

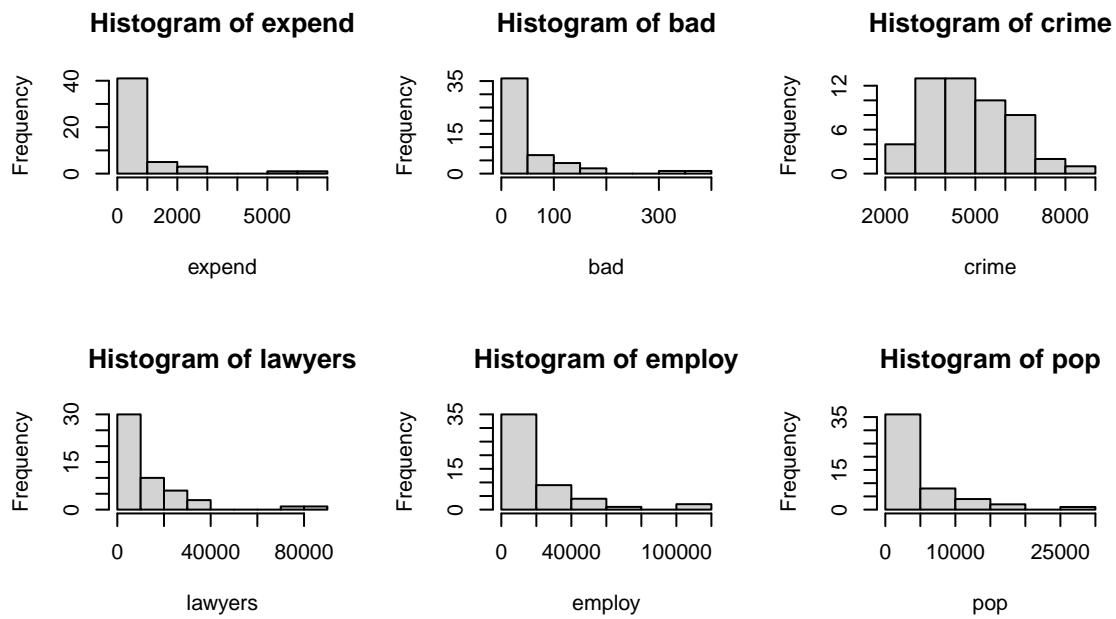
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Xinyu Hu

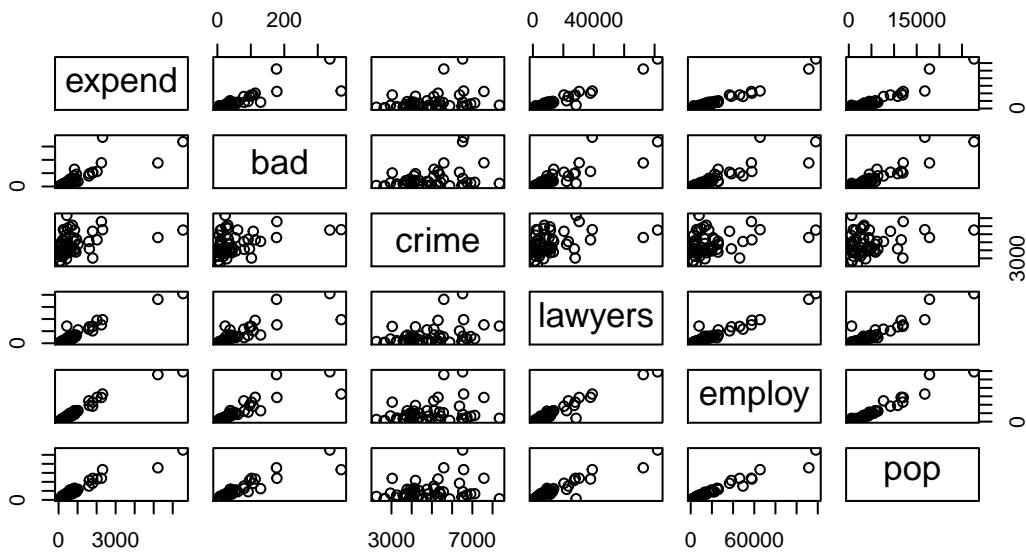
2021/3/5

Exercise 5

a)



In case of finding the potential and influence points, the histograms are shown above. It is clear that the crime factor is normally distributed, and the rest of factors have the similar curve. It shows that there exists collinearity.



```
##          expend  bad  crime lawyers employ  pop
## expend    1.00  0.83  0.33   0.97  0.98 0.95
## bad       0.83  1.00  0.37   0.83  0.87 0.92
## crime     0.33  0.37  1.00   0.38  0.31 0.28
## lawyers   0.97  0.83  0.38   1.00  0.97 0.93
## employ    0.98  0.87  0.31   0.97  1.00 0.97
## pop       0.95  0.92  0.28   0.93  0.97 1.00
```

In the graph, (expend, crime), (bad, crime), (crime, lawyers), (crime, employ), (crime, pop) are not linear independently. And we need to see which predictor variables are involved in collinearity.

```
exlm1 = lm(expend~bad+crime+lawyers+employ+pop, data=ex); vif(exlm1)
```

```
##      bad      crime      lawyers      employ      pop
## 8.364321 1.487978 16.967470 33.591361 32.937517
```

```
exlm2 = lm(expend~crime+lawyers+employ+pop, data=ex); vif(exlm2)
```

```
##      crime      lawyers      employ      pop
## 1.233263 16.372292 33.106158 17.576977
```

```
exlm3 = lm(expend~crime+employ+pop, data=ex); vif(exlm3)
```

```
##      crime      employ      pop
## 1.121163 17.967808 17.568906
```

```
exlm4 = lm(expend~crime+pop, data=ex); vif(exlm4)
```

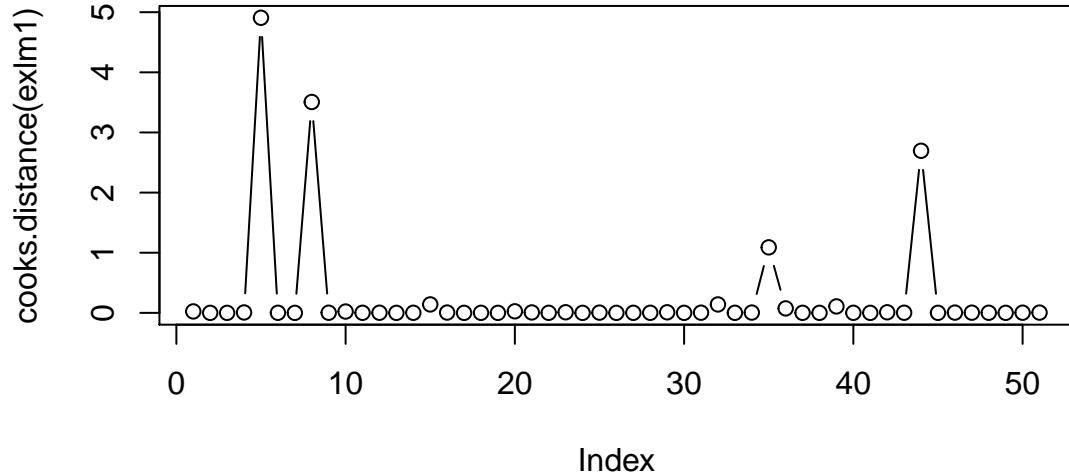
```
##      crime      pop
## 1.08213 1.08213
```

```
exlm5 = lm(expend~crime, data=ex); vif(exlm5)
```

```
# Error in vif.default(exlm5) : model contains fewer than 2 terms
```

In exlm1, exlm2 and exlm3, all VIF's are large, so there is a collinearity problem, but the exlm4 and exlm5 are OK.

```
plot(cooks.distance(exlm1), type="b")
```



```
round(cooks.distance(exlm1), 2)
```

```
##   1    2    3    4    5    6    7    8    9    10   11   12   13   14   15   16  
## 0.02 0.00 0.00 0.01 4.91 0.00 0.00 3.51 0.00 0.02 0.00 0.00 0.00 0.00 0.14 0.01  
##   17   18   19   20   21   22   23   24   25   26   27   28   29   30   31   32  
## 0.00 0.00 0.00 0.03 0.01 0.00 0.01 0.00 0.00 0.00 0.00 0.00 0.01 0.00 0.00 0.14  
##   33   34   35   36   37   38   39   40   41   42   43   44   45   46   47   48  
## 0.00 0.00 1.09 0.07 0.00 0.00 0.11 0.00 0.00 0.01 0.00 2.70 0.00 0.00 0.00 0.00  
##   49   50   51  
## 0.00 0.00 0.00
```

Thus, the potential and influence points are Point(5), Point(8), Point(35) and Point(44).

b)

First, we start with step-up method.

```
summary(lm(expend~bad, data=ex)) [[8]]  
  
## [1] 0.6963839  
summary(lm(expend~crime, data=ex)) [[8]]  
  
## [1] 0.1118564  
summary(lm(expend~lawyers, data=ex)) [[8]]  
  
## [1] 0.9372789  
summary(lm(expend~employ, data=ex)) [[8]]  
  
## [1] 0.9539745  
summary(lm(expend~pop, data=ex)) [[8]]  
  
## [1] 0.9073261
```

The employ has highest value: 0.9539745.

```
summary(lm(expend~employ+bad,data=ex))[[8]]  
  
## [1] 0.955097  
summary(lm(expend~employ+crime,data=ex))[[8]]  
  
## [1] 0.9550501  
summary(lm(expend~employ+pop,data=ex))[[8]]  
  
## [1] 0.95431  
summary(lm(expend~employ+lawyers,data=ex))[[8]]  
  
## [1] 0.9631745
```

The model of $\text{expend} \sim \text{employ} + \text{lawyers}$ has highest value: 0.9631745.

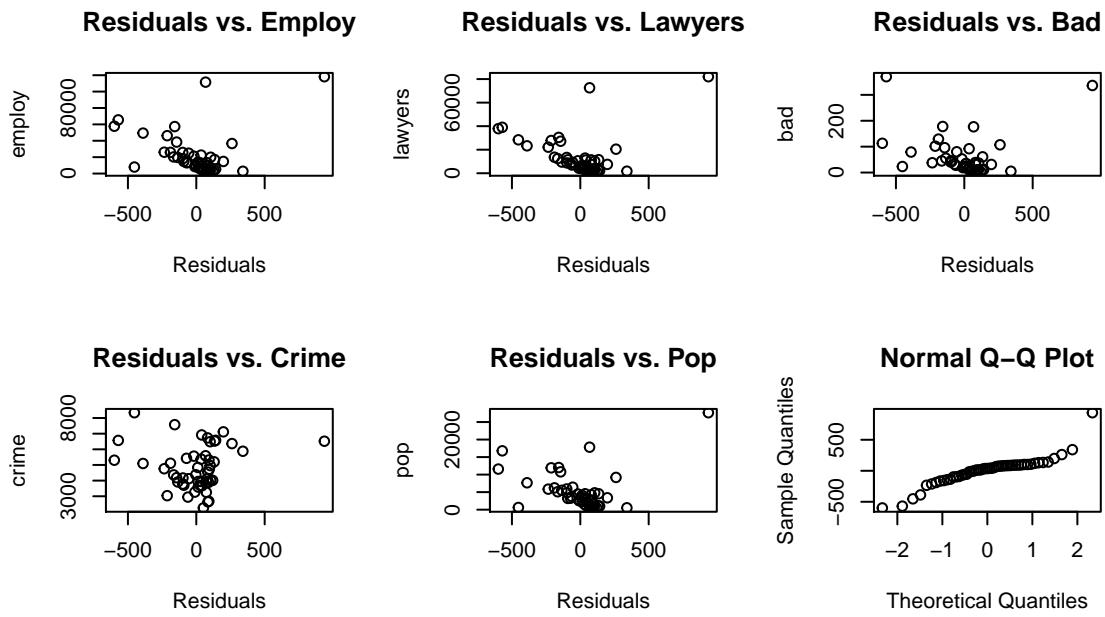
```
summary(lm(expend~employ+lawyers+bad,data=ex))[[8]]  
  
## [1] 0.9638741  
summary(lm(expend~employ+lawyers+crime,data=ex))[[8]]  
  
## [1] 0.9631881  
summary(lm(expend~employ+lawyers+pop,data=ex))[[8]]  
  
## [1] 0.9637326
```

Since the models did not yield any significant results, the step-up method stopped.

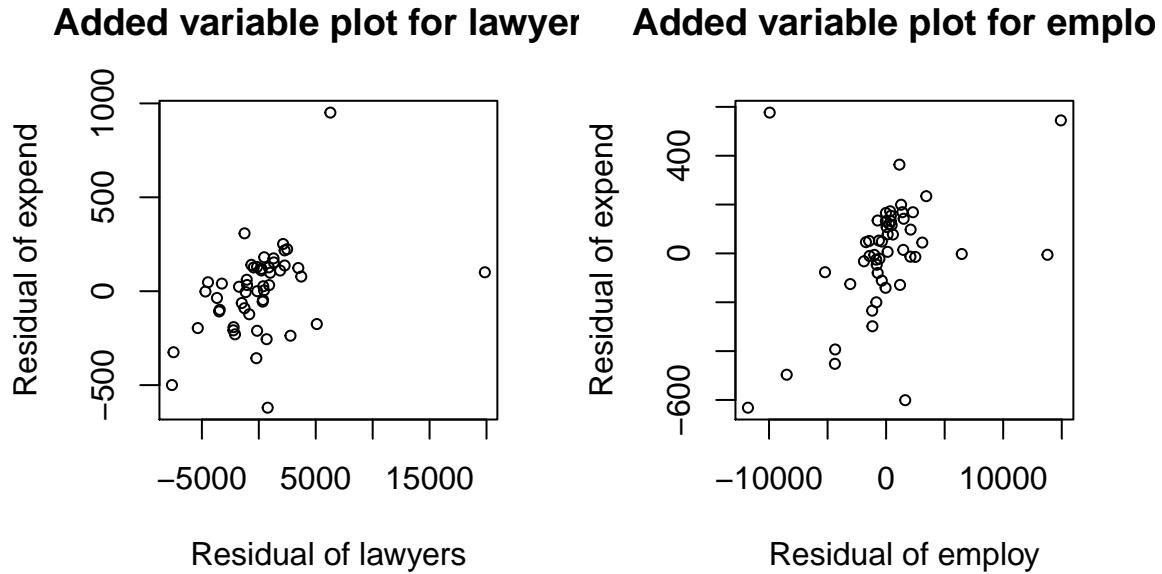
```
summary(lm(expend~bad+crime+lawyers+employ+pop,data=ex))[[8]]  
  
## [1] 0.9675314  
summary(lm(expend~bad+lawyers+employ+pop,data=ex))[[8]]  
  
## [1] 0.9665736  
summary(lm(expend~bad+lawyers+employ,data=ex))[[8]]  
  
## [1] 0.9638741  
summary(lm(expend~lawyers+employ,data=ex))[[8]]  
  
## [1] 0.9631745
```

All of these models did not yield any significant results, so the step-down method stopped. Hence, $\text{expend} \sim \text{lawyers} + \text{employ}$ is the final model for both methods, which $\text{expend} = -110.7 + 0.002971 \times \text{employ} + 0.02686 \times \text{lawyers}$.

c)



From question(a), it already shows the collinearity of dependent and independent variables. The above graphs claims that the spread of residuals against variables did not show such a pattern existing. And the QQ-plot shows the residuals are normally distributed.



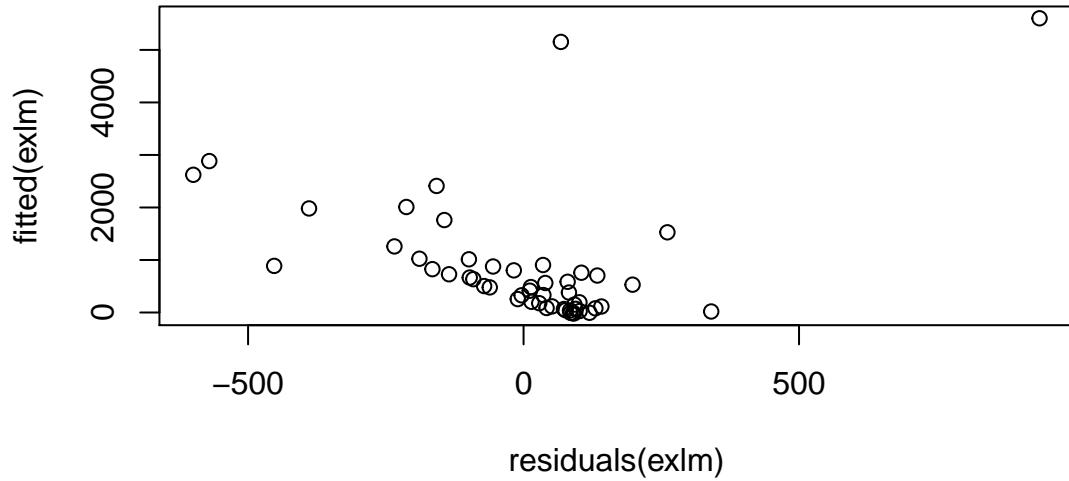
The added variable plots also show that there is no such specific curved pattern visible.

```
shapiro.test(residuals(exlm))
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: residuals(exlm)
```

```
## W = 0.8475, p-value = 1.118e-05
```

The Shapiro-Wilk normality test shows the same as the QQ-plot, which means it is still normally distributed since $p\text{-value}=1.118\text{e-}05 < 0.05$.



Moreover, there is no patterns or errors are visible in the scatter plot of residuals against Y (and \hat{Y}).