

Assignment 2

Mgmt 237E: Empirical Methods

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Problem 1

Lets load the data into R

```
options(warn = -1)
suppressMessages(library(lubridate))
suppressMessages(library(xts))
options(warn = 0)

# Retrieve Data
portfolio.data <- read.csv("49_Industry_Portfolios.csv",header=TRUE,sep = ",",stringsAsFactors = FALSE,
portfolio.data$X <- as.Date(paste(floor(as.numeric(portfolio.data$X)/100),as.numeric(portfolio.data$X) %
portfolio.data <- xts(portfolio.data[,-1],order.by = portfolio.data$X)

datafactors <- read.csv("F-F_Research_Data_Factors.csv",header=TRUE,sep = ",",stringsAsFactors = FALSE,
datafactors$X <- as.Date(paste(floor(as.numeric(datafactors$X)/100),as.numeric(datafactors$X) %% 100,"0
datafactors <- xts(datafactors[, -1],order.by = datafactors$X)

#Constraints
#Date
portfolio.data <- portfolio.data[index(portfolio.data)>="1960-01-01" & index(portfolio.data)<="2015-12-31",]
datafactors <- datafactors[index(datafactors)>="1960-01-01" & index(datafactors)<="2015-12-31",]

#Remove columns with -99.99 value
contains99 <- logical(dim(portfolio.data)[2])
for(col in 1:dim(portfolio.data)[2]){
  contains99[col] <- sum(portfolio.data[,col] %in% -99.99) > 0
}
portfolio.data <- portfolio.data[,!contains99]

#Excess Returns
excessReturns <- function(industryReturns){
  industryReturns - datafactors$RF
}

excess.returns <- matrix(nrow=dim(portfolio.data)[1],ncol=dim(portfolio.data)[2])
for(col in 1:dim(portfolio.data)[2]){
  excess.returns[,col] <- portfolio.data[,col] - datafactors$RF
}
colnames(excess.returns) <- colnames(portfolio.data)
```

For each industry, regress the excess return on market excess return

```
options(warn = -1)
suppressWarnings(library(DataAnalytics))
options(warn = 0)
regressionData <- function(excess.industryReturns){
  reg <- lm(excess.industryReturns~datafactors$Mkt.RF[index(excess.industryReturns)])
}
```

```

dummy <- capture.output(summ <- lmSumm(reg,HAC = TRUE))
result <- c(summ$coef.table[2,c(1,2)],summ$coef.table[1,c(1,2)],summary(summ$lmfit)$r.squared)
t(result)
}

excess.returns <- xts(excess.returns,order.by = as.Date(index(portfolio.data)))
regressionInfo <- apply(excess.returns, 2, regressionData)

```

a

Lets plot the betas of various industry

```

betaTable <- regressionInfo[c(1,2),]
betaBars <- barplot(height = betaTable[1,],
                    names.arg = colnames(betaTable),
                    beside = true, las = 2,
                    ylim = c(0, 1.75),
                    cex.names = 0.75, xaxt = "n",
                    main = "Plot of beta for various industries",
                    ylab = "Beta",
                    border = "black", axes = TRUE,width=35,space = 0.8)

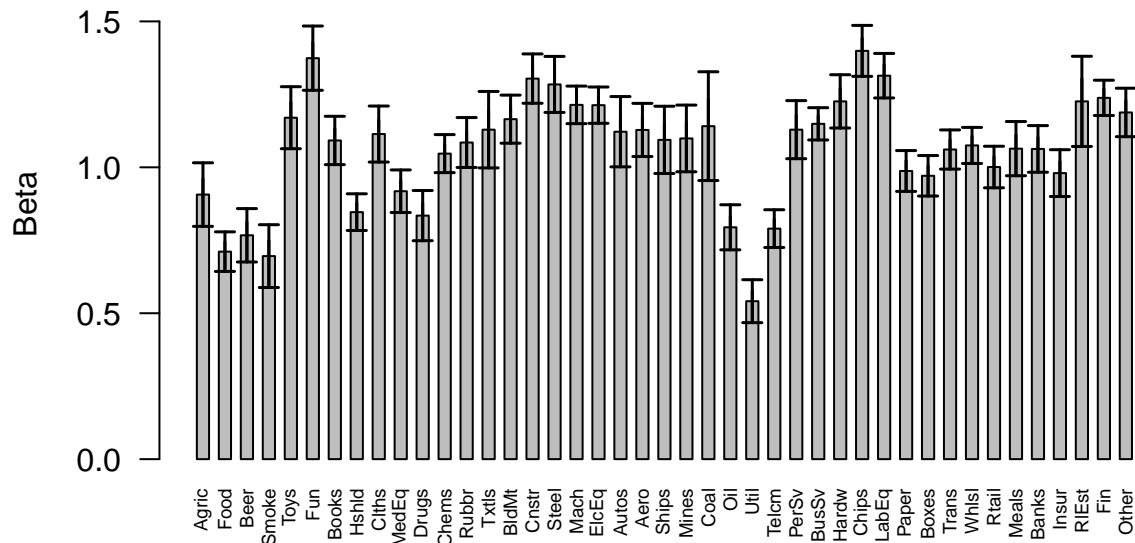
text(x = betaBars, y = -0.1, srt = 90,
     adj = 1, labels = colnames(betaTable), xpd = TRUE,cex=0.6)

segments(betaBars, betaTable[1,]- betaTable[2,] * 2, betaBars,
         betaTable[1,] + betaTable[2,], lwd = 1.5)

arrows(betaBars, betaTable[1,]- betaTable[2,] * 2, betaBars,
       betaTable[1,] + betaTable[2,]*2, lwd = 1.5, angle = 90,
       code = 3, length = 0.05)

```

Plot of beta for various industries



b

The range of the betas can be found by using the *range* function

```
range(betaTable[1,])
```

```
## [1] 0.5413 1.3990
```

The max, min and mean of the R^2 across industries is as below

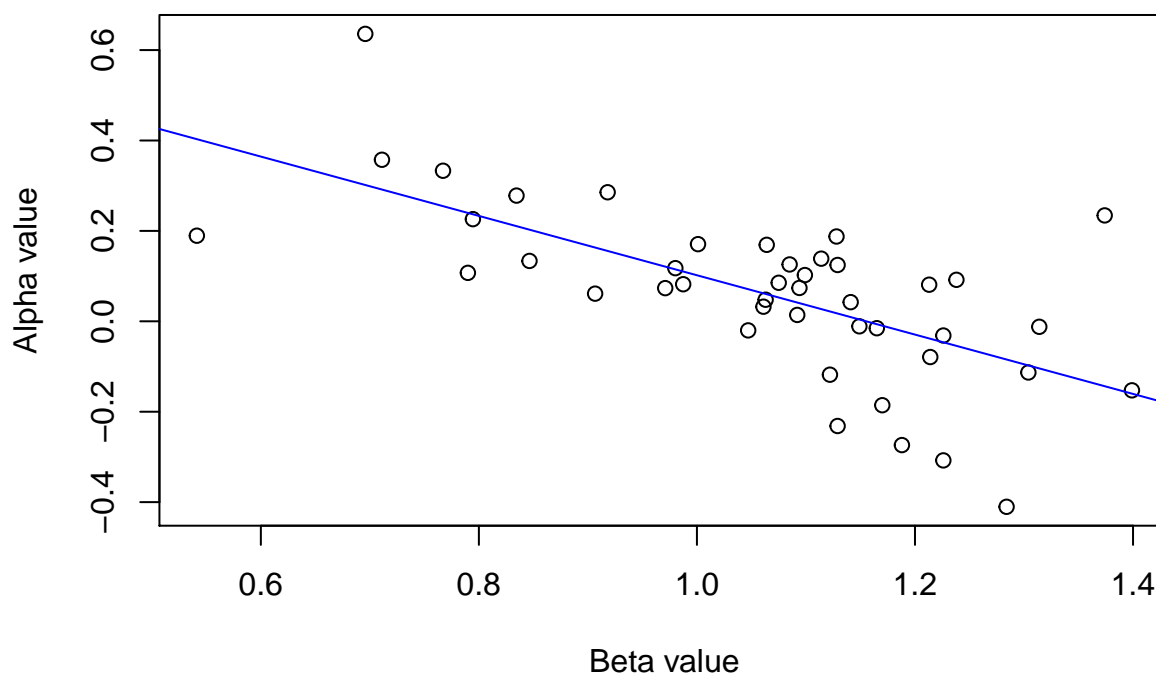
```
resultsR2 <- c()
resultsR2[1] <- min(regressionInfo[5,])
resultsR2[2] <- max(regressionInfo[5,])
resultsR2[3] <- mean(regressionInfo[5,])
names(resultsR2) <- c("Min", "Max", "Mean")
resultsR2
```

```
##      Min      Max      Mean
## 0.2505057 0.8055172 0.5776919
```

c

```
plot(regressionInfo[1,], regressionInfo[3,], xlab="Beta value", ylab="Alpha value", main = "data and regression plot of alpha vs beta of all industries")
alphalm <- lm(regressionInfo[3,]~regressionInfo[1,])
abline(alphalm$coefficients[1], alphalm$coefficients[2], col = "blue")
```

data and regression plot of alpha vs beta of all industries



We can see that for lower beta, we mostly get higher alpha.

In CAPM, we expect low beta to have low risk, which means low returns. But in this case, there is high alpha. So there is a chance that the total returns is higher for low beta stocks. This goes against CAPM.

Problem 2

```
acfBeta <- function(indsustryBeta){
  acf_Val <- acf(indsustryBeta, plot = FALSE)
  acf_Val$acf[2]
}

fiveYearBeta <- matrix(nrow=((2010-1960)/5)+1, ncol=dim(excess.returns)[2])
fiveYearBeta.Error <- matrix(nrow=((2010-1960)/5)+1, ncol=dim(excess.returns)[2])
fiveyears.Endpoints <- seq(1960, 2010, by=5)
for(fiveYearStart in fiveyears.Endpoints){
  excess.5year.returns <- excess.returns[paste0(fiveYearStart, "/", fiveYearStart+4)]
  results <- sapply(excess.5year.returns, regressionData)
  fiveYearBeta[((fiveYearStart-1960)/5)+1,] <- results[1,]
```

```

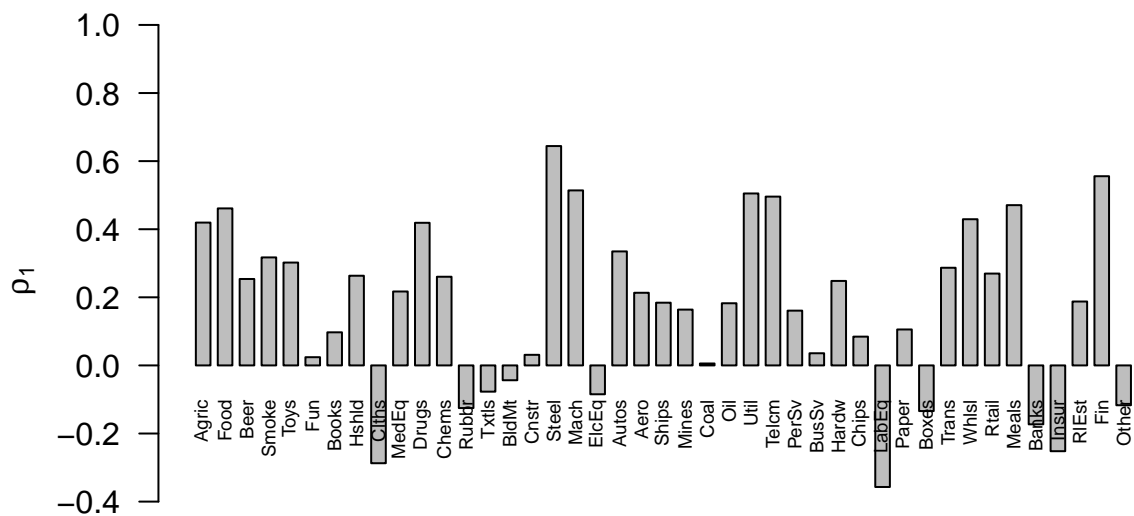
fiveYearBeta.Error[((fiveYearStart-1960)/5)+1,] <- results[2,]
}

fiveYearBeta <- xts(fiveYearBeta,order.by= as.Date(paste(fiveyears.Endpoints,"01","01",sep = "-")))
fiveYearBeta.Error <- xts(fiveYearBeta.Error,order.by= as.Date(paste(fiveyears.Endpoints,"01","01",sep = "-")))
colnames(fiveYearBeta) <- colnames(excess.returns)
colnames(fiveYearBeta.Error) <- colnames(excess.returns)

industry.acf <- apply(fiveYearBeta,2,acfBeta)
beta.5yearsCor <- barplot(height = industry.acf,
  names.arg = colnames(industry.acf),
  beside = true, las = 2,
  ylim = c(-0.5, 1),
  cex.names = 0.75, xaxt = "n",
  main = expression(paste("Auto correlation of one lag for industry ",beta[s])),
  ylab = expression(rho[1]),
  border = "black", axes = TRUE,width=35,space = 0.5)
text(x = beta.5yearsCor, y = -0.1, srt = 90,
  adj = 1, labels =colnames(excess.returns), xpd = TRUE,cex=0.6)

```

Auto correlation of one lag for industry β_s



Some of the industries are almost 0 correlation (like Coal, Cnstr, BusSv). For these industries, the Beta will be stable.

Some of the industries, have very high auto correlation (like Stell, Finance).

```
fiveYearBeta.Error[, "Steel"]
```

```
##           Steel
## 1960-01-01 0.09385
## 1965-01-01 0.10940
## 1970-01-01 0.12110
## 1975-01-01 0.12610
## 1980-01-01 0.11890
## 1985-01-01 0.20130
## 1990-01-01 0.11480
## 1995-01-01 0.13920
## 2000-01-01 0.14990
## 2005-01-01 0.17600
## 2010-01-01 0.14080
```