Investments HW 3

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Problem 1

a

$$Var(A + B + C + D) = w_a^2 Var(A) + w_b^2 Var(B) + w_c^2 Var(C) + w_d^2 Var(D) + 2 * (w_a w_b cov(A, B) + w_a w_c cov(A, C) + w_a w_d cov(A, D) + w_b w_c cov(B, C) + w_b w_d cov(B, D) + w_c w_d cov(C, D))$$

$$Var(Mrkt) = .0715$$

$$Cov(A, A + B + C + D) = Var(A) + Cov(A, B) + C(A, C) + Cov(A, D)$$

Similarly for B,C,D. Interestingly, this corresponds to the sum of the elements in any given row of the Variance Covariance matrix!

$$Cov(A, P) = 0.2308$$

 $Cov(B, P) = 0.2660$
 $Cov(C, P) = 0.3012$
 $Cov(D, P) = 0.3364$

b

$$\beta = \frac{Cov(P, A)}{\sigma_P^2}$$

$$\beta_A = 3.226$$

$$\beta_B = 3.718$$

$$\beta_C = 4.210$$

$$\beta_D = 4.702$$

 \mathbf{c}

$$r_A - r_f = \beta_A * [E(r_m) - r_f]$$

$$E(r_A - r_f) = 0.226$$

$$E(r_B - r_f) = 0.260$$

$$E(r_C - r_f) = 0.295$$

$$E(r_D - r_f) = 0.329$$

 \mathbf{d}

MVE Weights of the Portfolio:

A: 24%

```
C: 25%
D: 26%
The MVE weights are exactly the percentage of each firm over the value of the market.
e MVE Weights of the Portfolio:
A: 25%
B: 25%
C: 25%
D: 25%
The weights have changed.
f
MVE Weights of the Portfolio:
A: 29%
B: 23%
C: 24%
D: 24%
Var(P) = .0028
```

Problem 2

B: 25%

• Format Data

```
openFormat <- function(x, skips){
    y <- read.csv(x, skip=skips, nrows=1068)
    y <- rename(y, c("X"="Date"))
    y <- y[y$Date > 193200,]
    tmp <- y$Date
    y <- y[,-1]/100
    y$Date <- tmp
    y
}

mrkt <- openFormat("F-F_Research_Data_Factors.CSV", 3)
ports <- openFormat("25_Portfolios_5x5.CSV", 19)
dates <- ports$Date
ports <- ports - mrkt$RF
ports$Date <- dates</pre>
```

 \boldsymbol{a}

```
regressors <- names(ports)[-length(ports)]
coefs <- data.frame(Index=c(1,2), row.names = c("alpha", "beta"))
for(i in regressors){
   cs <- coefficients(lm(ports[,i] ~ mrkt$Mkt.RF))
   coefs <- cbind(coefs, rbind(round(as.numeric(cs["(Intercept)"]),5), round(as.numeric(cs["mrkt$Mkt.RF"]))</pre>
```

```
}
coefs <- coefs[,-1]
names(coefs) <- regressors
# pandoc.table(coefs,split.tables=90 )</pre>
```

	SMALL.LoBM	ME1.BM2	ME1.BM3	ME1.BM4	SMALL.HiBM
alpha	-0.00396	-0.00117	0.00199	0.00443	0.00524
beta	1.651	1.444	1.395	1.274	1.402
	ME2.BM1	ME2.BM2	ME2.BM3	ME2.BM4	ME2.BM5
alpha	-0.00161	0.0013	0.00259	0.00347	0.00375
\mathbf{beta}	1.306	1.262	1.23	1.243	1.394
	ME3.BM1	ME3.BM2	ME3.BM3	ME3.BM4	ME3.BM5
alpha	-0.00082	0.00183	0.00219	0.00293	0.00309
\mathbf{beta}	1.281	1.132	1.149	1.18	1.375
	ME4.BM1	ME4.BM2	ME4.BM3	ME4.BM4	ME4.BM5
alpha	0.00011	0.00046	0.0018	0.00245	0.0014
\mathbf{beta}	1.085	1.095	1.144	1.159	1.413
	BIG.LoBM	ME5.BM2	ME5.BM3	ME5.BM4	BIG.HiBM
alpha	-0.00029	2e-05	0.00084	0.00028	0.00147
\mathbf{beta}	0.9492	0.943	0.9748	1.089	1.245

b

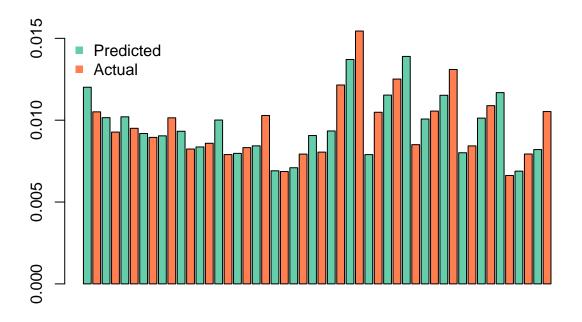
```
coefs <- rbind(coefs, sapply(ports, mean))
rownames(coefs) <- c("alpha", "beta", "actual")
coefs <- data.frame(t(coefs))
kable(coefs)</pre>
```

	alpha	beta	actual
SMALL.LoBM	-0.00396	1.65058	0.0080490
ME1.BM2	-0.00117	1.44401	0.0093410
ME1.BM3	0.00199	1.39526	0.0121487
ME1.BM4	0.00443	1.27417	0.0137078
SMALL.HiBM	0.00524	1.40240	0.0154456
ME2.BM1	-0.00161	1.30590	0.0078980
ME2.BM2	0.00130	1.26222	0.0104836
ME2.BM3	0.00259	1.22953	0.0115339
ME2.BM4	0.00347	1.24267	0.0125122

	alpha	beta	actual
ME2.BM5	0.00375	1.39384	0.0138949
ME3.BM1	-0.00082	1.28106	0.0085043
ME3.BM2	0.00183	1.13201	0.0100712
ME3.BM3	0.00219	1.14924	0.0105557
ME3.BM4	0.00293	1.18019	0.0115219
ME3.BM5	0.00309	1.37529	0.0130966
ME4.BM1	0.00011	1.08544	0.0080107
ME4.BM2	0.00046	1.09543	0.0084292
ME4.BM3	0.00180	1.14362	0.0101255
ME4.BM4	0.00245	1.15873	0.0108878
ME4.BM5	0.00140	1.41345	0.0116846
BIG.LoBM	-0.00029	0.94915	0.0066222
ME5.BM2	0.00002	0.94301	0.0068857
ME5.BM3	0.00084	0.97485	0.0079334
ME5.BM4	0.00028	1.08938	0.0082060
BIG.HiBM	0.00147	1.24520	0.0105276

 \mathbf{c}

```
lm.capm <- lm(actual ~ beta, coefs)</pre>
summary(lm.capm)
##
## Call:
## lm(formula = actual ~ beta, data = coefs)
##
## Residuals:
##
          Min
                      1Q
                             Median
                                             3Q
## -0.0050694 -0.0013277 0.0001523 0.0012318 0.0039872
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.002078
                          0.003135
                                      0.663
                                              0.5140
## beta
              0.006689
                          0.002521
                                      2.654
                                              0.0142 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.002092 on 23 degrees of freedom
## Multiple R-squared: 0.2344, Adjusted R-squared: 0.2011
## F-statistic: 7.042 on 1 and 23 DF, p-value: 0.01419
\mathbf{d}
\verb|coefs$predictions| <- \verb|mean(mrkt$Mkt.RF)| * coefs$beta|\\
barplot(c(coefs$predictions, coefs$actual), beside = T, col=c("aquamarine3","coral"))
legend("topleft", c("Predicted", "Actual"), pch=15, col=c("aquamarine3", "coral"), bty="n")
```



Problem 3

 \mathbf{a}

```
coefs <- data.frame(Index=c(1,2,3,4), row.names = c("alpha", "betaMrkt", "betaSMB", "betaHML"))
for(i in regressors){
   cs <- coefficients(lm(ports[,i] ~ mrkt$Mkt.RF + mrkt$SMB + mrkt$HML))
   coefs <- cbind(coefs, rbind(round(as.numeric(cs["(Intercept)"]),5), round(as.numeric(cs["mrkt$Mkt.RF"]))
}
coefs <- coefs[,-1]
names(coefs) <- regressors</pre>
```

b

```
coefs <- rbind(coefs, sapply(ports, mean))
rownames(coefs) <- c(rownames(coefs)[-length(rownames(coefs))], "actual")
# pandoc.table(coefs,split.tables=90)</pre>
```

	${\bf SMALL.LoBM}$	ME1.BM2	ME1.BM3	ME1.BM4	SMALL.HiBM
alpha	-0.00723	-0.00373	-0.00112	0.00105	0.00063
${f betaMrkt}$	1.251	1.07	1.051	0.918	0.9758
betaSMB	1.513	1.54	1.227	1.227	1.327
${f beta}{f H}{f M}{f L}$	0.4489	0.2297	0.5043	0.5848	0.917
actual	0.008049	0.009341	0.01215	0.01371	0.01545

	ME2.BM1	ME2.BM2	ME2.BM3	ME2.BM4	ME2.BM5
alpha	-0.00215	-0.00033	0.00038	0.00062	-0.00029
${f betaMrkt}$	1.088	1.027	0.9935	0.9797	1.066
${\bf betaSMB}$	1.136	0.9661	0.8235	0.8167	0.8945
${f beta}{f H}{f M}{f L}$	-0.2344	0.1497	0.3718	0.568	0.8982
actual	0.007898	0.01048	0.01153	0.01251	0.01389

	ME2.BM1	ME2.BM2	ME2.BM3	ME2.BM4	ME2.BM5
	ME3.BM1	ME3.BM2	ME3.BM3	ME3.BM4	ME3.BM5
alpha	-0.00098	0.00109	0.00061	0.00055	-0.00053
${f betaMrkt}$	1.134	1.013	1.007	1	1.116
${\bf betaSMB}$	0.8219	0.5067	0.4327	0.4491	0.5997
$\mathbf{betaHML}$	-0.2384	0.04692	0.3229	0.5565	0.8743
actual	0.008504	0.01007	0.01056	0.01152	0.0131
	ME4.BM1	ME4.BM2	ME4.BM3	ME4.BM4	ME4.BM5
alpha	0.00091	-0.00012	4e-04	0.00031	-0.00217
${f betaMrkt}$	1.064	1.034	1.048	1.027	1.207
${\bf betaSMB}$	0.3312	0.2162	0.2006	0.2261	0.305
$\mathbf{betaHML}$	-0.3546	0.09609	0.3503	0.5639	0.9602
actual	0.008011	0.008429	0.01013	0.01089	0.01168
	BIG.LoBM	ME5.BM2	ME5.BM3	ME5.BM4	BIG.HiBM
alpha	0.00073	0.00017	7e-05	-0.00166	-0.00164
${f betaMrkt}$	1.019	0.9854	0.981	1.034	1.134
${\bf betaSMB}$	-0.15	-0.2122	-0.2313	-0.1686	-0.1353

0.0313

0.006886

0.311

0.007933

0.6385

0.008206

0.9772

0.01053

coefs <- data.frame(t(coefs))
kable(coefs)</pre>

-0.2519

0.006622

 ${\bf beta HML}$

actual

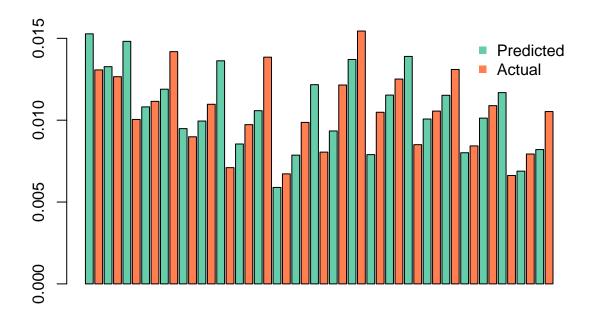
	alpha	${\bf betaMrkt}$	${\bf betaSMB}$	${\it betaHML}$	actual
SMALL.LoBM	-0.00723	1.25073	1.51317	0.44893	0.0080490
ME1.BM2	-0.00373	1.06993	1.54024	0.22968	0.0093410
ME1.BM3	-0.00112	1.05073	1.22688	0.50435	0.0121487
ME1.BM4	0.00105	0.91796	1.22694	0.58482	0.0137078
SMALL.HiBM	0.00063	0.97585	1.32734	0.91703	0.0154456
ME2.BM1	-0.00215	1.08840	1.13642	-0.23440	0.0078980
ME2.BM2	-0.00033	1.02676	0.96611	0.14971	0.0104836
ME2.BM3	0.00038	0.99346	0.82347	0.37182	0.0115339
ME2.BM4	0.00062	0.97966	0.81675	0.56803	0.0125122
ME2.BM5	-0.00029	1.06580	0.89446	0.89824	0.0138949
ME3.BM1	-0.00098	1.13374	0.82191	-0.23845	0.0085043
ME3.BM2	0.00109	1.01309	0.50674	0.04692	0.0100712
ME3.BM3	0.00061	1.00672	0.43267	0.32289	0.0105557
ME3.BM4	0.00055	1.00021	0.44911	0.55647	0.0115219
ME3.BM5	-0.00053	1.11593	0.59972	0.87431	0.0130966
ME4.BM1	0.00091	1.06353	0.33118	-0.35457	0.0080107
ME4.BM2	-0.00012	1.03368	0.21618	0.09609	0.0084292
ME4.BM3	0.00040	1.04850	0.20055	0.35027	0.0101255
ME4.BM4	0.00031	1.02701	0.22608	0.56395	0.0108878

	alpha	betaMrkt	betaSMB	betaHML	actual
ME4.BM5	-0.00217	1.20685	0.30503	0.96021	0.0116846
BIG.LoBM	0.00073	1.01884	-0.15002	-0.25193	0.0066222
ME5.BM2	0.00017	0.98542	-0.21217	0.03130	0.0068857
ME5.BM3	0.00007	0.98097	-0.23125	0.31101	0.0079334
ME5.BM4	-0.00166	1.03420	-0.16865	0.63847	0.0082060
BIG.HiBM	-0.00164	1.13357	-0.13526	0.97718	0.0105276

C

```
lm.ff3f <- lm(actual ~ betaMrkt + betaSMB + betaHML, coefs)</pre>
summary(lm.ff3f)
##
## lm(formula = actual ~ betaMrkt + betaSMB + betaHML, data = coefs)
## Residuals:
                     1Q
                           Median
                                          3Q
                                                    Max
## -0.0020975 -0.0001985 0.0001491 0.0006034 0.0014887
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.0190361 0.0031442 6.054 5.23e-06 ***
## betaMrkt
              ## betaSMB
               0.0018208 0.0004059
                                     4.485 0.000204 ***
## betaHML
               0.0043667 0.0005628
                                    7.758 1.34e-07 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.00109 on 21 degrees of freedom
## Multiple R-squared: 0.8102, Adjusted R-squared: 0.7831
## F-statistic: 29.88 on 3 and 21 DF, p-value: 9.103e-08
\mathbf{d}
coefs$predictions <- mean(mrkt$Mkt.RF) * coefs$betaMrkt +</pre>
 mean(mrkt$SMB) * coefs$betaSMB +
 mean(mrkt$HML) * coefs$betaHML
```

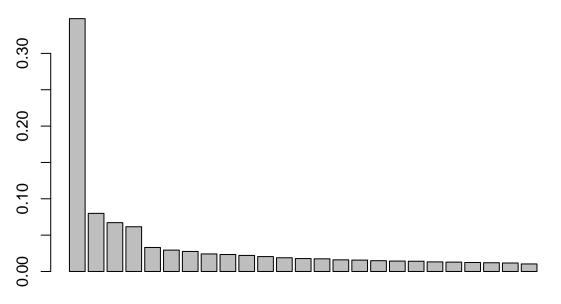
barplot(c(coefs\$predictions, coefs\$actual), beside = T, col=c("aquamarine3","coral"))
legend("topright", c("Predicted","Actual"), pch=15, col=c("aquamarine3","coral"), bty="n")



Problem 4

 \mathbf{a}

```
covMat <- cov(subset(ports, select = -(Date)))
eigens <- eigen(covMat)
barplot(sqrt(eigens$values))</pre>
```



```
loadings <- eigens$vectors[c(1:5),]
pc1 <- matrix(loadings[1,], nrow = 5, ncol = 5)
pc2 <- matrix(loadings[2,], nrow = 5, ncol = 5)
pc3 <- matrix(loadings[3,], nrow = 5, ncol = 5)
pc4 <- matrix(loadings[4,], nrow = 5, ncol = 5)
kable(pc1)</pre>
```

-0.2902098	0.2012026	0.0259154	-0.0405035	0.0058789
-0.6087921	-0.0772449	-0.0780085	0.0052111	0.0203759
-0.4981649	-0.0702421	-0.0427339	-0.0051010	0.0246795
0.4712854	0.0345290	0.0695276	0.0051879	-0.0053488
-0.0533622	-0.0088743	-0.0234012	-0.0509555	0.0010785

kable(pc2)

-0.2486496	0.0430704	-0.2147033	-0.0199065	0.0404258
-0.2721463	-0.0174940	-0.0976253	0.0292416	0.0248087
0.3545691	-0.0774821	0.0783678	0.0715920	-0.0152345
-0.2524731	0.1829801	-0.0551747	-0.0744769	0.0288245
-0.7391359	0.0564767	-0.0580795	-0.0081577	-0.0000387

kable(pc3)

-0.2385009	-0.0285400	0.0487068	-0.1692358	-0.0550929
-0.1457029	0.6923218	-0.0035629	-0.0359713	0.0163645
0.0228324	-0.3072733	0.3476169	0.0607575	-0.0168612
-0.1635031	0.1531875	-0.1650339	-0.0715323	-0.0103402
0.2937583	-0.1102879	-0.0125181	0.0699665	-0.0148553

kable(pc4)

-0.2224945	0.0215300	0.2229385	0.1656413	-0.2794113
-0.1454130	0.1185867	0.0827552	-0.0843345	0.0513415
0.0808813	-0.0126698	-0.2626865	-0.0738195	0.0865126
-0.2768872	-0.3047850	0.0058318	0.4801660	-0.2038050
0.0152032	0.0602267	-0.1348032	-0.4097434	0.1508923