

# Tierion

```

1. library(magrittr)
2. library(dplyr)
3. library(ggplot2)
4. library(readr)
5. library(fitdistrplus)
6. library(DAAG)
7. library("ggplot2")
8. library(anytime)
9.
10. tierion <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics/Project/Blockchain-Tokens-Data-Analytics/networktierionTX.txt', delim = " ", col_names = F)
11. names(tierion) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
12. decimals <- 10^8
13. supply <- 1 * 10^9
14. tierionFiltered <-
  filter(tierion, tokenAmount < decimals * supply) #filter out all outliers
15.
16. #figure out how many users induced those unnormal transaction
17. tierion_outliers <- filter(tierion, tokenAmount >= decimals * supply)
18. user_outliers <- tierion_outliers %>% group_by(toID) %>% summarise(nn = n()) %>% ungroup
19. number_users_outliers <- nrow(user_outliers)
20. number_users_outliers
21.
22. #get top X buyers data
23. buys <-
  tierionFiltered %>% group_by(toID) %>% summarise(nn = n()) %>% ungroup #change the supply and decimals amount
24. buys_sorted_dec <- buys[order(-buys$n),]
25. #top 30 active buyers and number of buys
26. top_30_buyers <- buys_sorted_dec %>% head(30)
27. top_30_buyers
28.
29. #####Question 1#####
30.
31. #####group by user pairs#####
32. buys_pairs <-
  tierionFiltered %>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
33. for (row in 1:nrow(buys_pairs)) {
34.   a <- buys_pairs[row, "fromID"]
35.   b <- buys_pairs[row, "toID"]
36.   for (inner_row in row:nrow(buys_pairs)) {
37.     c <- buys_pairs[inner_row, "fromID"]
38.     d <- buys_pairs[inner_row, "toID"]
39.     if (a == d && b == c) {
40.       buys_pairs[inner_row, "fromID"] <- d
41.       buys_pairs[inner_row, "toID"] <- c
42.     }
43.   }
44. }

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45. buys_pairs<-
  tierionFiltered%>% group_by(fromID*toID+fromID+toID) %>% summarise(nn = n()) %>% ungroup
  p
46. buys_pairs<-
  tierionFiltered%>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
47. buys_pair_sorted_asc<-buys_pairs[order(buys_pairs$n),]
48. buys_pair_less_30<-subset(buys_pair_sorted_asc,n<30)
49.
50. #####find out estimates of paramaters of several distribution based on the buys_pairs d
    ata set#####
51. exp_dis <- fitdist(buys_pair_less_30$n, 'exp')
52. exp_dis
53. gamma_dis <- fitdist(buys_pair_less_30$n, 'gamma')
54. gamma_dis
55. lnorm_dis <- fitdist(buys_pair_less_30$n, 'lnorm')
56. lnorm_dis
57. pois_dis <- fitdist(buys_pair_less_30$n, 'pois')
58. pois_dis
59. weibull_dis <- fitdist(buys_pair_less_30$n, 'weibull')
60. weibull_dis
61.
62. gofstat(list(exp_dis, gamma_dis, lnorm_dis))
63. descdist(buys_sorted_asc$n,boot=1000)
64.
65. #lognorm
66. fit_lnorm <- fitdist(buys_pair_less_30$n,"lnorm")
67. summary(fit_lnorm)
68. plot(fit_lnorm)
69. cdfcomp(fit_lnorm)
70.
71. #exp
72. fit_exp <- fitdist(buys_pair_less_30$n,"exp")
73. summary(fit_exp)
74. plot(fit_exp)
75. cdfcomp(fit_exp)
76.
77. #gamma
78. fit_gamma <- fitdist(buys_pair_less_30$n,"gamma")
79. summary(fit_gamma)
80. plot(fit_gamma)
81. cdfcomp(fit_gamma)
82.
83. #weibull
84. fit_weibull <- fitdist(buys_pair_less_30$n,"weibull")
85. summary(fit_weibull)
86.
87. #normal
88. fit_normal <- fitdist(buys_pair_less_30$n,"norm")
89. summary(fit_normal)
90.
91. #pois
92. #normal
93. fit_pois <- fitdist(buys_pair_less_30$n,"pois")
94. summary(fit_pois)
95.
96. #unif
97. fit_unif <- fitdist(buys_pair_less_30$n,"unif")
98. summary(fit_unif)
99. plot(fit_unif)
100.      cdfcomp(fit_unif)
101.

```

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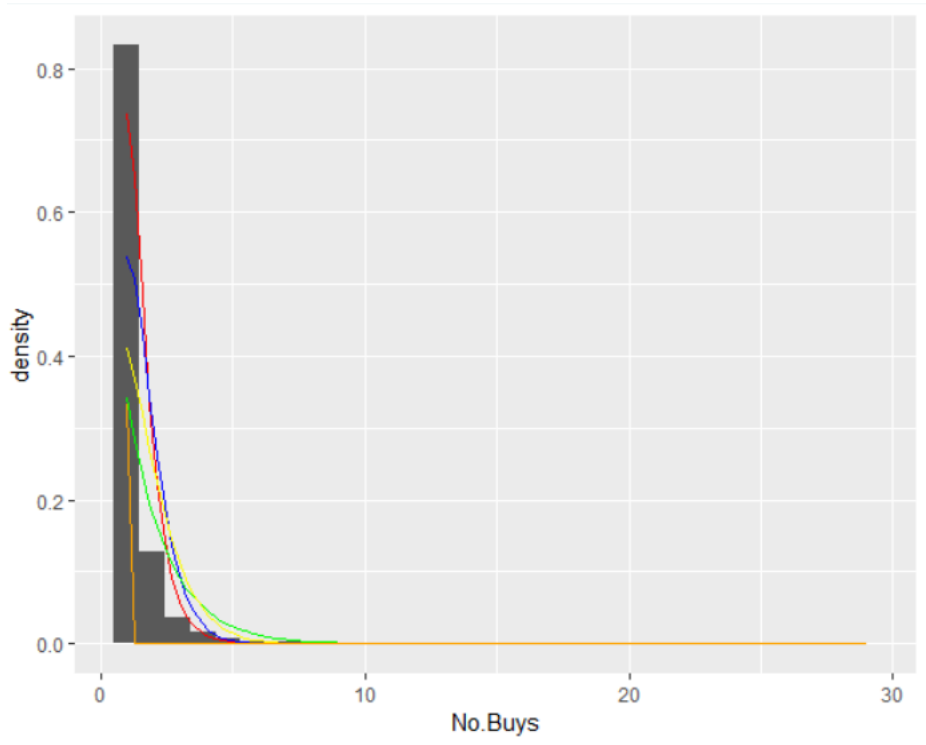
102. #####draw graph#####
103. all_density <- ggplot(data=buys_pair_less_30) +
104.   geom_histogram(bins=30,aes(x = buys_pair_less_30$n, ..density..)) +
105.   stat_function(fun = dlnorm, args = list(meanlog = 0.2373987, sdlog = 0.4791316
),
106.               colour = "red")+
107.   stat_function(fun = dgamma, args = list(shape = 3.020395, rate=2.000602),
108.               colour = "blue")+
109.   stat_function(fun=dexp, args=list(rate=0.6623558),colour="green")+
110.   stat_function(fun=dweibull, args=list(shape=1.360851, scale=1.678697),colour="
yellow")+
111.   stat_function(fun=dpois, args=list(lambda=1.509763),colour="orange")+ xlab("N
o.Buys")
112. all_density
113.
114. #####Question 1#####
115.
116.
117. #####Question 2#####
118. tierion_prices <- read_delim("C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/St
atistics/Project/Blockchain-Tokens-Data-
Analytics/tierion", delim = "\t", col_names = T) #load token price data
119. names(tierion_prices) <- make.names(names(tierion_prices))
120. tierion_prices <- tierion_prices %>% mutate(date = as.Date(Date, format = '%m/%d
/%Y'))
121.
122. tierion <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistic
s/Project/Blockchain-Tokens-Data-
Analytics/networktierionTX.txt', delim = " ", col_names = F)
123. names(tierion) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
124. decimals <- 10^8
125. supply <- 1 * 10^9
126. tierion_filtered <-filter(tierion,tokenAmount < decimals * supply)
127. ## convert data type of unixTime
128. tierion_filtered <- tierion_filtered %>%
129.   mutate(date = anydate(unixTime))
130. names(tierion_filtered) <- c('fromID', 'toID', 'unixTime', 'tokenAmount', 'date'
)
131.
132. ## merge the prices and edge
133. tierion_merged<-
  merge(x = tierion_prices, y = tierion_filtered, by = "date", all.x = TRUE)
134.
135. #####Determin K#####
136. top_30_buyers<-buys_sorted_dec%>%head(30)
137.
138. top_K<-c(1:30)
139. count <- 1
140. for (val in top_K) {
141.   top_K_buyers<-buys_sorted_dec%>%head(val)
142.   filter_K_tierion_merged<-filter(tierion_merged,toID %in% top_K_buyers$toID)
143.   filter_K_tierion_merged=transform(filter_K_tierion_merged,average_price= (Open
+Close)/2)
144.   filter_K_tierion_merged$num_Date <- as.numeric(as.POSIXct(filter_K_tierion_me
rged$date))
145.   filered<-
    filter_K_tierion_merged%>% group_by(num_Date) %>% summarise(nn = n(),Close=mean(Close),
    tokenAmount=sum(tokenAmount),Open=mean(Open))
146.   shift <- function(x, n){

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147.      c(x[-(seq(n))], rep(NA, n))
148.    }
149.    filered$new_Close<-shift(filered$Close,1)
150.    num_rows<-nrow(filered)
151.    filered[-num_rows,]
152.    regression<-lm(filered$new_Close~filered$tokenAmount+filered$n+filered$Open)
153.
154.    setwd("C:/Users/ygaoq/Desktop/Tierion")
155.    yourfilename=paste("W",val,".txt",sep="")
156.    capture.output(summary(regression),append = TRUE,file = "C:/Users/ygaoq/Desktop/
p/Tierion/Final_Result.txt")
157.
158.
159.    summary(regression)
160.    par(mfcol=c(2,2))
161.    setwd("C:/Users/ygaoq/Desktop/Tierion")
162.    yourfilename=paste("A",val,".png",sep="")
163.    png(file=yourfilename)
164.    opar <- par(mfrow=c(2,2))
165.    plot(regression)
166.    dev.off()
167.  }

```



```

> exp_dis <- fitdist(buys_pair_less_30$n, 'exp')
> exp_dis
Fitting of the distribution ' exp ' by maximum likelihood
Parameters:
      estimate Std. Error
rate 0.6623558 0.002165898
> gamma_dis <- fitdist(buys_pair_less_30$n, 'gamma')
> gamma_dis
Fitting of the distribution ' gamma ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 3.020395 0.013268906
rate 2.000602 0.009561027
> lnorm_dis <- fitdist(buys_pair_less_30$n, 'lnorm')
> lnorm_dis
Fitting of the distribution ' lnorm ' by maximum likelihood
Parameters:
      estimate Std. Error
meanlog 0.2373987 0.001566760
sdlog 0.4791316 0.001107845
> pois_dis <- fitdist(buys_pair_less_30$n, 'pois')
> pois_dis
Fitting of the distribution ' pois ' by maximum likelihood
Parameters:
      estimate Std. Error
lambda 1.509763 0.004017926
> weibull_dis <- fitdist(buys_pair_less_30$n, 'weibull')
> weibull_dis
Fitting of the distribution ' weibull ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 1.360851 0.002649363
scale 1.678697 0.004299194
> gofstat(list(exp_dis, gamma_dis, lnorm_dis, pois_dis, weibull_dis))
Goodness-of-fit statistics

```

	1-mle-exp	2-mle-gamma	3-mle-lnorm	4-mle-pois	5-mle-weibull
Kolmogorov-Smirnov statistic	5.293769e-01	0.4896604	4.938013e-01	0.6173609	0.428102
Cramer-von Mises statistic	5.463616e+03	4842.9805265	4.879257e+03	7366.9122883	4579.661605
Anderson-Darling statistic	2.492568e+04	Inf	2.281400e+04	Inf	Inf

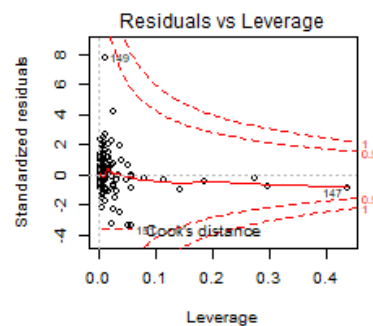
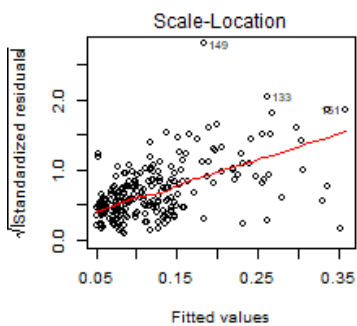
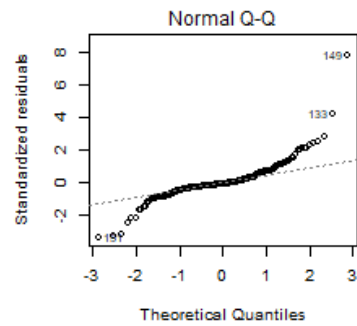
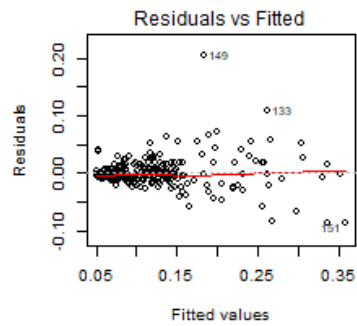
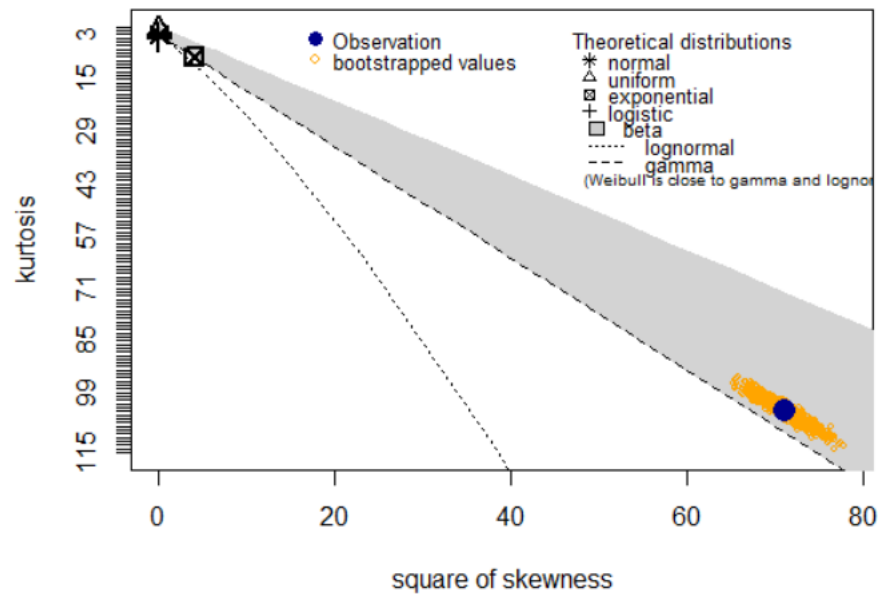
```

Goodness-of-fit criteria

```

	1-mle-exp	2-mle-gamma	3-mle-lnorm	4-mle-pois	5-mle-weibull
Akaike's Information Criterion	228918.9	163531.8	117685.7	237643.8	201455.2
Bayesian Information Criterion	228928.3	163550.6	117704.5	237653.2	201474.0

## Cullen and Frey graph



```

Call:
lm(formula = filtered$new_Close ~ filtered$tokenAmount + filtered$n +
    filtered$Open)

Residuals:
      Min       1Q   Median       3Q      Max
-0.088379 -0.009174 -0.003581  0.007236  0.206421

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    8.838e-03  3.386e-03   2.610  0.00959 **
filtered$tokenAmount -9.105e-19  2.324e-18  -0.392  0.69548
filtered$n       3.239e-06  9.140e-06   0.354  0.72334
filtered$Open    9.211e-01  2.441e-02  37.727 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.02653 on 248 degrees of freedom
(1 observation deleted due to missingness)
Multiple R-squared:  0.8571,    Adjusted R-squared:  0.8554
F-statistic: 495.8 on 3 and 248 DF,  p-value: < 2.2e-16

```

## Aragon

```

1. library(magrittr)
2. library(dplyr)
3. library(ggplot2)
4. library(readr)
5. library(fitdistrplus)
6. library(DAAG)
7. library("ggplot2")
8. library(anytime)
9.
10. aragon <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics/Project/Blockchain-Tokens-Data-Analytics/networkaragonTX.txt', delim = " ", col_names = F)
11. names(aragon) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
12. decimals <- 10^18
13. supply <- 39609523
14. aragonFiltered <-
    filter(aragon, tokenAmount < decimals * supply) #filter out all outliers
15.

```

```

16. #figure out how many users induced those unnormal transaction
17. aragon_outliers<- filter(aragon,tokenAmount >= decimals * supply)
18. user_outliers <- aragon_outliers %>% group_by(toID) %>% summarise(nn = n()) %>% ungroup

19. number_users_outliers<-nrow(user_outliers)
20. number_users_outliers
21.
22. #get top X buyers data
23. buys<-
  aragonFiltered%>% group_by(toID) %>% summarise(nn = n()) %>% ungroup #change the supply
  and decimals amount
24. buys_sorted_dec<-buys[order(-buys$n),]
25. #top 30 active buyers and number of buys
26. top_30_buyers<-buys_sorted_dec%>%head(30)
27. top_30_buyers
28.
29. #####Question 1#####
30.
31. #####group by user pairs####
32. buys_pairs<-
  aragonFiltered%>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
33. for (row in 1:nrow(buys_pairs)) {
34.   a<-buys_pairs[row,"fromID"]
35.   b<-buys_pairs[row,"toID"]
36.   for (inner_row in row:nrow(buys_pairs)) {
37.     c<-buys_pairs[inner_row,"fromID"]
38.     d<-buys_pairs[inner_row,"toID"]
39.     if(a==d&&b==c){
40.       buys_pairs[inner_row,"fromID"]<-d
41.       buys_pairs[inner_row,"toID"]<-c
42.     }
43.   }
44. }
45. buys_pairs<-
  aragonFiltered%>% group_by(fromID*toID+fromID+toID) %>% summarise(nn = n()) %>% ungroup

46. buys_pairs<-
  aragonFiltered%>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
47. buys_pair_sorted_asc<-buys_pairs[order(buys_pairs$n),]
48. buys_pair_less_30<-subset(buys_pair_sorted_asc,n<30)
49. buys_pair_data<-buys_pair_less_30
50.
51. #####find out estimates of parameters of several distribution based on the buys_pairs d
  ata set#####
52. exp_dis <- fitdist(buys_pair_data$n, 'exp')
53. exp_dis
54. gamma_dis <- fitdist(buys_pair_data$n, 'gamma')
55. gamma_dis
56. lnorm_dis <- fitdist(buys_pair_data$n, 'lnorm')
57. lnorm_dis
58. pois_dis <- fitdist(buys_pair_data$n, 'pois')
59. pois_dis
60. weibull_dis <- fitdist(buys_pair_data$n, 'weibull')
61. weibull_dis
62.
63. gofstat(list(exp_dis, gamma_dis, lnorm_dis))
64. descdist(buys_sorted_asc$n,boot=1000)
65.
66. #lognorm
67. fit_lnorm <- fitdist(buys_pair_less_30$n,"lnorm")

```



```

68. summary(fit_lnorm)
69. plot(fit_lnorm)
70. cdfcomp(fit_lnorm)
71.
72. #exp
73. fit_exp <- fitdist(buys_pair_less_30$n,"exp")
74. summary(fit_exp)
75. plot(fit_exp)
76. cdfcomp(fit_exp)
77.
78. #gamma
79. fit_gamma <- fitdist(buys_pair_less_30$n,"gamma")
80. summary(fit_gamma)
81. plot(fit_gamma)
82. cdfcomp(fit_gamma)
83.
84. #weibull
85. fit_weibull <- fitdist(buys_pair_less_30$n,"weibull")
86. summary(fit_weibull)
87.
88. #normal
89. fit_normal <- fitdist(buys_pair_less_30$n,"norm")
90. summary(fit_normal)
91.
92. #pois
93. #normal
94. fit_pois <- fitdist(buys_pair_less_30$n,"pois")
95. summary(fit_pois)
96.
97. #unif
98. fit_unif <- fitdist(buys_pair_less_30$n,"unif")
99. summary(fit_unif)
100.      plot(fit_unif)
101.      cdfcomp(fit_unif)
102.
103.      #####draw graph#####
104.      all_density <- ggplot(data=buys_pair_less_30) +
105.        geom_histogram(bins=30,aes(x = buys_pair_less_30$n, ..density..)) +
106.        stat_function(fun = dlnorm, args = list(meanlog = 0.1537565, sdlog = 0.4012176
107.      ),
108.        colour = "red")+
109.        stat_function(fun = dgamma, args = list(shape = 4.034757, rate=3.040920),
110.        colour = "blue")+
111.        stat_function(fun=dexp, args=list(rate=0.7536976),colour="green")+
112.        stat_function(fun=dweibull, args=list(shape=1.464309, scale=1.488009),colour="
113.      yellow")+
114.        stat_function(fun=dpois, args=list(lambda=1.326792),colour="orange")+
115.        xlab("No.Buys")
116.      all_density
117.
118.      #####Question 1#####
119.
120.      #####Question 2#####
121.      aragon_prices <- read_delim("C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics/Project/Blockchain-Tokens-Data-Analytics/aragon", delim = "\t", col_names = T) #load token price data
122.      names(aragon_prices) <- make.names(names(aragon_prices))

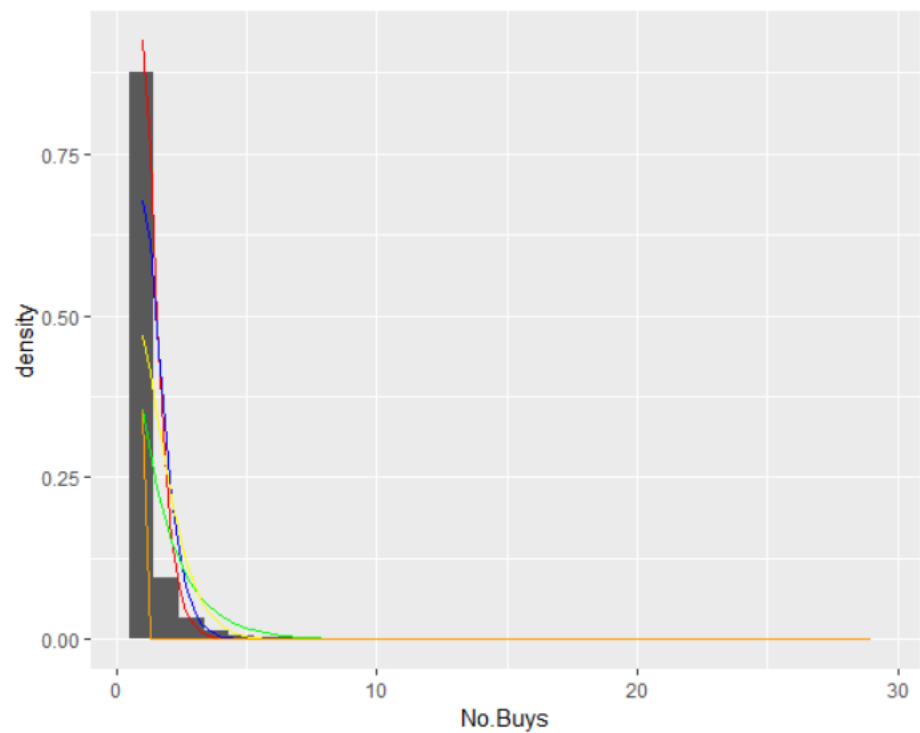
```

```

122.     aragon_prices <- aragon_prices %>% mutate(date = as.Date(Date, format = '%m/%d/%
Y'))
123.
124.     aragon <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics
/Project/Blockchain-Tokens-Data-
Analytics/networkkaragonTX.txt', delim = " ", col_names = F)
125.     names(aragon) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
126.     decimals <- 10^18
127.     supply <- 39609523
128.     aragon_filtered <- filter(aragon, tokenAmount < decimals * supply)
129.     ## convert data type of unixTime
130.     aragon_filtered <- aragon_filtered %>%
131.       mutate(date = anydate(unixTime))
132.     names(aragon_filtered) <- c('fromID', 'toID', 'unixTime', 'tokenAmount', 'date')

133.
134.     ## merge the prices and edge
135.     aragon_merged <-
merge(x = aragon_prices, y = aragon_filtered, by = "date", all.x = TRUE)
136.
137.     #####Determin K#####
138.     top_30_buyers <- buys_sorted_dec %>% head(30)
139.
140.     top_K <- c(1:30)
141.     count <- 1
142.     for (val in top_K) {
143.       top_K_buyers <- buys_sorted_dec %>% head(val)
144.       filter_K_aragon_merged <- filter(aragon_merged, toID %in% top_K_buyers$toID)
145.       filter_K_aragon_merged = transform(filter_K_aragon_merged, average_price = (Open + C
lose)/2)
146.       filter_K_aragon_merged$num_Date <- as.numeric(as.POSIXct(filter_K_aragon_merg
ed$date))
147.       filered <-
filter_K_aragon_merged %>% group_by(num_Date) %>% summarise(nn = n(), Close = mean(Close), t
okenAmount = sum(tokenAmount), Open = mean(Open))
148.       shift <- function(x, n){
149.         c(x[-(seq(n))], rep(NA, n))
150.       }
151.       filered$new_Close <- shift(filered$Close, 1)
152.       num_rows <- nrow(filered)
153.       filered[-num_rows,]
154.       regression <- lm(filered$new_Close ~ filered$tokenAmount + filered$n + filered$Open)
155.
156.       setwd("C:/Users/ygaoq/Desktop/aragon")
157.       yourfilename = paste("W", val, ".txt", sep = "")
158.       capture.output(summary(regression), append = TRUE, file = "C:/Users/ygaoq/Deskto
p/aragon/Final_Result.txt")
159.
160.
161.       summary(regression)
162.       par(mfcol = c(2, 2))
163.       setwd("C:/Users/ygaoq/Desktop/aragon")
164.       yourfilename = paste("A", val, ".png", sep = "")
165.       png(file = yourfilename)
166.       opar <- par(mfrow = c(2, 2))
167.       plot(regression)
168.       dev.off()
169.     }

```

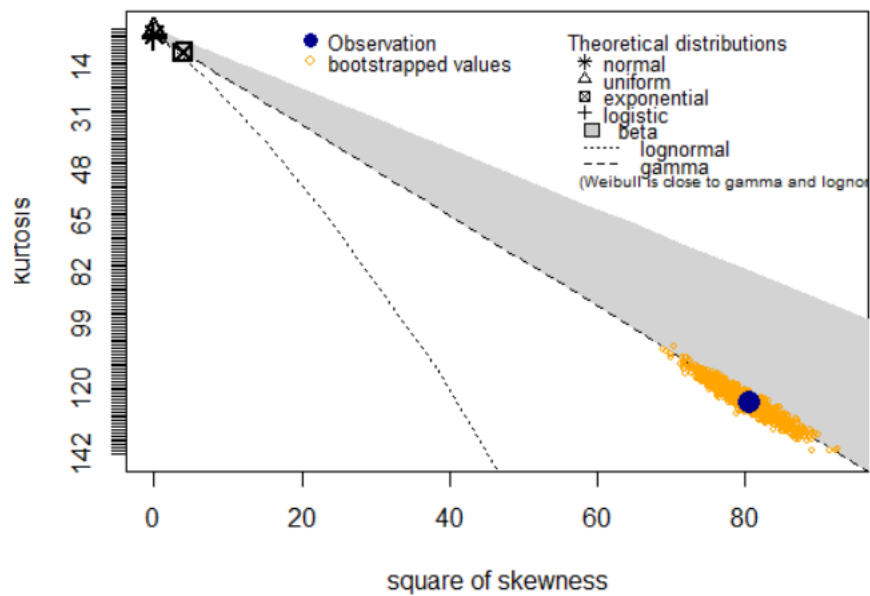


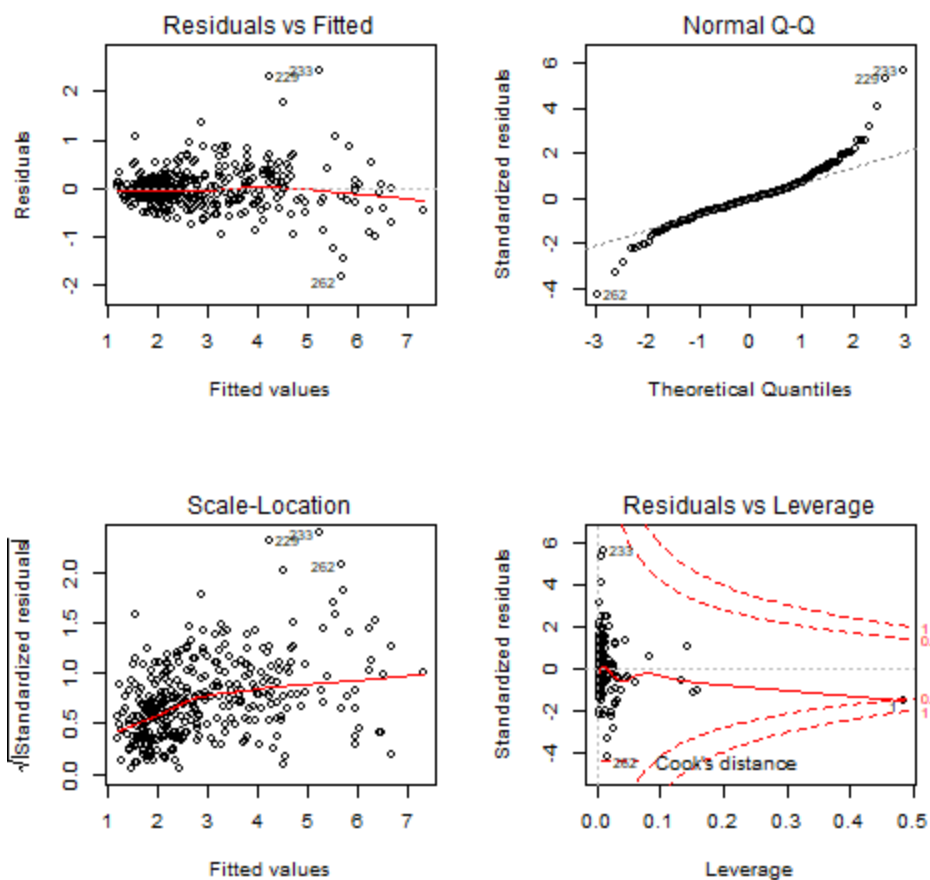
```
> exp_dis <- fitdist(buys_pair_data$n, 'exp')
> exp_dis
Fitting of the distribution ' exp ' by maximum likelihood
Parameters:
      estimate Std. Error
rate 0.7047821 0.002233351
> gamma_dis <- fitdist(buys_pair_data$n, 'gamma')
> gamma_dis
Fitting of the distribution ' gamma ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 3.354202 0.01434705
rate 2.364093 0.01090802
> lnorm_dis <- fitdist(buys_pair_data$n, 'lnorm')
> lnorm_dis
Fitting of the distribution ' lnorm ' by maximum likelihood
Parameters:
      estimate Std. Error
meanlog 0.1934339 0.001403909
sdlog 0.4430329 0.000992691
> pois_dis <- fitdist(buys_pair_data$n, 'pois')
> pois_dis
Fitting of the distribution ' pois ' by maximum likelihood
Parameters:
      estimate Std. Error
lambda 1.418878 0.003774639
> weibull_dis <- fitdist(buys_pair_data$n, 'weibull')
> weibull_dis
Fitting of the distribution ' weibull ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 1.381703 0.002538758
scale 1.581290 0.003864472
> |
```

```
> gofstat(list(exp_dis, gamma_dis, lnorm_dis, pois_dis, weibull_dis))
Goodness-of-fit statistics
              1-mle-exp  2-mle-gamma  3-mle-lnorm  4-mle-pois  5-mle-weibull
Kolmogorov-Smirnov statistic 5.293769e-01  0.4896604  4.938013e-01  0.6173609  0.428102
Cramer-von Mises statistic  5.463616e+03 4842.9805265  4.879257e+03 7366.9122883 4579.661605
Anderson-Darling statistic  2.492568e+04          Inf  2.281400e+04          Inf          Inf

Goodness-of-fit criteria
              1-mle-exp  2-mle-gamma  3-mle-lnorm  4-mle-pois  5-mle-weibull
Akaike's Information Criterion 228918.9  163531.8  117685.7  237643.8  201455.2
Bayesian Information Criterion 228928.3  163550.6  117704.5  237653.2  201474.0
> |
```

### Cullen and Frey graph





```
Call:
lm(formula = filered$new_Close ~ filered$tokenAmount + filered$n +
    filered$Open)
```

```
Residuals:
    Min       1Q   Median       3Q      Max
-1.83279 -0.22306 -0.02322  0.17976  2.44810
```

```
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)   1.470e-01  6.085e-02   2.416   0.0162 *
filered$tokenAmount  5.187e-25  6.673e-25   0.777   0.4375
filered$n        -6.161e-05  1.872e-04  -0.329   0.7422
filered$Open     9.511e-01  1.777e-02  53.518  <2e-16 ***
```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.433 on 348 degrees of freedom
(1 observation deleted due to missingness)
Multiple R-squared:  0.9002,    Adjusted R-squared:  0.8993
F-statistic: 1046 on 3 and 348 DF, p-value: < 2.2e-16
```

# Bitqy

```
1. library(magrittr)
2. library(dplyr)
3. library(ggplot2)
4. library(readr)
5. library(fitdistrplus)
6. library(DAAG)
7. library("ggplot2")
8. library(anytime)
9.
10. bitqy <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics/Project
    /Blockchain-Tokens-Data-Analytics/networkbitqyTX.txt', delim = " ", col_names = F)
11. names(bitqy) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
12. decimals <- 2
13. supply <- 1 * 10^10
14. bitqyFiltered <-
    filter(bitqy, tokenAmount < decimals * supply) #filter out all outliers
15.
16. #figure out how many users induced those unnormal transaction
17. bitqy_outliers <- filter(bitqy, tokenAmount >= decimals * supply)
18. user_outliers <- bitqy_outliers %>% group_by(toID) %>% summarise(nn = n()) %>% ungroup
19. number_users_outliers <- nrow(user_outliers)
20. number_users_outliers
21.
22. #get top X buyers data
23. buys <-
    bitqyFiltered %>% group_by(toID) %>% summarise(nn = n()) %>% ungroup #change the supply
    and decimals amount
24. buys_sorted_dec <- buys[order(-buys$nn),]
25. #top 30 active buyers and number of buys
26. top_30_buyers <- buys_sorted_dec %>% head(30)
27. top_30_buyers
28.
29. #####Question 1#####
    #####
30.
31. #####group by user pairs#####
32. buys_pairs <-
    bitqyFiltered %>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
33. for (row in 1:nrow(buys_pairs)) {
34.   a <- buys_pairs[row, "fromID"]
35.   b <- buys_pairs[row, "toID"]
36.   for (inner_row in row:nrow(buys_pairs)) {
37.     c <- buys_pairs[inner_row, "fromID"]
38.     d <- buys_pairs[inner_row, "toID"]
39.     if (a == d && b == c) {
40.       buys_pairs[inner_row, "fromID"] <- d
41.       buys_pairs[inner_row, "toID"] <- c
42.     }
43.   }
44. }
45. buys_pairs <-
    bitqyFiltered %>% group_by(fromID*toID+fromID+toID) %>% summarise(nn = n()) %>% ungroup
46. buys_pairs <-
    bitqyFiltered %>% group_by(fromID, toID) %>% summarise(nn = n()) %>% ungroup
```

```

47. buys_pair_sorted_asc<-buys_pairs[order(buys_pairs$n),]
48. buys_pair_less_30<-subset(buys_pair_sorted_asc,n<30)
49. buys_pair_data<-buys_pair_less_30
50.
51. #####find out estimates of paramaters of several distribution based on the buys_pairs d
    ata set#####
52. exp_dis <- fitdist(buys_pair_data$n, 'exp')
53. exp_dis
54. gamma_dis <- fitdist(buys_pair_data$n, 'gamma')
55. gamma_dis
56. lnorm_dis <- fitdist(buys_pair_data$n, 'lnorm')
57. lnorm_dis
58. pois_dis <- fitdist(buys_pair_data$n, 'pois')
59. pois_dis
60. weibull_dis <- fitdist(buys_pair_data$n, 'weibull')
61. weibull_dis
62.
63. gofstat(list(exp_dis, gamma_dis, lnorm_dis))
64. descdist(buys_sorted_asc$n,boot=1000)
65.
66. #lognorm
67. fit_lnorm <- fitdist(buys_pair_less_30$n,"lnorm")
68. summary(fit_lnorm)
69. plot(fit_lnorm)
70. cdfcomp(fit_lnorm)
71.
72. #exp
73. fit_exp <- fitdist(buys_pair_less_30$n,"exp")
74. summary(fit_exp)
75. plot(fit_exp)
76. cdfcomp(fit_exp)
77.
78. #gamma
79. fit_gamma <- fitdist(buys_pair_less_30$n,"gamma")
80. summary(fit_gamma)
81. plot(fit_gamma)
82. cdfcomp(fit_gamma)
83.
84. #weibull
85. fit_weibull <- fitdist(buys_pair_less_30$n,"weibull")
86. summary(fit_weibull)
87.
88. #normal
89. fit_normal <- fitdist(buys_pair_less_30$n,"norm")
90. summary(fit_normal)
91.
92. #pois
93. #normal
94. fit_pois <- fitdist(buys_pair_less_30$n,"pois")
95. summary(fit_pois)
96.
97. #unif
98. fit_unif <- fitdist(buys_pair_less_30$n,"unif")
99. summary(fit_unif)
100. plot(fit_unif)
101. cdfcomp(fit_unif)
102.
103. #####draw graph#####
104. all_density <- ggplot(data=buys_pair_less_30) +
105.   geom_histogram(bins=30,aes(x = buys_pair_less_30$n, ..density..)) +

```

```

106.      stat_function(fun = dlnorm, args = list(meanlog = 0.9122759, sdlog = 0.9075324
),
107.          colour = "red")+
108.      stat_function(fun = dgamma, args = list(shape = 1.2923088, rate=0.3361498),
109.          colour = "blue")+
110.      stat_function(fun=dexp, args=list(rate=0.2600901),colour="green")+
111.      stat_function(fun=dweibull, args=list(shape=1.083871, scale=3.982783),colour="
yellow")+
112.      stat_function(fun=dpois, args=list(lambda=3.844821),colour="orange")+
113.      xlab("No.Buys")
114.      all_density
115.
116.      #####Question 1#####
#####
117.
118.
119.      #####Question 2#####
#####
120.      bitqy_prices <- read_delim("C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Stat
istics/Project/Blockchain-Tokens-Data-
Analytics/bitqy", delim = "\t", col_names = T) #load token price data
121.      names(bitqy_prices) <- make.names(names(bitqy_prices))
122.      bitqy_prices <- bitqy_prices %>% mutate(date = as.Date(Date, format = '%m/%d/%Y'
))
123.
124.      bitqy <- read_delim('C:/Users/ygaoq/OneDrive/MyDocuments/2019 Spring/Statistics/
Project/Blockchain-Tokens-Data-
Analytics/networkbitqyTX.txt', delim = " ", col_names = F)
125.      names(bitqy) <- c('fromID', 'toID', 'unixTime', 'tokenAmount')
126.      decimals <- 2
127.      supply <- 1 * 10^10
128.      bitqy_filtered <-filter(bitqy,tokenAmount < decimals * supply)
129.      ## convert data type of unixTime
130.      bitqy_filtered <- bitqy_filtered %>%
131.          mutate(date = anydate(unixTime))
132.      names(bitqy_filtered) <- c('fromID', 'toID', 'unixTime', 'tokenAmount', 'date')
133.
134.      ## merge the prices and edge
135.      bitqy_merged<-
merge(x = bitqy_prices, y = bitqy_filtered, by = "date", all.x = TRUE)
136.
137.      #####Determin K#####
138.      top_30_buyers<-buys_sorted_dec%>%head(30)
139.
140.      top_K<-c(1:30)
141.      count <- 1
142.      for (val in top_K) {
143.          top_K_buyers<-buys_sorted_dec%>%head(val)
144.          filter_K_bitqy_merged<-filter(bitqy_merged,toID %in% top_K_buyers$toID)
145.          filter_K_bitqy_merged=transform(filter_K_bitqy_merged,average_price= (Open+Clo
se)/2)
146.          filter_K_bitqy_merged$num_Date <- as.numeric(as.POSIXct(filter_K_bitqy_merged
$date))
147.          filered<-
filter_K_bitqy_merged%>% group_by(num_Date) %>% summarise(nn = n(),Close=mean(Close),to
kenAmount=sum(tokenAmount),Open=mean(Open))
148.          shift <- function(x, n){
149.              c(x[-(seq(n))], rep(NA, n))
150.          }
151.          filered$new_Close<-shift(filered$Close,1)

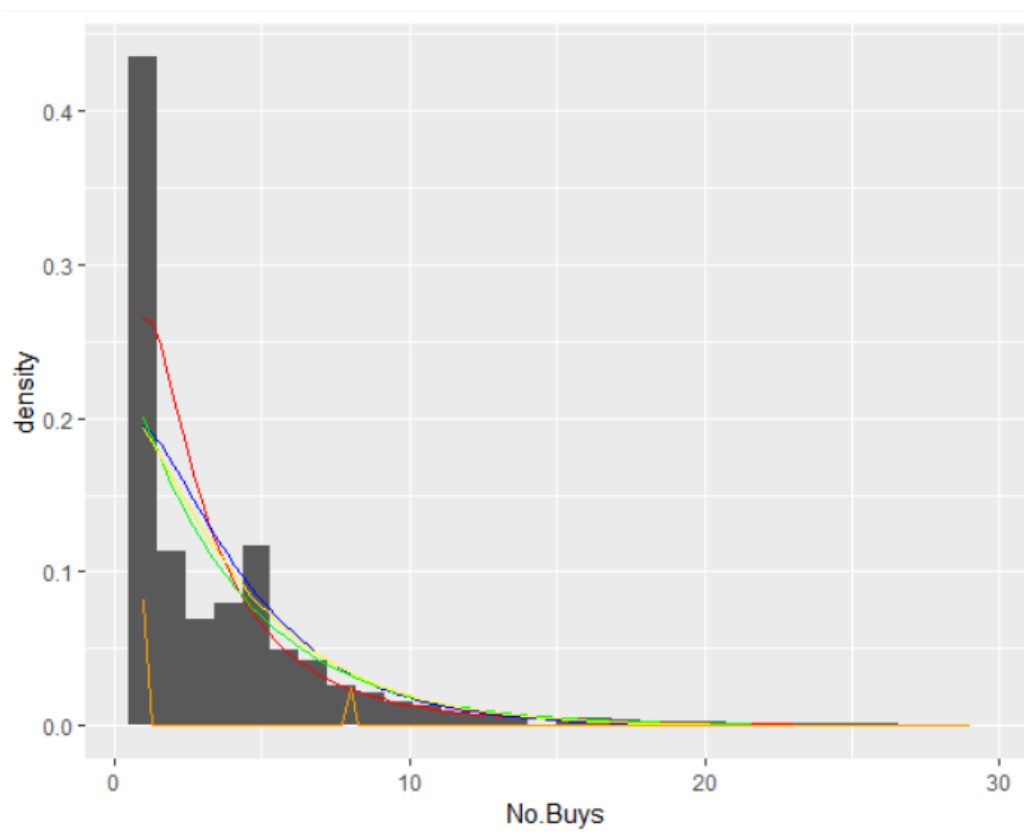
```



```

152.     num_rows<-nrow(filered)
153.     filered[-num_rows,]
154.     regression<-lm(filered$new_Close~filered$tokenAmount+filered$n+filered$Open)
155.
156.     setwd("C:/Users/ygaoq/Desktop/bitqy")
157.     yourfilename=paste("W",val,".txt",sep="")
158.     capture.output(summary(regression),append = TRUE,file = "C:/Users/ygaoq/Desktop/
p/bitqy/Final_Result.txt")
159.
160.
161.     summary(regression)
162.     par(mfcol=c(2,2))
163.     setwd("C:/Users/ygaoq/Desktop/bitqy")
164.     yourfilename=paste("A",val,".png",sep="")
165.     png(file=yourfilename)
166.     opar <- par(mfrow=c(2,2))
167.     plot(regression)
168.     dev.off()
169. }

```

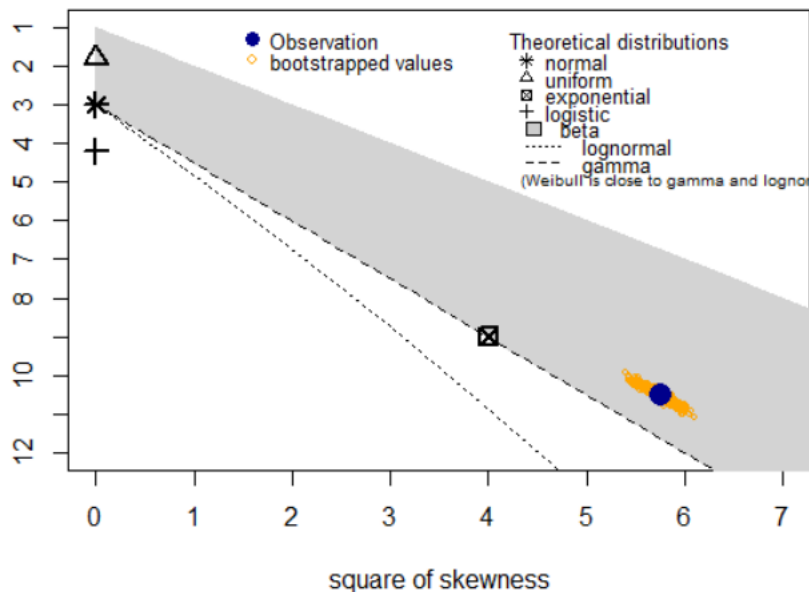


```

> exp_dis <- fitdist(buys_pair_data$n, 'exp')
> exp_dis
Fitting of the distribution ' exp ' by maximum likelihood
Parameters:
      estimate Std. Error
rate 0.2600901 0.001208959
> gamma_dis <- fitdist(buys_pair_data$n, 'gamma')
> gamma_dis
Fitting of the distribution ' gamma ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 1.2923088 0.007644403
rate  0.3361498 0.002417218
> lnorm_dis <- fitdist(buys_pair_data$n, 'lnorm')
> lnorm_dis
Fitting of the distribution ' lnorm ' by maximum likelihood
Parameters:
      estimate Std. Error
meanlog 0.9122759 0.004218481
sdlog   0.9075324 0.002982900
> pois_dis <- fitdist(buys_pair_data$n, 'pois')
> pois_dis
Fitting of the distribution ' pois ' by maximum likelihood
Parameters:
      estimate Std. Error
lambda 3.844821 0.009114482
> weibull_dis <- fitdist(buys_pair_data$n, 'weibull')
> weibull_dis
Fitting of the distribution ' weibull ' by maximum likelihood
Parameters:
      estimate Std. Error
shape 1.083871 0.003679308
scale 3.982783 0.018143064

```

### Cullen and Frey graph



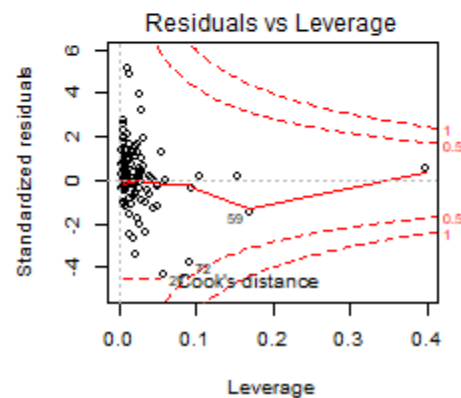
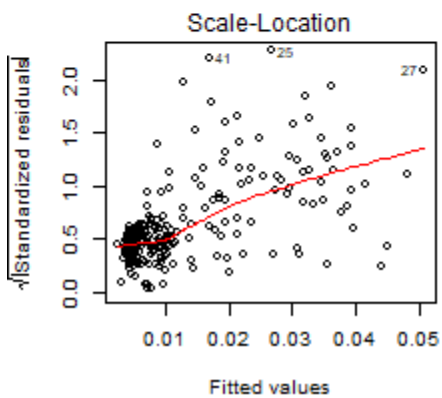
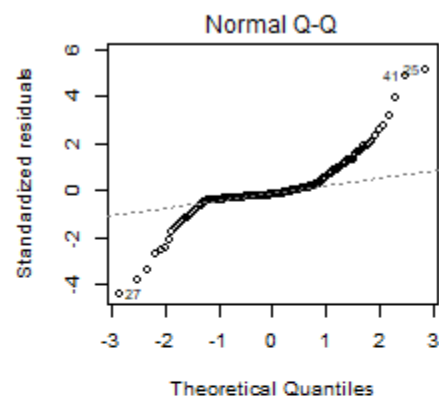
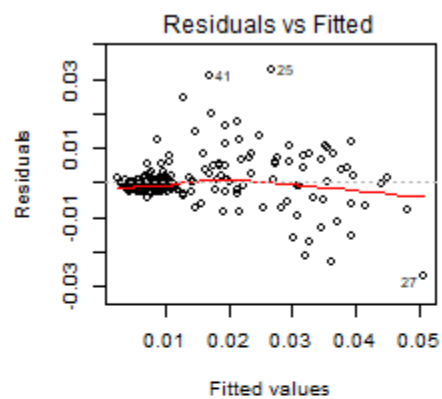
```
> gofstat(list(exp_dis, gamma_dis, lnorm_dis, pois_dis, weibull_dis))
```

Goodness-of-fit statistics

	1-mle-exp	2-mle-gamma	3-mle-lnorm	4-mle-pois	5-mle-weibull
Kolmogorov-Smirnov statistic	0.2290179	0.2443449	0.2618188	0.3155808	0.2188462
Cramer-von Mises statistic	332.9182743	397.9670717	407.9692848	876.8235031	354.4395405
Anderson-Darling statistic	2164.0769789	2510.0855460	2640.0179951	Inf	2259.2630269

Goodness-of-fit criteria

	1-mle-exp	2-mle-gamma	3-mle-lnorm	4-mle-pois	5-mle-weibull
Akaike's Information Criterion	217224.4	215474.7	206809.4	282512.5	216686.4
Bayesian Information Criterion	217233.2	215492.2	206826.9	282521.2	216703.9



```
Call:
lm(formula = filered$new_Close ~ filered$tokenAmount + filered$n +
    filered$Open)
```

```
Residuals:
```

```
      Min       1Q   Median       3Q      Max
-0.027199 -0.001834 -0.001114  0.000887  0.032898
```

```
Coefficients:
```

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    2.095e-03  6.423e-04   3.262  0.00127 **
filered$tokenAmount -4.231e-14  4.916e-14  -0.861  0.39030
filered$n       2.345e-06  9.470e-06   0.248  0.80461
filered$Open     8.171e-01  3.222e-02  25.356 < 2e-16 ***
```

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.0064 on 240 degrees of freedom
(1 observation deleted due to missingness)
```

```
Multiple R-squared:  0.7473,    Adjusted R-squared:  0.7442
```

```
F-statistic: 236.6 on 3 and 240 DF,  p-value: < 2.2e-16
```