

GrupoBimbo-InventoryDemanda

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7/17/2021

Carregando dados e pacotes

```
library(readr)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(caret)
```

```
## Loading required package: lattice
## Loading required package: ggplot2
```

```
library(ggplot2)
library(mltools)
library(e1071)
```

```
##
## Attaching package: 'e1071'
## The following object is masked from 'package:mltools':
##
##   skewness
```

```
library(stringr)
```

```
# Lendo os arquivos
df <- read_csv('C:/Users/hiago/Downloads/datasets/train.csv')
```

```
##
## -- Column specification -----
## cols(
##   Semana = col_double(),
##   Agencia_ID = col_double(),
##   Canal_ID = col_double(),
##   Ruta_SAK = col_double(),
##   Cliente_ID = col_double(),
```

```

## Producto_ID = col_double(),
## Venta_uni_hoy = col_double(),
## Venta_hoy = col_double(),
## Dev_uni_proxima = col_double(),
## Dev_proxima = col_double(),
## Demanda_uni_equil = col_double()
## )

df_cliente <- read_csv('C:/Users/hiago/Downloads/datasets/cliente_tabla.csv')

##
## -- Column specification -----
## cols(
##   Cliente_ID = col_double(),
##   NombreCliente = col_character()
## )

df_producto <- read_csv('C:/Users/hiago/Downloads/datasets/producto_tabla.csv')

##
## -- Column specification -----
## cols(
##   Producto_ID = col_double(),
##   NombreProducto = col_character()
## )

df_town <- read_csv('C:/Users/hiago/Downloads/datasets/town_state.csv')

##
## -- Column specification -----
## cols(
##   Agencia_ID = col_double(),
##   Town = col_character(),
##   State = col_character()
## )

df_test <- read_csv('C:/Users/hiago/Downloads/datasets/test.csv')

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   Semana = col_double(),
##   Agencia_ID = col_double(),
##   Canal_ID = col_double(),
##   Ruta_SAK = col_double(),
##   Cliente_ID = col_double(),
##   Producto_ID = col_double()
## )

```

Pré processamento

```

head(df)

## # A tibble: 6 x 11
##   Semana Agencia_ID Canal_ID Ruta_SAK Cliente_ID Producto_ID Venta_uni_hoy
##   <dbl>      <dbl>    <dbl>   <dbl>    <dbl>      <dbl>      <dbl>

```

```
## 1      3      1110      7      3301      15766      1212      3
## 2      3      1110      7      3301      15766      1216      4
## 3      3      1110      7      3301      15766      1238      4
## 4      3      1110      7      3301      15766      1240      4
## 5      3      1110      7      3301      15766      1242      3
## 6      3      1110      7      3301      15766      1250      5
## # ... with 4 more variables: Venta_hoy <dbl>, Dev_uni_proxima <dbl>,
## #   Dev_proxima <dbl>, Demanda_uni_equil <dbl>
```

```
str(df)
```

```
## spec_tbl_df[,11] [74,180,464 x 11] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Semana      : num [1:74180464] 3 3 3 3 3 3 3 3 3 3 ...
## $ Agencia_ID   : num [1:74180464] 1110 1110 1110 1110 1110 1110 1110 1110 1110 1110 ...
## $ Canal_ID     : num [1:74180464] 7 7 7 7 7 7 7 7 7 7 ...
## $ Ruta_SAK     : num [1:74180464] 3301 3301 3301 3301 3301 3301 ...
## $ Cliente_ID   : num [1:74180464] 15766 15766 15766 15766 15766 ...
## $ Producto_ID  : num [1:74180464] 1212 1216 1238 1240 1242 ...
## $ Venta_uni_hoy : num [1:74180464] 3 4 4 4 3 5 3 6 4 6 ...
## $ Venta_hoy    : num [1:74180464] 25.1 33.5 39.3 33.5 22.9 ...
## $ Dev_uni_proxima : num [1:74180464] 0 0 0 0 0 0 0 0 0 0 ...
## $ Dev_proxima   : num [1:74180464] 0 0 0 0 0 0 0 0 0 0 ...
## $ Demanda_uni_equil: num [1:74180464] 3 4 4 4 3 5 3 6 4 6 ...
## - attr(*, "spec")=
## .. cols(
## ..   Semana = col_double(),
## ..   Agencia_ID = col_double(),
## ..   Canal_ID = col_double(),
## ..   Ruta_SAK = col_double(),
## ..   Cliente_ID = col_double(),
## ..   Producto_ID = col_double(),
## ..   Venta_uni_hoy = col_double(),
## ..   Venta_hoy = col_double(),
## ..   Dev_uni_proxima = col_double(),
## ..   Dev_proxima = col_double(),
## ..   Demanda_uni_equil = col_double()
## .. )
```

```
head(df_cliente)
```

```
## # A tibble: 6 x 2
##   Cliente_ID NombreCliente
##   <dbl> <chr>
## 1      0 SIN NOMBRE
## 2      1 OXXO XINANTECATL
## 3      2 SIN NOMBRE
## 4      3 EL MORENO
## 5      4 SDN SER DE ALIM CUERPO SA CIA DE INT
## 6      4 SDN SER DE ALIM CUERPO SA CIA DE INT
```

```
str(df_cliente)
```

```
## spec_tbl_df[,2] [935,362 x 2] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Cliente_ID   : num [1:935362] 0 1 2 3 4 5 6 7 8 ...
## $ NombreCliente: chr [1:935362] "SIN NOMBRE" "OXXO XINANTECATL" "SIN NOMBRE" "EL MORENO" ...
## - attr(*, "spec")=
```

```
## .. cols(
## ..   Cliente_ID = col_double(),
## ..   NombreCliente = col_character()
## .. )
```

```
head(df_producto)
```

```
## # A tibble: 6 x 2
##   Producto_ID NombreProducto
##       <dbl> <chr>
## 1         0 NO IDENTIFICADO 0
## 2         9 Capuccino Moka 750g NES 9
## 3        41 Bimbollos Ext sAjonjoli 6p 480g BIM 41
## 4        53 Burritos Sincro 170g CU LON 53
## 5        72 Div Tira Mini Doradita 4p 45g TR 72
## 6        73 Pan Multigrano Linaza 540g BIM 73
```

```
str(df_producto)
```

```
## spec_tbl_df[,2] [2,592 x 2] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Producto_ID : num [1:2592] 0 9 41 53 72 73 98 99 100 106 ...
## $ NombreProducto: chr [1:2592] "NO IDENTIFICADO 0" "Capuccino Moka 750g NES 9" "Bimbollos Ext sAjonjoli 6p 480g BIM 41" ...
## - attr(*, "spec")=
## .. cols(
## ..   Producto_ID = col_double(),
## ..   NombreProducto = col_character()
## .. )
```

```
head(df_town)
```

```
## # A tibble: 6 x 3
##   Agencia_ID Town State
##       <dbl> <chr> <chr>
## 1    1110 2008 AG. LAGO FILT MÉXICO, D.F.
## 2    1111 2002 AG. AZCAPOTZALCO MÉXICO, D.F.
## 3    1112 2004 AG. CUAUTITLAN ESTADO DE MÉXICO
## 4    1113 2008 AG. LAGO FILT MÉXICO, D.F.
## 5    1114 2029 AG. IZTAPALAPA 2 MÉXICO, D.F.
## 6    1116 2011 AG. SAN ANTONIO MÉXICO, D.F.
```

```
str(df_town)
```

```
## spec_tbl_df[,3] [790 x 3] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ Agencia_ID: num [1:790] 1110 1111 1112 1113 1114 ...
## $ Town      : chr [1:790] "2008 AG. LAGO FILT" "2002 AG. AZCAPOTZALCO" "2004 AG. CUAUTITLAN" "2008 AG. IZTAPALAPA 2" ...
## $ State     : chr [1:790] "MÉXICO, D.F." "MÉXICO, D.F." "ESTADO DE MÉXICO" "MÉXICO, D.F." ...
## - attr(*, "spec")=
## .. cols(
## ..   Agencia_ID = col_double(),
## ..   Town = col_character(),
## ..   State = col_character()
## .. )
```

```
# Os datasets cliente, producto e town não serão relevantes para o processo de predição
# pois contêm apenas descrições de Ids referenciados no DF principal.
# Porém podem ser úteis no processo de análise
```

```
# Selecionado apenas variáveis preditoras + target
# As variáveis Venta_uni_hot, venta_hoy, dev_uni_proxima e dev_proxima não estão no
# df de teste, pois a variável target é resultado de cálculos a partir dessas variáveis
```

```
df <- df %>% select(Semana, Agencia_ID, Canal_ID, Ruta_SAK, Cliente_ID, Producto_ID, Demanda_uni_equil)
```

```
head(df)
```

```
## # A tibble: 6 x 7
```

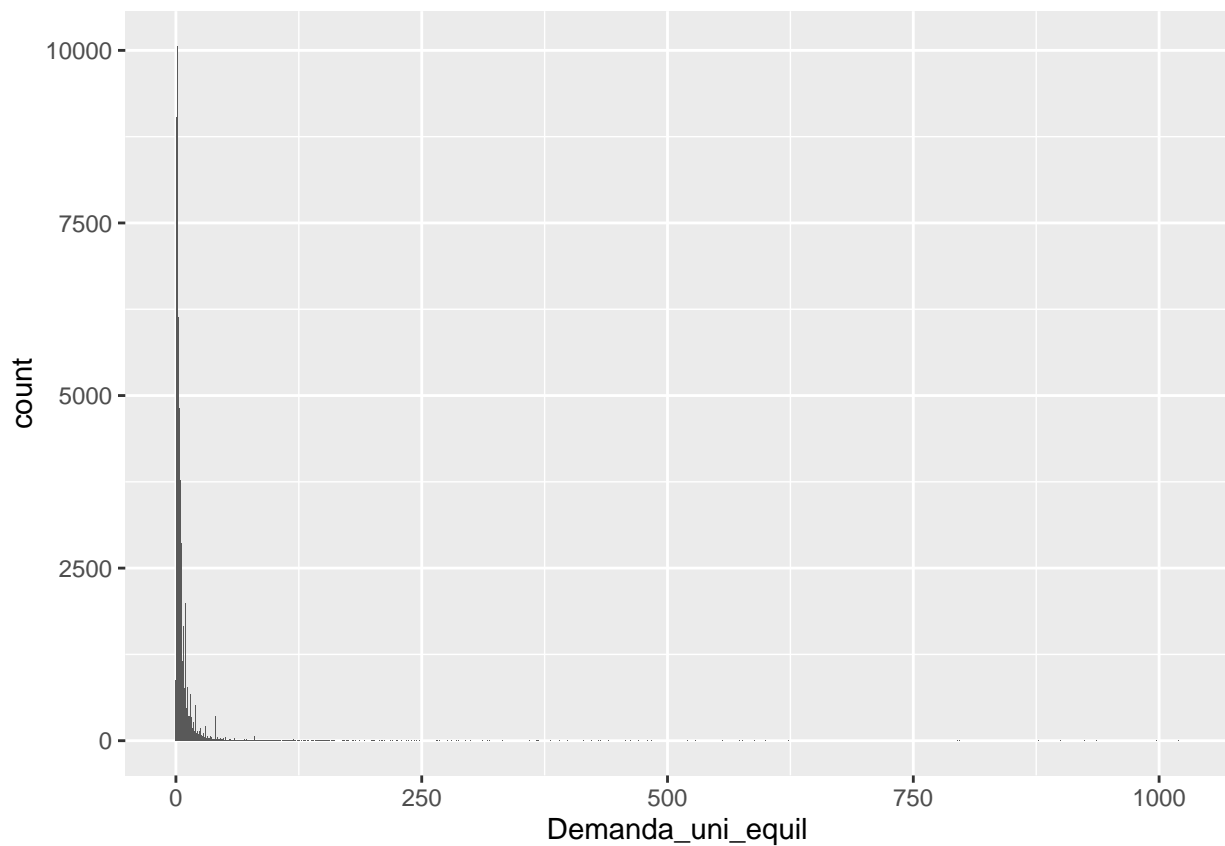
```
##   Semana Agencia_ID Canal_ID Ruta_SAK Cliente_ID Producto_ID Demanda_uni_equil
##   <dbl>      <dbl>    <dbl>   <dbl>    <dbl>      <dbl>      <dbl>
## 1     3         1110        7    3301    15766      1212         3
## 2     3         1110        7    3301    15766      1216         4
## 3     3         1110        7    3301    15766      1238         4
## 4     3         1110        7    3301    15766      1240         4
## 5     3         1110        7    3301    15766      1242         3
## 6     3         1110        7    3301    15766      1250         5
```

```
# Sub sampling
```

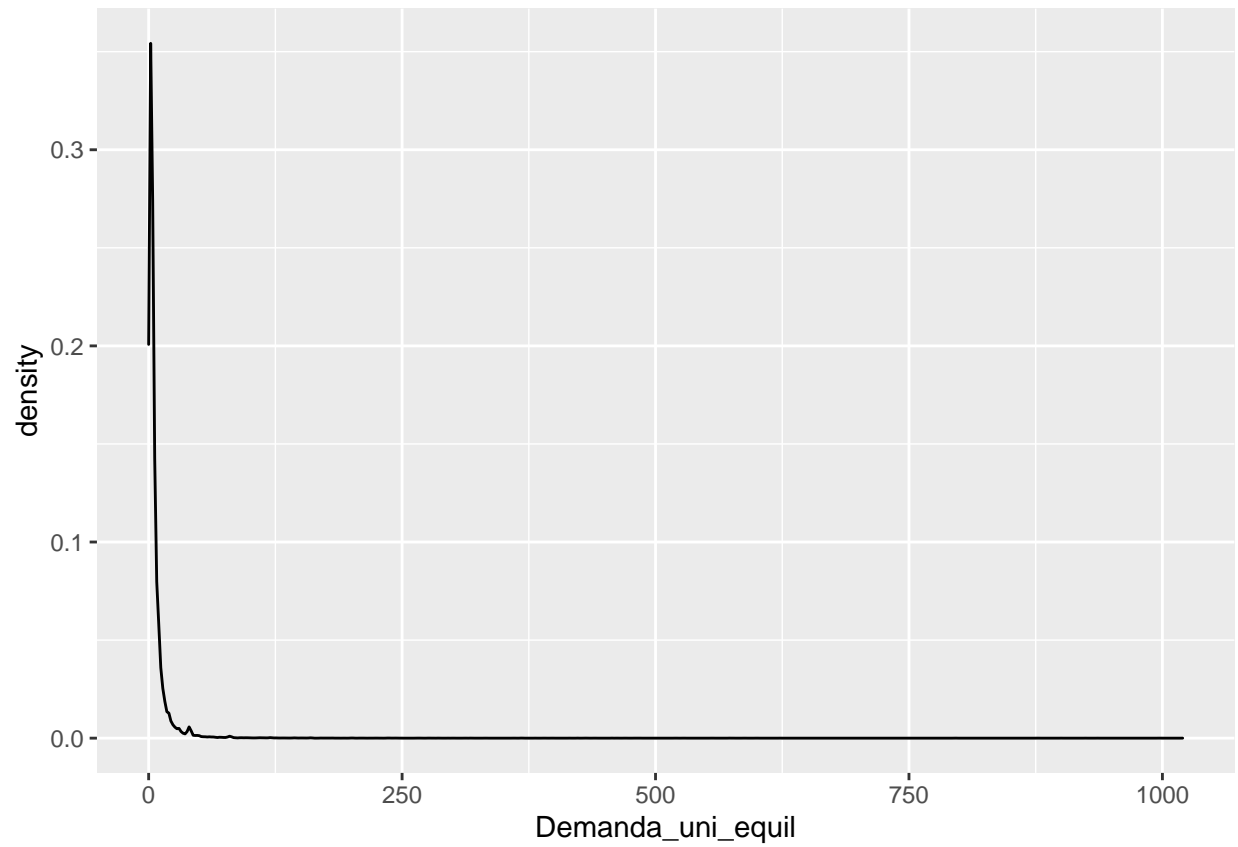
```
df_sample <- df %>% sample_n(50000)
```

```
# Analisando distribuição da variável target
```

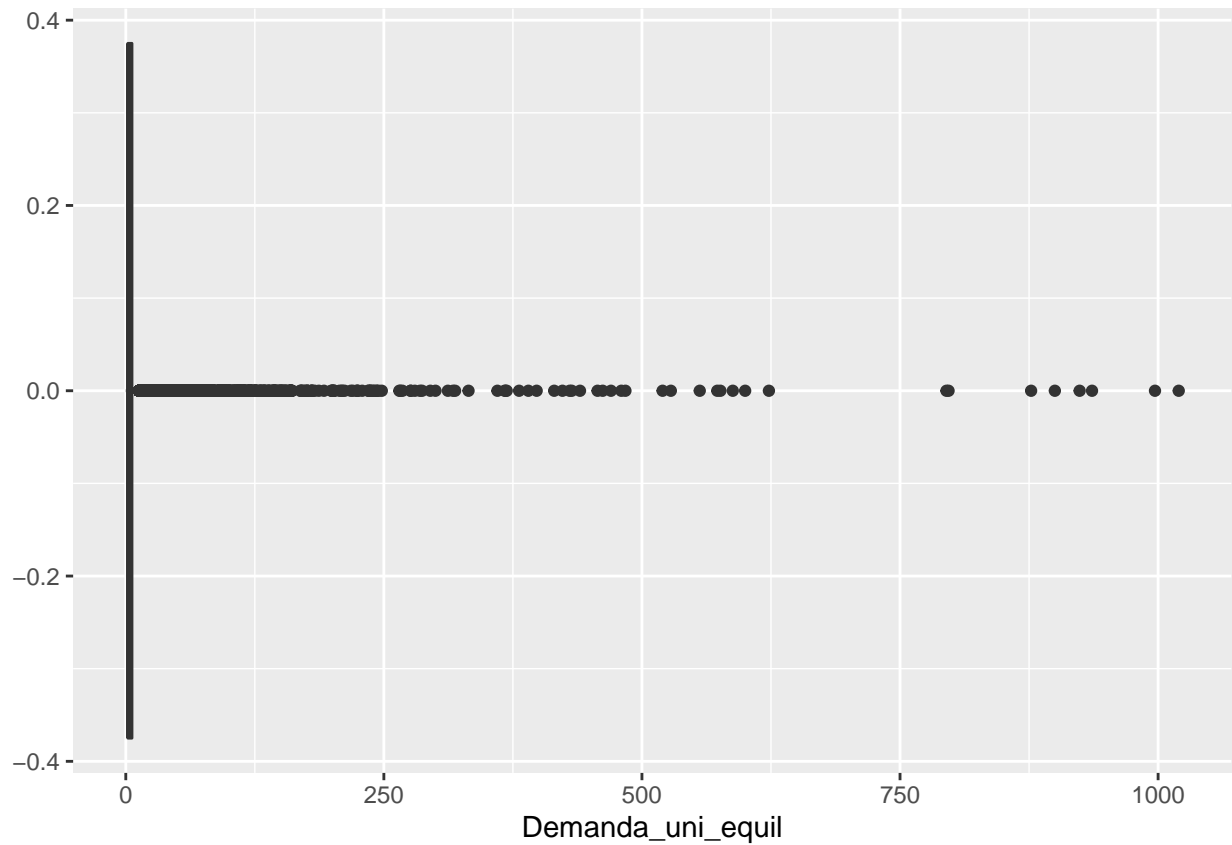
```
ggplot(df_sample, aes(x=Demanda_uni_equil)) +
  geom_bar()
```



```
ggplot(df_sample, aes(x=Demanda_uni_equil)) +
  geom_density()
```

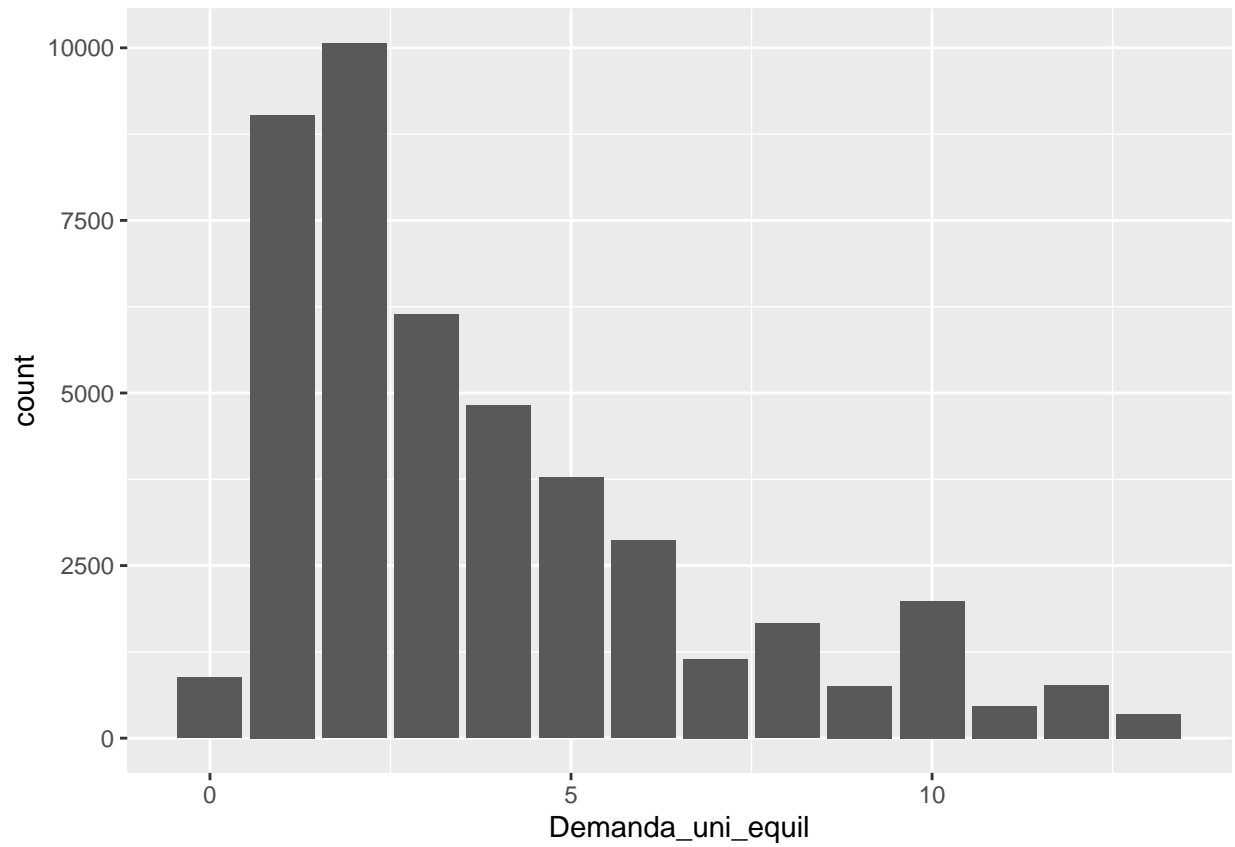


```
ggplot(df_sample,aes(x=Demanda_uni_equil)) +  
  geom_boxplot()
```

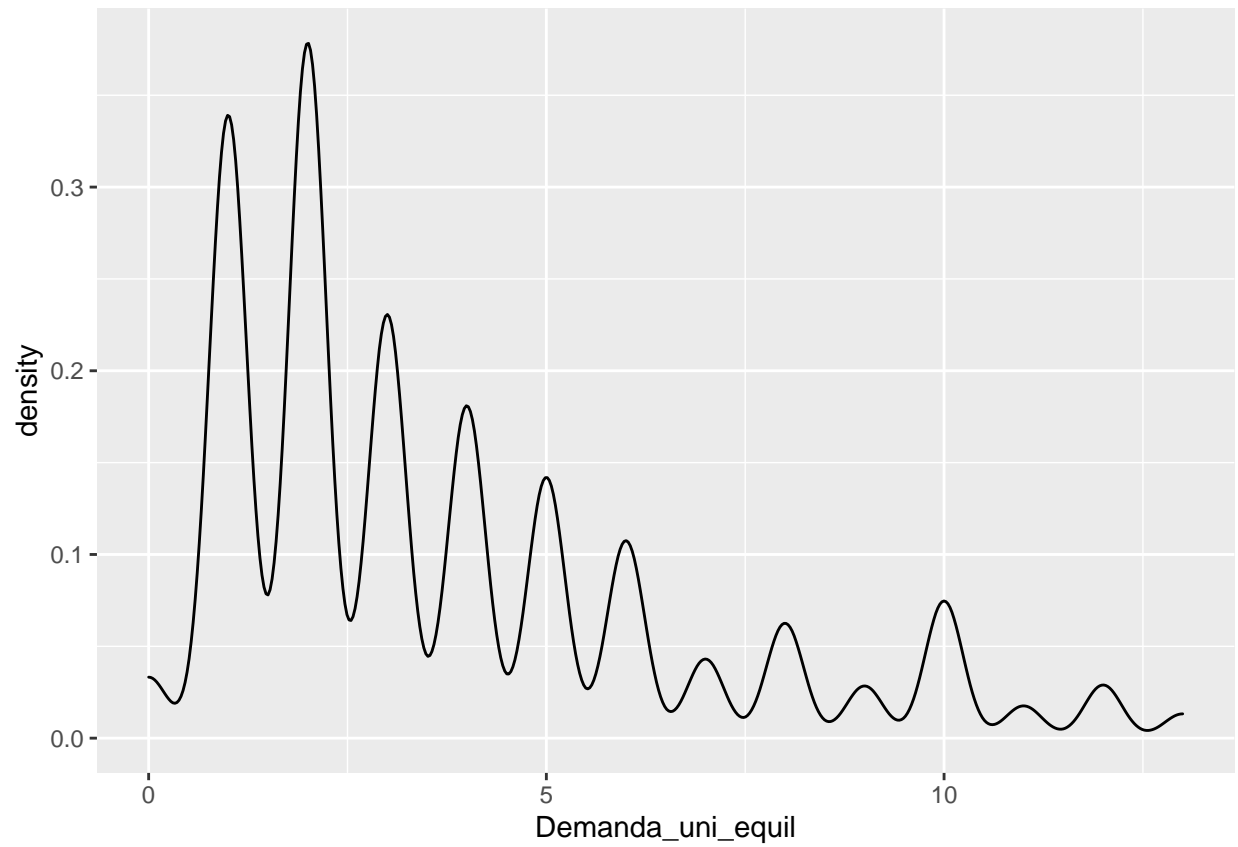


```
# Seleciona apenas os dados até o percentil 90
df_2 <- df_sample %>% filter(Demanda_uni_equil < quantile(df_sample$Demanda_uni_equil,0.9))

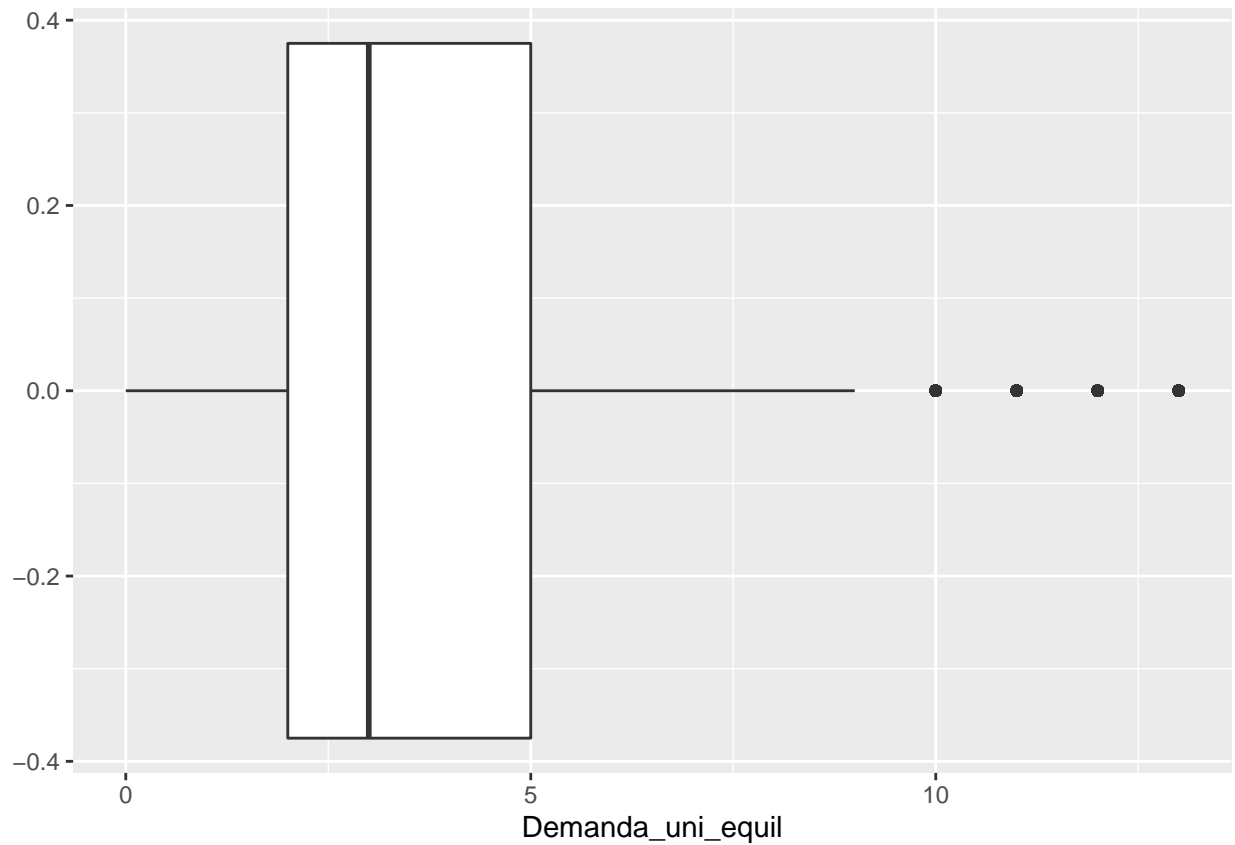
# Analisando novamente a distribuição da variável
ggplot(df_2,aes(x=Demanda_uni_equil)) +
  geom_bar()
```



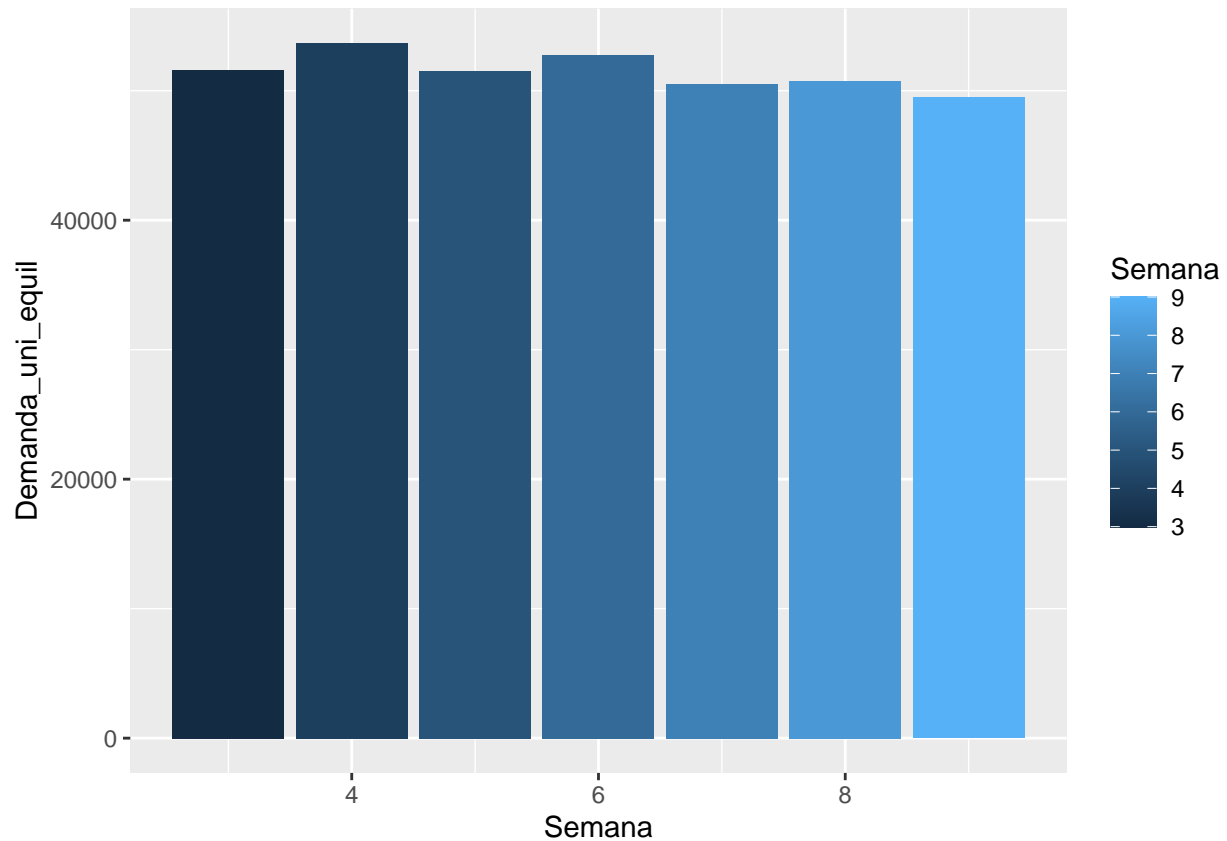
```
ggplot(df_2,aes(x=Demanda_uni_equil)) +  
  geom_density()
```

```
ggplot(df_2,aes(x=Demanda_uni_equil)) +  
  geom_boxplot()
```



```
# ----- Análise exploratória -----  
  
# Demandas por semana  
ggplot(df_sample,aes(fill=Semana,x=Semana, y=Demanda_uni_equil)) +  
  geom_bar(stat='identity')
```



```
# Clientes com mais demanda
```

```
df_demanda_por_cliente <- df_2 %>%
  group_by(Cliente_ID) %>% summarise(Total = sum(Demanda_uni_equil)) %>%
  arrange(desc(Total)) %>% top_n(10)
```

```
## Selecting by Total
```

```
head(df_demanda_por_cliente)
```

```
## # A tibble: 6 x 2
##   Cliente_ID Total
##       <dbl> <dbl>
## 1      653378    39
## 2     1584225    31
## 3      695724    30
## 4      319811    29
## 5     1249858    26
## 6      426566    25
```

```
df_demanda_por_cliente <- merge(df_demanda_por_cliente, df_cliente, by='Cliente_ID')
```

```
df_demanda_por_cliente <- df_demanda_por_cliente %>% arrange(desc(Total))
```

```
df_demanda_por_cliente$NombraclienteId <- str_c(df_demanda_por_cliente$NombreCliente, '-', df_demanda_por_cliente$Total)
```

```
head(df_demanda_por_cliente)
```

```
## Cliente_ID Total NombreCliente NombraclienteId
## 1 653378 39 PUEBLA REMISION PUEBLA REMISION-653378
## 2 1584225 31 VAZQUEZ VAZQUEZ-1584225
## 3 695724 30 MARY MARY-695724
## 4 319811 29 HUICHO HUICHO-319811
## 5 1249858 26 DANY DANY-1249858
## 6 426566 25 EL SOL EL SOL-426566
```

```
df_demanda_por_cliente$NombraclienteId <- factor(df_demanda_por_cliente$NombraclienteId, levels=df_dema
```

```
ggplot(df_demanda_por_cliente,aes(fill=NombreCliente,x=NombraclienteId, y=Total)) +
  geom_bar(stat='identity') +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1), legend.position = "none") +
  scale_fill_hue(l=40, c=35) +
  ggtitle("Demanda por cliente") +
  xlab("Cliente")
```



```
# Productos com mais demanda
```

```
df_demanda_por_producto <- df_2 %>% select(Producto_ID, Demanda_uni_equil) %>%
  group_by(Producto_ID) %>% summarise(Total = sum(Demanda_uni_equil)) %>%
  arrange(desc(Total)) %>% top_n(10)
```

```
## Selecting by Total
```

```
head(df_demanda_por_producto)
```

```
## # A tibble: 6 x 2
```

```
## Producto_ID Total
```

```
##      <dbl> <dbl>
## 1      1240  5667
## 2      1284  5619
## 3      1242  5553
## 4      1250  4678
## 5      2233  4448
## 6      1278  4259
```

```
df_demanda_por_producto <- merge(df_demanda_por_producto, df_producto, by='Producto_ID')
```

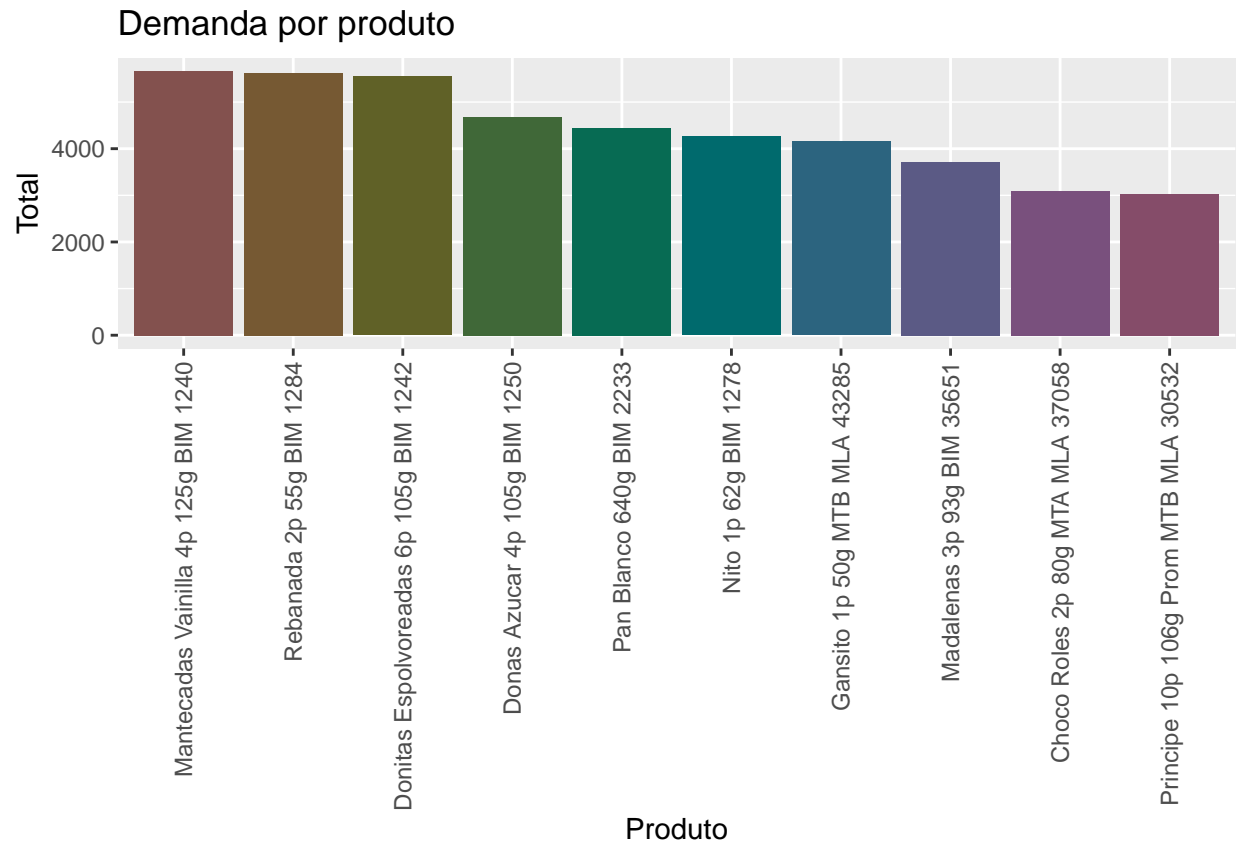
```
df_demanda_por_producto <- df_demanda_por_producto %>% arrange(desc(Total))
```

```
head(df_demanda_por_producto)
```

```
##  Producto_ID Total                               NombreProducto
## 1      1240  5667      Mantecadas Vainilla 4p 125g BIM 1240
## 2      1284  5619                Rebanada 2p 55g BIM 1284
## 3      1242  5553 Donitas Espolvoreadas 6p 105g BIM 1242
## 4      1250  4678                Donas Azucar 4p 105g BIM 1250
## 5      2233  4448                Pan Blanco 640g BIM 2233
## 6      1278  4259                Nito 1p 62g BIM 1278
```

```
df_demanda_por_producto$NombreProducto <- factor(df_demanda_por_producto$NombreProducto, levels=df_demanda_por_producto$NombreProducto)
```

```
ggplot(df_demanda_por_producto, aes(fill=NombreProducto, x=NombreProducto, y=Total)) +
  geom_bar(stat='identity') +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1), legend.position = "none") +
  scale_fill_hue(l=40, c=35) +
  ggtitle("Demanda por producto") +
  xlab("Producto")
```



Cidades com mais demanda

```
df_demanda_por_cidade <- df_2 %>% select(Agencia_ID, Demanda_uni_equil) %>%
  group_by(Agencia_ID) %>% summarise(Total = sum(Demanda_uni_equil)) %>%
  arrange(desc(Total)) %>% top_n(10)
```

Selecting by Total

```
head(df_demanda_por_cidade)
```

```
## # A tibble: 6 x 2
##   Agencia_ID Total
##   <dbl> <dbl>
## 1     1911  1985
## 2     2013  1701
## 3     1126  1535
## 4     1123  1512
## 5     1220  1431
## 6     1117  1414
```

```
df_demanda_por_cidade <- merge(df_demanda_por_cidade, df_town, by='Agencia_ID')
```

```
df_demanda_por_cidade <- df_demanda_por_cidade %>% arrange(desc(Total))
```

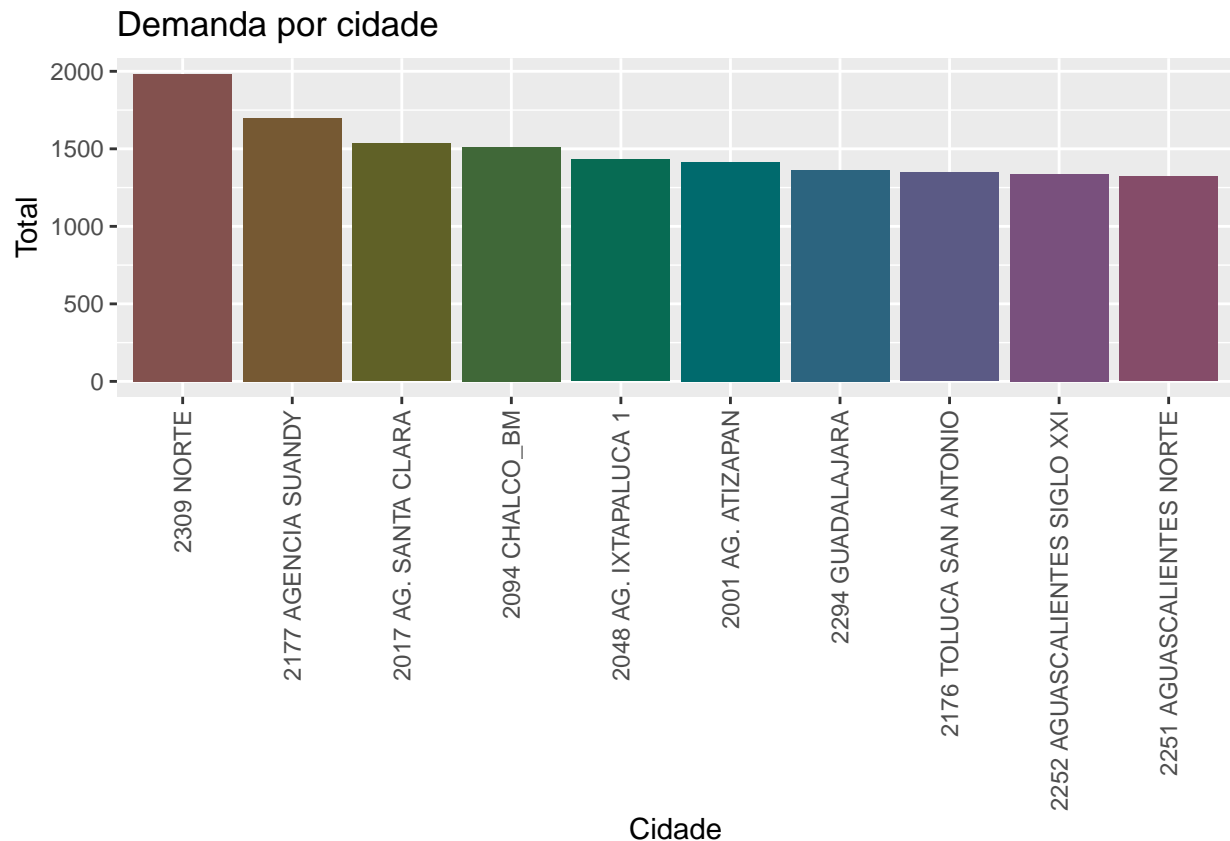
```
head(df_demanda_por_cidade)
```

```
##   Agencia_ID Total      Town      State
## 1     1911  1985    2309 NORTE    JALISCO
## 2     2013  1701    2177 AGENCIA SUANDY ESTADO DE MÉXICO
```

```
## 3      1126 1535 2017 AG. SANTA CLARA ESTADO DE MÉXICO
## 4      1123 1512      2094 CHALCO_BM ESTADO DE MÉXICO
## 5      1220 1431 2048 AG. IXTAPALUCA 1 ESTADO DE MÉXICO
## 6      1117 1414      2001 AG. ATIZAPAN ESTADO DE MÉXICO
```

```
df_demanda_por_cidade$Town <- factor(df_demanda_por_cidade$Town, levels=df_demanda_por_cidade$Town)
```

```
ggplot(df_demanda_por_cidade,aes(fill=Town,x=Town, y=Total)) +
  geom_bar(stat='identity') +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1), legend.position = "none") +
  scale_fill_hue(l=40, c=35) +
  ggtitle("Demanda por cidade") +
  xlab("Cidade")
```



```
# Estados com mais demanda
```

```
df_demanda_por_estado <- df_2 %>% select(Agencia_ID, Demanda_uni_equil)
```

```
head(df_demanda_por_estado)
```

```
## # A tibble: 6 x 2
##   Agencia_ID Demanda_uni_equil
##   <dbl>         <dbl>
## 1     25699             1
## 2      1238             2
## 3      1614             3
## 4      1955             1
## 5      1122             3
```

```
## 6      1235      1
df_demanda_por_estado <- merge(df_demanda_por_estado, df_town, by='Agencia_ID')

head(df_demanda_por_estado)

##   Agencia_ID Demanda_uni_equil      Town      State
## 1      1110      0 2008 AG. LAGO FILT MÉXICO, D.F.
## 2      1110     13 2008 AG. LAGO FILT MÉXICO, D.F.
## 3      1110     12 2008 AG. LAGO FILT MÉXICO, D.F.
## 4      1110      2 2008 AG. LAGO FILT MÉXICO, D.F.
## 5      1110      2 2008 AG. LAGO FILT MÉXICO, D.F.
## 6      1110      8 2008 AG. LAGO FILT MÉXICO, D.F.

df_demanda_por_estado <- df_demanda_por_estado %>%
  group_by(State) %>% summarise(Total = sum(Demanda_uni_equil)) %>%
  arrange(desc(Total)) %>% top_n(10)

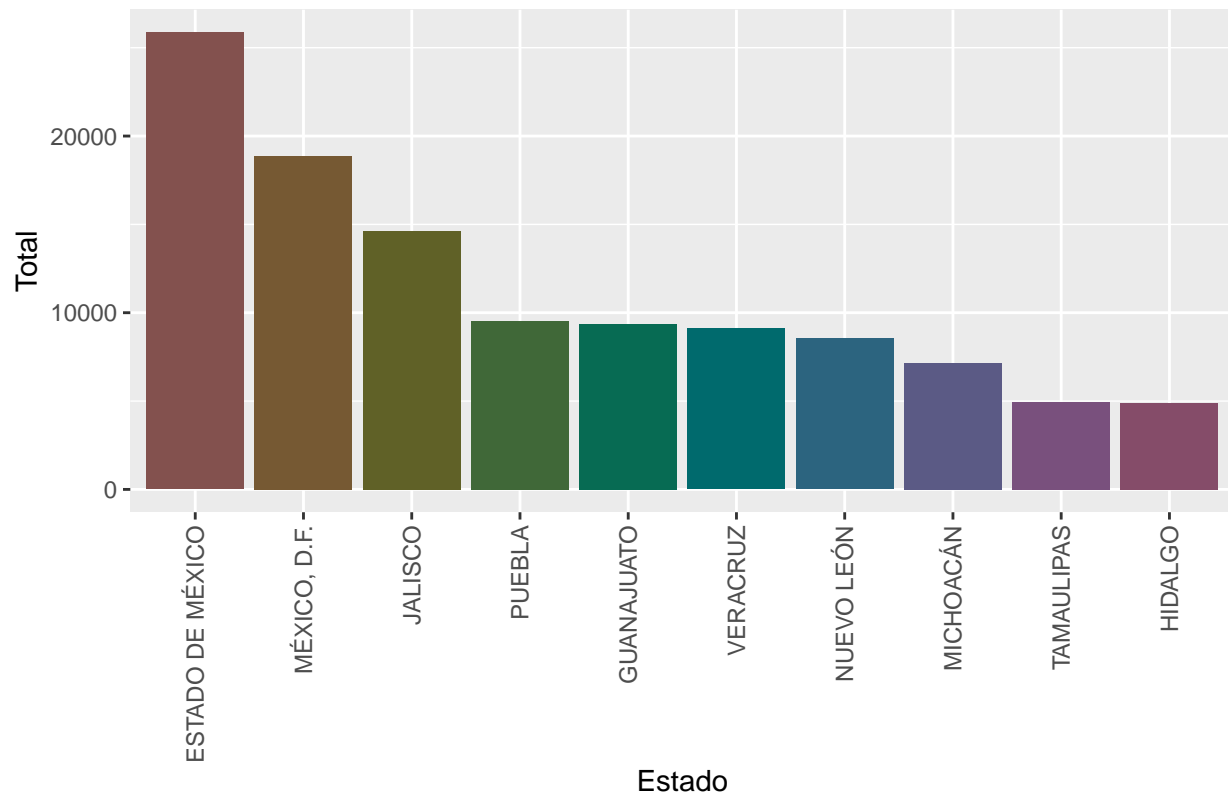
## Selecting by Total
head(df_demanda_por_estado)

## # A tibble: 6 x 2
##   State      Total
##   <chr>      <dbl>
## 1 ESTADO DE MÉXICO 25856
## 2 MÉXICO, D.F.    18894
## 3 JALISCO        14645
## 4 PUEBLA         9546
## 5 GUANAJUATO     9341
## 6 VERACRUZ       9102

df_demanda_por_estado$State <- factor(df_demanda_por_estado$State, levels=df_demanda_por_estado$State)

ggplot(df_demanda_por_estado, aes(fill=State, x=State, y=Total)) +
  geom_bar(stat='identity') +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1), legend.position = "none") +
  scale_fill_hue(l=40, c=35) +
  ggtitle("Demanda por estado") +
  xlab("Estado")
```


Demanda por estado



```
# ----- Dividindo em dados de treino e de teste -----
amostra <- createDataPartition(df_2$Demanda_uni_equil,p=0.7,list=F)
testData <- df_2[amostra,]
trainData <- df_2[-amostra,]
```

Modelos

```
# Criando um modelo LM com todas variáveis
modelo_lm_v1 <- lm(Demanda_uni_equil ~ .,data=trainData)
```

```
summary(modelo_lm_v1)
```

```
##
## Call:
## lm(formula = Demanda_uni_equil ~ ., data = trainData)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.9138 -2.0684 -0.8317  1.3804  9.7430
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  3.319e+00  9.314e-02  35.637 < 2e-16 ***
## Semana      -9.865e-03  1.245e-02  -0.793   0.428
## Agencia_ID   6.624e-06  6.031e-06   1.098   0.272
## Canal_ID     1.854e-01  2.184e-02   8.488 < 2e-16 ***
```

```
## Ruta_SAK      1.319e-04  2.065e-05   6.386 1.76e-10 ***
## Cliente_ID   -6.983e-08  1.375e-08  -5.077 3.88e-07 ***
## Producto_ID  5.843e-06  1.410e-06   4.144 3.43e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.888 on 13403 degrees of freedom
## Multiple R-squared:  0.01984,    Adjusted R-squared:  0.0194
## F-statistic: 45.22 on 6 and 13403 DF,  p-value: < 2.2e-16

predictions <- predict(modelo_lm_v1,testData)

# rmsle: Root Mean Square Logarithmic Error
# 0.599
rmsle(predictions, testData$Demanda_uni_equil)

## [1] 0.5977337

# Modelo 2 com variáveis mais significativas
modelo_lm_v2 <- lm(Demanda_uni_equil ~ Canal_ID + Ruta_SAK + Cliente_ID + Producto_ID, data=trainData)

summary(modelo_lm_v2)

##
## Call:
## lm(formula = Demanda_uni_equil ~ Canal_ID + Ruta_SAK + Cliente_ID +
##      Producto_ID, data = trainData)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.9416 -2.0646 -0.8283  1.3845  9.7061
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  3.278e+00  5.317e-02  61.652 < 2e-16 ***
## Canal_ID     1.844e-01  2.183e-02   8.447 < 2e-16 ***
## Ruta_SAK     1.322e-04  2.065e-05   6.401 1.60e-10 ***
## Cliente_ID  -6.976e-08  1.375e-08  -5.072 3.98e-07 ***
## Producto_ID  5.852e-06  1.410e-06   4.150 3.34e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.888 on 13405 degrees of freedom
## Multiple R-squared:  0.01971,    Adjusted R-squared:  0.01942
## F-statistic: 67.37 on 4 and 13405 DF,  p-value: < 2.2e-16

predictions_2 <- predict(modelo_lm_v2,testData)

# RMSLE = 0.599 -> igual anterior
rmsle(predictions_2, testData$Demanda_uni_equil)

## [1] 0.5977544

# KNN
modelo_knn_v1 <- knnreg(Demanda_uni_equil ~ ., data=trainData)

summary(modelo_knn_v1)
```

```
##          Length Class  Mode
## learn    2      -none- list
## k         1      -none- numeric
## terms    3      terms  call
## xlevels   0      -none- list
## theDots   0      -none- list
```

```
predictions_3 <- predict(modelo_knn_v1, testData)
```

```
# 0.6526 -> Pior que os outros
rmsle(predictions_3, testData$Demanda_uni_equil)
```

```
## [1] 0.6424282
```

```
# SVR
modelo_svr_v1 <- svm(Demanda_uni_equil ~ ., data=trainData)

summary(modelo_svr_v1)
```

```
##
## Call:
## svm(formula = Demanda_uni_equil ~ ., data = trainData)
##
##
## Parameters:
##   SVM-Type:  eps-regression
##   SVM-Kernel: radial
##       cost:  1
##       gamma: 0.1666667
##   epsilon:  0.1
##
##
## Number of Support Vectors: 12496
```

```
predictions_4 <- predict(modelo_svr_v1, testData)
```

```
# 0.577 -> Melhor resultado até o momento
rmsle(predictions_4, testData$Demanda_uni_equil)
```

```
## [1] 0.5774893
```

```
# SVR - v2
modelo_svr_v2 <- svm(Demanda_uni_equil ~ Canal_ID + Ruta_SAK + Cliente_ID + Producto_ID, data=trainData)

summary(modelo_svr_v2)
```

```
##
## Call:
## svm(formula = Demanda_uni_equil ~ Canal_ID + Ruta_SAK + Cliente_ID +
##   Producto_ID, data = trainData)
##
##
## Parameters:
##   SVM-Type:  eps-regression
##   SVM-Kernel: radial
```

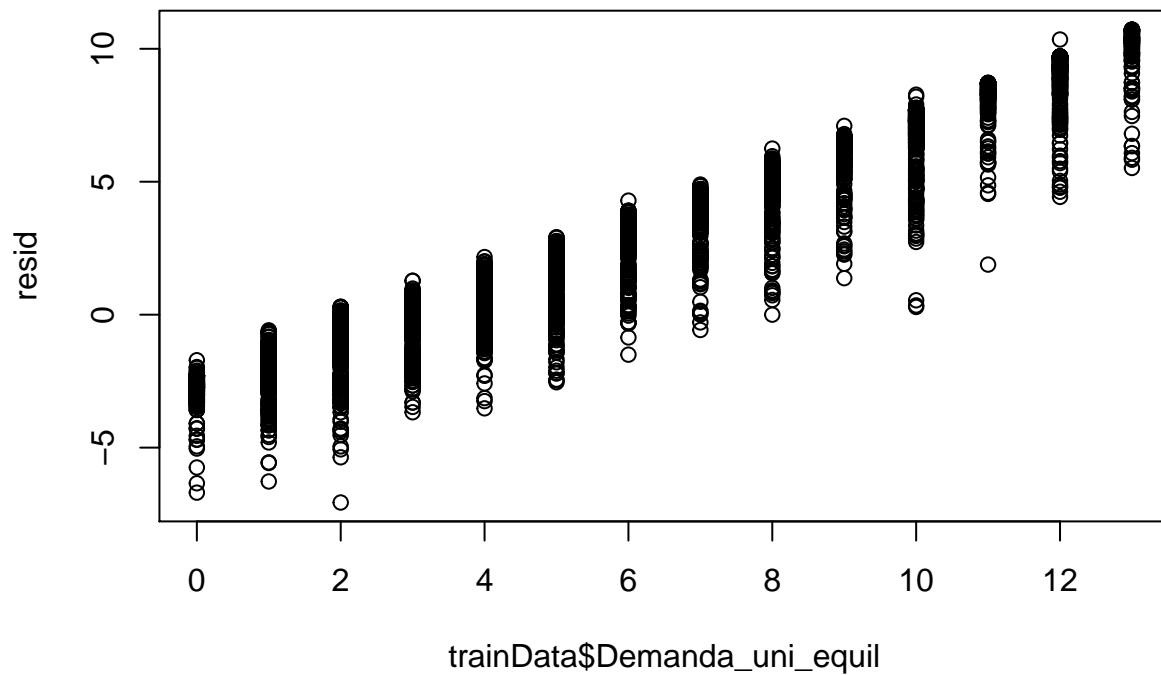
```
##      cost: 1
##      gamma: 0.25
##      epsilon: 0.1
##
## Number of Support Vectors: 12526
predictions_5 <- predict(modelo_svr_v2, testData)

# 0.576 -> Um pouco melhor que o anterior
rmsle(predictions_5, testData$Demanda_uni_equil)

## [1] 0.57646
```

Avaliação

```
# Plot Residuals
resid <- resid(modelo_svr_v2)
plot(trainData$Demanda_uni_equil, resid)
```



```
plot(y=predictions_5, x=testData$Demanda_uni_equil)
```

