# Data Understanding and Data Exploration

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### PDF config

```
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=60),tidy=TRUE)
```

### **EuroCom**

### Dependencies

```
# install.packages('ggplot2')
library(ggplot2)
```

### Reading the dataset

```
phones <- read.csv("./eurocomPHONEchurners.csv")
head(phones)</pre>
```

##		CUST_ID A	REA_CODE	: MI	NUTES_CU	RR_MONTH	MINUTE	S_PF	REV_MON	TH	
##	1	129	45987	•		60			4	56	
##	2	130	15563	}		2				0	
##	3	131	10040	)		2				0	
##	4	132 21750		)	678			1222			
##	5	133 55166		3	110			98			
##	6	134	36785	5		97				56	
##		MINUTES_3	MONTHS_A	GO	CUST_MOS	LONGDIST	_FLAG	CALL	WAITIN	G_FLAG	NUM_LINES
##	1		3	398	13		0			1	1
##	2			4	4		0			0	1
##	3			0	1		0			0	1
##	4		5	98	30		1			1	2
##	5			56	15		1			0	1
##	6			97	8		0			0	1
##		VOICEMAIL	_FLAG MC	BII	LE_PLAN C	ONVERGENT	_BILLI	NG C	ENDER	INCOME	
##	1		1		0		Y	es	M	88000	
##	2		0		0		Y	es	M	53000	
##	3		1		0			No	F	29000	
##	_	0			1			es	M		
##	_		1		0			No	M	98000	
##	6		0		1			No		125000	
##		_			CATION TO	T_MINUTES	_		JRNER		
		Internati			asters		914		yes		
		Internati					6		no		
##			onal Hig	•			0		no		
		Internati	-				2498		yes		
##	5	Promo_	plan Hig	gh S	School		264	ŀ	no		

## 6 National 250 no

1

### Data pre-processing

#### 1.a

Getting how many null values, or empty string values there is per column.

```
count_na <- sapply(phones, function(y) sum(length(which(is.na(y) |</pre>
    y == ""))))
na_df <- data.frame(count_na)</pre>
subset(na_df, na_df$count_na > 0)
##
                         count_na
## MINUTES_3MONTHS_AGO
## CUST MOS
                                 3
## PHONE_PLAN
                                 4
## EDUCATION
                                 8
## TOT_MINUTES_USAGE
1.b
Replacing na numerics with medians
replace_na_with_median <- function(col) {</pre>
    median_without_na <- median(col, na.rm = TRUE)</pre>
```

#### MINUTES\_3MONTHS\_AGO

return(col)

}

col[is.na(col)] <- median\_without\_na</pre>

```
phones$MINUTES_3MONTHS_AGO <- replace_na_with_median(phones$MINUTES_3MONTHS_AGO)</pre>
```

### $CUST\_MOS$

```
phones$CUST_MOS <- replace_na_with_median(phones$CUST_MOS)
```

#### TOT\_MINUTES\_USAGE

```
phones$TOT_MINUTES_USAGE <- replace_na_with_median(phones$TOT_MINUTES_USAGE)
```

```
Getting the mode for categorical columns PER GENDER
```

```
get_mode <- function(x) {
    xtable <- table(x)
    idx <- xtable == max(xtable)
    names(xtable)[idx]
}</pre>
```

Function to get all modes from a data frame

```
get_modes <- function(x) {
   if (class(x) == "numeric" | class(x) == "integer")
        return("X")
   xtable <- table(x)
   idx <- xtable == max(xtable)
   names(xtable)[idx]
}</pre>
```

Displaying modes for males

```
phones_male <- phones[phones$GENDER == "M", ]

modes_male <- data.frame(sapply(phones_male, get_modes))
names(modes_male)[1] <- "MODE_MALE"
modes_male <- subset(modes_male, MODE_MALE != "X")

modes_male</pre>
```

```
## CONVERGENT_BILLING Yes
## GENDER M
## PHONE_PLAN International
## EDUCATION Post Primary
## CHURNER yes
```

Displaying modes for females

```
phones_female <- phones[phones$GENDER == "F", ]

modes_female <- data.frame(sapply(phones_female, get_modes))
names(modes_female)[1] <- "MODE_FEMALE"
modes_female <- subset(modes_female, MODE_FEMALE != "X")

modes_female</pre>
```

#### PHONE PLAN

```
phones$PHONE_PLAN[phones$PHONE_PLAN == "" & phones$GENDER ==
    "M"] <- get_mode(phones$PHONE_PLAN[phones$GENDER == "M"])
phones$PHONE_PLAN[phones$PHONE_PLAN == "" & phones$GENDER ==
    "F"] <- get_mode(phones$PHONE_PLAN[phones$GENDER == "F"])</pre>
```

#### **EDUCATION**

```
phones$EDUCATION[phones$EDUCATION == "" & phones$GENDER == "M"] <- get_mode(phones$EDUCATION[phones$GENDER == "M"])
phones$EDUCATION[phones$EDUCATION == "" & phones$GENDER == "F"] <- get_mode(phones$EDUCATION[phones$GENDER == "F"])</pre>
```

### AREA\_CODE

```
phones$AREA_CODE[phones$AREA_CODE == "" & phones$GENDER == "M"] <- get_mode(phones$AREA_CODE[phones$GENDER == "M"])
phones$AREA_CODE[phones$AREA_CODE == "" & phones$GENDER == "F"] <- get_mode(phones$AREA_CODE[phones$GENDER == "F"])</pre>
```

### LONGDIST\_FLAG

```
phones$LONGDIST_FLAG[phones$LONGDIST_FLAG == "" & phones$GENDER ==
   "M"] <- get_mode(phones$LONGDIST_FLAG[phones$GENDER == "M"])
phones$LONGDIST_FLAG[phones$LONGDIST_FLAG == "" & phones$GENDER ==
   "F"] <- get_mode(phones$LONGDIST_FLAG[phones$GENDER == "F"])</pre>
```

#### CALLWAITING\_FLAG

```
phones$CALLWAITING_FLAG[phones$CALLWAITING_FLAG == "" & phones$GENDER ==
    "M"] <- get_mode(phones$CALLWAITING_FLAG[phones$GENDER ==
    "M"])
phones$CALLWAITING_FLAG[phones$CALLWAITING_FLAG == "" & phones$GENDER ==
    "F"] <- get_mode(phones$CALLWAITING_FLAG[phones$GENDER ==
    "F"])</pre>
```

#### VOICEMAIL FLAG

```
phones$VOICEMAIL_FLAG[phones$VOICEMAIL_FLAG == "" & phones$GENDER ==
    "M"] <- get_mode(phones$VOICEMAIL_FLAG[phones$GENDER == "M"])
phones$VOICEMAIL_FLAG[phones$VOICEMAIL_FLAG == "" & phones$GENDER ==
    "F"] <- get_mode(phones$VOICEMAIL_FLAG[phones$GENDER == "F"])</pre>
```

#### MOBILE\_PLAN

```
phones$MOBILE_PLAN[phones$MOBILE_PLAN === "" & phones$GENDER ==
    "M"] <- get_mode(phones$MOBILE_PLAN[phones$GENDER == "M"])
phones$MOBILE_PLAN[phones$MOBILE_PLAN === "" & phones$GENDER ==
    "F"] <- get_mode(phones$MOBILE_PLAN[phones$GENDER == "F"])</pre>
```

2

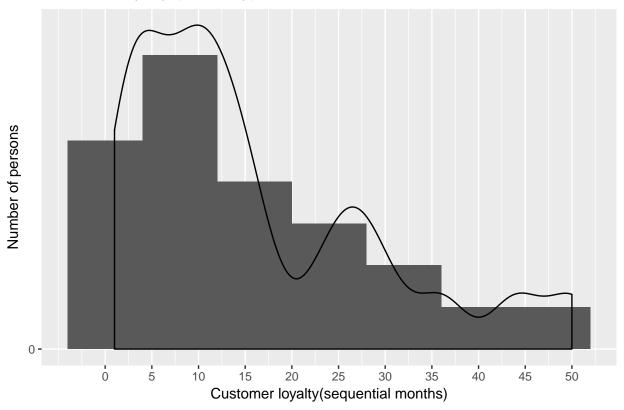
#### Discretising Income predictor values

```
head(phones$INCOME)
## [1] 88000 53000 29000 46000 98000 125000
phones$INCOME <- cut(phones$INCOME, breaks = c(0, 37999, 88000,</pre>
    max(phones$INCOME)), include.lowest = TRUE, labels = c("Low Income",
    "Medium Income", "High Income"))
head(phones$INCOME)
## [1] Medium Income Medium Income Low Income
                                                  Medium Income High Income
## [6] High Income
## Levels: Low Income Medium Income High Income
3.c
get_mode(phones$AREA_CODE)
## [1] "10040"
summary(phones$CUST_MOS)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                               Max.
      1.00
              6.00
                     11.00
                             16.05
                                      26.00
                                              50.00
get_mode(phones$LONGDIST_FLAG)
## [1] "1"
get_mode(phones$CALLWAITING_FLAG)
## [1] "0"
summary(phones$NUM_LINES)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                               Max.
             1.000
                     1.000
                             1.391
                                      2.000
                                              3.000
get_mode(phones$VOICEMAIL_FLAG)
## [1] "1"
get_mode(phones$MOBILE_PLAN)
## [1] "0"
```

```
get_mode(phones$CONVERGENT_BILLING)
## [1] "No"
get_mode(phones$GENDER)
## [1] "M"
get_mode(phones$INCOME)
## [1] "Medium Income"
get_mode(phones$PHONE_PLAN)
## [1] "International"
get_mode(phones$EDUCATION)
## [1] "Post Primary"
summary(phones$TOT_MINUTES_USAGE)
##
     Min. 1st Qu. Median
                             Mean 3rd Qu.
                                             Max.
##
              116
                      264
                             2036
                                     1677
                                            36240
        0
3.d
CUST MOS
```

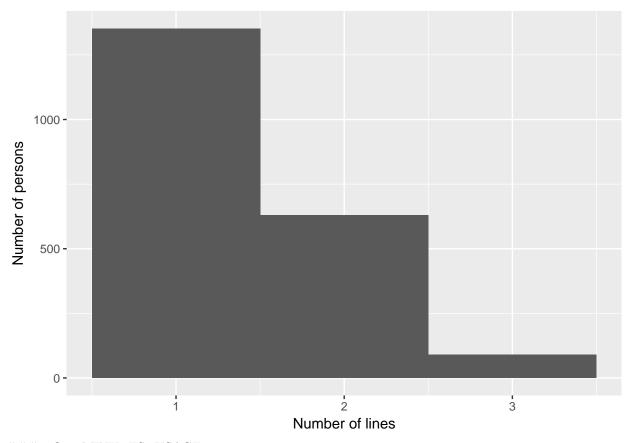
```
ggplot(data = phones, aes(phones$CUST_MOS)) + geom_histogram(binwidth = 8,
   aes(y = ..density..)) + scale_x_continuous(breaks = seq(0,
   60, 5)) + scale_y_continuous(breaks = seq(0, 1000, 50)) +
   labs(x = "Customer loyalty(sequential months)", y = "Number of persons",
       title = "Customer loyalty (+ density)") + geom_density()
```

## Customer loyalty (+ density)



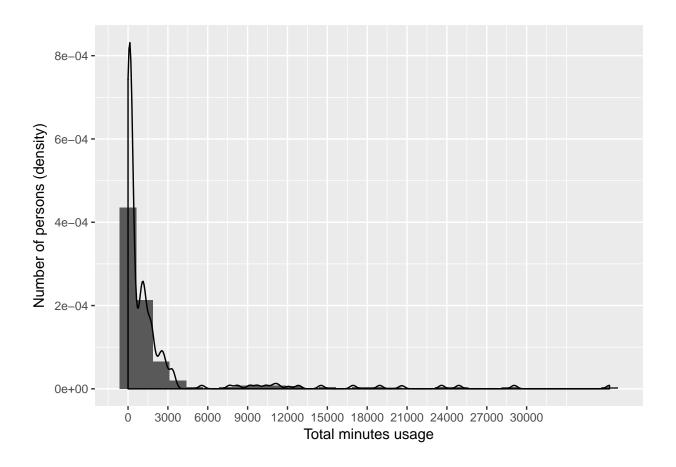
### ### NUM\_LINES

```
ggplot(data = phones, aes(phones$NUM_LINES)) + geom_histogram(binwidth = 1) +
    scale_x_continuous(breaks = 0:3) + labs(x = "Number of lines",
    y = "Number of persons")
```



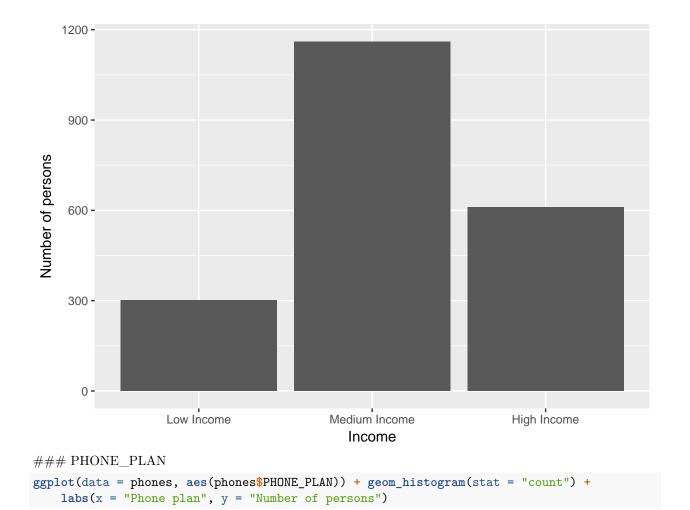
### ### TOT\_MINUTES\_USAGE

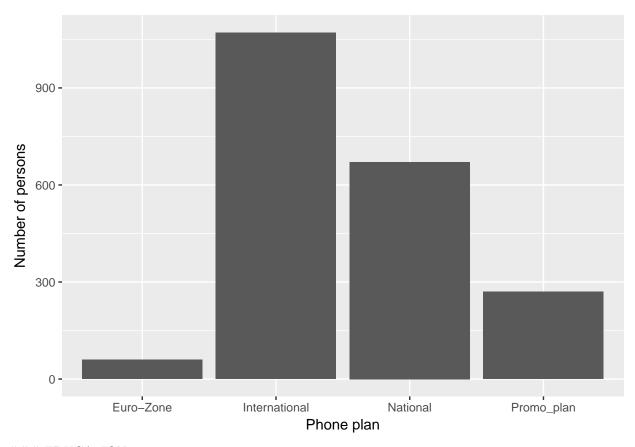
```
ggplot(data = phones, aes(phones$TOT_MINUTES_USAGE)) + geom_histogram(bins = 30,
    aes(y = ..density..)) + scale_x_continuous(breaks = seq(0,
    30000, 3000)) + labs(x = "Total minutes usage", y = "Number of persons (density)") +
    geom_density()
```



### INCOME

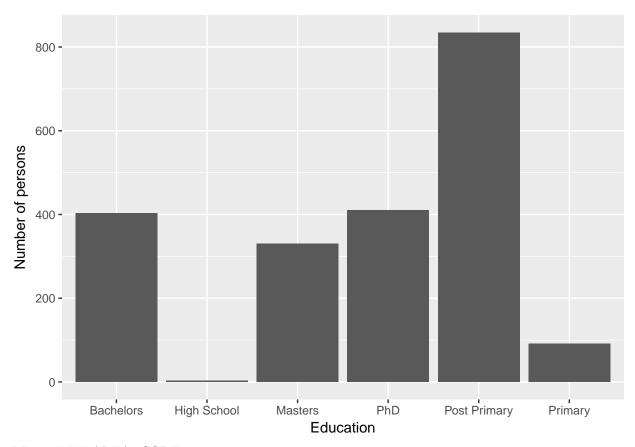
```
ggplot(data = phones, aes(phones$INCOME)) + geom_histogram(stat = "count") +
labs(x = "Income", y = "Number of persons")
```





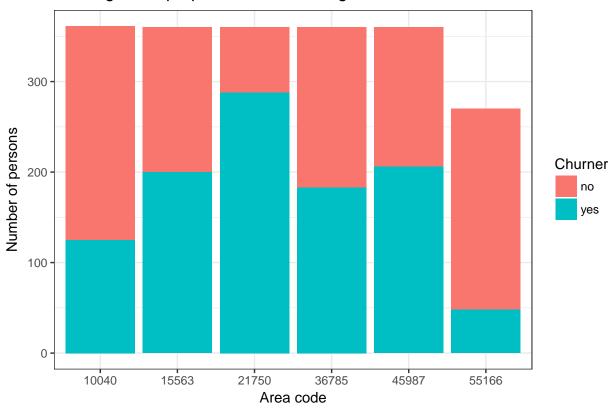
```
\#\#\# EDUCATION
```

```
ggplot(data = phones, aes(phones$EDUCATION)) + geom_histogram(stat = "count") +
    labs(x = "Education", y = "Number of persons")
```



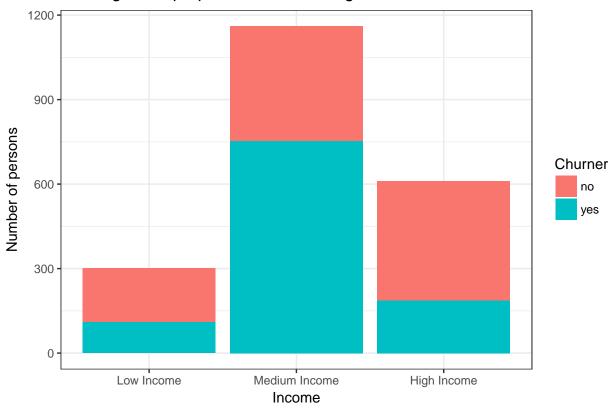
## 3.e ### AREA\_CODE

```
ggplot(data = phones, aes(x = phones$AREA_CODE, group = phones$CHURNER,
  fill = phones$CHURNER)) + geom_histogram(stat = "count") +
  theme_bw() + labs(x = "Area code", y = "Number of persons",
  title = "Looking for disproportions in Churning rates", fill = "Churner")
```



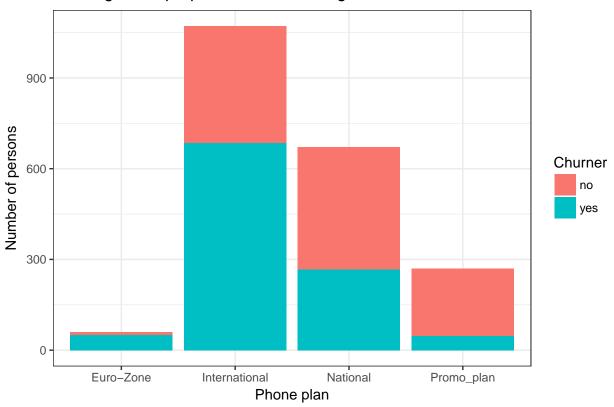
### ### INCOME

```
ggplot(data = phones, aes(x = phones$INCOME, group = phones$CHURNER,
  fill = phones$CHURNER)) + geom_histogram(stat = "count") +
  theme_bw() + labs(x = "Income", y = "Number of persons",
  title = "Looking for disproportions in Churning rates", fill = "Churner")
```



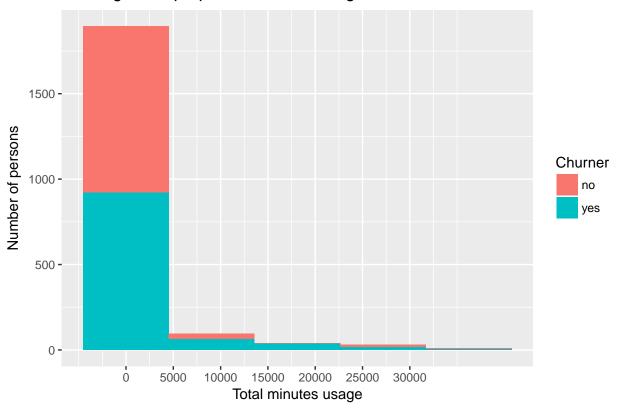
```
\#\#\# PHONE_PLAN
```

```
ggplot(data = phones, aes(x = phones$PHONE_PLAN, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(stat = "count") +
    theme_bw() + labs(x = "Phone plan", y = "Number of persons",
    title = "Looking for disproportions in Churning rates", fill = "Churner")
```



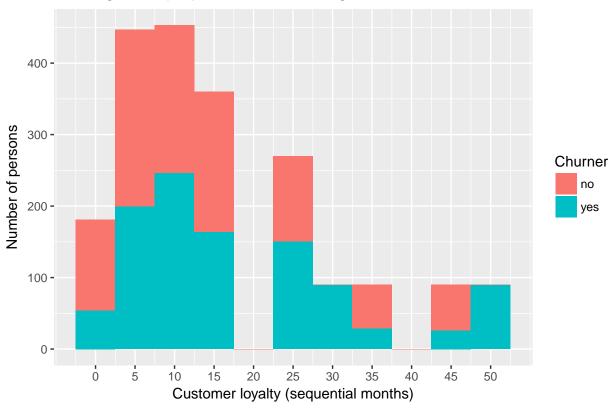
### TOT\_MINUTES\_USAGE

```
ggplot(data = phones, aes(x = phones$TOT_MINUTES_USAGE, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(bins = 5) + scale_x_continuous(breaks = seq(0, 30000, 5000)) + labs(x = "Total minutes usage", y = "Number of persons",
    title = "Looking for disproportions in Churning rates", fill = "Churner")
```



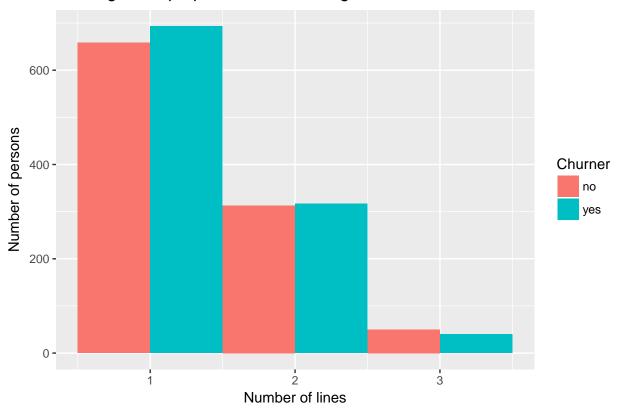
### CUST\_MOS

```
ggplot(data = phones, aes(x = phones$CUST_MOS, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(binwidth = 5) +
    scale_x_continuous(breaks = seq(0, 50, 5)) + labs(x = "Customer loyalty (sequential months)",
    y = "Number of persons", title = "Looking for disproportions in Churning rates",
    fill = "Churner")
```



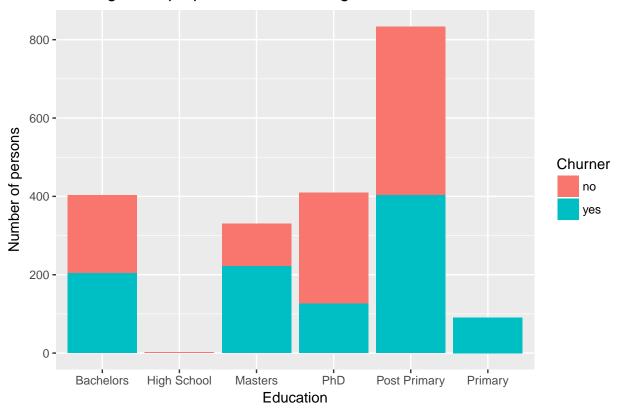
### ### NUM\_LINES

```
ggplot(data = phones, aes(x = phones$NUM_LINES, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(binwidth = 1, position = "dodge") +
    scale_x_continuous(breaks = 0:3) + labs(x = "Number of lines",
    y = "Number of persons", title = "Looking for disproportions in Churning rates",
    fill = "Churner")
```



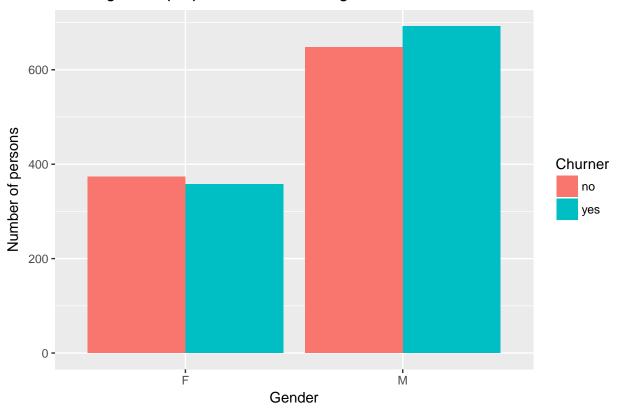
### ### EDUCATION

```
ggplot(data = phones, aes(x = phones$EDUCATION, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(stat = "count") +
    labs(x = "Education", y = "Number of persons", title = "Looking for disproportions in Churning rate
    fill = "Churner")
```



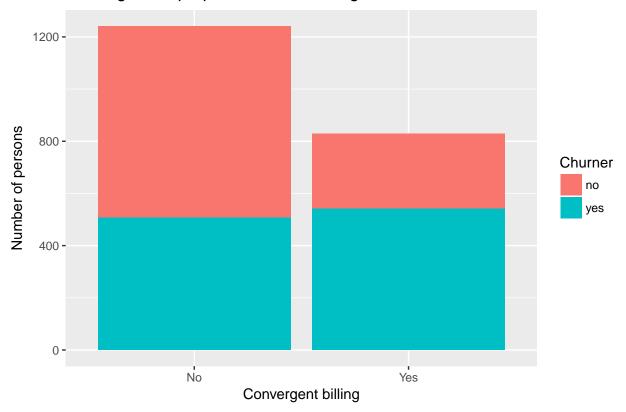
### ### GENDER

```
ggplot(data = phones, aes(x = phones$GENDER, group = phones$CHURNER,
  fill = phones$CHURNER)) + geom_histogram(stat = "count",
  position = "dodge") + labs(x = "Gender", y = "Number of persons",
  title = "Looking for disproportions in Churning rates", fill = "Churner")
```



```
### CONVERGENT_BILLING
```

```
ggplot(data = phones, aes(x = phones$CONVERGENT_BILLING, group = phones$CHURNER,
    fill = phones$CHURNER)) + geom_histogram(stat = "count") +
    labs(x = "Convergent billing", y = "Number of persons", title = "Looking for disproportions in Churrefill = "Churner")
```



## 3.f ### CUST\_MOS

## [1] 1.131224

### NUM\_LINES

```
num_lines_skew <- (3 * (mean(phones$NUM_LINES) - median(phones$NUM_LINES)))/sd(phones$NUM_LINES)
num_lines_skew</pre>
```

## [1] 2.057503

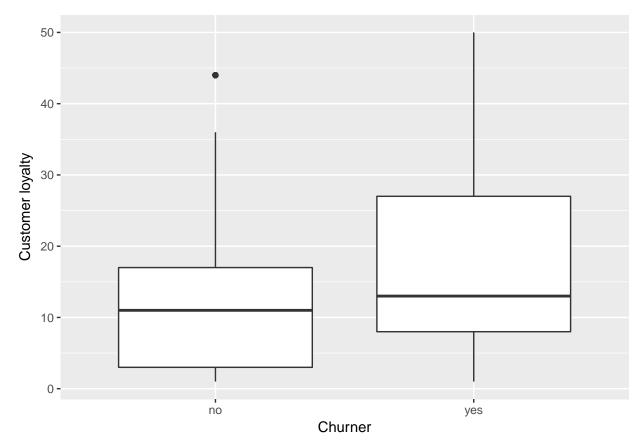
#### TOT\_MINUTES\_USAGE

```
tot_minutes_usage_skew <- (3 * (mean(phones$TOT_MINUTES_USAGE) -
    median(phones$TOT_MINUTES_USAGE)))/sd(phones$TOT_MINUTES_USAGE)
tot_minutes_usage_skew</pre>
```

## [1] 1.088757

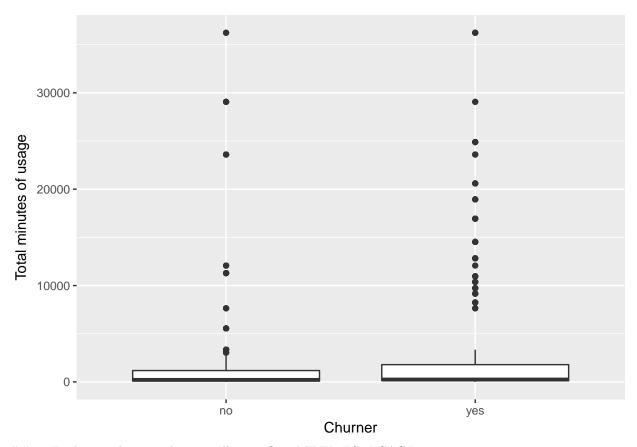
3.gCUST\_MOS

```
ggplot(data = phones, aes(phones$CHURNER, phones$CUST_MOS)) +
   geom_boxplot() + labs(x = "Churner", y = "Customer loyalty")
```



```
### TOT_MINUTES_USAGE
```

```
ggplot(data = phones, aes(phones$CHURNER, phones$TOT_MINUTES_USAGE)) +
geom_boxplot() + labs(x = "Churner", y = "Total minutes of usage")
```



## 4. Finding outliers mathematically in TOT\_MINUTES\_USAGE

IQR method

```
summary(phones$TOT_MINUTES_USAGE)
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                               Max.
##
         0
                               2036
                                              36240
               116
                       264
                                       1677
IQR <- 1677 - 116
lower_bound <- 116 - (IQR * 1.5)
upper_bound <- 1677 + (IQR * 1.5)
nrow(phones[phones$TOT_MINUTES_USAGE < lower_bound | phones$TOT_MINUTES_USAGE >
    upper_bound, ])
## [1] 176
Z standardisation method
z_score_tot_minutes_usage <- scale(phones$TOT_MINUTES_USAGE,</pre>
    center = TRUE, scale = TRUE)
# same as (phones$TOT_MINUTES_USAGE -
# mean(phones$TOT_MINUTES_USAGE))/sd(phones$TOT_MINUTES_USAGE)
summary(z_score_tot_minutes_usage)
##
          ۷1
          :-0.41698
  Min.
## 1st Qu.:-0.39323
```

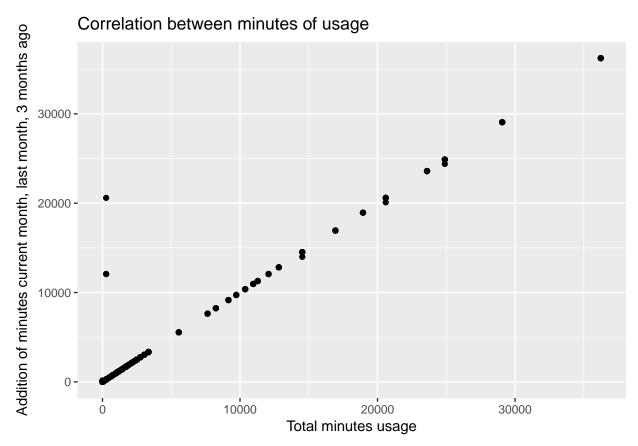
```
## Median :-0.36292
## Mean : 0.00000
## 3rd Qu.:-0.07354
## Max. : 7.00417
z_range <- table(z_score_tot_minutes_usage > -3 & z_score_tot_minutes_usage 
z_range[names(z_range) == FALSE]
## FALSE
##
      69
5
tot_mins_before_transfo <- (3 * (mean(phones$TOT_MINUTES_USAGE) -</pre>
    median(phones$TOT_MINUTES_USAGE)))/sd(phones$TOT_MINUTES_USAGE)
tot_mins_before_transfo
## [1] 1.088757
5.a
Z-score standardisation see above, we reduced the number of outliers from 176 to 69
tot_mins_z_score <- (3 * (mean(z_score_tot_minutes_usage) - median(z_score_tot_minutes_usage)))/sd(z_sc
tot_mins_z_score
## [1] 1.088757
5.b
Natural log
natural_log_transfo <- log(phones$TOT_MINUTES_USAGE[phones$TOT_MINUTES_USAGE !=
natural_log_transfo_skewness <- (3 * (mean(natural_log_transfo) -</pre>
    median(natural_log_transfo)))/sd(natural_log_transfo)
natural_log_transfo_skewness
## [1] -0.7042918
5.c
Square root
square_root_transfo <- sqrt(phones$TOT_MINUTES_USAGE)</pre>
square_root_transfo_skewness <- (3 * (mean(square_root_transfo) -</pre>
    median(square_root_transfo)))/sd(square_root_transfo)
square_root_transfo_skewness
## [1] 1.288432
```

### 7.a.

#### Correlation

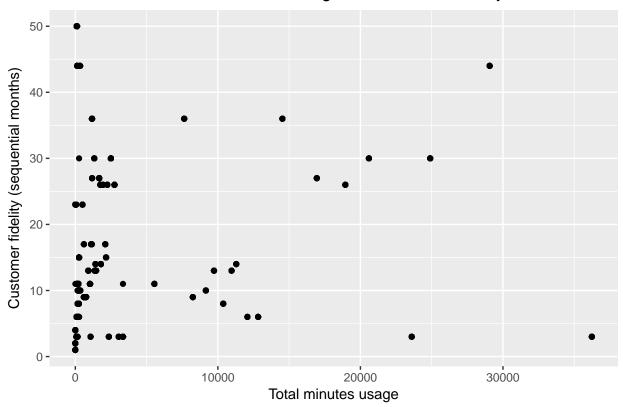
Minutes of usage

```
ggplot(data = phones, aes(x = phones$TOT_MINUTES_USAGE, y = phones$MINUTES_CURR_MONTH +
    phones$MINUTES_PREV_MONTH + phones$MINUTES_3MONTHS_AGO)) +
    geom_point() + labs(x = "Total minutes usage", y = "Addition of minutes current month, last month,
    title = "Correlation between minutes of usage")
```



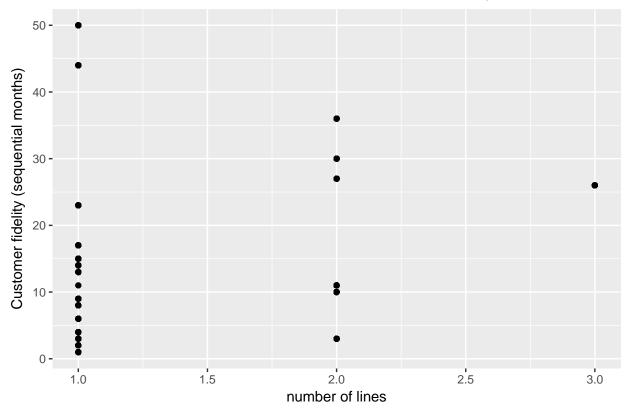
```
ggplot(data = phones, aes(x = phones$TOT_MINUTES_USAGE, y = phones$CUST_MOS)) +
    geom_point() + labs(y = "Customer fidelity (sequential months)",
    x = "Total minutes usage", title = "Correlation between minutes of usage and customer fidelity")
```

### Correlation between minutes of usage and customer fidelity



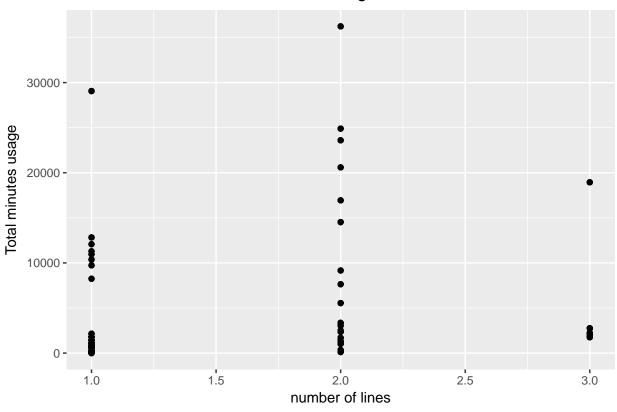
```
ggplot(data = phones, aes(x = phones$NUM_LINES, y = phones$CUST_MOS)) +
    geom_point() + labs(y = "Customer fidelity (sequential months)",
    x = "number of lines", title = "Correlation between number of lines and customer fidelity")
```

### Correlation between number of lines and customer fidelity



```
ggplot(data = phones, aes(x = phones$NUM_LINES, y = phones$TOT_MINUTES_USAGE)) +
    geom_point() + labs(y = "Total minutes usage", x = "number of lines",
    title = "Correlation between total minutes usage and number of lines")
```

### Correlation between total minutes usage and number of lines



#### 7.b

```
Minutes usage metrics correlation
```

```
covariance_minutes <- cov(phones$TOT_MINUTES_USAGE, phones$MINUTES_CURR_MONTH +
    phones$MINUTES_PREV_MONTH + phones$MINUTES_3MONTHS_AGO)
covariance_minutes</pre>
```

#### ## [1] 23778254

```
correlation_minutes <- covariance_minutes/(sd(phones$TOT_MINUTES_USAGE) *
    sd(phones$MINUTES_CURR_MONTH + phones$MINUTES_PREV_MONTH +
        phones$MINUTES_3MONTHS_AGO))
correlation_minutes</pre>
```

### ## [1] 0.9916396

Usage and customer fidelity

```
covariance_minutes_fid <- cov(phones$TOT_MINUTES_USAGE, phones$CUST_MOS)
covariance minutes fid</pre>
```

#### ## [1] 5931.69

```
correlation_minutes_fid <- covariance_minutes_fid/(sd(phones$TOT_MINUTES_USAGE) *
    sd(phones$CUST_MOS))
correlation_minutes_fid</pre>
```

```
## [1] 0.09075367
covariance_lines_fid <- cov(phones$NUM_LINES, phones$CUST_MOS)</pre>
covariance_lines_fid
## [1] 1.550566
correlation_lines_fid <- covariance_lines_fid/(sd(phones$NUM_LINES) *</pre>
    sd(phones$CUST_MOS))
correlation_lines_fid
## [1] 0.2031285
covariance_lines_minutes <- cov(phones$NUM_LINES, phones$TOT_MINUTES_USAGE)
covariance_lines_minutes
## [1] 685.4576
correlation_lines_minutes <- covariance_lines_minutes/(sd(phones$NUM_LINES) *
    sd(phones$TOT_MINUTES_USAGE))
correlation lines minutes
## [1] 0.2461581
Part 2
Preparating Data for learning
keep <- c("INCOME", "PHONE_PLAN", "EDUCATION", "AREA_CODE", "CUS_MOS",
    "CHURNER", "CONVERGENT_BILLING")
phones_learning <- phones[, (names(phones) %in% keep)]</pre>
phones_learning$AREA_CODE <- as.factor(phones_learning$AREA_CODE)
head(phones_learning)
##
     AREA_CODE CONVERGENT_BILLING
                                           INCOME
                                                     PHONE_PLAN
                                                                    EDUCATION
## 1
         45987
                               Yes Medium Income International
                                                                      Masters
                               Yes Medium Income International
## 2
         15563
                                                                    Bachelors
## 3
         10040
                                      Low Income
                                                       National High School
## 4
         21750
                               Yes Medium Income International High School
## 5
         55166
                                     High Income
                                                     Promo_plan High School
## 6
         36785
                                No
                                     High Income
                                                       National Post Primary
##
     CHURNER
## 1
         yes
## 2
          no
## 3
          no
## 4
         yes
## 5
          no
## 6
          no
Writing the learning data to csv
```

write.csv(phones\_learning, file = "./learning\_churners.csv")