## x2            -2934.758   4115.950  -0.713   0.478
## x3            13356.157   2742.171   4.871 4.32e-06 ***
## x4              3101.042   1887.071   1.643   0.104
## x5            -3000.954   1862.062  -1.612   0.110
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 56930 on 97 degrees of freedom
## Multiple R-squared:  0.3355, Adjusted R-squared:  0.3012
## F-statistic: 9.794 on 5 and 97 DF,  p-value: 1.341e-07
```

```
summary(loader2)
```

```
##
## Call:
## lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data = data_demo)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -67551 -14420   -4908   7774 347455
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  55890.278 146968.941   0.380 0.704524
## x1              8.493     2.236   3.798 0.000248 ***
## x2            -1110.576   2736.906  -0.406 0.685756
## x3            5350.915   2043.762   2.618 0.010186 *
## x4            1229.761   1320.021   0.932 0.353732
## x5            -3138.775   1742.387  -1.801 0.074591 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 40770 on 102 degrees of freedom
## Multiple R-squared:  0.2934, Adjusted R-squared:  0.2588
## F-statistic: 8.472 on 5 and 102 DF,  p-value: 9.81e-07
```

```
summary(loader3)
```

```
##
```

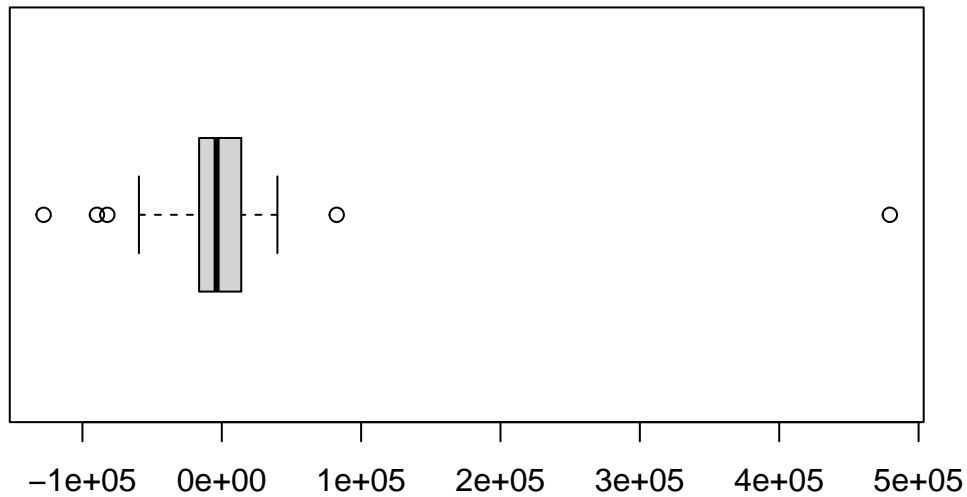
```
## Call:
## lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data = data_demo)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -61813 -14846  -7166   2314 215373
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -77889.315  63697.010  -1.223  0.22337
## x1              4.047     1.352   2.993  0.00324 **
## x2             1721.363   2080.442   0.827  0.40936
## x3             1983.880    924.544   2.146  0.03354 *
## x4              -7.352    755.049  -0.010  0.99224
## x5              35.081    763.520   0.046  0.96342
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 36800 on 146 degrees of freedom
## Multiple R-squared:  0.1132, Adjusted R-squared:  0.08287
## F-statistic: 3.729 on 5 and 146 DF,  p-value: 0.003313
```

```
summary(loader4)
```

```
##
## Call:
## lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data = data_demo)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -91456 -35694  -6947  12592 580752
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 735587.395 366561.289   2.007  0.0486 *
## x1              1.170     4.464   0.262  0.7939
## x2            -9009.014   4879.278  -1.846  0.0690 .
## x3            -4101.237   5517.308  -0.743  0.4597
## x4             3144.073   2918.537   1.077  0.2850
## x5            -8602.818   3722.965  -2.311  0.0238 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 81060 on 71 degrees of freedom
## Multiple R-squared:  0.128, Adjusted R-squared:  0.06661
## F-statistic: 2.085 on 5 and 71 DF,  p-value: 0.07732
```

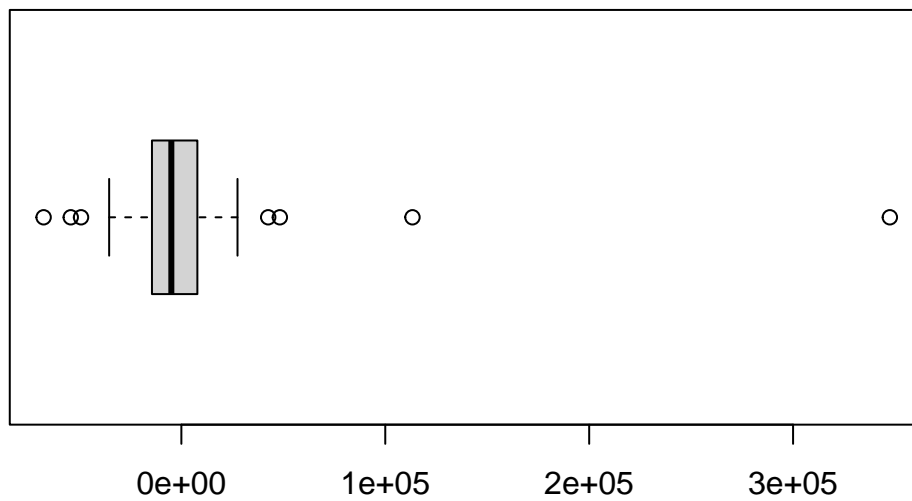
```
boxplot(loader1$residuals, horizontal = T, main = "Residuals for Reigon 1")
```

Residuals for Reigon 1



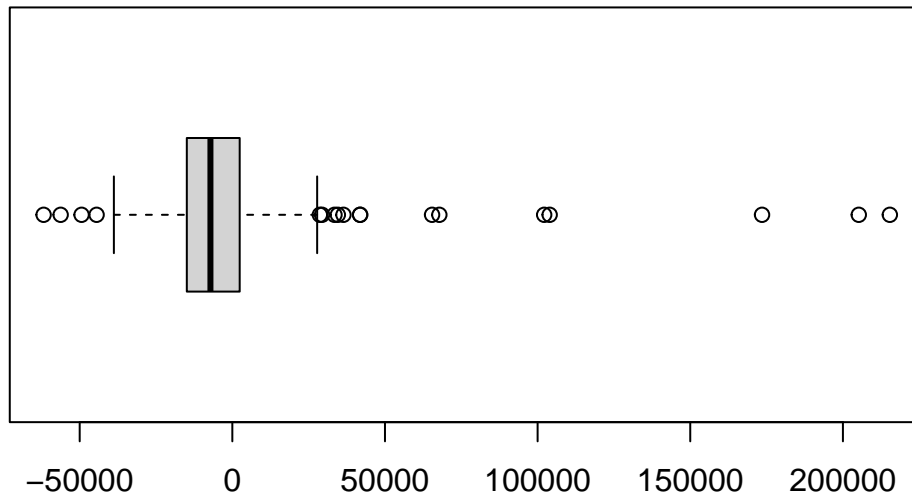
```
boxplot(loader2$residuals, horizontal = T, main = "Residuals for Reigon 2")
```

Residuals for Reigon 2



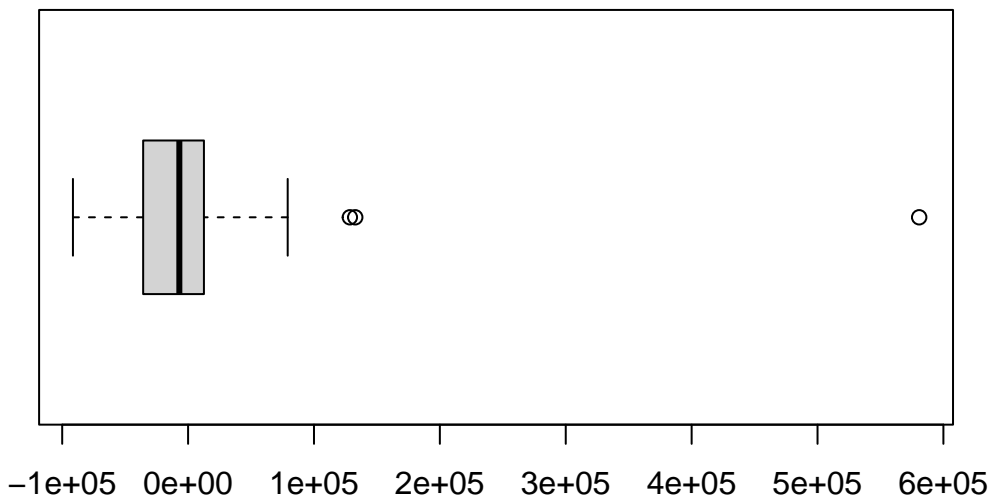
```
boxplot(loader3$residuals, horizontal = T, main = "Residuals for Reigon 3")
```

Residuals for Reigon 3



```
boxplot(loader4$residuals, horizontal = T, main = "Residuals for Reigon 4")
```

Residuals for Reigon 4



Question 2:

Functions to re-initialize variables and construct AIC and BIC models:

```
AICfunction <- function(data_demo, region){
  loader <- data_demo[data_demo$geo_region == region,]
  library(MASS)
  y <- as.numeric(loader$crime)
  x1 <- as.numeric(loader$per_cap_income)
  x2 <- as.numeric(loader$unemployment)
  x3 <- as.numeric(loader$below_poverty)
  x4 <- as.numeric(loader$bachelor)
  x5 <- as.numeric(loader$highschool)
  loader_mod <- lm(y~x1+x2+x3+x4+x5, data = loader)
```

```

AIC_loader <- stepAIC(loader_mod, k =2)
return(AIC_loader)
}

BICfunction <- function(data_demo, region){
  loader <- data_demo[data_demo$geo_region == region,]
  library(MASS)
  y <- as.numeric(loader$crime)
  x1 <- as.numeric(loader$per_cap_income)
  x2 <- as.numeric(loader$unemployment)
  x3 <- as.numeric(loader$below_poverty)
  x4 <- as.numeric(loader$bachelor)
  x5 <- as.numeric(loader$highschool)
  loader_mod <- lm(y~x1+x2+x3+x4+x5, data = loader)
  BIC_loader <- stepAIC(loader_mod, k =log(nrow(loader)))
  return(BIC_loader)
}

```

Construct AIC and BIC for each model per geographic region:

```
AIC_loader1 = AICfunction(data_demo, 1)
```

```

## Start:  AIC=2261.42
## y ~ x1 + x2 + x3 + x4 + x5
##
##           Df Sum of Sq      RSS   AIC
## - x2      1 1.6475e+09 3.1598e+11 2259.9
## <none>                3.1433e+11 2261.4
## - x1      1 7.9952e+09 3.2233e+11 2262.0
## - x5      1 8.4169e+09 3.2275e+11 2262.1
## - x4      1 8.7510e+09 3.2309e+11 2262.2
## - x3      1 7.6877e+10 3.9121e+11 2281.9
##
## Step:  AIC=2259.95
## y ~ x1 + x3 + x4 + x5
##
##           Df Sum of Sq      RSS   AIC
## <none>                3.1598e+11 2259.9
## - x1      1 7.8106e+09 3.2379e+11 2260.5
## - x5      1 8.3993e+09 3.2438e+11 2260.7
## - x4      1 1.0755e+10 3.2674e+11 2261.4
## - x3      1 7.6837e+10 3.9282e+11 2280.4

```

```
AIC_loader1$coefficients
```

```

## (Intercept)          x1          x3          x4          x5
## 24461.535401    3.625788 12798.923610 3369.023096 -2997.812291

```

```
AIC_loader2 = AICfunction(data_demo, 2)
```

```

## Start:  AIC=2298.8
## y ~ x1 + x2 + x3 + x4 + x5
##
##           Df Sum of Sq      RSS   AIC
## - x2      1 2.7364e+08 1.6979e+11 2297.0

```

```

## - x4      1 1.4424e+09 1.7095e+11 2297.7
## <none>                1.6951e+11 2298.8
## - x5      1 5.3930e+09 1.7491e+11 2300.2
## - x3      1 1.1392e+10 1.8090e+11 2303.8
## - x1      1 2.3976e+10 1.9349e+11 2311.1
##
## Step: AIC=2296.97
## y ~ x1 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x4      1 2.7082e+09 1.7249e+11 2296.7
## <none>                1.6979e+11 2297.0
## - x5      1 5.4201e+09 1.7521e+11 2298.4
## - x3      1 1.1280e+10 1.8107e+11 2301.9
## - x1      1 2.5189e+10 1.9498e+11 2309.9
##
## Step: AIC=2296.68
## y ~ x1 + x3 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x5      1 2.8323e+09 1.7533e+11 2296.4
## <none>                1.7249e+11 2296.7
## - x3      1 2.3323e+10 1.9582e+11 2308.4
## - x1      1 4.9408e+10 2.2190e+11 2321.9
##
## Step: AIC=2296.44
## y ~ x1 + x3
##
##      Df Sum of Sq      RSS      AIC
## <none>                1.7533e+11 2296.4
## - x1      1 4.6810e+10 2.2214e+11 2320.0
## - x3      1 5.1372e+10 2.2670e+11 2322.2

```

```
AIC_loader2$coefficients
```

```

##      (Intercept)          x1          x3
## -2.024345e+05  8.932537e+00  7.685012e+03

```

```
AIC_loader3 = AICfunction(data_demo, 3)
```

```

## Start: AIC=3201.9
## y ~ x1 + x2 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x4      1 1.2838e+05 1.9771e+11 3199.9
## - x5      1 2.8587e+06 1.9771e+11 3199.9
## - x2      1 9.2706e+08 1.9864e+11 3200.6
## <none>                1.9771e+11 3201.9
## - x3      1 6.2352e+09 2.0395e+11 3204.6
## - x1      1 1.2133e+10 2.0984e+11 3209.0
##
## Step: AIC=3199.9
## y ~ x1 + x2 + x3 + x5
##
##      Df Sum of Sq      RSS      AIC

```

```
## - x5      1 3.4997e+06 1.9771e+11 3197.9
## - x2      1 1.0447e+09 1.9876e+11 3198.7
## <none>                1.9771e+11 3199.9
## - x3      1 9.1501e+09 2.0686e+11 3204.8
## - x1      1 2.0546e+10 2.1826e+11 3212.9
##
## Step: AIC=3197.9
## y ~ x1 + x2 + x3
##
##      Df Sum of Sq      RSS      AIC
## - x2      1 1.1429e+09 1.9886e+11 3196.8
## <none>                1.9771e+11 3197.9
## - x3      1 1.0401e+10 2.0811e+11 3203.7
## - x1      1 2.3238e+10 2.2095e+11 3212.8
##
## Step: AIC=3196.78
## y ~ x1 + x3
##
##      Df Sum of Sq      RSS      AIC
## <none>                1.9886e+11 3196.8
## - x3      1 1.5230e+10 2.1409e+11 3206.0
## - x1      1 2.2194e+10 2.2105e+11 3210.9
```

```
AIC_loader3$coefficients
```

```
## (Intercept)          x1          x3
## -64787.42054      3.91437  2206.90440
```

```
AIC_loader4 = AICfunction(data_demo, 4)
```

```
## Start: AIC=1746.42
## y ~ x1 + x2 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x1      1 4.5169e+08 4.6703e+11 1744.5
## - x3      1 3.6311e+09 4.7020e+11 1745.0
## - x4      1 7.6263e+09 4.7420e+11 1745.7
## <none>                4.6657e+11 1746.4
## - x2      1 2.2403e+10 4.8898e+11 1748.0
## - x5      1 3.5089e+10 5.0166e+11 1750.0
##
## Step: AIC=1744.49
## y ~ x2 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x3      1 9.8347e+09 4.7686e+11 1744.1
## <none>                4.6703e+11 1744.5
## - x2      1 2.3284e+10 4.9031e+11 1746.2
## - x4      1 2.6717e+10 4.9374e+11 1746.8
## - x5      1 6.0540e+10 5.2757e+11 1751.9
##
## Step: AIC=1744.09
## y ~ x2 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
```



```
## <none>          4.7686e+11 1744.1
## - x2    1 2.3032e+10 4.9989e+11 1745.7
## - x4    1 2.8423e+10 5.0528e+11 1746.5
## - x5    1 5.3254e+10 5.3011e+11 1750.2
```

```
AIC_loader4$coefficients
```

```
## (Intercept)          x2          x4          x5
## 581617.594   -9087.879    3847.990   -7060.223
```

```
BIC_loader1 = BICfunction(data_demo, 1)
```

```
## Start:  AIC=2277.22
## y ~ x1 + x2 + x3 + x4 + x5
##
##          Df  Sum of Sq      RSS    AIC
## - x2     1 1.6475e+09 3.1598e+11 2273.1
## - x1     1 7.9952e+09 3.2233e+11 2275.2
## - x5     1 8.4169e+09 3.2275e+11 2275.3
## - x4     1 8.7510e+09 3.2309e+11 2275.4
## <none>          3.1433e+11 2277.2
## - x3     1 7.6877e+10 3.9121e+11 2295.1
##
## Step:  AIC=2273.13
## y ~ x1 + x3 + x4 + x5
##
##          Df  Sum of Sq      RSS    AIC
## - x1     1 7.8106e+09 3.2379e+11 2271.0
## - x5     1 8.3993e+09 3.2438e+11 2271.2
## - x4     1 1.0755e+10 3.2674e+11 2271.9
## <none>          3.1598e+11 2273.1
## - x3     1 7.6837e+10 3.9282e+11 2290.9
##
## Step:  AIC=2271.01
## y ~ x3 + x4 + x5
##
##          Df  Sum of Sq      RSS    AIC
## <none>          3.2379e+11 2271.0
## - x5     1 1.8073e+10 3.4187e+11 2272.0
## - x4     1 6.1535e+10 3.8533e+11 2284.3
## - x3     1 6.9756e+10 3.9355e+11 2286.5
```

```
BIC_loader1$coefficients
```

```
## (Intercept)          x3          x4          x5
## 147968.881   11134.647    5475.472   -4078.540
```

```
BIC_loader2 = BICfunction(data_demo, 2)
```

```
## Start:  AIC=2314.89
## y ~ x1 + x2 + x3 + x4 + x5
##
##          Df  Sum of Sq      RSS    AIC
## - x2     1 2.7364e+08 1.6979e+11 2310.4
## - x4     1 1.4424e+09 1.7095e+11 2311.1
## - x5     1 5.3930e+09 1.7491e+11 2313.6
## <none>          1.6951e+11 2314.9
```

```

## - x3      1 1.1392e+10 1.8090e+11 2317.2
## - x1      1 2.3976e+10 1.9349e+11 2324.5
##
## Step: AIC=2310.38
## y ~ x1 + x3 + x4 + x5
##
##          Df Sum of Sq      RSS      AIC
## - x4      1 2.7082e+09 1.7249e+11 2307.4
## - x5      1 5.4201e+09 1.7521e+11 2309.1
## <none>                                1.6979e+11 2310.4
## - x3      1 1.1280e+10 1.8107e+11 2312.7
## - x1      1 2.5189e+10 1.9498e+11 2320.6
##
## Step: AIC=2307.41
## y ~ x1 + x3 + x5
##
##          Df Sum of Sq      RSS      AIC
## - x5      1 2.8323e+09 1.7533e+11 2304.5
## <none>                                1.7249e+11 2307.4
## - x3      1 2.3323e+10 1.9582e+11 2316.4
## - x1      1 4.9408e+10 2.2190e+11 2329.9
##
## Step: AIC=2304.49
## y ~ x1 + x3
##
##          Df Sum of Sq      RSS      AIC
## <none>                                1.7533e+11 2304.5
## - x1      1 4.6810e+10 2.2214e+11 2325.4
## - x3      1 5.1372e+10 2.2670e+11 2327.6

```

```
BIC_loader2$coefficients
```

```

##      (Intercept)          x1          x3
## -2.024345e+05  8.932537e+00  7.685012e+03

```

```
BIC_loader3 = BICfunction(data_demo, 3)
```

```

## Start: AIC=3220.04
## y ~ x1 + x2 + x3 + x4 + x5
##
##          Df Sum of Sq      RSS      AIC
## - x4      1 1.2838e+05 1.9771e+11 3215.0
## - x5      1 2.8587e+06 1.9771e+11 3215.0
## - x2      1 9.2706e+08 1.9864e+11 3215.7
## - x3      1 6.2352e+09 2.0395e+11 3219.7
## <none>                                1.9771e+11 3220.0
## - x1      1 1.2133e+10 2.0984e+11 3224.1
##
## Step: AIC=3215.02
## y ~ x1 + x2 + x3 + x5
##
##          Df Sum of Sq      RSS      AIC
## - x5      1 3.4997e+06 1.9771e+11 3210.0
## - x2      1 1.0447e+09 1.9876e+11 3210.8
## <none>                                1.9771e+11 3215.0

```

```
## - x3      1 9.1501e+09 2.0686e+11 3216.9
## - x1      1 2.0546e+10 2.1826e+11 3225.0
##
## Step: AIC=3210
## y ~ x1 + x2 + x3
##
##      Df Sum of Sq      RSS      AIC
## - x2      1 1.1429e+09 1.9886e+11 3205.9
## <none>                1.9771e+11 3210.0
## - x3      1 1.0401e+10 2.0811e+11 3212.8
## - x1      1 2.3238e+10 2.2095e+11 3221.9
##
## Step: AIC=3205.85
## y ~ x1 + x3
##
##      Df Sum of Sq      RSS      AIC
## <none>                1.9886e+11 3205.9
## - x3      1 1.5230e+10 2.1409e+11 3212.0
## - x1      1 2.2194e+10 2.2105e+11 3216.9
```

```
BIC_loader3$coefficients
```

```
## (Intercept)          x1          x3
## -64787.42054      3.91437    2206.90440
```

```
BIC_loader4 = BICfunction(data_demo, 4)
```

```
## Start: AIC=1760.48
## y ~ x1 + x2 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x1      1 4.5169e+08 4.6703e+11 1756.2
## - x3      1 3.6311e+09 4.7020e+11 1756.7
## - x4      1 7.6263e+09 4.7420e+11 1757.4
## - x2      1 2.2403e+10 4.8898e+11 1759.8
## <none>                4.6657e+11 1760.5
## - x5      1 3.5089e+10 5.0166e+11 1761.7
##
## Step: AIC=1756.21
## y ~ x2 + x3 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x3      1 9.8347e+09 4.7686e+11 1753.5
## - x2      1 2.3284e+10 4.9031e+11 1755.6
## - x4      1 2.6717e+10 4.9374e+11 1756.2
## <none>                4.6703e+11 1756.2
## - x5      1 6.0540e+10 5.2757e+11 1761.2
##
## Step: AIC=1753.47
## y ~ x2 + x4 + x5
##
##      Df Sum of Sq      RSS      AIC
## - x2      1 2.3032e+10 4.9989e+11 1752.8
## <none>                4.7686e+11 1753.5
## - x4      1 2.8423e+10 5.0528e+11 1753.6
```

```
## - x5      1 5.3254e+10 5.3011e+11 1757.3
##
## Step: AIC=1752.76
## y ~ x4 + x5
##
##          Df Sum of Sq      RSS      AIC
## <none>                4.9989e+11 1752.8
## - x4      1 2.9733e+10 5.2963e+11 1752.9
## - x5      1 3.0760e+10 5.3065e+11 1753.0
```

```
BIC_loader4$coefficients
```

```
## (Intercept)          x4          x5
## 247217.155      3934.480     -3686.151
```

Question 3:

Estimate a 90% confidence interval for parameters B_j , $j=1, \dots, p$

90% Confidence Intervals for each region listed below:

```
confint(AIC_loader1, level = 0.9)
```

```
##                5 %                95 %
## (Intercept) -2.250082e+05 2.739312e+05
## x1          -2.425978e-01 7.494175e+00
## x3           8.445211e+03 1.715264e+04
## x4           3.059355e+02 6.432111e+03
## x5          -6.082087e+03 8.646288e+01
```

```
x1 -2.425978e-01 7.494175e+00 x3 8.445211e+03 1.715264e+04 x4 3.059355e+02 6.432111e+03 x5
-6.082087e+03 8.646288e+01
```

```
confint(AIC_loader2, level = 0.9)
```

```
##                5 %                95 %
## (Intercept) -2.654457e+05 -139423.35914
## x1           6.132835e+00      11.73224
## x3           5.385754e+03     9984.27016
```

```
x1 6.132835e+00 11.73224 x3 5.385754e+03 9984.27016
```

```
confint(AIC_loader3, level = 0.9)
```

```
##                5 %                95 %
## (Intercept) -1.008800e+05 -28694.890532
## x1           2.325627e+00      5.503114
## x3           1.125616e+03     3288.192401
```

```
x1 2.325627e+00 5.503114 x3 1.125616e+03 3288.192401
```

```
confint(AIC_loader4, level = 0.9)
```

```
##                5 %                95 %
## (Intercept) 232751.4801 930483.708
## x2          -17151.0469 -1024.712
## x4           774.6599    6921.321
## x5          -11179.7778 -2940.668
```

```
x2 -17151.0469 -1024.712 x4 774.6599 6921.321 x5 -11179.7778 -2940.668
```

```
confint(BIC_loader1, level = 0.9)
```

```
##              5 %          95 %
## (Intercept) -65355.845 361293.608
## x3           7131.412 15137.883
## x4           3379.490 7571.454
## x5          -6959.379 -1197.701
x3 7131.412 15137.883 x4 3379.490 7571.454 x5 -6959.379 -1197.701
```

```
confint(BIC_loader2, level = 0.9)
```

```
##              5 %          95 %
## (Intercept) -2.654457e+05 -139423.35914
## x1           6.132835e+00 11.73224
## x3           5.385754e+03 9984.27016
x1 6.132835e+00 11.73224 x3 5.385754e+03 9984.27016
```

```
confint(BIC_loader3, level = 0.9)
```

```
##              5 %          95 %
## (Intercept) -1.008800e+05 -28694.890532
## x1           2.325627e+00 5.503114
## x3           1.125616e+03 3288.192401
x1 2.325627e+00 5.503114 x3 1.125616e+03 3288.192401
```

```
confint(BIC_loader4, level = 0.9)
```

```
##              5 %          95 %
## (Intercept) 60623.396 433810.9147
## x4           810.656 7058.3043
## x5          -6563.551 -808.7508
x4 810.656 7058.3043 x5 -6563.551 -808.7508
```

Using $\alpha = 0.01$, compute the p-value for the two alternatives in the formula:

The P-Value computations and comparisons are listed below:

```
summary(AIC_loader1)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -125137  -16763   -1610   13404  481036
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  24461.535 150233.059   0.163  0.8710
## x1              3.626     2.330   1.556  0.1228
## x3            12798.924   2621.848   4.882 4.09e-06 ***
## x4             3369.023   1844.621   1.826  0.0708 .
## x5            -2997.812   1857.380  -1.614  0.1097
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 56780 on 98 degrees of freedom
## Multiple R-squared:  0.332, Adjusted R-squared:  0.3047
## F-statistic: 12.18 on 4 and 98 DF,  p-value: 4.481e-08
x1 P-Value: 0.1228 > 0.01 x3 P-Value: 4.09e-0 < 0.01 x4 P-Value: 0.0708 > 0.01 x5 P-Value: 0.1097 > 0.01
From this test, We can conclude that we can reject H0 for x1, x4, and x5, but fail to reject H0 for x3
summary(AIC_loader2)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -53569 -14118  -3370   6912 359972
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.024e+05  3.797e+04  -5.331 5.62e-07 ***
## x1           8.933e+00  1.687e+00   5.295 6.59e-07 ***
## x3           7.685e+03  1.386e+03   5.547 2.19e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 40860 on 105 degrees of freedom
## Multiple R-squared:  0.2692, Adjusted R-squared:  0.2553
## F-statistic: 19.34 on 2 and 105 DF,  p-value: 7.076e-08
x1 P-Value: 6.59e-07 < 0.01 x3 P-Value: 2.19e-07 < 0.01
```

From this test we can conclude that we fail to reject H0 for both x1 and x3

```
summary(AIC_loader3)

##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -61130 -15702  -7705   2106 214330
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.479e+04  2.181e+04  -2.971 0.003461 **
## x1           3.914e+00  9.599e-01   4.078 7.37e-05 ***
## x3           2.207e+03  6.533e+02   3.378 0.000931 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 36530 on 149 degrees of freedom
## Multiple R-squared:  0.1081, Adjusted R-squared:  0.09612
## F-statistic: 9.029 on 2 and 149 DF,  p-value: 0.0001989
```

x1 P-Value: $7.37\text{e-}05 < 0.01$ x3 P-Value: $0.000931 < 0.01$

From this test, we conclude that we fail to reject H_0 for either x1 or x3

```
summary(AIC_loader4)
```

```
##
## Call:
## lm(formula = y ~ x2 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -92749 -32731 -10488   9860 588427
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   581618     209404   2.777  0.00696 **
## x2             -9088        4840  -1.878  0.06441 .
## x4              3848        1845   2.086  0.04048 *
## x5            -7060        2473  -2.855  0.00560 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 80820 on 73 degrees of freedom
## Multiple R-squared:  0.1088, Adjusted R-squared:  0.07217
## F-statistic:  2.97 on 3 and 73 DF,  p-value: 0.03733
```

x2 P-Value: $0.06441 > 0.01$ x4 P-Value: $0.04048 > 0.01$ x5 P-Value: $0.00560 < 0.01$

From this test we can conclude that we reject H_0 for x2 and x4, but fail to reject H_0 for x5

```
summary(BIC_loader1)
```

```
##
## Call:
## lm(formula = y ~ x3 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -114724 -17914  -1981   11517 484782
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   147969     128479   1.152  0.2522
## x3             11135        2411   4.618 1.17e-05 ***
## x4              5476        1262   4.338 3.47e-05 ***
## x5            -4078        1735  -2.351  0.0207 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 57190 on 99 degrees of freedom
## Multiple R-squared:  0.3155, Adjusted R-squared:  0.2947
## F-statistic: 15.21 on 3 and 99 DF,  p-value: 3.258e-08
```

x3 P-Value: $1.17\text{e-}05 < 0.01$ x4 P-Value: $3.47\text{e-}05 < 0.01$ x5 P-Value: $0.0207 > 0.01$

From this test we conclude that we reject H_0 for x3 and x4, but fail to reject H_0 for x5

```
summary(BIC_loader2)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -53569 -14118  -3370    6912 359972
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.024e+05  3.797e+04  -5.331 5.62e-07 ***
## x1           8.933e+00  1.687e+00   5.295 6.59e-07 ***
## x3           7.685e+03  1.386e+03   5.547 2.19e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 40860 on 105 degrees of freedom
## Multiple R-squared:  0.2692, Adjusted R-squared:  0.2553
## F-statistic: 19.34 on 2 and 105 DF,  p-value: 7.076e-08
x1 P-Value: 6.59e-07 < 0.01 x3 P-Value: 2.19e-07 < 0.01
```

From this test we conclude that we fail to reject H_0 for either parameter

```
summary(BIC_loader3)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -61130 -15702  -7705    2106 214330
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.479e+04  2.181e+04  -2.971 0.003461 **
## x1           3.914e+00  9.599e-01   4.078 7.37e-05 ***
## x3           2.207e+03  6.533e+02   3.378 0.000931 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 36530 on 149 degrees of freedom
## Multiple R-squared:  0.1081, Adjusted R-squared:  0.09612
## F-statistic: 9.029 on 2 and 149 DF,  p-value: 0.0001989
x1 P-Value: 7.37e-05 < 0.01 x3 P-Value: 0.000931 < 0.01
```

From this test we conclude that we fail to reject H_0 for either parameter

```
summary(BIC_loader4)
```

```
##
## Call:
## lm(formula = y ~ x4 + x5, data = loader)
```



```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -81236 -28955 -14033   3792 612011
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  247217      112021   2.207  0.0304 *
## x4           3934        1875   2.098  0.0393 *
## x5          -3686        1727  -2.134  0.0362 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 82190 on 74 degrees of freedom
## Multiple R-squared:  0.06575,    Adjusted R-squared:  0.0405
## F-statistic: 2.604 on 2 and 74 DF,  p-value: 0.08076
x4 P-Value: 0.0393 > 0.01 x5 P-Value: 0.0362 > 0.01
```

From this test we conclude that we reject H_0 for both x_4 and x_5

Test whether or not $B_1 = B_2 = \dots = B_{p-1} = 0$ with $\alpha = 0.05$. State the decision rule and conclusion. Are these measures similar for the four regions?

$H_0: B_1 = B_2 = \dots = B_{p-1} = 0$ $H_a: B_1 \neq B_2 \neq \dots = B_{p-1} \neq 0$

We fail to reject H_0 with the majority of the tests, with the exceptions of x_1 , x_4 , x_5 of Region 1, and x_2 of Region 4. For these parameters, we fail to reject H_0 and conclude that these parameters are not the same and do not equal 0.

```
summary(AIC_loader1)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -125137 -16763   -1610   13404  481036
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  24461.535 150233.059   0.163  0.8710
## x1           3.626      2.330   1.556  0.1228
## x3          12798.924  2621.848   4.882 4.09e-06 ***
## x4           3369.023  1844.621   1.826  0.0708 .
## x5          -2997.812  1857.380  -1.614  0.1097
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 56780 on 98 degrees of freedom
## Multiple R-squared:  0.332,    Adjusted R-squared:  0.3047
## F-statistic: 12.18 on 4 and 98 DF,  p-value: 4.481e-08
```

```
summary(AIC_loader2)
```

```
##
```

```
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -53569 -14118  -3370    6912 359972
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.024e+05  3.797e+04  -5.331 5.62e-07 ***
## x1           8.933e+00  1.687e+00   5.295 6.59e-07 ***
## x3           7.685e+03  1.386e+03   5.547 2.19e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 40860 on 105 degrees of freedom
## Multiple R-squared:  0.2692, Adjusted R-squared:  0.2553
## F-statistic: 19.34 on 2 and 105 DF,  p-value: 7.076e-08
```

```
summary(AIC_loader3)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -61130 -15702  -7705    2106 214330
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.479e+04  2.181e+04  -2.971 0.003461 **
## x1           3.914e+00  9.599e-01   4.078 7.37e-05 ***
## x3           2.207e+03  6.533e+02   3.378 0.000931 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 36530 on 149 degrees of freedom
## Multiple R-squared:  0.1081, Adjusted R-squared:  0.09612
## F-statistic: 9.029 on 2 and 149 DF,  p-value: 0.0001989
```

```
summary(AIC_loader4)
```

```
##
## Call:
## lm(formula = y ~ x2 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -92749 -32731 -10488    9860 588427
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   581618    209404   2.777  0.00696 **
## x2            -9088      4840  -1.878  0.06441 .
## x4             1000      4840   0.207  0.84111
## x5             1000      4840   0.207  0.84111
```

```
## x4          3848      1845   2.086  0.04048 *
## x5          -7060      2473  -2.855  0.00560 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 80820 on 73 degrees of freedom
## Multiple R-squared:  0.1088, Adjusted R-squared:  0.07217
## F-statistic:  2.97 on 3 and 73 DF,  p-value: 0.03733
summary(BIC_loader1)
```

```
##
## Call:
## lm(formula = y ~ x3 + x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -114724  -17914   -1981   11517   484782
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  147969     128479   1.152  0.2522
## x3           11135       2411    4.618 1.17e-05 ***
## x4           5476       1262    4.338 3.47e-05 ***
## x5          -4078       1735   -2.351  0.0207 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 57190 on 99 degrees of freedom
## Multiple R-squared:  0.3155, Adjusted R-squared:  0.2947
## F-statistic: 15.21 on 3 and 99 DF,  p-value: 3.258e-08
summary(BIC_loader2)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -53569  -14118   -3370    6912   359972
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.024e+05  3.797e+04  -5.331 5.62e-07 ***
## x1           8.933e+00  1.687e+00   5.295 6.59e-07 ***
## x3           7.685e+03  1.386e+03   5.547 2.19e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 40860 on 105 degrees of freedom
## Multiple R-squared:  0.2692, Adjusted R-squared:  0.2553
## F-statistic: 19.34 on 2 and 105 DF,  p-value: 7.076e-08
summary(BIC_loader3)
```

```
##
## Call:
## lm(formula = y ~ x1 + x3, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -61130 -15702  -7705    2106 214330
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.479e+04  2.181e+04  -2.971 0.003461 **
## x1           3.914e+00  9.599e-01   4.078 7.37e-05 ***
## x3           2.207e+03  6.533e+02   3.378 0.000931 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 36530 on 149 degrees of freedom
## Multiple R-squared:  0.1081, Adjusted R-squared:  0.09612
## F-statistic: 9.029 on 2 and 149 DF,  p-value: 0.0001989

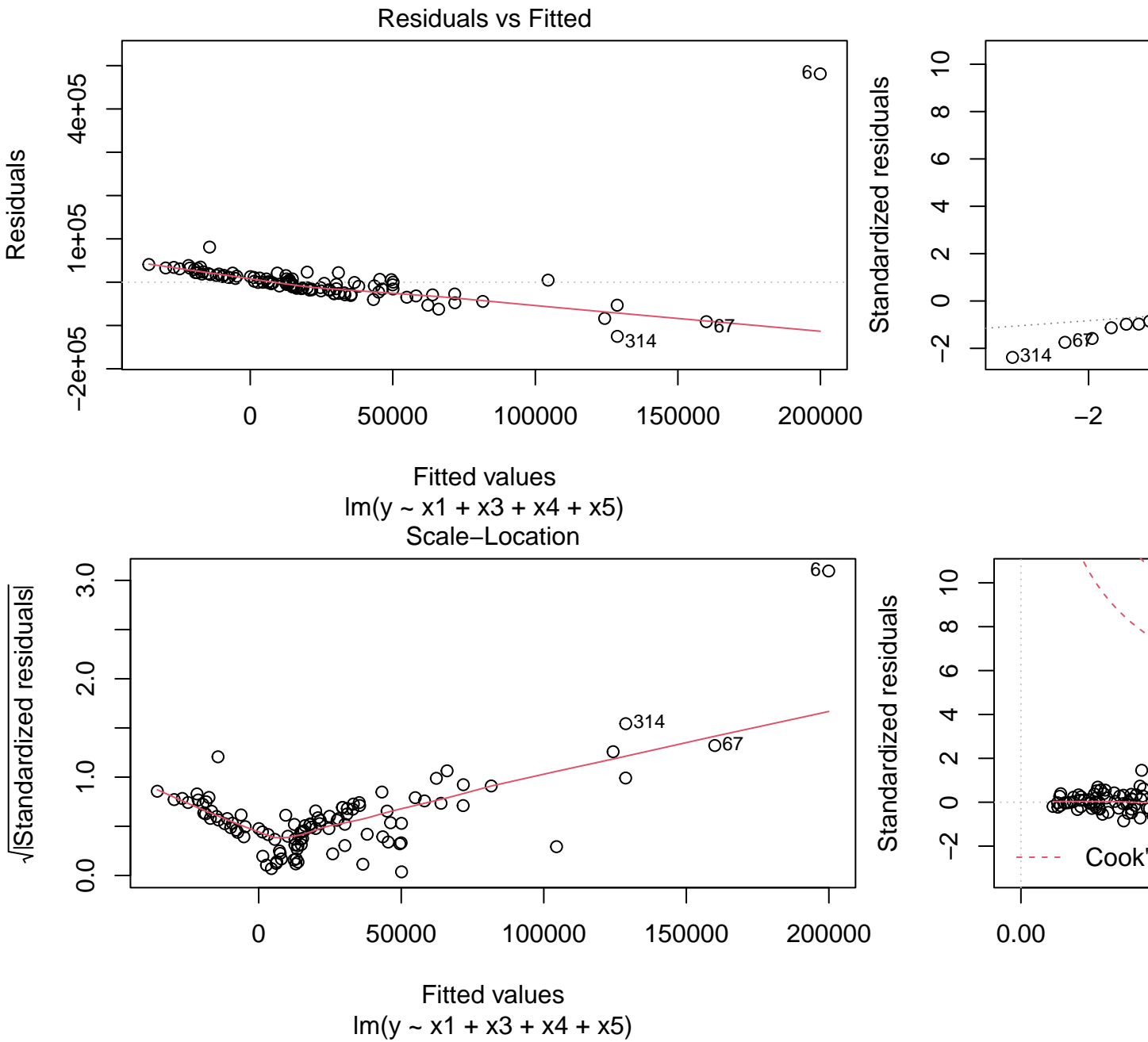
summary(BIC_loader4)

##
## Call:
## lm(formula = y ~ x4 + x5, data = loader)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -81236 -28955 -14033    3792 612011
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  247217     112021   2.207  0.0304 *
## x4           3934         1875   2.098  0.0393 *
## x5          -3686         1727  -2.134  0.0362 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 82190 on 74 degrees of freedom
## Multiple R-squared:  0.06575,    Adjusted R-squared:  0.0405
## F-statistic: 2.604 on 2 and 74 DF,  p-value: 0.08076
```

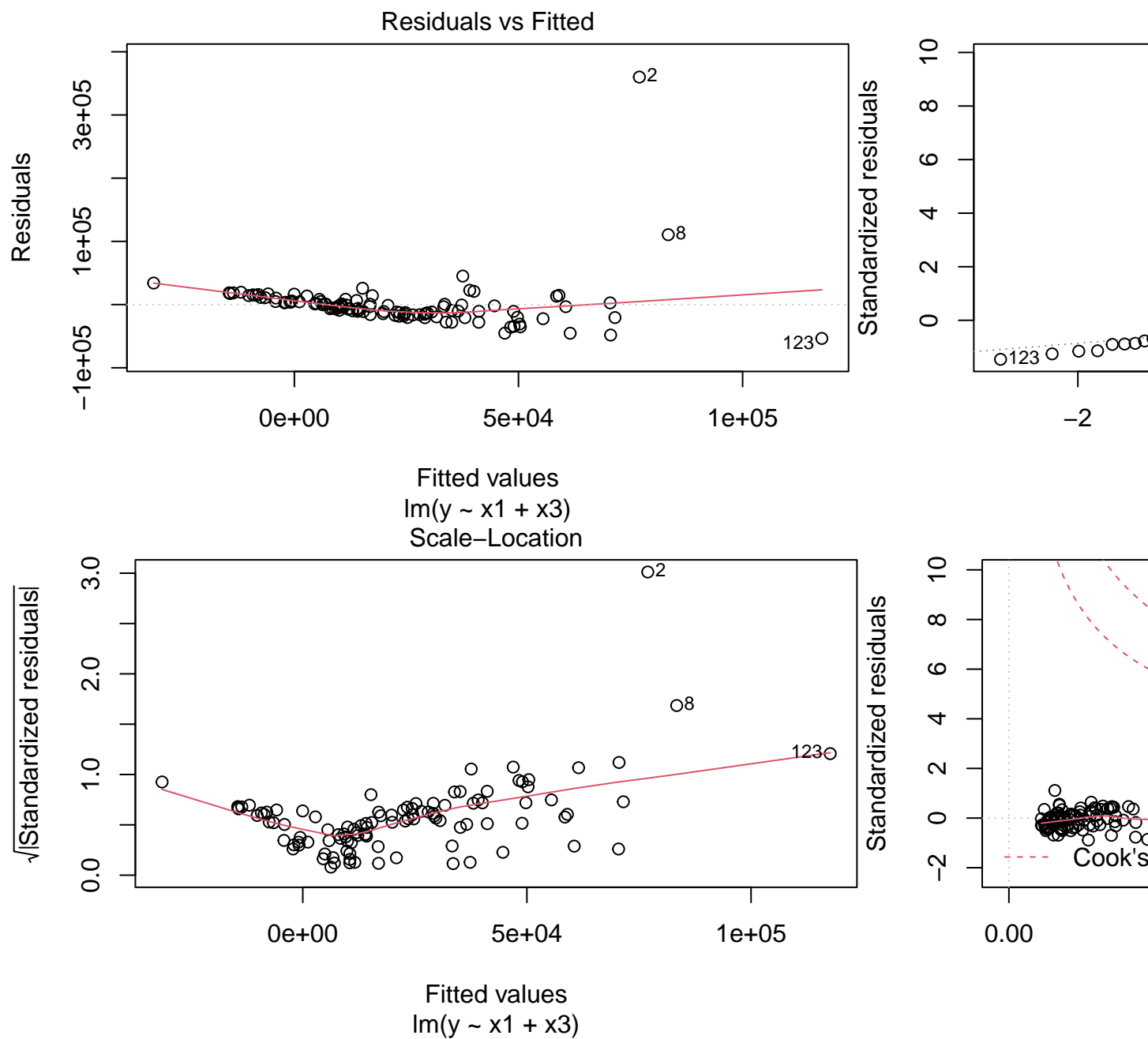
Obtain the residuals for each fitted model and prepare the diagnostic plots for each fitted model. State the conclusions.

Despite being categorized differently because of the AIC vs BIC method, the residual plots for the 4 geographic regions remain the same between AIC and BIC models. In terms of the plots compared between regions, Region 1 and Region 2 have very similar residual plots, with the only noticeable difference being that Region 2 has less extreme outliers in the Cook's Distance plot when compared to Region 1's plot. Region 4 also closely follows the trends of Region's 1 and 2. The most deviation comes from region 3, who's residuals deviate from the trends in a much more extreme fashion.

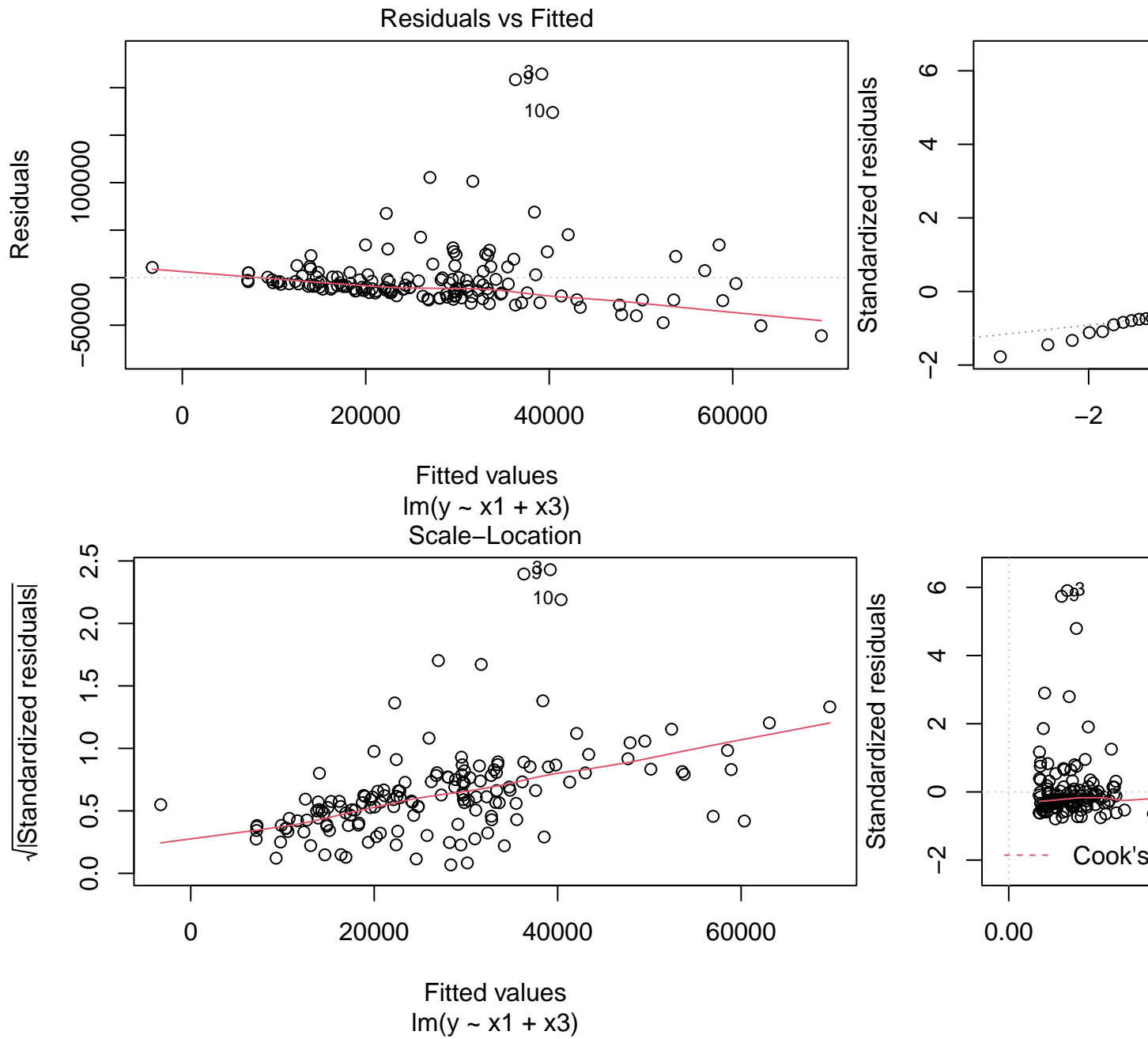
```
plot(AIC_loader1)
```



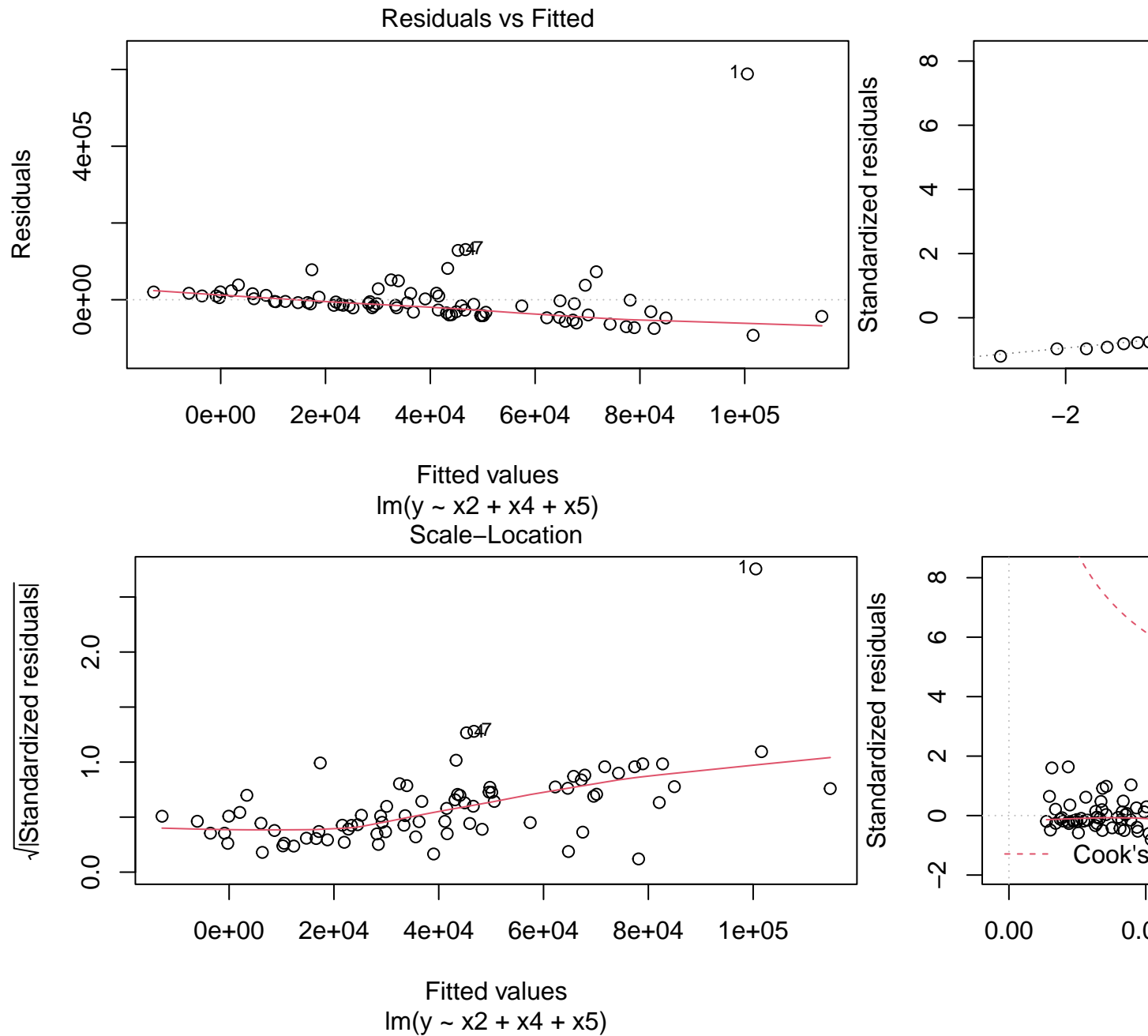
```
plot(AIC_loader2)
```



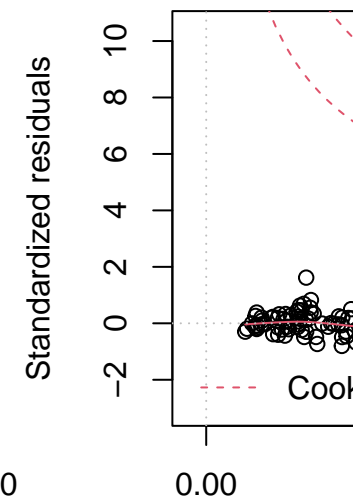
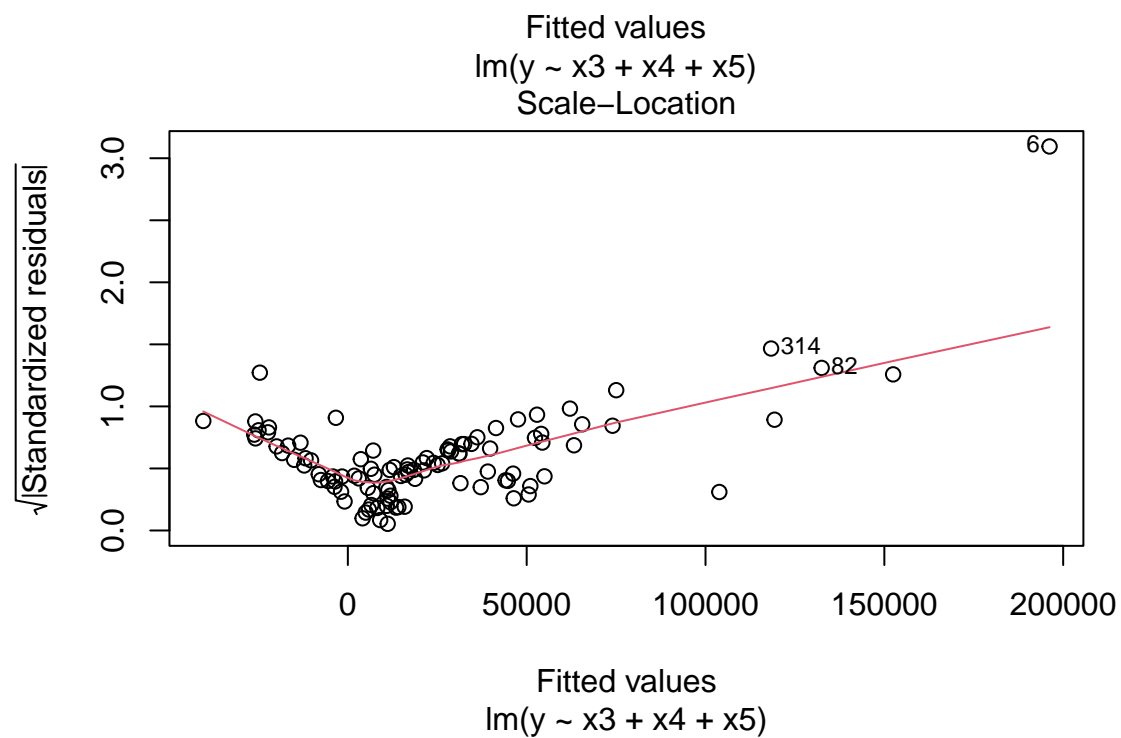
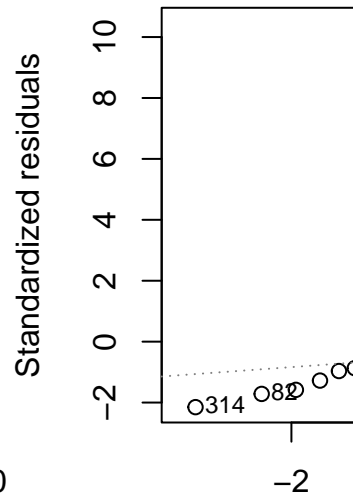
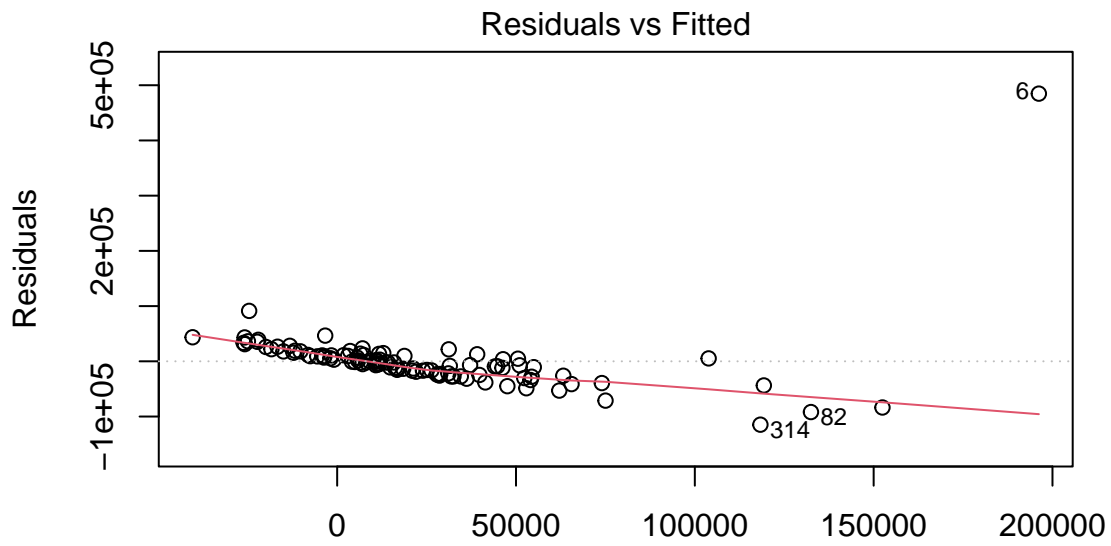
```
plot(AIC_loader3)
```



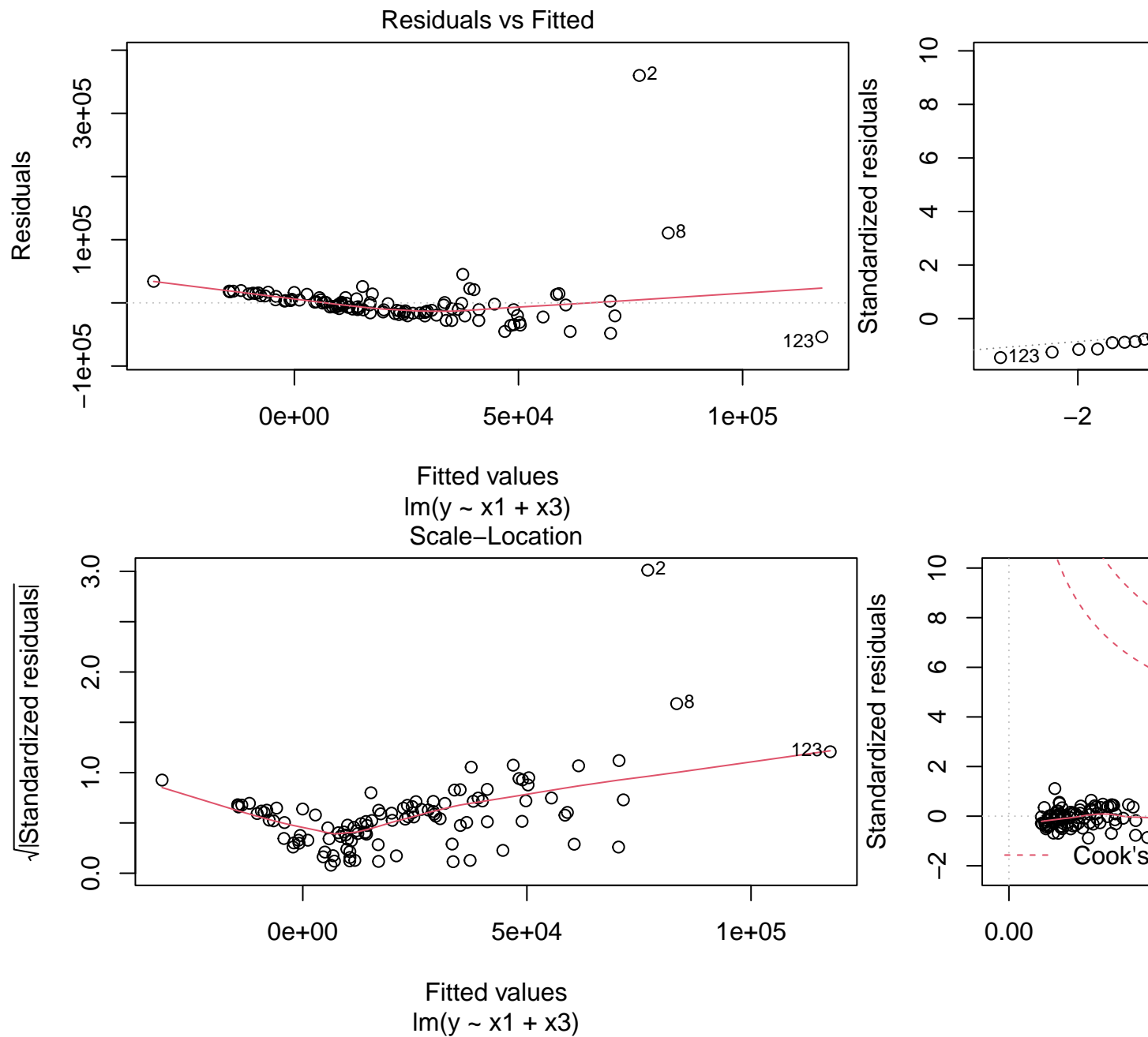
```
plot(AIC_loader4)
```



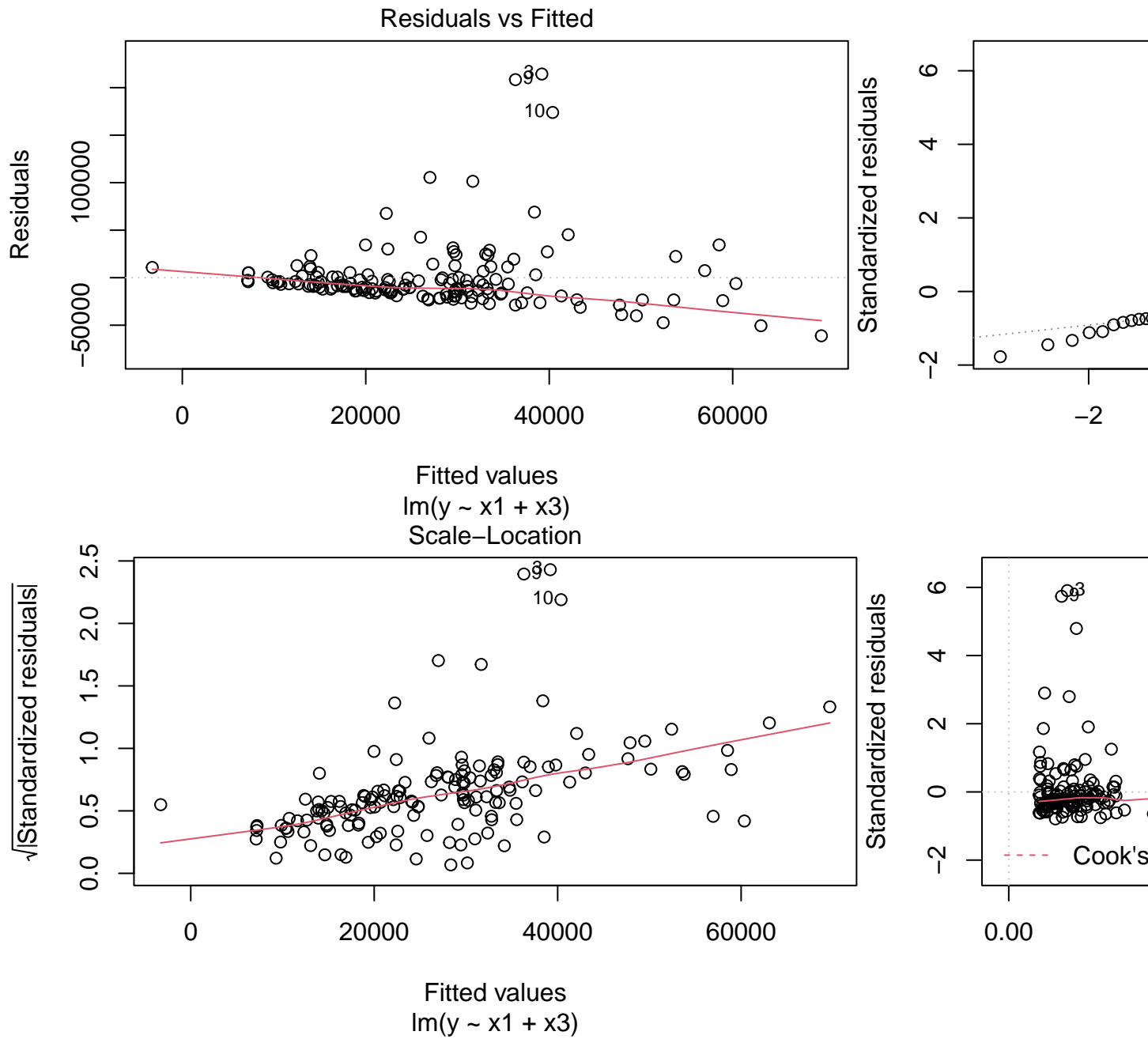
```
plot(BIC_loader1)
```

```
plot(BIC_loader2)
```



```
plot(BIC_loader3)
```



```
plot(BIC_loader4)
```

