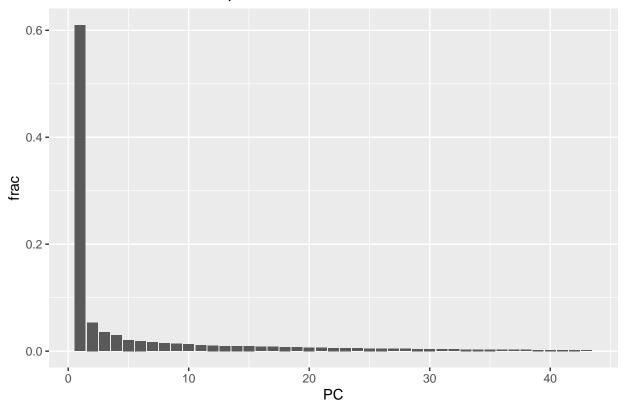
# Empirical Methods in Finance Homework 7

Group 9: Linqi Huang, Abhesh Kumar, Yu Onohara, Maitrayee Patil, Redmond Xia

#### Problem 1

```
# Importing libraries and organizing data
suppressMessages(library(tidyverse))
suppressMessages(library(readxl))
suppressMessages(library(ggrepel))
suppressMessages(library(stargazer))
suppressMessages(library(xtable))
options(xtable.comment = FALSE)
ff <- read csv('F-F Research Data Factors.csv', skip=3)%>%
 filter(X1>=196001 & X1<=201512)
ind <- read_csv('48_Industry_Portfolios.csv', skip=11) %>%
 filter(X1>=196001 & X1<=201512) %>%
 slice(1:nrow(ff)) %>%
 mutate_if(is.numeric, list(~na_if(., "-99.99"))) %>%
 select_if(~!any(is.na(.))) %>%
 mutate(RF = ff$RF)
Ind = ind[2:44] - ind$RF
# Problem 1
cov.mat = cov(Ind) # variance-covariance matrix of the excess return
eigen.cov = eigen(cov.mat) # solving eigen problem
eigen.val = eigen.cov$values # eigen value
eigen.vec = eigen.cov$vectors # eigen vector
frac.var = eigen.val / sum(eigen.val) # fraction of variance explained
eigen.val
## [1] 1063.588563 93.701238 61.827898
                                            51.712939
                                                        36.598735
        32.659992 28.721750 25.630900 23.971119
## [6]
                                                        22.374542
        19.468473 18.634690 16.902934
## [11]
                                           16.655990
                                                        16.462992
## [16] 14.635944 14.081187 13.595840 12.905028 12.132211
## [21] 11.212621 10.432142 10.065567 9.291008 8.862178
## [26]
        8.399305
                    7.913556
                               7.687662 6.964257 6.926746
                               5.671311
## [31]
         6.540604
                     6.043709
                                           5.499267
                                                        4.895282
## [36]
          4.807527
                               4.422294
                                           4.067772 3.820311
                     4.615763
## [41]
          3.616439
                      3.478702
                                 3.077869
# Plotting the fractions
df = data.frame(matrix(ncol = 0, nrow = length(frac.var))) %>%
 mutate(PC = seq(1, length(frac.var), 1)) %>%
 mutate(frac = frac.var)
plot1 = ggplot(data = df, aes(x = PC, y = frac)) + geom_bar(stat = "identity")
plot1 = plot1 + ggtitle("Fraction of Variance Explained")
plot1
```

## Fraction of Variance Explained



## Problem 2

```
# Problem 2
# accumulative variances from PC1 to PC3
cum.frac.var = sum(frac.var[1:3])
cum.frac.var
## [1] 0.698805
# b
# Implementing PCA
pca = prcomp(Ind)
# Picking weights and calculating sample mean, stddev, and correlation
weight = pca$rotation
ind.mean = apply(Ind, 2, mean)
ret = as.matrix(Ind) %*% weight
sample.mean = apply(ret, 2, mean)
sample.stddev = sqrt(apply(ret, 2, var))
cor.mat = cor(ret)
cor.mat = cor.mat[1:3,1:3]
print(xtable(as.data.frame(sample.mean[1:3])))
print(xtable(as.data.frame(sample.stddev[1:3])))
```

	sample.mean[1:3]
PC1	3.77
PC2	0.21
PC3	-0.50

	sample.stddev[1:3]
PC1	32.61
PC2	9.68
PC3	7.86

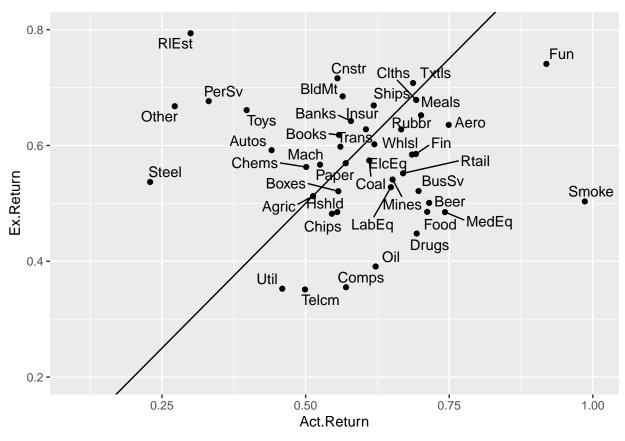
#### print(xtable(cor.mat, digit = 6))

	PC1	PC2	PC3
PC1	1.000000	0.000000	-0.000000
PC2	0.000000	1.000000	-0.000000
PC3	-0.000000	-0.000000	1.000000

```
# c
# Calculating E[Rt] and E[Ft]. Beta equals to weights
beta = weight[, 1:3]
E.Ft = sample.mean[1:3]
E.Rt = beta %*% E.Ft
act.ret = apply(Ind, 2, mean)

# Plotting the results
df2 = data.frame(matrix(ncol = 0, nrow = length(E.Rt))) %>%
    mutate(Industry = colnames(Ind)) %>%
    mutate(Ex.Return = E.Rt) %>%
    mutate(Act.Return = act.ret)

plot2 = ggplot(data = df2, aes(x = Act.Return, y = Ex.Return, label = Industry)) + geom_point()
plot2 = plot2 + geom_abline(intercept = 0) + xlim(c(0.1,1)) + ylim(c(0.2,0.8))
plot2 = plot2 + geom_text_repel()
plot2
```



```
# d
diff = act.ret - E.Rt
diff.va = diff - mean(diff)
test = act.ret - mean(act.ret)
var.d = var(diff)
var.act = var(df2$Act.Return)
Rsquare = 1 - var.d / var.act
Rsquare
```

```
## [1,] -0.5947519
```

### Problem 3

```
# a
# preparing the data
ff25 <- read_csv('25_Portfolios_5x5.CSV', skip=15)%>%
    filter(X1>=196001 & X1<=201512) %>%
    slice(1:nrow(ff))
FF25 = ff25[,2:26]
FF25 = FF25 - ind$RF

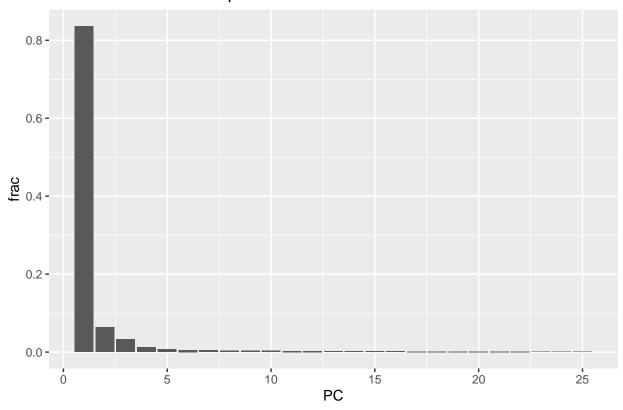
# getting the fraction of variance in the similar way of problem 1
cov.matff = cov(FF25)
eigen.covff = eigen(cov.matff)
eigen.valff = eigen.covff$values
```

```
eigen.vecff = eigen.covff$vectors
frac.varff = eigen.valff / sum(eigen.valff)

# Organizing the data
df3 = data.frame(matrix(ncol = 0, nrow = length(frac.varff))) %>%
    mutate(PC = seq(1, length(frac.varff), 1)) %>%
    mutate(frac = frac.varff)

# plotting the fraction of variance explained
plot3 = ggplot(data = df3, aes(x = PC, y = frac)) + geom_bar(stat = "identity")
plot3 = plot3 + ggtitle("Fraction of Variance Explained")
plot3
```

## Fraction of Variance Explained



By the PC5, the accumulative fraction of variance explained reaches 95%.

```
# b
# accumulative variances
cum.frac.varff = cumsum(frac.varff[1:25])
cum.frac.varff

## [1] 0.8362506 0.8998043 0.9325635 0.9450868 0.9522204 0.9577525 0.9627125
## [8] 0.9664420 0.9700398 0.9733405 0.9762290 0.9788752 0.9812569 0.9835512
## [15] 0.9855775 0.9874736 0.9892079 0.9908919 0.9924712 0.9939822 0.9953437
## [22] 0.9966451 0.9979014 0.9989749 1.0000000
```