

STAT 331 Final Project

Maxine, Estella, Judy, Weiwei

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Requirement of the project

Your 7–10 page report must contain the following components:

- 1. Summary: A maximum of 200 words describing the objective of the report, an overview of the statistical analysis, and summary of the main results.
- 2. Objective: Describe your goals for the analysis.
- 3. Exploratory Data Analysis: Conduct exploratory data analyses: report summary statistics, visualize data (histograms, scatter plots, etc.). Report on any interesting findings and comment on how these inform the rest of your analysis.
- 4. Methods: Describe your statistical analysis: What is your model? Did you use any transformations or extensions of the basic multiple linear regression model? How did you select a model? Does the model fit the data well? Are the necessary assumptions met? Be sure to explain and justify your decisions.
- 5. Results: Report on the findings of your analysis
- 6. Discussion: Comment on your findings/conclusions; describe any limitations of your analysis.

1. Summary

A maximum of 200 words describing the objective of the report, an overview of the statistical analysis, and summary of the main results.

2. Objective

The goal of this project is to analyze the pollutants.csv data and write a report on your analysis. The specific goals of your analysis are up to you to decide.

3. Exploratory Data Analysis

Conduct exploratory data analyses: report summary statistics, visualize data (histograms, scatter plots, etc.). Report on any interesting findings and comment on how these inform the rest of your analysis.

can use this as a tutorial <https://r4ds.had.co.nz/exploratory-data-analysis.html>

Take a peak at the first 5 entries

```
# CHANGE ABSOLUTE PATH
# setwd("~/Desktop/stat341/R331project/data")
# setwd("~/School/4A/STAT 331/R331project/data")
setwd("~/Desktop/R331project/data")

pollutants <- read.csv("pollutants.csv", header = TRUE)
head(pollutants)
```

```
##   X      length POP_PCB1 POP_PCB2 POP_PCB3 POP_PCB4 POP_PCB5 POP_PCB6 POP_PCB7
## 1 1 1.1587651   20000     7600     3700    14700    18900     5300     5500
## 2 2 0.9011283   43900    14900     9700    32300    55500    13400    18700
## 3 3 1.2753948    3300     3300     3300     3300     3300     3300     3300
## 4 4 0.9369063    8500     4100     6000    11500    13500     6900    13500
## 5 5 0.7027998  159000    60200    29800   170000   215000    79200    47400
## 6 6 1.1516147   14400     7100    16900    28200    37200    22000    10200
##   POP_PCB8 POP_PCB9 POP_PCB10 POP_PCB11 POP_dioxin1 POP_dioxin2 POP_dioxin3
## 1      5700     2000     15.6     23.1       70.9       50.0       173
## 2     12000    16200     35.4     31.1      116.0      129.0       709
## 3      3300     3300      1.8      9.3       29.9       5.4       148
## 4      4100     4100      4.5     21.1       50.4       29.4       668
## 5     41400    53900     59.2     80.3       98.1       80.1       875
## 6      3800     6400     19.2     70.0      106.0       47.4       533
##   POP_furan1 POP_furan2 POP_furan3 POP_furan4 whitecell_count lymphocyte_pct
## 1          6.9         5.6         0.8        15.6             5.4           33.8
## 2         18.5        15.4        20.3         2.3             5.6           16.8
## 3          1.3         1.4         1.2         2.9             6.3           35.3
## 4          2.2         2.4         2.3        43.2             8.4           23.0
## 5         13.7         1.2         0.8        11.0             6.7           24.5
## 6          8.3         7.0         3.4        19.4             4.7           39.5
##   monocyte_pct eosinophils_pct basophils_pct neutrophils_pct   BMI edu_cat
## 1          8.1             51.2           6.2           0.6 27.50      2
## 2         10.2             69.4           3.2           0.5 27.46      3
## 3          7.3             54.9           1.6           0.9 36.13      1
## 4          6.4             68.8           1.7           0.2 21.79      4
## 5          7.5             64.3           3.0           0.8 31.46      2
## 6          4.4             54.2           1.3           0.8 40.68      1
##   race_cat male ageyrs yrssmoke smokenow ln_lbxcot
## 1         4     1     41         0         0 -2.312635
## 2         4     0     77         0         0 -4.509860
## 3         2     0     22         0         0 -4.017384
## 4         4     0     27         0         0 -3.863233
## 5         4     1     78         0         0 -1.826351
## 6         3     0     35         0         0 -2.207275
```

Covariates

```
names(pollutants)
```

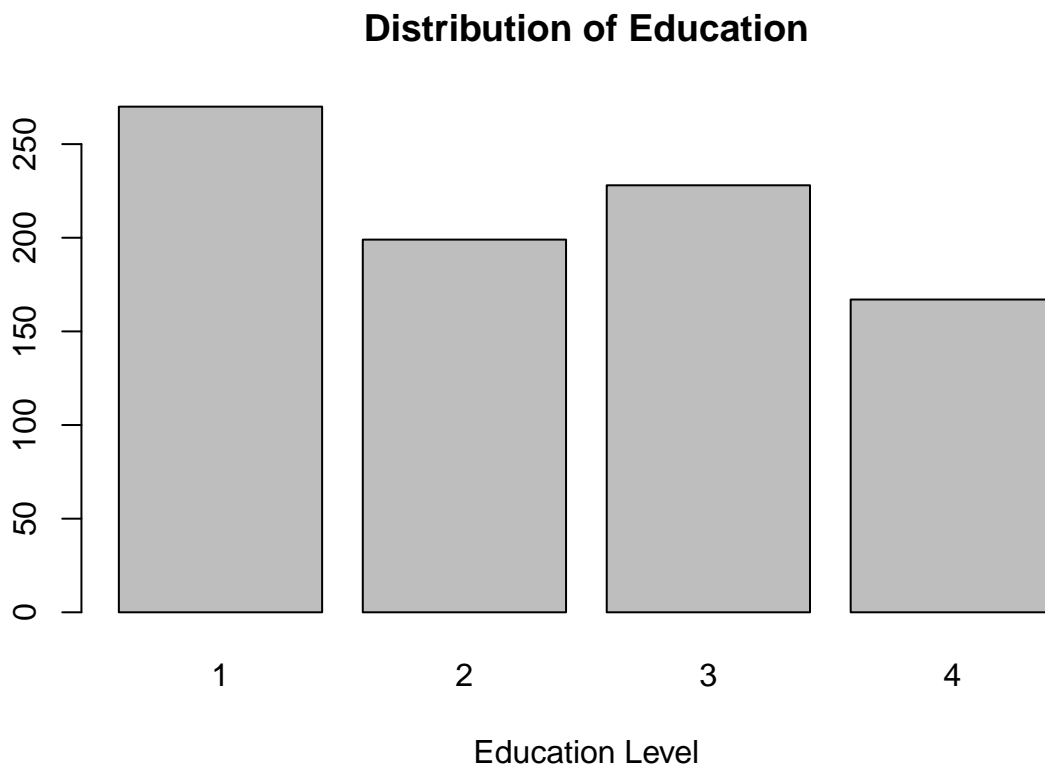
```
## [1] "X"                "length"           "POP_PCB1"         "POP_PCB2"
## [5] "POP_PCB3"         "POP_PCB4"         "POP_PCB5"         "POP_PCB6"
## [9] "POP_PCB7"         "POP_PCB8"         "POP_PCB9"         "POP_PCB10"
## [13] "POP_PCB11"        "POP_dioxin1"      "POP_dioxin2"      "POP_dioxin3"
## [17] "POP_furan1"      "POP_furan2"      "POP_furan3"      "POP_furan4"
```

```
## [21] "whitecell_count" "lymphocyte_pct" "monocyte_pct" "eosinophils_pct"
## [25] "basophils_pct" "neutrophils_pct" "BMI" "edu_cat"
## [29] "race_cat" "male" "ageyrs" "yrssmoke"
## [33] "smokenow" "ln_lbxcot"
```

Note that “edu_cat”, “race_cat”, “male”, “smokenow” are categorical data.

```
# 1 = Less Than 9th Grade or 9-11th Grade (Includes 12th grade with no diploma)
# 2 = High School Grad/GED or Equivalent
# 3 = Some College or AA degree
# 4 = College Graduate
```

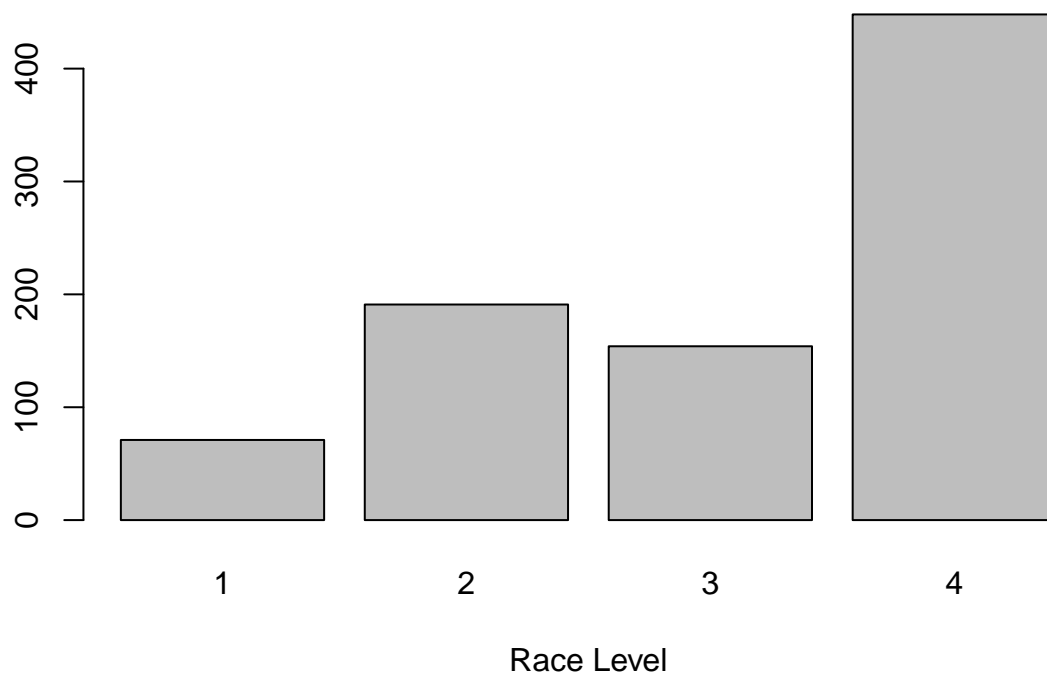
```
edu_factor=as.factor(pollutants$edu_cat)
plot(edu_factor,
     main="Distribution of Education",
     xlab="Education Level")
```



```
# 1 = Other Race (Including Multi-Racial);
# 2 = Mexican American;
# 3 = Non-Hispanic Black;
# 4 = Non-Hispanic White
```

```
race_factor=as.factor(pollutants$race_cat)
plot(race_factor,
     main="Distribution of Race",
     xlab="Race Level")
```

Distribution of Race



```
# Estella's work 1
library(corrplot)
```

```
## corrplot 0.84 loaded
```

```
library(ggplot2)
```

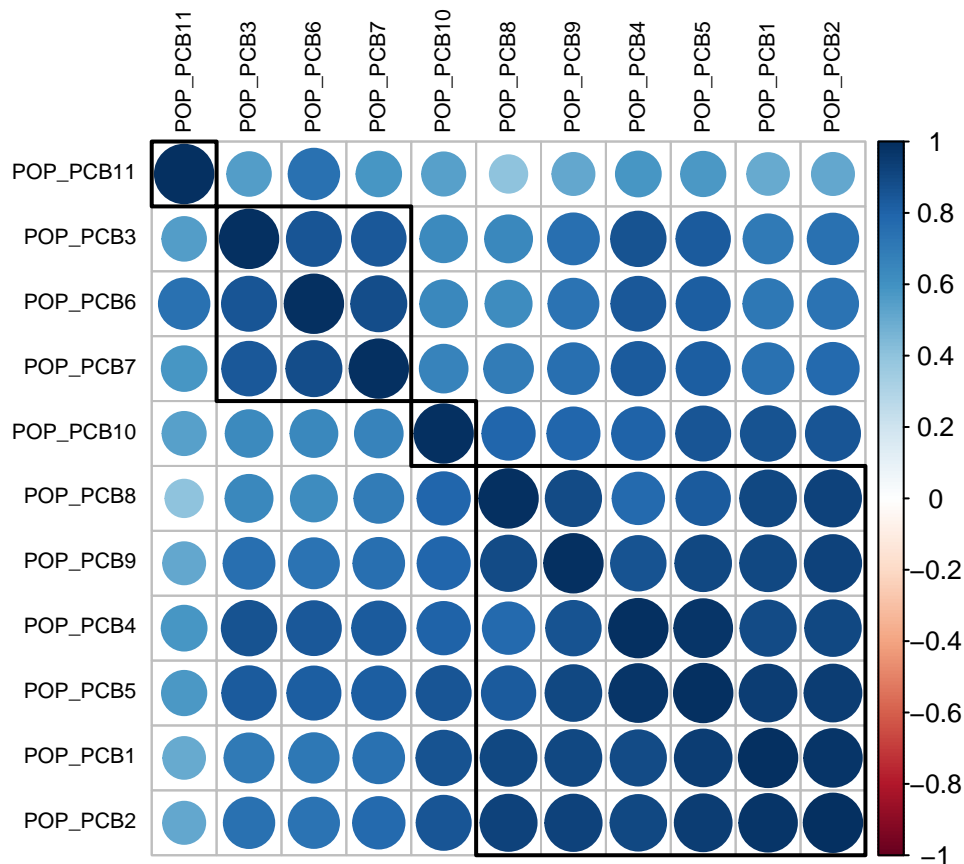
```
summary(pollutants)
```

```
##           X           length      POP_PCB1      POP_PCB2
## Min.      : 1.0      Min.      :0.5266      Min.      : 2000      Min.      : 2000
## 1st Qu.:216.8      1st Qu.:0.8754      1st Qu.: 9975      1st Qu.: 4800
## Median :432.5      Median :1.0286      Median : 27600     Median : 11500
## Mean      :432.5      Mean      :1.0543      Mean      : 38082     Mean      : 15637
## 3rd Qu.:648.2      3rd Qu.:1.2095      3rd Qu.: 53325     3rd Qu.: 21825
## Max.      :864.0      Max.      :2.3512      Max.      :572000     Max.      :165000
## POP_PCB3      POP_PCB4      POP_PCB5      POP_PCB6
## Min.      : 2000      Min.      : 2100      Min.      : 2100      Min.      : 2000
## 1st Qu.: 3700      1st Qu.: 11475      1st Qu.: 15600      1st Qu.: 4400
## Median : 6200      Median : 25550      Median : 36300      Median : 9400
## Mean      : 10158     Mean      : 38456      Mean      : 52650      Mean      : 16820
## 3rd Qu.: 12000      3rd Qu.: 50650      3rd Qu.: 68625      3rd Qu.: 19500
## Max.      :123000     Max.      :487000      Max.      :708000      Max.      :319000
## POP_PCB7      POP_PCB8      POP_PCB9      POP_PCB10
## Min.      : 1100      Min.      : 1100      Min.      : 1100      Min.      : 1.70
## 1st Qu.: 4000      1st Qu.: 3800      1st Qu.: 3900      1st Qu.: 9.10
## Median : 7450      Median : 6950      Median : 8050      Median : 18.35
## Mean      : 12682     Mean      : 10530      Mean      : 12220      Mean      : 24.49
## 3rd Qu.: 15625      3rd Qu.: 14425      3rd Qu.: 16025      3rd Qu.: 34.90
## Max.      :144000     Max.      :187000      Max.      :144000      Max.      :172.00
```

```
## POP_PCB11 POP_dioxin1 POP_dioxin2 POP_dioxin3
## Min. : 1.30 Min. : 1.90 Min. : 1.40 Min. : 36.8
## 1st Qu.: 14.80 1st Qu.: 23.90 1st Qu.: 21.27 1st Qu.: 197.0
## Median : 24.50 Median : 41.35 Median : 37.80 Median : 342.5
## Mean : 38.15 Mean : 57.65 Mean : 47.81 Mean : 494.4
## 3rd Qu.: 42.95 3rd Qu.: 71.62 3rd Qu.: 62.42 3rd Qu.: 603.0
## Max. :845.00 Max. :760.00 Max. :281.00 Max. :8190.0
## POP_furan1 POP_furan2 POP_furan3 POP_furan4
## Min. : 1.000 Min. : 0.800 Min. : 0.700 Min. : 0.90
## 1st Qu.: 3.200 1st Qu.: 2.600 1st Qu.: 2.200 1st Qu.: 6.40
## Median : 5.200 Median : 4.200 Median : 5.050 Median : 9.65
## Mean : 6.371 Mean : 5.390 Mean : 6.669 Mean : 11.54
## 3rd Qu.: 7.700 3rd Qu.: 6.825 3rd Qu.: 9.300 3rd Qu.: 14.00
## Max. :44.400 Max. :33.500 Max. :38.300 Max. :234.00
## whitecell_count lymphocyte_pct monocyte_pct eosinophils_pct
## Min. : 2.300 Min. : 5.80 Min. : 1.600 Min. :21.60
## 1st Qu.: 5.600 1st Qu.:24.00 1st Qu.: 6.600 1st Qu.:52.35
## Median : 6.900 Median :28.95 Median : 7.700 Median :59.30
## Mean : 7.191 Mean :29.92 Mean : 7.936 Mean :58.62
## 3rd Qu.: 8.300 3rd Qu.:35.42 3rd Qu.: 9.100 3rd Qu.:65.22
## Max. :20.100 Max. :73.40 Max. :23.800 Max. :88.10
## basophils_pct neutrophils_pct BMI edu_cat
## Min. : 0.000 Min. :0.0000 Min. :16.16 Min. :1.000
## 1st Qu.: 1.500 1st Qu.:0.4000 1st Qu.:23.88 1st Qu.:1.000
## Median : 2.300 Median :0.6000 Median :27.38 Median :2.000
## Mean : 2.903 Mean :0.6669 Mean :28.09 Mean :2.338
## 3rd Qu.: 3.700 3rd Qu.:0.8000 3rd Qu.:31.17 3rd Qu.:3.000
## Max. :28.200 Max. :5.5000 Max. :62.99 Max. :4.000
## race_cat male ageyrs yrssmoke
## Min. :1.000 Min. :0.0000 Min. :20.00 Min. : 0.0
## 1st Qu.:2.000 1st Qu.:0.0000 1st Qu.:34.00 1st Qu.: 0.0
## Median :4.000 Median :0.0000 Median :46.00 Median : 0.0
## Mean :3.133 Mean :0.4329 Mean :48.36 Mean :10.6
## 3rd Qu.:4.000 3rd Qu.:1.0000 3rd Qu.:63.00 3rd Qu.:20.0
## Max. :4.000 Max. :1.0000 Max. :85.00 Max. :69.0
## smokenow ln_lbxcot
## Min. :0.0000 Min. : -4.5099
## 1st Qu.:0.0000 1st Qu.: -4.0745
## Median :0.0000 Median : -2.7334
## Mean :0.2315 Mean : -0.9804
## 3rd Qu.:0.0000 3rd Qu.: 2.8000
## Max. :1.0000 Max. : 6.5848
```

```
POP_PCB = c("POP_PCB1", "POP_PCB2", "POP_PCB3", "POP_PCB4", "POP_PCB5", "POP_PCB6", "POP_PCB7", "POP_PCB8")
POP_PCB_data <- pollutants[, POP_PCB]
cc = cor(POP_PCB_data, method = "spearman")
```

```
# cluster my POP_PCB so that those with similar patterns of correlation coefficients are closer together
corrplot(cc, tl.col = "black", order = "hclust", hclust.method = "average", addrect = 4, tl.cex = 0.7)
```



<https://jkkorz.github.io/2019/06/11/Correlation-heatmaps.html>

Estella's work 3

```
f <- as.formula((paste("length", paste("(", paste(POP_PCB, collapse = " + "), ")^2"), sep = " ~"))
m <- lm(f, data = pollutants)
summary(m)
```

```
##
## Call:
## lm(formula = f, data = pollutants)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.53819 -0.16080 -0.01896  0.12149  1.20671
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.153e+00  2.892e-02  39.876 < 2e-16 ***
## POP_PCB1     -6.741e-06  3.521e-06  -1.915  0.05591 .
## POP_PCB2      3.801e-06  9.328e-06   0.407  0.68378
## POP_PCB3      6.747e-06  6.701e-06   1.007  0.31431
## POP_PCB4      1.373e-06  3.278e-06   0.419  0.67539
## POP_PCB5      1.920e-06  3.267e-06   0.588  0.55680
## POP_PCB6     -3.673e-06  4.336e-06  -0.847  0.39729
## POP_PCB7     -5.281e-06  4.697e-06  -1.124  0.26126
## POP_PCB8     -1.073e-05  8.331e-06  -1.288  0.19796
## POP_PCB9     -1.833e-06  5.806e-06  -0.316  0.75232
```

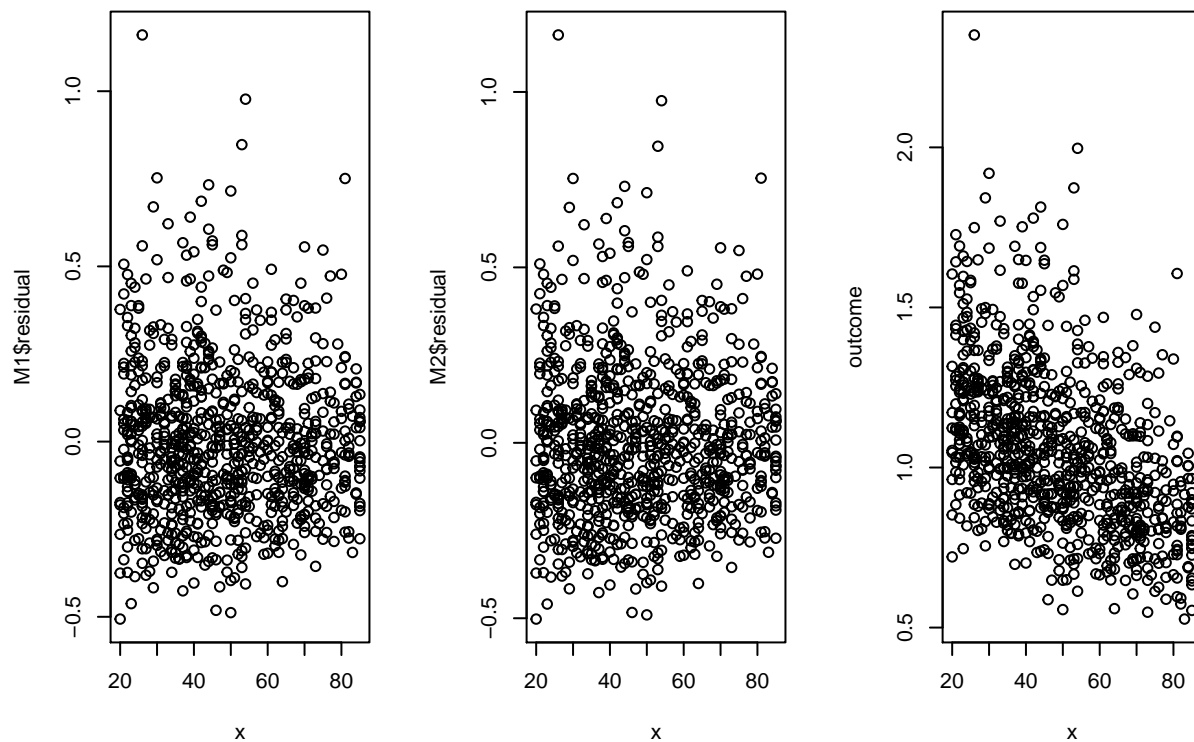
## POP_PCB10	2.720e-03	2.088e-03	1.303	0.19311
## POP_PCB11	4.644e-04	9.916e-04	0.468	0.63969
## POP_PCB1:POP_PCB2	9.529e-11	2.113e-10	0.451	0.65216
## POP_PCB1:POP_PCB3	-6.580e-10	4.156e-10	-1.583	0.11377
## POP_PCB1:POP_PCB4	1.116e-10	1.917e-10	0.582	0.56080
## POP_PCB1:POP_PCB5	-1.621e-11	1.318e-10	-0.123	0.90218
## POP_PCB1:POP_PCB6	6.244e-11	2.176e-10	0.287	0.77423
## POP_PCB1:POP_PCB7	2.221e-11	2.742e-10	0.081	0.93548
## POP_PCB1:POP_PCB8	-5.209e-10	2.693e-10	-1.935	0.05340 .
## POP_PCB1:POP_PCB9	4.146e-10	2.287e-10	1.813	0.07020 .
## POP_PCB1:POP_PCB10	1.675e-07	1.311e-07	1.277	0.20183
## POP_PCB1:POP_PCB11	-6.663e-08	7.321e-08	-0.910	0.36303
## POP_PCB2:POP_PCB3	1.673e-09	8.717e-10	1.919	0.05537 .
## POP_PCB2:POP_PCB4	-6.761e-10	4.688e-10	-1.442	0.14963
## POP_PCB2:POP_PCB5	3.840e-10	3.632e-10	1.057	0.29069
## POP_PCB2:POP_PCB6	-1.426e-09	5.834e-10	-2.444	0.01474 *
## POP_PCB2:POP_PCB7	1.532e-09	6.770e-10	2.264	0.02387 *
## POP_PCB2:POP_PCB8	2.135e-09	8.207e-10	2.602	0.00945 **
## POP_PCB2:POP_PCB9	-1.356e-09	7.249e-10	-1.870	0.06183 .
## POP_PCB2:POP_PCB10	-1.232e-06	4.242e-07	-2.904	0.00378 **
## POP_PCB2:POP_PCB11	3.388e-07	2.013e-07	1.683	0.09270 .
## POP_PCB3:POP_PCB4	-3.996e-11	1.199e-10	-0.333	0.73900
## POP_PCB3:POP_PCB5	4.665e-11	2.413e-10	0.193	0.84674
## POP_PCB3:POP_PCB6	-3.741e-10	2.662e-10	-1.405	0.16029
## POP_PCB3:POP_PCB7	6.438e-10	2.896e-10	2.223	0.02649 *
## POP_PCB3:POP_PCB8	7.340e-10	8.821e-10	0.832	0.40563
## POP_PCB3:POP_PCB9	-4.221e-10	5.470e-10	-0.772	0.44059
## POP_PCB3:POP_PCB10	-4.835e-07	2.555e-07	-1.892	0.05885 .
## POP_PCB3:POP_PCB11	7.155e-08	7.874e-08	0.909	0.36382
## POP_PCB4:POP_PCB5	3.002e-12	6.669e-11	0.045	0.96410
## POP_PCB4:POP_PCB6	1.788e-10	1.543e-10	1.159	0.24694
## POP_PCB4:POP_PCB7	-2.117e-10	1.579e-10	-1.341	0.18019
## POP_PCB4:POP_PCB8	-4.525e-11	3.961e-10	-0.114	0.90908
## POP_PCB4:POP_PCB9	1.217e-10	2.625e-10	0.464	0.64294
## POP_PCB4:POP_PCB10	1.345e-07	8.933e-08	1.505	0.13265
## POP_PCB4:POP_PCB11	1.685e-08	5.047e-08	0.334	0.73861
## POP_PCB5:POP_PCB6	4.714e-11	1.390e-10	0.339	0.73458
## POP_PCB5:POP_PCB7	-1.555e-10	1.446e-10	-1.076	0.28244
## POP_PCB5:POP_PCB8	-4.639e-10	3.185e-10	-1.457	0.14562
## POP_PCB5:POP_PCB9	-1.626e-11	1.822e-10	-0.089	0.92890
## POP_PCB5:POP_PCB10	9.703e-08	9.241e-08	1.050	0.29406
## POP_PCB5:POP_PCB11	-5.549e-08	4.079e-08	-1.360	0.17407
## POP_PCB6:POP_PCB7	-2.248e-11	1.147e-10	-0.196	0.84474
## POP_PCB6:POP_PCB8	7.086e-10	3.808e-10	1.861	0.06310 .
## POP_PCB6:POP_PCB9	4.295e-10	3.267e-10	1.315	0.18895
## POP_PCB6:POP_PCB10	2.152e-07	1.182e-07	1.820	0.06909 .
## POP_PCB6:POP_PCB11	-4.299e-08	2.038e-08	-2.109	0.03523 *
## POP_PCB7:POP_PCB8	-1.029e-09	4.279e-10	-2.404	0.01645 *
## POP_PCB7:POP_PCB9	-2.467e-10	3.622e-10	-0.681	0.49603
## POP_PCB7:POP_PCB10	-3.893e-08	1.308e-07	-0.298	0.76608
## POP_PCB7:POP_PCB11	4.226e-08	3.690e-08	1.145	0.25246
## POP_PCB8:POP_PCB9	1.317e-10	5.297e-10	0.249	0.80373
## POP_PCB8:POP_PCB10	5.264e-07	3.029e-07	1.738	0.08265 .
## POP_PCB8:POP_PCB11	-5.764e-08	1.285e-07	-0.449	0.65382

```
## POP_PCB9:POP_PCB10 -2.240e-08 1.448e-07 -0.155 0.87712
## POP_PCB9:POP_PCB11 7.916e-08 6.811e-08 1.162 0.24548
## POP_PCB10:POP_PCB11 -5.384e-05 2.694e-05 -1.999 0.04599 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2377 on 797 degrees of freedom
## Multiple R-squared:  0.1666, Adjusted R-squared:  0.09763
## F-statistic: 2.415 on 66 and 797 DF,  p-value: 1.316e-08
```

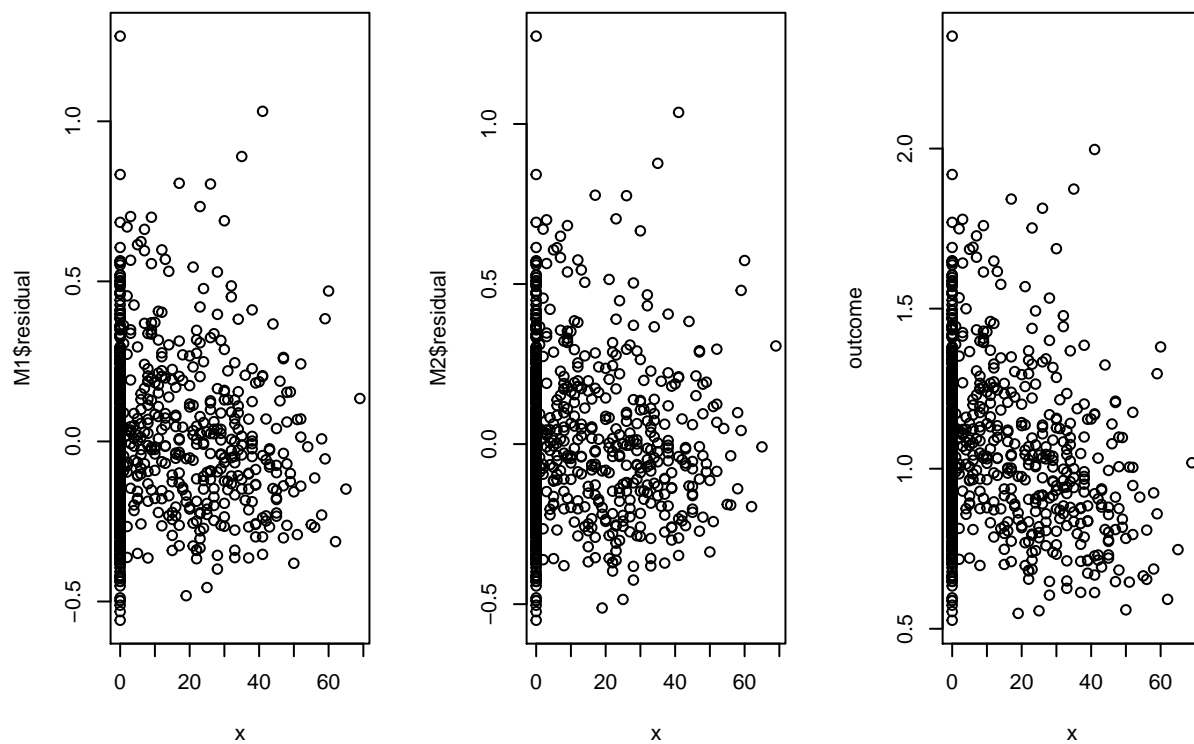
```
# Judy's work Part 1
# testing non-linearity in SLR
# if for any covariate, residual vs x for M1 has a pattern and
# residual vs x for M2 seems random, then y has a nonlinear
# relationship with with x.
# M1: fitting y to x
# M2: fitting y to x^2

par(mfrow=c(1, 3))
outcome <- pollutants$length
check <- function(x) {
  M1 <- lm(outcome ~ x)
  print(paste("residual for M1: ", sigma(M1)))
  M2 <- lm(outcome ~ x + I(x^2))
  print(paste("residual for M2: ", sigma(M2)))
  plot(x, M1$residual)
  plot(x, M2$residual)
  plot(x, outcome)
}
list <- list(pollutants$ageyrs, pollutants$yrssmoke,
             pollutants$BMI, pollutants$ln_lbxcot,
             pollutants$whitecell_count, pollutants$lymphocyte_pct,
             pollutants$monocyte_pct, pollutants$eosinophils_pct,
             pollutants$basophils_pct, pollutants$neutrophils_pct)
for (column in list) {
  check(column)
}
```

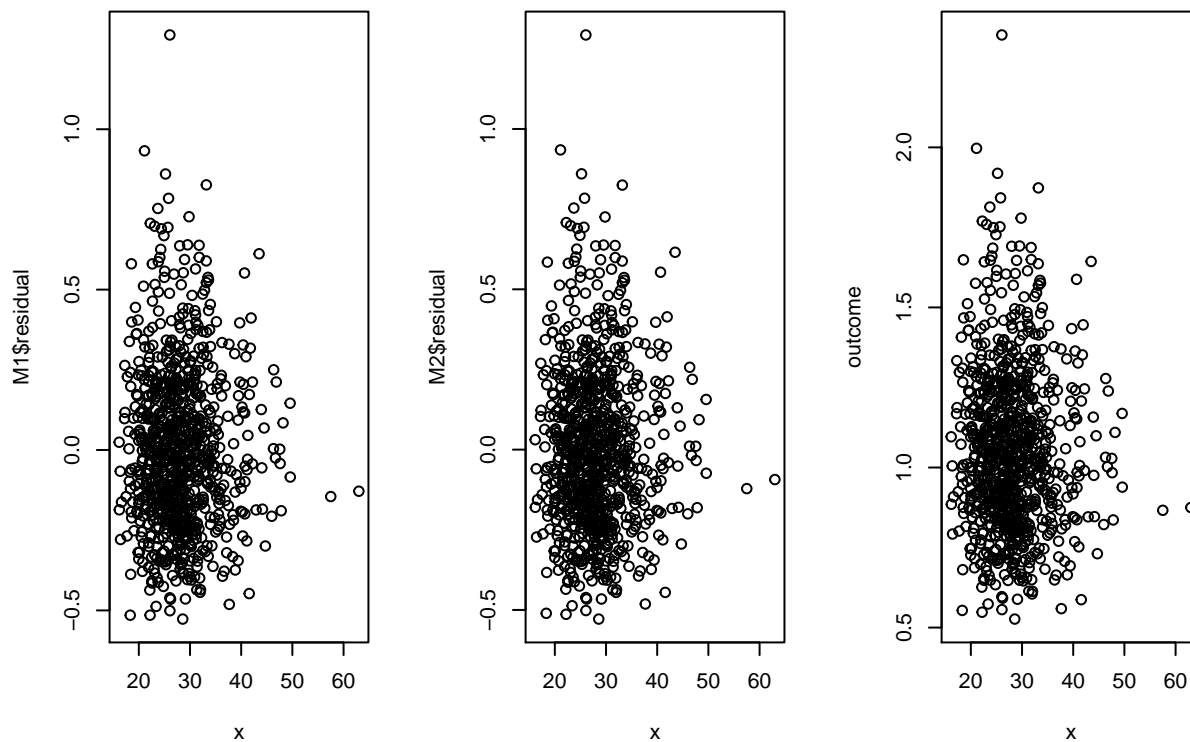
```
## [1] "residual for M1: 0.224172364185412"
## [1] "residual for M2: 0.22429269961392"
```

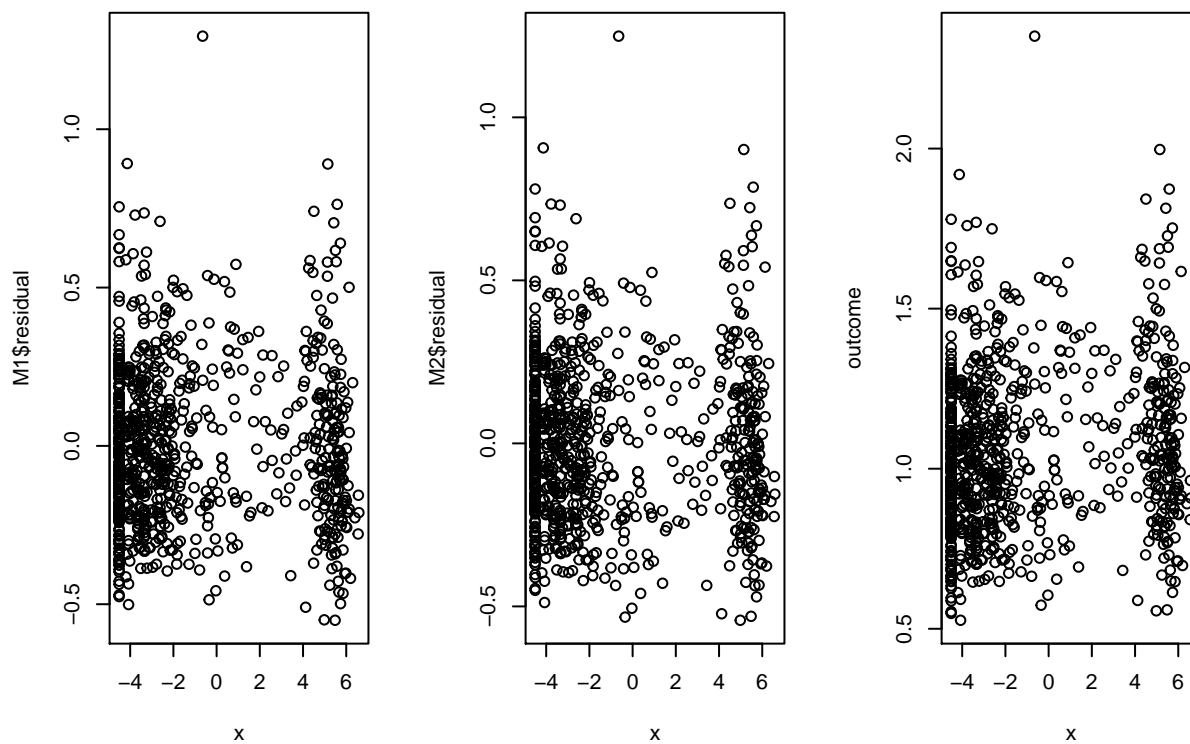
```
## [1] "residual for M1: 0.246320733146214"
## [1] "residual for M2: 0.245622720856213"
```



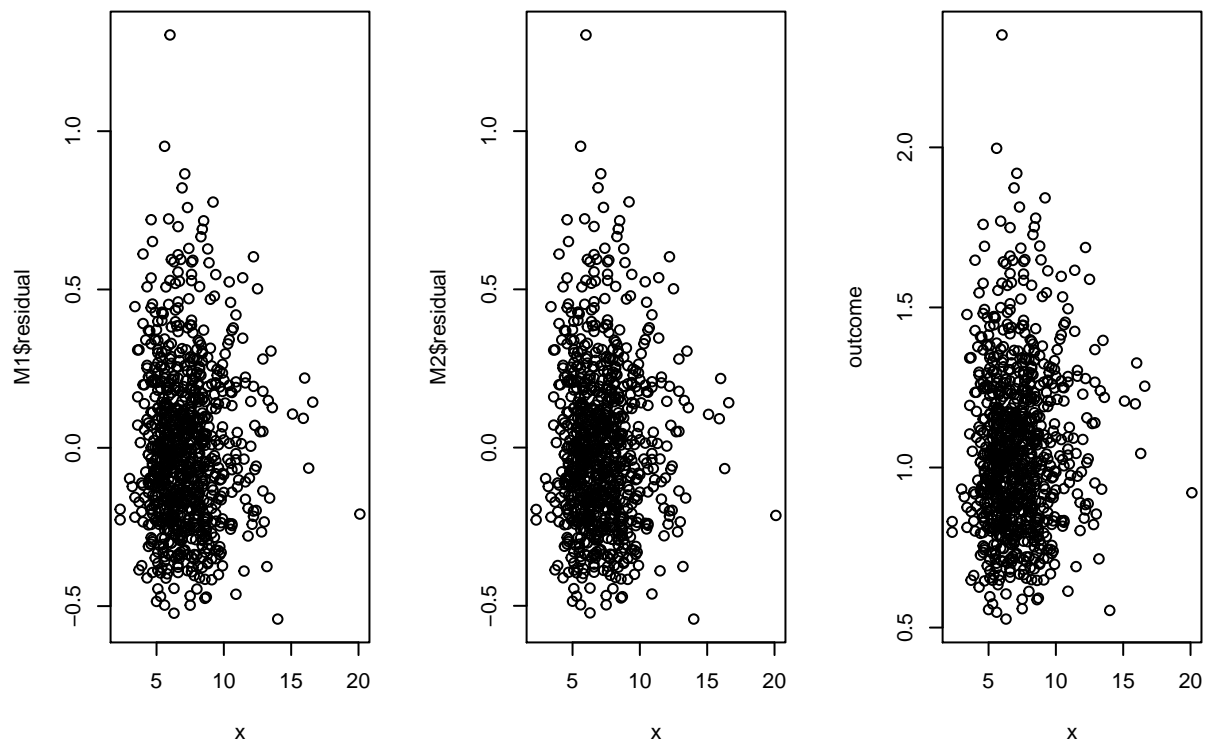
```
## [1] "residual for M1: 0.250228706427173"
## [1] "residual for M2: 0.25036248052387"
```



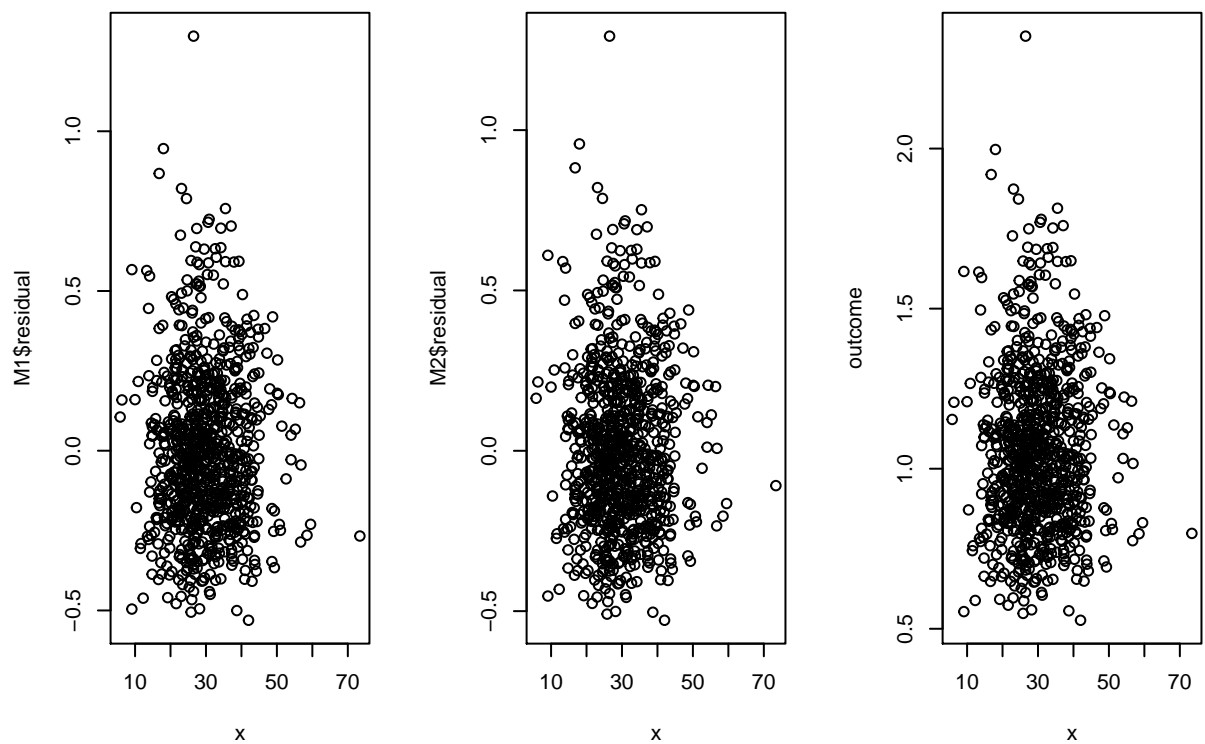
```
## [1] "residual for M1: 0.248212063673837"
## [1] "residual for M2: 0.24710732733351"
```



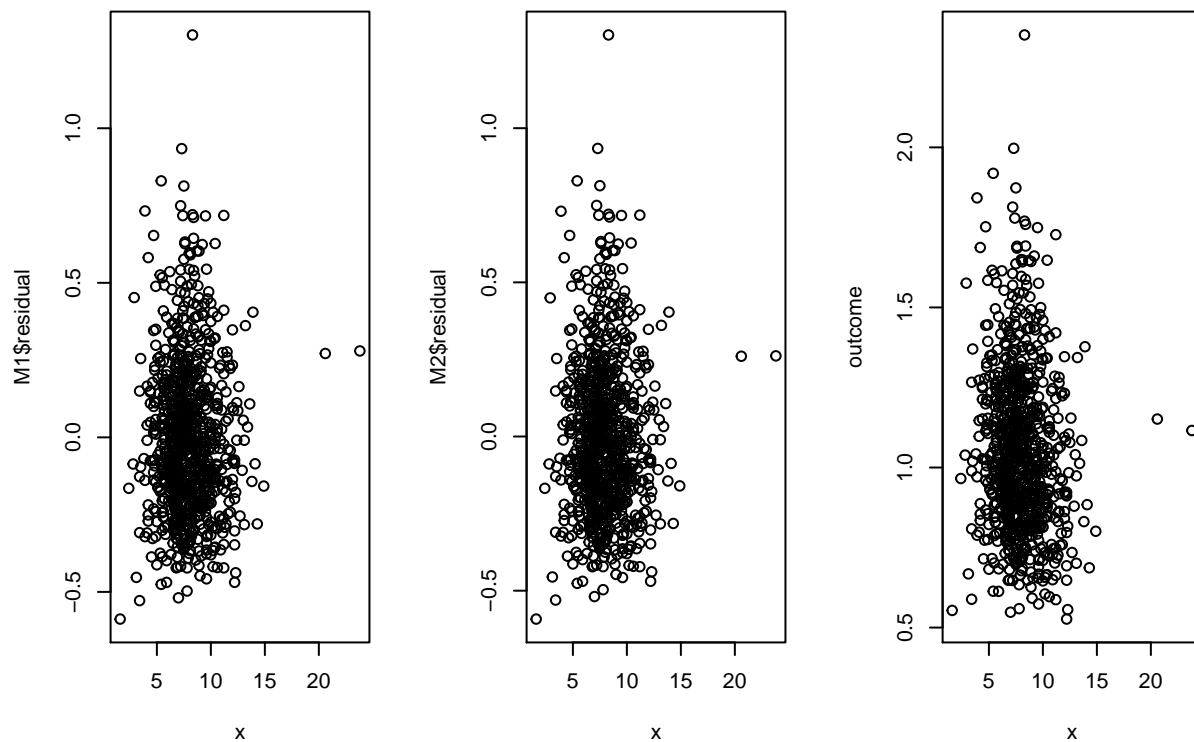
```
## [1] "residual for M1: 0.250065445847753"
## [1] "residual for M2: 0.250210403543218"
```



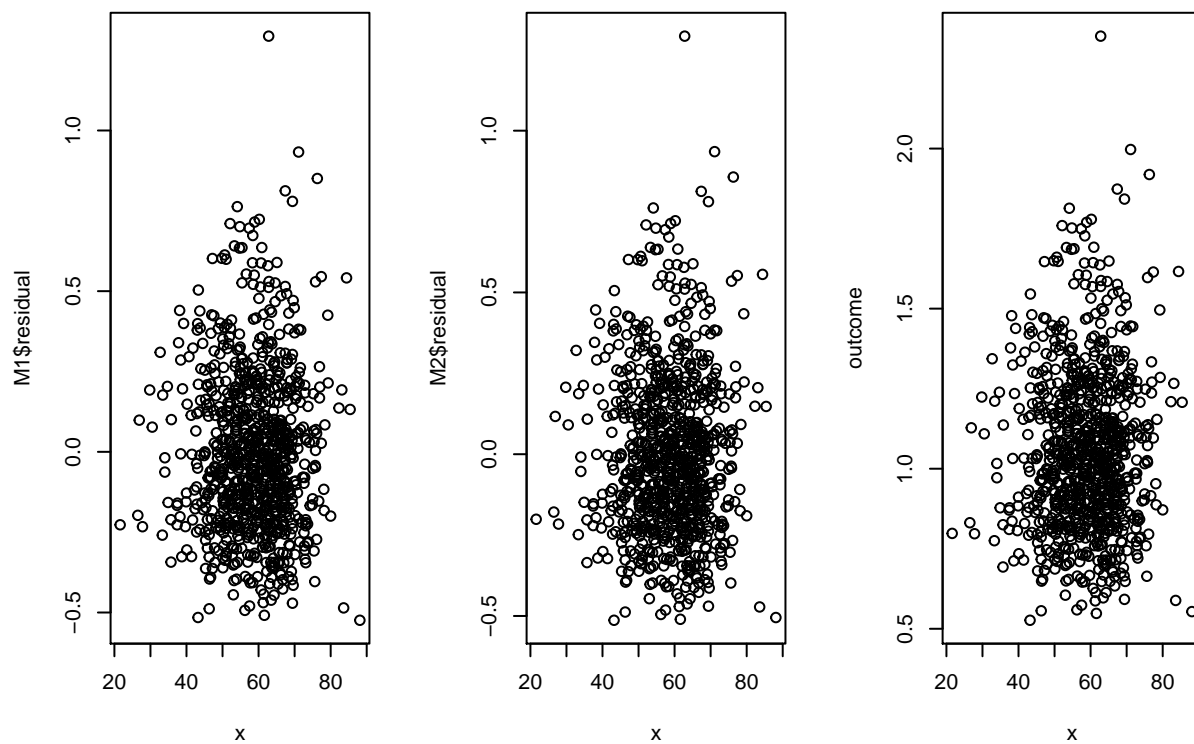
```
## [1] "residual for M1: 0.250373616826691"
## [1] "residual for M2: 0.250255208638358"
```



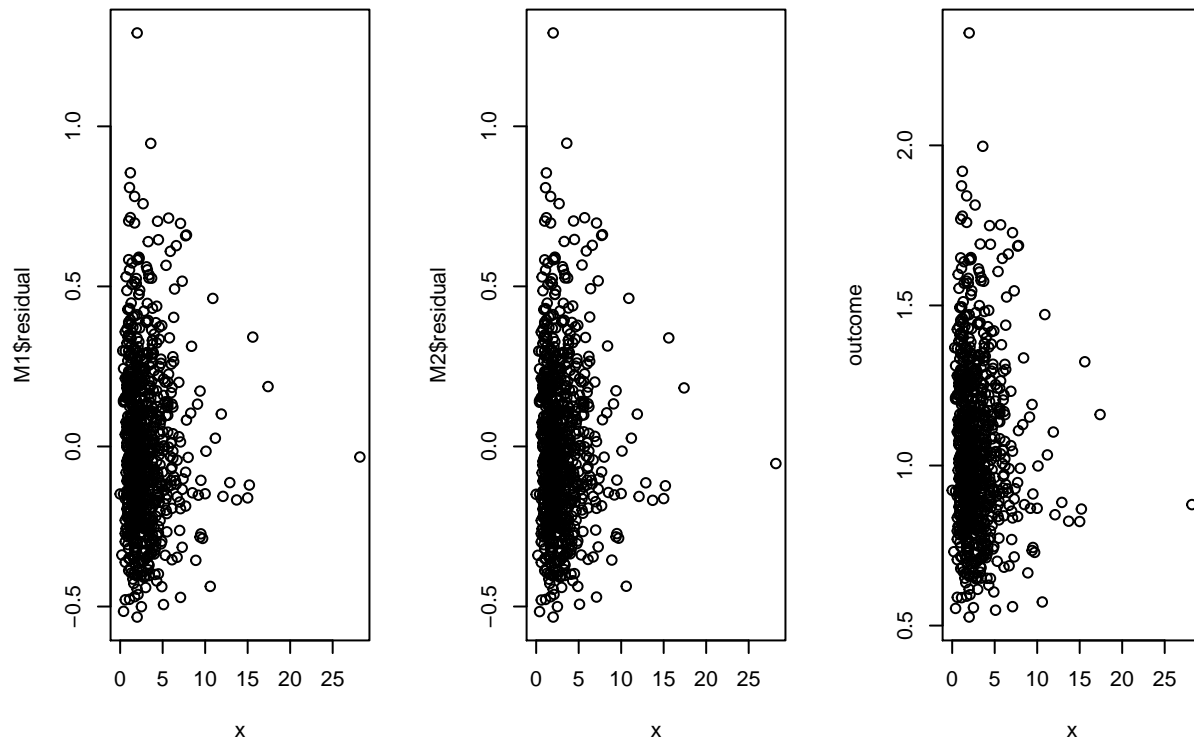
```
## [1] "residual for M1: 0.248704466454944"
## [1] "residual for M2: 0.248847192837983"
```



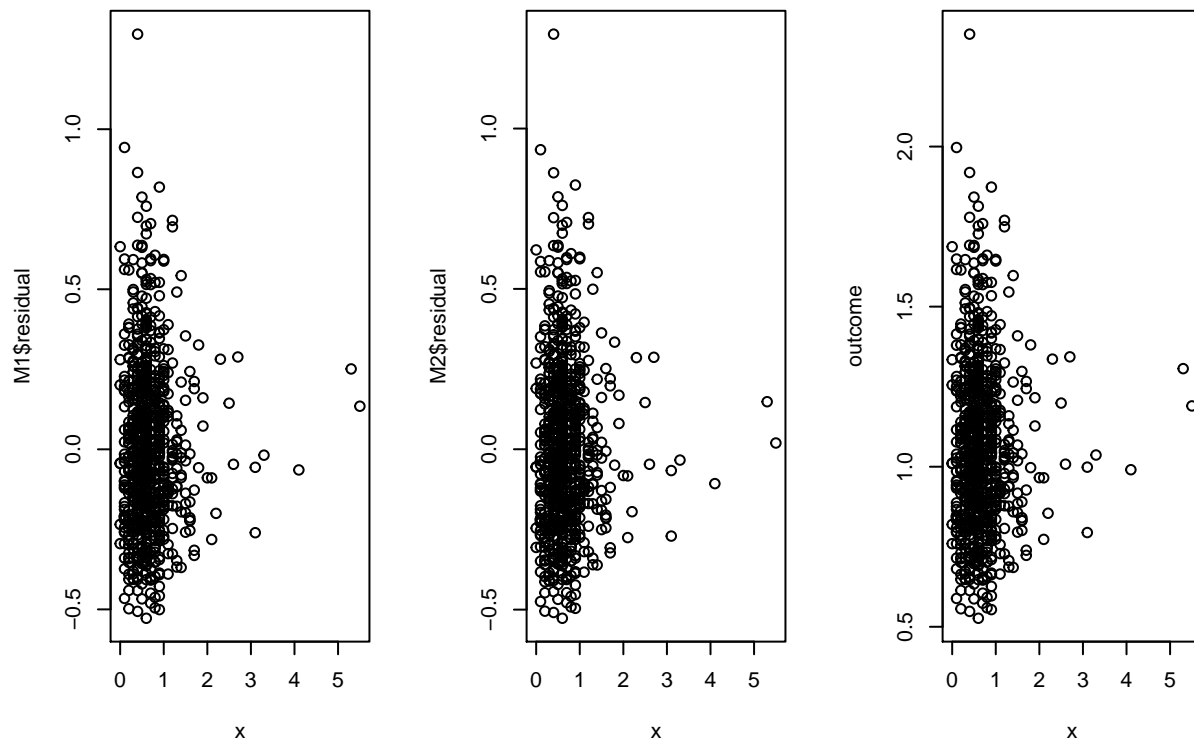
```
## [1] "residual for M1: 0.25026710930793"
## [1] "residual for M2: 0.250393729526099"
```



```
## [1] "residual for M1: 0.250043388210667"
## [1] "residual for M2: 0.25018695270193"
```



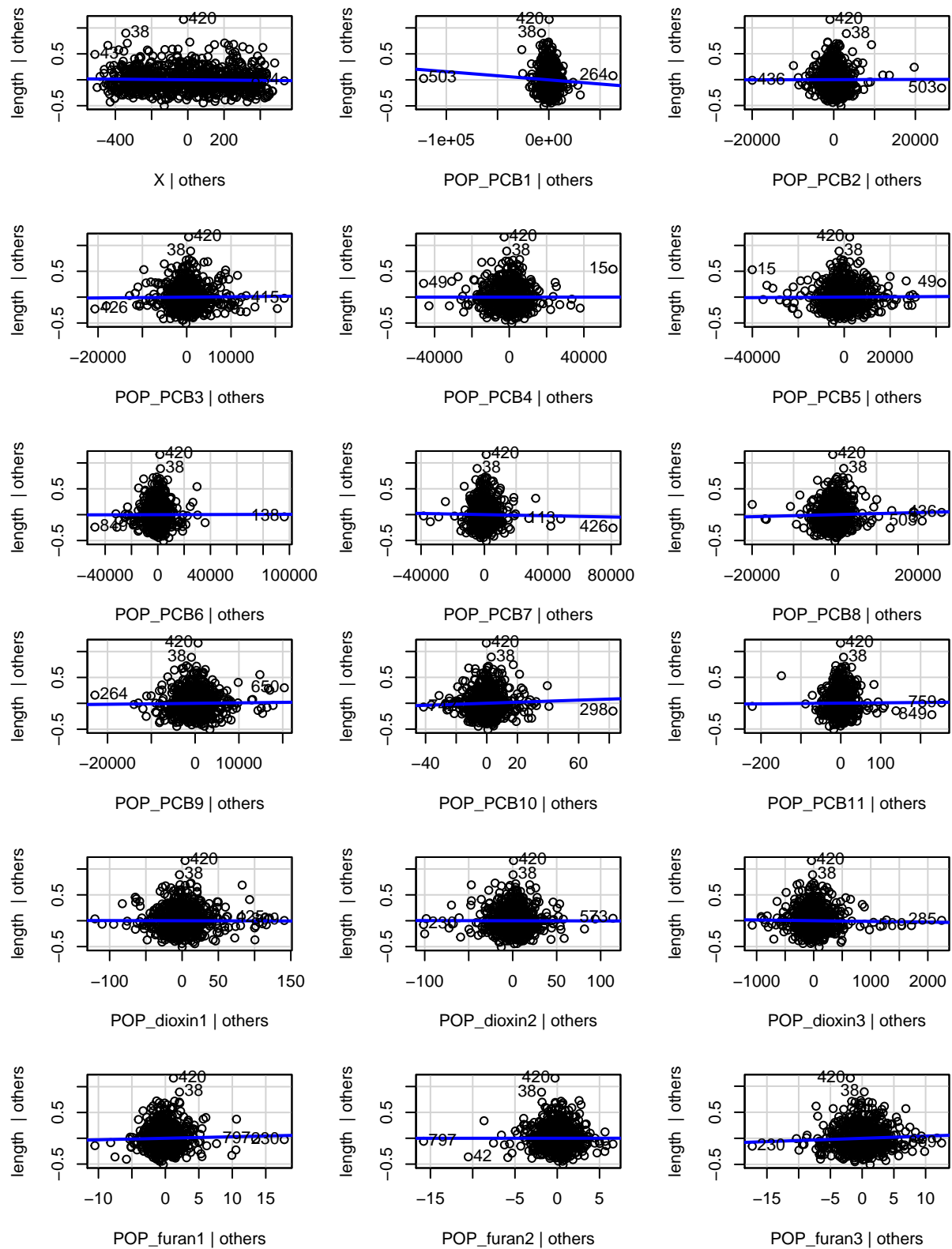
```
## [1] "residual for M1: 0.250382476371691"
## [1] "residual for M2: 0.25042580861039"
```

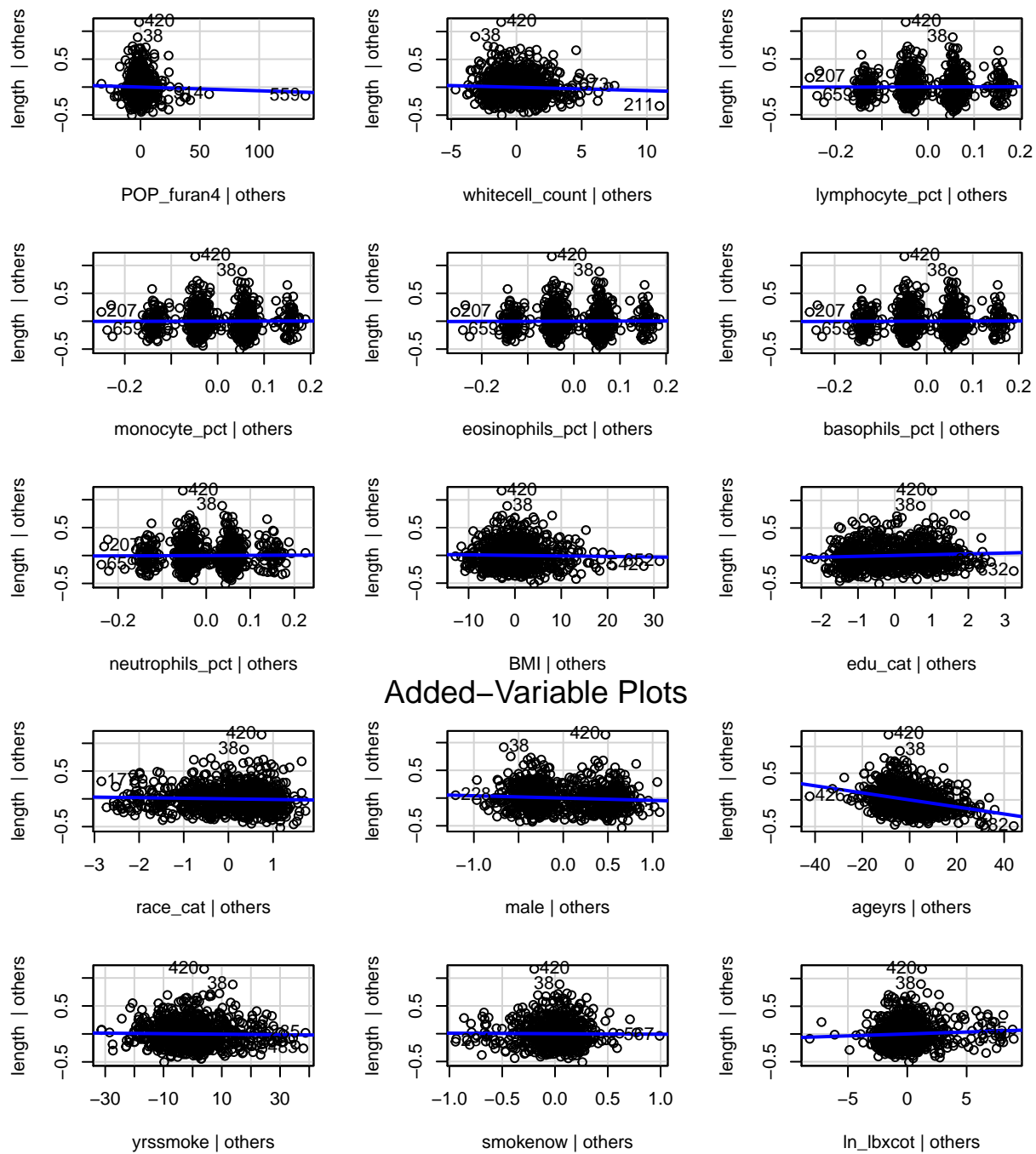


```
# Judy's work Part 2
# testing non-linearity in MLR
library(car)
```

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

```
M <- lm (length ~ ., data=pollutants)
avPlots(M)
```





Added-Variable Plots