

MGMT 237E:
Empirical Methods in Finance
Homework 5

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1.a. The null hypothesis is: $\beta_1^{n+1} = 1$ for all n.

1.b. OLS and HAC test results shows that the coefficient is significantly different from 1. The HAC result is more meaningful, as the time series data demonstrate heteroscedasticity.

Table 1: Classic OLS assumes homoscedasticity.

β	OLS SE	T-stat
0.2736	0.086388	2.6178
0.53016	0.083435	5.2238
0.71412	0.076759	9.7909
0.78976	0.074268	11.773

Table 2: Variance covariance estimation is heteroscedastic robust.

β	OLS White SE	T-stat
0.2736	0.090767	2.6178
0.53016	0.10149	5.2238
0.71412	0.072937	9.7909
0.78976	0.067082	11.773

Table 3: HAC regression result.

β	HAC SE	T-stat
0.2736	0.20735	1.1459
0.53016	0.26637	1.9903
0.71412	0.18802	3.798
0.78976	0.17014	4.6419

2.a. The null hypothesis is: $\gamma_1^{n+1} = 0$ for all n.

2.b. OLS and HAC test results shows that the coefficient is significantly different from 0.

Table 4: Classic OLS assumes homoscedasticity.

γ	OLS SE	T-stat
0.7624	0.086388	8.3995
1.0118	0.10801	8.5929
1.2821	0.11977	9.8876
1.0561	0.13546	6.9011

Table 5: Variance covariance estimation is heteroscedastic robust.

γ	OLS White SE	T-stat
0.7624	0.090767	8.3995
1.0118	0.11775	8.5929
1.2821	0.12966	9.8876
1.0561	0.15304	6.9011

Table 6: HAC regression result.

γ	HAC SE	T-stat
0.7624	0.20735	3.6769
1.0118	0.26474	3.8218
1.2821	0.30213	4.2435
1.0561	0.34352	3.0745

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1 %% MGMT237E HW5
2 % Zhaofang Shi
3 clear; clc; close all;
4 % this part is subject to changes
5 addpath('C:\Users\SallyShi\Desktop\MGMT237E- Empirical Methods in
  Finance\MFEToolbox\timeseries ');
6 addpath('C:\Users\SallyShi\Desktop\MGMT237E- Empirical Methods in
  Finance\MFEToolbox\crosssection ');
7 addpath('C:\Users\SallyShi\Desktop\MGMT237E- Empirical Methods in
  Finance\MFEToolbox\utility ');
8 addpath('C:\Users\SallyShi\Desktop\MGMT237E- Empirical Methods in
  Finance\HW5')
9 %% question 1
10 prices = xlsread('Fama_bond_prices.xlsx');
11 % get annually price based on 1 dollar face value
12 yrprices = prices(1:length(prices(:,1)),2:6)/100;
13 % generate matrix for log price based on 1 dollar face value
14 logprices = log(yrprices);
15 % generate matrix for yields
16 yields = zeros(length(logprices(:,1)),5);
17 for n=1:5
18     yields(:,n)= -logprices(:,n)/n;
19 end
20 % generate matrix for forward rate
21 fwd = zeros(length(logprices(:,1)),4);
22 for n=1:4
23     fwd(:,n)= logprices(:,n)-logprices(:,n+1);
24 end
25 % generate matrix for dependent variables for regression 1
26 yieldchanges=NaN(length(logprices(:,1)),4);
27 for n=1:4
28     yieldchanges(1:end-n*12,n)= yields(1+n*12:end,1)-yields(1:end-n
      *12,1);
29 end
30 % generate matrix for independent variables
31 fwdspread=NaN(length(logprices(:,1)),4);
32 for n=1:4
33     fwdspread(:,n)=fwd(:,n)-yields(:,1);
34 end
35
36 % figure(1)
37 % plot(fwdspread)
38 %
39 % figure(2)
40 % plot(yieldchanges)
41
42 % compute OLS standard error
43 B = NaN(4,2);
44 TSTAT = NaN(4,2);
45 S2 = NaN(4,1);
46 SE = NaN(4,2);
47 SEWHITE = NaN(4,2);
48 VCV= NaN(2,2);
49 VCVWHITE = NaN(2,2);
50 for i=1:4
51     [B(i,:),TSTAT(i,:),S2(i,1),VCV,VCVWHITE] = ols(yieldchanges(1
      :end-i*12,i),fwdspread(1:end-i*12,i));

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52     SE(i,:) = sqrt(diag(VCV));
53     SEWHITE(i,:) = sqrt(diag(VCVWHITE));
54 end
55
56 OLS = table(B(:,2), SE(:,2), TSTAT(:,2));
57 OLS.Properties.VariableNames = {'Beta' 'OLS_SE' 'T_stat' };
58 disp(OLS);
59
60 OLSWHITE = table(B(:,2), SEWHITE(:,2), TSTAT(:,2));
61 OLSWHITE.Properties.VariableNames = {'Beta' 'OLS_SEWHITE' 'T_stat' };
62 disp(OLSWHITE);
63
64 % compute HAC standard erros
65 B_hac = NaN(4,2);
66 TSTAT_hac = NaN(4,2);
67 S2_hac = NaN(4,1);
68 SE_hac = NaN(4,2);
69 VCVNW_hac = NaN(2,2);
70
71 for i=1:4
72     [B_hac(i,:), TSTAT_hac(i,:), S2_hac(i,1), VCVNW_hac] = olsnw(
73         yieldchanges(1:end-i*12,i), fwdspread(1:end-i*12,i));
74     SE_hac(i,:) = sqrt(diag(VCVNW_hac));
75 end
76
77 HAC = table(B_hac(:,2), SE_hac(:,2), TSTAT_hac(:,2));
78 HAC.Properties.VariableNames = {'Beta_hac' 'Hac_SE' 'T_stat' };
79 disp(HAC);
80
81 %% question 2
82 % generate matrix for holding period returns
83 hpr = NaN(length(logprices(:,1)),4);
84 for n=1:4
85     hpr(1:end-12,n) = logprices(13:end,n) - logprices(1:end-12,n+1);
86 end
87 % generate matrix for dependent variables for regression 2
88 hprchanges = NaN(length(logprices(:,1)),4);
89 for n=1:4
90     hprchanges(:,n) = hpr(:,n) - yields(:,1);
91 end
92
93 % figure(1)
94 % plot(fwdspread)
95 %
96 % figure(3)
97 % plot(hprchanges)
98
99 % compute OLS standard error
100 B2 = NaN(4,2);
101 TSTAT2 = NaN(4,2);
102 S22 = NaN(4,1);
103 SE22 = NaN(4,2);
104 SEWHITE2 = NaN(4,2);
105 VCV2 = NaN(2,2);
106 VCVWHITE2 = NaN(2,2);

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107 for i=1:4
108     [B2(i,:),TSTAT2(i,:),S22(i,1),VCV2,VCVWHITE2] = ols(hprchanges(1
        :end-12,i),fwdspread(1:end-12,i));
109     SE2(i,:)=sqrt(diag(VCV2));
110     SEWHITE2(i,:)=sqrt(diag(VCVWHITE2));
111 end
112
113 OLS2 = table(B2(:,2),SE2(:,2),TSTAT2(:,2));
114 OLS2.Properties.VariableNames = {'Gamma' 'OLS_SE' 'T_stat' };
115 disp(OLS2);
116
117 OLSWHITE2 = table(B2(:,2),SEWHITE2(:,2),TSTAT2(:,2));
118 OLSWHITE2.Properties.VariableNames = {'Gamma' 'OLS_SEWHITE' 'T_stat'
        ' '};
119 disp(OLSWHITE2);
120
121 % compute HAC standard erros
122 B2_hac = NaN(4,2);
123 TSTAT2_hac = NaN(4,2);
124 S22_hac = NaN(4,1);
125 SE2_hac = NaN(4,2);
126 VCVNW2_hac = NaN(2,2);
127
128 for i=1:4
129     [B2_hac(i,:),TSTAT2_hac(i,:),S22_hac(i,1),VCVNW2_hac] = olsnw(
        hprchanges(1:end-12,i),fwdspread(1:end-12,i));
130     SE2_hac(i,:)=sqrt(diag(VCVNW2_hac));
131 end
132
133 HAC2 = table(B2_hac(:,2),SE2_hac(:,2),TSTAT2_hac(:,2));
134 HAC2.Properties.VariableNames = {'Gamma_hac' 'Hac_SE' 'T_stat' };
135 disp(HAC2);

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