

Projeto2 - Previsão de demanda Grupo Bimbo

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Carregando pacotes necessários para a avaliação

```
library(data.table)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:data.table':
##
##   between, first, last

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(DataExplorer)
library(ggplot2)
library(biglm)

## Loading required package: DBI

library(Metrics)
```

Carregando e limpando os dados

De acordo com a descrição dos dados as variáveis “Venta_uni_hoy” e “Venta_hoy” representam a mesma informação porém em unidades diferentes, o mesmo ocorre com as variáveis “Dev_uni_proxima” e “Dev_proxima”. Dessa forma para essa análise deve ser retirada uma das variáveis que representam a mesma informação. Então serão retiradas as variáveis “Venta_hoy” e “Dev_proxima”. A variável “Ruta_SAK” corresponde a forma como o produto chegou até o cliente dessa forma não será relevante para a previsão de demanda.

```
train <- fread(file = 'train.csv', header = TRUE, verbose = TRUE, data.table = FALSE)

##   omp_get_num_procs()           8
##   R_DATATABLE_NUM_PROCS_PERCENT unset (default 50)
##   R_DATATABLE_NUM_THREADS      unset
##   omp_get_thread_limit()       2147483647
##   omp_get_max_threads()        8
##   OMP_THREAD_LIMIT             unset
##   OMP_NUM_THREADS              unset
##   RestoreAfterFork             true
```

```

## data.table is using 4 threads. See ?setDTthreads.
## Input contains no \n. Taking this to be a filename to open
## [01] Check arguments
## Using 4 threads (omp_get_max_threads()=8, nth=4)
## NAstrings = [<<NA>>]
## None of the NAstrings look like numbers.
## show progress = 0
## 0/1 column will be read as integer
## [02] Opening the file
## Opening file train.csv
## File opened, size = 2.980GB (3199358223 bytes).
## Memory mapped ok
## [03] Detect and skip BOM
## [04] Arrange mmap to be \0 terminated
## \n has been found in the input and different lines can end with different line endings (e.g. mixed)
## [05] Skipping initial rows if needed
## Positioned on line 1 starting: <<Semana,Agencia_ID,Canal_ID,Rut>>
## [06] Detect separator, quoting rule, and ncolumns
## Detecting sep automatically ...
## sep=', ' with 100 lines of 11 fields using quote rule 0
## Detected 11 columns on line 1. This line is either column names or first data row. Line starts as:
## Quote rule picked = 0
## fill=false and the most number of columns found is 11
## [07] Detect column types, good nrow estimate and whether first row is column names
## 'header' changed by user from 'auto' to true
## Number of sampling jump points = 100 because (3199358222 bytes from row 1 to eof) / (2 * 4296 jump points)
## Type codes (jump 000) : 55555557575 Quote rule 0
## Type codes (jump 100) : 55555557575 Quote rule 0
## =====
## Sampled 10047 rows (handled \n inside quoted fields) at 101 jump points
## Bytes from first data row on line 2 to the end of last row: 3199358093
## Line length: mean=43.10 sd=1.46 min=39 max=60
## Estimated number of rows: 3199358093 / 43.10 = 74227396
## Initial alloc = 81650135 rows (74227396 + 9%) using bytes/max(mean-2*sd,min) clamped between [1.1*min, 1.1*max]
## =====
## [08] Assign column names
## [09] Apply user overrides on column types
## After 0 type and 0 drop user overrides : 55555557575
## [10] Allocate memory for the datatable
## Allocating 11 column slots (11 - 0 dropped) with 81650135 rows
## [11] Read the data
## jumps=[0..3052), chunk_size=1048282, total_size=3199358093
## Read 74180464 rows x 11 columns from 2.980GB (3199358223 bytes) file in 01:22.649 wall clock time
## [12] Finalizing the datatable
## Type counts:
## 9 : int32 '5'
## 2 : float64 '7'
## =====
## 0.001s ( 0%) Memory map 2.980GB file
## 1.051s ( 1%) sep=', ' ncol=11 and header detection
## 0.000s ( 0%) Column type detection using 10047 sample rows
## 0.450s ( 1%) Allocation of 81650135 rows x 11 cols (3.954GB) of which 74180464 ( 91%) rows used
## 81.147s (98%) Reading 3052 chunks (0 swept) of 1.000MB (each chunk 24305 rows) using 4 threads
## + 75.088s ( 91%) Parse to row-major thread buffers (grown 0 times)

```

```
## + 1.214s ( 1%) Transpose
## + 4.845s ( 6%) Waiting
## 0.000s ( 0%) Rereading 0 columns due to out-of-sample type exceptions
## 82.649s Total
```

```
head(train)
```

```
##   Semana Agencia_ID Canal_ID Ruta_SAK Cliente_ID Producto_ID Venta_uni_hoy
## 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
##   Venta_hoy Dev_uni_proxima Dev_proxima Demanda_uni_equil
## 1    25.14      0      0      3
## 2    33.52      0      0      4
## 3    39.32      0      0      4
## 4    33.52      0      0      4
## 5    22.92      0      0      3
## 6    38.20      0      0      5
```

```
train$Venta_hoy <- NULL
train$Dev_proxima <- NULL
train$Ruta_SAK <- NULL
```

Dividindo o objeto “train” pelas semanas

O objeto “train” é muito grande sendo oneroso trabalhar com ele, então o objeto foi dividido pelas semanas, sendo da semana 3 a semana 8, para treino do modelo de machine learning e a semana 9 sendo para teste e a validação do modelo.

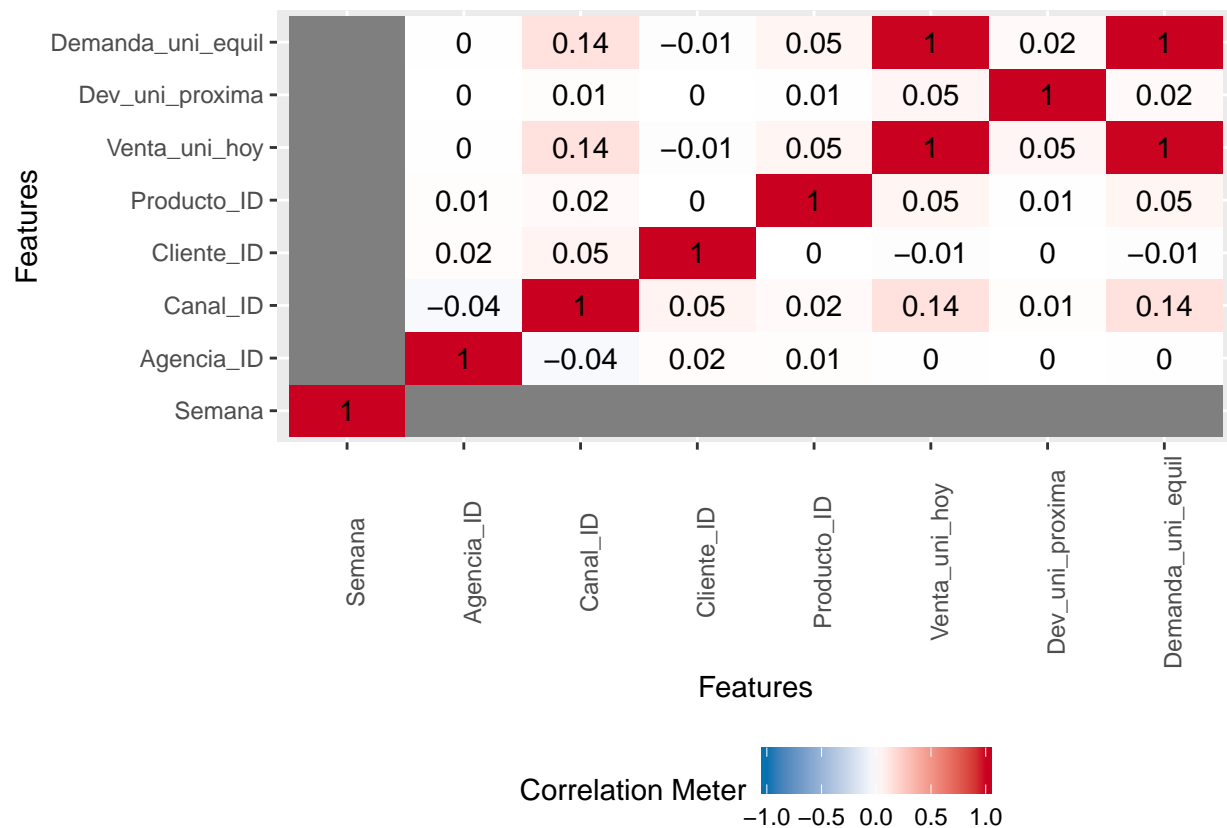
```
train_sample3 <- dplyr::filter(train, train$Semana == 3)
```

Análise de correlação

Algumas variáveis apresentam uma correlação mais forte que outras, ou seja, algumas irão apresentar uma maior relevância para a previsão de demanda. A semana 3 foi utilizada como amostra para representar todo restante.

```
plot_correlation(data = train_sample3, type = 'all')
```

```
## Warning in cor(x = structure(list(Semana = c(3L, 3L, 3L, 3L, 3L, 3L, 3L, : the
## standard deviation is zero
## Warning: Removed 14 rows containing missing values (geom_text).
```



Como observado através do gráfico de correlação, as variáveis “Agencia_ID” e “Cliente_ID” não serão representativas podendo ser retiradas.

```
train_sample3$Agencia_ID <- NULL
train_sample3$Cliente_ID <- NULL
train$Agencia_ID <- NULL
train$Cliente_ID <- NULL

train_sample3 <- dplyr::filter(train, train$Semana == 3)
train_sample4 <- dplyr::filter(train, train$Semana == 4)
train_sample5 <- dplyr::filter(train, train$Semana == 5)
train_sample6 <- dplyr::filter(train, train$Semana == 6)
train_sample7 <- dplyr::filter(train, train$Semana == 7)
train_sample8 <- dplyr::filter(train, train$Semana == 8)
test <- dplyr::filter(train, train$Semana == 9)
rm(train)
```

Análise exploratória

A análise exploratória foi realizada para cada objeto, ou seja, por semana.

```
table(train_sample3$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7
## 10815787 174159 71531 30105 17844 13715 7966 4145
##      8      9     10     11     12     13     14     15
##   4223   2652   4811   1560   1990   1058   1034   1436
```

##	16	17	18	19	20	21	22	23
##	767	622	723	427	1113	411	349	289
##	24	25	26	27	28	29	30	31
##	449	354	225	248	248	147	452	149
##	32	33	34	35	36	37	38	39
##	150	117	123	144	139	106	87	85
##	40	41	42	43	44	45	46	47
##	756	67	187	73	63	78	61	50
##	48	49	50	51	52	53	54	55
##	85	31	115	46	51	39	53	36
##	56	57	58	59	60	61	62	63
##	37	13	25	32	72	28	31	32
##	64	65	66	67	68	69	70	71
##	36	33	28	11	25	22	37	10
##	72	73	74	75	76	77	78	79
##	22	17	11	19	17	15	12	12
##	80	81	82	83	84	85	86	87
##	117	12	13	10	24	11	14	9
##	88	89	90	91	92	94	95	96
##	7	12	29	6	8	14	9	16
##	97	98	99	100	101	102	103	104
##	5	13	7	23	10	9	2	9
##	105	106	107	108	109	110	111	112
##	5	6	1	8	3	15	4	8
##	113	114	115	117	118	119	120	121
##	2	7	10	4	1	6	46	2
##	122	123	124	125	126	127	128	129
##	5	5	3	1	5	4	6	2
##	130	131	132	133	134	135	136	137
##	3	1	8	2	5	8	4	1
##	138	139	140	141	142	143	144	145
##	3	4	5	5	4	2	6	2
##	148	149	150	152	153	154	155	156
##	2	3	12	3	2	2	1	1
##	157	158	160	161	165	166	167	168
##	1	3	35	4	4	1	1	3
##	169	171	172	175	176	177	178	179
##	1	1	1	1	1	1	1	1
##	180	181	183	184	185	186	187	188
##	6	2	1	4	2	3	4	1
##	190	191	192	195	196	198	199	200
##	3	1	2	2	1	2	1	12
##	202	207	208	209	210	211	212	214
##	1	1	1	1	2	1	2	1
##	216	219	220	222	225	226	228	229
##	1	1	2	3	2	1	1	1
##	230	231	234	235	237	238	239	240
##	1	1	1	1	2	1	1	3
##	241	243	250	252	253	256	258	260
##	1	1	1	2	1	2	1	3
##	264	265	268	273	274	280	281	284
##	1	1	3	1	1	6	1	1
##	285	288	291	295	300	310	316	320
##	1	1	1	1	3	2	1	3

```
##      323      324      326      330      336      341      345      346
##        1        1        1        2        1        1        1        3
##      348      353      360      370      376      384      392      400
##        1        1        4        1        1        1        1        1
##      404      410      440      470      474      500      509      520
##        1        1        1        1        1        1        1        2
##      555      570      576      587      591      600      608      660
##        1        1        1        1        1        2        1        1
##      672      697      705      750      778      800      807      900
##        1        1        1        1        1        1        1        1
##      904     1000     1008     1100     1160     1230     1250     1258
##        1        1        1        1        1        1        1        1
##     1370     1630     2400     2418     4104     16345
##        1        1        1        1        1        1
```

```
summary(train_sample3$Dev_uni_proxima)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
##    0.000    0.000    0.000    0.115    0.000 16345.000
```

```
quantile(train_sample3$Dev_uni_proxima, .999)
```

```
## 99.9%
##      16
```

```
any(is.na(train_sample3))
```

```
## [1] FALSE
```

```
any(train_sample3$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

```
table(train_sample4$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7
## 10668140 167414 69439 29631 17526 13225 8150 4201
##      8      9     10     11     12     13     14     15
##  4349  2827  5088  1652  2073  1027  1099 1587
##     16     17     18     19     20     21     22     23
##   864   589   684   487  1194   429   371   300
##     24     25     26     27     28     29     30     31
##   437   383   258   270   254   188   467   138
##     32     33     34     35     36     37     38     39
##   180   166   137   163   149    88   103    88
##     40     41     42     43     44     45     46     47
##   964    75   171    82    81    92    57    79
##     48     49     50     51     52     53     54     55
##   109    38   120    51    63    46    52    45
##     56     57     58     59     60     61     62     63
##     45     31     33     32     87     18     30     27
##     64     65     66     67     68     69     70     71
##     42     26     34     25     37     18     32     21
##     72     73     74     75     76     77     78     79
##     33     18     17     22     19     16     25     18
##     80     81     82     83     84     85     86     87
##    174     20      6      7     23     19     10     13
```

##	88	89	90	91	92	93	94	95
##	9	7	21	10	6	8	15	11
##	96	97	98	99	100	101	102	103
##	11	9	1	6	37	4	13	4
##	104	105	106	107	108	109	110	111
##	5	5	9	9	10	5	16	4
##	112	113	114	115	116	117	118	119
##	5	4	4	6	4	3	4	6
##	120	121	122	123	124	125	126	127
##	56	5	5	5	2	6	6	2
##	128	129	130	131	132	133	134	135
##	5	2	8	1	4	3	4	5
##	136	137	138	140	141	142	143	144
##	1	6	3	7	5	3	4	6
##	145	146	147	148	149	150	151	152
##	1	1	3	5	1	7	1	2
##	153	154	155	156	158	159	160	162
##	2	1	3	3	1	2	20	3
##	163	164	165	166	167	168	170	171
##	1	2	3	1	1	6	2	1
##	174	175	176	177	179	180	181	182
##	1	2	2	1	2	2	1	3
##	184	186	187	188	189	191	192	193
##	1	3	2	2	1	1	1	2
##	194	195	198	200	205	207	208	210
##	1	1	2	12	4	1	1	2
##	212	214	215	216	217	218	219	220
##	1	1	1	2	1	1	1	3
##	223	225	228	230	232	233	237	239
##	1	2	2	1	1	1	1	1
##	240	241	242	244	245	246	250	255
##	4	1	1	1	1	2	4	1
##	258	260	269	270	272	277	279	280
##	3	2	1	2	1	1	1	1
##	282	284	288	291	293	294	300	302
##	2	1	1	1	1	1	2	1
##	303	310	316	320	322	324	325	332
##	1	1	1	4	1	3	1	1
##	345	355	360	362	371	380	398	400
##	1	1	3	2	1	1	1	1
##	402	416	420	450	452	456	485	489
##	1	1	1	1	1	1	1	1
##	496	500	510	520	534	551	603	619
##	1	1	1	2	1	1	1	1
##	648	656	667	709	713	720	800	807
##	1	1	1	1	1	2	2	1
##	843	845	854	909	1146	1189	1232	1288
##	1	1	1	1	1	1	1	1
##	1371	1550	1588	1853	1973	3838	3960	5030
##	1	1	1	1	1	1	1	1
##	12760							
##	1							

```
summary(train_sample4$Dev_uni_proxima)
```

```
##      Min.   1st Qu.   Median     Mean  3rd Qu.     Max.
##      0.00     0.00     0.00     0.12     0.00  12760.00
```

```
quantile(train_sample4$Dev_uni_proxima, .999)
```

```
## 99.9%
##      17
```

```
any(is.na(train_sample4))
```

```
## [1] FALSE
```

```
any(train_sample4$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

```
table(train_sample5$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7
## 10292311 158652 64464 28226 16774 12749 8007 4030
##      8      9     10     11     12     13     14     15
##  4407   2657   4977   1548   1995   1037   1047  1423
##     16     17     18     19     20     21     22     23
##    833    604    660    486   1148    391    367    273
##     24     25     26     27     28     29     30     31
##    407    342    237    220    239    163    436    131
##     32     33     34     35     36     37     38     39
##    156    135     94    135    167     83     84     70
##     40     41     42     43     44     45     46     47
##    904     75    165     48     71     77     49     63
##     48     49     50     51     52     53     54     55
##     79     43    106     33     50     28     62     32
##     56     57     58     59     60     61     62     63
##     28     25     16     24     73     16     26     26
##     64     65     66     67     68     69     70     71
##     32     39     21     18     19     12     29     19
##     72     73     74     75     76     77     78     79
##     30      8      7     20      9     16     12     15
##     80     81     82     83     84     85     86     87
##    146      8     14      3     23      8      8      8
##     88     89     90     91     92     93     94     95
##      8      7     18      2      8     11      7      7
##     96     97     98     99    100    101    102    103
##      7      8      7      7     27      4      7      7
##    104    105    106    107    108    109    110    111
##      3      8      5      2      9      4     15      2
##    112    113    114    115    116    117    118    119
##      4      3      4      3      3      4      5      2
##    120    121    122    123    124    125    126    127
##     52      5      3      4      3      3      4      3
##    128    129    130    131    132    133    134    135
##      6      2      4      5      6      3      2      5
##    136    137    138    139    140    141    142    143
##      2      3      1      1      6      4      2      2
```



```
##      144      145      147      149      150      151      152      153
##         4         3         1         1         5         2         2         3
##      154      155      156      157      158      159      160      161
##         1         1         3         1         1         2        18         2
##      162      164      165      168      169      170      171      175
##         2         3         1         5         2         3         1         4
##      176      177      178      179      180      183      184      185
##         1         1         4         1         8         1         3         1
##      186      187      188      189      190      191      192      196
##         2         2         2         2         4         1         2         1
##      197      198      200      201      203      204      205      206
##         1         1         8         2         2         1         1         1
##      209      210      211      212      213      218      220      225
##         2         3         3         1         2         1         1         1
##      230      234      235      236      240      241      242      245
##         2         1         2         1         7         1         1         1
##      246      250      270      273      280      282      284      285
##         1         2         1         1         5         2         1         1
##      288      289      297      300      302      303      310      314
##         6         1         1         3         1         1         1         1
##      315      317      319      320      336      342      346      364
##         1         1         3         4         1         1         1         1
##      368      370      377      396      400      406      427      432
##         1         1         1         1         3         1         1         1
##      440      443      450      480      494      497      500      523
##         3         1         1         1         1         1         1         1
##      530      547      642      656      680      720      750      868
##         1         1         1         1         2         3         1         1
##      896      980      1080      1128      1175      1200      1241      1440
##         1         1         1         1         1         1         1         1
##     1740     2100     5400     5795
##         1         1         1         2
```

```
summary(train_sample5$Dev_uni_proxima)
```

```
##      Min.   1st Qu.   Median     Mean  3rd Qu.     Max.
##      0.000    0.000    0.000    0.115    0.000 5795.000
```

```
quantile(train_sample5$Dev_uni_proxima, .999)
```

```
## 99.9%
##      16
```

```
any(is.na(train_sample5))
```

```
## [1] FALSE
```

```
any(train_sample5$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

```
table(train_sample6$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7      8      9
## 9834238 178723 71036 31104 17995 13378 8644 4241 4674 2820
##      10      11      12      13      14      15      16      17      18      19
##   5086   1693   2174   1094   1117   1612   850   640   790   478
```

##	20	21	22	23	24	25	26	27	28	29
##	1181	427	371	329	483	381	301	252	248	185
##	30	31	32	33	34	35	36	37	38	39
##	476	165	184	138	128	135	162	115	115	81
##	40	41	42	43	44	45	46	47	48	49
##	895	66	159	68	87	88	61	75	93	55
##	50	51	52	53	54	55	56	57	58	59
##	118	53	81	44	57	36	44	43	31	28
##	60	61	62	63	64	65	66	67	68	69
##	83	28	25	31	28	28	28	21	24	26
##	70	71	72	73	74	75	76	77	78	79
##	37	20	28	15	16	20	12	18	18	13
##	80	81	82	83	84	85	86	87	88	89
##	172	6	10	10	25	7	11	8	12	3
##	90	91	92	93	94	95	96	97	98	99
##	16	6	4	11	14	12	13	8	4	5
##	100	101	102	103	104	105	106	107	108	109
##	30	6	7	4	17	8	6	5	8	4
##	110	111	112	113	114	115	116	117	118	119
##	12	1	9	5	4	8	7	7	3	6
##	120	121	122	123	124	125	126	127	128	129
##	52	3	2	1	4	3	9	3	4	3
##	130	131	132	133	134	135	137	138	139	140
##	5	2	8	1	2	3	6	2	5	4
##	141	142	144	145	146	147	148	150	151	152
##	3	2	5	1	3	2	3	6	2	4
##	154	155	156	157	158	160	161	162	163	165
##	1	2	2	1	2	16	3	2	1	3
##	166	167	168	169	170	171	172	173	174	175
##	1	4	7	2	5	2	2	1	2	4
##	176	178	180	181	182	184	186	187	188	189
##	2	1	3	2	1	1	2	2	3	1
##	190	192	193	194	195	198	200	202	203	208
##	1	4	2	1	1	1	10	2	1	1
##	209	210	212	213	216	218	219	220	226	229
##	1	3	1	2	1	1	1	1	2	1
##	230	231	234	235	240	242	244	249	250	251
##	1	2	1	1	11	1	1	1	1	1
##	260	263	266	267	268	270	272	273	280	282
##	1	1	1	1	1	1	1	1	5	2
##	288	289	291	297	298	300	305	309	310	312
##	2	1	1	1	1	3	1	1	1	1
##	321	323	330	347	350	353	360	364	373	376
##	1	1	1	1	1	1	2	1	1	1
##	381	388	390	394	396	400	410	415	416	428
##	1	1	1	1	1	3	1	1	1	1
##	435	440	444	446	453	463	470	480	482	489
##	1	1	1	1	1	1	1	2	1	1
##	520	521	532	549	598	627	705	720	819	882
##	2	1	1	1	1	1	1	1	1	1
##	960	980	1017	1200	1258	1484	1550	1680	1848	2044
##	2	1	1	1	1	1	1	1	1	1
##	3600									
##	1									

```
summary(train_sample6$Dev_uni_proxima)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00   0.00   0.00   0.13   0.00 3600.00
```

```
quantile(train_sample6$Dev_uni_proxima, .999)
```

```
## 99.9%
##      18
```

```
any(is.na(train_sample6))
```

```
## [1] FALSE
```

```
any(train_sample6$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

```
table(train_sample7$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7      8      9
## 9995019 193473 78631 33797 19528 14660 9191 4518 4837 3031
##      10      11      12      13      14      15      16      17      18      19
## 5527 1780 2283 1167 1168 1608 899 664 769 488
##      20      21      22      23      24      25      26      27      28      29
## 1236 417 436 361 471 407 284 250 233 185
##      30      31      32      33      34      35      36      37      38      39
## 520 157 184 140 146 132 153 79 126 97
##      40      41      42      43      44      45      46      47      48      49
## 1021 76 167 69 74 82 60 85 95 40
##      50      51      52      53      54      55      56      57      58      59
## 115 37 92 45 60 34 50 20 42 28
##      60      61      62      63      64      65      66      67      68      69
## 92 18 31 23 25 42 31 23 21 12
##      70      71      72      73      74      75      76      77      78      79
## 40 9 22 18 14 23 18 11 16 13
##      80      81      82      83      84      85      86      87      88      89
## 161 6 17 7 28 17 19 12 8 8
##      90      91      92      93      94      95      96      97      98      99
## 27 6 10 8 20 7 13 9 8 6
## 100 101 102 103 104 105 106 107 108 109
## 39 5 1 4 24 5 5 4 9 5
## 110 111 112 113 114 115 116 117 118 119
## 11 6 6 2 9 4 5 7 2 6
## 120 121 122 123 124 125 126 127 128 129
## 65 4 4 4 3 1 5 5 5 2
## 130 131 132 134 135 137 138 139 140 141
## 6 4 11 3 2 3 4 3 5 5
## 142 143 144 145 147 148 149 150 151 152
## 4 1 1 9 3 3 4 7 2 4
## 153 155 156 157 158 160 161 163 164 165
## 2 5 9 2 5 27 3 2 1 2
## 166 168 171 172 174 175 176 179 180 181
## 1 7 1 1 1 1 2 1 9 1
## 184 185 186 187 188 189 191 192 193 195
## 2 1 1 1 2 1 2 3 2 1
```

```
##      198      199      200      201      203      206      208      209      210      213
##        1        1       15        1        1        2        5        1        3        1
##      214      216      218      219      220      221      223      226      227      228
##        1        1        1        2        2        1        2        2        1        1
##      230      231      233      234      235      240      241      244      245      249
##        1        1        1        1        2        8        1        2        2        1
##      251      253      255      256      257      258      259      260      261      270
##        2        1        1        1        2        2        1        2        1        2
##      272      274      276      279      281      282      284      288      289      290
##        1        1        1        1        1        1        1        3        1        1
##      301      302      304      306      308      311      312      320      324      329
##        1        1        1        1        1        2        1        3        2        1
##      332      337      344      349      350      356      358      360      370      374
##        1        1        1        1        1        1        1        2        1        1
##      377      390      400      401      404      416      417      420      423      428
##        1        1        2        1        1        1        1        1        1        1
##      443      460      463      469      470      480      487      494      500      510
##        1        1        1        1        1        1        1        1        1        1
##      528      588      596      600      632      640      682      705      720      740
##        1        1        1        1        1        1        1        1        1        1
##      750      766      779      949      1000      1020      1105      1116      1148      1200
##        1        1        1        1        2        1        1        1        1        1
##     1300     1478     1551     2240     2520     2772     2800
##        1        1        1        1        1        1        1
```

```
summary(train_sample7$Dev_uni_proxima)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
##    0.0000   0.0000   0.0000   0.1365   0.0000 2800.0000
```

```
quantile(train_sample7$Dev_uni_proxima, .999)
```

```
## 99.9%
##    18
```

```
any(is.na(train_sample7))
```

```
## [1] FALSE
```

```
any(train_sample7$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

```
table(train_sample8$Dev_uni_proxima)
```

```
##
##      0      1      2      3      4      5      6      7
## 10010732 198902 80638 34273 19838 14570 9088 4807
##      8      9     10     11     12     13     14     15
##  4859   2941   5413   1797   2259   1191   1164   1656
##     16     17     18     19     20     21     22     23
##    935    670    751    472   1294    475    378    335
##     24     25     26     27     28     29     30     31
##    453    384    274    287    247    188    447    165
##     32     33     34     35     36     37     38     39
##    172    144    119    154    167    102    114     93
##     40     41     42     43     44     45     46     47
##   1098     63    177     72     78     76     57     75
```

##	48	49	50	51	52	53	54	55
##	96	49	130	44	72	44	79	39
##	56	57	58	59	60	61	62	63
##	42	29	42	18	74	24	27	19
##	64	65	66	67	68	69	70	71
##	29	36	35	16	15	15	29	12
##	72	73	74	75	76	77	78	79
##	33	24	22	23	10	16	24	16
##	80	81	82	83	84	85	86	87
##	189	14	9	10	36	19	11	15
##	88	89	90	91	92	93	94	95
##	8	8	24	12	9	12	23	10
##	96	97	98	99	100	101	102	103
##	6	7	6	5	26	8	10	6
##	104	105	106	107	108	109	110	111
##	21	8	2	4	13	10	10	4
##	112	113	114	115	116	117	119	120
##	7	8	3	5	3	4	6	59
##	122	124	125	126	127	128	129	130
##	6	2	3	8	4	3	5	8
##	131	132	134	135	136	137	138	139
##	2	6	1	4	4	1	5	1
##	140	141	142	143	144	145	146	147
##	6	4	2	4	4	5	1	1
##	148	149	150	151	152	153	156	157
##	3	2	6	2	4	3	6	4
##	158	159	160	162	163	164	165	166
##	1	2	31	1	1	3	3	3
##	167	168	169	170	171	172	173	174
##	2	6	1	3	2	1	1	1
##	175	176	177	178	180	181	182	183
##	7	1	1	2	7	1	1	1
##	184	185	187	188	189	190	191	192
##	1	2	1	3	1	1	1	4
##	194	195	196	197	198	199	200	201
##	1	2	2	3	2	2	15	1
##	202	203	204	206	207	208	209	210
##	2	1	2	1	1	6	1	2
##	212	215	216	218	220	221	223	224
##	1	1	1	1	3	1	2	1
##	225	229	230	233	235	236	238	239
##	2	1	1	1	1	2	1	1
##	240	249	250	252	253	255	256	257
##	8	1	1	2	2	1	1	1
##	258	260	263	264	270	271	273	276
##	1	3	1	1	3	1	1	1
##	280	289	290	295	300	312	317	320
##	5	1	2	1	2	4	1	1
##	325	330	334	340	342	359	360	361
##	1	1	1	3	1	1	1	1
##	362	373	397	400	403	405	406	415
##	2	1	1	1	1	2	2	1
##	416	427	431	432	440	460	480	492
##	2	1	1	1	1	1	1	2

```
##      500      505      530      533      564      596      599      600
##      1       1       1       1       1       1       1       1
##     629     630     647     652     739     800     833     840
##      1       1       1       1       1       1       1       2
##     985    1120    1159    1197    1240    1389    1410    1686
##      1       1       1       1       1       1       1       2
##    1708    2604    2701    3360    6768    9765
##      1       1       1       1       1       1
```

```
summary(train_sample8$Dev_uni_proxima)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00   0.00   0.00   0.14   0.00 9765.00
```

```
quantile(train_sample8$Dev_uni_proxima, .999)
```

```
## 99.9%
```

```
##      18
```

```
any(is.na(train_sample8))
```

```
## [1] FALSE
```

```
any(train_sample8$Dev_uni_proxima<0)
```

```
## [1] FALSE
```

Eliminando outliers

Como pode ser observado pela análise exploratória, há vários valores extremos que podem ser julgados como erros ou valores que apresentam um fato raro e pontual. Então esses valores serão retirados do objeto de treino para não interferirem negativamente no modelo de machine learning.

```
train_sample3 <- dplyr::filter(train_sample3, train_sample3$Dev_uni_proxima < 25)
train_sample4 <- dplyr::filter(train_sample4, train_sample4$Dev_uni_proxima < 30)
train_sample5 <- dplyr::filter(train_sample5, train_sample5$Dev_uni_proxima < 20)
train_sample6 <- dplyr::filter(train_sample6, train_sample6$Dev_uni_proxima < 20)
train_sample7 <- dplyr::filter(train_sample7, train_sample7$Dev_uni_proxima < 20)
train_sample8 <- dplyr::filter(train_sample8, train_sample8$Dev_uni_proxima < 20)
```

Treinando o modelo

Agora com os dados limpos e estruturados pode se realizar o treinamento do modelo. Foi utilizado o modelo de regressão linear para essa previsão. O método de avaliação do modelo foi o Root Mean Squared Logarithmic Error (RMSLE).

```
treino <- full_join(train_sample3, train_sample4)
```

```
## Joining, by = c("Semana", "Canal_ID", "Producto_ID", "Venta_uni_hoy",
## "Dev_uni_proxima", "Demanda_uni_equil")
```

```
rm(train_sample3)
```

```
rm(train_sample4)
```

```
treino <- full_join(treino, train_sample5)
```

```
## Joining, by = c("Semana", "Canal_ID", "Producto_ID", "Venta_uni_hoy",
## "Dev_uni_proxima", "Demanda_uni_equil")
```

```
rm(train_sample5)
treino <- full_join(treino, train_sample6)

## Joining, by = c("Semana", "Canal_ID", "Producto_ID", "Venta_uni_hoy",
## "Dev_uni_proxima", "Demanda_uni_equil")

rm(train_sample6)
treino <- full_join(treino, train_sample7)

## Joining, by = c("Semana", "Canal_ID", "Producto_ID", "Venta_uni_hoy",
## "Dev_uni_proxima", "Demanda_uni_equil")

rm(train_sample7)
treino <- full_join(treino, train_sample8)

## Joining, by = c("Semana", "Canal_ID", "Producto_ID", "Venta_uni_hoy",
## "Dev_uni_proxima", "Demanda_uni_equil")

rm(train_sample8)
treino$Semana <- NULL
fit_lm <- bigglm(formula = Demanda_uni_equil ~ Canal_ID + Producto_ID + Venta_uni_hoy + Dev_uni_proxima
actual <- test$Demanda_uni_equil
test$Demanda_uni_equil <- NULL
test$Demanda_uni_equil <- 0
test$Semana <- NULL
prev <- predict(fit_lm, test)
```

Avaliação do modelo

Algumas previsões tiveram valores negativos mas que em sua maioria foram bem próximas de zero, como nesse caso não há valores de demanda inferiores a zero então os valores negativos foram igualados a zero.

```
any(prev < 0)

## [1] TRUE

df_prev <- as.data.frame(prev)
df_prev$V1[df_prev$V1 < 0] <- 0
df_actual <- as.data.frame(actual)
rmsle(actual = as.numeric(df_actual$actual), predicted = as.numeric(df_prev$V1))

## [1] 0.0586998
```