

Principal Component Analysis and Self Organized Maps

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In this part, I employ Principal Component Analysis (PCA) and Self-Organized Map (SOM) algorithm to gain insights about 20 companies based on their exposures to Fama French 3-factors, their alphas, volatility, adjusted R-squared, Sharpe Ratio, and Information Ratio.

Installing relevant packages.

```
warning = FALSE
library('kohonen') #SOM algorithm

## Warning: package 'kohonen' was built under R version 4.3.3
library('gmm') #data set 'Finance'

## Warning: package 'gmm' was built under R version 4.3.3
## Loading required package: sandwich
library('data.table') #data manipulation

## Warning: package 'data.table' was built under R version 4.3.2
library('ggplot2') #visualization

## Warning: package 'ggplot2' was built under R version 4.3.3
library('frenchdata') #optional library

## Warning: package 'frenchdata' was built under R version 4.3.3
```

Creating a function that estimates model and some additional coefficients.

```
mylm<-function(y,x1,x2,x3)
{
  fit<-lm(y~x1+x2+x3,na.action='na.omit')
  IVOLhat<-summary(fit)$sigma
  adjR2<-summary(fit)$adj.r.squared
  m<-mean(y,na.rm=TRUE)
  s<-sd(y,na.rm=TRUE)
  SRhat<-m/s
  IRhat<-coef(fit)[1]/IVOLhat

  coefficients<-c(coef(fit),IVOLhat,adjR2,SRhat,IRhat)
  names(coefficients)<-c('alphahat','betahat','shat','hhat',
                        'sigmahat','adjR2','SRhat','IRhat')

  return(coefficients)
}
```

Using the Finance dataset to extract relevant variables including risk free rate, market return, and fama french factors. Then taking the differences and creating a dataframe for the resulting values.

```
data(Finance)
rstock<-Finance[,1:(ncol(Finance)-4)]
rf<-Finance[, 'rf']
rm<-Finance[, 'rm']
rmexcess<-rm-rf
smb<-Finance[, 'smb']
hml<-Finance[, 'hml']

rstockexcess<-sweep(x=rstock,MARGIN=1,STATS=rf,FUN="-")

EST<-apply(X=rstockexcess,MARGIN=2,FUN="mylm",x1=rmexcess,x2=smb,x3=hml)
EST<-t(EST)
X<-data.frame(EST)
X
```

##	alphahat	betahat	shat	hhat	sigmahat	adjR2
## WMK	-0.014210460	0.5840843	0.36911382	0.511735796	1.358008	0.18523640
## UIS	-0.102665430	1.3988750	0.53399668	0.145826813	3.318372	0.18928825
## ORB	-0.021097135	1.0938957	0.72947598	0.007998101	3.454441	0.12450783
## MAT	-0.011163040	0.7526733	0.24607336	0.181042201	2.145538	0.13388211
## ABAX	-0.024328233	0.8789724	1.04726575	0.588227848	4.468547	0.05252895
## T	-0.013939992	0.8260146	-0.53430829	0.302761692	1.576504	0.29505140
## EMR	0.004530843	0.9821885	-0.23923261	0.329103711	1.334034	0.41823207
## JCS	-0.007303786	0.3986744	0.17779106	-0.146393605	2.927464	0.02844975
## VOXX	-0.089042429	1.5221496	1.68674327	0.455771776	3.647425	0.20574968
## ZOOM	-0.116170587	0.6869271	0.63936729	-0.432720695	5.830268	0.02703197
## ROG	0.007446174	1.0472297	1.00092477	0.473072594	2.024605	0.26698611
## GGG	0.030862329	0.8654158	0.58254971	0.429910587	1.656164	0.25437910
## PC	-0.021085353	0.7644425	0.19917982	0.181481864	1.738315	0.19561619
## GCD	-0.025764752	1.1857563	0.97370253	0.532418981	3.343363	0.14077712
## EBF	-0.020935622	0.8211295	0.68184392	0.773311728	1.898858	0.19403897
## F	-0.073920555	1.2835707	-0.15107610	0.954673073	2.218184	0.29803587
## FNM	-0.122733423	1.2229485	-0.34080598	0.969911593	4.760113	0.08058095
## NHP	0.011309042	0.9495598	0.69759993	0.917312172	1.489851	0.33972671
## AA	-0.041765915	1.3217150	-0.09407038	0.755323149	1.932294	0.36847230
## TDW	-0.015788479	1.0459957	0.32450483	0.721491343	2.376836	0.18721783
##	SRhat	IRhat				
## WMK	0.0033174561	-0.010464195				
## UIS	-0.0209572487	-0.030938489				
## ORB	-0.0007116718	-0.006107250				
## MAT	0.0018833769	-0.005202910				
## ABAX	0.0004499701	-0.005444328				
## T	0.0019147315	-0.008842343				
## EMR	0.0145806160	0.003396348				
## JCS	-0.0011732275	-0.002494919				
## VOXX	-0.0131834043	-0.024412411				
## ZOOM	-0.0190067214	-0.019925429				
## ROG	0.0144853833	0.003677841				
## GGG	0.0277944915	0.018634827				
## PC	-0.0028225406	-0.012129765				
## GCD	0.0011786500	-0.007706239				
## EBF	0.0034695652	-0.011025374				

```
## F      -0.0138021589 -0.033324807
## FNM    -0.0173938483 -0.025783719
## NHP     0.0241465678  0.007590720
## AA     -0.0030334421 -0.021614678
## TDW     0.0054656144 -0.006642646
```

Annualizing and converting to unit-less value for better interpretability.

```
sweep(x=X[,c('alphahat','sigmahat','SRhat','IRhat')],
      STATS=c(252*0.01,sqrt(252)*0.01,sqrt(252),sqrt(252)),MARGIN=2,FUN='*')
```

```
##          alphahat  sigmahat          SRhat          IRhat
## WMK  -0.03581036  0.2155771  0.052662982 -0.16611394
## UIS  -0.25871688  0.5267753 -0.332686009 -0.49113329
## ORB  -0.05316478  0.5483755 -0.011297439 -0.09694959
## MAT  -0.02813086  0.3405936  0.029897681 -0.08259363
## ABAX -0.06130715  0.7093598  0.007143054 -0.08642603
## T    -0.03512878  0.2502623  0.030395420 -0.14036784
## EMR   0.01141773  0.2117713  0.231460103  0.05391535
## JCS  -0.01840554  0.4647206 -0.018624409 -0.03960561
## VOXX -0.22438692  0.5790107 -0.209280055 -0.38753502
## ZOOM -0.29274988  0.9255263 -0.301722348 -0.31630639
## ROG   0.01876436  0.3213960  0.229948331  0.05838392
## GGG   0.07777307  0.2629079  0.441223873  0.29581871
## PC   -0.05313509  0.2759490 -0.044806443 -0.19255405
## GCO  -0.06492717  0.5307424  0.018710489 -0.12233275
## EBF  -0.05275777  0.3014344  0.055077640 -0.17502238
## F    -0.18627980  0.3521258 -0.219102481 -0.52901491
## FNM  -0.30928823  0.7556446 -0.276118781 -0.40930385
## NHP   0.02849879  0.2365065  0.383314880  0.12049894
## AA   -0.10525011  0.3067422 -0.048154400 -0.34312238
## TDW  -0.03978697  0.3773110  0.086763939 -0.10544874
```

```
pca_results <- prcomp(X, scale = TRUE)
print(summary(pca_results))
```

Principal Component Analysis

Importance of components:

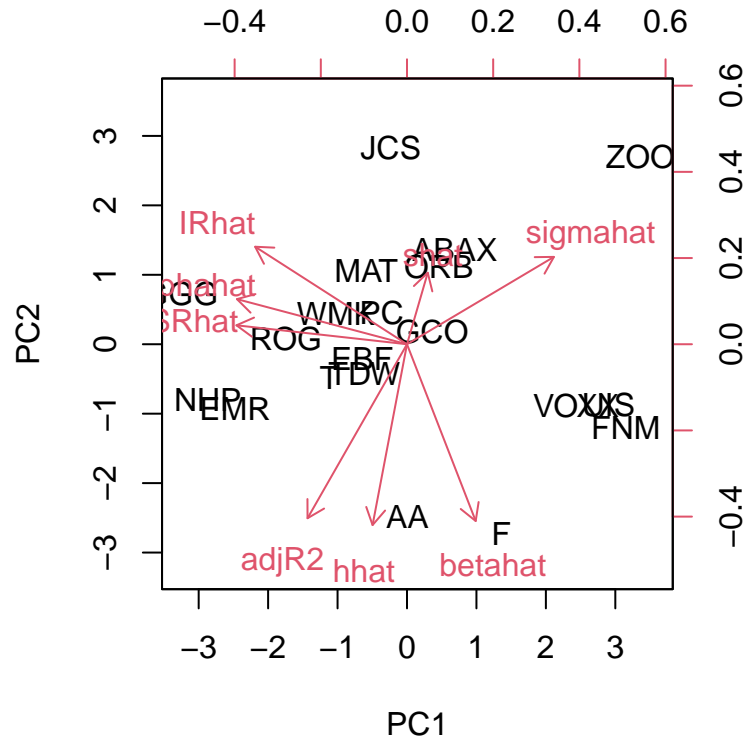
```
##          PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation  1.9355  1.4323  1.0979  0.74263  0.54412  0.3382  0.18026
## Proportion of Variance 0.4683  0.2564  0.1507  0.06894  0.03701  0.0143  0.00406
## Cumulative Proportion 0.4683  0.7247  0.8754  0.94432  0.98133  0.9956  0.99969
##          PC8
## Standard deviation   0.04952
## Proportion of Variance 0.00031
## Cumulative Proportion 1.00000
```

```
pca_loadings <- pca_results$rotation
round(pca_loadings,4)
```

```
##          PC1      PC2      PC3      PC4      PC5      PC6      PC7      PC8
## alphahat -0.4933  0.1309 -0.0605  0.0077  0.1261  0.5043  0.6823 -0.0090
## betahat  0.1993 -0.5127 -0.4059 -0.2668 -0.3573  0.5585 -0.1365  0.0571
## shat     0.0602  0.2074 -0.8248 -0.1763  0.4316 -0.2277  0.0203 -0.0591
## hhat     -0.0999 -0.5245 -0.2021  0.7888  0.1214 -0.1090  0.0577 -0.1482
```

```
## sigmahat  0.4261  0.2533 -0.1927  0.2096 -0.5953 -0.2215  0.5146  0.0701
## adjR2     -0.2885 -0.5047  0.0480 -0.4665 -0.1390 -0.5425  0.3223 -0.1569
## SRhat     -0.4946  0.0549 -0.1948  0.0984 -0.2628 -0.1633 -0.2073  0.7523
## IRhat     -0.4404  0.2828 -0.1819  0.0656 -0.4612  0.0060 -0.3167 -0.6129
```

```
biplot(pca_results, scale=0)
```



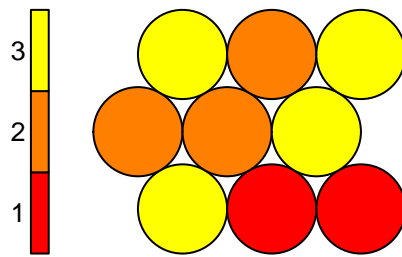
Self Organizing Maps (SOM) Using unsupervised SOMs to visually inspect common variables.

```
xdim<-ydim<-3
Xst<-scale(as.matrix(X))#convert X to a matrix and standardize the columns of X

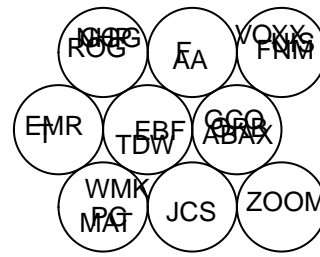
mysom <- som(X=Xst, grid = somgrid(xdim=xdim, ydim=ydim, 'hexagonal'))

par(mfrow=c(2,2),mar=c(2, 4, 1, 1),cex=0.8)
plot(mysom,type='counts')
plot(mysom,type='mapping',labels=rownames(X))
plot(mysom,type='codes')
```

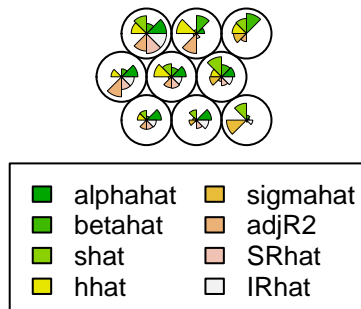
Counts plot



Mapping plot



Codes plot



Extracting relevant nodes and centroids for each stock.

```
info<-data.frame(ticker=rownames(X),node=mysom$unit.classif)
info
```

```
##      ticker node
## 1      WMT     1
## 2      UIS     9
## 3      ORB     6
## 4      MAT     1
## 5     ABAX     6
## 6        T     4
## 7     EMR     4
## 8      JCS     2
## 9     VOXX     9
## 10     ZOOM     3
## 11     ROG     7
## 12     GGG     7
## 13      PC     1
## 14     GCO     6
## 15     EBF     5
## 16      F      8
## 17     FNM     9
## 18     NHP     7
## 19      AA     8
## 20     TDW     5
```

```
table(x=info$node)
```

```
## x
## 1 2 3 4 5 6 7 8 9
## 3 1 1 2 2 3 3 2 3
```

```
mysom$codes
```

```
## [[1]]
##      alphahat      betahat      shat      hhat      sigmahat      adjR2
## V1  0.4008721 -0.93740870 -0.29388775 -0.4148888 -0.7126300 -0.24526069
## V2  0.4182431 -1.49883676 -0.22410674 -1.1642135  0.1915622 -1.20599538
## V3 -1.2350603 -0.80464620  0.50327122 -1.6800166  2.0331034 -1.41857769
## V4  0.6056278 -0.35732579 -1.33635166 -0.2396122 -0.9106607  1.14331559
## V5  0.3704263 -0.15921165  0.08145582  0.7427249 -0.4389867 -0.02068377
## V6  0.1833441  0.24525350  0.84435446 -0.2047074  0.8458508 -0.82891788
## V7  1.1039710 -0.05986914  0.59320021  0.4271646 -0.7434531  0.82381744
## V8 -0.5356395  1.07781845 -0.92531996  1.0780626 -0.4299136  1.13437401
## V9 -1.6297899  1.37504768  0.30330658  0.3022795  0.9918710 -0.38332335
##      SRhat      IRhat
## V1  0.01558959  0.04909615
## V2 -0.11561168  0.38276032
## V3 -1.02534237 -0.42826551
## V4  0.50391569  0.44454336
## V5  0.34044435  0.13491506
## V6 -0.03663810  0.23296916
## V7  1.57159730  1.43573896
## V8 -0.61029619 -1.22964871
## V9 -1.31260862 -1.28750184
```

```
df.centroids<-data.frame(mysom$codes)
df.centroids$node<-paste0('Node',1:nrow(df.centroids))
df.centroids$node<-factor(df.centroids$node,labels=paste0('Node',1:nrow(df.centroids)))
dt<-data.table(df.centroids)
dt
```

```
##      alphahat      betahat      shat      hhat      sigmahat      adjR2
## 1:  0.4008721 -0.93740870 -0.29388775 -0.4148888 -0.7126300 -0.24526069
## 2:  0.4182431 -1.49883676 -0.22410674 -1.1642135  0.1915622 -1.20599538
## 3: -1.2350603 -0.80464620  0.50327122 -1.6800166  2.0331034 -1.41857769
## 4:  0.6056278 -0.35732579 -1.33635166 -0.2396122 -0.9106607  1.14331559
## 5:  0.3704263 -0.15921165  0.08145582  0.7427249 -0.4389867 -0.02068377
## 6:  0.1833441  0.24525350  0.84435446 -0.2047074  0.8458508 -0.82891788
## 7:  1.1039710 -0.05986914  0.59320021  0.4271646 -0.7434531  0.82381744
## 8: -0.5356395  1.07781845 -0.92531996  1.0780626 -0.4299136  1.13437401
## 9: -1.6297899  1.37504768  0.30330658  0.3022795  0.9918710 -0.38332335
##      SRhat      IRhat      node
## 1:  0.01558959  0.04909615 Node1
## 2: -0.11561168  0.38276032 Node2
## 3: -1.02534237 -0.42826551 Node3
## 4:  0.50391569  0.44454336 Node4
## 5:  0.34044435  0.13491506 Node5
## 6: -0.03663810  0.23296916 Node6
## 7:  1.57159730  1.43573896 Node7
## 8: -0.61029619 -1.22964871 Node8
```

```
## 9: -1.31260862 -1.28750184 Node9
```

Melting data from wide to long.

```
dt.molten<-melt.data.table(data=dt, id.vars='node')
dt.molten
```

##	node	variable	value
## 1:	Node1	alphahat	0.40087212
## 2:	Node2	alphahat	0.41824308
## 3:	Node3	alphahat	-1.23506032
## 4:	Node4	alphahat	0.60562778
## 5:	Node5	alphahat	0.37042630
## 6:	Node6	alphahat	0.18334414
## 7:	Node7	alphahat	1.10397100
## 8:	Node8	alphahat	-0.53563945
## 9:	Node9	alphahat	-1.62978990
## 10:	Node1	betahat	-0.93740870
## 11:	Node2	betahat	-1.49883676
## 12:	Node3	betahat	-0.80464620
## 13:	Node4	betahat	-0.35732579
## 14:	Node5	betahat	-0.15921165
## 15:	Node6	betahat	0.24525350
## 16:	Node7	betahat	-0.05986914
## 17:	Node8	betahat	1.07781845
## 18:	Node9	betahat	1.37504768
## 19:	Node1	shat	-0.29388775
## 20:	Node2	shat	-0.22410674
## 21:	Node3	shat	0.50327122
## 22:	Node4	shat	-1.33635166
## 23:	Node5	shat	0.08145582
## 24:	Node6	shat	0.84435446
## 25:	Node7	shat	0.59320021
## 26:	Node8	shat	-0.92531996
## 27:	Node9	shat	0.30330658
## 28:	Node1	hhhat	-0.41488883
## 29:	Node2	hhhat	-1.16421347
## 30:	Node3	hhhat	-1.68001664
## 31:	Node4	hhhat	-0.23961215
## 32:	Node5	hhhat	0.74272489
## 33:	Node6	hhhat	-0.20470742
## 34:	Node7	hhhat	0.42716465
## 35:	Node8	hhhat	1.07806264
## 36:	Node9	hhhat	0.30227949
## 37:	Node1	sigmahat	-0.71263001
## 38:	Node2	sigmahat	0.19156219
## 39:	Node3	sigmahat	2.03310341
## 40:	Node4	sigmahat	-0.91066071
## 41:	Node5	sigmahat	-0.43898672
## 42:	Node6	sigmahat	0.84585080
## 43:	Node7	sigmahat	-0.74345311
## 44:	Node8	sigmahat	-0.42991365
## 45:	Node9	sigmahat	0.99187104
## 46:	Node1	adjR2	-0.24526069
## 47:	Node2	adjR2	-1.20599538

```
## 48: Node3      adjR2 -1.41857769
## 49: Node4      adjR2  1.14331559
## 50: Node5      adjR2 -0.02068377
## 51: Node6      adjR2 -0.82891788
## 52: Node7      adjR2  0.82381744
## 53: Node8      adjR2  1.13437401
## 54: Node9      adjR2 -0.38332335
## 55: Node1      SRhat  0.01558959
## 56: Node2      SRhat -0.11561168
## 57: Node3      SRhat -1.02534237
## 58: Node4      SRhat  0.50391569
## 59: Node5      SRhat  0.34044435
## 60: Node6      SRhat -0.03663810
## 61: Node7      SRhat  1.57159730
## 62: Node8      SRhat -0.61029619
## 63: Node9      SRhat -1.31260862
## 64: Node1      IRhat  0.04909615
## 65: Node2      IRhat  0.38276032
## 66: Node3      IRhat -0.42826551
## 67: Node4      IRhat  0.44454336
## 68: Node5      IRhat  0.13491506
## 69: Node6      IRhat  0.23296916
## 70: Node7      IRhat  1.43573896
## 71: Node8      IRhat -1.22964871
## 72: Node9      IRhat -1.28750184
##      node variable      value
```

Plotting the centroids on a bar graph.

```
g<-ggplot(data=dt.molten, aes(x=variable,y=value))
g<-g+stat_summary(fun='mean', geom='bar')+facet_wrap(~node,nrow=xdim,as.table=FALSE)
g<-g+theme(axis.text.x=element_text(angle=-90))
print(g)
```