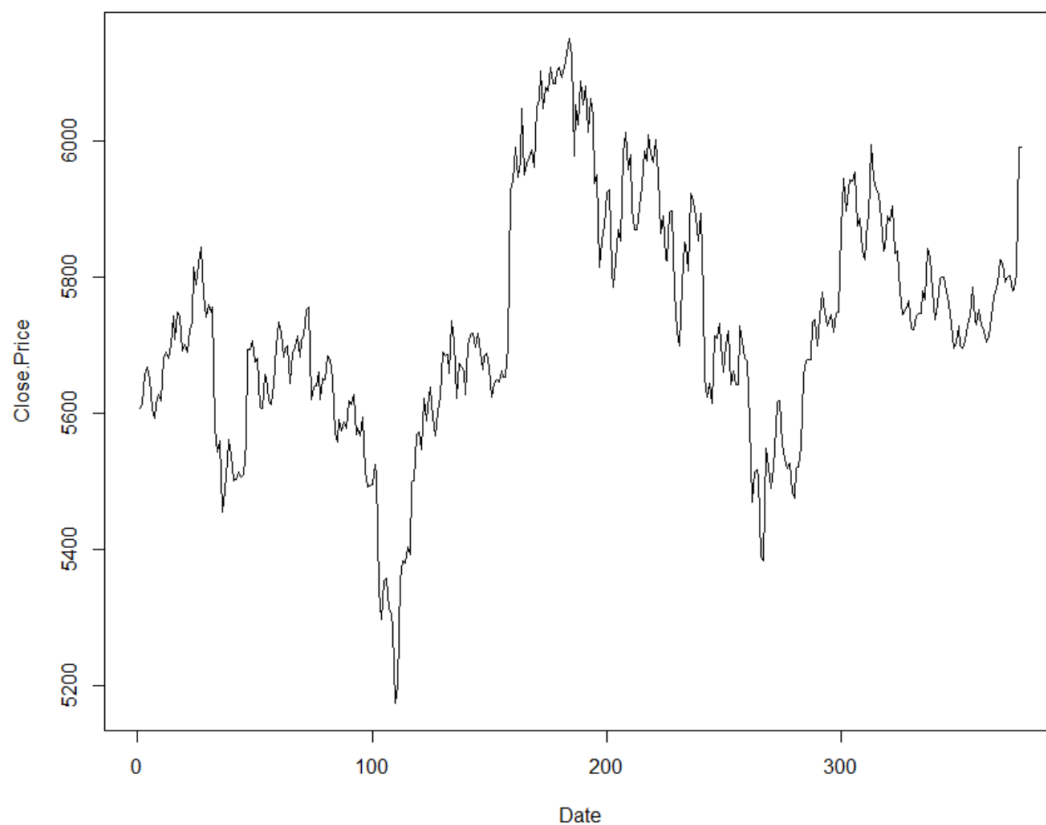
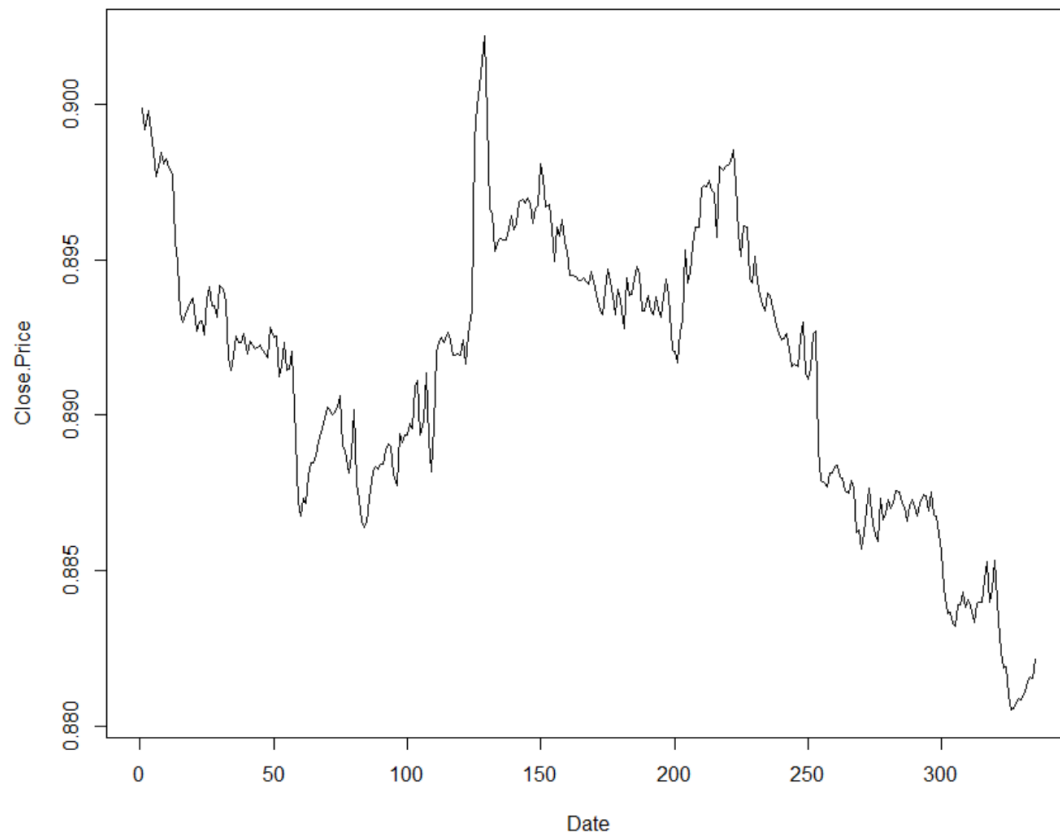


## Exercise 7 - Multi-Model Comparison of Cryptocurrency vs Forex (in R)

November 22, 2017

*Note: in all predictions below the same FTS data is used (400h EUR/GBP October 2017 & 400h BTC/USD October 2017)*



## 1. AR Yule-Walker model predictions

Forex (EUR/GBP) accuracy:

1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 19. 20.  
0.52 0.52 0.51 0.51 0.51 0.53 0.51 0.50 0.50 0.49 0.49 0.50 0.48 0.49 0.48 0.49 0.48 0.49 0.50 0.50,  $\mu = 0.50$ .  
(234 predictions)

Cryptocurrency (BTC/USD) accuracy:

1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 19. 20.  
0.50 0.50 0.53 0.53 0.54 0.54 0.54 0.55 0.54 0.54 0.54 0.55 0.54 0.53 0.53 0.53 0.53 0.54 0.54 0.55,  $\mu = 0.53$ .  
(276 predictions)

Script used:

#Author: Thomas Hollis

#Subject: Bachelor Thesis

#1. Data Import

```
data <- read.csv("R/BTC.csv")
```

#2. Data Split

```
data_train <- data[1:100,2]
```

```
data_test <- data[101:377,2]
```

#3. Model Train & apply Model

```
performance <- 0
```

```
for (j in 1:20)
```

```
{
```

```
  data_train <- data[1:100,2]
```

```
  c_count <- 0
```

```
  for (i in 1:277)
```

```
  {
```

```
    btc_ar <- ar.yw(data_train, order.max = j, aic = FALSE)
```

```
    btc_pred <- predict(object = btc_ar, n.ahead = 1)
```

```
    if(i > 1)
```

```
    {
```

```
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||
```

```
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))
```

```
      {
```

```
        c_count = c_count+1
```

```
      }
```

```
    }
```

```
    data_train <- c(data_train, data_test[i])
```

```
  }
```

```
  performance[j] <- c_count/276.00
```

```
}
```

```
round(performance, digits = 2)
```

```
round(mean(performance), digits =2)
```

## 2. AR Burg model predictions

Forex (EUR/GBP) accuracy:

1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 19. 20.  
0.52 0.52 0.53 0.52 0.52 0.53 0.51 0.50 0.50 0.50 0.50 0.49 0.48 0.52 0.47 0.47 0.48 0.47 0.47 0.47,  $\mu = 0.50$   
(234 predictions)

Cryptocurrency (BTC/USD) accuracy:

1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 19. 20.  
0.50 0.50 0.54 0.54 0.54 0.52 0.53 0.51 0.53 0.51 0.52 0.54 0.54 0.53 0.53 0.51 0.50 0.54 0.54 0.53,  $\mu = 0.52$ .  
(276 predictions)

Script used:

#Author: Thomas Hollis

#Subject: Bachelor Thesis

### #1. Data Import

```
data <- read.csv("R/BTC.csv")
```

### #2. Data Split

```
data_train <- data[1:100,2]  
data_test <- data[101:377,2]
```

### #3. Model Train & apply Model

```
performance <- 0  
for (j in 1:20)  
{  
  data_train <- data[1:100,2]  
  c_count <- 0  
  for (i in 1:277)  
  {  
    btc_ar <- ar.burg(data_train, order.max = j, aic = FALSE)  
    btc_pred <- predict(object = btc_ar, n.ahead = 1)  
  
    if(i > 1)  
    {  
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||  
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))  
      {  
        c_count = c_count+1  
      }  
    }  
  
    data_train <- c(data_train, data_test[i])  
  }  
  performance[j] <- c_count/276.00  
}  
round(performance, digits = 2)  
round(mean(performance), digits =2)
```

### 3. ARMA model predictions

Forex (EUR/GBP) accuracy:

|       |        |        |        |        |        |        |        |       |        |        |        |        |        |       |       |        |        |        |       |
|-------|--------|--------|--------|--------|--------|--------|--------|-------|--------|--------|--------|--------|--------|-------|-------|--------|--------|--------|-------|
| 404.0 | 204.00 | 502.00 | 306.00 | 103.00 | 501.00 | 606.00 | 103.00 | 206.0 | 503.00 | 405.00 | 405.00 | 405.00 | 504.00 | 404.0 | 402.0 | 403.00 | 603.00 | 505.00 | 302.0 |
| 0.5   | 0.53   | 0.52   | 0.49   | 0.52   | 0.55   | 0.52   | 0.52   | 0.5   | 0.54   | 0.48   | 0.48   | 0.48   | 0.49   | 0.5   | 0.5   | 0.51   | 0.55   | 0.49   | 0.5   |

$\mu = 0.51$

(234 predictions)

Cryptocurrency (BTC/USD) accuracy:

|        |        |        |        |        |        |       |        |        |        |        |        |        |        |        |        |       |        |        |        |
|--------|--------|--------|--------|--------|--------|-------|--------|--------|--------|--------|--------|--------|--------|--------|--------|-------|--------|--------|--------|
| 401.00 | 503.00 | 505.00 | 404.00 | 205.00 | 505.00 | 202.0 | 405.00 | 103.00 | 501.00 | 103.00 | 305.00 | 303.00 | 304.00 | 503.00 | 302.00 | 406.0 | 303.00 | 302.00 | 502.00 |
| 0.53   | 0.48   | 0.54   | 0.51   | 0.56   | 0.54   | 0.5   | 0.51   | 0.53   | 0.51   | 0.53   | 0.54   | 0.51   | 0.53   | 0.48   | 0.51   | 0.5   | 0.51   | 0.51   | 0.48   |

$\mu = 0.52$

(276 predictions)

Script used:

#Author: Thomas Hollis

#Subject: Bachelor Thesis

#1. Data Import

```
data <- read.csv("R/BTC.csv")
```

#2. Data Split

```
data_train <- data[1:100,2]
```

```
data_test <- data[101:377,2]
```

#3. Model Train & apply Model

```
performance <- matrix(0,2,20)
```

```
for (j in 1:20)
```

```
{
```

```
  data_train <- data[1:100,2]
```

```
  c_count <- 0
```

```
  p <- round(runif(1, 1, 6))
```

```
  d <- 0
```

```
  q <- round(runif(1, 1, 6))
```

```
  for (i in 1:277)
```

```
  {
```

```
    btc_arma <- arima(data_train, order = c(p,d,q), method = "CSS")
```

```
    btc_pred <- predict(btc_arma, 1)
```

```
    if(i >= 2)
```

```
    {
```

```
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||
```

```
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))
```

```
      {
```

```
        c_count = c_count+1
```

```
      }
```

```
    }
```

```
    data_train <- c(data_train, data_test[i])
```

```
  }
```

```
  performance[1,j] <- p*100+d*10+q
```

```
  performance[2,j] <- c_count/276.00
```

```
}
```

```
round(performance, digits = 2)
```

```
round(mean(performance[2,]), digits = 2)
```

#### 4. ARIMA model predictions

Forex (EUR/GBP) accuracy:

|        |        |        |        |       |        |        |        |        |        |        |       |       |        |        |        |        |        |        |        |
|--------|--------|--------|--------|-------|--------|--------|--------|--------|--------|--------|-------|-------|--------|--------|--------|--------|--------|--------|--------|
| 556.00 | 532.00 | 433.00 | 364.00 | 522.0 | 521.00 | 434.00 | 233.00 | 334.00 | 455.00 | 254.00 | 444.0 | 656.0 | 243.00 | 144.00 | 625.00 | 141.00 | 425.00 | 452.00 | 441.00 |
| 0.51   | 0.54   | 0.52   | 0.54   | 0.5   | 0.51   | 0.51   | 0.51   | 0.51   | 0.52   | 0.51   | 0.5   | 0.5   | 0.51   | 0.51   | 0.51   | 0.49   | 0.53   | 0.52   | 0.52   |

$\mu = 0.51$

(234 predictions)

Cryptocurrency (BTC/USD) accuracy:

|        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |        |
|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|--------|
| 255.00 | 512.00 | 235.00 | 151.00 | 453.00 | 463.00 | 215.00 | 355.00 | 545.00 | 542.00 | 433.00 | 144.00 | 554.00 | 264.00 | 663.00 | 312.00 | 643.00 | 555.00 | 533.00 | 345.00 |
| 0.47   | 0.58   | 0.46   | 0.52   | 0.46   | 0.46   | 0.57   | 0.49   | 0.46   | 0.44   | 0.49   | 0.47   | 0.51   | 0.48   | 0.45   | 0.55   | 0.46   | 0.49   | 0.41   | 0.47   |

$\mu = 0.48$

(276 predictions)

Script used:

#Author: Thomas Hollis

#Subject: Bachelor Thesis

#1. Data Import

```
data <- read.csv("R/BTC.csv")
```

#2. Data Split

```
data_train <- data[1:100,2]
```

```
data_test <- data[101:377,2]
```

#3. Model Train & apply Model

```
performance <- matrix(0,2,20)
```

```
for (j in 1:20)
```

```
{
```

```
  data_train <- data[1:100,2]
```

```
  c_count <- 0
```

```
  p <- round(runif(1, 1, 6))
```

```
  d <- round(runif(1, 1, 6))
```

```
  q <- round(runif(1, 1, 6))
```

```
  for (i in 1:277)
```

```
  {
```

```
    btc_arima <- arima(data_train, order = c(p,d,q), method = "CSS")
```

```
    btc_pred <- predict(btc_arima, 1)
```

```
    if(i >= 2)
```

```
    {
```

```
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||
```

```
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))
```

```
      {
```

```
        c_count = c_count+1
```

```
      }
```

```
    }
```

```
    data_train <- c(data_train, data_test[i])
```

```
  }
```

```
  performance[1,j] <- p*100+d*10+q
```

```
  performance[2,j] <- c_count/276.00
```

```
}
```

```
round(performance, digits = 2)
```

```
round(mean(performance[2,]), digits = 2)
```

## 5. ARCH model predictions

Forex (EUR/GBP) accuracy: ???% (??? predictions)

Cryptocurrency (BTC/USD) accuracy: ???% (??? predictions)

Script used:

```
#Author: Thomas Hollis
```

```
#Subject: Bachelor Thesis
```

```
#0. Package Import
```

```
library(fGarch)
```

```
#1. Data Import
```

```
data <- read.csv("R/BTC.csv")
```

```
#2. Data Split
```

```
data_train <- data[1:100,2]
```

```
data_test <- data[101:377,2]
```

```
#3. Model Train & apply Model
```

```
performance <- matrix(0,2,20)
```

```
for (j in 1:20)
```

```
{
```

```
  data_train <- data[1:100,2]
```

```
  c_count <- 0
```

```
  p <- 1
```

```
  q <- round(runif(1, 1, 6))
```

```
  for (i in 1:277)
```

```
  {
```

```
    btc_arch <- garchFit(~1+garch(p,q),data = data_train,trace=F)
```

```
    btc_pred <- predict(btc_arch, 1)
```

```
    if(i >= 2)
```

```
    {
```

```
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||
```

```
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))
```

```
      {
```

```
        c_count = c_count+1
```

```
      }
```

```
    }
```

```
    data_train <- c(data_train, data_test[i])
```

```
  }
```

```
  performance[1,j] <- p*100+d*10+q
```

```
  performance[2,j] <- c_count/276.00
```

```
}
```

```
round(performance, digits = 2)
```

```
round(mean(performance[2,]), digits = 2)
```

## 6. GARCH model predictions

Forex (EUR/GBP) accuracy: ???% (??? predictions)

Cryptocurrency (BTC/USD) accuracy: ???% (??? predictions)

Script used:

```
#Author: Thomas Hollis
```

```
#Subject: Bachelor Thesis
```

```
#0. Package Import
```

```
library(fGarch)
```

```
#1. Data Import
```

```
data <- read.csv("R/BTC.csv")
```

```
#2. Data Split
```

```
data_train <- data[1:100,2]
```

```
data_test <- data[101:377,2]
```

```
#3. Model Train & apply Model
```

```
performance <- matrix(0,2,20)
```

```
for (j in 1:20)
```

```
{
```

```
  data_train <- data[1:100,2]
```

```
  c_count <- 0
```

```
  p <- round(runif(1, 1, 6))
```

```
  q <- round(runif(1, 1, 6))
```

```
  for (i in 1:277)
```

```
  {
```

```
    btc_arch <- garchFit(~1+garch(p,q),data = data_train,trace=F)
```

```
    btc_pred <- predict(btc_arch, 1)
```

```
    if(i >= 2)
```

```
    {
```

```
      if(((btc_pred$pred[1] > data_test[i-1]) && (data_test[i] > data_test[i-1])) ||
```

```
((btc_pred$pred[1] < data_test[i-1]) && (data_test[i] < data_test[i-1])))
```

```
      {
```

```
        c_count = c_count+1
```

```
      }
```

```
    }
```

```
    data_train <- c(data_train, data_test[i])
```

```
  }
```

```
  performance[1,j] <- p*100+d*10+q
```

```
  performance[2,j] <- c_count/276.00
```

```
}
```

```
round(performance, digits = 2)
```

```
round(mean(performance[2,]), digits = 2)
```

## 7. SLP model predictions

Forex (EUR/GBP) accuracy:

0.58 0.58 0.58 0.58 0.58 0.59 0.58 0.58 0.59 0.59 0.58 0.58 0.58 0.59 0.58 0.58 0.58 0.58 0.58 0.58

$\mu = 0.587$

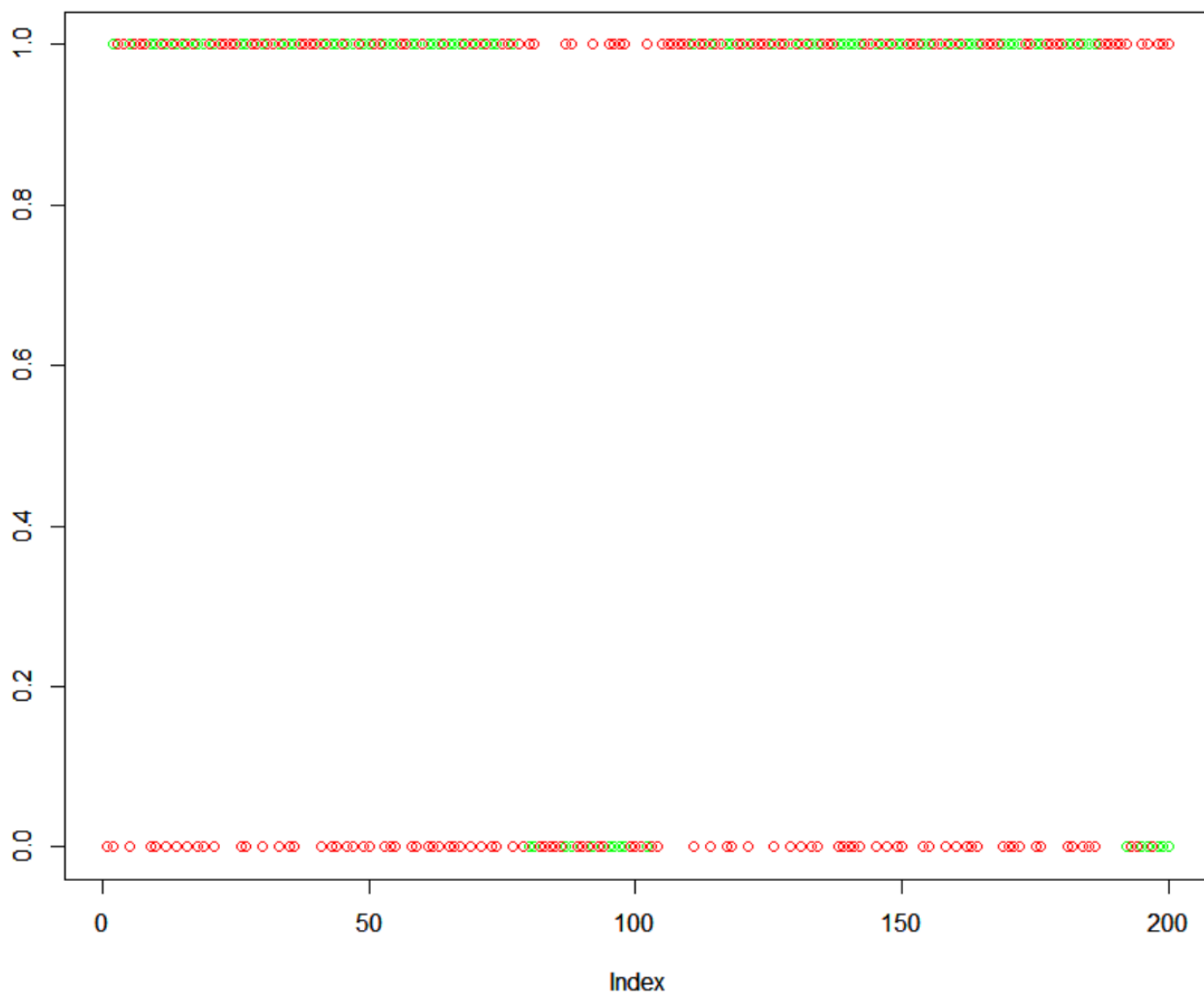
(200 predictions)

Cryptocurrency (BTC/USD) accuracy:

0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52 0.52

$\mu = 0.518$

(200 predictions)





Script used:

```
#Author: Thomas Hollis
#Subject: Bachelor Thesis

#1. Data Import & Global Variables
data <- read.csv("R/DigitData2.csv") #import data

#2. Data post-processing
data_train <- data[1:300,2] #split data to training set
data_test <- data[301:501,2] #split data to testing set
acc = 0

#3. Model application
normalize <- function(x) {
  return ((x - min(x)) / (max(x) - min(x)))
}

data_train <- normalize(data_train)
data_test <- normalize(data_test)

w <- runif(10, -0.2, 0.2)
b = 0
alpha = 0.1
y = integer(8)
right_value = 0

for(i in 1:290)
{
  trainingData <- data_train[i:(i+9)]

  if(data_train[i+10] > data_train[i+9])
  {
    right_value = 1
  }else
  {
    right_value = 0
  }

  if((trainingData%%w + b) > 0)
  {
    y[i] = 1
  }else
  {
    y[i] = 0
  }

  w = w + alpha*(right_value - y[i])*right_value
  b = b + alpha*(right_value - y[i])
}

#prediction
real_result = integer(200)
result = integer(200)
```

```
for(i in 2:200)
{
  if(data_test[i] > data_test[i-1])
  {
    real_result[i] = 1
  }else
  {
    real_result[i] = 0
  }
}
```

```
for(i in 2:191)
{
  testData <- data_test[i:(i+9)]

  if((testData%%w + b) > 0)
  {
    result[i] = 1
  }else
  {
    result[i] = 0
  }

  if(result[i] == real_result[i])
  {
    acc = acc+1
  }
}
```

```
plot(result, col = "green")
par(new = TRUE)
plot(real_result, col = "red")
```

```
result
real_result
acc/200
```

## 8. MLP model predictions

Forex (EUR/GBP) accuracy: ???% (??? predictions)

Cryptocurrency (BTC/USD) accuracy: ???% (??? predictions)

Script used:

(insert code here)

## 9. NN model predictions

Forex (EUR/GBP) accuracy: ???% (??? predictions)

Cryptocurrency (BTC/USD) accuracy: ???% (??? predictions)

Script used:

(insert code here)

## 10. SOM model predictions

Forex (EUR/GBP) accuracy: ???% (??? predictions)

Cryptocurrency (BTC/USD) accuracy: ???% (??? predictions)

Script used:

(insert code here)