## Homework

# SHENDUO ZHANG

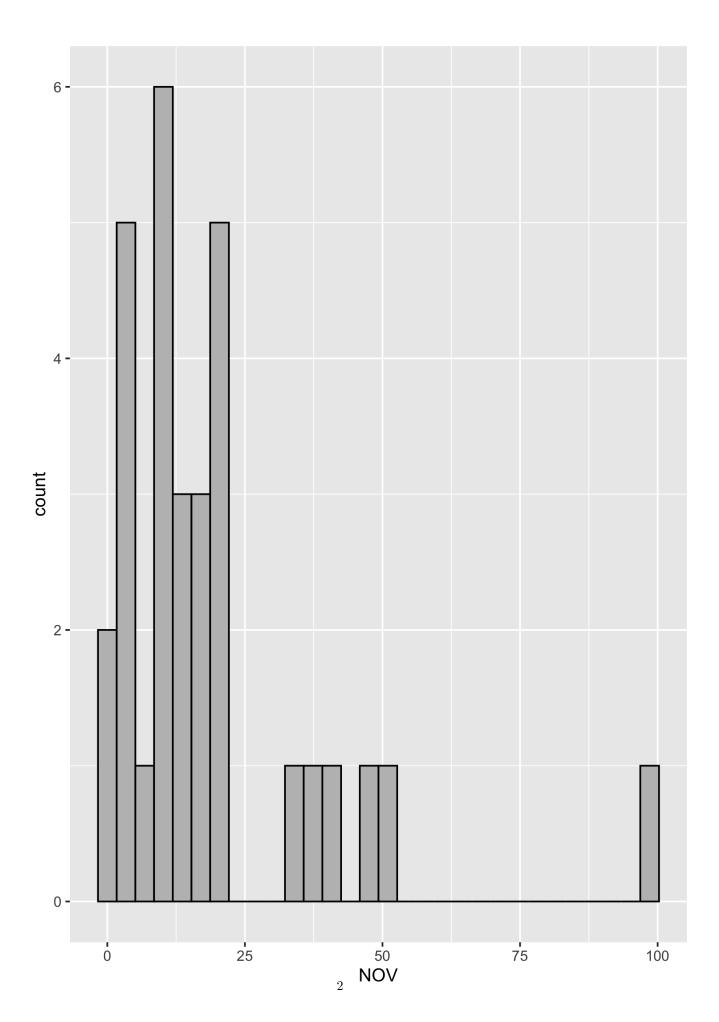
I omit the library and read.table calls because the tex template do not work in Chinese while the file path are in Chinese. Two '.txt' files are imported with the read.table with the original name "exersice1\_4" and "exersice1\_7". One would be expected to have packages ggplot2, Hmisc, moments and their dependency installed if they like to see the code running.

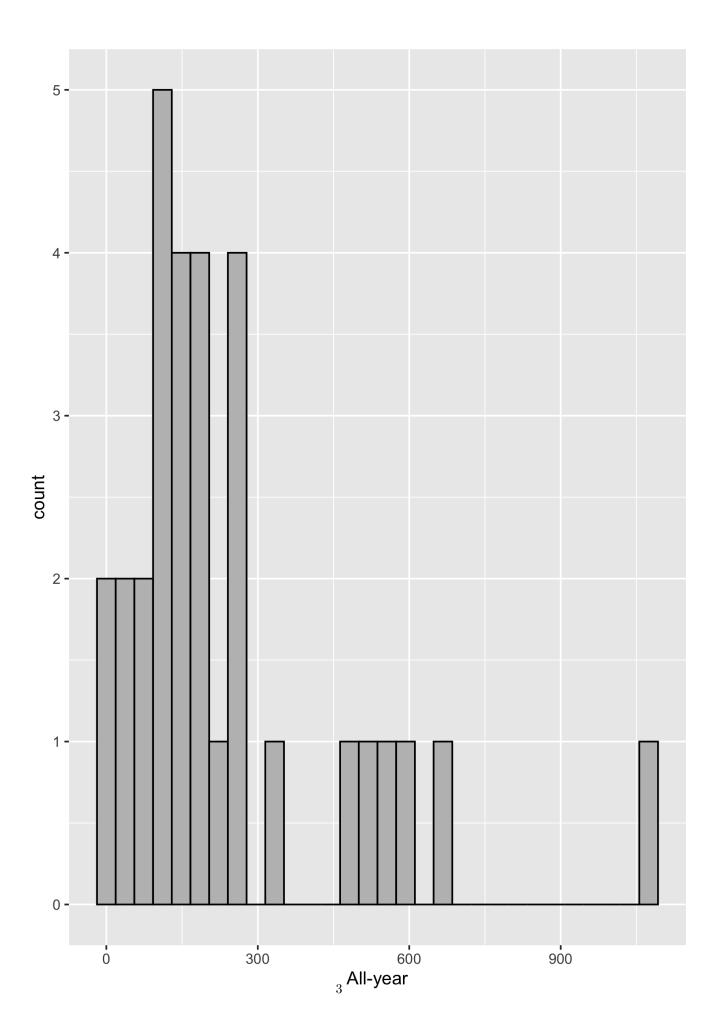
### Problem 1

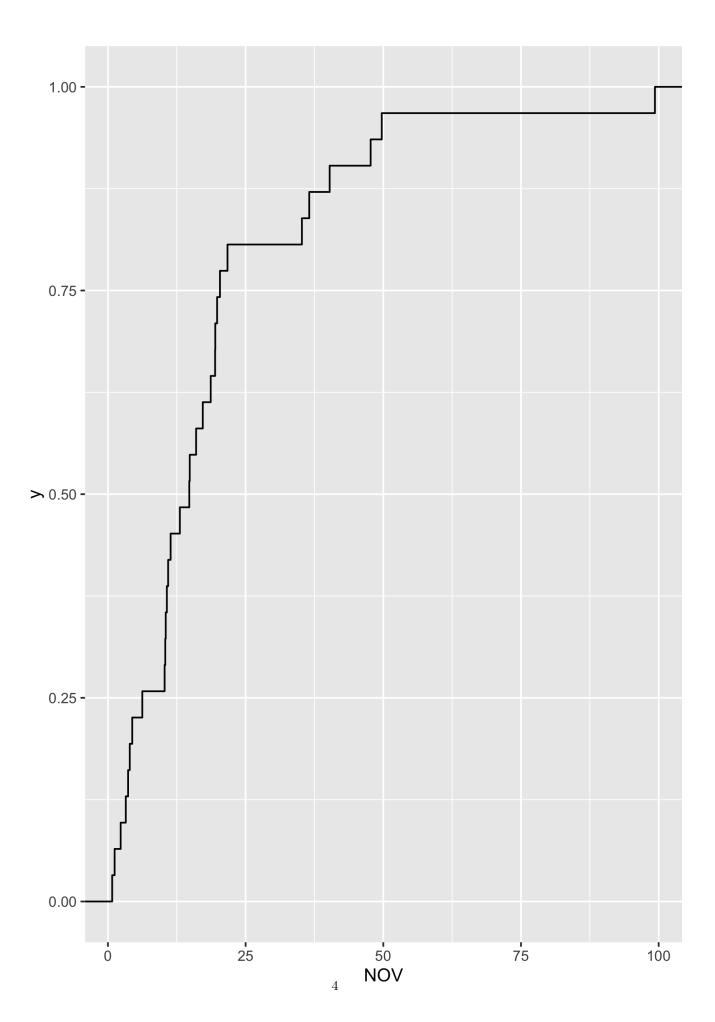
#### Solution 1.a

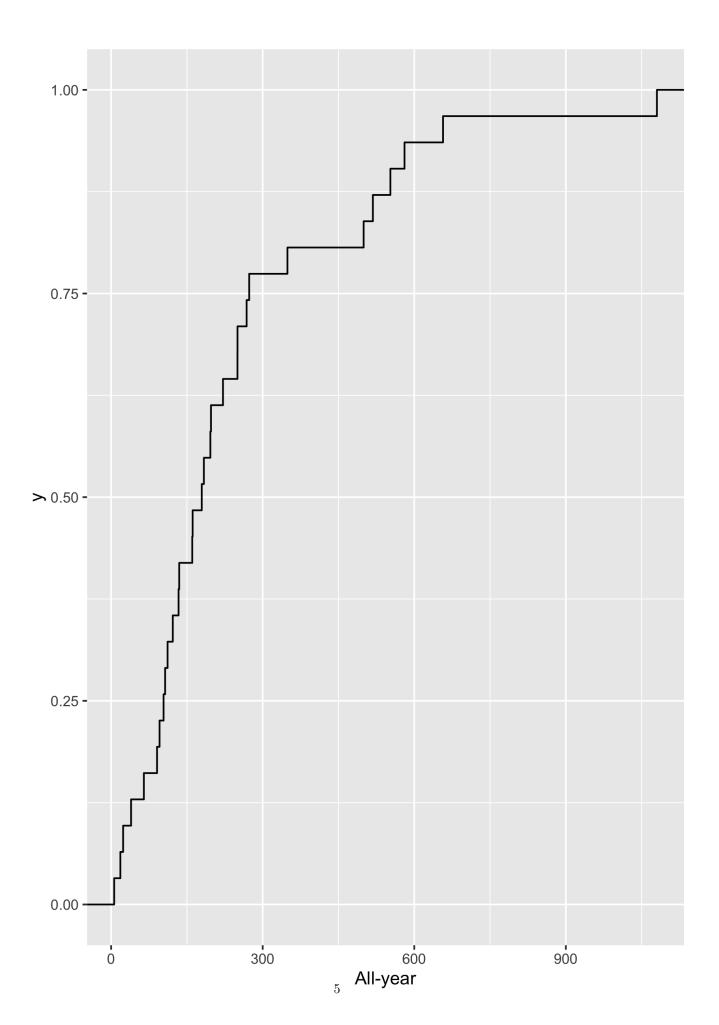
I would gave my result ahead to make everyone's life easier. The vector stands for the result of applying the certain function to  $X_1, X_2$  respectively.

```
\operatorname{Mean}(X_1, X_2) = (19.2, 246.2)
\operatorname{Var}(X_1, X_2) = (392.0308, 54276.9982)
\operatorname{Standard\ deviation}(X_1, X_2) = (19.79977, 233.97210)
\operatorname{Coefficients\ of\ variation}(X_1, X_2) = (1.033043, 12.155202)
\operatorname{Skewness}(X_1, X_2) = (2.391967, 1.821974)
\operatorname{Kurtosis}(X_1, X_2) = (9.804995, 6.521676)
\operatorname{Median}(X_1, X_2) = (14.77, 179.41)
\operatorname{Quantile}(0.25, 0.75)X_1 = (8.265, 20.080)
\operatorname{Quantile}(0.25, 0.75)X_2 = (105.350, 270.745)
\operatorname{Interquantilerange}(X_1, X_2) = (11.815, 165.395)
\operatorname{Correlation}_{peason} = 0.9762474
\operatorname{Correlation}_{spearson} = 0.9278226
```









```
> X_1 <- exersice1_4[,2]</pre>
> X_2 <- exersice1_4[,3]</pre>
> Mean <- c(mean(X_1, na.rm=FALSE),mean(X_1,na.rm=FALSE))</pre>
> Var <- c(var(X_1), var(X_2))
> Standard_Deviation <- c(sd(X_1, na.rm=FALSE),sd(X_2, na.rm=FALSE))
> Coefficents_of_variation <- Standard_Deviation/Mean
> Skewness <- c(skewness(X_1), skewness(X_2))</pre>
> Kurtosis <- c(kurtosis(X_1), kurtosis(X_2))</pre>
> Median <- c(median(X_1),median(X_2))</pre>
> QuantileX_1 = c(quantile(X_1, probs = seq(0, 1, 0.25)), quantile(X_1,0.75))
> QuantileX_2 = c(quantile(X_2, probs = seq(0, 1, 0.25)), quantile(X_2,0.75))
> Interquantilerange <- c(QuantileX_1[2]-QuantileX_1[1],QuantileX_2[2]-QuantileX_2[1])
> Mean
[1] 19.16645 19.16645
> Var
     392.0308 54275.9982
[1]
> Standard_Deviation
[1] 19.79977 232.97210
> Coefficents_of_variation
[1] 1.033043 12.155202
> Skewness
[1] 2.391967 1.821974
> Kurtosis
[1] 9.804995 6.521676
> Median
[1] 14.77 179.41
> QuantileX_1
                 50%
                        75%
                             100%
    0%
          25%
                                       75%
0.770 8.265 14.770 20.080 99.320 20.080
> QuantileX_2
                       50%
                                 75%
      0%
              25%
                                         100%
                                                    75%
   6.080 105.350 179.410 270.745 1080.260 270.745
> Interquantilerange
   25%
          25%
7.495 99.270
> ggplot(exersice1_4,aes(V2)) + geom_histogram(color="black",fill="gray")+xlab("NOV")
'stat_bin()' using 'bins = 30'. Pick better value with
'binwidth'.
> ggsave("Plot1.png")
Saving 5.68 x 8.08 in image
'stat_bin()' using 'bins = 30'. Pick better value with
'binwidth'.
> ggplot(exersice1_4,aes(V3)) + geom_histogram(color="black",fill="gray")+xlab("All-year")
'stat_bin()' using 'bins = 30'. Pick better value with
'binwidth'.
> ggsave("Plot2.png")
Saving 5.68 x 8.08 in image
'stat_bin()' using 'bins = 30'. Pick better value with
'binwidth'.
> ggplot(exersice1_4,aes(V2)) + stat_ecdf(geom = 'step')+xlab("NOV")
> ggsave("Plot3.png")
Saving 5.68 x 8.08 in image
> ggplot(exersice1_4,aes(V3)) + stat_ecdf(geom = 'step')+xlab("All-year")
> ggsave("Plot4.png")
Saving 5.68 x 8.08 in image
> Correlation_pearson <- cor(X_1, X_2, method='pearson')</pre>
> Correlation_spearman <- cor(X_1, X_2, method='spearman')</pre>
> Correlation_pearson
[1] 0.9762474
```

```
> Correlation_spearman
[1] 0.9278226
```

#### Problem 2

I believe this part of code is neat and self explained. By writing the result ahead won't help in any way of convenience.

```
> apply(exercise1_7,MARGIN = 2,FUN=mean )
  V 1
       V2
              V3
14.41 16.02 4.23
> apply(exercise1_7,MARGIN = 2,FUN=median )
V1 V2 V3
15 15 4
> rcorr(as.matrix(exercise1_7), type = 'spearman')
        V2
V1 1.00 0.55 0.51
V2 0.55 1.00 0.53
V3 0.51 0.53 1.00
n = 50
   V 1
        V2
              V3
         0e+00 2e-04
V 1
V2 0e+00
               0e+00
V3 2e-04 0e+00
> rcorr(as.matrix(exersice1_7),type = 'pearson')
     V1 V2 V3
V1 1.00 0.62 0.52
V2 0.62 1.00 0.46
V3 0.52 0.46 1.00
n = 50
  V 1
        V2
             V3
V 1
         0e+00 1e-04
V2 0e+00
              7e-04
V3 1e-04 7e-04
```

No matter which techniques we choose, the P-value is so small that with significance level as low as 0.001, one can infer that there's a linear correlation among those three features  $X_1, X_2, X_3$ .