

Regression_Analysis_Food

2024-01-18

```
### import libraries
```

```
library(car)
```

```
## Loading required package: carData
```

```
library(MASS)  
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:MASS':
```

```
##
```

```
##      select
```

```
## The following object is masked from 'package:car':
```

```
##
```

```
##      recode
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
library(tidyr)  
library(fastDummies)  
library(lubridate)
```

```
##
```

```
## Attaching package: 'lubridate'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      date, intersect, setdiff, union
```

```
library(coefplot)
```

```
## Loading required package: ggplot2
```

```
library(ggplot2)
library(leaps)
library(lmtest)
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      as.Date, as.Date.numeric
```

Loading the data

```
df = read.csv("data_cleaned_R_final.csv", head = TRUE)
```

```
head(df, 10)
```

```
##      X age income      political_party
## 1  25  65  3000      CDU/CSU
## 2  26  59   800      Keine Angabe
## 3  27  60  1750      Keine Angabe
## 4  28  73  2500      SPD
## 5  30  43  2500 Einer anderen Partei
## 6  31  49  2300      CDU/CSU
## 7  32  57   600      CDU/CSU
## 8  33  39  5000      SPD
## 9  34  62    0      Keine Angabe
## 10 36  45  2600      Keine Angabe
##
##                                     education
## 1 (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
## 2      Allgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
## 3      Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 4      Realschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## 5      Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 6      Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 7      Realschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## 8 (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
## 9 (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
## 10      Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
##      EUROSTAT      RLK2022      KTU2022
## 1      PU      zentral      Städtischer Kreis
## 2      PU sehr zentral      kreisfreie Großstadt
## 3      IN      peripher Ländlicher Kreis mit Verdichtungsansätzen
```

## 4	IN sehr zentral		Städtischer Kreis
## 5	PU sehr zentral		kreisfreie Großstadt
## 6	IN zentral		kreisfreie Großstadt
## 7	IN zentral		Städtischer Kreis
## 8	PU sehr zentral		kreisfreie Großstadt
## 9	PU sehr zentral		kreisfreie Großstadt
## 10	PU sehr zentral		kreisfreie Großstadt
##	federal_state	C02_housing	C02_electricity C02_housing_electricity
## 1	Saarland	5038.2000	1053.000 6091.2000
## 2	Hessen	1785.0000	487.500 2272.5000
## 3	Bayern	200.1024	663.000 863.1024
## 4	Bayern	648.4800	975.000 1623.4800
## 5	Berlin	1923.4862	390.000 2313.4862
## 6	Sachsen-Anhalt	2793.0960	663.000 3456.0960
## 7	Baden-Württemberg	1620.0000	112.000 1732.0000
## 8	Berlin	902.6745	26.320 928.9945
## 9	Nordrhein-Westfalen	2340.0000	825.825 3165.8250
## 10	Hessen	868.1526	47.600 915.7526
##	C02_cruise	C02_flight	C02_public_transport C02_car1 C02_car2 C02_car3
## 1	0	2440.0	0.0 1432.728 0.000 0
## 2	2710	5985.0	107.8 1944.608 1037.124 0
## 3	0	598.5	107.8 0.000 0.000 0
## 4	0	2287.6	0.0 1432.728 0.000 0
## 5	0	0.0	107.8 0.000 0.000 0
## 6	0	532.0	107.8 3581.820 0.000 0
## 7	0	0.0	0.0 0.000 0.000 0
## 8	4878	2074.8	107.8 5185.620 5185.620 0
## 9	0	0.0	107.8 2226.012 2782.515 0
## 10	0	3894.0	107.8 0.000 0.000 0
##	C02_car4	C02_car5	C02_car_total C02_mobility C02_food C02_other_consumption
## 1	0	0	1432.728 3872.728 1494.628 3766.100
## 2	0	0	2981.731 11784.531 1731.025 1444.879
## 3	0	0	0.000 706.300 1180.241 2433.480
## 4	0	0	1432.728 3720.328 1709.007 4152.125
## 5	0	0	0.000 107.800 1735.132 3766.100
## 6	0	0	3581.820 4221.620 1033.474 2317.600
## 7	0	0	0.000 0.000 1295.785 1520.925
## 8	0	0	10371.240 17431.840 2384.497 1216.740
## 9	0	0	5008.527 5116.327 1790.341 1376.075
## 10	0	0	0.000 4001.800 1407.010 3398.905
##	public_emission	C02_total	belief_diff_housing_electricity
## 1	1152	16376.656	-31
## 2	1152	18384.935	-38
## 3	1152	6335.123	40
## 4	1152	12356.940	-2
## 5	1152	9074.518	-43
## 6	1152	12180.790	-6
## 7	1152	5700.710	-1
## 8	1152	23114.072	5
## 9	1152	12600.568	-48
## 10	1152	10875.468	-1
##	belief_diff_mobility	belief_diff_food	belief_diff_other_consumption
## 1	-14	5	-68
## 2	-42	-26	23

## 3	11	49	9
## 4	-31	-9	-36
## 5	-2	-26	-53
## 6	22	93	24
## 7	72	60	37
## 8	-67	-61	12
## 9	-34	-5	18
## 10	-48	11	-64
##	belief_diff_total		
## 1	-15		
## 2	-76		
## 3	57		
## 4	-8		
## 5	-1		
## 6	13		
## 7	68		
## 8	-66		
## 9	-16		
## 10	-2		

Hypotheses for the regression model

1. The first dependent variable: actual CO2 emission H1a: age makes differences in the actual CO2 emission from everyday activity.

H1b: income makes differences in the actual CO2 emission from everyday activity.

H1c: education level makes differences in the actual CO2 emission from everyday activity.

H1d: the place of residence (city or countryside) in the actual CO2 emission from every day activity. H1e: the region (the federal state) makes differences in the actual CO2 emission from everyday activity.

H1f: the political party that the respondent supports makes differences in the actual CO2 emission from everyday activity.

2. The second dependent variable: cons H2a: age makes differences in the consumers' belief about CO2 emission from everyday activity.

H2b: income makes differences in the consumers' belief about CO2 emission from everyday activity.

H2c: education level makes differences in the consumers' belief about CO2 emission from everyday activity.

H2d: the place of residence (city or countryside) makes differences in the consumers' belief about CO2 emission from everyday activity.

H2e: the region (the federal state) makes differences in the consumers' belief about CO2 emission from everyday activity.

H2f: the political party that the respondent supports makes differences in the consumers' belief about CO2 emission from everyday activity.

Independent variables in the dataset

1. age: age, numerical variable
2. income: monthly net income in Euro, numerical variable, less than 10,000 EUR only (outlier removed)
3. education: categorical variable
4. urban_rural_class: categorical variable
5. federal_state: federal state, categorical variable
6. political_party: political_party, categorical variable

Dependent variables in the dataset

1. Actual CO2 from housing, electricity, mobility, food, other consumption

- 1) CO2_housing_electricity
- 2) CO2_mobility
- 3) CO2_food
- 4) CO2_other_consumption
- 5) CO2_total

2. Belief about CO2

- 1) belief_diff_housing_electricity
- 2) belief_diff_mobility
- 3) belief_diff_food
- 4) belief_diff_other_consumption
- 5) belief_diff_total

Data preparation

```
# change into categorical variable
```

```
df$education <-as.factor(df$education)
df$EUROSTAT <-as.factor(df$EUROSTAT)
df$RLK2022 <-as.factor(df$RLK2022)
df$KTU2022 <-as.factor(df$KTU2022)
df$political_party <-as.factor(df$political_party)
df$federal_state <-as.factor(df$federal_state)
```

```
## Select the classification for the urban_rural
```

```
#df1_1<- subset(df, select = -c(KTU2022, RLK2022) #EUROSTATS
```

```
df1_1<- subset(df, select = -c(KTU2022, EUROSTAT)) #RLK2022
```

```
#df1_1<- subset(df, select = -c(RLK2022, EUROSTAT)) #KTU2022
```

```
names(df1_1)[names(df1_1) == 'RLK2022'] <- 'urban_rural_class' #change the variable name!!
```

```
head(df1_1)
```

```
##      X age income      political_party
## 1 25  65   3000          CDU/CSU
## 2 26  59    800        Keine Angabe
## 3 27  60   1750        Keine Angabe
## 4 28  73   2500             SPD
## 5 30  43   2500 Einer anderen Partei
## 6 31  49   2300          CDU/CSU
##
##                                     education
## 1 (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
## 2      Allgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
```

```

## 3          Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 4          Realschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## 5          Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 6          Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
##  urban_rural_class  federal_state C02_housing C02_electricity
## 1          zentral      Saarland   5038.2000      1053.0
## 2      sehr zentral      Hessen    1785.0000      487.5
## 3          peripher      Bayern    200.1024      663.0
## 4      sehr zentral      Bayern    648.4800      975.0
## 5      sehr zentral      Berlin    1923.4862      390.0
## 6          zentral Sachsen-Anhalt 2793.0960      663.0
##  C02_housing_electricity C02_cruise C02_flight C02_public_transport C02_car1
## 1          6091.2000          0      2440.0          0.0 1432.728
## 2          2272.5000      2710      5985.0          107.8 1944.608
## 3          863.1024          0      598.5          107.8 0.000
## 4          1623.4800          0      2287.6          0.0 1432.728
## 5          2313.4862          0          0.0          107.8 0.000
## 6          3456.0960          0      532.0          107.8 3581.820
##  C02_car2 C02_car3 C02_car4 C02_car5 C02_car_total C02_mobility C02_food
## 1  0.000      0      0      0      1432.728      3872.728 1494.628
## 2 1037.124      0      0      0      2981.731      11784.531 1731.025
## 3  0.000      0      0      0          0.000      706.300 1180.241
## 4  0.000      0      0      0      1432.728      3720.328 1709.007
## 5  0.000      0      0      0          0.000      107.800 1735.132
## 6  0.000      0      0      0      3581.820      4221.620 1033.474
##  C02_other_consumption public_emission C02_total
## 1          3766.100          1152 16376.656
## 2          1444.879          1152 18384.935
## 3          2433.480          1152 6335.123
## 4          4152.125          1152 12356.940
## 5          3766.100          1152 9074.518
## 6          2317.600          1152 12180.790
##  belief_diff_housing_electricity belief_diff_mobility belief_diff_food
## 1          -31          -14          5
## 2          -38          -42         -26
## 3          40           11          49
## 4          -2          -31          -9
## 5          -43          -2         -26
## 6          -6           22          93
##  belief_diff_other_consumption belief_diff_total
## 1          -68          -15
## 2          23          -76
## 3          9           57
## 4         -36          -8
## 5         -53          -1
## 6          24          13

```

```
## Creating a demo-dataset for a quick regression model building
```

```

# Independent variables: age, income, political_party, education, urban_rural, federal_state
# Dependent variables: C02_food

```

```
df1 <- as_tibble(df1_1)
```

```
head(df1)
```

```
## # A tibble: 6 x 29
##       X   age income political~1 educa~2 urban~3 feder~4 C02_h~5 C02_e~6 C02_h~7
##   <int> <int>   <dbl> <fct>         <fct>   <fct>   <fct>   <dbl>   <dbl>   <dbl>
## 1    25    65   3000 CDU/CSU      (Fach-- zentral Saarla~  5038.   1053   6091.
## 2    26    59    800 Keine Anga~ Allgem~ sehr z~ Hessen    1785    488.   2272.
## 3    27    60   1750 Keine Anga~ Berufs~ periph~ Bayern     200.    663    863.
## 4    28    73   2500 SPD          Realsc~ sehr z~ Bayern     648.    975   1623.
## 5    30    43   2500 Einer ande~ Berufs~ sehr z~ Berlin   1923.    390   2313.
## 6    31    49   2300 CDU/CSU      Berufs~ zentral Sachse~  2793.    663   3456.
## # ... with 19 more variables: C02_cruise <dbl>, C02_flight <dbl>,
## #   C02_public_transport <dbl>, C02_car1 <dbl>, C02_car2 <dbl>, C02_car3 <dbl>,
## #   C02_car4 <dbl>, C02_car5 <dbl>, C02_car_total <dbl>, C02_mobility <dbl>,
## #   C02_food <dbl>, C02_other_consumption <dbl>, public_emission <dbl>,
## #   C02_total <dbl>, belief_diff_housing_electricity <dbl>,
## #   belief_diff_mobility <dbl>, belief_diff_food <dbl>,
## #   belief_diff_other_consumption <dbl>, belief_diff_total <dbl>, and ...
```

```
df1 <- df1 %>% select(2, 3, 4, 5, 6, 7, 21) #10, 20, 21, 22, 24
```

```
df1
```

```
## # A tibble: 588 x 7
##       age income political_party      education      urban~1 feder~2 C02_f~3
##   <int>   <dbl> <fct>         <fct>         <fct>   <fct>   <dbl>
## 1    65   3000 CDU/CSU      (Fach-) Hochschula~ zentral Saarla~  1495.
## 2    59    800 Keine Angabe Allgemeine oder fa~ sehr z~ Hessen    1731.
## 3    60   1750 Keine Angabe Berufsausbildung, ~ periph~ Bayern    1180.
## 4    73   2500 SPD          Realschulabschluss~ sehr z~ Bayern    1709.
## 5    43   2500 Einer anderen Partei Berufsausbildung, ~ sehr z~ Berlin    1735.
## 6    49   2300 CDU/CSU      Berufsausbildung, ~ zentral Sachse~  1033.
## 7    57    600 CDU/CSU      Realschulabschluss~ zentral Baden~  1296.
## 8    39   5000 SPD          (Fach-) Hochschula~ sehr z~ Berlin    2384.
## 9    62     0 Keine Angabe (Fach-) Hochschula~ sehr z~ Nordrh~  1790.
## 10   45   2600 Keine Angabe Berufsausbildung, ~ sehr z~ Hessen    1407.
## # ... with 578 more rows, and abbreviated variable names 1: urban_rural_class,
## #   2: federal_state, 3: C02_food
```

```
## Creating a demo-dataset for a quick regression model building
```

```
# Independent variables: age, income, political_party, education, urban_rural, federal_state
# Dependent variables: belief_diff_food
```

```
df2 <- as_tibble(df1_1)
```

```
head(df1_1)
```

```
##       X age income      political_party
## 1 25  65   3000          CDU/CSU
## 2 26  59    800          Keine Angabe
```

```

## 3 27 60 1750 Keine Angabe
## 4 28 73 2500 SPD
## 5 30 43 2500 Einer anderen Partei
## 6 31 49 2300 CDU/CSU
##
## education
## 1 (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
## 2 Allgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
## 3 Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 4 Realschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## 5 Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## 6 Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
## urban_rural_class federal_state CO2_housing CO2_electricity
## 1 zentral Saarland 5038.2000 1053.0
## 2 sehr zentral Hessen 1785.0000 487.5
## 3 peripher Bayern 200.1024 663.0
## 4 sehr zentral Bayern 648.4800 975.0
## 5 sehr zentral Berlin 1923.4862 390.0
## 6 zentral Sachsen-Anhalt 2793.0960 663.0
## CO2_housing_electricity CO2_cruise CO2_flight CO2_public_transport CO2_car1
## 1 6091.2000 0 2440.0 0.0 1432.728
## 2 2272.5000 2710 5985.0 107.8 1944.608
## 3 863.1024 0 598.5 107.8 0.000
## 4 1623.4800 0 2287.6 0.0 1432.728
## 5 2313.4862 0 0.0 107.8 0.000
## 6 3456.0960 0 532.0 107.8 3581.820
## CO2_car2 CO2_car3 CO2_car4 CO2_car5 CO2_car_total CO2_mobility CO2_food
## 1 0.000 0 0 0 1432.728 3872.728 1494.628
## 2 1037.124 0 0 0 2981.731 11784.531 1731.025
## 3 0.000 0 0 0 0.000 706.300 1180.241
## 4 0.000 0 0 0 1432.728 3720.328 1709.007
## 5 0.000 0 0 0 0.000 107.800 1735.132
## 6 0.000 0 0 0 3581.820 4221.620 1033.474
## CO2_other_consumption public_emission CO2_total
## 1 3766.100 1152 16376.656
## 2 1444.879 1152 18384.935
## 3 2433.480 1152 6335.123
## 4 4152.125 1152 12356.940
## 5 3766.100 1152 9074.518
## 6 2317.600 1152 12180.790
## belief_diff_housing_electricity belief_diff_mobility belief_diff_food
## 1 -31 -14 5
## 2 -38 -42 -26
## 3 40 11 49
## 4 -2 -31 -9
## 5 -43 -2 -26
## 6 -6 22 93
## belief_diff_other_consumption belief_diff_total
## 1 -68 -15
## 2 23 -76
## 3 9 57
## 4 -36 -8
## 5 -53 -1
## 6 24 13

```



```
df2 <- df2 %>% select(2, 3, 4, 5, 6, 7, 27) #25, 26, 27, 28, 29
```

```
df2
```

```
## # A tibble: 588 x 7
##   age income political_party education urban~1 feder~2 belie~3
##   <int> <dbl> <fct> <fct> <fct> <fct> <dbl>
## 1 65 3000 CDU/CSU (Fach-) Hochschule~ zentral Saarla~ 5
## 2 59 800 Keine Angabe Allgemeine oder fa~ sehr z~ Hessen -26
## 3 60 1750 Keine Angabe Berufsausbildung, ~ periph~ Bayern 49
## 4 73 2500 SPD Realschulabschluss~ sehr z~ Bayern -9
## 5 43 2500 Einer anderen Partei Berufsausbildung, ~ sehr z~ Berlin -26
## 6 49 2300 CDU/CSU Berufsausbildung, ~ zentral Sachse~ 93
## 7 57 600 CDU/CSU Realschulabschluss~ zentral Baden-- 60
## 8 39 5000 SPD (Fach-) Hochschule~ sehr z~ Berlin -61
## 9 62 0 Keine Angabe (Fach-) Hochschule~ sehr z~ Nordrh~ -5
## 10 45 2600 Keine Angabe Berufsausbildung, ~ sehr z~ Hessen 11
## # ... with 578 more rows, and abbreviated variable names 1: urban_rural_class,
## # 2: federal_state, 3: belief_diff_food
```

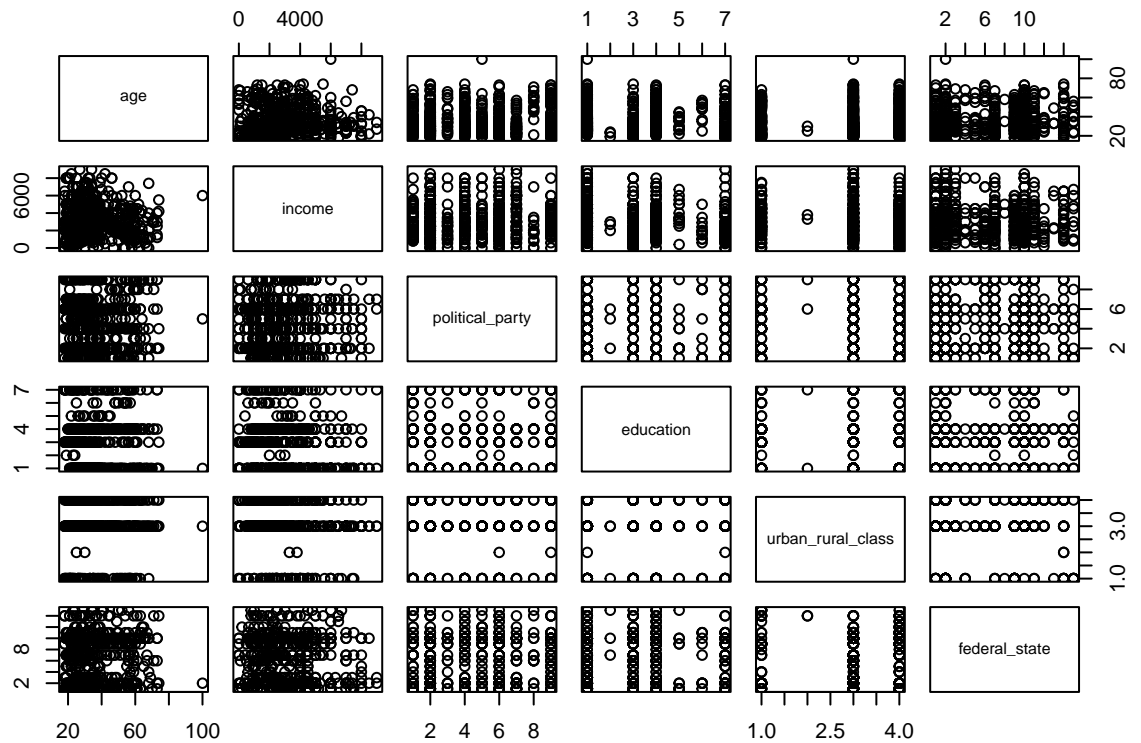
I. Exploratory Data Analysis

Check the Jupyter notebook: EDA_scatter_plot_actual_belief

II. Multivariate Regression: CO2 food

```
# Checking the possible correlation in the data
```

```
plot(df1[1:6])
```



1. Modeling

```
table(df1$political_party)
```

```
##
##           AfD      Bündnis 90/Die Grünen Bündnis Sarah Wagenknecht
##           58           143           23
##      CDU/CSU           Die Linke      Einer anderen Partei
##           75           44           111
##           FDP      Keine Angabe      SPD
##           48           15           71
```

```
table(df1$education)
```

```
##
## (Fach-) Hochschulabschluss (Bachelor, Master, Magister, Diplom, Staatsexamen)
##                                           253
##                                           (Noch) kein Abschluss
##                                           3
## Allgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
##                                           131
## Berufsausbildung, Lehre oder Ausbildung an einer Fachschule
##                                           118
##                                           Doktorgrad oder Habilitation
##                                           13
## Hauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss
##                                           11
```

```
## Realschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## 59
```

```
table(df1$urban_rural_class)
```

```
##
## peripher sehr peripher sehr zentral zentral
## 79 2 350 157
```

```
table(df1$federal_state)
```

```
##
## Baden-Württemberg Bayern Berlin
## 94 100 44
## Brandenburg Bremen Hamburg
## 8 15 25
## Hessen Mecklenburg-Vorpommern Niedersachsen
## 50 2 58
## Nordrhein-Westfalen Rheinland-Pfalz Saarland
## 117 30 10
## Sachsen-Anhalt Schleswig-Holstein Thüringen
## 4 22 9
```

```
## defining a reference level
```

```
df1$political_party <- relevel(df1$political_party, ref='Bündnis 90/Die Grünen')
df1$education <- relevel(df1$education, ref='(Fach-) Hochschulabschluss (Bachelor, Master, Magister, D
df1$urban_rural_class <- relevel(df1$urban_rural_class, ref='sehr zentral')
df1$federal_state <- relevel(df1$federal_state, ref='Nordrhein-Westfalen')
```

```
# regression model with all variables
```

```
modell1 <- lm(CO2_food ~ age + income + political_party + education + urban_rural_class + federal_state
summary(modell1)
```

```
##
## Call:
## lm(formula = CO2_food ~ age + income + political_party + education +
## urban_rural_class + federal_state, data = df1)
##
## Residuals:
## Min 1Q Median 3Q Max
## -1228.42 -351.29 -80.26 261.35 3024.52
##
## Coefficients:
## Estimate
## (Intercept) 1649.95700
## age -4.75323
## income 0.01770
## political_partyAfD 494.41873
## political_partyBündnis Sarah Wagenknecht 237.97895
```

## political_partyCDU/CSU	401.30696
## political_partyDie Linke	60.97770
## political_partyEiner anderen Partei	269.78750
## political_partyFDP	338.27776
## political_partyKeine Angabe	52.06262
## political_partySPD	390.52818
## education(Noch) kein Abschluss	1036.26216
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	80.48354
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	87.69220
## educationDoktorgrad oder Habilitation	10.56860
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	156.98337
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	102.58551
## urban_rural_classperipher	-182.03001
## urban_rural_classsehr peripher	-112.32325
## urban_rural_classzentral	-57.63702
## federal_stateBaden-Württemberg	-44.78441
## federal_stateBayern	81.71124
## federal_stateBerlin	-23.55651
## federal_stateBrandenburg	79.06846
## federal_stateBremen	-114.86223
## federal_stateHamburg	-152.23293
## federal_stateHessen	-63.76098
## federal_stateMecklenburg-Vorpommern	-164.39711
## federal_stateNiedersachsen	226.47128
## federal_stateRheinland-Pfalz	-8.27212
## federal_stateSaarland	-43.16881
## federal_stateSachsen-Anhalt	-512.64010
## federal_stateSchleswig-Holstein	120.69573
## federal_stateThüringen	122.87544
##	Std. Error
## (Intercept)	102.10932
## age	1.80773
## income	0.01208
## political_partyAfD	86.49161
## political_partyBündnis Sarah Wagenknecht	121.22905
## political_partyCDU/CSU	77.68269
## political_partyDie Linke	93.63026
## political_partyEiner anderen Partei	69.49607
## political_partyFDP	89.96842
## political_partyKeine Angabe	155.83677
## political_partySPD	79.10575
## education(Noch) kein Abschluss	316.48391
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	61.51900
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	62.75856
## educationDoktorgrad oder Habilitation	153.41144
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	175.11760
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	79.81045
## urban_rural_classperipher	80.42637
## urban_rural_classsehr peripher	398.15281
## urban_rural_classzentral	58.96386
## federal_stateBaden-Württemberg	76.59576
## federal_stateBayern	78.64226
## federal_stateBerlin	95.53706
## federal_stateBrandenburg	200.66496

## federal_stateBremen	146.36473
## federal_stateHamburg	119.11131
## federal_stateHessen	91.38668
## federal_stateMecklenburg-Vorpommern	383.60044
## federal_stateNiedersachsen	92.97044
## federal_stateRheinland-Pfalz	114.66837
## federal_stateSaarland	180.05848
## federal_stateSachsen-Anhalt	277.55695
## federal_stateSchleswig-Holstein	134.12031
## federal_stateThüringen	203.29023
##	t value
## (Intercept)	16.159
## age	-2.629
## income	1.465
## political_partyAfD	5.716
## political_partyBündnis Sarah Wagenknecht	1.963
## political_partyCDU/CSU	5.166
## political_partyDie Linke	0.651
## political_partyEiner anderen Partei	3.882
## political_partyFDP	3.760
## political_partyKeine Angabe	0.334
## political_partySPD	4.937
## education(Noch) kein Abschluss	3.274
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	1.308
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	1.397
## educationDoktorgrad oder Habilitation	0.069
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	0.896
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	1.285
## urban_rural_classperipher	-2.263
## urban_rural_classsehr peripher	-0.282
## urban_rural_classzentral	-0.977
## federal_stateBaden-Württemberg	-0.585
## federal_stateBayern	1.039
## federal_stateBerlin	-0.247
## federal_stateBrandenburg	0.394
## federal_stateBremen	-0.785
## federal_stateHamburg	-1.278
## federal_stateHessen	-0.698
## federal_stateMecklenburg-Vorpommern	-0.429
## federal_stateNiedersachsen	2.436
## federal_stateRheinland-Pfalz	-0.072
## federal_stateSaarland	-0.240
## federal_stateSachsen-Anhalt	-1.847
## federal_stateSchleswig-Holstein	0.900
## federal_stateThüringen	0.604
##	Pr(> t)
## (Intercept)	< 2e-16
## age	0.008791
## income	0.143391
## political_partyAfD	1.78e-08
## political_partyBündnis Sarah Wagenknecht	0.050140
## political_partyCDU/CSU	3.34e-07
## political_partyDie Linke	0.515148
## political_partyEiner anderen Partei	0.000116

## political_partyFDP	0.000188
## political_partyKeine Angabe	0.738442
## political_partySPD	1.05e-06
## education(Noch) kein Abschluss	0.001125
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	0.191324
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	0.162884
## educationDoktorgrad oder Habilitation	0.945102
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	0.370404
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	0.199202
## urban_rural_classperipher	0.024003
## urban_rural_classsehr peripher	0.777964
## urban_rural_classzentral	0.328750
## federal_stateBaden-Württemberg	0.558997
## federal_stateBayern	0.299247
## federal_stateBerlin	0.805333
## federal_stateBrandenburg	0.693709
## federal_stateBremen	0.432926
## federal_stateHamburg	0.201759
## federal_stateHessen	0.485654
## federal_stateMecklenburg-Vorpommern	0.668408
## federal_stateNiedersachsen	0.015167
## federal_stateRheinland-Pfalz	0.942517
## federal_stateSaarland	0.810614
## federal_stateSachsen-Anhalt	0.065284
## federal_stateSchleswig-Holstein	0.368561
## federal_stateThüringen	0.545803
##	
## (Intercept)	***
## age	**
## income	
## political_partyAfD	***
## political_partyBündnis Sarah Wagenknecht	.
## political_partyCDU/CSU	***
## political_partyDie Linke	
## political_partyEiner anderen Partei	***
## political_partyFDP	***
## political_partyKeine Angabe	
## political_partySPD	***
## education(Noch) kein Abschluss	**
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	
## educationDoktorgrad oder Habilitation	
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	
## urban_rural_classperipher	*
## urban_rural_classsehr peripher	
## urban_rural_classzentral	
## federal_stateBaden-Württemberg	
## federal_stateBayern	
## federal_stateBerlin	
## federal_stateBrandenburg	
## federal_stateBremen	
## federal_stateHamburg	
## federal_stateHessen	

```
## federal_stateMecklenburg-Vorpommern
## federal_stateNiedersachsen
## federal_stateRheinland-Pfalz
## federal_stateSaarland
## federal_stateSachsen-Anhalt
## federal_stateSchleswig-Holstein
## federal_stateThüringen
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 528.4 on 554 degrees of freedom
## Multiple R-squared:  0.1539, Adjusted R-squared:  0.1035
## F-statistic: 3.053 on 33 and 554 DF,  p-value: 6.253e-08
```

```
# Checking the VIFs for multicollinearity
```

```
vif(model1)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## age           1.313360  1      1.146019
## income        1.099357  1      1.048502
## political_party 1.794759  8      1.037231
## education      1.848270  6      1.052520
## urban_rural_class 2.066166  3      1.128568
## federal_state   3.002832 14      1.040051
```

```
# threshold for multicollinearity
# Calculating the threshold
```

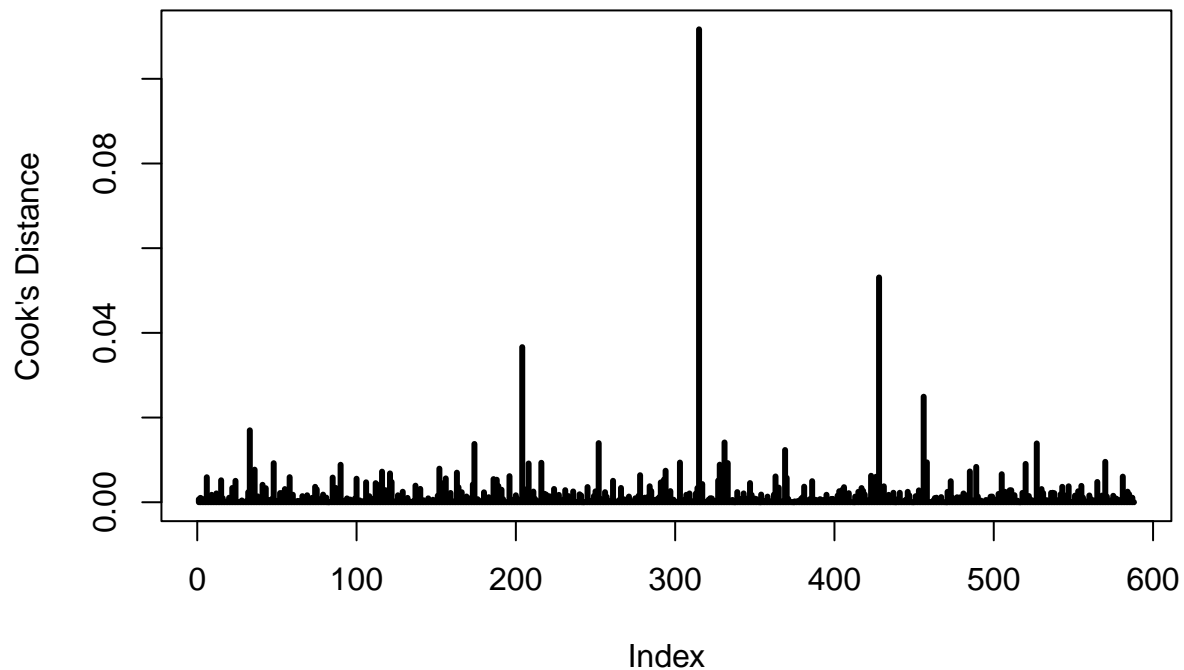
```
max(10, 1/(1-summary(model1)$r.square))
```

```
## [1] 10
```

```
# Checking outliers: estimate of the influence of data point; summary of how much a regression model ch
```

```
cook = cooks.distance(model1)
plot(cook,
      type="h",
      lwd=3,
      ylab = "Cook's Distance",
      main="Cook's Distance")
abline(h = 1)
```

Cook's Distance



```
influential = cooks.distance(model1)[which(cook > 3*mean(cook, na.rm=TRUE))]  
influential
```

```
##          6          33          36          48          58          85  
## 0.005905500 0.016987374 0.007701172 0.009222091 0.005922695 0.005829832  
##          90         100         116         121         152         156  
## 0.008849586 0.005559505 0.007228999 0.006809638 0.007934072 0.005632054  
##         163         174         186         196         204         208  
## 0.006989085 0.013764261 0.005469170 0.006165627 0.036633543 0.009152248  
##         216         252         278         294         303         315  
## 0.009319069 0.013966868 0.006389329 0.007449710 0.009375852 0.111679185  
##         328         330         331         333         363         369  
## 0.008878170 0.005651801 0.014119699 0.009246118 0.006092358 0.012333071  
##         370         423         426         428         456         458  
## 0.005769210 0.006248660 0.006002794 0.053089194 0.024925336 0.009419272  
##         485         489         505         520         527         570  
## 0.007270443 0.008350133 0.006622905 0.009055882 0.013890920 0.009547564  
##         581  
## 0.006051068
```

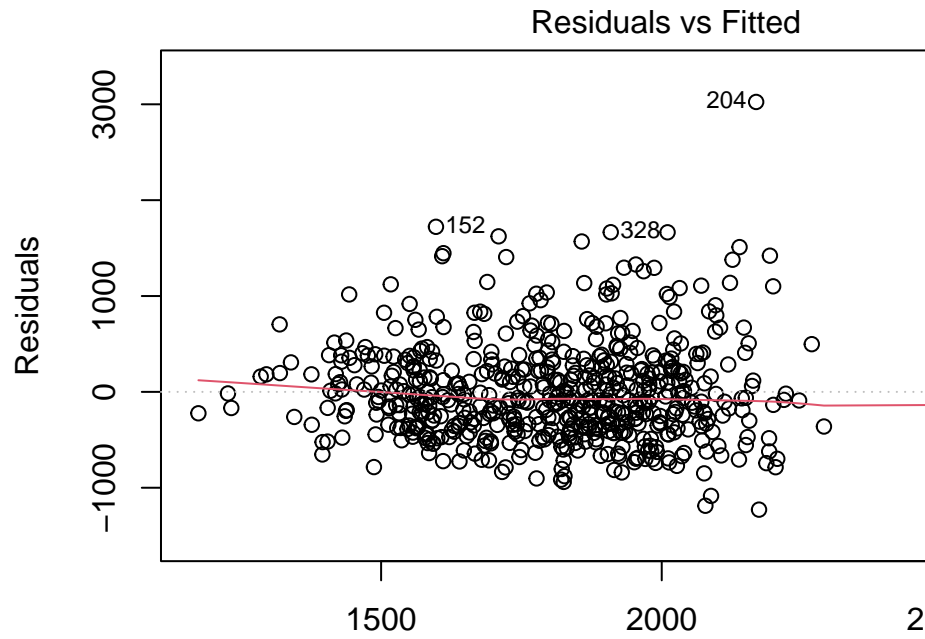
```
influential = influential[!is.na(influential)]  
influential_vector = c(as.numeric(rownames(data.frame(influential))))
```

```
df1[influential_vector, ]
```



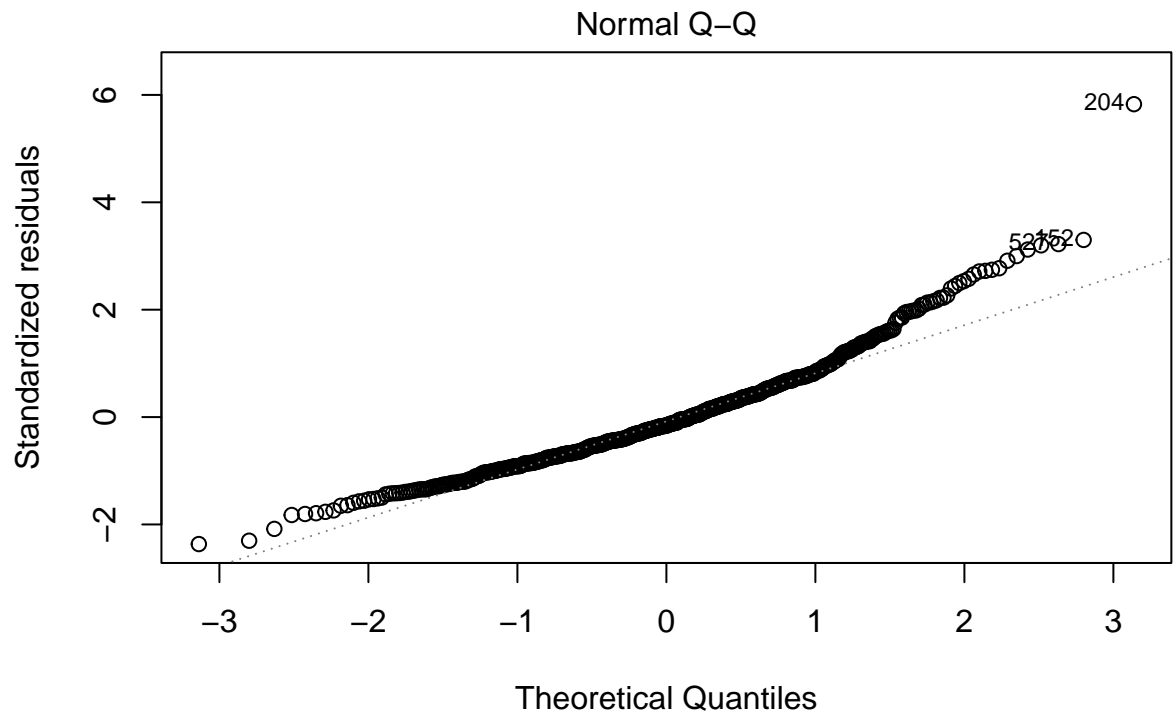
```
## # A tibble: 43 x 7
##   age income political_party education urban-1 feder-2 CO2_f-3
##   <int> <dbl> <fct> <fct> <fct> <fct> <dbl>
## 1 49 2300 CDU/CSU Berufsausbildung, Lehre~ zentral Sachse~ 1033.
## 2 37 3500 Keine Angabe Hauptschulabschluss (Vo~ sehr z~ Bayern 948.
## 3 19 1000 FDP Allgemeine oder fachgeb~ sehr z~ Bayern 890.
## 4 19 5000 CDU/CSU Allgemeine oder fachgeb~ sehr z~ Rheinl~ 3259.
## 5 53 1500 AfD Hauptschulabschluss (Vo~ periph~ Bayern 1363.
## 6 57 1000 AfD Hauptschulabschluss (Vo~ periph~ Baden~ 1248.
## 7 19 3000 FDP Allgemeine oder fachgeb~ sehr z~ Hessen 3226.
## 8 50 2500 AfD Berufsausbildung, Lehre~ periph~ Thürin~ 1418.
## 9 23 2000 CDU/CSU Realschulabschluss (Mit~ zentral Rheinl~ 3003.
## 10 58 0 CDU/CSU Realschulabschluss (Mit~ zentral Bayern 2981.
## # ... with 33 more rows, and abbreviated variable names 1: urban_rural_class,
## # 2: federal_state, 3: CO2_food
```

```
plot(model1)
```

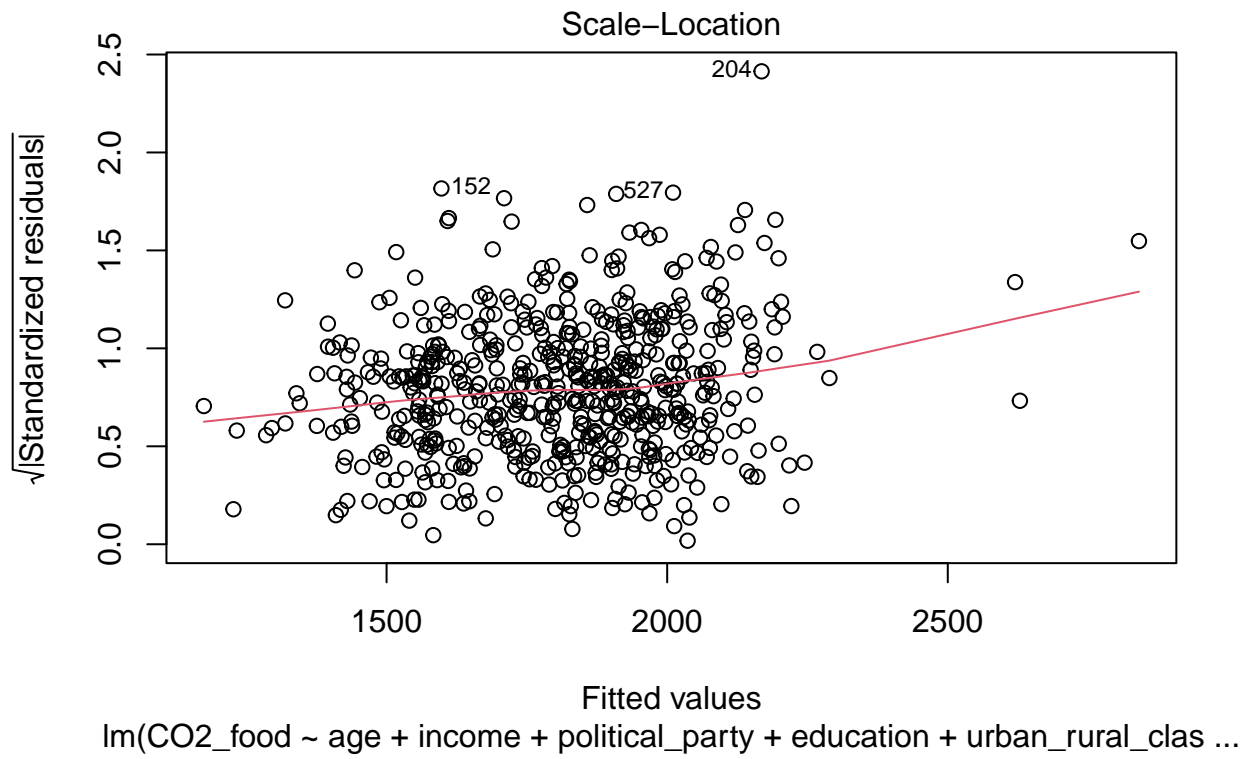


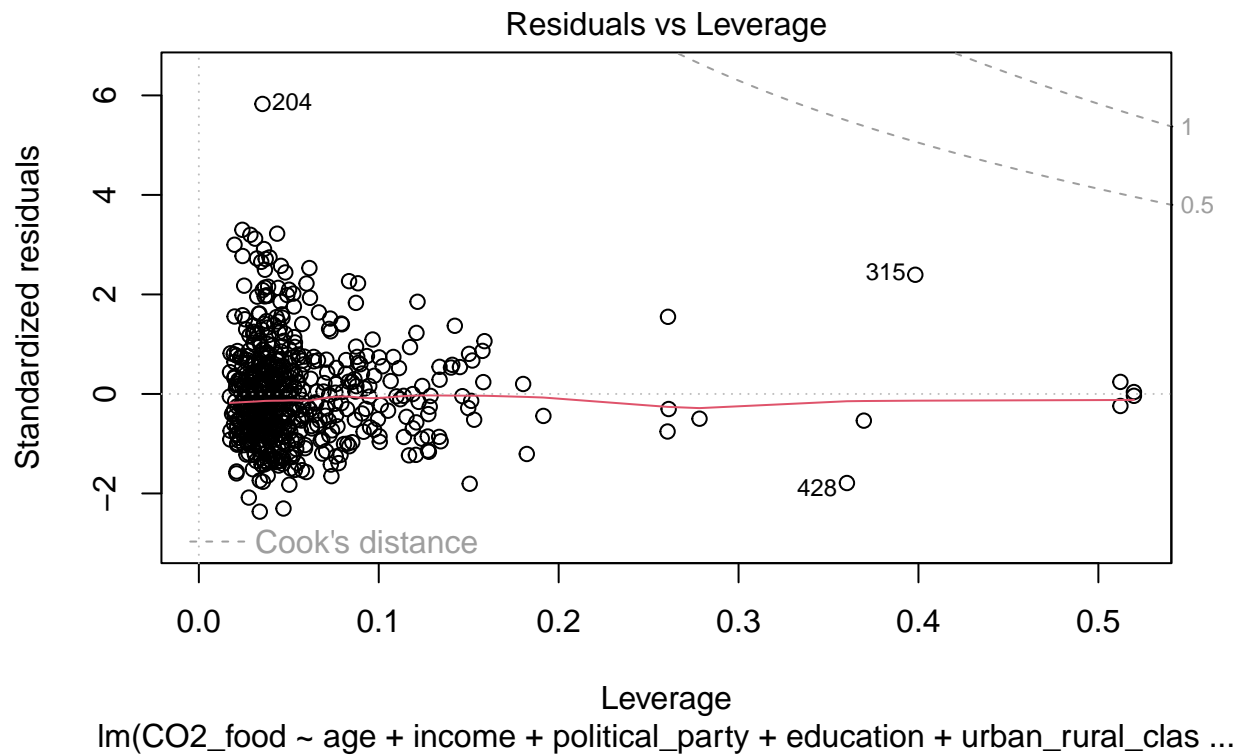
2. Assumptions check in the residuals

$\text{lm}(\text{CO2_food} \sim \text{age} + \text{income} + \text{political_party} + \text{education})$



$\text{lm}(\text{CO2_food} \sim \text{age} + \text{income} + \text{political_party} + \text{education} + \text{urban_rural_clas} \dots)$

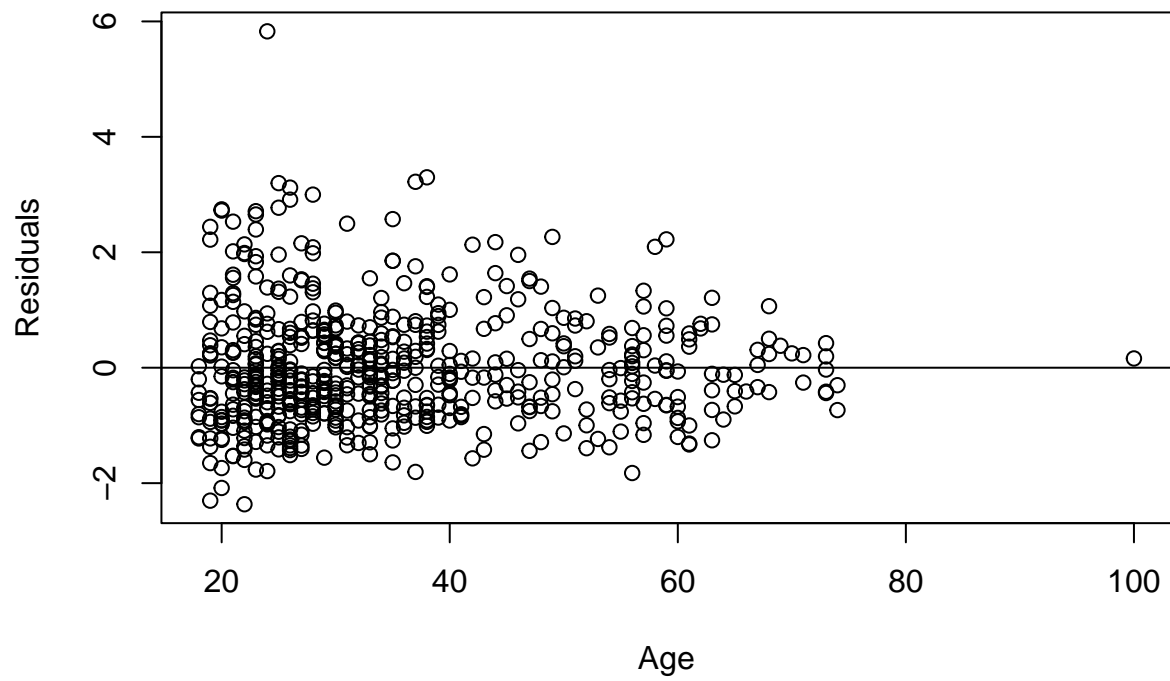




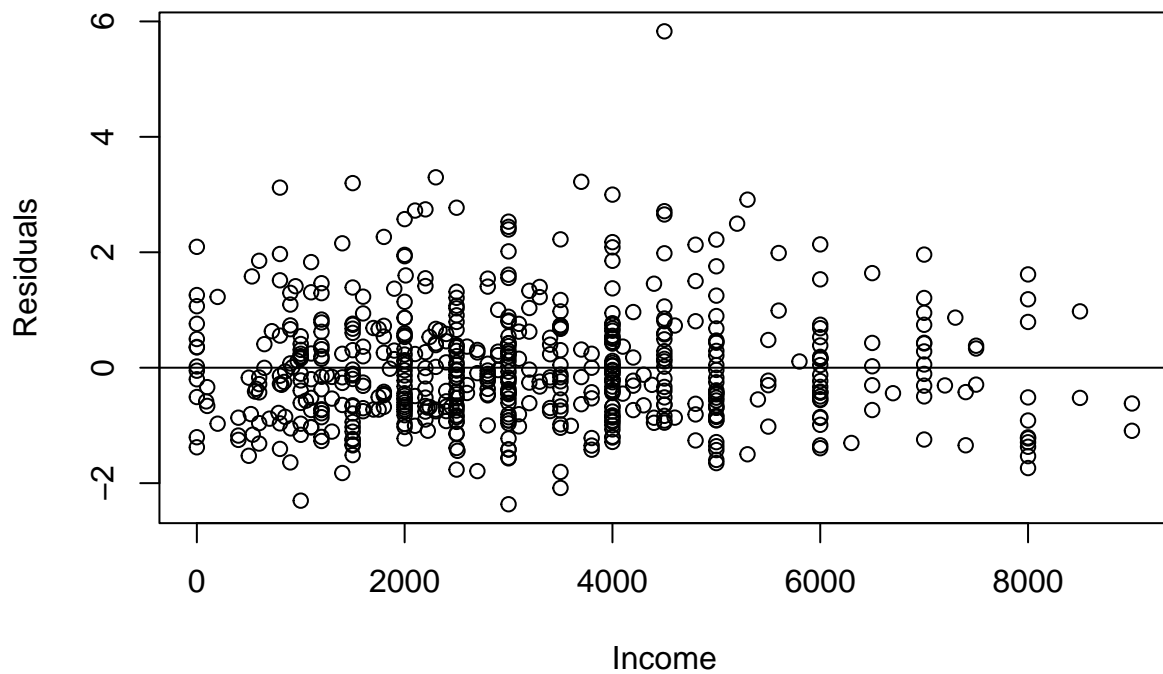
```
res1 = stdres(model1) ## (Standardized) Residuals

# Linearity assumption/Mean zero assumption

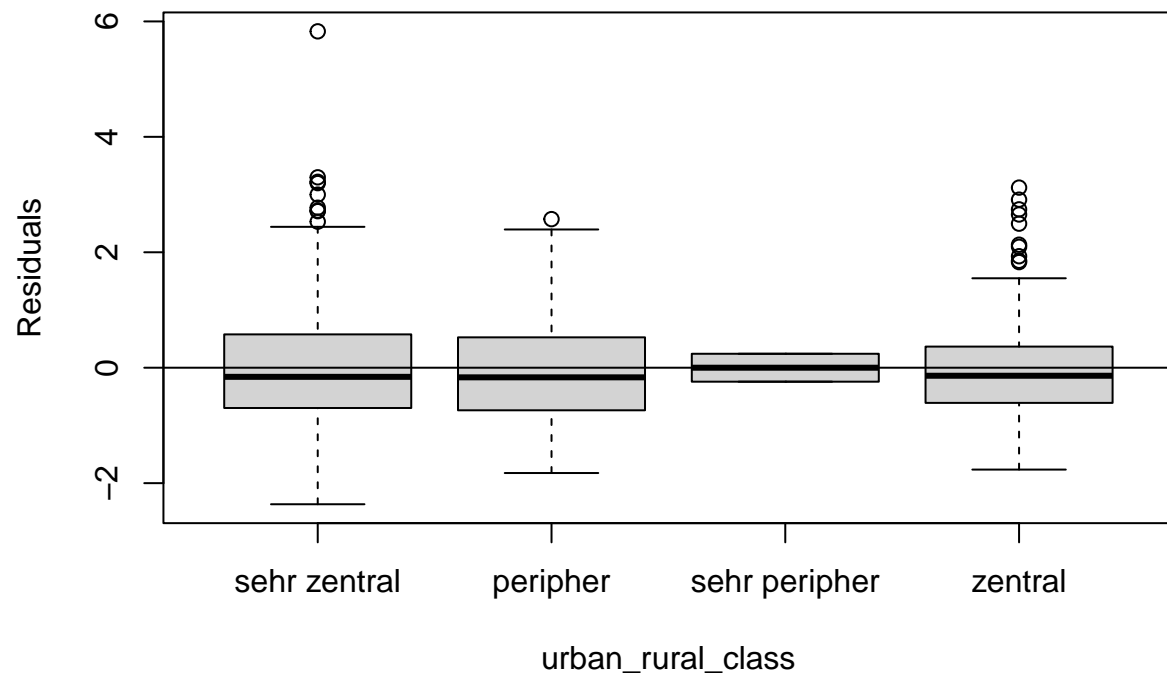
plot(df1$age, res1, xlab = "Age", ylab = "Residuals")
abline(h = 0)
```



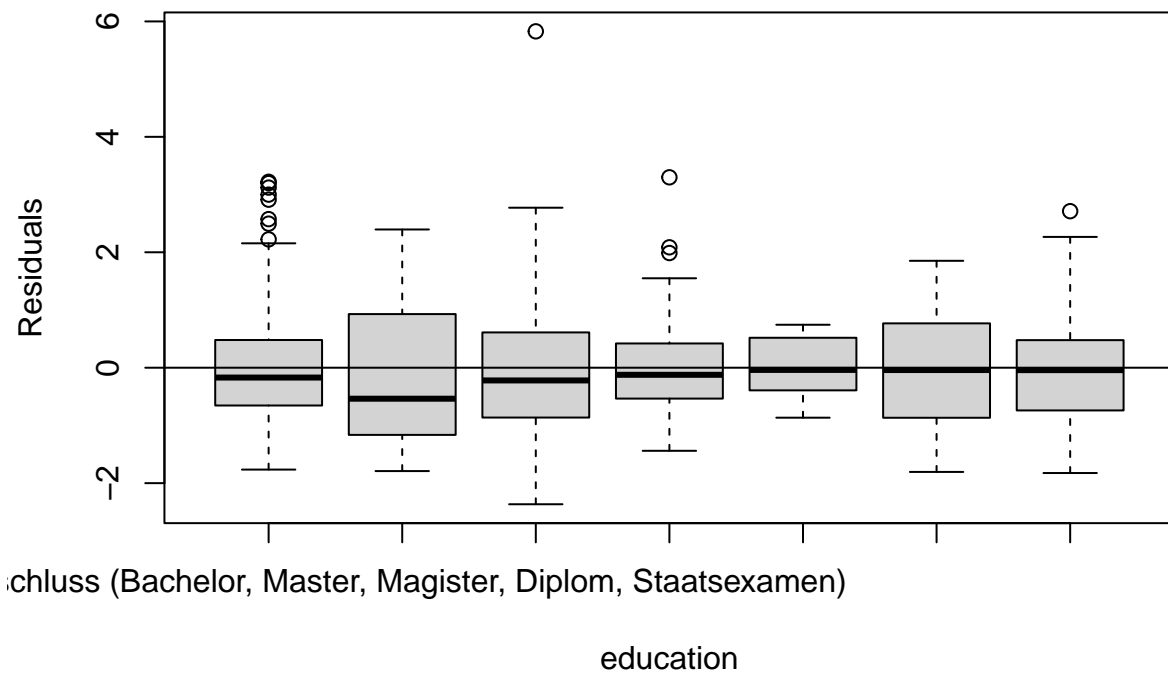
```
plot(df1$income, res1, xlab = "Income", ylab = "Residuals")  
abline(h = 0)
```



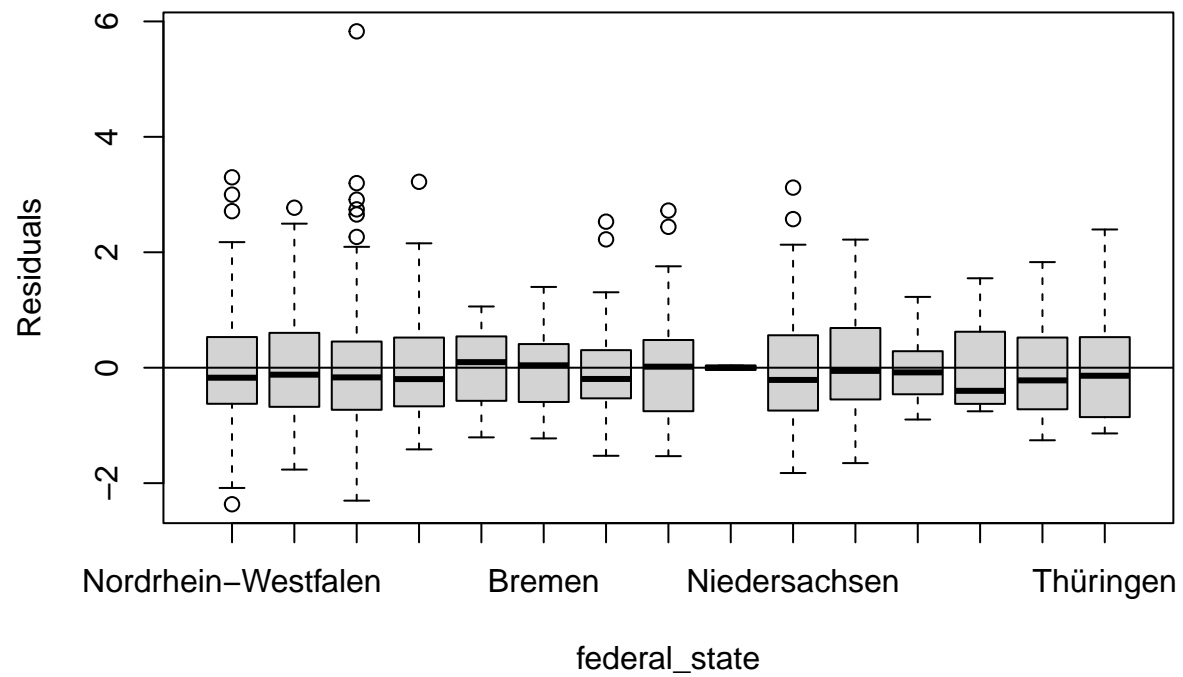
```
plot(df1$urban_rural_class, res1, xlab = "urban_rural_class", ylab = "Residuals")  
abline(h = 0)
```



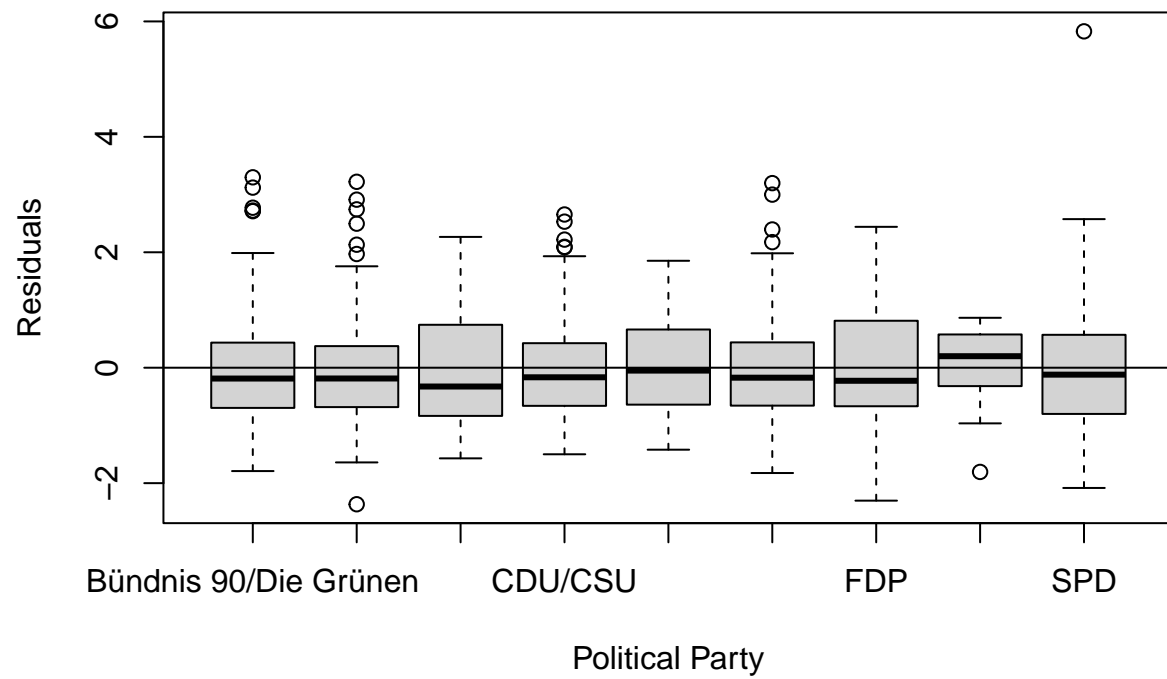
```
plot(df1$education, res1, xlab = "education", ylab = "Residuals")  
abline(h = 0)
```



```
plot(df1$federal_state, res1, xlab = "federal_state", ylab = "Residuals")
abline(h = 0)
```

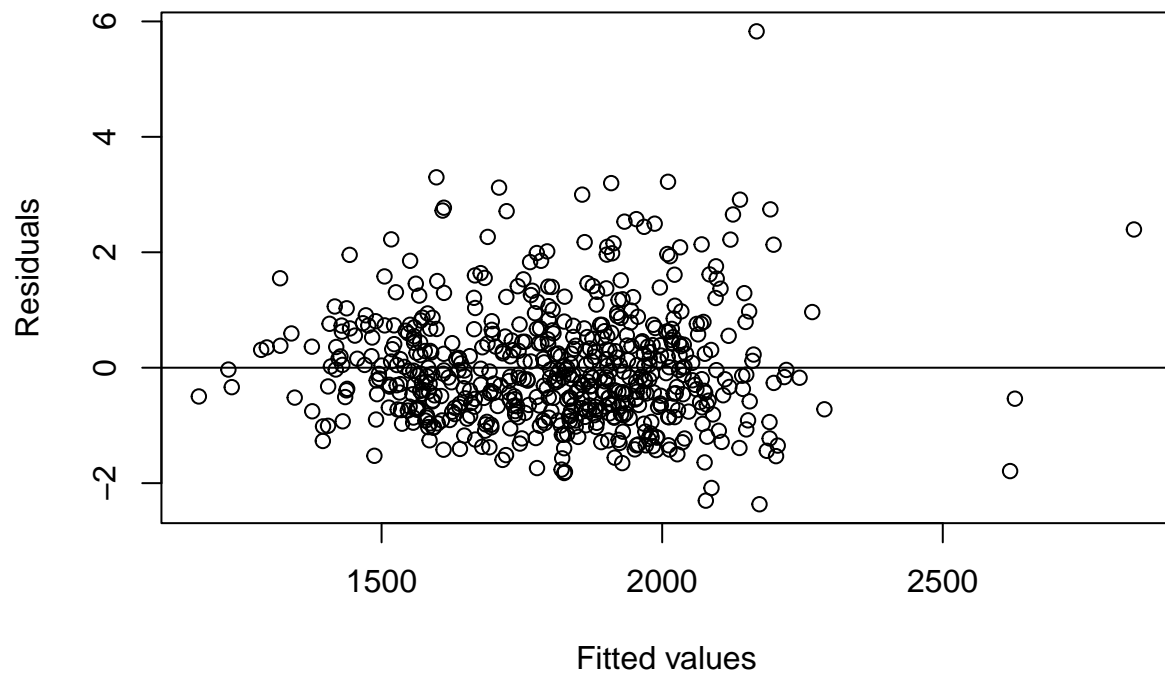



```
plot(df1$political_party, res1, xlab = "Political Party", ylab = "Residuals")
abline(h = 0)
```



Constant variance and independent error term assumption

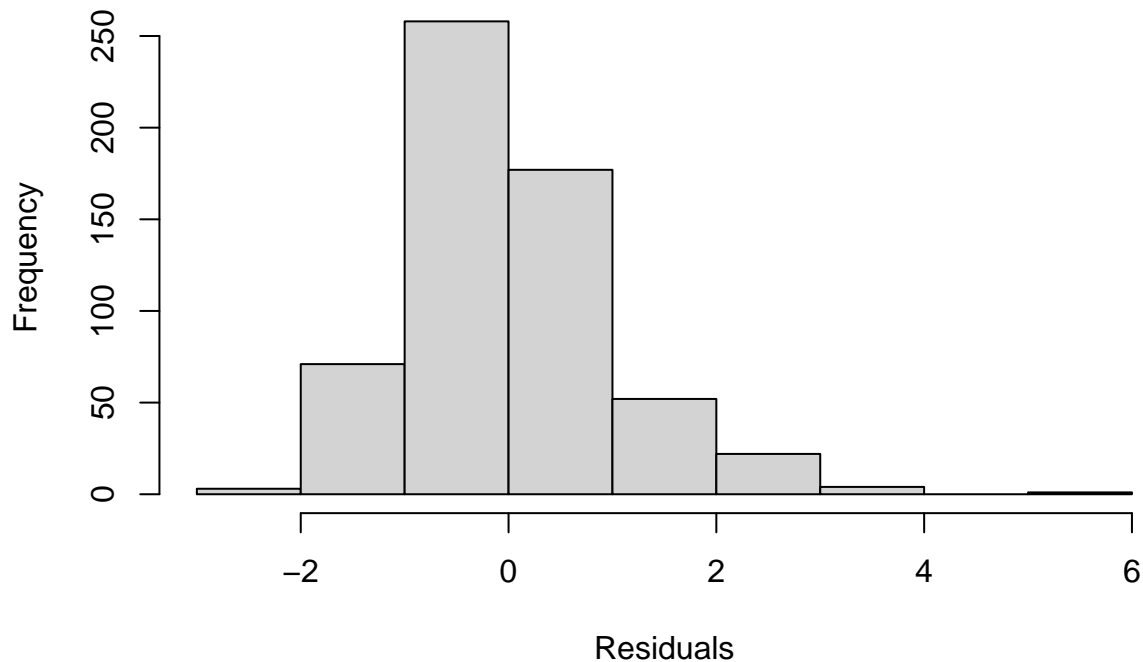
```
plot(fitted(model1), res1, xlab = "Fitted values", ylab = "Residuals")
abline(h = 0)
```



```
# Normality assumption
```

```
hist(res1, xlab="Residuals", main= "Histogram of Residuals")
```

Histogram of Residuals



```
## normality test using shapiro-test: reject the H0  
#H0: the sample comes from a normal distribution
```

```
res1_num = res1[is.finite(res1)]  
shapiro.test(res1_num)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: res1_num  
## W = 0.94981, p-value = 3.066e-13
```

```
### Backward regression using AIC: starting with all of the variables - best
```

```
step_model1 <- stepAIC(model1, trace=TRUE, direction= "backward")
```

3. Variable Selection, model outcome and assumption check

```
## Start: AIC=7406.36  
## CO2_food ~ age + income + political_party + education + urban_rural_class +  
## federal_state  
##
```

```
##              Df Sum of Sq      RSS      AIC
## - federal_state 14  5461417 160151258 7398.8
## - urban_rural_class 3   1438123 156127963 7405.8
## <none>                                154689841 7406.4
## - income          1    599569 155289409 7406.6
## - education        6    3798811 158488651 7408.6
## - age              1    1930466 156620307 7411.7
## - political_party  8   16816209 171506050 7451.0
##
## Step: AIC=7398.76
## CO2_food ~ age + income + political_party + education + urban_rural_class
##
##              Df Sum of Sq      RSS      AIC
## - urban_rural_class 3    361311 160512568 7394.1
## <none>                                160151258 7398.8
## - income          1    627914 160779172 7399.1
## - education        6    3576894 163728151 7399.8
## - age              1    1757161 161908418 7403.2
## - political_party  8   15973782 176125039 7438.7
##
## Step: AIC=7394.09
## CO2_food ~ age + income + political_party + education
##
##              Df Sum of Sq      RSS      AIC
## <none>                                160512568 7394.1
## - income          1    641618 161154186 7394.4
## - education        6    3420504 163933072 7394.5
## - age              1    1828133 162340701 7398.7
## - political_party  8   15721609 176234177 7433.0
```

```
summary(step_model1)
```

```
##
## Call:
## lm(formula = CO2_food ~ age + income + political_party + education,
##     data = df1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1176.66  -364.06   -76.22   279.85  3146.72
##
## Coefficients:
##              Estimate
## (Intercept) 1643.75236
## age        -4.56653
## income      0.01796
## political_partyAfD 472.66377
## political_partyBündnis Sarah Wagenknecht 266.31954
## political_partyCDU/CSU 370.52532
## political_partyDie Linke 71.31061
## political_partyEiner anderen Partei 246.60463
## political_partyFDP 322.43817
## political_partyKeine Angabe 11.89768
## political_partySPD 379.28628
```

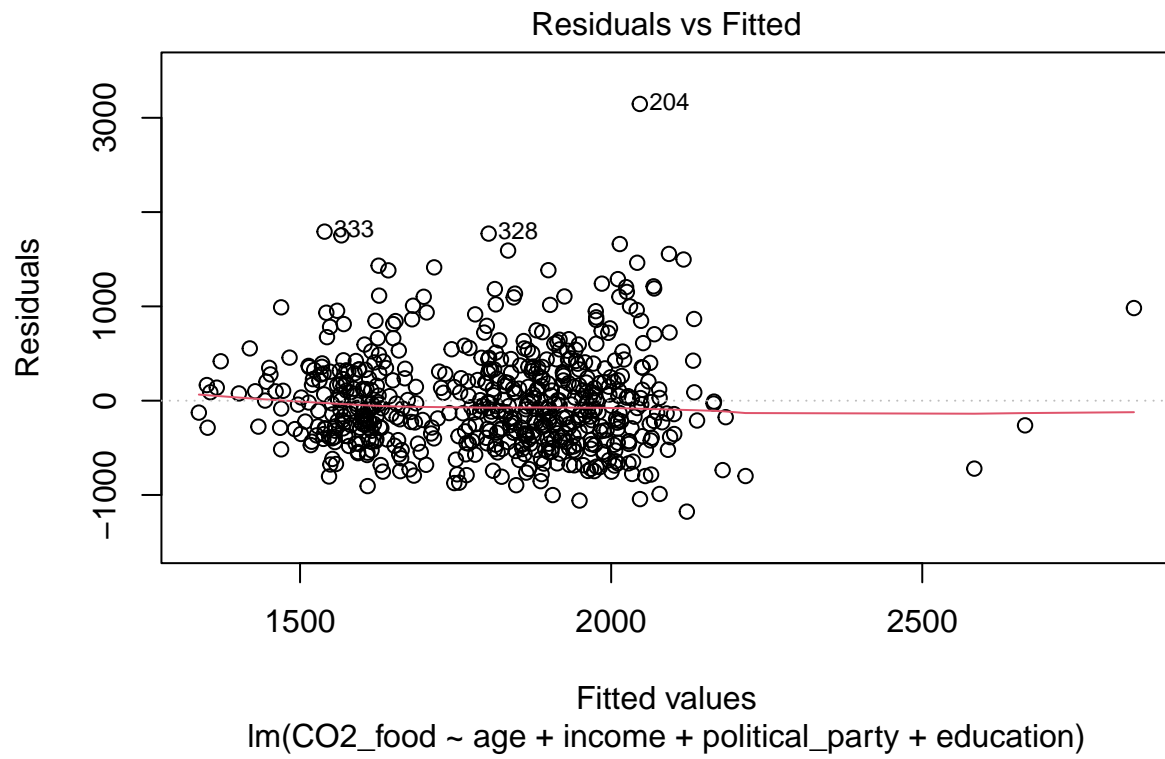
## education(Noch) kein Abschluss	1001.01093
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	51.79976
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	54.90458
## educationDoktorgrad oder Habilitation	8.00367
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	126.51540
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	95.99489
##	Std. Error
## (Intercept)	89.01839
## age	1.79068
## income	0.01189
## political_partyAfD	84.86621
## political_partyBündnis Sarah Wagenknecht	120.42916
## political_partyCDU/CSU	76.92301
## political_partyDie Linke	92.15183
## political_partyEiner anderen Partei	68.07532
## political_partyFDP	89.55848
## political_partyKeine Angabe	152.28022
## political_partySPD	78.14095
## education(Noch) kein Abschluss	309.93720
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	59.83958
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	61.48889
## educationDoktorgrad oder Habilitation	151.68538
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	170.72390
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	78.53385
##	t value
## (Intercept)	18.465
## age	-2.550
## income	1.511
## political_partyAfD	5.570
## political_partyBündnis Sarah Wagenknecht	2.211
## political_partyCDU/CSU	4.817
## political_partyDie Linke	0.774
## political_partyEiner anderen Partei	3.623
## political_partyFDP	3.600
## political_partyKeine Angabe	0.078
## political_partySPD	4.854
## education(Noch) kein Abschluss	3.230
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	0.866
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	0.893
## educationDoktorgrad oder Habilitation	0.053
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	0.741
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	1.222
##	Pr(> t)
## (Intercept)	< 2e-16
## age	0.011027
## income	0.131397
## political_partyAfD	3.94e-08
## political_partyBündnis Sarah Wagenknecht	0.027402
## political_partyCDU/CSU	1.87e-06
## political_partyDie Linke	0.439347
## political_partyEiner anderen Partei	0.000318
## political_partyFDP	0.000345
## political_partyKeine Angabe	0.937752
## political_partySPD	1.56e-06

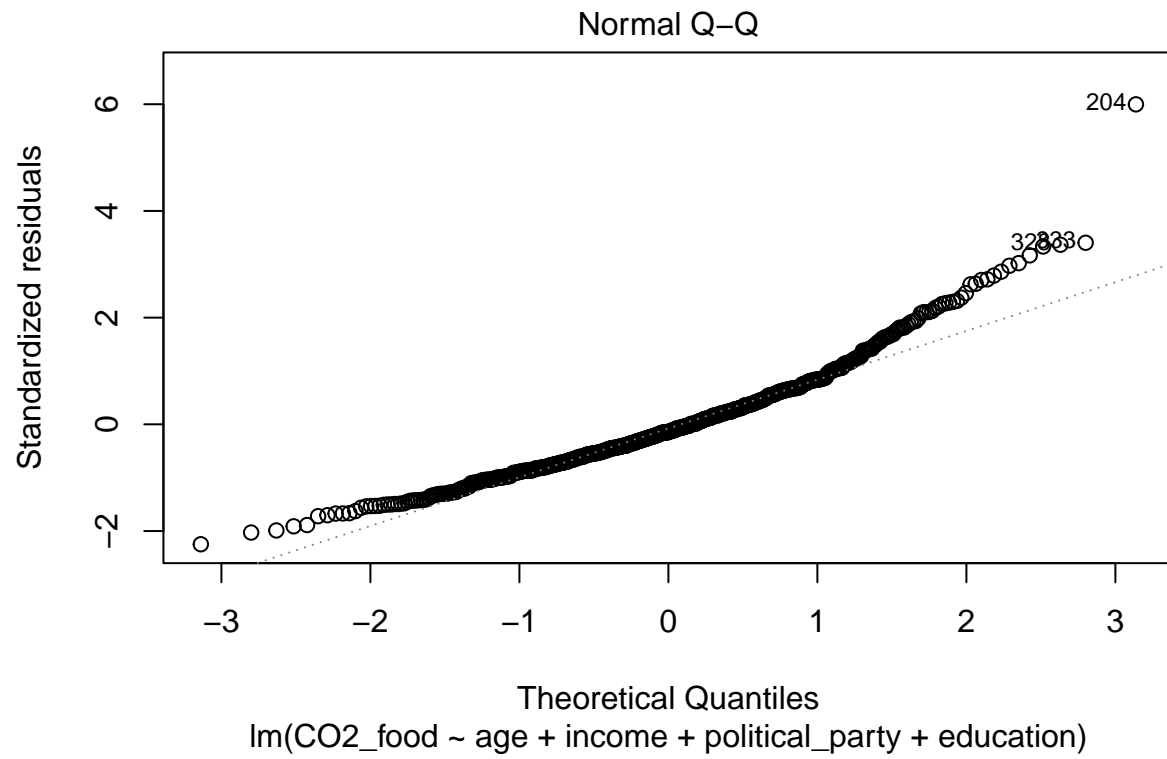
```

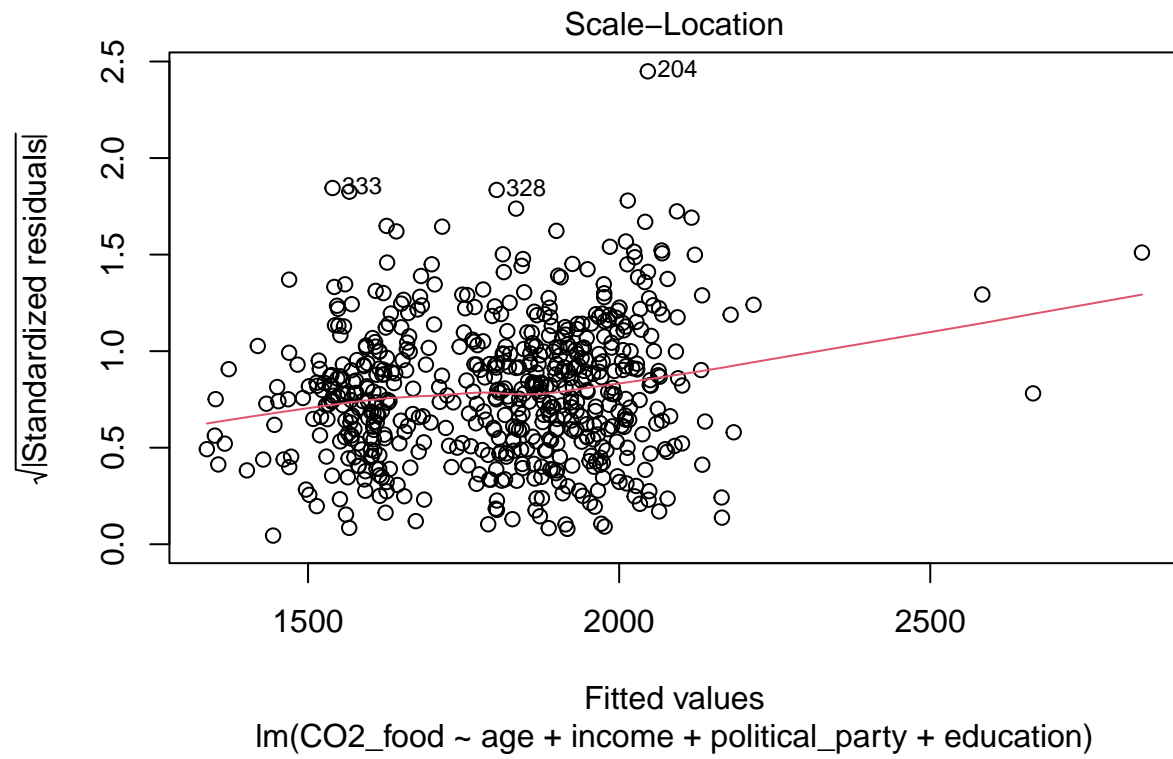
## education(Noch) kein Abschluss 0.001310
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 0.387049
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule 0.372277
## educationDoktorgrad oder Habilitation 0.957938
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss 0.458966
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss 0.222084
##
## (Intercept) ***
## age *
## income
## political_partyAfD ***
## political_partyBündnis Sarah Wagenknecht *
## political_partyCDU/CSU ***
## political_partyDie Linke
## political_partyEiner anderen Partei ***
## political_partyFDP ***
## political_partyKeine Angabe
## political_partySPD ***
## education(Noch) kein Abschluss **
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule
## educationDoktorgrad oder Habilitation
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 530.2 on 571 degrees of freedom
## Multiple R-squared:  0.122, Adjusted R-squared:  0.0974
## F-statistic: 4.959 on 16 and 571 DF, p-value: 1.207e-09

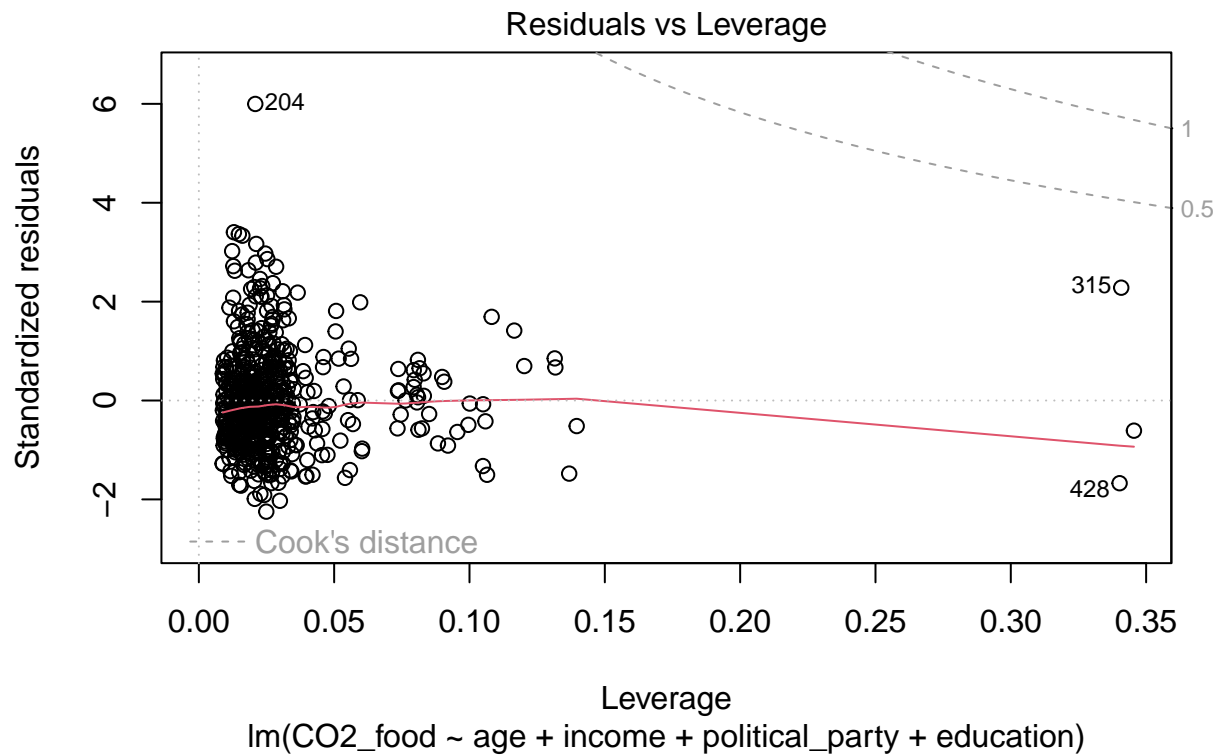
```

```
plot(step_model1)
```



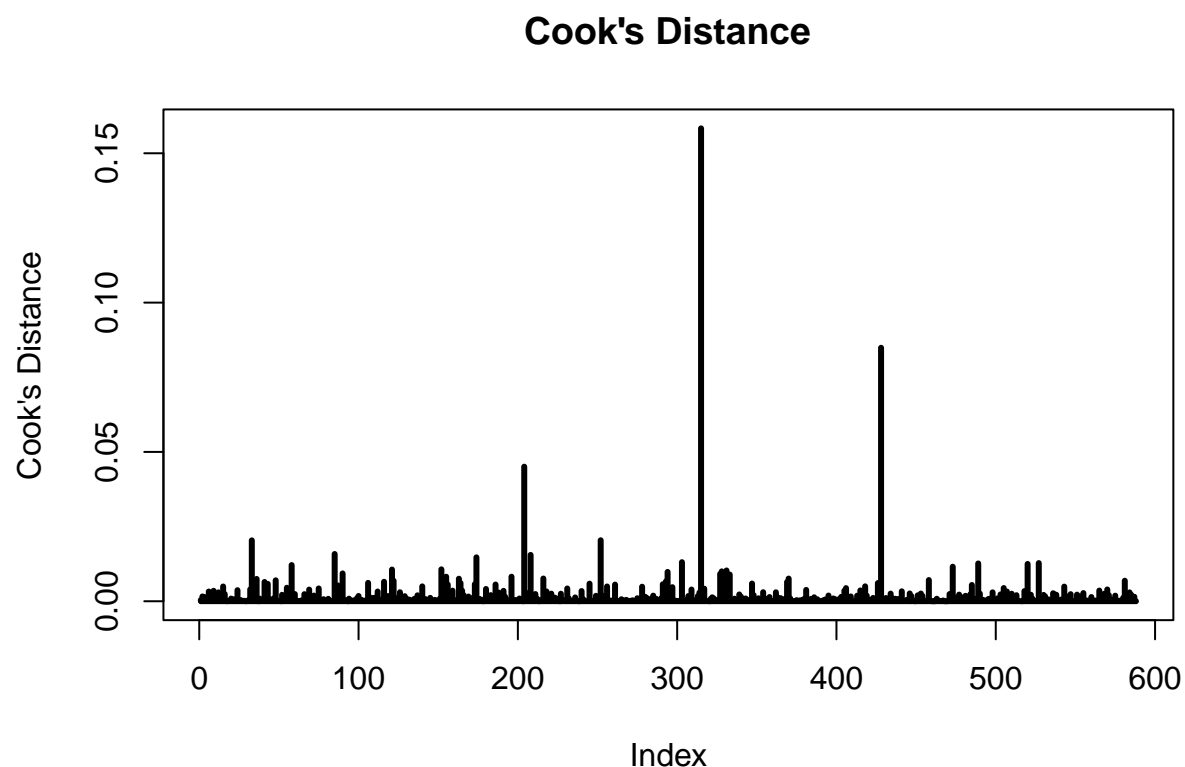






Checking outliers: estimate of the influence of data point; summary of how much a regression model changes when a point is removed

```
cook = cooks.distance(step_model1)
plot(cook,
     type="h",
     lwd=3,
     ylab = "Cook's Distance",
     main="Cook's Distance")
abline(h = 1)
```



```
influential = cooks.distance(step_model1)[which(cook > 1)]
influential
```

```
## named numeric(0)
```

```
influential = influential[!is.na(influential)]
influential_vector = c(as.numeric(rownames(data.frame(influential))))
```

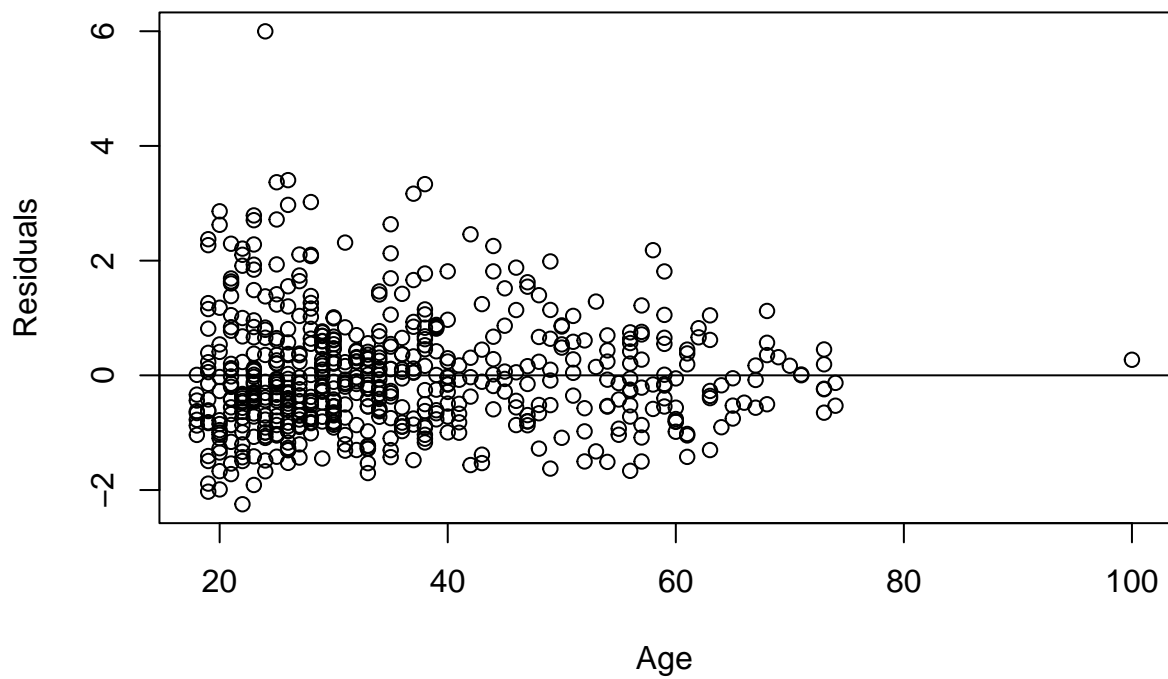
```
df1[influential_vector, ]
```

```
## # A tibble: 0 x 7
## # ... with 7 variables: age <int>, income <dbl>, political_party <fct>,
## #   education <fct>, urban_rural_class <fct>, federal_state <fct>,
## #   CO2_food <dbl>
```

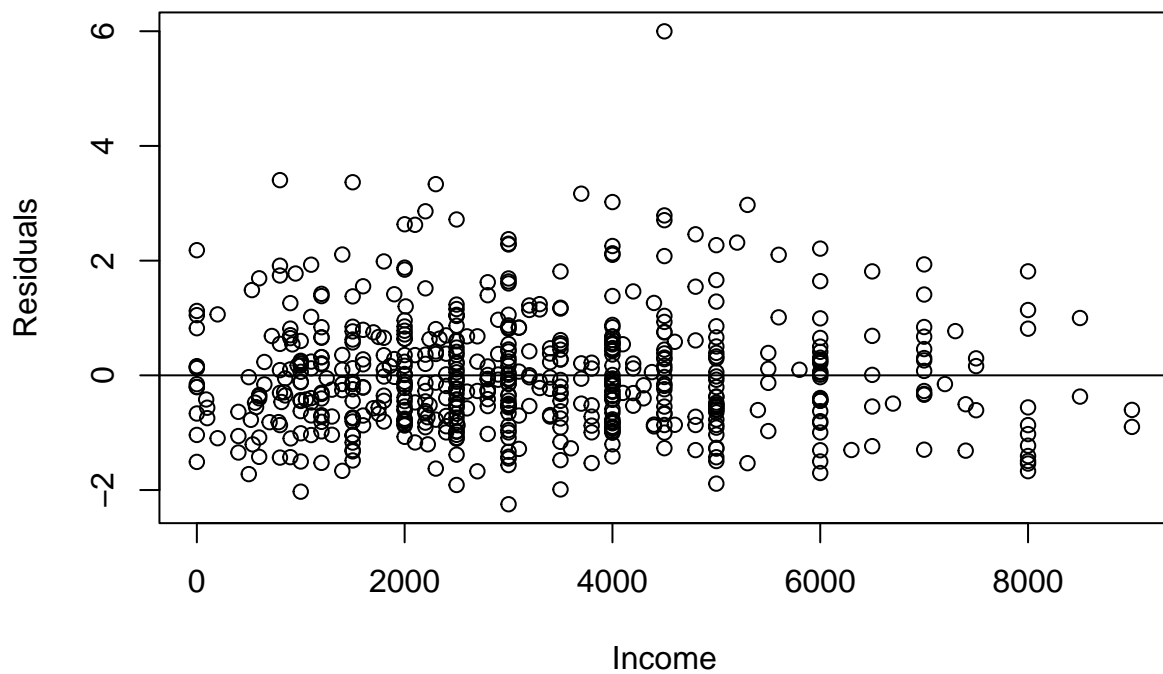
```
res1 = stdres(step_model1) ## (Standardized) Residuals
```

```
# Linearity assumption/Mean zero assumption
```

```
plot(df1$age, res1, xlab = "Age", ylab = "Residuals")
abline(h = 0)
```

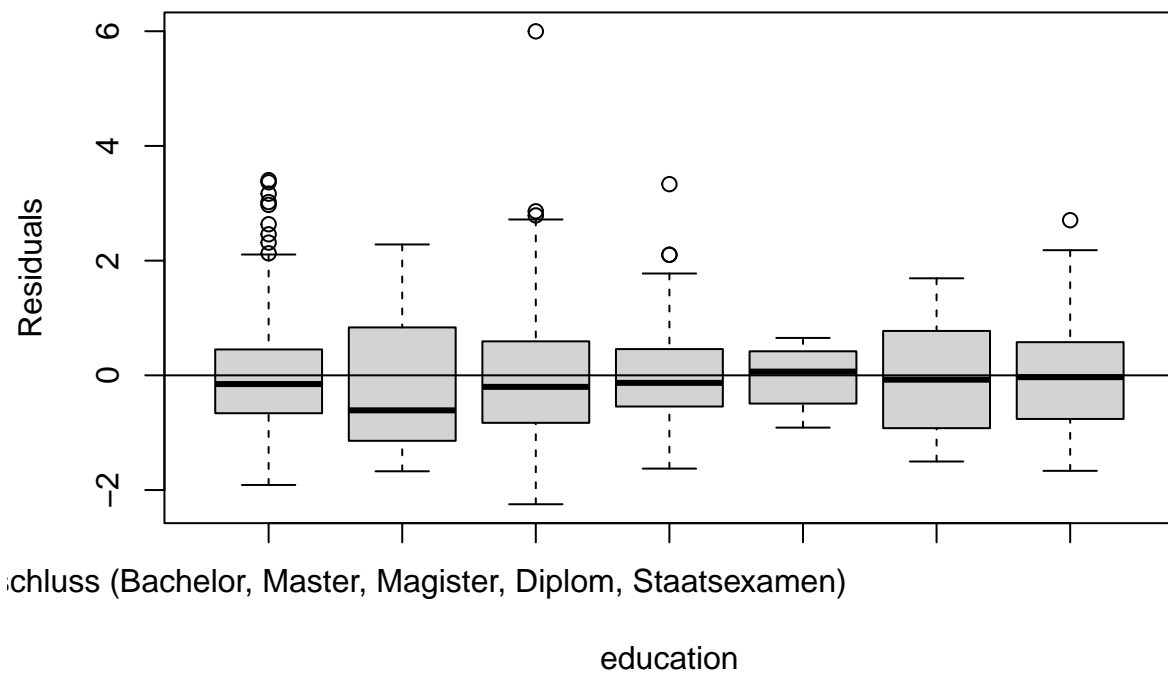


```
plot(df1$income, res1, xlab = "Income", ylab = "Residuals")  
abline(h = 0)
```



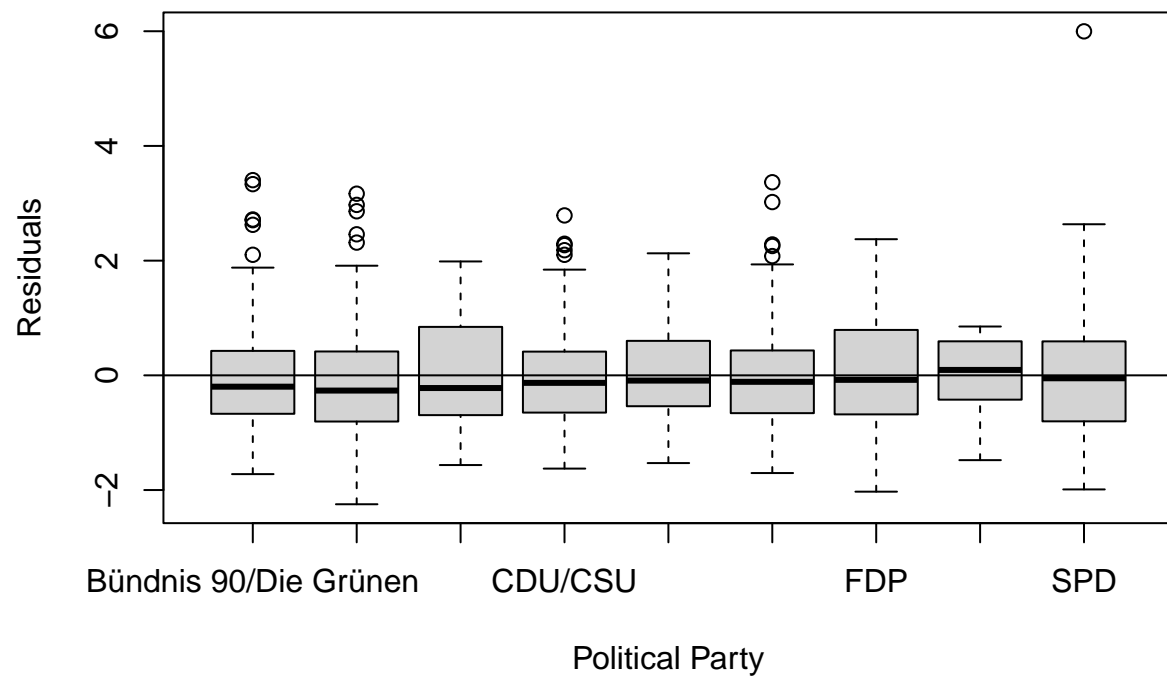
```
#plot(df1$urban_rural_class, res1, xlab = "urban_rural_class", ylab = "Residuals")  
#abline(h = 0)
```

```
plot(df1$education, res1, xlab = "education", ylab = "Residuals")  
abline(h = 0)
```



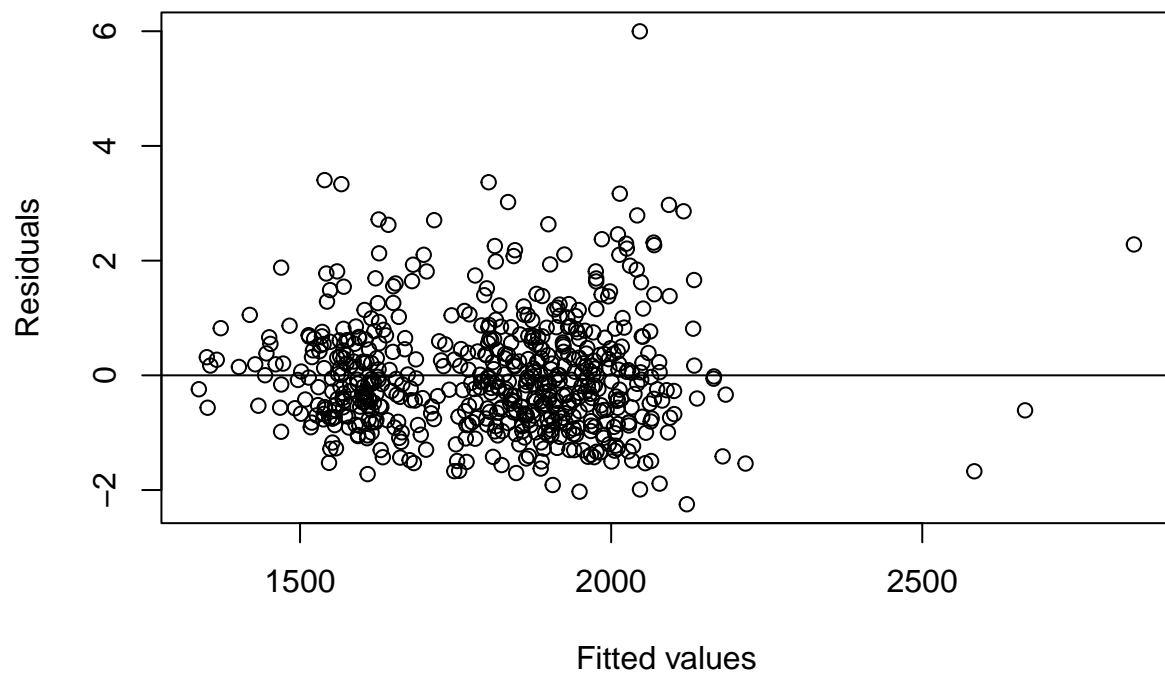
```
#plot(df1$federal_state, res1, xlab = "federal_state", ylab = "Residuals")
#abline(h = 0)

plot(df1$political_party, res1, xlab = "Political Party", ylab = "Residuals")
abline(h = 0)
```



Constant variance and independent error term assumption: clustering observed

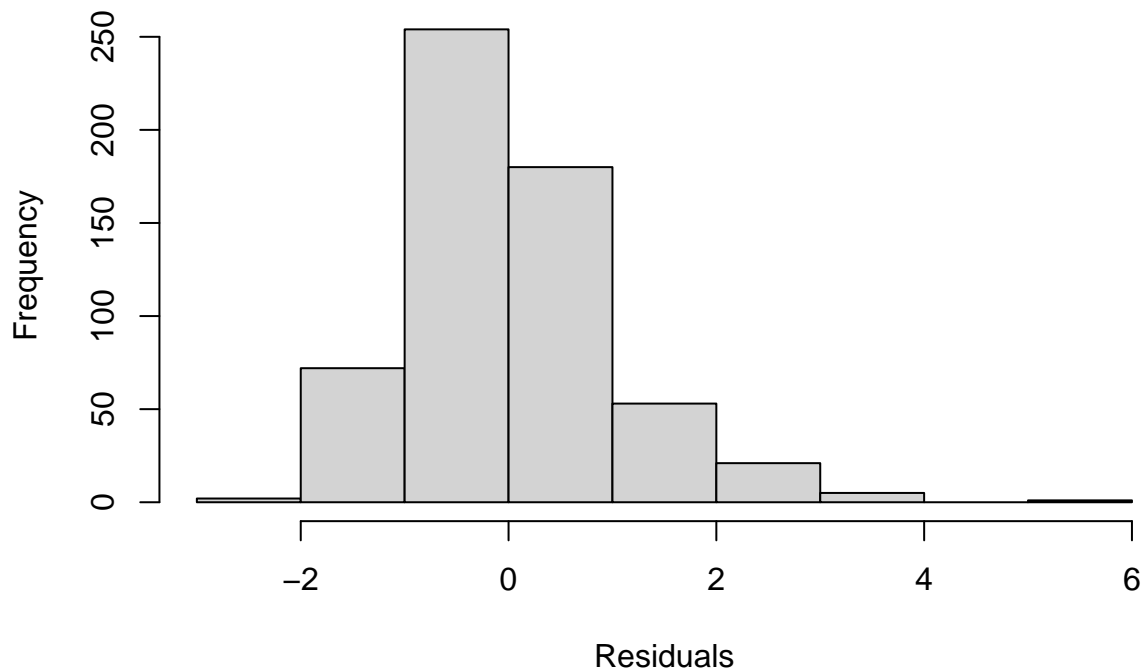
```
plot(fitted(step_model1), res1, xlab = "Fitted values", ylab = "Residuals")
abline(h = 0)
```

```
# Normality assumption
```

```
hist(res1, xlab="Residuals", main= "Histogram of Residuals")
```

Histogram of Residuals

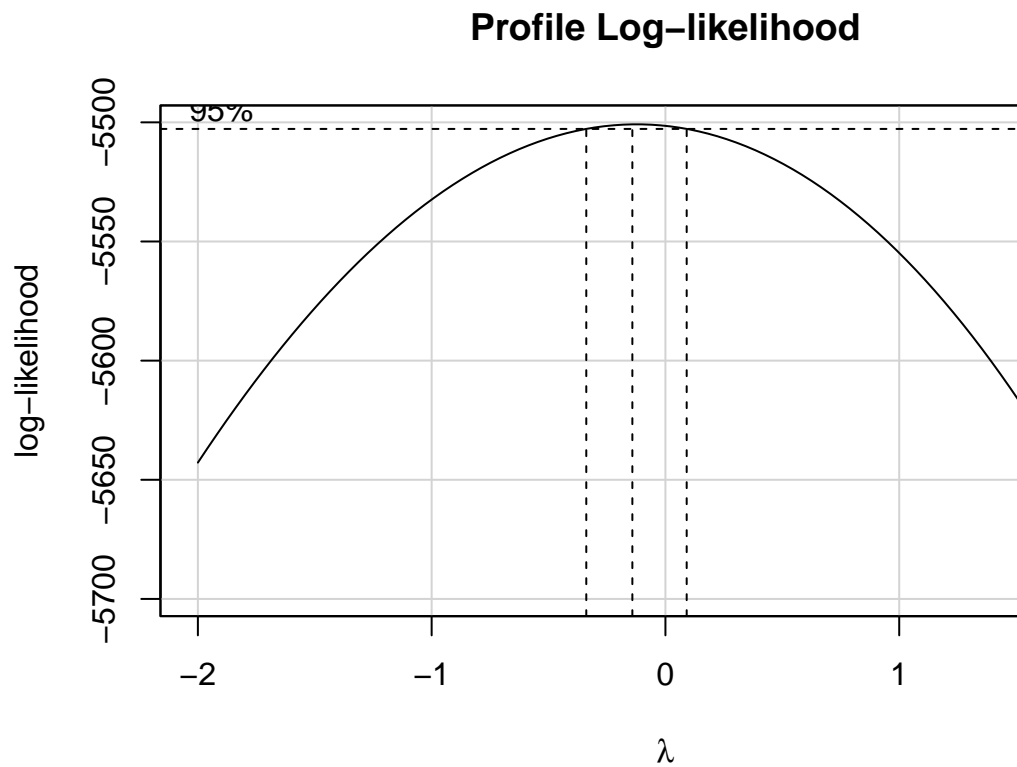


```
## normality test using shapiro-test: reject the H0  
#H0: the sample comes from a normal distribution
```

```
res1_num = res1[is.finite(res1)]  
shapiro.test(res1_num)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: res1_num  
## W = 0.94742, p-value = 1.335e-13
```

```
# box-cox transformation  
bc = boxCox(step_model1)
```



4. Improving the regression fit

```
opt.lambda = bc$x[which.max(bc$y)]
round(opt.lambda/0.5)*0.5 # round it to the nearest 0.5
```

```
## [1] 0
```

```
# regression with non-linear transformation

options(scipen=0, digits=2)

modell_trans = lm(log(CO2_food+1) ~ age + income + political_party + education, data = df1)

summary(modell_trans)
```

FINAL MODEL

```
##
## Call:
## lm(formula = log(CO2_food + 1) ~ age + income + political_party +
##     education, data = df1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```

## -0.7622 -0.1852 -0.0072  0.1840  0.9877
##
## Coefficients:
##
## (Intercept) 7.34e+00
## age -2.05e-03
## income 9.34e-06
## political_partyAfD 2.62e-01
## political_partyBündnis Sarah Wagenknecht 1.58e-01
## political_partyCDU/CSU 2.18e-01
## political_partyDie Linke 5.51e-02
## political_partyEiner anderen Partei 1.57e-01
## political_partyFDP 1.90e-01
## political_partyKeine Angabe 9.90e-03
## political_partySPD 2.18e-01
## education(Noch) kein Abschluss 4.62e-01
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 1.34e-02
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule 3.70e-02
## educationDoktorgrad oder Habilitation 3.71e-02
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss 7.01e-02
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss 4.99e-02
## Std. Error
## (Intercept) 4.70e-02
## age 9.45e-04
## income 6.27e-06
## political_partyAfD 4.48e-02
## political_partyBündnis Sarah Wagenknecht 6.36e-02
## political_partyCDU/CSU 4.06e-02
## political_partyDie Linke 4.86e-02
## political_partyEiner anderen Partei 3.59e-02
## political_partyFDP 4.73e-02
## political_partyKeine Angabe 8.04e-02
## political_partySPD 4.12e-02
## education(Noch) kein Abschluss 1.64e-01
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 3.16e-02
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule 3.24e-02
## educationDoktorgrad oder Habilitation 8.00e-02
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss 9.01e-02
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss 4.14e-02
## t value
## (Intercept) 156.33
## age -2.17
## income 1.49
## political_partyAfD 5.84
## political_partyBündnis Sarah Wagenknecht 2.49
## political_partyCDU/CSU 5.37
## political_partyDie Linke 1.13
## political_partyEiner anderen Partei 4.38
## political_partyFDP 4.01
## political_partyKeine Angabe 0.12
## political_partySPD 5.28
## education(Noch) kein Abschluss 2.82
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 0.42
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule 1.14

```

```

## educationDoktorgrad oder Habilitation 0.46
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss 0.78
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss 1.20
## Pr(>|t|)
## (Intercept) < 2e-16
## age 0.0303
## income 0.1372
## political_partyAfD 8.8e-09
## political_partyBündnis Sarah Wagenknecht 0.0131
## political_partyCDU/CSU 1.1e-07
## political_partyDie Linke 0.2573
## political_partyEiner anderen Partei 1.4e-05
## political_partyFDP 6.9e-05
## political_partyKeine Angabe 0.9020
## political_partySPD 1.8e-07
## education(Noch) kein Abschluss 0.0049
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 0.6724
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule 0.2545
## educationDoktorgrad oder Habilitation 0.6429
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss 0.4367
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss 0.2292
##
## (Intercept) ***
## age *
## income
## political_partyAfD ***
## political_partyBündnis Sarah Wagenknecht *
## political_partyCDU/CSU ***
## political_partyDie Linke
## political_partyEiner anderen Partei ***
## political_partyFDP ***
## political_partyKeine Angabe
## political_partySPD ***
## education(Noch) kein Abschluss **
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule
## educationDoktorgrad oder Habilitation
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.28 on 571 degrees of freedom
## Multiple R-squared:  0.128, Adjusted R-squared:  0.104
## F-statistic: 5.24 on 16 and 571 DF, p-value: 2.37e-10

```

```
# Checking the VIFs for multicollinearity
```

```
vif(model1_trans)
```

```

##          GVIF Df GVIF^(1/(2*Df))
## age          1.3  1          1.1
## income        1.1  1          1.0
## political_party 1.3  8          1.0

```

```
## education      1.4  6      1.0
```

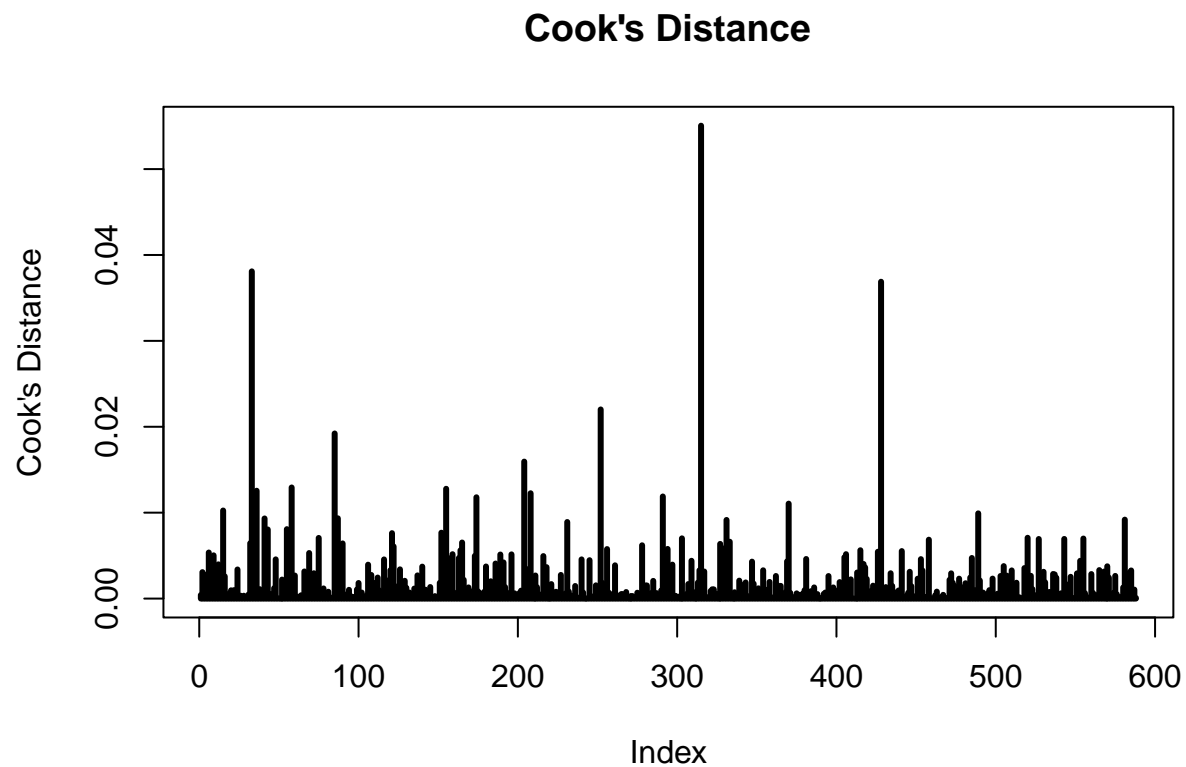
```
# threshold for multicollinearity  
# Calculating the threshold
```

```
max(10, 1/(1-summary(model1_trans)$r.square))
```

```
## [1] 10
```

```
# Checking outliers: estimate of the influence of data point; summary of how much a regression model ch
```

```
cook = cooks.distance(model1_trans)  
plot(cook,  
      type="h",  
      lwd=3,  
      ylab = "Cook's Distance",  
      main="Cook's Distance")  
abline(h = 1)
```

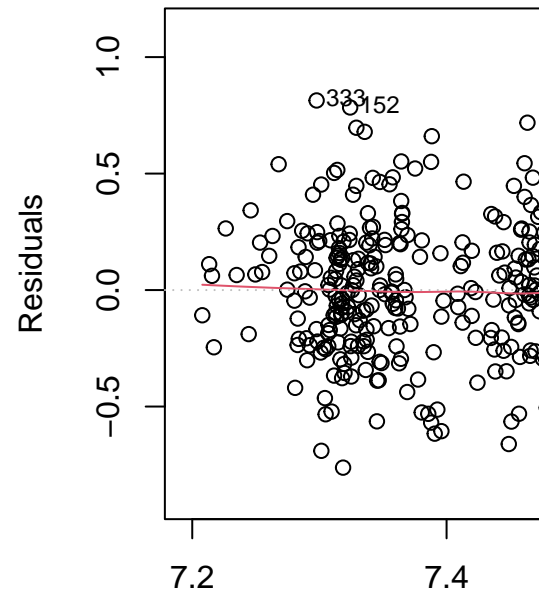


```
influential = cooks.distance(model1_trans)[which(cook >1)]
```

```
influential
```

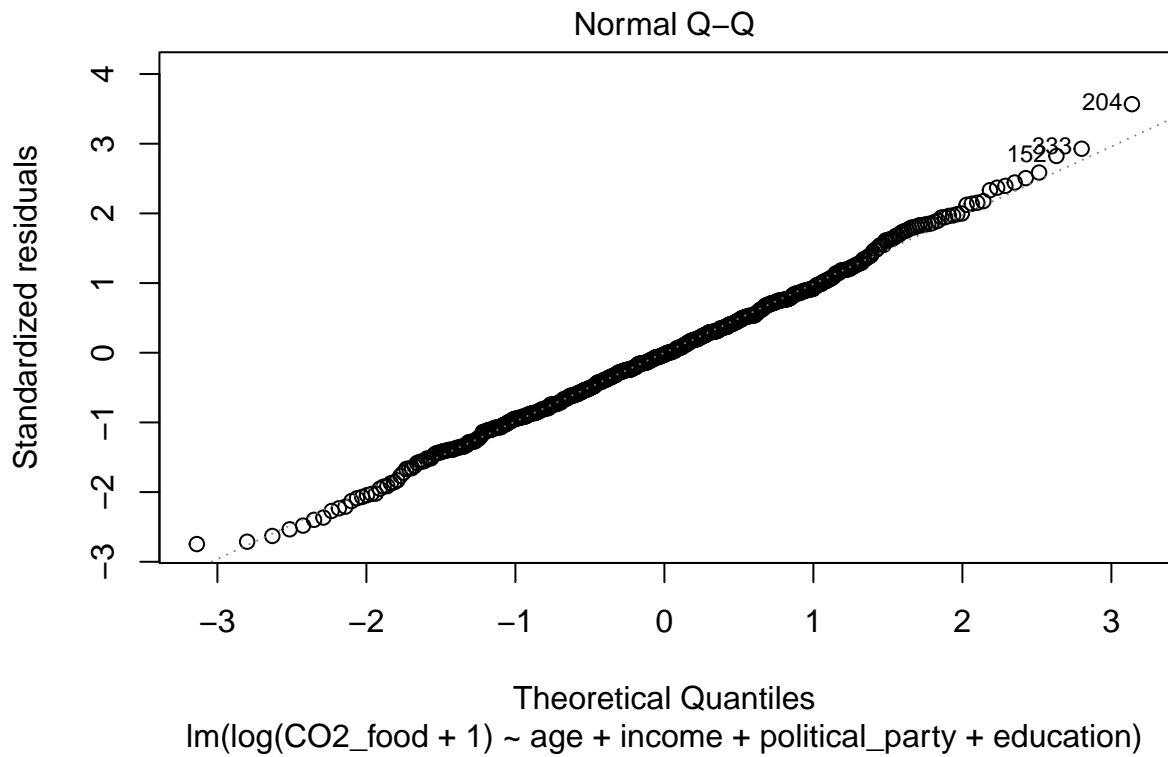
```
## named numeric(0)
```

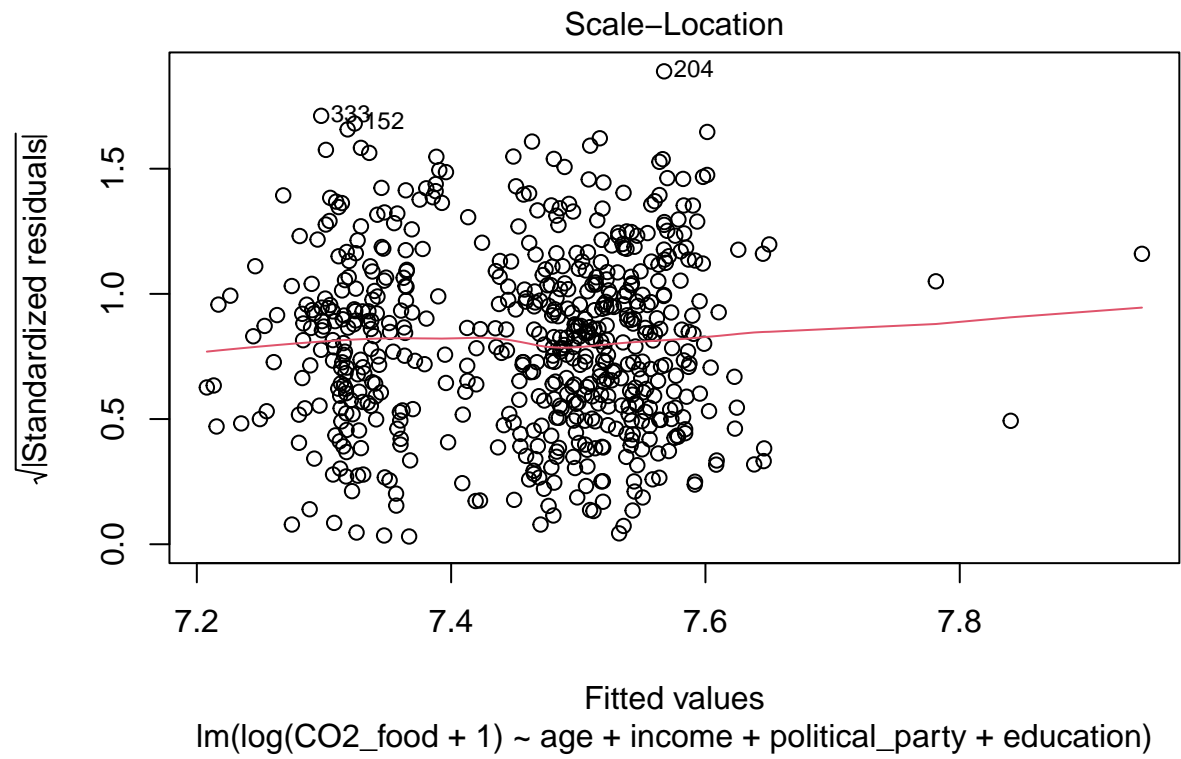
```
plot(model1_trans)
```

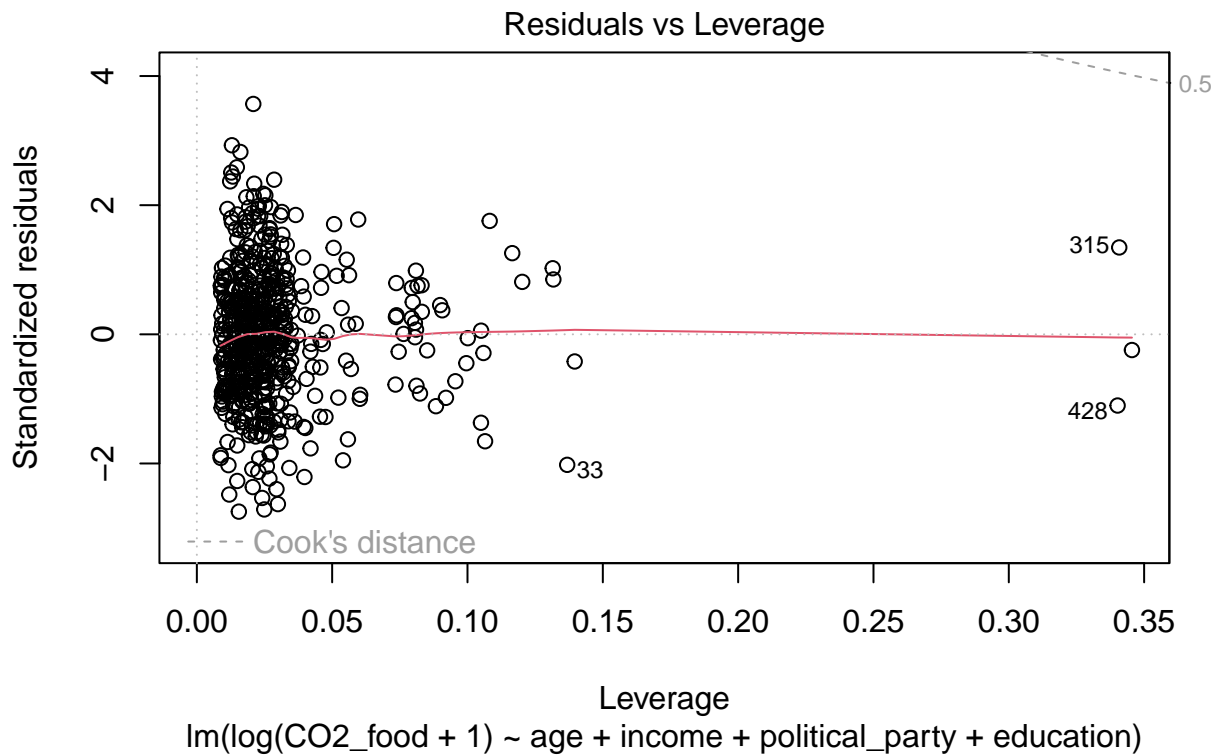


5. Assumptions check in the residuals of the transformed regression

$\ln(\log(\text{CO2_food} + 1)) \sim \text{age} + \text{income} + \text{political_party} + \text{education}$



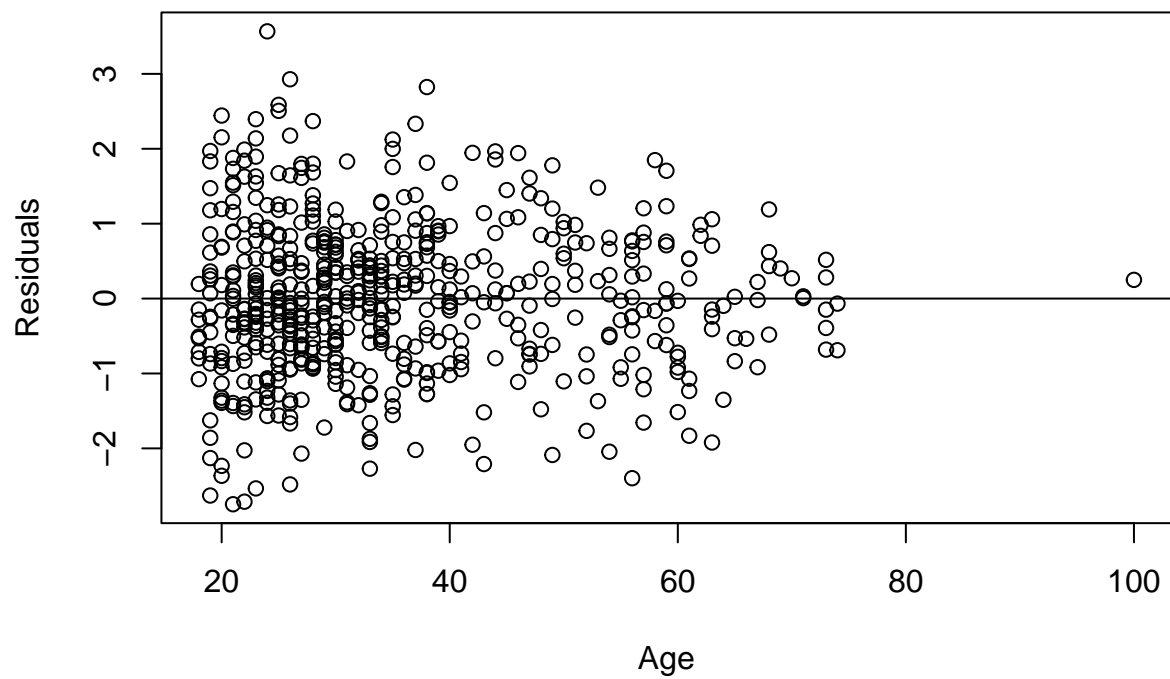




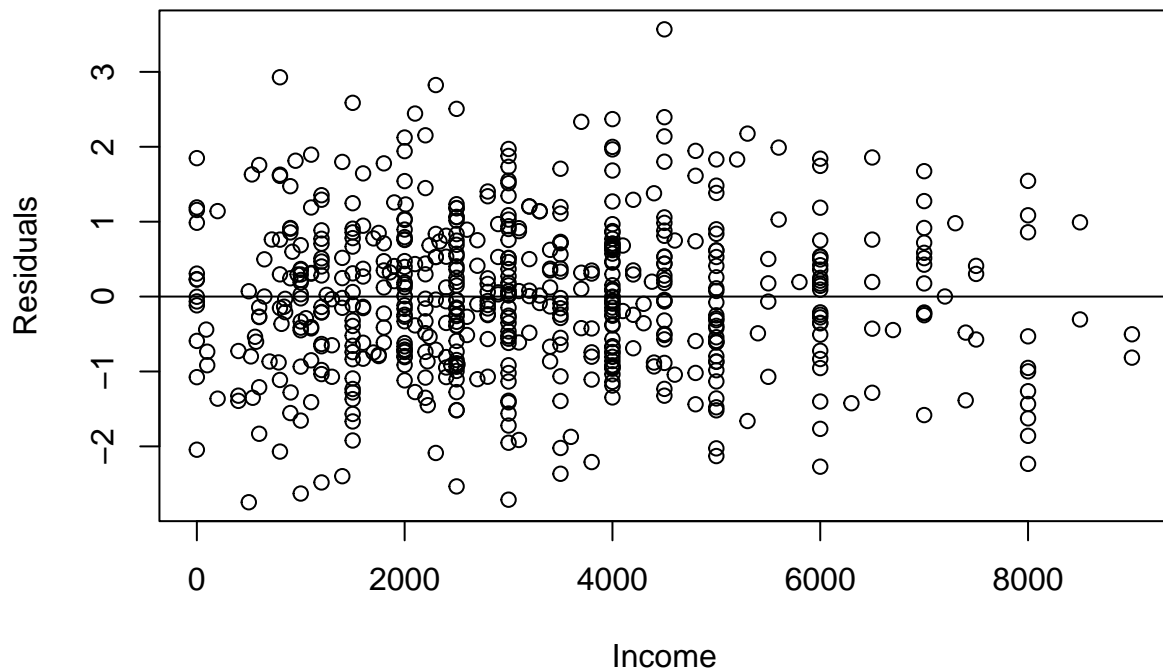
```
res1 = stdres(model1_trans) ## (Standardized) Residuals

# Linearity assumption/Mean zero assumption

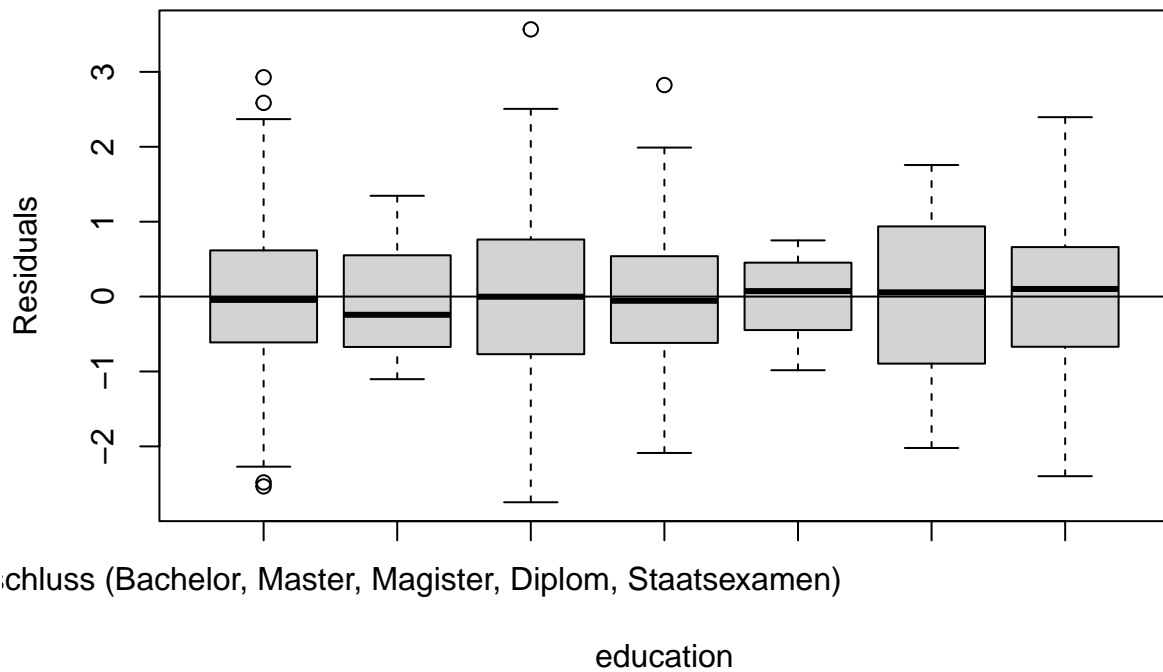
plot(df1$age, res1, xlab = "Age", ylab = "Residuals")
abline(h = 0)
```



```
plot(df1$income, res1, xlab = "Income", ylab = "Residuals")  
abline(h = 0)
```

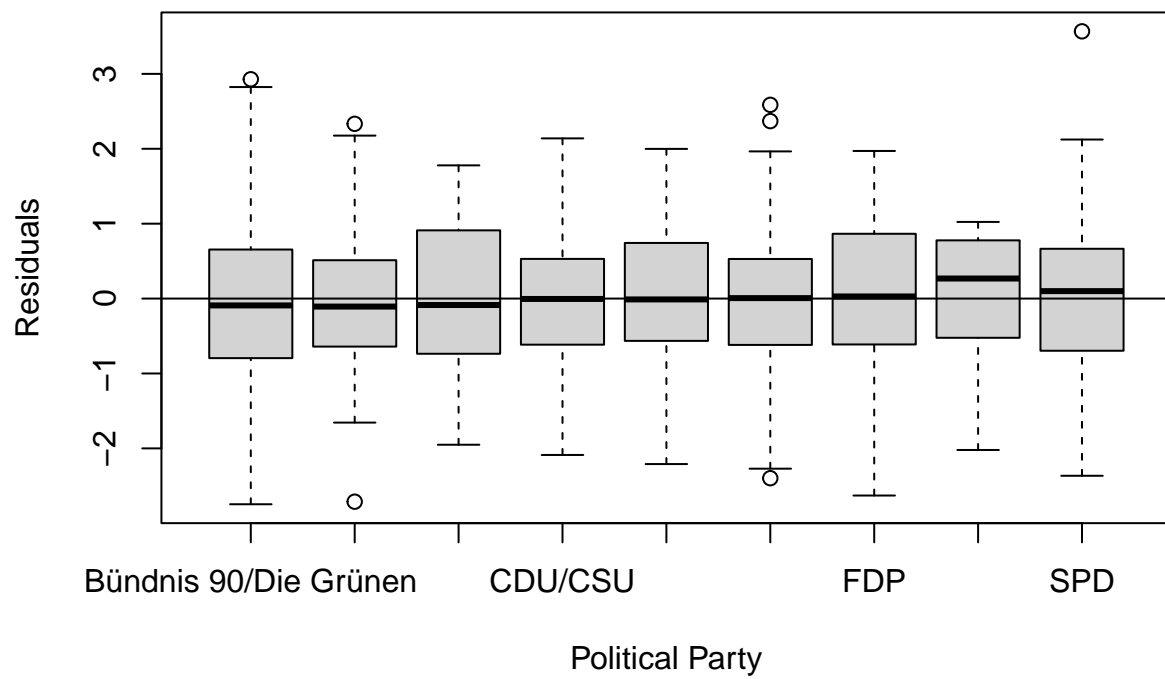


```
#plot(df1$urban_rural_class, res1, xlab = "urban_rural_class", ylab = "Residuals")  
#abline(h = 0)  
  
plot(df1$education, res1, xlab = "education", ylab = "Residuals")  
abline(h = 0)
```



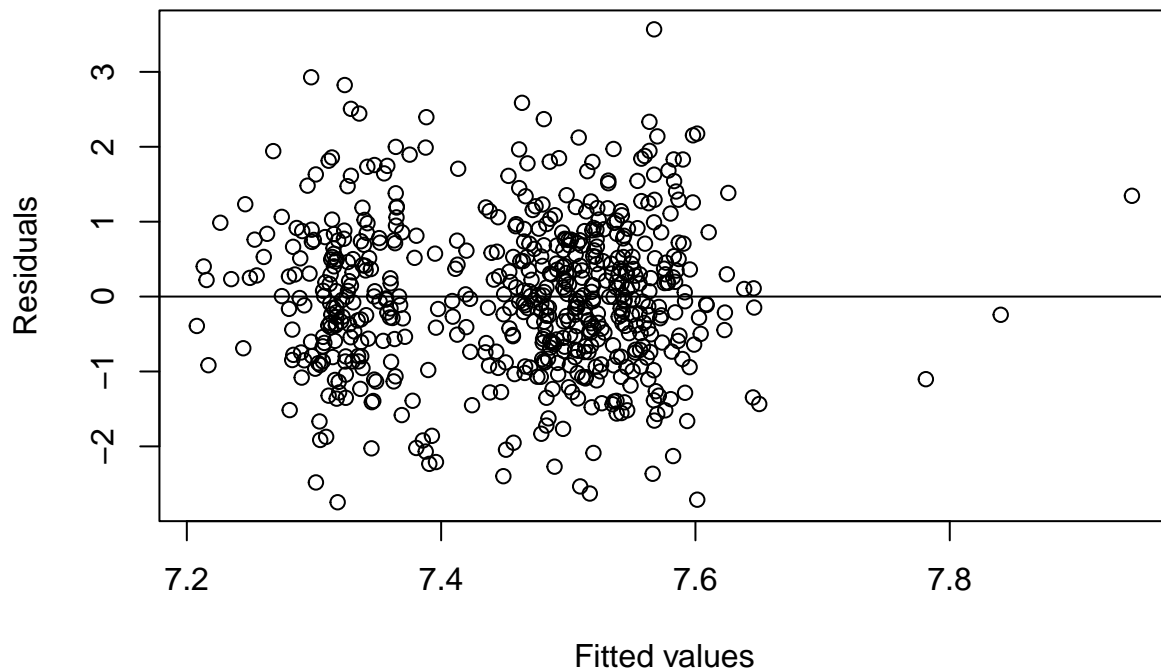
```
#plot(df1$federal_state, res1, xlab = "federal_state", ylab = "Residuals")
#abline(h = 0)

plot(df1$political_party, res1, xlab = "Political Party", ylab = "Residuals")
abline(h = 0)
```



Constant variance and independent error term assumption

```
plot(fitted(model1_trans), res1, xlab = "Fitted values", ylab = "Residuals")
abline(h = 0)
```



```
# Durbin-Watson Test: Independence of the error terms
# H0 (null hypothesis): There is no correlation among the residuals
```

```
durbinWatsonTest(model1_trans)
```

```
## lag Autocorrelation D-W Statistic p-value
## 1 0.13 1.7 0
## Alternative hypothesis: rho != 0
```

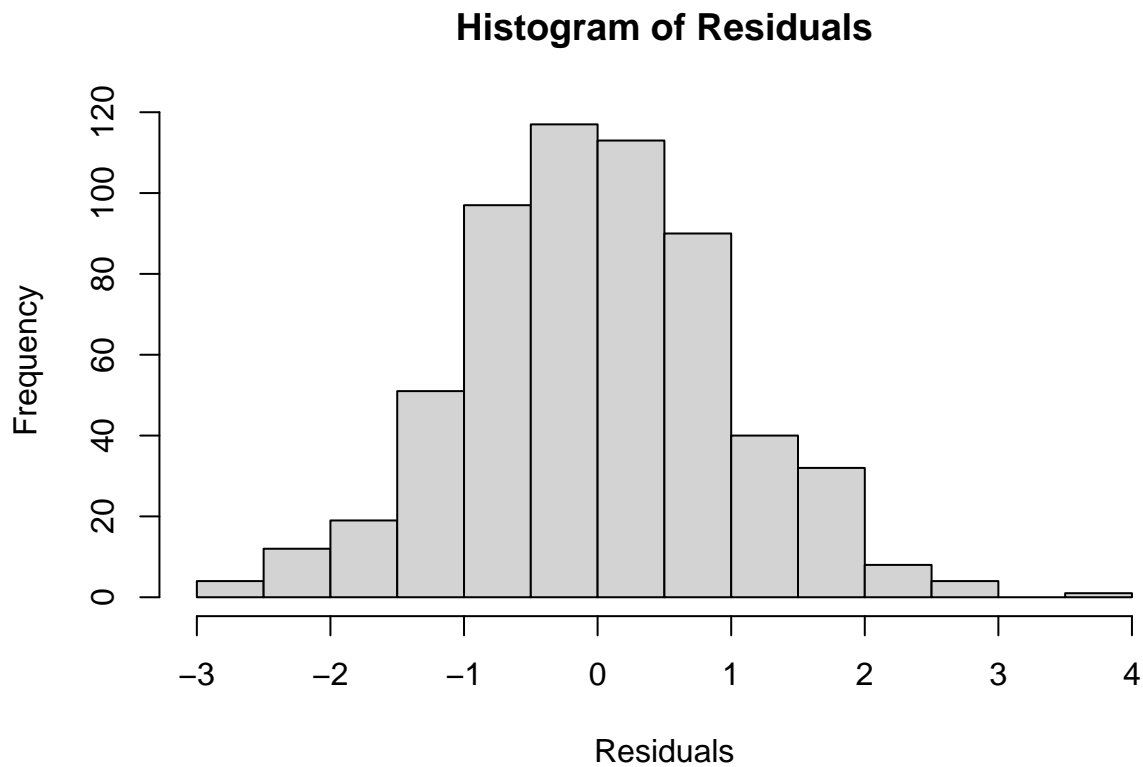
```
# Breusch-Pagan Test: Heteroscedasticity
# H0: Homoscedasticity is present
```

```
library(lmtest)
bptest(model1_trans)
```

```
##
## studentized Breusch-Pagan test
##
## data: model1_trans
## BP = 27, df = 16, p-value = 0.04
```

```
# Normality assumption
```

```
hist(res1, xlab="Residuals", main= "Histogram of Residuals")
```

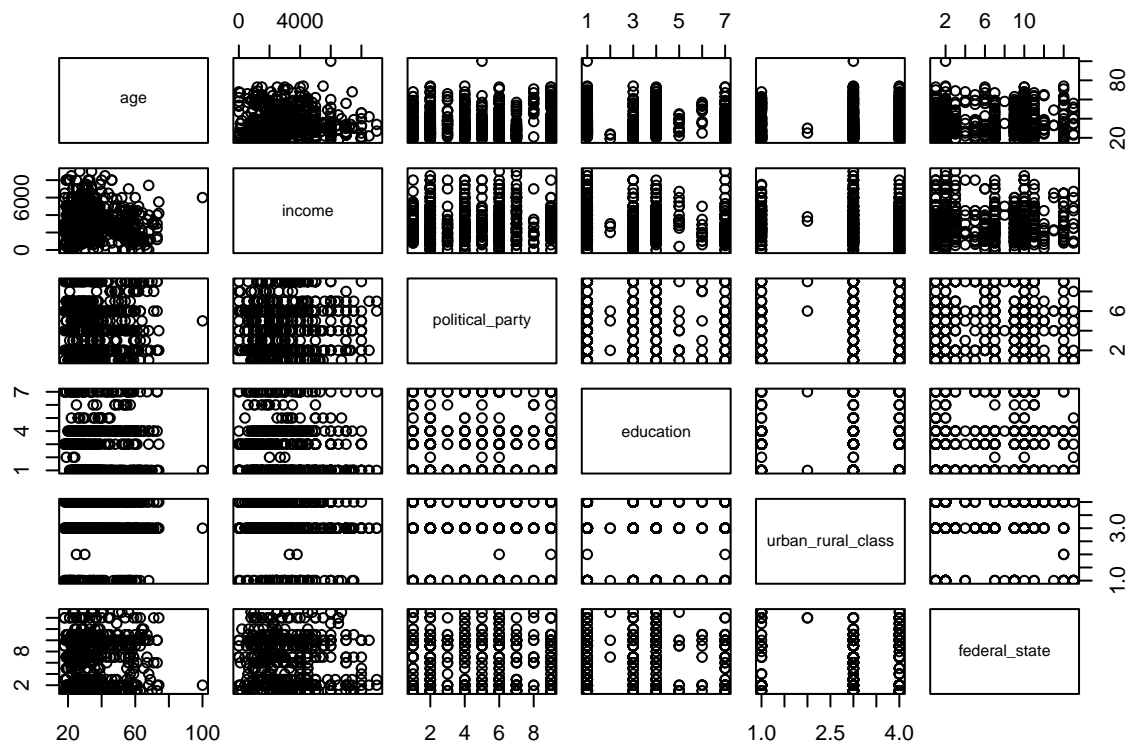


```
## normality test using shapiro-test: reject the H0  
#H0: the sample comes from a normal distribution  
  
res1_num = res1[is.finite(res1)]  
shapiro.test(res1_num)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: res1_num  
## W = 1, p-value = 0.5
```

III. Multivariate Regression: belief diff food

```
# Checking the possible correlation in the data  
  
plot(df2[1:6])
```

1. Modeling

```
## defining a reference level
```

```
df2$political_party <- relevel(df2$political_party, ref='Bündnis 90/Die Grünen')
df2$education <- relevel(df2$education, ref='(Fach-) Hochschulabschluss (Bachelor, Master, Magister, D
df2$urban_rural_class <- relevel(df2$urban_rural_class, ref='sehr zentral')
df2$federal_state <- relevel(df2$federal_state, ref='Nordrhein-Westfalen')
```

```
nrow(df2)
```

```
## [1] 588
```

```
# regression model
```

```
options(scipen = 0, digits=2)
```

```
model2 <- lm(belief_diff_food ~ age + income + political_party + education + urban_rural_class + feder
summary(model2)
```

```
##
```

```
## Call:
```

```
## lm(formula = belief_diff_food ~ age + income + political_party +
##     education + urban_rural_class + federal_state, data = df2)
```

```
##
```

```

## Residuals:
##      Min       1Q   Median       3Q      Max
## -87.63 -19.85   0.21  21.05  80.03
##
## Coefficients:
##                                     Estimate
## (Intercept)                        3.97e+00
## age                                1.27e-01
## income                             -7.52e-04
## political_partyAfD                 -1.73e+01
## political_partyBündnis Sarah Wagenknecht -1.12e+01
## political_partyCDU/CSU              -1.14e+01
## political_partyDie Linke            -8.53e-01
## political_partyEiner anderen Partei  -9.61e+00
## political_partyFDP                  -9.24e+00
## political_partyKeine Angabe         1.15e+01
## political_partySPD                  -1.77e+01
## education(Noch) kein Abschluss      -3.21e+01
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) -2.87e+00
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule -1.91e+00
## educationDoktorgrad oder Habilitation -3.82e+00
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss -9.60e+00
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss -2.88e+00
## urban_rural_classperipher           6.40e+00
## urban_rural_classsehr peripher      1.04e+01
## urban_rural_classzentral            2.99e+00
## federal_stateBaden-Württemberg      -1.50e+00
## federal_stateBayern                 1.71e+00
## federal_stateBerlin                 4.83e+00
## federal_stateBrandenburg            9.87e+00
## federal_stateBremen                 1.16e+01
## federal_stateHamburg                4.50e+00
## federal_stateHessen                 4.96e-01
## federal_stateMecklenburg-Vorpommern 3.18e+01
## federal_stateNiedersachsen          -8.54e+00
## federal_stateRheinland-Pfalz       -5.09e+00
## federal_stateSaarland               7.10e-01
## federal_stateSachsen-Anhalt         4.06e+01
## federal_stateSchleswig-Holstein     -4.57e+00
## federal_stateThüringen              1.45e+00
##                                     Std. Error
## (Intercept)                        5.84e+00
## age                                1.03e-01
## income                             6.91e-04
## political_partyAfD                 4.95e+00
## political_partyBündnis Sarah Wagenknecht 6.93e+00
## political_partyCDU/CSU              4.44e+00
## political_partyDie Linke            5.36e+00
## political_partyEiner anderen Partei  3.97e+00
## political_partyFDP                  5.15e+00
## political_partyKeine Angabe         8.91e+00
## political_partySPD                  4.52e+00
## education(Noch) kein Abschluss      1.81e+01
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS) 3.52e+00

```

## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	3.59e+00
## educationDoktorgrad oder Habilitation	8.77e+00
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	1.00e+01
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	4.56e+00
## urban_rural_classperipher	4.60e+00
## urban_rural_classsehr peripher	2.28e+01
## urban_rural_classzentral	3.37e+00
## federal_stateBaden-Württemberg	4.38e+00
## federal_stateBayern	4.50e+00
## federal_stateBerlin	5.46e+00
## federal_stateBrandenburg	1.15e+01
## federal_stateBremen	8.37e+00
## federal_stateHamburg	6.81e+00
## federal_stateHessen	5.23e+00
## federal_stateMecklenburg-Vorpommern	2.19e+01
## federal_stateNiedersachsen	5.32e+00
## federal_stateRheinland-Pfalz	6.56e+00
## federal_stateSaarland	1.03e+01
## federal_stateSachsen-Anhalt	1.59e+01
## federal_stateSchleswig-Holstein	7.67e+00
## federal_stateThüringen	1.16e+01
##	t value
## (Intercept)	0.68
## age	1.23
## income	-1.09
## political_partyAfD	-3.50
## political_partyBündnis Sarah Wagenknecht	-1.61
## political_partyCDU/CSU	-2.57
## political_partyDie Linke	-0.16
## political_partyEiner anderen Partei	-2.42
## political_partyFDP	-1.80
## political_partyKeine Angabe	1.29
## political_partySPD	-3.91
## education(Noch) kein Abschluss	-1.77
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	-0.82
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	-0.53
## educationDoktorgrad oder Habilitation	-0.44
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	-0.96
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	-0.63
## urban_rural_classperipher	1.39
## urban_rural_classsehr peripher	0.46
## urban_rural_classzentral	0.89
## federal_stateBaden-Württemberg	-0.34
## federal_stateBayern	0.38
## federal_stateBerlin	0.88
## federal_stateBrandenburg	0.86
## federal_stateBremen	1.38
## federal_stateHamburg	0.66
## federal_stateHessen	0.09
## federal_stateMecklenburg-Vorpommern	1.45
## federal_stateNiedersachsen	-1.61
## federal_stateRheinland-Pfalz	-0.78
## federal_stateSaarland	0.07
## federal_stateSachsen-Anhalt	2.56

## federal_stateSchleswig-Holstein	-0.60
## federal_stateThüringen	0.12
##	Pr(> t)
## (Intercept)	0.4967
## age	0.2204
## income	0.2770
## political_partyAfD	0.0005
## political_partyBündnis Sarah Wagenknecht	0.1083
## political_partyCDU/CSU	0.0105
## political_partyDie Linke	0.8734
## political_partyEiner anderen Partei	0.0159
## political_partyFDP	0.0732
## political_partyKeine Angabe	0.1968
## political_partySPD	0.0001
## education(Noch) kein Abschluss	0.0765
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	0.4151
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	0.5940
## educationDoktorgrad oder Habilitation	0.6635
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	0.3383
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss	0.5283
## urban_rural_classperipher	0.1645
## urban_rural_classsehr peripher	0.6489
## urban_rural_classzentral	0.3763
## federal_stateBaden-Württemberg	0.7319
## federal_stateBayern	0.7040
## federal_stateBerlin	0.3771
## federal_stateBrandenburg	0.3904
## federal_stateBremen	0.1672
## federal_stateHamburg	0.5091
## federal_stateHessen	0.9245
## federal_stateMecklenburg-Vorpommern	0.1474
## federal_stateNiedersachsen	0.1086
## federal_stateRheinland-Pfalz	0.4383
## federal_stateSaarland	0.9451
## federal_stateSachsen-Anhalt	0.0107
## federal_stateSchleswig-Holstein	0.5518
## federal_stateThüringen	0.9010
##	
## (Intercept)	
## age	
## income	
## political_partyAfD	***
## political_partyBündnis Sarah Wagenknecht	
## political_partyCDU/CSU	*
## political_partyDie Linke	
## political_partyEiner anderen Partei	*
## political_partyFDP	.
## political_partyKeine Angabe	
## political_partySPD	***
## education(Noch) kein Abschluss	.
## educationAllgemeine oder fachgebundene Hochschulreife/Abitur (Gymnasium bzw. EOS)	
## educationBerufsausbildung, Lehre oder Ausbildung an einer Fachschule	
## educationDoktorgrad oder Habilitation	
## educationHauptschulabschluss (Volksschulabschluss) oder gleichwertiger Abschluss	

```
## educationRealschulabschluss (Mittlere Reife) oder gleichwertiger Abschluss
## urban_rural_classperipher
## urban_rural_classsehr peripher
## urban_rural_classzentral
## federal_stateBaden-Württemberg
## federal_stateBayern
## federal_stateBerlin
## federal_stateBrandenburg
## federal_stateBremen
## federal_stateHamburg
## federal_stateHessen
## federal_stateMecklenburg-Vorpommern
## federal_stateNiedersachsen
## federal_stateRheinland-Pfalz
## federal_stateSaarland
## federal_stateSachsen-Anhalt
## federal_stateSchleswig-Holstein
## federal_stateThüringen
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 30 on 554 degrees of freedom
## Multiple R-squared:  0.0945, Adjusted R-squared:  0.0405
## F-statistic: 1.75 on 33 and 554 DF,  p-value: 0.00663
```

```
# Checking the VIFs for multicollinearity
```

```
vif(model2)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## age           1.3  1             1.1
## income        1.1  1             1.0
## political_party 1.8  8             1.0
## education     1.8  6             1.1
## urban_rural_class 2.1  3             1.1
## federal_state  3.0 14             1.0
```

```
# threshold for multicollinearity
```

```
# Calculating the threshold
```

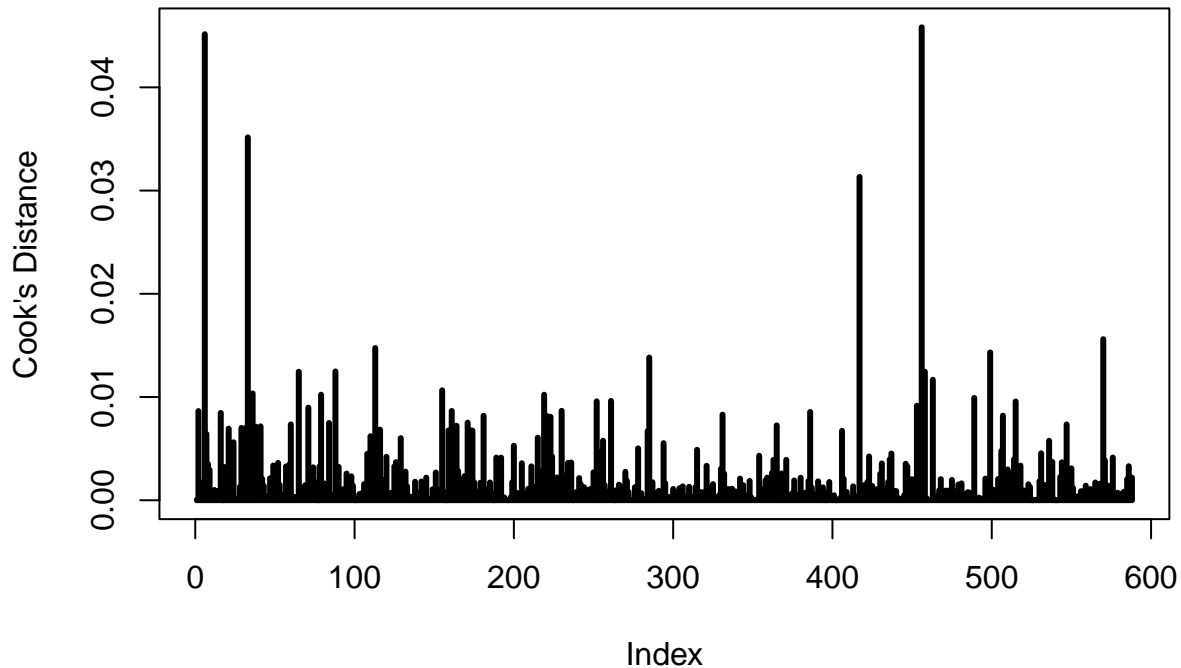
```
max(10, 1/(1-summary(model2)$r.square))
```

```
## [1] 10
```

```
# Checking outliers
```

```
cook = cooks.distance(model2)
plot(cook,
     type="h",
     lwd=3,
     ylab = "Cook's Distance",
     main="Cook's Distance")
abline(h = 1)
```

Cook's Distance



```
influential = cook[(cook > (3 * mean(cook, na.rm = TRUE)))]
influential
```

```
##      2      6      7     16     21     29     32     33     36     38     41
## 0.0086 0.0451 0.0064 0.0085 0.0069 0.0070 0.0070 0.0352 0.0104 0.0071 0.0072
##      60     65     71     79     84     88     110     113     116     129     155
## 0.0074 0.0125 0.0090 0.0102 0.0075 0.0125 0.0062 0.0147 0.0068 0.0060 0.0106
##     159     161     164     171     173     174     181     215     219     221     223
## 0.0068 0.0086 0.0072 0.0075 0.0068 0.0067 0.0082 0.0061 0.0102 0.0082 0.0081
##     230     252     261     284     285     331     365     386     406     417     453
## 0.0087 0.0096 0.0096 0.0067 0.0138 0.0083 0.0073 0.0085 0.0067 0.0313 0.0092
##     456     458     463     489     499     507     515     547     570
## 0.0458 0.0125 0.0117 0.0099 0.0143 0.0082 0.0096 0.0073 0.0156
```

```
influential = influential[!is.na(influential)]
```

```
influential_vector = c(as.numeric(rownames(data.frame(influential))))
```

```
df2[influential_vector, ]
```

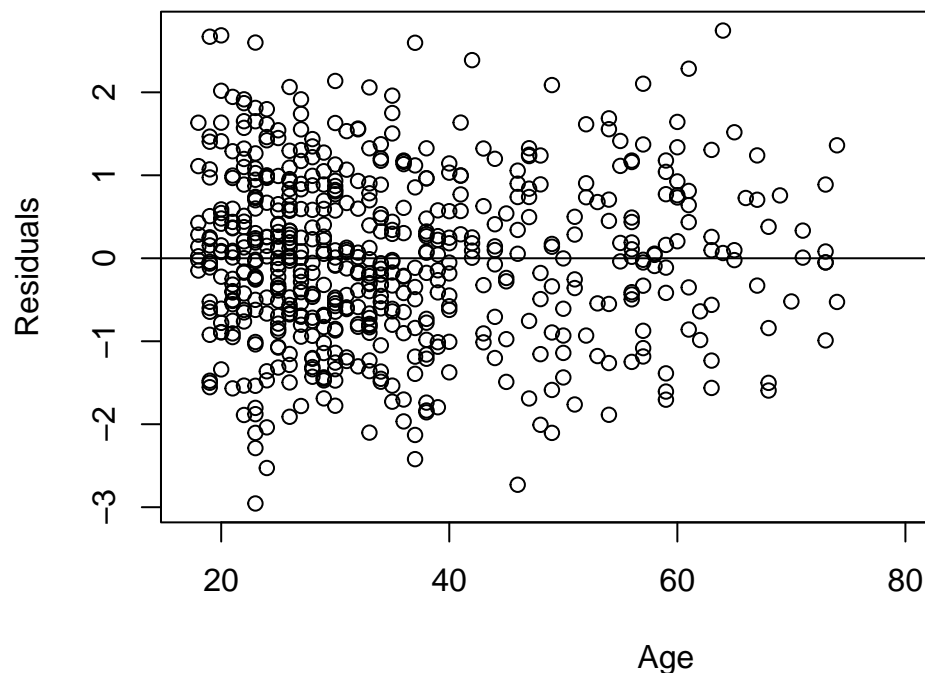
```
## # A tibble: 53 x 7
##   age income political_party education urban~1 feder~2 belie~3
##   <int> <dbl> <fct>          <fct>          <fct> <fct>    <dbl>
## 1    59    800 Keine Angabe Allgemeine oder fachgeb~ sehr z~ Hessen    -26
```

```
## 2 49 2300 CDU/CSU Berufsausbildung, Lehre~ zentral Sachse~ 93
## 3 57 600 CDU/CSU Realschulabschluss (Mit~ zentral Baden~ 60
## 4 33 6500 CDU/CSU Berufsausbildung, Lehre~ zentral Sachse~ 57
## 5 54 2900 AfD Hauptschulabschluss (Vo~ zentral Rheinl~ -56
## 6 59 2000 SPD (Fach-) Hochschulabschl~ zentral Brande~ 38
## 7 48 1750 Keine Angabe Hauptschulabschluss (Vo~ sehr z~ Hessen -21
## 8 37 3500 Keine Angabe Hauptschulabschluss (Vo~ sehr z~ Bayern 82
## 9 19 1000 FDP Allgemeine oder fachgeb~ sehr z~ Bayern 74
## 10 65 1500 CDU/CSU Berufsausbildung, Lehre~ sehr z~ Bremen 53
## # ... with 43 more rows, and abbreviated variable names 1: urban_rural_class,
## # 2: federal_state, 3: belief_diff_food
```

```
res2 = stdres(model2) ## (Standardized) Residuals

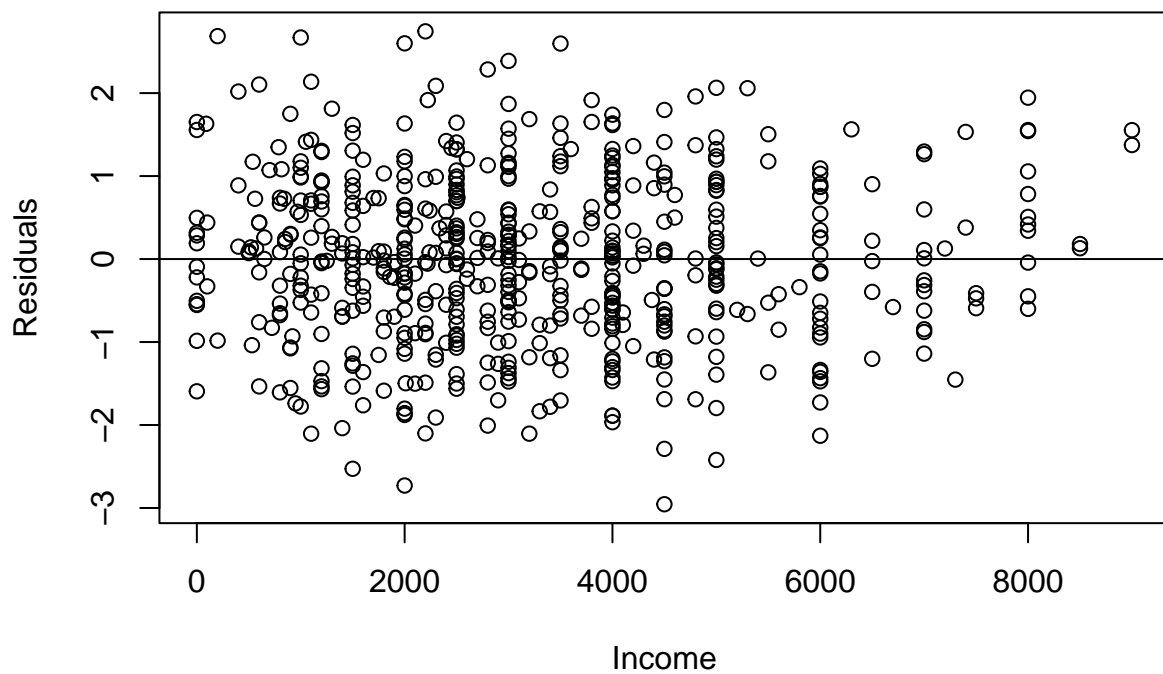
# Linearity assumption/Mean zero assumption

plot(df2$age, res2, xlab = "Age", ylab = "Residuals")
abline(h = 0)
```

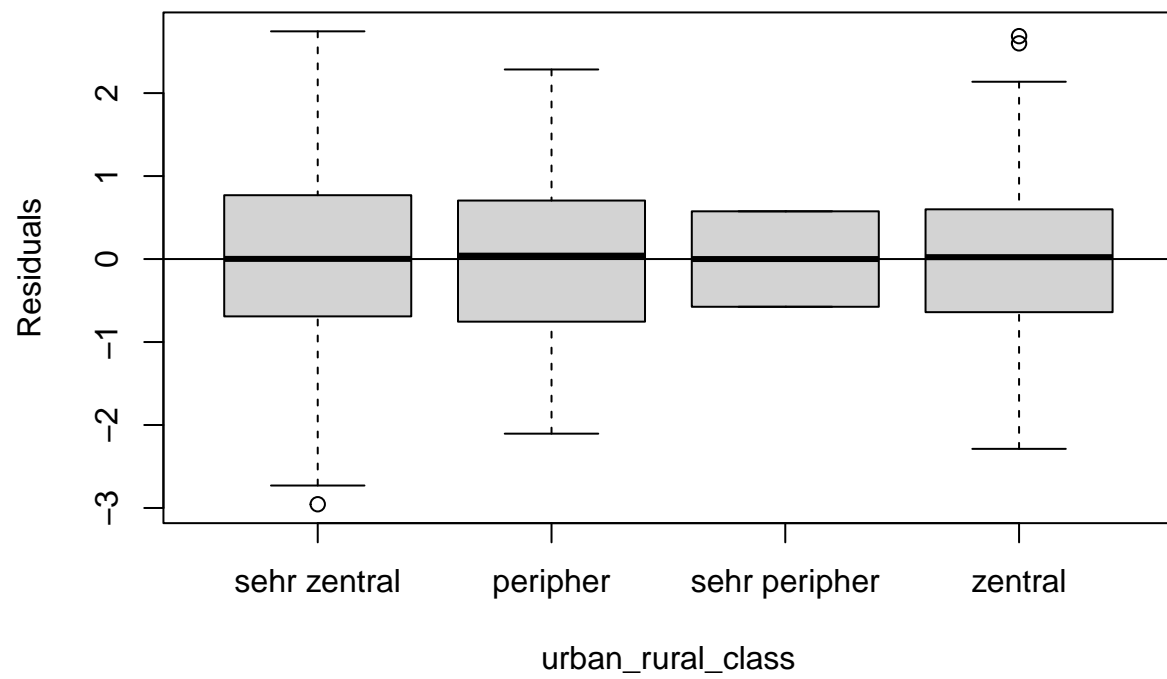


2. Assumptions check in the residuals

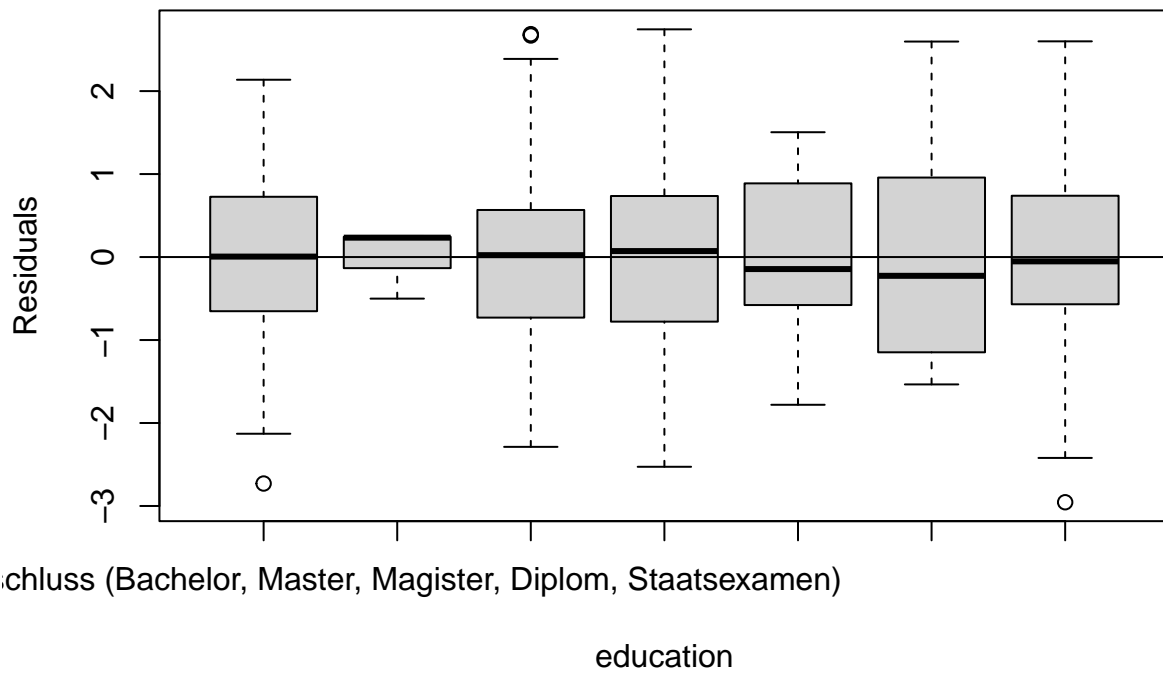
```
plot(df2$income, res2, xlab = "Income", ylab = "Residuals")
abline(h = 0)
```



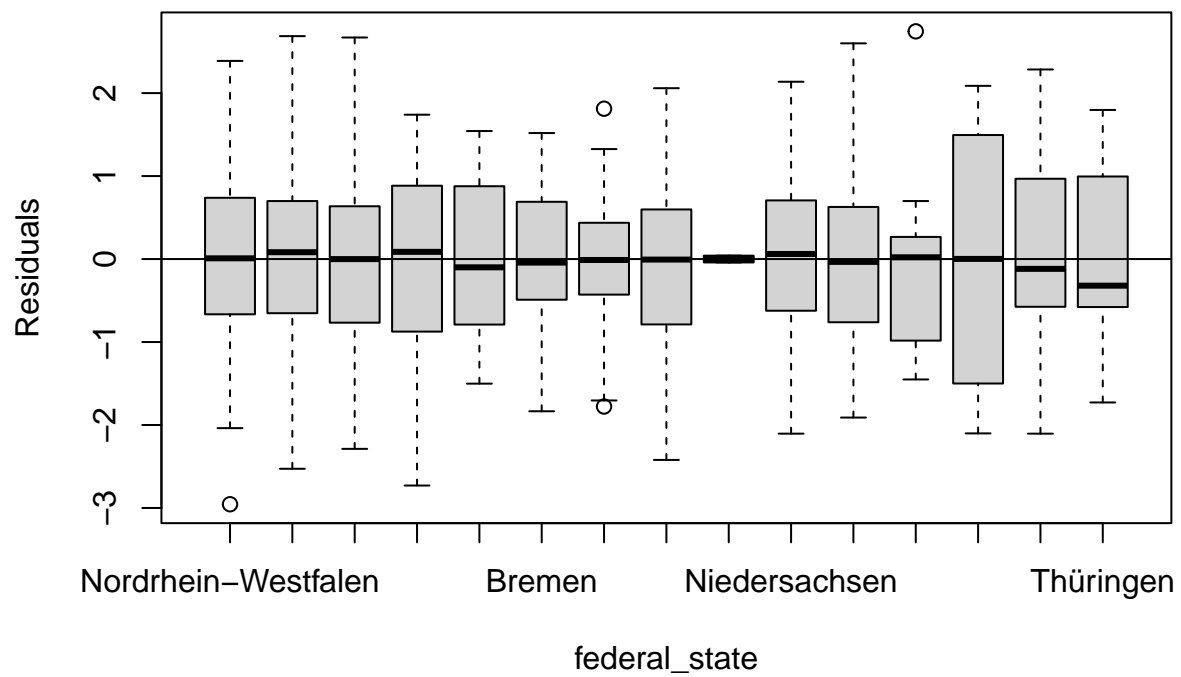
```
plot(df2$urban_rural_class, res2, xlab = "urban_rural_class", ylab = "Residuals")  
abline(h = 0)
```

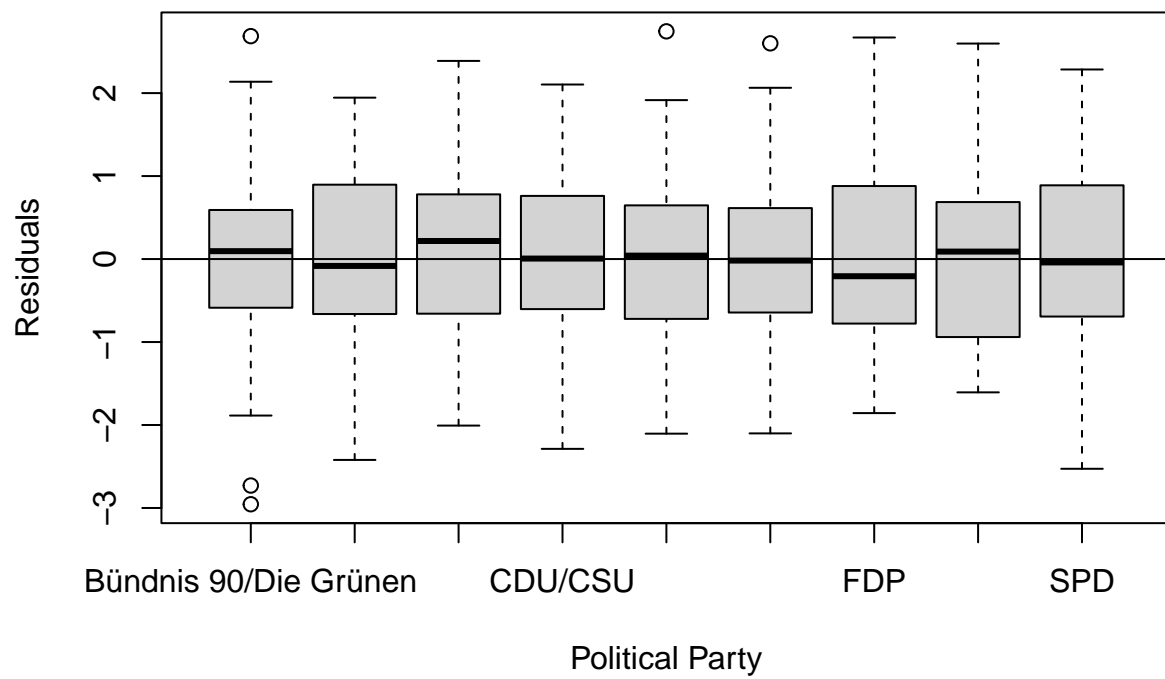
```
plot(df2$education, res2, xlab = "education", ylab = "Residuals")  
abline(h = 0)
```



```
plot(df2$federal_state, res2, xlab = "federal_state", ylab = "Residuals")  
abline(h = 0)
```

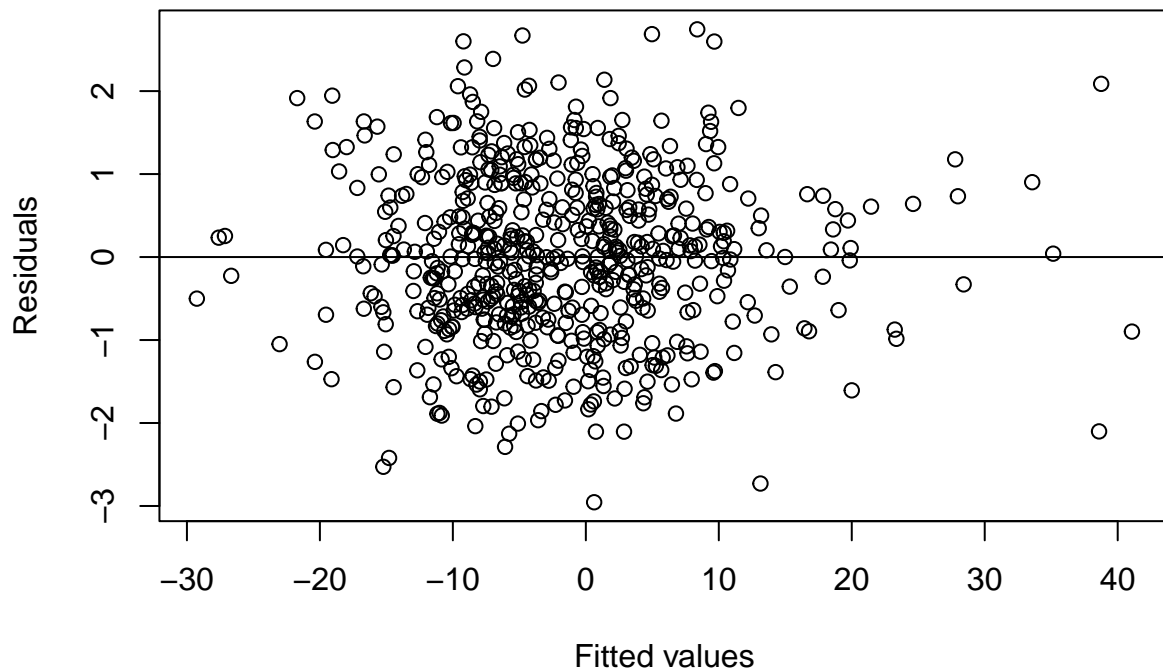


```
plot(df2$political_party, res2, xlab = "Political Party", ylab = "Residuals")
abline(h = 0)
```



Constant variance and independent error term assumption

```
plot(fitted(model2), res2, xlab = "Fitted values", ylab = "Residuals")
abline(h = 0)
```



```
# Durbin-Watson Test: Independence of the error terms
# H0 (null hypothesis): There is no correlation among the residuals
```

```
durbinWatsonTest(model2)
```

```
## lag Autocorrelation D-W Statistic p-value
## 1 0.091 1.8 0.016
## Alternative hypothesis: rho != 0
```

```
# Breusch-Pagan Test: Heteroscedasticity
# H0: Homoscedasticity is present
```

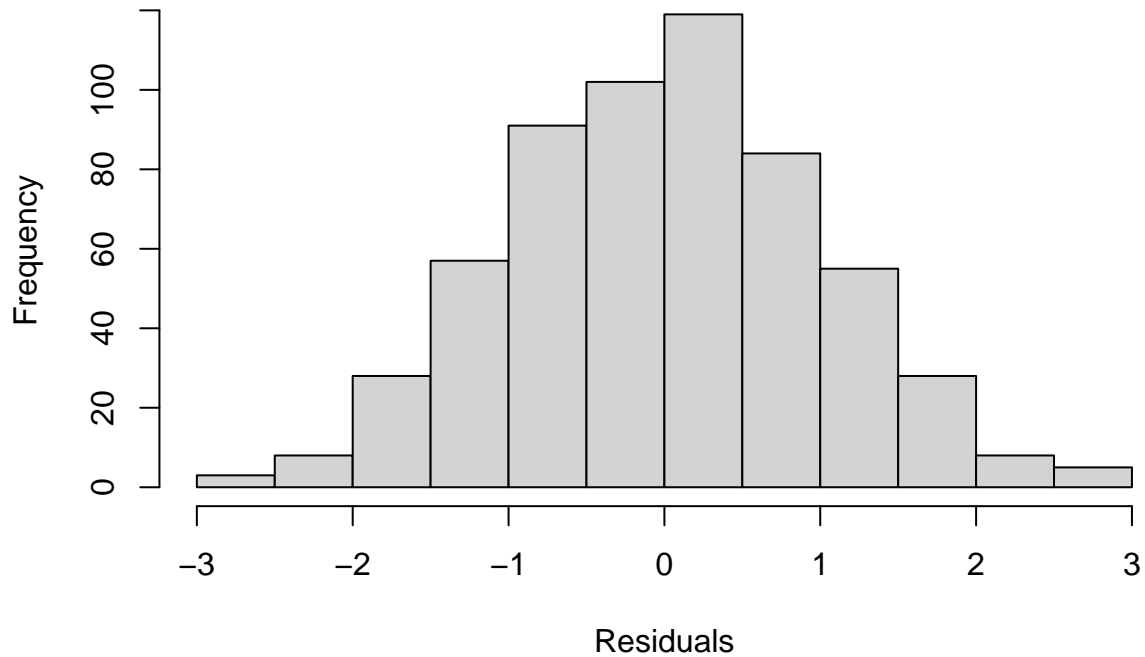
```
bptest(model2)
```

```
##
## studentized Breusch-Pagan test
##
## data: model2
## BP = 25, df = 33, p-value = 0.8
```

```
# Normality assumption
```

```
hist(res2, xlab="Residuals", main= "Histogram of Residuals")
```

Histogram of Residuals



```
## normality test using shapiro-test: reject the H0  
#H0: the sample comes from a normal distribution
```

```
res2_num = res2[is.finite(res2)]  
shapiro.test(res2_num)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: res2_num  
## W = 1, p-value = 0.6
```

3. Variable selection

```
### Backward regression using AIC: starting with all of the variables  
options(scipen = -2)  
  
step_model2 <- stepAIC(model2, trace=TRUE, direction= "backward")
```

FINAL MODEL

```

## Start: AIC=4e+03
## belief_diff_food ~ age + income + political_party + education +
##   urban_rural_class + federal_state
##
##           Df Sum of Sq   RSS   AIC
## - federal_state 14    18145 524152 4034
## - education      6     4129 510136 4034
## - urban_rural_class 3     2002 508009 4038
## - income          1     1082 507088 4041
## - age             1     1375 507382 4041
## <none>                        506007 4041
## - political_party  8     28531 534538 4058
##
## Step: AIC=4e+03
## belief_diff_food ~ age + income + political_party + education +
##   urban_rural_class
##
##           Df Sum of Sq   RSS   AIC
## - education      6     4837 528988 4028
## - urban_rural_class 3     1548 525700 4030
## - income          1      614 524766 4033
## - age             1     1239 525391 4034
## <none>                        524152 4034
## - political_party  8     26653 550805 4047
##
## Step: AIC=4e+03
## belief_diff_food ~ age + income + political_party + urban_rural_class
##
##           Df Sum of Sq   RSS   AIC
## - urban_rural_class 3     1096 530085 4023
## - income            1      353 529342 4026
## <none>                        528988 4028
## - age              1     2038 531027 4028
## - political_party  8     26735 555724 4041
##
## Step: AIC=4e+03
## belief_diff_food ~ age + income + political_party
##
##           Df Sum of Sq   RSS   AIC
## - income          1      392 530477 4021
## <none>                        530085 4023
## - age             1     2212 532297 4023
## - political_party  8     26043 556127 4035
##
## Step: AIC=4e+03
## belief_diff_food ~ age + political_party
##
##           Df Sum of Sq   RSS   AIC
## <none>                        530477 4021
## - age          1     2210 532687 4022
## - political_party  8     26403 556880 4034

```

```
summary(step_model2)
```

```
##
## Call:
## lm(formula = belief_diff_food ~ age + political_party, data = df2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -90.92 -20.96  -0.04   22.33   95.12
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.46e-01   4.00e+00   0.14  8.9e-01
## age            1.47e-01   9.45e-02   1.55  1.2e-01
## political_partyAfD      -1.65e+01   4.74e+00  -3.48  5.4e-04
## political_partyBündnis Sarah Wagenknecht -1.21e+01   6.83e+00  -1.77  7.7e-02
## political_partyCDU/CSU    -9.85e+00   4.35e+00  -2.26  2.4e-02
## political_partyDie Linke   -1.70e+00   5.22e+00  -0.33  7.5e-01
## political_partyEiner anderen Partei    -8.60e+00   3.83e+00  -2.24  2.5e-02
## political_partyFDP        -8.97e+00   5.06e+00  -1.77  7.7e-02
## political_partyKeine Angabe    1.12e+01   8.43e+00   1.32  1.9e-01
## political_partySPD       -1.66e+01   4.43e+00  -3.74  2.0e-04
##
## (Intercept)
## age
## political_partyAfD ***
## political_partyBündnis Sarah Wagenknecht .
## political_partyCDU/CSU *
## political_partyDie Linke
## political_partyEiner anderen Partei *
## political_partyFDP .
## political_partyKeine Angabe
## political_partySPD ***
## ---
## Signif. codes:  0 '***' 1e-03 '**' 1e-02 '*' 5e-02 '.' 0.1 ' ' 1
##
## Residual standard error: 30 on 578 degrees of freedom
## Multiple R-squared:  0.0507, Adjusted R-squared:  0.0359
## F-statistic: 3.43 on 9 and 578 DF,  p-value: 3.98e-04
```

```
vif(step_model2)
```

```
##              GVIF Df GVIF^(1/(2*Df))
## age              1.1  1              1
## political_party  1.1  8              1
```

```
# Checking outliers
```

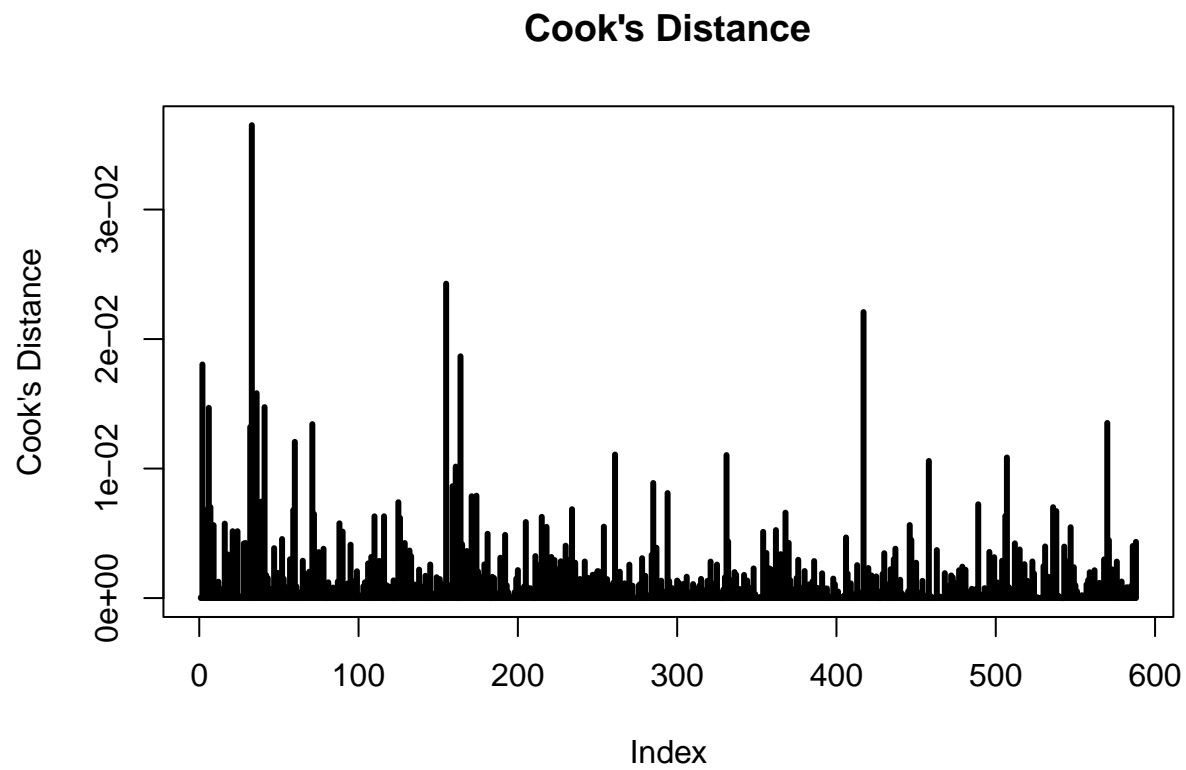
```
cook = cooks.distance(step_model2)
plot(cook,
      type="h",
      lwd=3,
```



```

    ylab = "Cook's Distance",
    main="Cook's Distance")
abline(h = 1)

```



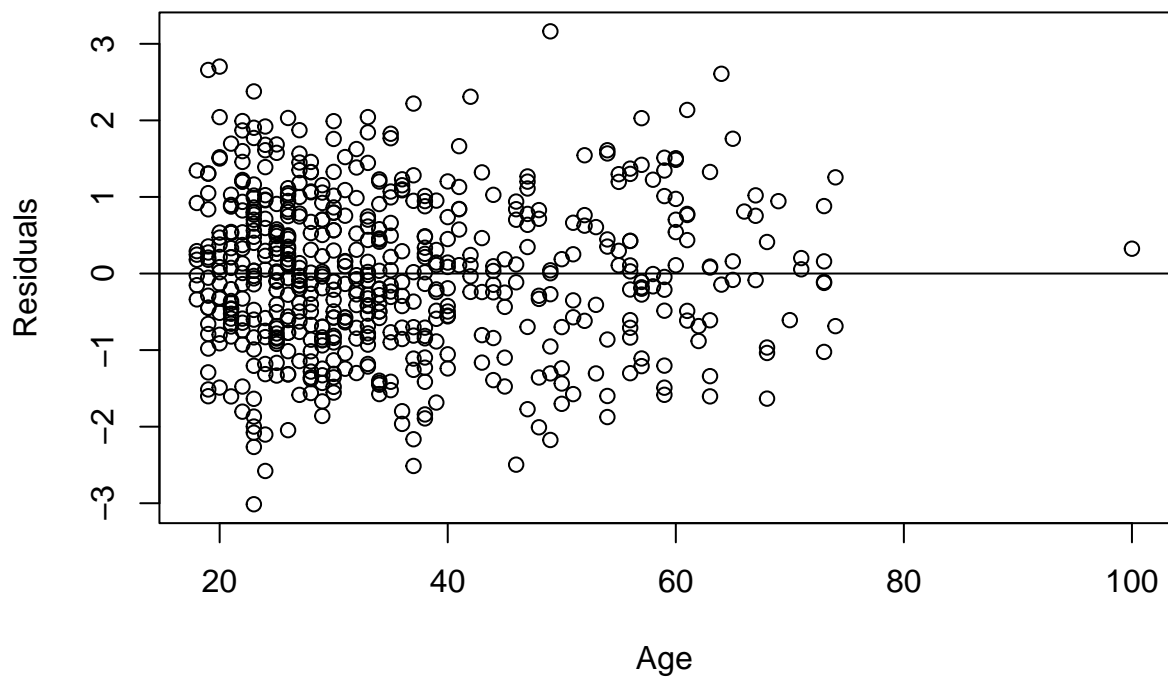
```

res2 = stdres(step_model2) ## (Standardized) Residuals

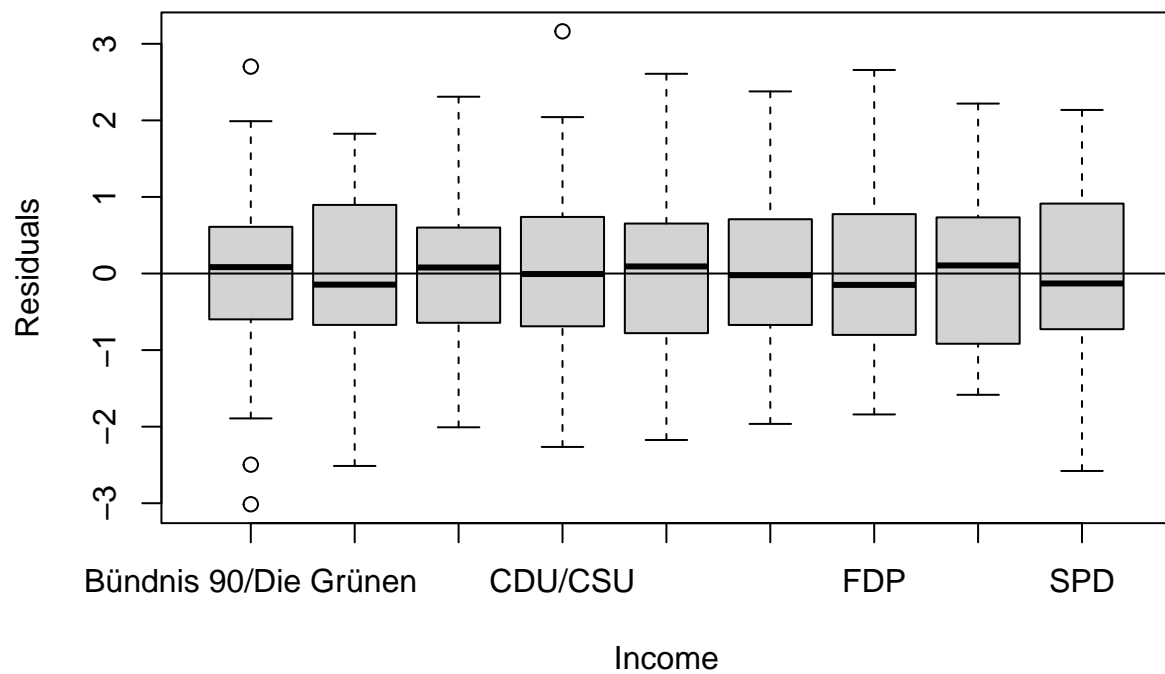
# Linearity assumption/Mean zero assumption

plot(df2$age, res2, xlab = "Age", ylab = "Residuals")
abline(h = 0)

```



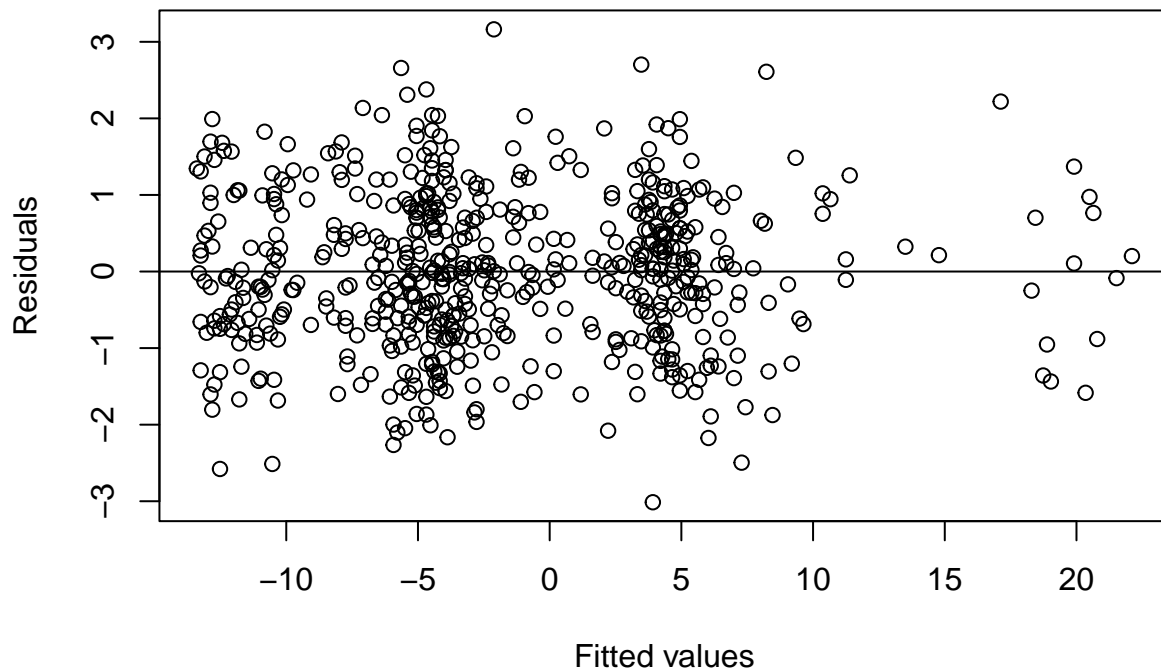
```
plot(df2$political_party, res2, xlab = "Income", ylab = "Residuals")  
abline(h = 0)
```



```
#plot(df2$federal_state, res2, xlab = "Income", ylab = "Residuals")
#abline(h = 0)
```

```
# Constant variance and independent error term assumption
```

```
plot(fitted(step_model2), res2, xlab = "Fitted values", ylab = "Residuals")
abline(h = 0)
```



```
# Durbin-Watson Test: Independence of the error terms
# H0 (null hypothesis): There is no correlation among the residuals
```

```
durbinWatsonTest(step_model2)
```

```
## lag Autocorrelation D-W Statistic p-value
## 1 0.085 1.8 0.032
## Alternative hypothesis: rho != 0
```

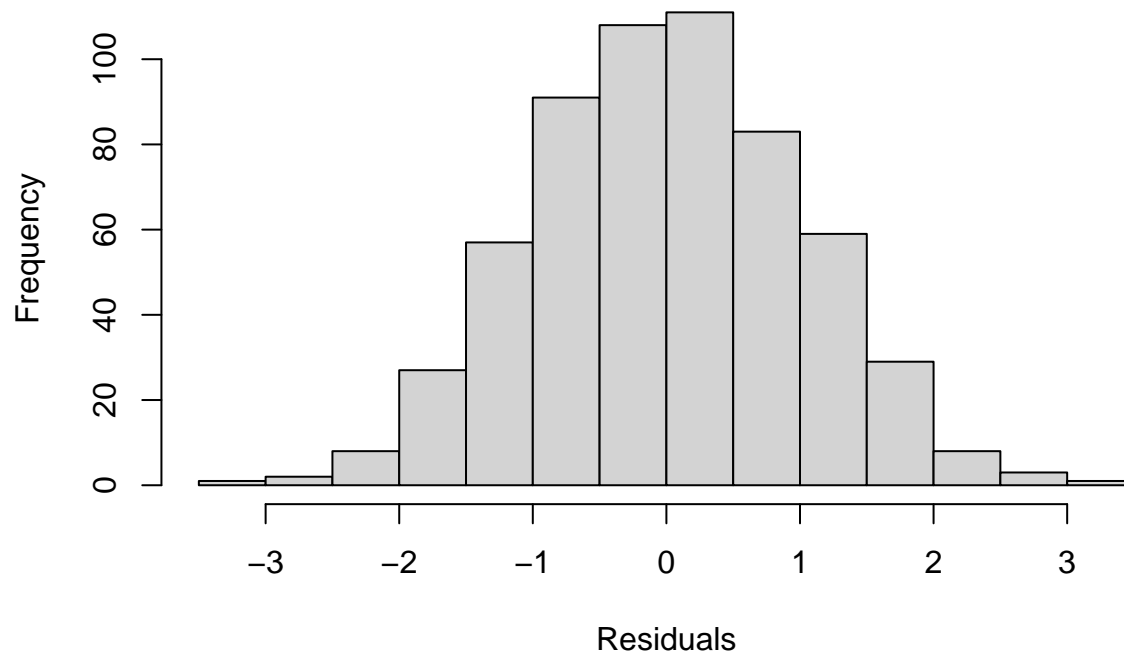
```
# Breusch-Pagan Test: Heteroscedasticity
# H0: Homoscedasticity is present
```

```
bptest(step_model2)
```

```
##
## studentized Breusch-Pagan test
##
## data: step_model2
## BP = 7, df = 9, p-value = 0.6
```

```
hist(res2, xlab="Residuals", main= "Histogram of Residuals")
```

Histogram of Residuals



```
## normality test using shapiro-test: reject the H0  
#H0: the sample comes from a normal distribution
```

```
res2_num = res2[is.finite(res2)]  
shapiro.test(res2_num)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: res2_num  
## W = 1, p-value = 0.8
```