Financial Econometrics Assignment 3

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Due data: November 11, 2024

Question 1. Please see JD.csv data which includes some monthly stock price information like adjusted closing price etc. for JD.com.

a. Write R code to calculate monthly returns for JD stock using adjusted closing prices. > # Read data JD_data <- read.csv("JD.csv")</pre> # 计算每日对数收益率 JD_data\$Return <- c(NA, diff(log(JD_data\$Adj.Close)))</pre> # 移除包含 NA 的行 JD_data <- na.omit(JD_data)</pre> # 输出计算后的收益率 print(JD_data\$Return) 0.0038923462 0.00000000000 -0.0436730013 -0.0050257136 -0.0208492401 -0.0410057449 0.0073745025 -0.0277439320 0.0139681787 0.0005136364 -0.0041162908 -0.0106259906 0.0235907863 -0.0015216082 0.0041601723 -0.0207726915 -0.0167236454 0.0149080567 -0.0065112875 -0.0060280355 -0.0013152441 -0.0121824333 -0.0174711607 -0.0378551352 0.0445871172 0.0197134232 0.0042153218 -0.0098633742 -0.0076128430 -0.0294262233 -0.0118052164 0.0116441016 0.0002756511 -0.0033122581 -0.0074926395 0.0165749094 -0.0064057870 -0.0203224886 0.0077894770 -0.0137933221 0.0005554013 -0.0121251913 -0.0141236308 -0.0348164268 -0.0456065142 -0.0072904940 0.0251315701 -0.0298155152 0.0003185539 -0.0157337201 0.0396492715 -0.0065512322 0.0096308611 -0.0301850076 -0.0616035034 -0.1124456611 0.0354846199 -0.0110701866 -0.0157071286 0.0208884503 $\begin{array}{c} 0.0046940329 \\ 0.0444348777 \end{array}$ 0.0197374466 -0.0236934352 0.0007770008 0.0003714020 -0.0164736035 [69] -0.0776861057 0.0012232825 0.0265395542 0.0090855452 73 0.0256225197 -0.0232658808 -0.0234184149 0.0095923599 -0.0331611063 -0.0292052040 0.0113660664 -0.0046149862 0.0404812619 -0.0148907240 0.0460396212 -0.0127600826 0.0333436829 0.0200949973 0.0393343615 0.0138111820 -0.0132565975 -0.0653711248 0.0227722403 0.0361049603 93 -0.0269316097 0.0008861764 0.0180347092 0.0496834648 0.0788646593 -0.0565808273 0.0041494251 0.0230376681 0.0197195398 -0.0719052929 101 -0.0317486522 0.0141845889 0.0163902233 0.0613480652 105 0.0171174904 -0.0486869484 109 -0.0879188168 -0.0798453957 0.0426878771 -0.0540399082 0.0599222292 0.0184025052 0.0255816419 113 -0.0193630616 -0.0257722035 0.0193149414 117 0.0123224785 0.0360816198 -0.0103094163 0.0142317393 0.0202709644 0.0458023524 125 0.0367754462 -0.0240652885 -0.0376882840 -0.0146263043 -0.0422290834 129 -0.0129742339 0.0571026056 -0.0651712323 133 137 0.0661195491 0.0280392661 0.0004606312 -0.0365894014 0.0161140928 -0.0442167173 0.0901595694 0.0217641280 141 0.0083133708 0.0317350803 -0.0123170380 -0.0224728996 0.0094829790 0.0035890572 0.0374040931 -0.0085451837 0.0493492650 -0.0703449705 0.0132876643 149 0.0058997676 0.0669394404 0.0038014831 -0.0084674089 0.0465199751

-0.0088925810

-0.0139003873

0.0004169273

0.0120159305

0.0084866545

-0.0122700528 -0.0460460634

```
-0.0028402940 -0.0114427482
[165]
       0.0163401917
                                                   -0.0153244253
       0.0279853666
[169]
                      0.0016221010
                                   -0.0167521793
                                                    0.0669306284
[173]
                      0.0151575086
                                   -0.0235932181
                                                    0.0648513356
       0.0092060486
177
       0.0361485502
                      0.0055536069
                                     0.0157970664
                                                   -0.0109627338
[181]
      -0.0418559770
                     -0.0287868671
                                     0.0208532159
                                                    0.0175852363
185
       0.0007111665
                     -0.0085684212
                                     0.0014332142
                                                    0.0074906362
189
       0.0060230477
                     -0.0081574846
                                     0.0239279537
                                                    0.0267810227
193
                                                    0.0085368454
       0.0138324251
                      0.0146855791
                                     0.0120753928
197
       0.0248501706
                      0.0361544018
                                    -0.0315216669
                                                    0.0006600000
201
      -0.0019814736
                      0.0322031405
                                    -0.0145093306
                                                    0.0035661895
205
       0.0239081307
                      0.0009939043
                                                    0.0302197992
                                     0.0121771784
209
       0.0083935359
                     -0.0020080328
                                     0.0053458196
                                                   -0.0120685340
213
       0.0043749354
                     -0.0040377241
                                    -0.0146037576
                                                   -0.0034270425
217
221
       0.0249186895
                      0.0134698968
                                    -0.0023151655
                                                   -0.0069779647
       0.0019986349
                     -0.0459654701
                                    -0.0562570678
                                                   -0.0088856567
225
                                     0.0003549246
       0.0227951418
                      0.0237080070
                                                    0.0515239278
229
       0.0256211315
                                    -0.0384855940
                                                    0.0329489837
                      0.0006567488
233
       0.0240387119
                     -0.0286332066
                                    -0.0557216394
                                                    0.0143344964
                                     0.0129871955
237
       0.0011404297
                     -0.0107075973
                                                   -0.0226453738
241
                      0.0395789490
                                    -0.0230241535
                                                    0.0019072674
       0.000000000
245]
       0.0211168861
                      0.0170185761
                                     0.0267840839
                                                   -0.0275180289
249]
       0.0109530483
                     -0.0179525613
```

b. Fit the generalized t distribution (see the functions that I defined in the Appendix) to the daily returns of JD. You are not allowed to use external packages. (You can of course use **optim** function of R.) You need to use maximum likelihood method to estimate the three parameters (location, scale, and degrees of freedom) of the t distribution. For the same purpose we can use **fitdistrplus** package. Please compare the results from these two different ways of distribution fitting.



Observing the graph, the data shows a clear downward trend followed by a relatively stable fluctuation phase, which may suggest the presence of some outliers or data skewness. These characteristics could make the t-distribution a suitable choice for fitting the data, particularly when the sample size is small. So I anticipate that the fit of the t-distribution should be relatively good. Next, actual fitting will be conducted to further determine the degree of fit.

```
> # Define the PDF and CDF
> dt_G <- function(x, params) {</pre>
      location <- params[1]</pre>
      scale <- params[2]</pre>
      shape <- params[3]
      return(dt((x - \overline{location}) / scale, df = shape) / scale)
   Create a function for the probability density function (PDF) of the Gene
ralized t-distribution
 pt_G <- function(q, location, scale, shape)</pre>
      return(pt((q - location) / scale, df = shape))
+
  # Create a function for the cumulative distribution function (CDF) of the
Generalized t-distribution
  # Fit a Generalized t Distribution Using MLE
  neg_log_likelihood <- function(params, data) {</pre>
      log_density <- sapply(data, function(x) log(dt_G(x, params)))</pre>
      return(-sum(log_density))
```

```
> # Define the negative log-likelihood function for Maximum Likelihood Estim
ation (MLE)
> returns_data <- JD_data$Return</pre>
> initial_params <- c(location = mean(returns_data), scale = sd(returns_dat</pre>
a), shape = 5)
> mle_result <- optim(par = initial_params, fn = neg_log_likelihood, data =</pre>
returns_data,
                          method = "L-BFGS-B", lower = c(-Inf, 0.001, 2), upper
= c(Inf, Inf, Inf))
> mle_params <- mle_result$par</pre>
> # Perform MLE to estimate the parameters of the Generalized t-distribution
> # Utilize the fitdistrplus package for comparison
> library(fitdistrplus)
> # Load the fitdistrplus package for additional fitting methods
> # Define the PDF and CDF required for fitdistrplus
> ddt_G_fitdistr <- function(x, location, scale, shape) {
+ return(dt((x - location) / scale, df = shape) / scale)</pre>
+ }
 pdt_G_fitdistr <- function(q, location, scale, shape) {
    return(pt((q - location) / scale, df = shape))</pre>
+
+
  # Create functions for fitdistrplus to use the Generalized t-distribution
> fit_result <- fitdist(JD_data$Return, "dt_G_fitdistr", start = list(locati
on = mean(returns_data), scale = sd(returns_data), shape = 5))
> # Fit the distribution using fitdistrplus with initial parameter estimates
> # Compare the results of MLE and fitdistrplus
> cat("MLE Results:\n")
MLE Results:
> print(mle_params)
     location
                          scale
                                          shape
-0.001561329 0.025491936 6.711448675
> # Print the parameters estimated by MLE
  cat("\nfitdistrplus Results:\n")
fitdistrplus Results:
  print(fit_result$estimate)
     location
                                          shape
                         scale
-0.001564296 0.025456360 6.708586251
> # Print the parameters estimated by fitdistrplus
```

Conclusion:

The closeness of the parameter estimates obtained from MLE and the 'fitdistrplus' package indicates the robustness of the parameter estimation and the rationality of model selection, providing a solid foundation for further statistical analysis.

1. Consistency in Parameter Estimation:

The parameter estimates obtained through Maximum Likelihood Estimation (MLE) and the `fitdistrplus` package are very close. This indicates that both methods are consistent when estimating the parameters of the Generalized t-distribution.

2. Reliability of the Methods:

The similarity in results from two distinct methods bolsters confidence in the reliability of the parameter estimates, indicating that both direct optimization and specialized statistical software yield robust estimates.

Appendix.

```
# for maximum likelihood method
# Density of t distribution with location, scale, and degrees of freedom
dt_G<-function(x,params){#density</pre>
 loc<-params[1]</pre>
 sc<-params[2]</pre>
 nu<-params[3]
 dt((x-loc)/sc,nu)/sc
}
# for fitdistrplus package
dt_G<-
 function(x,loc,sc,nu){ dt((x
 -loc)/sc,nu)/sc
# CDF of t distribution with location, scale, and degrees of freedom
 function(q,loc,sc,nu){ pt((q
 -loc)/sc,nu)
}
```