

# Math23C Spring 2018 Final Project

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```
library(xts)
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      as.Date, as.Date.numeric
```

```
library(quantmod)
```

```
## Loading required package: TTR
```

```
## Version 0.4-0 included new data defaults. See ?getSymbols.
```

```
library(ggplot2)
```

## Abstract

Cryptocurrencies such as Bitcoin, Ethereum, etc. generated significant attention in 2017. Cryptocurrencies have significant volatility as there is rampant speculation. Given the high variance in prices, can data science methods explored in this class be used to model the market dynamics?

## Data Source

We obtained a history cryptocurrency price data from <https://www.kaggle.com/sudalairajkumar/cryptocurrencypricehistory/data> for four different cryptocurrencies (BTC, ETH, XMR, XRP)

(REQ: a dataframe, at least two numeric columns, at least 20 rows)

## Data Ingestion, clean up and normalization

```
# Read one price history file per currency
```

```
BTCdf = read.csv("data/bitcoin_price.csv", stringsAsFactors = F)
```

```
ETHdf = read.csv("data/ethereum_price.csv", stringsAsFactors = F)
```

```
XMRdf = read.csv("data/monero_price.csv", stringsAsFactors = F)
```

```
XRPdf = read.csv("data/ripple_price.csv", stringsAsFactors = F)
```

```
# Fix rest of data:
```

```
# 1- Make dates native format
```

```
# 2- Convert Volume and market cap:
```

```
#     a) From string ("123,456") to numeric (123456).
```

```
#     b) Convert "-" to 0.
```

```
# 3- Sort chronologically
```

```
fixVolCap = function(df) {
```

```

df$Date = as.Date(df$Date,"%b %d, %Y")
df$Volume = as.numeric(gsub("-", "0", gsub(",","",df$Volume)))
df$Market.Cap = as.numeric(gsub("-", "0", gsub(",","",df$Market.Cap)))
return (df[order(df$Date),])
}
BTCdf = fixVolCap(BTCdf)
ETHdf = fixVolCap(ETHdf)
XMRdf = fixVolCap(XMRdf)
XRPdf = fixVolCap(XRPdf)

# Ensure that all data start from the same date
earliestCommonDate = max(min(BTCdf$Date),
                           min(ETHdf$Date),
                           min(XMRdf$Date),
                           min(XRPdf$Date))
BTCdf = BTCdf[BTCdf$Date>=earliestCommonDate,]
ETHdf = ETHdf[ETHdf$Date>=earliestCommonDate,]
XMRdf = XMRdf[XMRdf$Date>=earliestCommonDate,]
XRPdf = XRPdf[XRPdf$Date>=earliestCommonDate,]

```

## Exploratory Data Analysis

```

plotSeries = function(df){
  dfdata = xts(df[,2:7],order.by = df[,1])
  # TODO fix weekly plot
  #wk = dfdata
  #data.wk = to.weekly(wk)
  #plot(data.wk)
  #plot.new()
  OHLC = as.quantmod.OHLC(dfdata)
  chartSeries(OHLC)
}
plotSeries(BTCdf)

```



`plotSeries(ETHdf)`



`plotSeries(XMRdf)`



`plotSeries(XRPdf)`

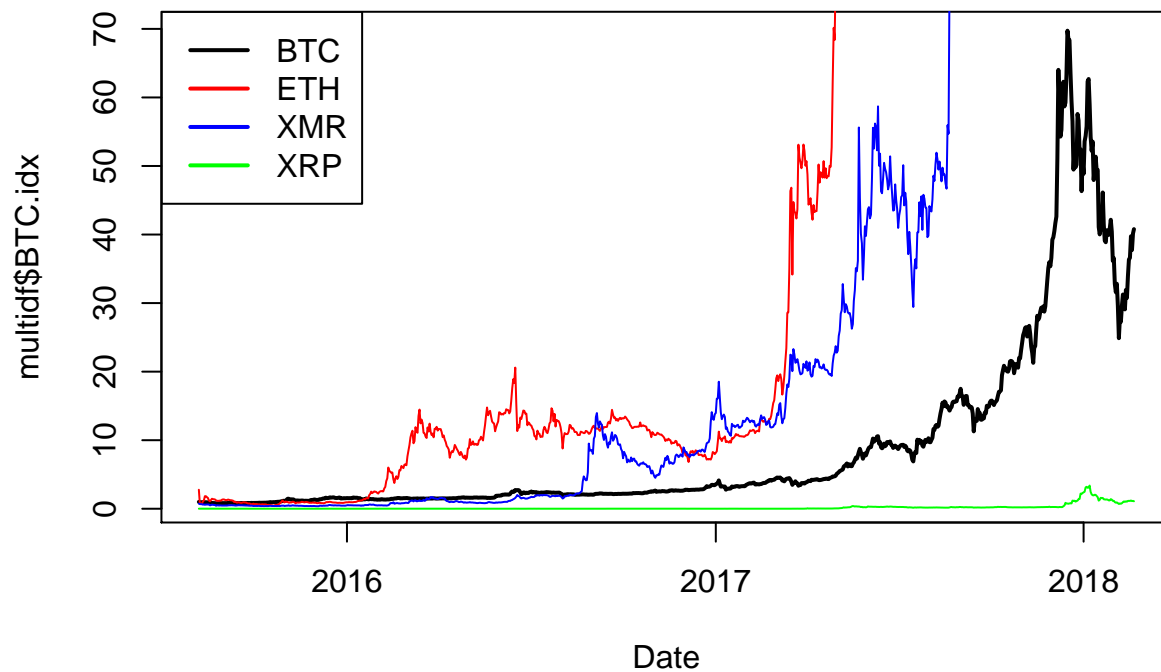


```
BTCdata = xts(BTCdf[,2:7],order.by = BTCdf[,1])
# wk = BTCdata
# data.wk = to.weekly(wk)
# plot(data.wk)
# OHLC = as.quantmod.OHLC(BTCdata)
# chartSeries(OHLC)
```

```
close.prices = BTCdata$Close
close.prices = cbind(close.prices,ETHdf$Close,XMRdf$Close,XRPdf$Close)
```

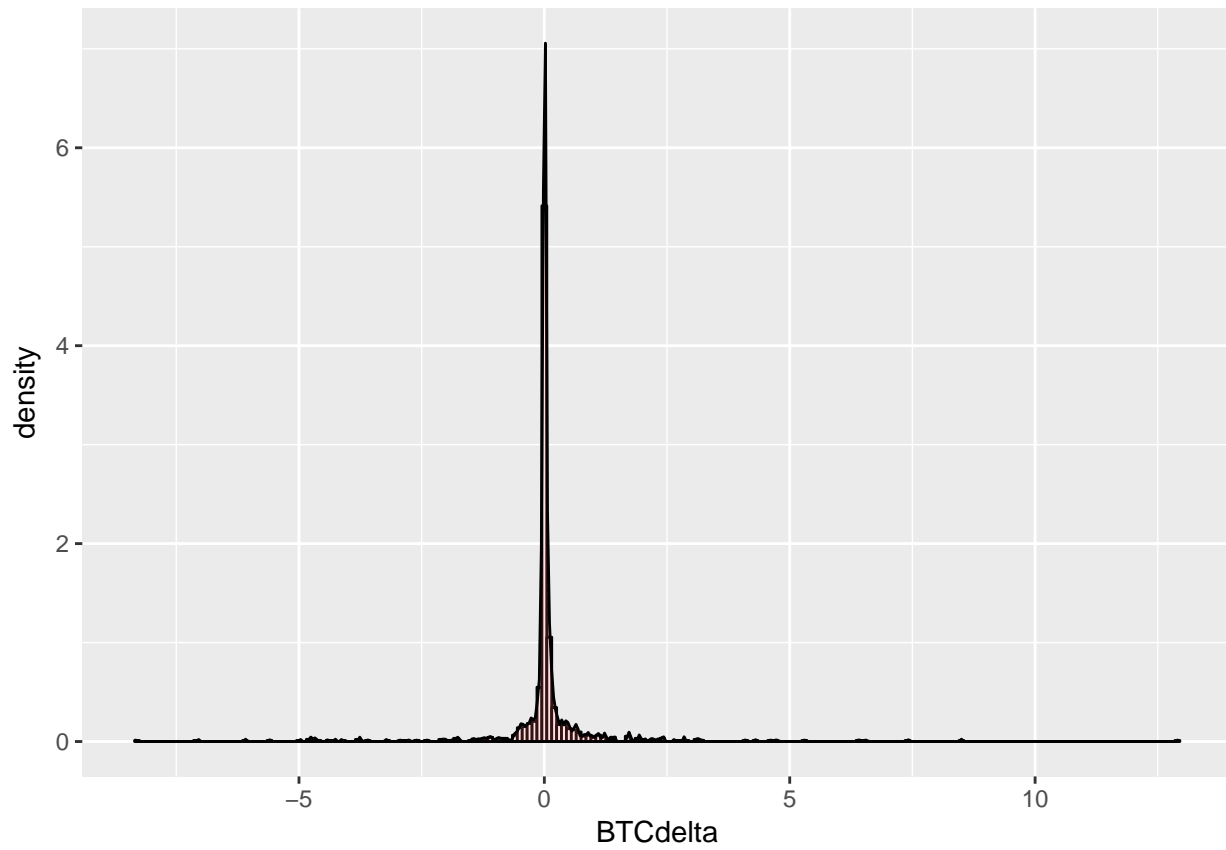
```
multidf = cbind(index(close.prices), data.frame(close.prices))
names(multidf) = paste(c("Date","BTC","ETH","XMR","XRP"))
multidf$BTC.idx = multidf$BTC / multidf$BTC[1]
multidf$ETH.idx = multidf$ETH / multidf$ETH[1]
multidf$XMR.idx = multidf$XMR / multidf$XMR[1]
multidf$XRP.idx = multidf$XRP / multidf$XRP[1]
```

```
plot(x = multidf$Date,y=multidf$BTC.idx,type="l",xlab="Date",col="black",lty=1,lwd=2)
lines(x=multidf$Date,y=multidf$ETH,col="red")
lines(x=multidf$Date,y=multidf$XMR,col="blue")
lines(x=multidf$Date,y=multidf$XRP,col="green")
legend("topleft",c("BTC","ETH","XMR","XRP"),col=c("black","red","blue","green"),
      lty=c(1,1,1,1),
      lwd=c(2,2,2,2))
```



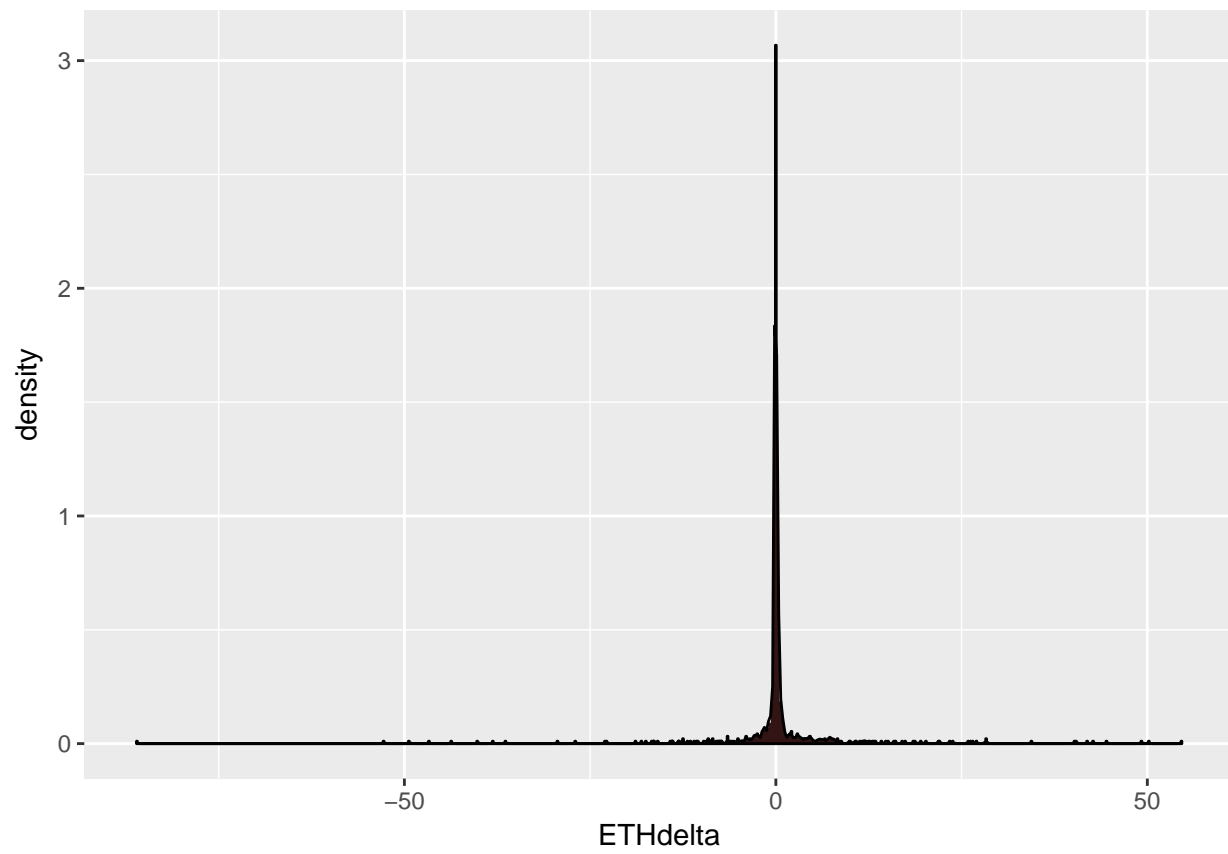
```
# distribution of relative price changes
multidf$BTCdelta = c(0,diff(multidf$BTC.idx))
multidf$ETHdelta = c(0,diff(multidf$ETH.idx))
multidf$XMRdelta = c(0,diff(multidf$XMR.idx))
multidf$XRDelta = c(0,diff(multidf$XRP.idx))
```

```
# REQ: display a histogram
ggplot(multidf, aes(x=BTCdelta)) +
  geom_histogram(aes(y=..density..),
    binwidth=0.1,colour="black",
    fill="white") +
  geom_density(alpha=0.2,fill="#FF6666")
```

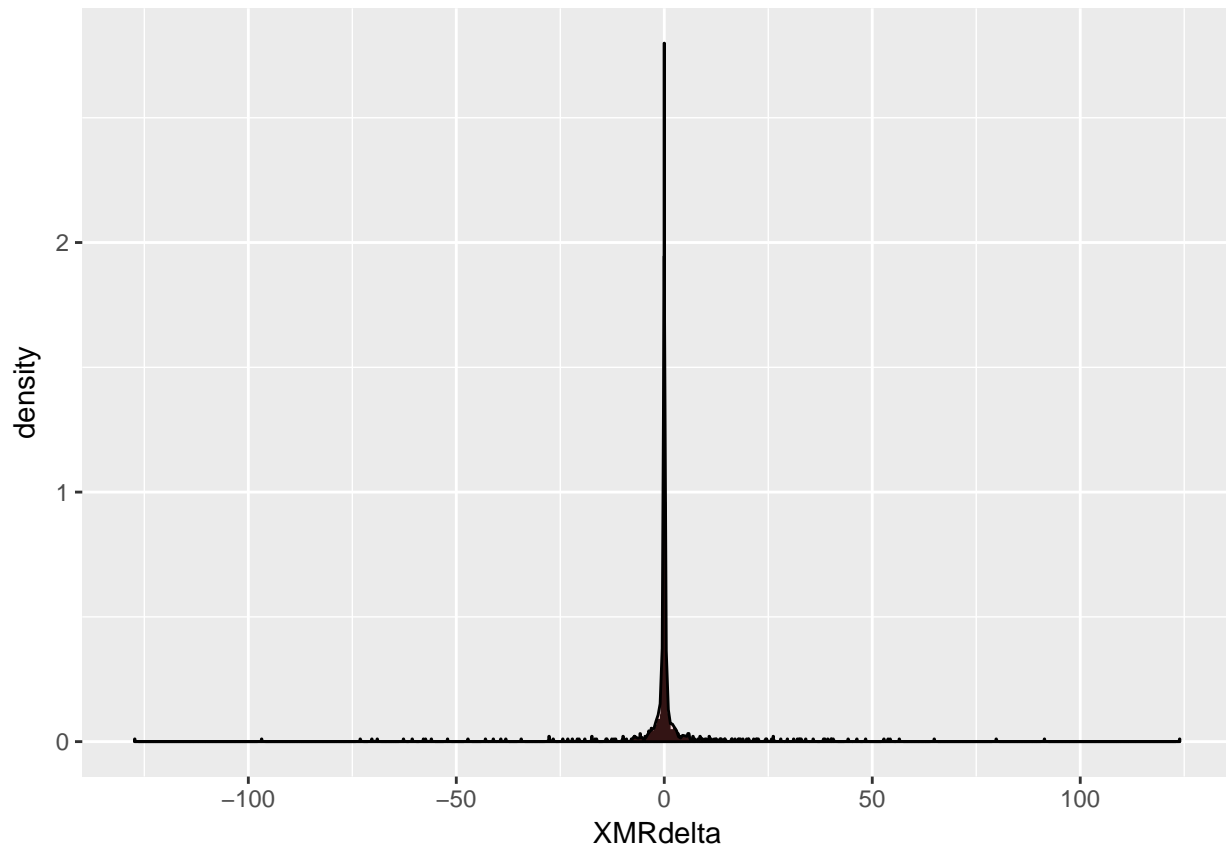


```
ggplot(multidf, aes(x=ETHdelta)) +  
  geom_histogram(aes(y=..density..),  
    binwidth=0.1, colour="black",  
    fill="white") +  
  geom_density(alpha=0.2, fill="#FF6666")
```

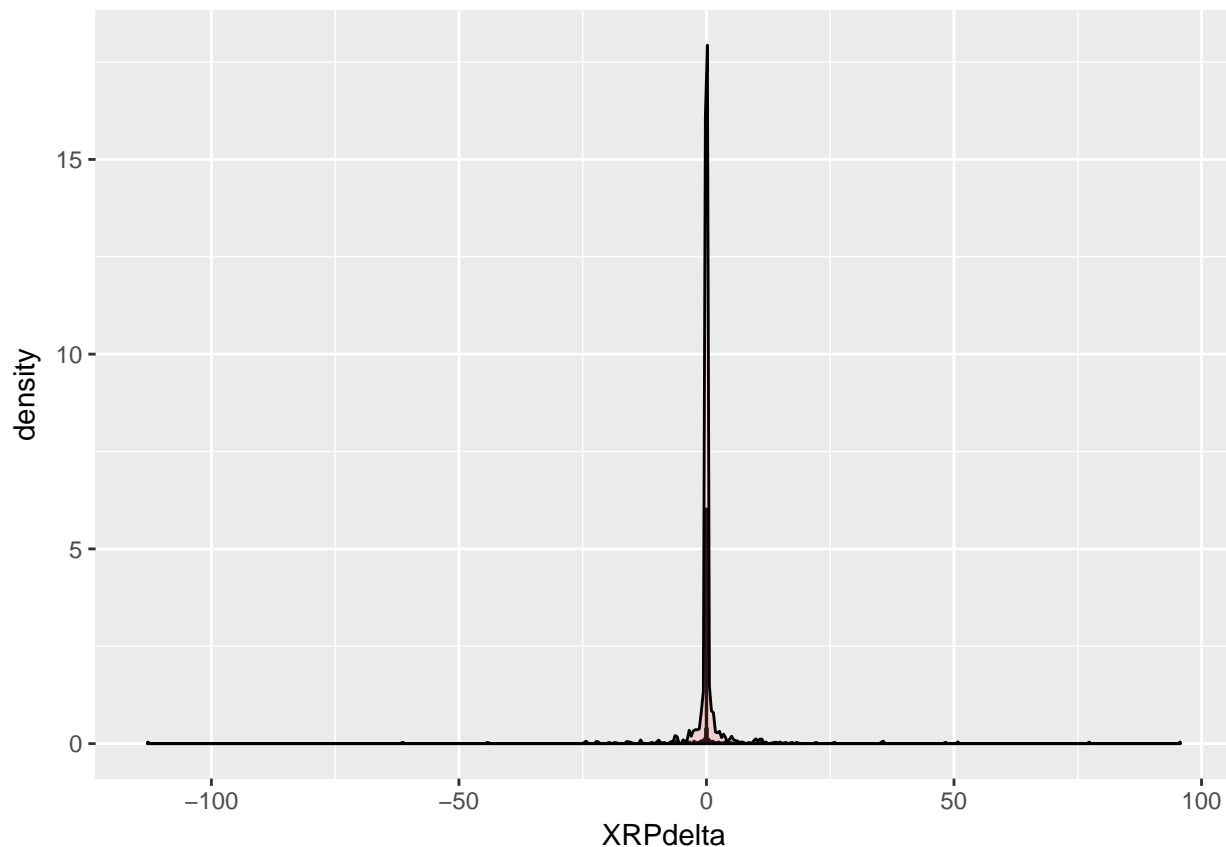




```
ggplot(multidf, aes(x=XMRdelta)) +  
  geom_histogram(aes(y=..density..),  
    binwidth=0.1, colour="black",  
    fill="white") +  
  geom_density(alpha=0.2, fill="#FF6666")
```



```
ggplot(multidf, aes(x=XRPdelta)) +  
  geom_histogram(aes(y=..density..),  
    binwidth=0.1, colour="black",  
    fill="white") +  
  geom_density(alpha=0.2, fill="#FF6666")
```



```
# Display a bar plot

overallReturn = function(df){
  return ((df$Close[nrow(df)] - df$Close[1]) / df$Close[1])
}

volIncrease = function(df){
  return ((df$Volume[nrow(df)] - df$Volume[1]) / df$Volume[1])
}

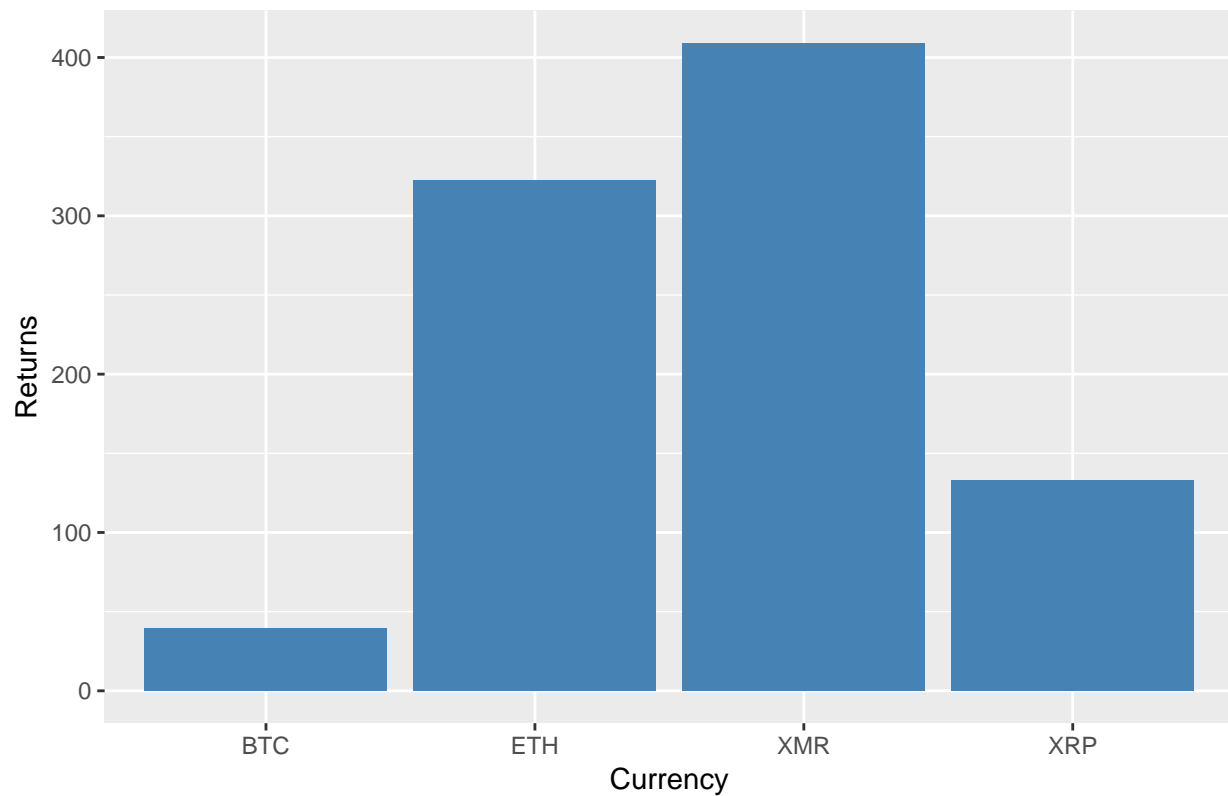
returns = c(overallReturn(BTCdf), overallReturn(ETHdf), overallReturn(XMRdf), overallReturn(XRPdf))
volumes = c(volIncrease(BTCdf), volIncrease(ETHdf), volIncrease(XMRdf), volIncrease(XRPdf))
close.prices = BTCdata$Close
close.prices = cbind(close.prices, ETHdf$Close, XMRdf$Close, XRPdf$Close)

barData = data.frame(Currency = c("BTC", "ETH", "XMR", "XRP"), Returns=returns, Volumes=volumes)
barData

##   Currency   Returns   Volumes
## 1      BTC  39.78868  232.6492
## 2      ETH 322.23827 15487.8060
## 3      XMR 409.08121  551.9270
## 4      XRP 132.70952  2217.3103

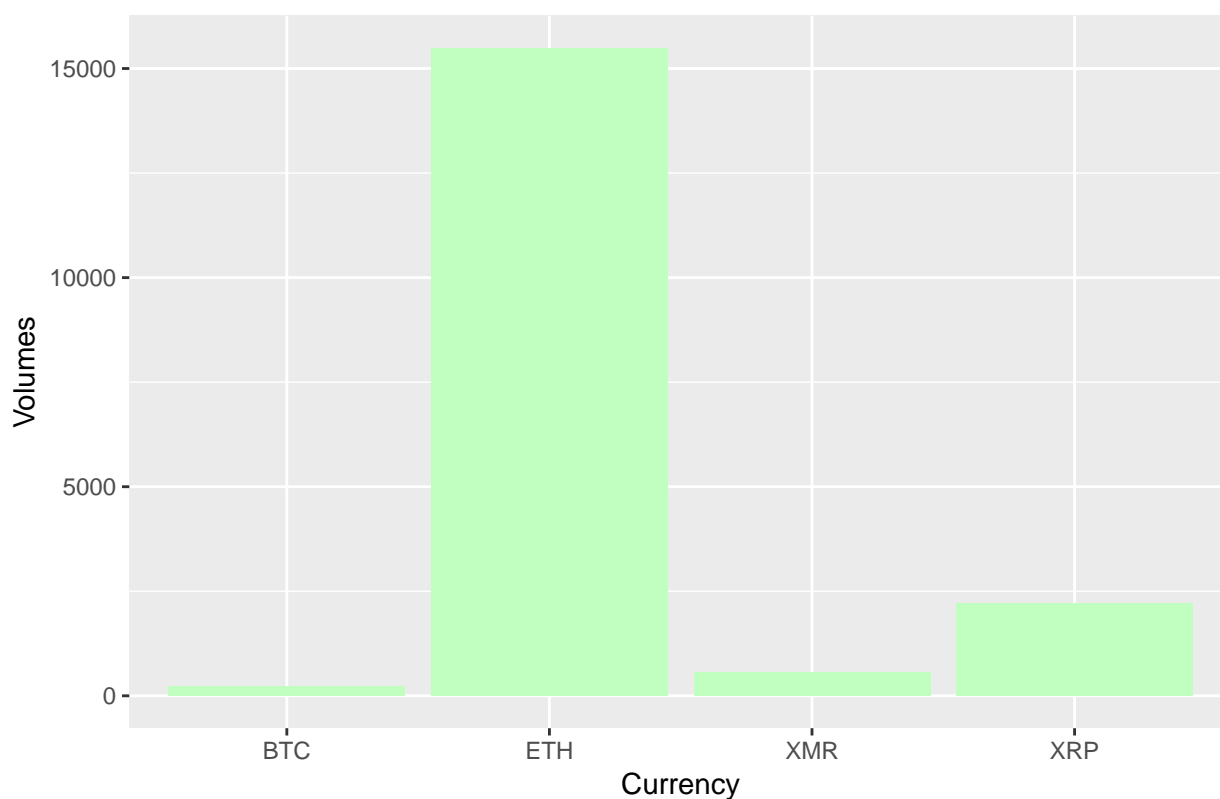
# REQ: barplot
ggplot(data=barData, aes(x=Currency, y>Returns)) +
  geom_bar(stat="identity", fill="steelblue") +
  ggtitle("Cumulative Return on Investment (not percent!) Aug 2015 - Feb 2018")
```

Cumulative Return on Investment (not percent!) Aug 2015 – Feb 2018



```
ggplot(data=barData, aes(x=Currency, y=Volumes)) +  
  geom_bar(stat="identity",fill="darkseagreen1") +  
  ggtitle("Change in Daily Volume Aug 2015 - Feb 2018")
```

Change in Daily Volume Aug 2015 – Feb 2018



“ ## Topic 1 - Volatility Exploration

**Topic 2 -**

**Topic 3 - Correlation between Cryptocurrencies**

```
cor(BTCdf$Close, ETHdf$Close)
```

```
## [1] 0.9060949
```

```
cor(BTCdf$Close, XMRdf$Close)
```

```
## [1] 0.9691732
```

```
cor(BTCdf$Close, XRPdf$Close)
```

```
## [1] 0.8049156
```

```
cor(ETHdf$Close, XMRdf$Close)
```

```
## [1] 0.9525516
```

```
cor(ETHdf$Close, XRPdf$Close)
```

```
## [1] 0.8798746
```

```
cor(XMRdf$Close, XRPdf$Close)
```

```
## [1] 0.8847865
```

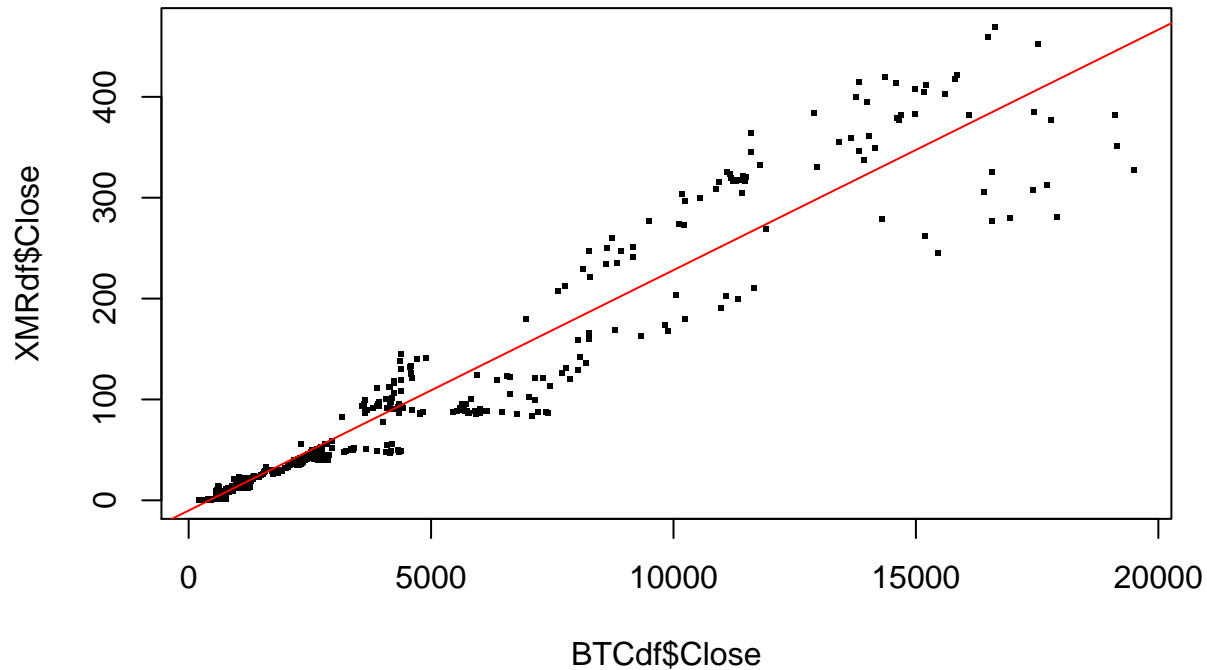
```

#Largest correlation between BTC and XMR
plot(BTCdf$Close,XMRdf$Close,pch = ".",cex = 3)
#b is slope
b <- cov(BTCdf$Close,XMRdf$Close)/var(BTCdf$Close)
#a is intercept
a <- mean(XMRdf$Close) - b*mean(BTCdf$Close);a

## [1] -10.16858

#We can add this regression line to the plot of the data
abline(a, b, col = "red")

```



##

Summary of Project Requirements

#### Required dataset standards

- [x] A dataframe
- [ ] At least two categorical or logical columns
- [x] At least two numeric columns
- [x] At least 20 rows, preferably more, but real-world data may be limited

#### Required graphical displays (all graphs must be colored and nicely labeled)

- [x] A barplot
- [x] A histogram
- [ ] A probability density graph overlaid on a histogram
- [ ] A contingency table

#### Required Analysis

- [ ] A permutation test
- [ ] A p-value or other statistic based on a distribution function

- [ ] Analysis of a contingency table
- [ ] Comparison of analysis by classical methods (chi-square, CLT) and simulation methods

### **Required submission uploads**

- [ ] A .csv with the dataset
- [ ] A long, well-commented script that loads the dataset, explores it, and does all the analysis.
- [ ] A shorter .Rmd with compiled .pdf or .html file that presents highlights in ten minutes.
- [ ] A one-page handout that explains the dataset and summarizes the analysis.

### **Additional points for creativity or complexity (up to 10 points)**

### **References**

Clifford S. Ang, *Analyzing Financial Data and Implementing Financial Models Using R*, Springer, 2015

Berlinger et al. *Mastering R for Quantitative Finance*, Packt Publishing, 2015