A Simple Application of Pairs Trading Strategy

Qingmin Shi

May 10, 2016

1. Introduction

The basic idea of pairs trading is to find two similar financial assets which trade at some spread. If there exists equilibrium between two assets and an anomaly is observed in the relationship, one can seek to profit from the comparative mispricing by selling the relative overvalued asset and simultaneously buying the undervalued asset.

The main theoretical foundation of my project is cointegration. Specifically, suppose we have asset X and Y, we need to find weights a and b, such that the linear combination of X and Y is a stationary noise process. That is to say, we need to make $aX + bY = \epsilon$, where ϵ is a stationary noise process.

This paper explored several different approaches for cointegration. It is natural to think about linear regression, i.e., $Y = aX + \epsilon$. However, the diagnostic plot indicate that the residuals are not stationary. We also tried AR models and got better performance. Finally, we decide on the Johansen procedure, which uses VAR (vector autoregressive) models instead of regression. Using this method, we can obtain a stationary noise process.

After getting the stationary noise process, we can begin to trade the spread. Firstly, we can normalize the process aX + bY. Then, we observe the process. When its value is below 1.5 standard error, we build a long position. Once its value is within 0.3 standard error, we cover the long position we built. When its value is above 1.5 standard error, we build a short position. Once its value is within 0.3 standard error, we cover the short position we built.

The above is the main framework of the project. We can see all the details in the following contents.

2. Data Description

Our choice of the stock pair is AT&T and Verizon, starting from May 1st in 2008, to April 1st in 2011. We use the daily adjusted closing price, so we have totally 737 prices for each stock. Denote log price of AT&T as X, log price of Verizon as Y, we can see that X is a 737 \times 1 vector, Y is also a 737 \times 1 vector. We can see from the plots that these two stocks do have similar trend of prices. The correlation coefficient is 0.82.

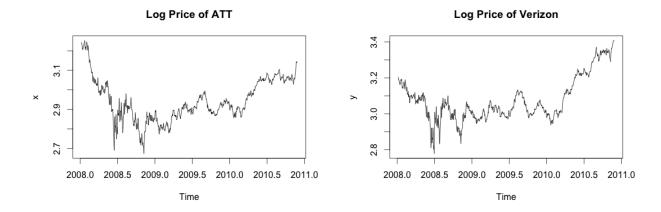


Figure 1: Price trend of AT&T and Verizon

3. Cointegration

3.1 The linear regression approach and AR(3) approach

First, we consider the simple linear regression model. The residual plot shows the residual series has certain patterns and it is not stationary. To further confirm it, we plot the ACF, which shows that it decays very slowly. So we fit the AR(3) model to the residual series. The coefficients are not significant. Otherwise, although the ADF test shows that the residuals of the AR(3) model is stationary, the ACF and PACF plot are not clean. After we plot the ACF and PACF of the squared of residual, it is obvious to tell that there is ARCH effect. Thus, the simple linear regression model with the AR(3) model is not satisfying. So we decide to use the Johansen procedure to fit the series.

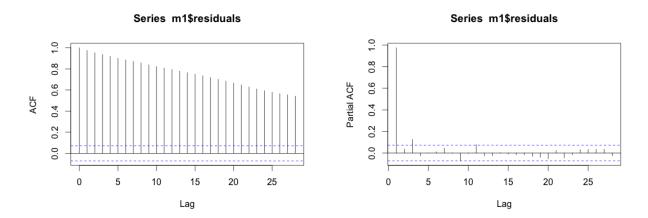


Figure 2: ACF and PACF of linear residuals indicating it is not a good model

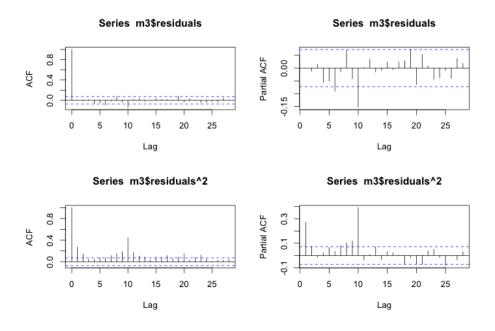


Figure 3: ACF and PACF of AR(3) squared residuals indicating it is not a good model

3.2 The Johansen procedure

```
Eigenvectors, normalised to first column:
# Johansen-Procedure #
********
                                                     (These are the cointegration relations)
Test type: trace statistic , with linear trend
                                                               ATTad.13 Verad.13
                                                     ATTad.13 1.0000000 1.00000
                                                     Verad.13 -0.3394999 12.10922
Eigenvalues (lambda):
[1] 0.4902120 0.1794881
                                                     Weights W:
                                                     (This is the loading matrix)
Values of teststatistic and critical values of test:
                                                              ATTad.13
                                                                         Verad.13
         test 10pct 5pct 1pct
                                                     ATTad.d -0.9852927 -0.01696224
         5.54 6.50 8.18 11.65
  <= 1 |
      | 24.40 15.66 17.95 23.52
                                                     Verad.d 0.4193060 -0.02548269
```

Figure 4: Results of the Johansen procedure

We can see from results of Johansen procedure that the estimated cointegration coefficient γ is -0.34.

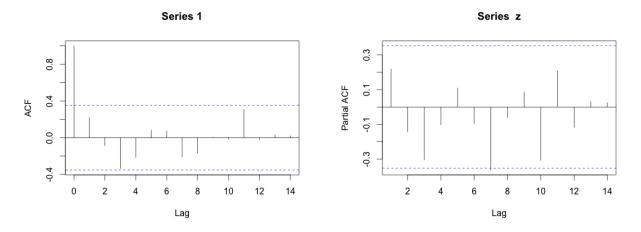


Figure 5: ACF and PACF of the residuals generated by Johansen procedure

We can see that this model is satisfactory. We can treat the residuals as a stationary noise process.

4. Trading Strategy and Application

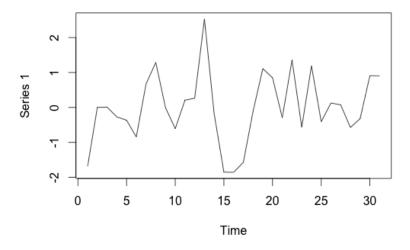


Figure 6: A glance of the normalized residuals

4.1 The fixed gamma approach

Using fixed gamma approach, our trading strategy can be expressed as follows. Firstly, we run Johansen procedure from day 1 to day 30 and obtain gamma. Then, in the next 30 trading days, i.e., day 31 to day 60, we use this fixed gamma to build the stationary noise process.

4.2 The floating gamma approach

Using floating gamma approach, we use gamma computed from the most recent 30 days and update our gamma every 10 days or 5 days. For example, if we update gamma every 10 days, the gamma used in days 31-40 is computed from days 1-30, the gamma used in days 41-50 is computed from days 11 to 40, and the gamma used in days 51-60 is computed from days 21-50.

4.3 Trading Strategy and Performance Comparation

Our trading process is as follows: Go over the residual process day by day. When the value of the residual is below 1.5 standard error, we build a long position. Once its value is within 0.3 standard error, we cover the long position we built. When its value is above 1.5 standard error, we build a short position. Once its value is within 0.3 standard error, we cover the short position we built.

Backtesting performances of our trading strategies are as follows:

Apply the Strategy to the next 30 Days Using Fixed Gamma

	31-60 Days
Gamma	-0.3394999
Gain	0.06663566

Apply the Strategy to the next 30 Days
Using Floating Gamma
(Changing Every 10 Days)

	31-40 days	41-50 days	51-60 days		
Gamma	-0.3394999	-1.420094	-1.558607		
Gain	0.3372988	-0.3185177	0.401776		
Total Gain	0.4205571				

Apply the Strategy to the next 30 Days
Using Floating Gamma
(Changing Every 5 Days)

	31-35 days	36-40 days	41-45 days	46-50 days	51-55 days	56-60 days
Gamma	0.3395	0.6975	1.4201	1.3287	1.5586	1.8208
Gain	0.3728	0.0367	0.0445	0.0579	0.4987	0.5359
Total Gain	1.268230477					

We can see clearly that the floating gamma approach has a better performance than fixed gamma approach. We can conclude that updating gamma every 5 trading days, and each gamma generating from the previous 30 days of price data is a good choice.

Code

```
1 library(tseries)
 2 library(forecast)
 3 library(urca)
 4 ###Data Selection
 5 ATT=ATT <- read.csv("~/Desktop/ts project/ATT.csv", sep=";")
 6 Ver=Verizon <- read.csv("~/Desktop/ts project/Verizon.csv", sep="")
 7 par(mfrow=c(1,1))
 8 x=ts(log(ATT[85:821,]),start=c(2008,5,1),freq=255)
9 plot(x,main='Log Price of ATT')
10 y=ts(log(Ver[85:821,]),start=c(2008,5,1),freq=255)
plot(y,main='Log Price of Verizon')
12 cor(log(ATT[85:821,]),log(Ver[85:821,]))
14 ###Simple Linear Regression with AR(3) Model
15 m1=lm(log(ATT[85:821,]) ~ log(Ver[85:821,]))
16 summary (m1)
17 wt=m1$residuals
18 plot(ts(wt), start=c(2008,5,1), freq=255)
19 acf(wt)
20 pacf(wt)
m3=arima(wt,order=c(3,0,0),include.mean = F)
22 m3
23 summary (m3)
24 adf.test(m3\residuals)
25 par(mfrow=c(2,2))
26 acf(m3$residuals)
27 acf(m3$residuals^2)
28 pacf(m3$residuals)
29 pacf(m3$residuals^2)
30 par(mfrow=c(1,1))
33 ###Cointegration
34 ATTad=log(ATT[684:714,])
35 Verad=log(Ver[684:714,])
36 xx=ts(ATTad)
37 plot(xx,main='Selected Log Price of ATT')
38 yy=ts(Verad)
39 plot(yy, main='Selected Log Price of Verizon')
40 xxy=cbind(ATTad, Verad)
41 mx=ar(xxy)
42 mx$order
43 cot=ca.jo(xxy, ecdet='none', type='trace', K=3, spec='longrun')
44 summary(cot)
45 cot@V[2,1] #gamma
47 ###Strategy Application
48 zz=cbind(xx,yy)
49 weight=c(1,cot@V[2,1])
z=ts(zz%*%weight)
51 plot(z)
52 acf(z)
```

```
53 pacf(z)
54 sd(z)
55 mean(z)
abs((z-mean(z))/sd(z))
57 plot((z-mean(z))/sd(z))
59 ####Fixed Gamma
spread = (z-mean(z))/sd(z)
61 longpos = rep(0,31)
62 i = 1
63 while(i != 31)
    if (spread[i] <-1.5)</pre>
66
   {
     longpos[i] = 1
67
      for(j in (i+1):31)
68
       if((spread[j]>-0.3)|| j == 31)
70
71
          break
72
     }
     longpos[j] = -1
73
     i = j + 1
74
     }
     else
76
     {
      i = i + 1
78
     }
80 }
shortpos = rep(0,31)
82 i = 1
83 while(i != 31)
    if(spread[i] > 1.5)
85
      shortpos[i] = 1
87
     for(j in (i+1):31)
88
89
      if((spread[j]<0.3)|| j == 31)</pre>
          break
91
      }
     shortpos[j] = -1
93
      i = j + 1
94
   }
95
     else
97
     i = i + 1
99
     }
100 }
102 gain1 = 0
103 buy=c(0,0)
104 \text{ sell=c}(0,0)
105 for(i in 1:31)
106 {
```

```
if (longpos[i]==1)
107
108
      buy = c(ATT[714+i,], Ver[714+i,])
109
     if(longpos[i] == -1)
112
      sell = c(ATT[714+i,], Ver[714+i,])
113
       gain1=gain1+c(1,-Ver[714,]^(cot@V[2,1]-1))%*%(sell-buy)
115
116 }
117
118 \text{ gain2} = 0
119 buy=c(0,0)
120 sell=c(0,0)
121 for(i in 1:31)
122 {
     if (shortpos[i]==1)
     {
124
      sell = c(ATT[714+i,], Ver[714+i,])
125
126
127
     if(shortpos[i] == -1)
128
      buy = c(ATT[714+i,], Ver[714+i,])
      gain2=gain2+c(1,-Ver[714,]^(cot@V[2,1]-1))%*%(sell-buy)
130
132 }
134 gain = gain1 + gain2
135 gain
####Floating Gamma(Changing Every 5 Days)
^{138} ###For simplification, I only post the calculation of the first two gains.
139 ####The method of using floating gamma changing every 10 days to calculate the gain is the same.
140 ####First Gamma
141 ATTad=log(ATT[684:714,])
142 Verad=log(Ver[684:714,])
143 xx=ts(ATTad)
144 plot(xx,main='Selected Log Price of ATT')
145 yy=ts(Verad)
146 plot(yy, main='Selected Log Price of Verizon')
147 xxy=cbind(ATTad, Verad)
148 mx=ar(xxy)
149 mx$order
cot=ca.jo(xxy, ecdet='none', type='trace', K=3, spec='longrun')
151 summary(cot)
152 cot@V[2,1] #gamma
153 zz=cbind(xx,yy)
154 weight=c(1,cot@V[2,1])
155 z=ts(zz%*%weight)
156 plot(z)
157 acf(z)
158 pacf(z)
159 sd(z)
160 mean(z)
```

```
abs((z-mean(z))/sd(z))
162 plot((z-mean(z))/sd(z))
spread = (z-mean(z))/sd(z)
165 longpos = rep(0,6)
166 i = 1
167 while(i != 6)
     if (spread[i] <-1.5)</pre>
169
      longpos[i] = 1
171
172
      for(j in (i+1):6)
173
        if((spread[j]>-0.3)|| j == 6)
174
           break
175
      }
176
      longpos[j] = -1
177
      i = j + 1
178
179
     }
180
     else
      i = i + 1
182
     }
183
184 }
shortpos = rep(0,6)
186 i = 1
187 while(i != 6)
188 {
189
     if(spread[i] > 1.5)
190
    {
191
      shortpos[i] = 1
      for(j in (i+1):6)
192
193
       if((spread[j]<0.3)|| j == 6)</pre>
194
           break
195
      }
196
197
      shortpos[j] = -1
      i = j + 1
198
     }
199
     else
     {
201
      i = i + 1
202
     }
203
204 }
205
206 gain1 = 0
207 buy=c(0,0)
208 \text{ sell=c}(0,0)
209 for(i in 1:6)
210 {
211    if (longpos[i] == 1)
212 {
     buy = c(ATT[714+i,], Ver[714+i,])
213
214 }
```

```
if (longpos[i] == -1)
215
216
     {
      sell = c(ATT[714+i,], Ver[714+i,])
217
      gain1=gain1+c(1,-Ver[714,]^(cot@V[2,1]-1))%*%(sell-buy)
219
220 }
221
222 gain2 = 0
223 buy=c(0,0)
224 sell=c(0,0)
225 for(i in 1:6)
226 {
    if(shortpos[i]==1)
227
228
      sell = c(ATT[714+i,], Ver[714+i,])
229
230
231
     if(shortpos[i] == -1)
    {
232
      buy = c(ATT[714+i,], Ver[714+i,])
233
      gain2=gain2+c(1,-Ver[714,]^(cot@V[2,1]-1))%*%(sell-buy)
234
235
236 }
238 gain = gain1 + gain2
239 gain
241 ####Second Gamma
242 ATTad=log(ATT[689:719,])
243 Verad=log(Ver[689:719,])
244 xx=ts(ATTad)
245 plot(xx,main='Selected Log Price of ATT')
246 yy=ts(Verad)
247 plot(yy,main='Selected Log Price of Verizon')
248 xxy=cbind(ATTad, Verad)
249 mx=ar(xxy)
250 mx$order
cot=ca.jo(xxy, ecdet='none', type='trace', K=3, spec='longrun')
252 summary(cot)
253 cot@V[2,1] #gamma
spread = (z-mean(z))/sd(z)
256 longpos = rep(0,6)
257 i = 1
258 while(i != 6)
259 {
     if (spread[i] <-1.5)</pre>
     {
261
262
      longpos[i] = 1
      for(j in (i+1):6)
264
265
        if((spread[j]>-0.3)|| j == 6)
266
           break
       }
267
     longpos[j] = -1
268
```

```
i = j + 1
269
270
     }
271
     else
      i = i + 1
273
274
275 }
shortpos = rep(0,6)
277 i = 1
278 while(i != 6)
279 {
280
     if(spread[i] > 1.5)
281
      shortpos[i] = 1
282
       for(j in (i+1):6)
283
284
       if((spread[j]<0.3)|| j == 6)
         break
286
287
      }
      shortpos[j] = -1
288
      i = j + 1
289
     }
290
     else
292
     {
      i = i + 1
294
     }
295 }
296
297 \text{ gain1} = 0
298 buy=c(0,0)
299 sell=c(0,0)
300 for(i in 1:6)
301 {
     if (longpos[i]==1)
302
303
      buy = c(ATT[719+i,], Ver[719+i,])
304
305
     if(longpos[i] == -1)
     {
307
      sell = c(ATT[719+i,], Ver[719+i,])
      gain1=gain1+c(1,-Ver[719,]^(cot@V[2,1]-1))%*%(sel1-buy)
309
311 }
313 \text{ gain2} = 0
314 buy=c(0,0)
315 sell=c(0,0)
316 for(i in 1:6)
317 {
     if (shortpos[i]==1)
318
320
      sell = c(ATT[719+i,], Ver[719+i,])
     }
321
322 if(shortpos[i] == -1)
```