

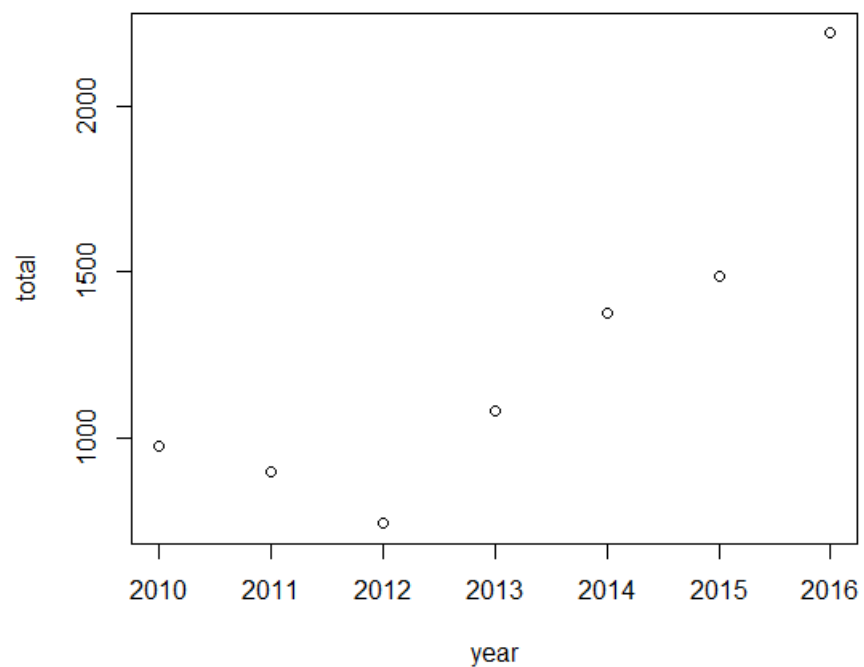
16

a) `> data <- read.csv(file = "https://raw.githubusercontent.com/galaamn/source-code-on-statistics/master/Data%20Sets/Controversy/data.csv")`

b) `> data <- read.csv(file = "https://raw.githubusercontent.com/galaamn/source-code-on-statistics/master/Data%20Sets/Controversy/data.csv")`  
`> data <- data[1:2]`  
`> print(data)`

	year	total
1	2010	978
2	2011	897
3	2012	742
4	2013	1080
5	2014	1377
6	2015	1487
7	2016	2220

c)

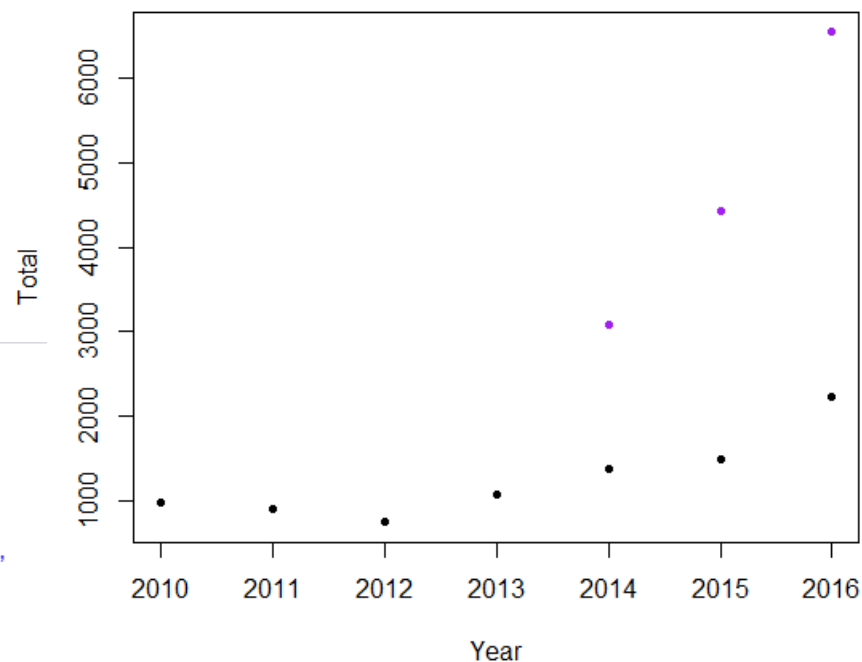


d)

```

1 data <- read.csv(file = "https://raw.githubusercontent.com/galaamn/source-code-on-statistics/master/Data%20Sets/Controversy/data.csv")
2 data <- data[1:2]
3 print(data)
4 plot(data)
5 partial_autocorrelations <- acf(x = data$total, lag.max
6                               = 5, type = "partial", plot = TRUE)
7 print(partial_autocorrelations)
8 fit.ar <- ar.ols(x=data$total, order.max=1)
9 forecast <- predict(fit.ar, n.ahead = 3)$pred
10 plot(x = c(data$year, 2014:2016), y = c(data$total,
11                                         forecast), xlab="Year", ylab="Total",
12       pch = 20, col = c("black", "purple")[rep.int(x=1:2,
13                                                    times=c(length(data$year), 3))])

```

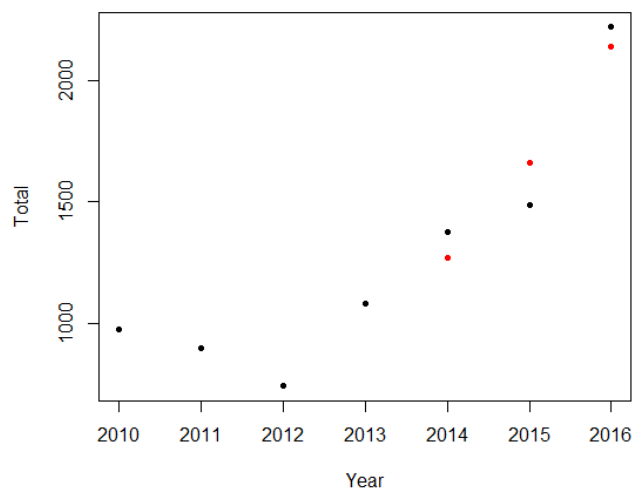


e)

```

> plot(data)
> fit <- lm(formula = log2(total) ~ poly(x=year, degree = 3),
+          data = data, subset = year >= 2010)
> forecast <- 2 ** predict(object = fit, newdata =
+                        data.frame(
+                          year = 2014:2016
+                        ))
> plot(x = c(data$year, 2014:2016), y = log2(c(data$total,
+                                                forecast)), xlab = "Year", ylab = "Total",
+       pch = 20, col = c("black", "red")[rep.int(x = 1:2,
+                                                  times = c(length(data$year), 3))])
> plot(x = c(data$year, 2014:2016), y = c(data$total,
+                                                forecast), xlab = "Year", ylab = "Total",
+       pch = 20, col = c("black", "red")[rep.int(x = 1:2,
+                                                  times = c(length(data$year), 3))])

```



f) Авторегрессион загвар илүү үнэмшилтэй байна.

16.5

a)

```
The downloaded binary packages are in
  C:\Users\ldava\AppData\Local\Temp\Rtmpodza9t\downloaded_packages
> data <- epitools::expand.table(HairEyeColor[,2])
+ install.packages("epitools")
```

b)

```
n <- nrow(data)
set.seed(0)
ind <- sample.int(n = n, size = round(0.2 * n))
training.dataset <- data[-ind,]
testing.dataset <- data[ind,]
```

c)

```
1 data <- epitools::expand.table(HairEyeColor[,2])
2 n <- nrow(data)
3 set.seed(0)
4 ind <- sample.int(n = n, size = round(0.2 * n))
5 training.dataset <- data[-ind,]
6 testing.dataset <- data[ind,]
7 classifier <- klaR::NaiveBayes(formula = class ~ ., data =
8                               training.dataset)
9
```

d)

```
1 data <- epitools::expand.table(HairEyeColor[,2])
2 n <- nrow(data)
3 set.seed(0)
4 ind <- sample.int(n = n, size = round(0.2 * n))
5 training.dataset <- data[-ind,]
6 testing.dataset <- data[ind,]
7 classifier <- klaR::NaiveBayes(formula = class ~ ., data =
8                               training.dataset)
9 test <- predict(classifier, testing.dataset)
10 table("true" = testing.dataset$class, "classifier" =
11       test$class)
```

16.4

```
1 data <- read.csv(file = "https://www.datahub.io/machine-learning/optdigits/r/1.html", header = TRUE)
2 data$class <- as.factor(data$class)
3 tapply(X = data$input64, INDEX = data$class, FUN = var)
4 data$input64 <- NULL
5 set.seed(0)
6 subset <- sample.int(n = nrow(data), size =
7                     round(nrow(data) * 0.8), replace = FALSE)
8 training.dataset <- data[subset,]
9 testing.dataset <- data[-subset,]
10 classifier <- klaR::NaiveBayes(formula = class ~ ., data =
11                               training.dataset)
12 test <- predict(classifier, testing.dataset)
13 table("true" = testing.dataset$class, "classifier" =
14       test$class)
```