

Study3

Zhe GUAN

2024-07-11

Prepare, Process

```
options(repos = "https://cran.rstudio.com/")
install.packages("data.table")
```

```
##
## The downloaded binary packages are in
## /var/folders/pt/0nf2m3pj1f3b373wsyfc6glh0000gn/T//RtmppClujN/downloaded_packages
```

```
install.packages("tidyverse")
```

```
## Warning in download.file(url, destfile, method, mode = "wb", ...): URL
## 'https://cran.rstudio.com/bin/macosx/big-sur-arm64/contrib/4.4/tidyverse_2.0.0.tgz':
## status was 'SSL connect error'
```

```
## Error in download.file(url, destfile, method, mode = "wb", ...) :
## cannot open URL 'https://cran.rstudio.com/bin/macosx/big-sur-arm64/contrib/4.4/tidyverse_2.0.0.tgz'
```

```
## Warning in download.packages(pkgs, destdir = tmpd, available = available, :
## download of package 'tidyverse' failed
```

```
install.packages("dplyr")
```

```
## Warning in download.file(url, destfile, method, mode = "wb", ...): URL
## 'https://cran.rstudio.com/bin/macosx/big-sur-arm64/contrib/4.4/dplyr_1.1.4.tgz':
## status was 'SSL connect error'
```

```
## Error in download.file(url, destfile, method, mode = "wb", ...) :
## cannot open URL 'https://cran.rstudio.com/bin/macosx/big-sur-arm64/contrib/4.4/dplyr_1.1.4.tgz'
```

```
## Warning in download.packages(pkgs, destdir = tmpd, available = available, :
## download of package 'dplyr' failed
```

```
install.packages("ggmosaic")
```

```
##
## The downloaded binary packages are in
## /var/folders/pt/0nf2m3pj1f3b373wsyfc6glh0000gn/T//RtmppClujN/downloaded_packages
```

```
library(data.table)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.1      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.1
## v purrr      1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::between()      masks data.table::between()
## x dplyr::filter()       masks stats::filter()
## x dplyr::first()        masks data.table::first()
## x lubridate::hour()     masks data.table::hour()
## x lubridate::isoweek()  masks data.table::isoweek()
## x dplyr::lag()          masks stats::lag()
## x dplyr::last()         masks data.table::last()
## x lubridate::mday()     masks data.table::mday()
## x lubridate::minute()   masks data.table::minute()
## x lubridate::month()    masks data.table::month()
## x lubridate::quarter()  masks data.table::quarter()
## x lubridate::second()   masks data.table::second()
## x purrr::transpose()    masks data.table::transpose()
## x lubridate::wday()     masks data.table::wday()
## x lubridate::week()     masks data.table::week()
## x lubridate::yday()     masks data.table::yday()
## x lubridate::year()     masks data.table::year()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(dplyr)
library(ggmosaic)
```

read data from link:

```
filepath <- "/Users/zheguan/CWR_fig/QVI_case/"
transaction_data1 <- fread(paste(filepath,"QVI_transaction_data.csv",sep = ""))
customer_data1 <- fread(paste0(filepath,"QVI_purchase_behaviour.csv"))
str(customer_data1)
```

```
## Classes 'data.table' and 'data.frame': 72637 obs. of 3 variables:
## $ LYLTY_CARD_NBR : int 1000 1002 1003 1004 1005 1007 1009 1010 1011 1012 ...
## $ LIFESTAGE : chr "YOUNG SINGLES/COUPLES" "YOUNG SINGLES/COUPLES" "YOUNG FAMILIES" "OLDER SINGLES" ...
## $ PREMIUM_CUSTOMER: chr "Premium" "Mainstream" "Budget" "Mainstream" ...
## - attr(*, ".internal.selfref")=<externalptr>
```

Primary key is LYLTY_CARD_NBR, then we can check transaction data:

```
transaction_data1[!is.na(LYLTY_CARD_NBR),]
```

```
##      DATE STORE_NBR LYLTY_CARD_NBR TXN_ID PROD_NBR
##      <int>      <int>          <int> <int>    <int>
```

```
##      1: 43390      1      1000      1      5
##      2: 43599      1      1307     348     66
##      3: 43605      1      1343     383     61
##      4: 43329      2      2373     974     69
##      5: 43330      2      2426    1038    108
##      ---
## 264832: 43533      272      272319 270088      89
## 264833: 43325      272      272358 270154      74
## 264834: 43410      272      272379 270187      51
## 264835: 43461      272      272379 270188      42
## 264836: 43365      272      272380 270189      74
##
##                                PROD_NAME PROD_QTY TOT_SALES
##                                <char>    <int>    <num>
##      1:  Natural Chip      Compny SeaSalt175g      2      6.0
##      2:              CCs Nacho Cheese    175g      3      6.3
##      3:  Smiths Crinkle Cut  Chips Chicken 170g      2      2.9
##      4:  Smiths Chip Thinly  S/Cream&Union 175g      5     15.0
##      5:  Kettle Tortilla ChpsHny&Jlpno Chili 150g      3     13.8
##      ---
## 264832:  Kettle Sweet Chilli And Sour Cream 175g      2     10.8
## 264833:              Tostitos Splash Of  Lime 175g      1      4.4
## 264834:              Doritos Mexicana    170g      2      8.8
## 264835:  Doritos Corn Chip Mexican Jalapeno 150g      2      7.8
## 264836:              Tostitos Splash Of  Lime 175g      2      8.8
```

Select the shops from stores 77,86,88 in which we are intrested:

```
selected_shops_transaction <- transaction_data1[STORE_NBR %in% c(77,86,88), ]
tibble(selected_shops_transaction)
```

```
## # A tibble: 3,974 x 8
##   DATE STORE_NBR LYLTY_CARD_NBR TXN_ID PROD_NBR PROD_NAME  PROD_QTY TOT_SALES
##   <int>   <int>       <int> <int>   <int> <chr>      <int>    <dbl>
## 1 43605      88       88320 87811    113 Twisties C~      2      9.2
## 2 43633      77       77069 74987     70 Tyrrells C~      2      8.4
## 3 43552      77       77000 74911     18 Cheetos Ch~      1      3.3
## 4 43568      77       77000 74912     69 Smiths Chi~      1      3
## 5 43298      77      2330211 236744     94 Burger Rin~      1      2.3
## 6 43527      77       77063 74977    112 Tyrrells C~      2      8.4
## 7 43516      77       77069 74985     98 NCC Sour C~      2      6
## 8 43532      77       77069 74986      8 Smiths Cri~      2      5.8
## 9 43524      77       77310 75254      9 Kettle Tor~      1      4.6
## 10 43375      77       77502 75463     94 Burger Rin~      2      4.6
## # i 3,964 more rows
```

left-link transaction data with the customer details:

```
merged_data_selectedshops <- merge(selected_shops_transaction, customer_data1, by = "LYLTY_CARD_NBR", all..
tibble(merged_data_selectedshops)
```

```
## # A tibble: 3,974 x 10
##   LYLTY_CARD_NBR DATE STORE_NBR TXN_ID PROD_NBR PROD_NAME  PROD_QTY TOT_SALES
```

```
##           <int> <int>           <int> <int>           <int> <chr>           <int>           <dbl>
## 1           77000 43552           77 74911           18 Cheetos Ch~           1           3.3
## 2           77000 43568           77 74912           69 Smiths Chi~           1           3
## 3           77000 43369           77 74910           36 Kettle Chi~           2          10.8
## 4           77001 43523           77 74913            7 Smiths Cri~           2          11.4
## 5           77001 43486           77 74914            9 Kettle Tor~           2           9.2
## 6           77002 43602           77 74915           63 Kettle 135~           1           4.2
## 7           77002 43634           77 74916          107 Smiths Cri~           2           5.2
## 8           77003 43542           77 74917           80 Natural Ch~           1           3
## 9           77004 43299           77 74918           30 Doritos Co~           1           4.4
## 10          77004 43307           77 74919           46 Kettle Ori~           2          10.8
## # i 3,964 more rows
## # i 2 more variables: LIFESTAGE <chr>, PREMIUM_CUSTOMER <chr>
```

the stakeholder asks us to focus on these measures:

consider the monthly sales experience of each store.

This can be broken down by:

1. total sales revenue 2. total number of customers 3. average number of transactions per customer

by checking, the date should be transformed with proper kind.

```
merged_data_selectedshops[,DATE := as.Date(DATE,origin = "1899/12/30")]
tibble(merged_data_selectedshops)
```

```
## # A tibble: 3,974 x 10
##   LYLTY_CARD_NBR DATE      STORE_NBR TXN_ID PROD_NBR PROD_NAME      PROD_QTY
##   <int> <date>           <int> <int>   <int> <chr>           <int>
## 1           77000 2019-03-28           77 74911           18 Cheetos Chs & B~           1
## 2           77000 2019-04-13           77 74912           69 Smiths Chip Thi~           1
## 3           77000 2018-09-26           77 74910           36 Kettle Chillli 1~           2
## 4           77001 2019-02-27           77 74913            7 Smiths Crinkle ~           2
## 5           77001 2019-01-21           77 74914            9 Kettle Tortilla~           2
## 6           77002 2019-05-17           77 74915           63 Kettle 135g Swt~           1
## 7           77002 2019-06-18           77 74916          107 Smiths Crinkle ~           2
## 8           77003 2019-03-18           77 74917           80 Natural ChipCo ~           1
## 9           77004 2018-07-18           77 74918           30 Doritos Corn Ch~           1
## 10          77004 2018-07-26           77 74919           46 Kettle Original~           2
## # i 3,964 more rows
## # i 3 more variables: TOT_SALES <dbl>, LIFESTAGE <chr>, PREMIUM_CUSTOMER <chr>
```

1. check the total sales revenue per month among these stores:

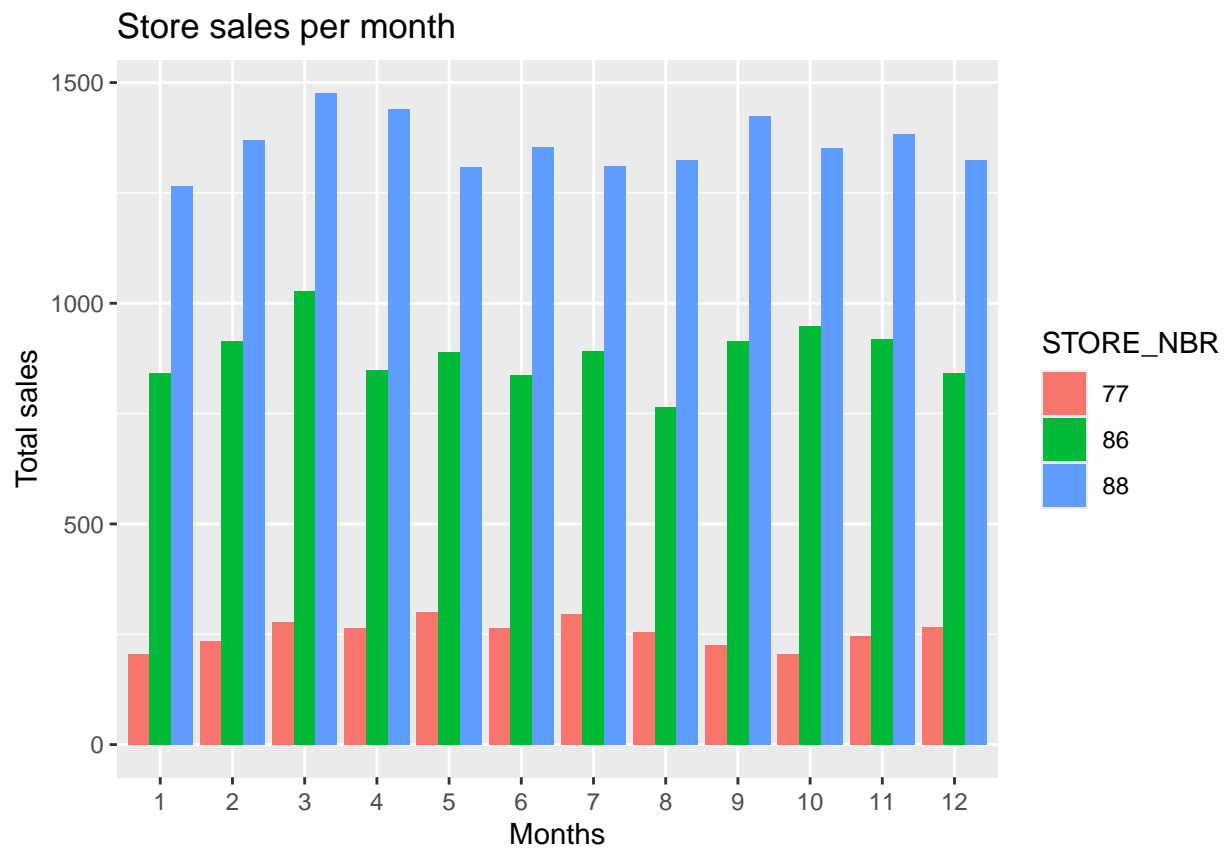
```
summary_table_1 <- merged_data_selectedshops[,.(month_income = sum(TOT_SALES)),.(month(DATE),STORE_NBR)]
tibble(summary_table_1)
```

```
## # A tibble: 36 x 3
##   month STORE_NBR month_income
##   <fct> <fct>           <dbl>
## 1 3      77          278.
## 2 4      77          263.
## 3 9      77          225.
```

```
## 4 2      77      235
## 5 1      77      204.
## 6 5      77      299.
## 7 6      77      265.
## 8 7      77      297.
## 9 8      77      256.
## 10 11     77      245.
## # i 26 more rows
```

plot the contributions

```
ggplot(data = summary_table_1) + geom_bar(position = position_dodge(), aes(x=month, fill= STORE_NBR, weigh
```



2. We count the total number of customers in separate stores, and only account the customers with unique loyalty number.

```
summary_table_2 <- merged_data_selectedshops[,.(num_of_customer=uniqueN(LYLT_CARD_NBR)),.(month(DATE)),  
tibble(summary_table_2)
```

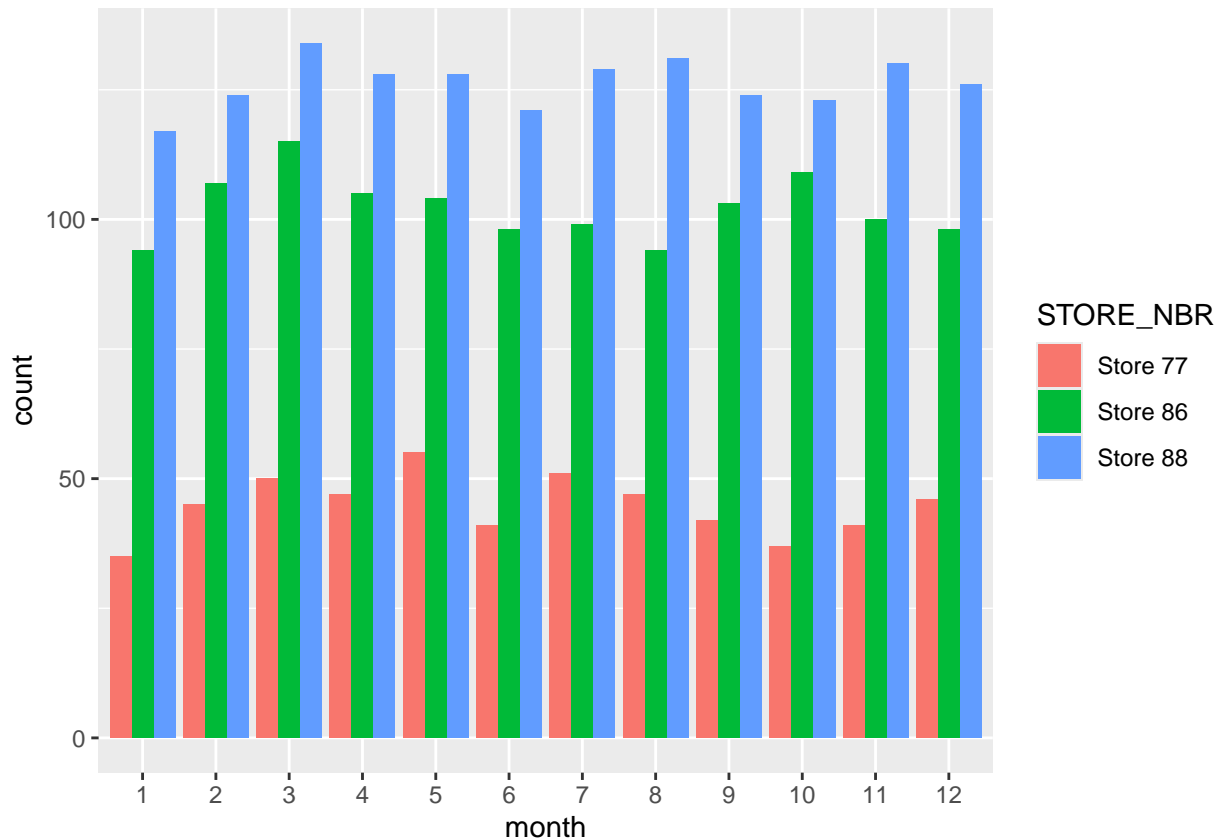
```
## # A tibble: 36 x 3
##   month STORE_NBR num_of_customer
##   <dbl>   <int>         <int>
## 1     3         77             50
## 2     4         77             47
```

```
## 3      9      77      42
## 4      2      77      45
## 5      1      77      35
## 6      5      77      55
## 7      6      77      41
## 8      7      77      51
## 9      8      77      47
## 10     11      77      41
## # i 26 more rows
```

```
summary_table_2[,STORE_NBR := as.factor(paste0("Store ",STORE_NBR))][,month := as.factor(month)]
```

plot1: bar plot

```
ggplot(data = summary_table_2) +geom_bar(position = position_dodge(), aes(x= month,weight = num_of_cust
```



plot2 mosaic:

```
p <- ggplot( data = summary_table_2) +geom_mosaic(aes(weight = num_of_customer, x = product(month,STORE_NBR)))
tibble(ggplot_build(p)$data[[1]])
```

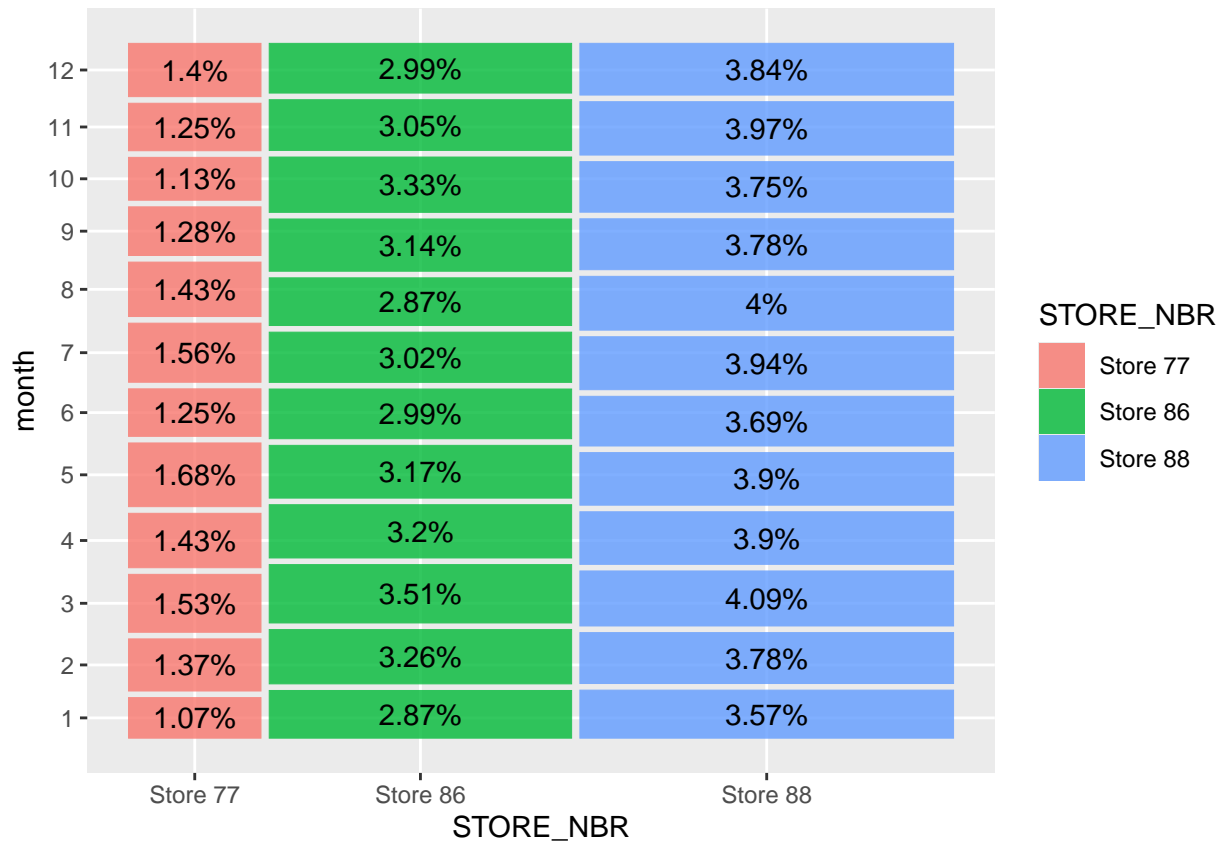
```
## Warning: The 'scale_name' argument of 'continuous_scale()' is deprecated as of ggplot2
## 3.5.0.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## Warning: The 'trans' argument of 'continuous_scale()' is deprecated as of ggplot2 3.5.0.
## i Please use the 'transform' argument instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## Warning: 'unite()' was deprecated in tidyr 1.2.0.
## i Please use 'unite()' instead.
## i The deprecated feature was likely used in the ggmosaic package.
## Please report the issue at <https://github.com/haleyjeppson/ggmosaic>.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## # A tibble: 36 x 25
##   fill x_fill__STORE_NBR .wt xmin xmax ymin ymax level x__month .n
##   <chr> <fct>          <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <fct> <dbl>
## 1 #F87~ Store 77      35    0 0.161 0      0.0583    2 1      35
## 2 #F87~ Store 77      45    0 0.161 0.0679 0.143    2 2      45
## 3 #F87~ Store 77      50    0 0.161 0.152 0.236    2 3      50
## 4 #F87~ Store 77      47    0 0.161 0.245 0.324    2 4      47
## 5 #F87~ Store 77      55    0 0.161 0.333 0.425    2 5      55
## 6 #F87~ Store 77      41    0 0.161 0.434 0.503    2 6      41
## 7 #F87~ Store 77      51    0 0.161 0.512 0.597    2 7      51
## 8 #F87~ Store 77      47    0 0.161 0.607 0.685    2 8      47
## 9 #F87~ Store 77      42    0 0.161 0.695 0.765    2 9      42
## 10 #F87~ Store 77     37    0 0.161 0.774 0.836    2 10     37
## # i 26 more rows
## # i 15 more variables: label <chr>, x <dbl>, y <dbl>, group <dbl>, PANEL <fct>,
## #   width <dbl>, linetype <chr>, fontsize <dbl>, shape <dbl>, colour <lgl>,
## #   size <dbl>, alpha <dbl>, stroke <dbl>, linewidth <dbl>, weight <dbl>
```

```
p + geom_text(data = ggplot_build(p)$data[[1]], aes(x=xmin/2+xmax/2,y=ymin/2+ymax/2,label = paste0(round
```



plot3:

```
install.packages("viridis")
```

```
##
## The downloaded binary packages are in
## /var/folders/pt/0nf2m3pj1f3b373wsyfc6glh0000gn/T//RtmppClujN/downloaded_packages
```

```
library(viridis)
```

```
## Loading required package: viridisLite
```

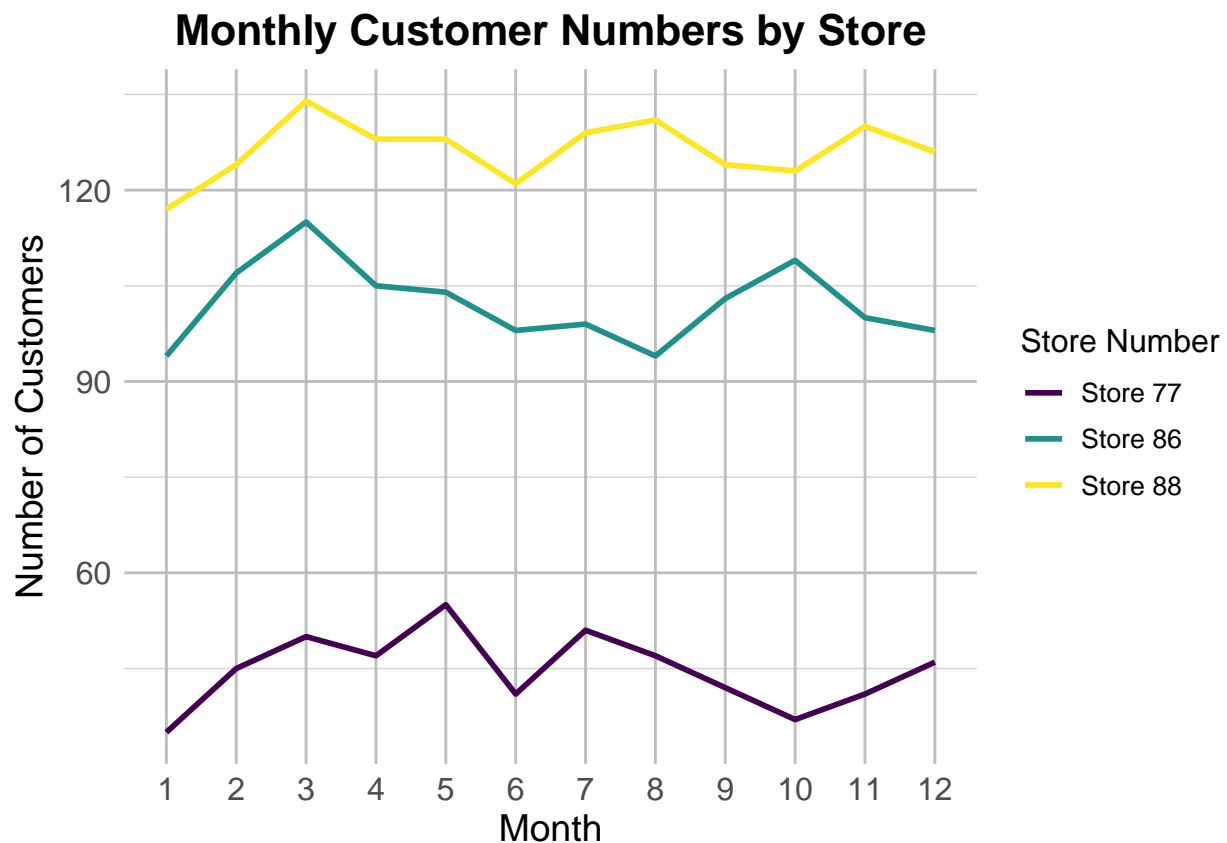
```
ggplot(data = summary_table_2) +
  geom_line(aes(x = month, y = num_of_customer, color = factor(STORE_NBR), group = STORE_NBR), size = 1,
  scale_color_viridis(discrete = TRUE) + #
  labs(x = "Month", y = "Number of Customers", title = "Monthly Customer Numbers by Store", color = "Store",
  theme_minimal() + #
  theme(
    plot.title = element_text(hjust = 0.5, size = 16, face = "bold"), #
    axis.title.x = element_text(size = 14),
    axis.title.y = element_text(size = 14),
    axis.text = element_text(size = 12),
    legend.title = element_text(size = 12),
    legend.text = element_text(size = 10),
    panel.grid.major = element_line(color = "gray", size = 0.5),
```



```
panel.grid.minor = element_line(color = "lightgray", size = 0.25)
)
```

```
## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## Warning: The 'size' argument of 'element_line()' is deprecated as of ggplot2 3.4.0.
## i Please use the 'linewidth' argument instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```



3. calculate the average number of transactions of customer:

```
summary_table_3 <- merged_data_selectedshops[,.(total_num = sum(PROD_QTY)),.(month(Date),STORE_NBR)]
tibble(summary_table_3)
```

```
## # A tibble: 36 x 3
##   month STORE_NBR total_num
##   <dbl>     <int>     <int>
```

```
## 1      3      77      82
## 2      4      77      78
## 3      9      77      70
## 4      2      77      74
## 5      1      77      65
## 6      5      77      84
## 7      6      77      70
## 8      7      77      84
## 9      8      77      74
## 10     11      77      67
## # i 26 more rows
```

```
summary_table_3[,STORE_NBR := as.factor(paste0("Store ",STORE_NBR))][,month := as.factor(month)]
```

merged table:

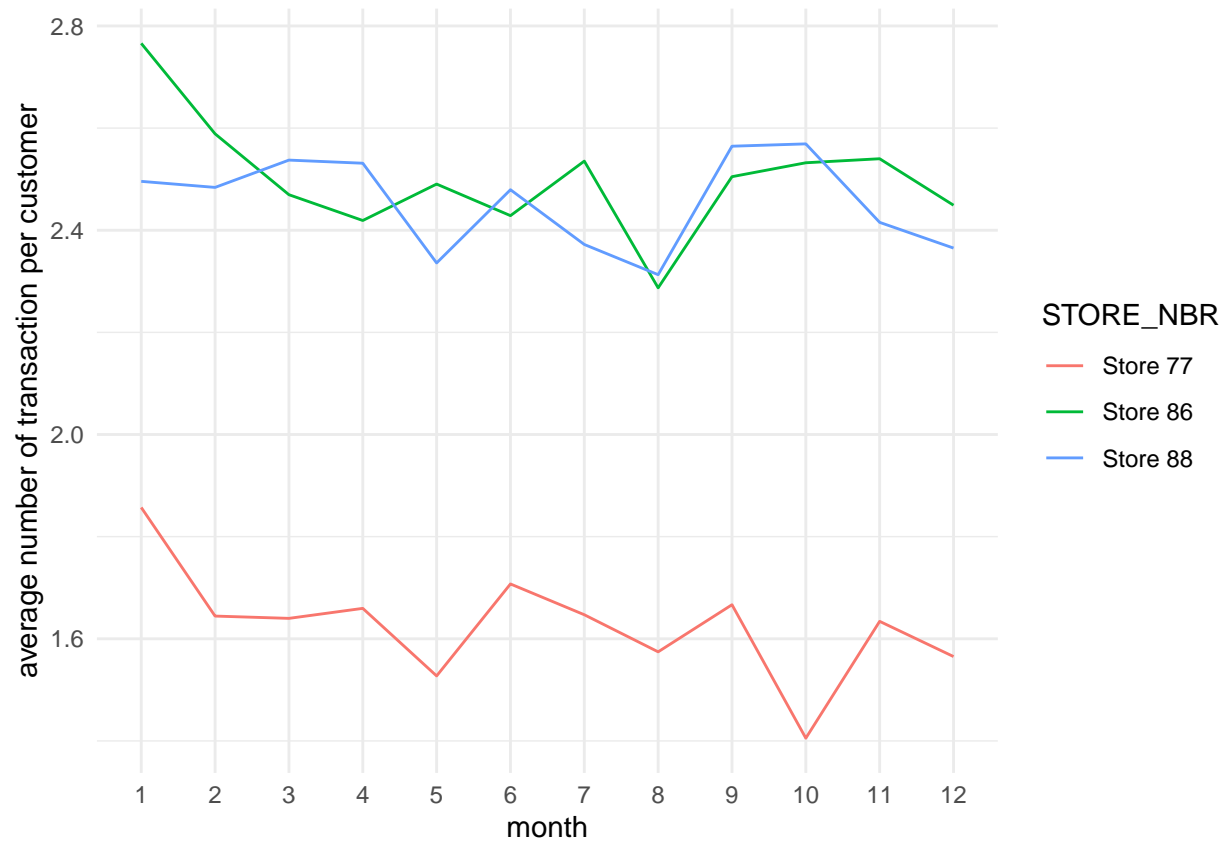
```
summary_table_3_link_2 <- merge(summary_table_3,summary_table_2,by = c("STORE_NBR","month"),all.x = TRUE)
tibble(summary_table_3_link_2)
```

```
## # A tibble: 36 x 4
##   STORE_NBR month total_num num_of_customer
##   <fct>      <fct>      <int>          <int>
## 1 Store 77  1          65             35
## 2 Store 77  2          74             45
## 3 Store 77  3          82             50
## 4 Store 77  4          78             47
## 5 Store 77  5          84             55
## 6 Store 77  6          70             41
## 7 Store 77  7          84             51
## 8 Store 77  8          74             47
## 9 Store 77  9          70             42
## 10 Store 77 10          52             37
## # i 26 more rows
```

```
summary_table_3_link_2[,avg_num := total_num/num_of_customer]
```

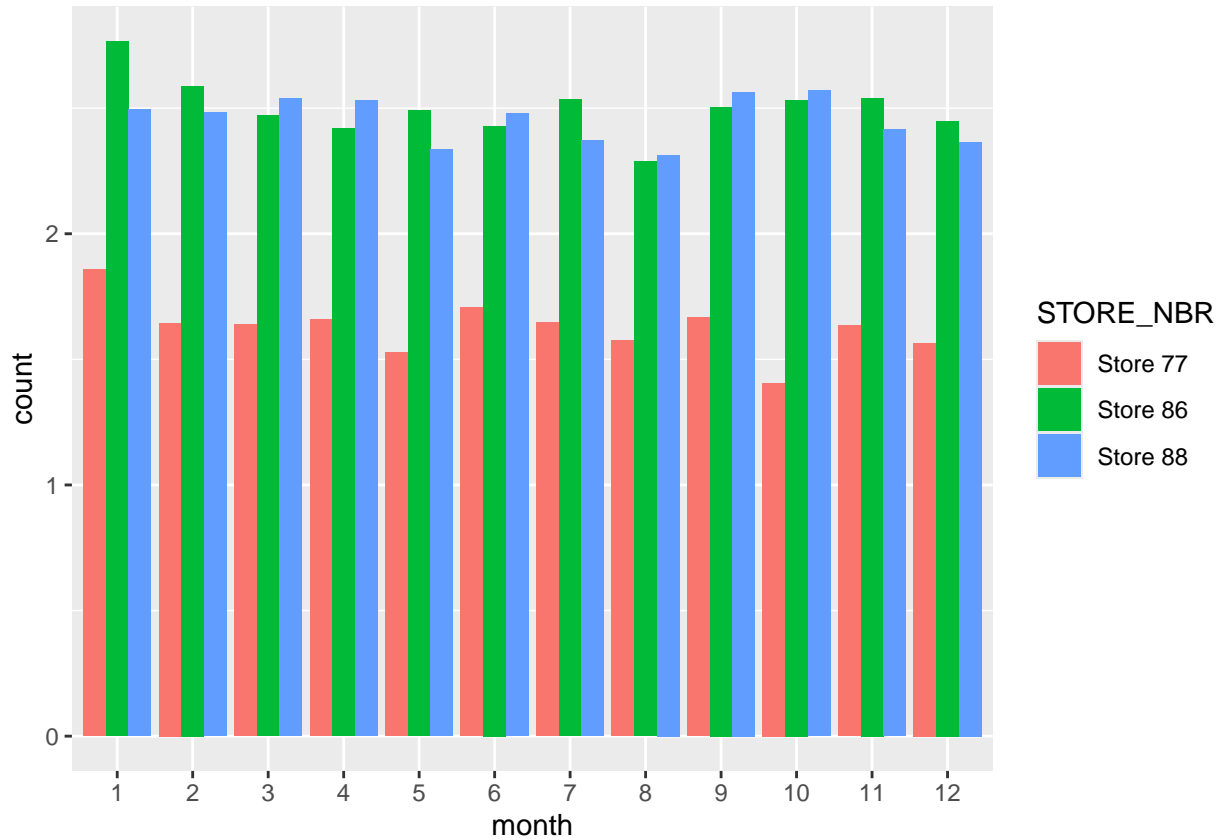
plot:

```
ggplot(data = summary_table_3_link_2) +geom_line(aes(x= month, y = avg_num, group = STORE_NBR, colour =
```



bar plot:

```
ggplot(data = summary_table_3_link_2) + geom_bar(position = position_dodge(), aes(x=month, weight = avg_r
```



Next important step is to choose the Control Stores, here we define a function to calculate the correlations between trial stores and other stores.

firstly, we need to similarly calculate a summary table:

```
transaction_data1[,DATE := as.Date(DATE,origin = "1899/12/30")]
measureOverTime <- transaction_data1[,.(totalSales = sum(TOT_SALES),nCustomers= uniqueN(LYLTY_CARD_NBR))
tibble(measureOverTime)
```

```
## # A tibble: 3,169 x 6
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit
##   <dbl>   <int>      <dbl>      <int>      <int>          <dbl>
## 1     1       1      155.        35         42           3.69
## 2     1       2      163.        43         49           3.32
## 3     1       3     1052.       102        236           4.46
## 4     1       4     1525.       134        335           4.55
## 5     1       5      838.        92        236           3.55
## 6     1       7      975.       102        221           4.41
## 7     1      10      879.       102        252           3.49
## 8     1      15      874.        99        248           3.52
## 9     1      17      365.        39         83           4.39
## 10    1      19      992.       111        224           4.43
## # i 3,159 more rows
```

Analyze

define a function which can calculate the correlation between two stores by using the information in separate 12 months.

```
testN <- measureOverTime[,N := .N ,.(STORE_NBR)][order(N)]
```

```
filtered_stores <- measureOverTime[N==12,]  
filtered_stores
```

```
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N  
##      <num>    <int>    <num>    <int>        <int>        <num> <int>  
##    1:      1        1    154.8        35          42    3.685714    12  
##    2:      1        2    162.8        43          49    3.322449    12  
##    3:      1        3   1051.7       102         236    4.456356    12  
##    4:      1        4   1525.0       134         335    4.552239    12  
##    5:      1        5    838.0        92         236    3.550847    12  
##    ---  
## 3116:     12       224     20.5         5          8    2.562500    12  
## 3117:     12       177      5.3         2          2    2.650000    12  
## 3118:     12       204     33.6         5          9    3.733333    12  
## 3119:     12       159     11.0         3          4    2.750000    12  
## 3120:     12       244      3.0         1          1    3.000000    12
```

```
tibble(testN)
```

```
## # A tibble: 3,169 x 7  
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N  
##   <dbl>    <int>    <dbl>    <int>        <int>        <dbl> <int>  
## 1     3       92      9.2         1          2         4.6     1  
## 2     7       85     13.9         3          5         2.78    1  
## 3    10       76      6         1          2          3     1  
## 4     3      252      3.7         1          1         3.7     2  
## 5     4      211      2.6         1          1         2.6     2  
## 6     4      206      4.6         1          1         4.6     2  
## 7     7      211      2.6         1          1         2.6     2  
## 8     7      206      3         1          1          3     2  
## 9     8      252      3.7         1          1         3.7     2  
## 10    9       31      6         1          2          3     2  
## # i 3,159 more rows
```

```
comparison_vector1 <- measureOverTime[STORE_NBR == 1]$totalSales  
comparison_vector2 <- measureOverTime[STORE_NBR == 77]$totalSales  
unique(measureOverTime$STORE_NBR)
```

```
## [1] 1 2 3 4 5 7 10 15 17 19 20 22 23 24 25 26 27 28  
## [19] 29 33 34 36 37 38 39 40 41 43 45 46 47 48 49 52 53 54  
## [37] 55 56 57 58 59 61 65 69 71 72 75 78 80 81 84 86 90 91  
## [55] 94 95 96 97 100 101 102 105 106 107 108 109 110 111 112 113 114 116  
## [73] 118 119 120 121 122 123 124 125 126 127 128 129 130 131 133 136 138 142  
## [91] 144 145 148 150 151 152 153 154 155 156 157 160 162 164 165 166 168 169  
## [109] 173 175 178 181 182 183 184 185 186 187 188 189 190 191 194 196 199 200
```

```
## [127] 201 203 205 207 208 210 212 213 214 216 217 218 219 223 225 226 227 229
## [145] 230 231 232 233 234 235 236 237 238 241 243 247 249 250 253 255 256 257
## [163] 259 260 261 262 264 265 266 267 269 270 272 6 8 9 12 13 16 18
## [181] 21 30 32 35 50 60 62 63 64 66 67 68 73 74 79 82 83 87
## [199] 88 89 93 99 103 104 115 134 135 137 141 143 146 147 149 158 163 170
## [217] 172 174 179 180 195 197 202 209 215 220 221 222 228 239 240 242 245 246
## [235] 248 251 254 271 14 51 77 98 132 159 171 244 268 70 176 177 192 42
## [253] 167 204 224 44 140 258 139 263 161 198 193 117 92 252 211 206 85 31
## [271] 76 11
```

```
cor(comparison_vector1, comparison_vector2, method = "kendall")
```

```
## [1] 0.2424242
```

```
tibble(filtered_stores)
```

```
## # A tibble: 3,120 x 7
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##   <dbl>   <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1     1       1        155.        35         42         3.69    12
## 2     1       2        163.        43         49         3.32    12
## 3     1       3       1052.       102        236         4.46    12
## 4     1       4       1525.       134        335         4.55    12
## 5     1       5        838.        92        236         3.55    12
## 6     1       7        975.       102        221         4.41    12
## 7     1      10        879.       102        252         3.49    12
## 8     1      15        874.        99        248         3.52    12
## 9     1      17        365.        39         83         4.39    12
## 10    1      19        992.       111        224         4.43    12
## # i 3,110 more rows
```

```
calculate_cor1 <- function(store1,metrics,method){
  #calculate the total number of stores
  unique_num <- unique(filtered_stores$STORE_NBR)
  result <- numeric(length = length(unique_num))

  for (i in seq_along(unique_num)){
    trail_vector <- filtered_stores[STORE_NBR == store1,..metrics]
    comparison_vector <- filtered_stores[STORE_NBR == unique_num[i],..metrics]
    result[i] <- cor(trail_vector,comparison_vector,method = method)
  }
  result_table <- data.table(STORE_NBR = unique_num,Correlation = result)
  return(result_table)
}

calculate_cor <- function(store1, metrics, method) {
  if (!exists("filtered_stores") || !all(c("STORE_NBR", metrics) %in% colnames(filtered_stores))) {
    stop("filtered_stores don't exist")
  }

  unique_num <- unique(filtered_stores$STORE_NBR)
  result <- numeric(length = length(unique_num))
```

```

for (i in seq_along(unique_num)) {
  trail_vector <- filtered_stores[STORE_NBR == store1, ..metrics, with = FALSE]
  comparison_vector <- filtered_stores[STORE_NBR == unique_num[i], ..metrics, with = FALSE]

  if (all(sapply(trail_vector, function(x) all(is.na(x)))) || all(sapply(comparison_vector, function(x) all(is.na(x))))) {
    result[i] <- NA
  } else {
    result[i] <- cor(trail_vector, comparison_vector, method = method, use = "complete.obs")
  }
}

result_table <- data.table(STORE_NBR = unique_num, Correlation = result)
return(result_table)
}

```

```

metrics <- c("totalSales", "nCustomers", "nChipsPerTxn", "avgPricePerUnit")
result_cor_77_pearson <- sapply(metrics, function(m) calculate_cor1(77, m, "pearson"), simplify = FALSE)
result_cor_77_pearson <- rbindlist(result_cor_77_pearson, idcol = "Metric")
result_cor_86_pearson <- sapply(metrics, function(m) calculate_cor1(86, m, "pearson"), simplify = FALSE)
result_cor_86_pearson <- rbindlist(result_cor_86_pearson, idcol = "Metric")
result_cor_88_pearson <- sapply(metrics, function(m) calculate_cor1(88, m, "pearson"), simplify = FALSE)
result_cor_88_pearson <- rbindlist(result_cor_88_pearson, idcol = "Metric")
tibble(result_cor_77_pearson[order(-Correlation)])

```

```

## # A tibble: 1,040 x 3
##   Metric      STORE_NBR Correlation
##   <chr>      <int>      <dbl>
## 1 totalSales      77          1
## 2 nCustomers      77          1
## 3 nChipsPerTxn    77          1
## 4 avgPricePerUnit 77          1
## 5 nCustomers      35        0.788
## 6 totalSales      41        0.762
## 7 nCustomers      41        0.761
## 8 nCustomers     167        0.749
## 9 nChipsPerTxn    41        0.746
## 10 nCustomers     71        0.737
## # i 1,030 more rows

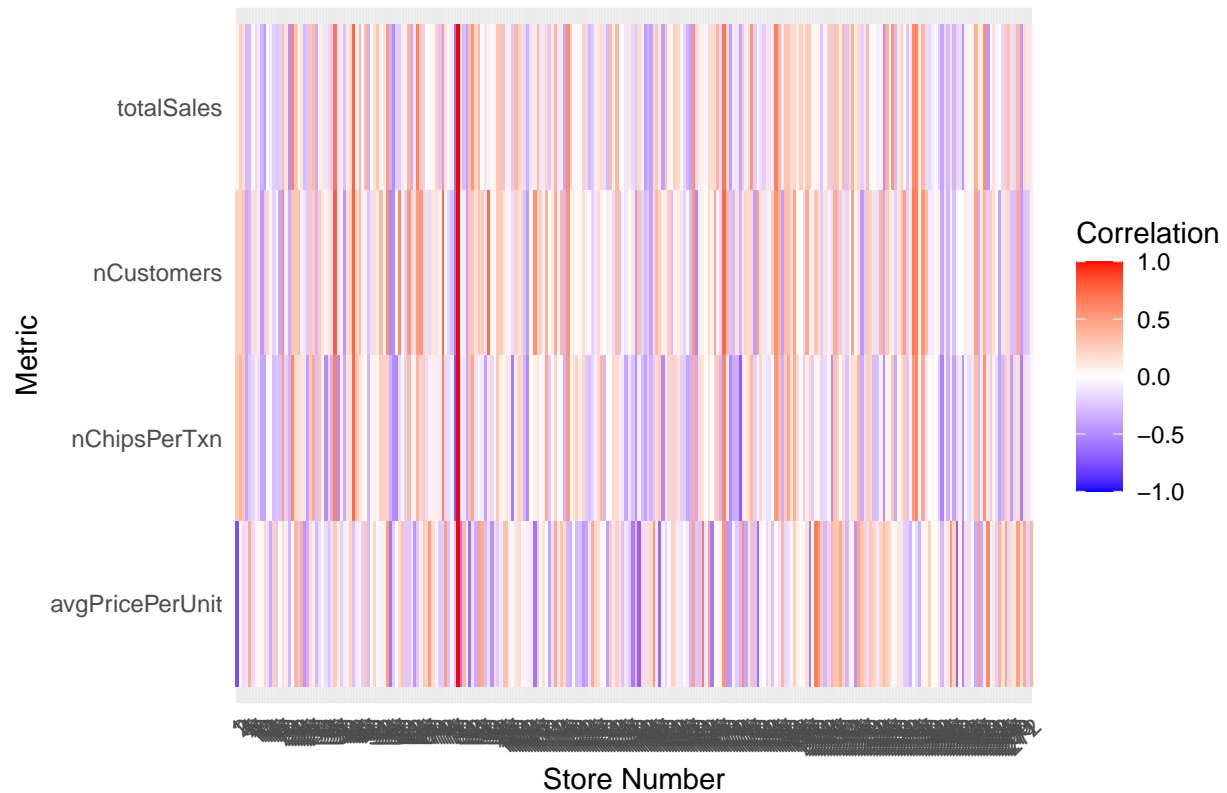
```

```

ggplot(data = result_cor_77_pearson, aes(x = as.factor(STORE_NBR), y = Metric, fill = Correlation)) +
  geom_tile() +
  scale_fill_gradient2(low = "blue", high = "red", mid = "white", midpoint = 0,
    limit = c(-1, 1), space = "Lab", name = "Correlation") +
  theme_minimal() +
  labs(title = "Heatmap of Correlation Between Stores and Metrics",
    x = "Store Number",
    y = "Metric") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

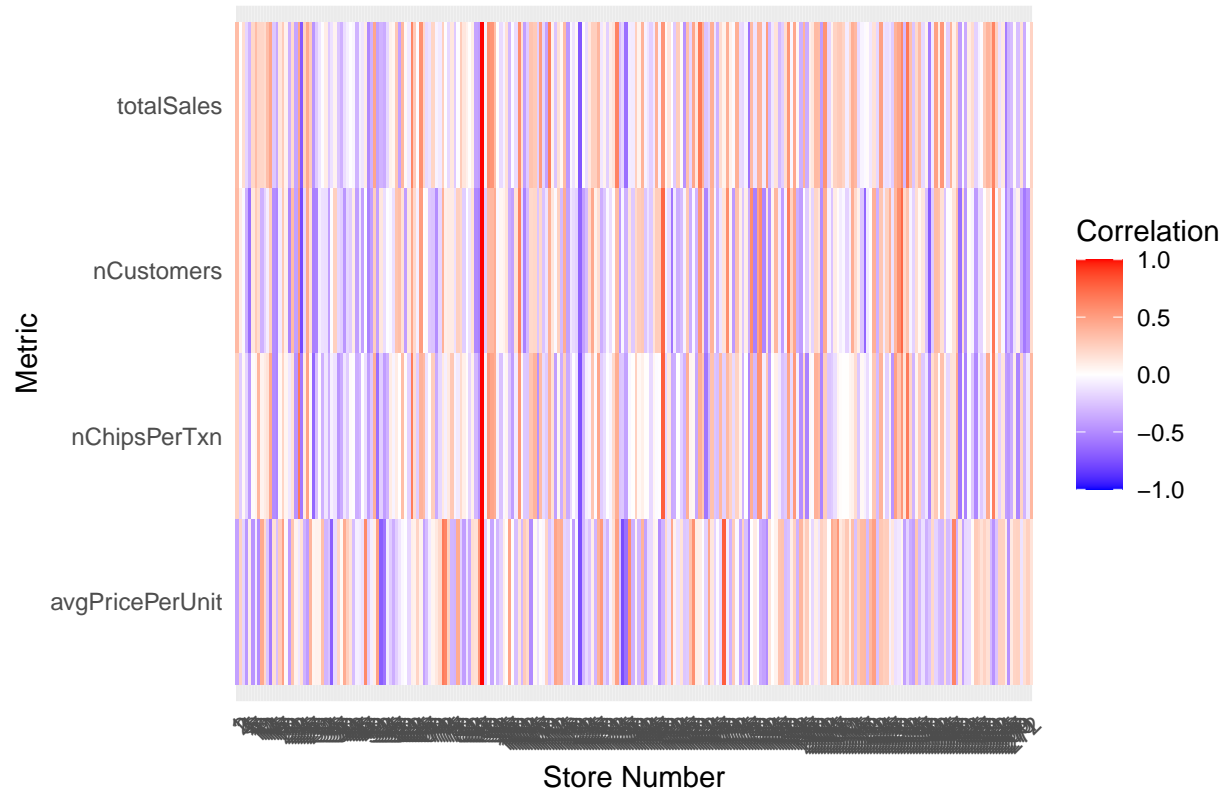
```

Heatmap of Correlation Between Stores and Metrics



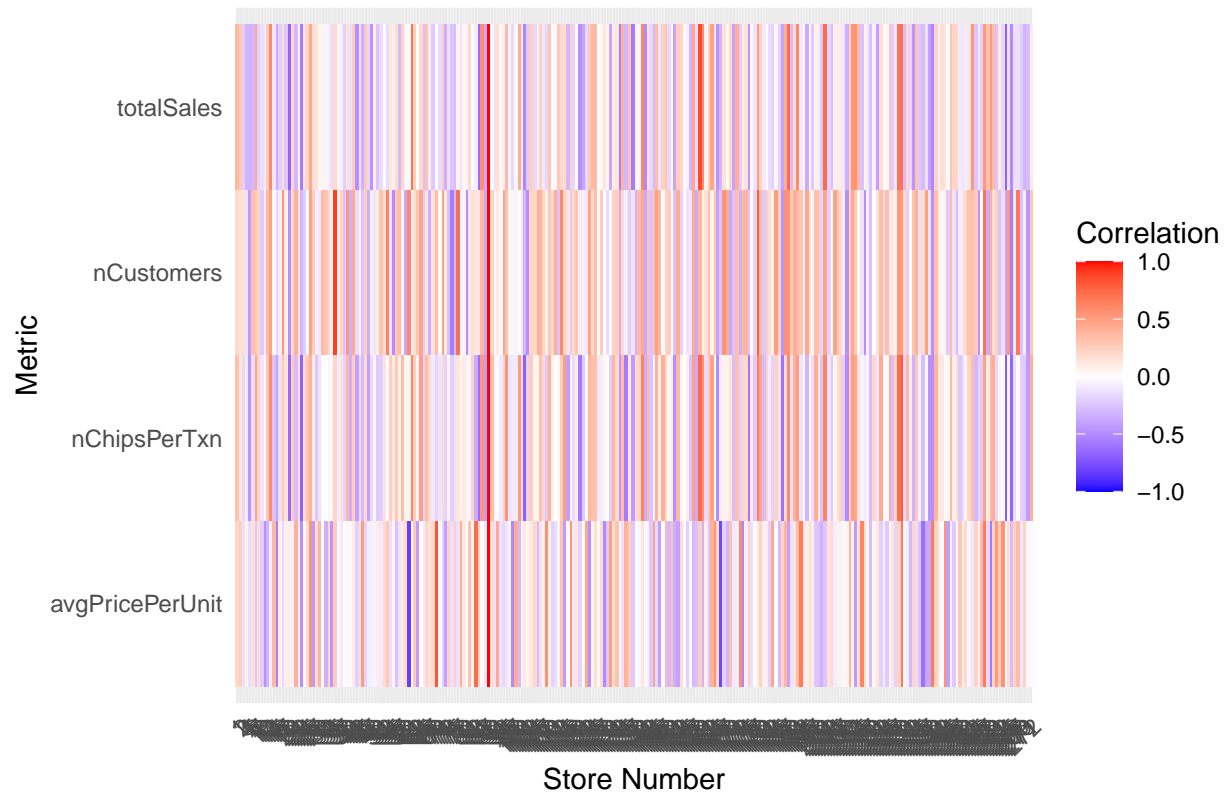
```
ggplot(data = result_cor_86_pearson, aes(x = as.factor(STORE_NBR), y = Metric, fill = Correlation)) +
  geom_tile() +
  scale_fill_gradient2(low = "blue", high = "red", mid = "white", midpoint = 0,
    limit = c(-1, 1), space = "Lab", name="Correlation") +
  theme_minimal() +
  labs(title = "Heatmap of Correlation Between Stores and Metrics",
    x = "Store Number",
    y = "Metric") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```


Heatmap of Correlation Between Stores and Metrics



```
ggplot(data = result_cor_88_pearson, aes(x = as.factor(STORE_NBR), y = Metric, fill = Correlation)) +
  geom_tile() +
  scale_fill_gradient2(low = "blue", high = "red", mid = "white", midpoint = 0,
    limit = c(-1, 1), space = "Lab", name="Correlation") +
  theme_minimal() +
  labs(title = "Heatmap of Correlation Between Stores and Metrics",
    x = "Store Number",
    y = "Metric") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Heatmap of Correlation Between Stores and Metrics



#we can define a distance which means the difference of metrics between trial and other stores.
`tibble(result_cor_77_pearson)`

```
## # A tibble: 1,040 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>         <dbl>
## 1 totalSales      1      0.0644
## 2 totalSales      2      0.262
## 3 totalSales      3      0.163
## 4 totalSales      4     -0.295
## 5 totalSales      5     -0.285
## 6 totalSales      7     -0.159
## 7 totalSales     10     -0.424
## 8 totalSales     15     -0.236
## 9 totalSales     17      0.338
## 10 totalSales     19     -0.485
## # i 1,030 more rows
```

```
sum_cor_central <- result_cor_77_pearson[STORE_NBR == 77,]
tibble(sum_cor_central)
```

```
## # A tibble: 4 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>         <dbl>
## 1 totalSales      77          1
```

```
## 2 nCustomers          77          1
## 3 nChipsPerTxn        77          1
## 4 avgPricePerUnit     77          1
```

```
distance_avg <- result_cor_77_pearson[,.(distance = sum((Correlation - sum_cor_central$Correlation)^2))
tibble(distance_avg[order(distance)])
```

```
## # A tibble: 260 x 2
##   STORE_NBR distance
##   <int>     <dbl>
## 1      77      0
## 2     167  0.505
## 3      35  0.787
## 4     157  1.01
## 5      53  1.22
## 6     184  1.49
## 7      17  1.51
## 8     233  1.54
## 9     268  1.67
## 10     111  1.68
## # i 250 more rows
```

```
#we can define a distance which means the difference of metrics between trial and other stores.
tibble(result_cor_86_pearson)
```

```
## # A tibble: 1,040 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>         <dbl>
## 1 totalSales      1      0.344
## 2 totalSales      2      0.00110
## 3 totalSales      3      0.178
## 4 totalSales      4     -0.179
## 5 totalSales      5     -0.322
## 6 totalSales      7      0.347
## 7 totalSales     10      0.184
## 8 totalSales     15     -0.375
## 9 totalSales     17      0.270
## 10 totalSales     19     -0.337
## # i 1,030 more rows
```

```
sum_cor_central <- result_cor_86_pearson[STORE_NBR == 86,]
tibble(sum_cor_central)
```

```
## # A tibble: 4 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>         <dbl>
## 1 totalSales      86          1
## 2 nCustomers      86          1
## 3 nChipsPerTxn    86          1
## 4 avgPricePerUnit 86          1
```

```
distance_avg <- result_cor_86_pearson[,.(distance = sum((Correlation - sum_cor_central$Correlation)^2))
tibble(distance_avg[order(distance)])
```

```
## # A tibble: 260 x 2
##   STORE_NBR distance
##   <int>     <dbl>
## 1      86         0
## 2     132     1.17
## 3      99     1.27
## 4     229     1.32
## 5     190     1.38
## 6      61     1.40
## 7     159     1.50
## 8     222     1.62
## 9      64     1.66
## 10    201     1.71
## # i 250 more rows
```

```
#we can define a distance which means the difference of metrics between trial and other stores.
tibble(result_cor_88_pearson)
```

```
## # A tibble: 1,040 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>     <dbl>
## 1 totalSales         1     0.371
## 2 totalSales         2     0.284
## 3 totalSales         3    -0.208
## 4 totalSales         4    -0.320
## 5 totalSales         5    -0.325
## 6 totalSales         7     0.444
## 7 totalSales        10    -0.0938
## 8 totalSales        15    -0.357
## 9 totalSales        17     0.170
## 10 totalSales        19    -0.668
## # i 1,030 more rows
```

```
sum_cor_central <- result_cor_88_pearson[STORE_NBR == 88,]
tibble(sum_cor_central)
```

```
## # A tibble: 4 x 3
##   Metric      STORE_NBR Correlation
##   <chr>         <int>     <dbl>
## 1 totalSales         88         1
## 2 nCustomers         88         1
## 3 nChipsPerTxn       88         1
## 4 avgPricePerUnit    88         1
```

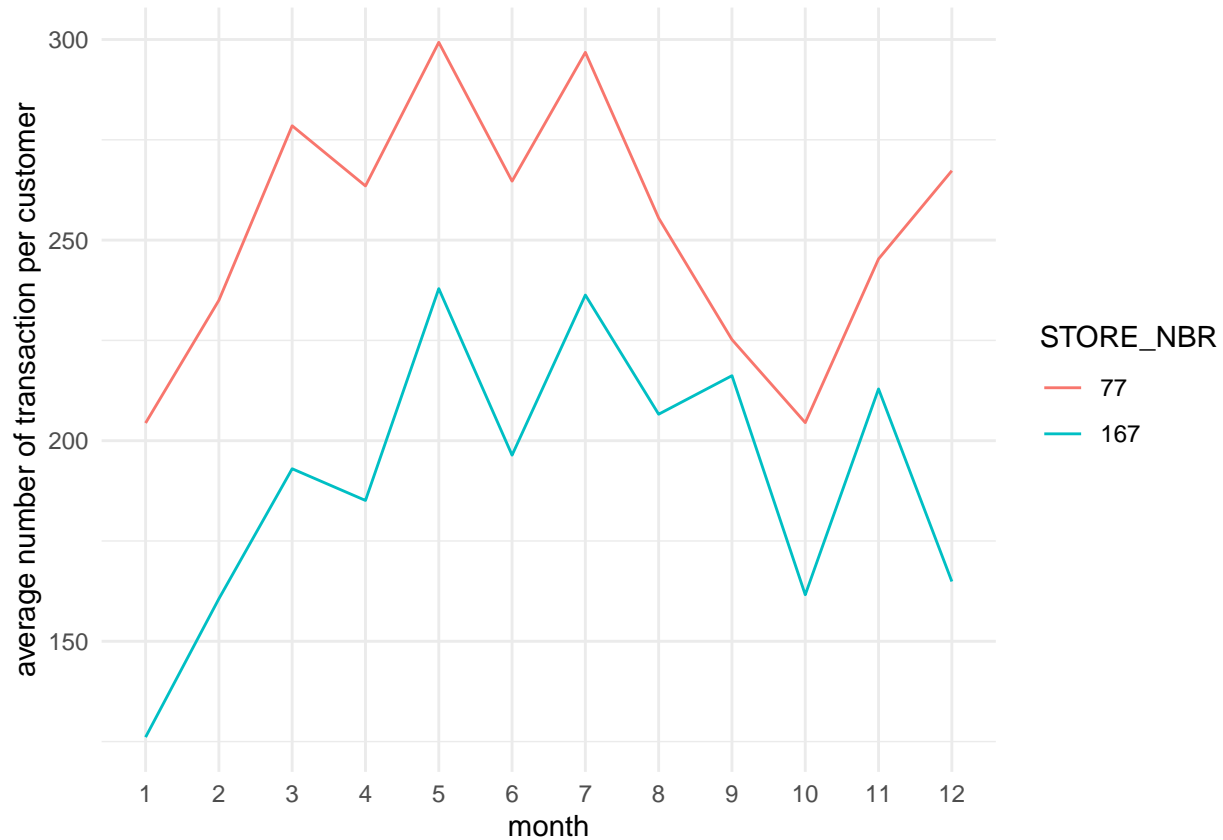
```
distance_avg <- result_cor_88_pearson[,.(distance = sum((Correlation - sum_cor_central$Correlation)^2))
tibble(distance_avg[order(distance)])
```

```
## # A tibble: 260 x 2
```

```
##      STORE_NBR distance
##      <int>    <dbl>
## 1         88      0
## 2        229    0.486
## 3        257    0.694
## 4        228    0.980
## 5        159    1.04
## 6         26    1.17
## 7        178    1.18
## 8        191    1.25
## 9        140    1.36
## 10       187    1.57
## # i 250 more rows
```

We can get the store 167 has the minmum distance to store 77. Then we can compare their metrics:

```
ggplot(data = measureOverTime[STORE_NBR %in% c(77,167),]) +geom_line(aes(x= as.factor(month), y = totalSales))
```



we can find the store 77's total sales are more than store 167, then we can compare more metrics:

```
tibble(measureOverTime[STORE_NBR %in% c(77,167),])
```

```
## # A tibble: 24 x 7
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit     N
##   <dbl>   <int>    <dbl>    <int>      <int>      <dbl> <int>
## 1     1       77    204.      35        65        3.14   12
```

```
## 2      1      167      126.      30      40      3.15      12
## 3      2       77      235      45      74      3.18      12
## 4      2      167      161.      35      50      3.21      12
## 5      3       77      278.      50      82      3.40      12
## 6      3      167      193      44      53      3.64      12
## 7      4       77      263.      47      78      3.38      12
## 8      4      167      185.      42      55      3.37      12
## 9      5      167      238.      55      72      3.30      12
## 10     5       77      299.      55      84      3.56      12
## # i 14 more rows
```

```
measureOverTime[STORE_NBR %in% c(77,167),][order(month)]
```

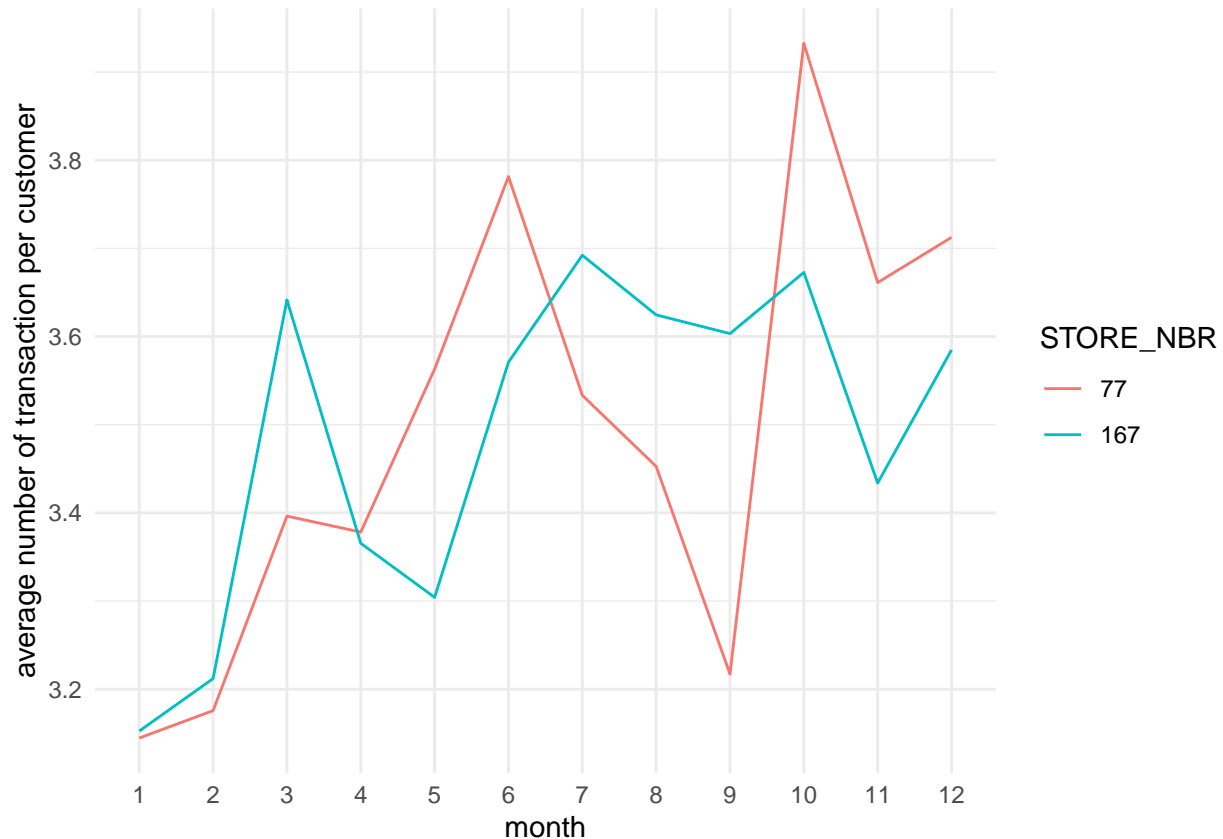
```
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <num>      <int>      <num>      <int>      <int>      <num> <int>
## 1:      1       77      204.4      35      65      3.144615      12
## 2:      1      167      126.1      30      40      3.152500      12
## 3:      2       77      235.0      45      74      3.175676      12
## 4:      2      167      160.6      35      50      3.212000      12
## 5:      3       77      278.5      50      82      3.396341      12
## 6:      3      167      193.0      44      53      3.641509      12
## 7:      4       77      263.5      47      78      3.378205      12
## 8:      4      167      185.1      42      55      3.365455      12
## 9:      5      167      237.9      55      72      3.304167      12
## 10:     5       77      299.3      55      84      3.563095      12
## 11:     6       77      264.7      41      70      3.781429      12
## 12:     6      167      196.4      41      55      3.570909      12
## 13:     7       77      296.8      51      84      3.533333      12
## 14:     7      167      236.3      53      64      3.692188      12
## 15:     8       77      255.5      47      74      3.452703      12
## 16:     8      167      206.6      37      57      3.624561      12
## 17:     9       77      225.2      42      70      3.217143      12
## 18:     9      167      216.2      45      60      3.603333      12
## 19:    10       77      204.5      37      52      3.932692      12
## 20:    10      167      161.6      34      44      3.672727      12
## 21:    11      167      212.9      46      62      3.433871      12
## 22:    11       77      245.3      41      67      3.661194      12
## 23:    12      167      164.9      40      46      3.584783      12
## 24:    12       77      267.3      46      72      3.712500      12
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
```

```
tibble(measureOverTime)
```

```
## # A tibble: 3,169 x 7
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <dbl>      <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1      1         1      155.      35      42      3.69      12
## 2      1         2      163.      43      49      3.32      12
## 3      1         3     1052.     102     236      4.46      12
## 4      1         4     1525     134     335      4.55      12
## 5      1         5      838      92     236      3.55      12
## 6      1         7      975.     102     221      4.41      12
## 7      1        10      879.     102     252      3.49      12
```

```
## 8      1      15      874.      99      248      3.52      12
## 9      1      17      365.      39      83      4.39      12
## 10     1      19      992      111     224      4.43      12
## # i 3,159 more rows
```

```
ggplot(data = measureOverTime[STORE_NBR %in% c(77,167),]) +geom_line(aes(x= as.factor(month), y = avgPr
```



the average sales are similar

```
tibble(measureOverTime[STORE_NBR %in% c(77,167),])
```

```
## # A tibble: 24 x 7
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit   N
##   <dbl>   <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1     1       77      204.        35         65       3.14    12
## 2     1      167      126.        30         40       3.15    12
## 3     2       77      235.        45         74       3.18    12
## 4     2      167      161.        35         50       3.21    12
## 5     3       77      278.        50         82       3.40    12
## 6     3      167      193.        44         53       3.64    12
## 7     4       77      263.        47         78       3.38    12
## 8     4      167      185.        42         55       3.37    12
## 9     5      167      238.        55         72       3.30    12
## 10    5       77      299.        55         84       3.56    12
## # i 14 more rows
```

```
measureOverTime[STORE_NBR %in% c(77,167),][order(month)]
```

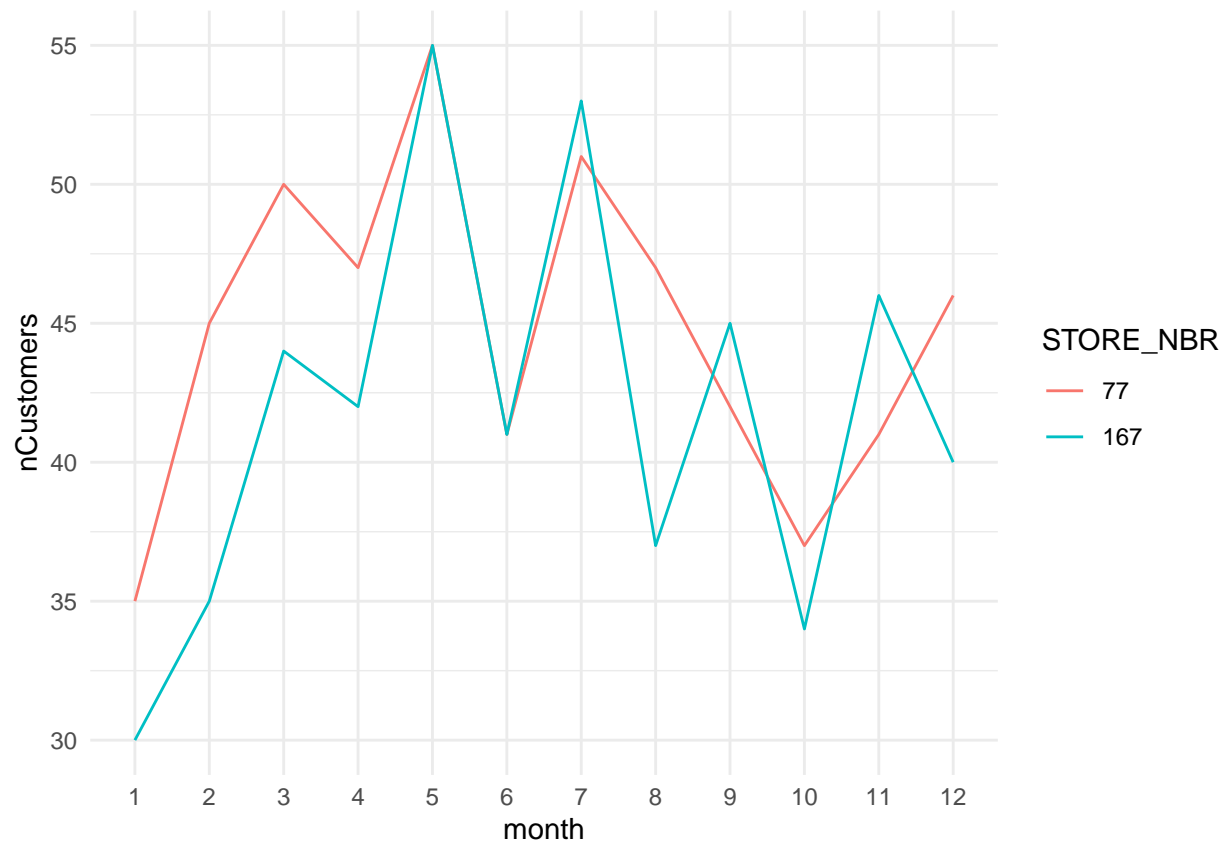
```
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <num>      <int>      <num>      <int>      <int>      <num> <int>
## 1:      1         77      204.4         35         65      3.144615    12
## 2:      1        167      126.1         30         40      3.152500    12
## 3:      2         77      235.0         45         74      3.175676    12
## 4:      2        167      160.6         35         50      3.212000    12
## 5:      3         77      278.5         50         82      3.396341    12
## 6:      3        167      193.0         44         53      3.641509    12
## 7:      4         77      263.5         47         78      3.378205    12
## 8:      4        167      185.1         42         55      3.365455    12
## 9:      5        167      237.9         55         72      3.304167    12
##10:      5         77      299.3         55         84      3.563095    12
##11:      6         77      264.7         41         70      3.781429    12
##12:      6        167      196.4         41         55      3.570909    12
##13:      7         77      296.8         51         84      3.533333    12
##14:      7        167      236.3         53         64      3.692188    12
##15:      8         77      255.5         47         74      3.452703    12
##16:      8        167      206.6         37         57      3.624561    12
##17:      9         77      225.2         42         70      3.217143    12
##18:      9        167      216.2         45         60      3.603333    12
##19:     10         77      204.5         37         52      3.932692    12
##20:     10        167      161.6         34         44      3.672727    12
##21:     11        167      212.9         46         62      3.433871    12
##22:     11         77      245.3         41         67      3.661194    12
##23:     12        167      164.9         40         46      3.584783    12
##24:     12         77      267.3         46         72      3.712500    12
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
```

```
tibble(measureOverTime)
```

```
## # A tibble: 3,169 x 7
```

```
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <dbl>      <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1      1         1      155.         35         42      3.69    12
## 2      1         2      163.         43         49      3.32    12
## 3      1         3     1052.        102        236      4.46    12
## 4      1         4     1525         134        335      4.55    12
## 5      1         5      838         92        236      3.55    12
## 6      1         7      975.        102        221      4.41    12
## 7      1        10      879.        102        252      3.49    12
## 8      1        15      874.         99        248      3.52    12
## 9      1        17      365.         39         83      4.39    12
##10     1        19      992         111        224      4.43    12
## # i 3,159 more rows
```

```
ggplot(data = measureOverTime[STORE_NBR %in% c(77,167),]) +geom_line(aes(x= as.factor(month), y = nCust
```

```
tibble(measureOverTime[STORE_NBR %in% c(77,167),])
```

```
## # A tibble: 24 x 7
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit N
##   <dbl>   <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1     1       77      204.        35         65        3.14    12
## 2     1      167      126.        30         40        3.15    12
## 3     2       77      235.        45         74        3.18    12
## 4     2      167      161.        35         50        3.21    12
## 5     3       77      278.        50         82        3.40    12
## 6     3      167      193.        44         53        3.64    12
## 7     4       77      263.        47         78        3.38    12
## 8     4      167      185.        42         55        3.37    12
## 9     5      167      238.        55         72        3.30    12
## 10    5       77      299.        55         84        3.56    12
## # i 14 more rows
```

```
measureOverTime[STORE_NBR %in% c(77,167),][order(month)]
```

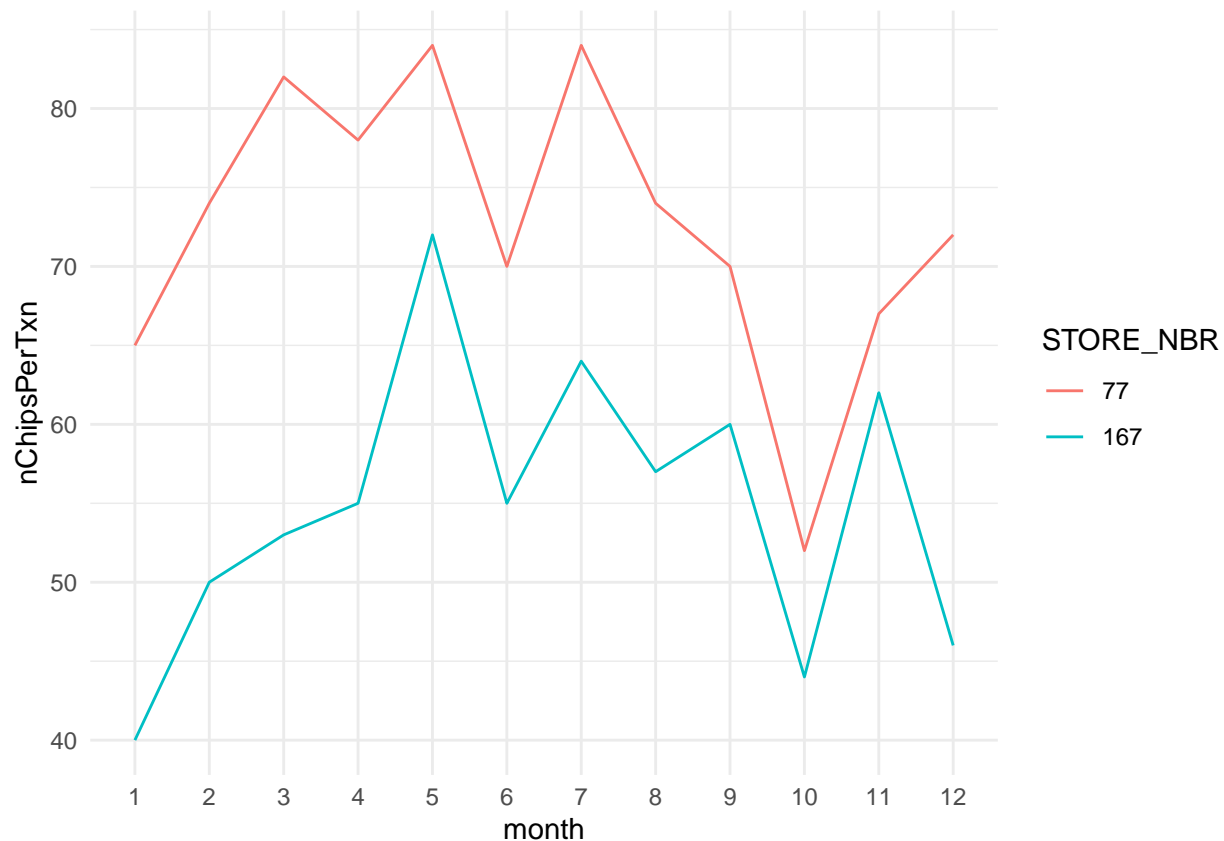
```
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit N
##   <num>   <int>      <num>      <int>      <int>      <num> <int>
## 1:     1       77      204.4        35         65      3.144615    12
## 2:     1      167      126.1        30         40      3.152500    12
## 3:     2       77      235.0        45         74      3.175676    12
## 4:     2      167      160.6        35         50      3.212000    12
```

```
## 5:      3      77      278.5      50      82      3.396341      12
## 6:      3     167     193.0      44      53      3.641509      12
## 7:      4      77     263.5      47      78      3.378205      12
## 8:      4     167     185.1      42      55      3.365455      12
## 9:      5     167     237.9      55      72      3.304167      12
## 10:     5      77     299.3      55      84      3.563095      12
## 11:     6      77     264.7      41      70      3.781429      12
## 12:     6     167     196.4      41      55      3.570909      12
## 13:     7      77     296.8      51      84      3.533333      12
## 14:     7     167     236.3      53      64      3.692188      12
## 15:     8      77     255.5      47      74      3.452703      12
## 16:     8     167     206.6      37      57      3.624561      12
## 17:     9      77     225.2      42      70      3.217143      12
## 18:     9     167     216.2      45      60      3.603333      12
## 19:    10      77     204.5      37      52      3.932692      12
## 20:    10     167     161.6      34      44      3.672727      12
## 21:    11     167     212.9      46      62      3.433871      12
## 22:    11      77     245.3      41      67      3.661194      12
## 23:    12     167     164.9      40      46      3.584783      12
## 24:    12      77     267.3      46      72      3.712500      12
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
```

```
tibble(measureOverTime)
```

```
## # A tibble: 3,169 x 7
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <dbl>      <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1      1          1      155.         35         42         3.69      12
## 2      1          2      163.         43         49         3.32      12
## 3      1          3     1052.        102        236         4.46      12
## 4      1          4     1525         134        335         4.55      12
## 5      1          5      838          92        236         3.55      12
## 6      1          7      975.         102        221         4.41      12
## 7      1         10      879.         102        252         3.49      12
## 8      1         15      874.          99        248         3.52      12
## 9      1         17      365.          39         83         4.39      12
## 10     1         19      992         111        224         4.43      12
## # i 3,159 more rows
```

```
ggplot(data = measureOverTime[STORE_NBR %in% c(77,167),]) +geom_line(aes(x= as.factor(month), y = nChipsPerTxn))
```



we found the answer is that the customer from store 77 tend to buy more chips every times which result in the total sales much larger than store 167 (although the total number of customer is similar).

By correlations, we can also get that store 132 is similar with store 86 comprehensively. but we can compare the difference:

```
tibble(measureOverTime[STORE_NBR %in% c(86,132),])
```

```
## # A tibble: 24 x 7
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit    N
##   <dbl>   <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1     1       86      841.         94        260        3.24    12
## 2     1      132       29.6          5          9        3.29    12
## 3     2       86      913.        107        277        3.30    12
## 4     2      132       15.4          3          6        2.57    12
## 5     3       86     1027.        115        284        3.62    12
## 6     3      132       72.8          9         18        4.04    12
## 7     4       86      848.        105        254        3.34    12
## 8     4      132        22          3          6        3.67    12
## 9     5       86      889.        104        259        3.43    12
## 10    5      132       75.1         11         23        3.27    12
## # i 14 more rows
```

```
measureOverTime[STORE_NBR %in% c(86,132),][order(month)]
```

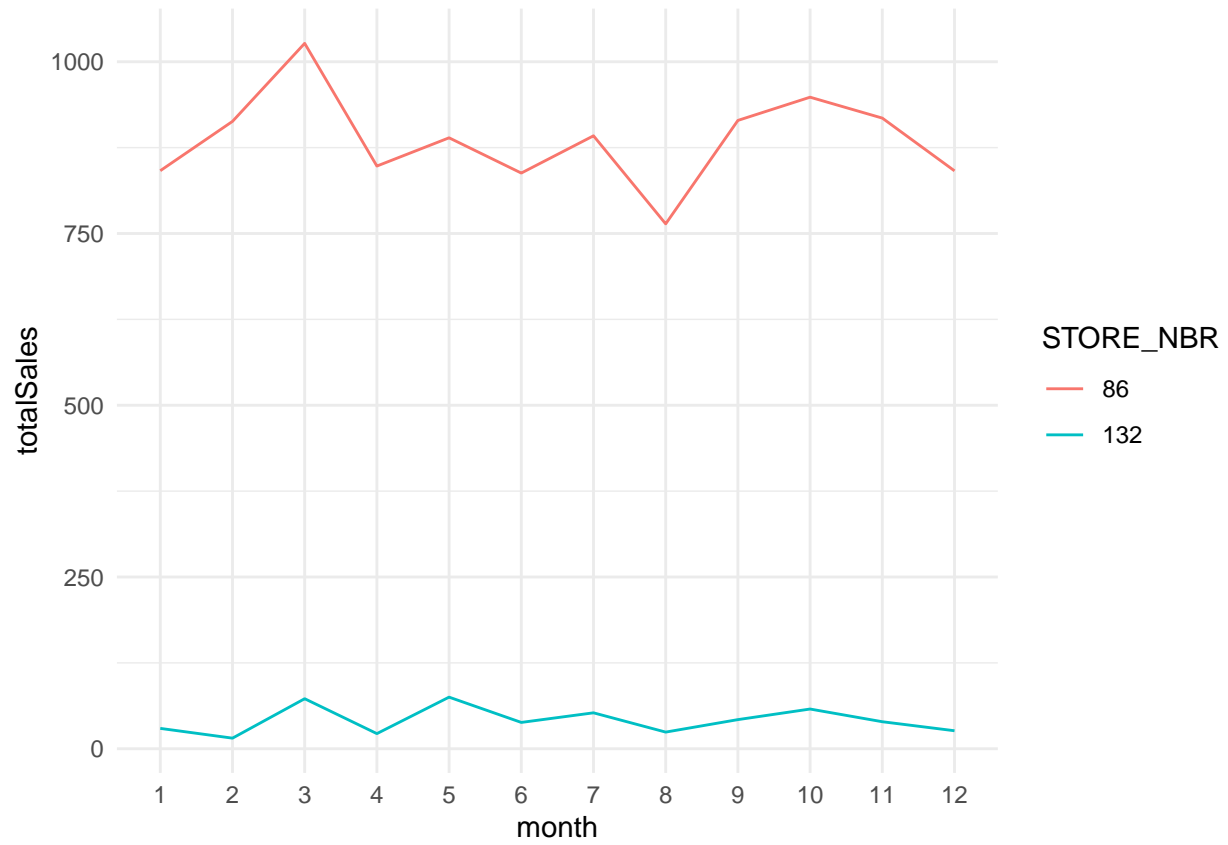
```
##   month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit    N
```

```
##      <num>      <int>      <num>      <int>      <int>      <num> <int>
## 1:      1      86      841.40      94      260      3.236154  12
## 2:      1     132      29.60       5       9      3.288889  12
## 3:      2      86     913.20     107     277      3.296751  12
## 4:      2     132      15.40       3       6      2.566667  12
## 5:      3      86     1026.80     115     284      3.615493  12
## 6:      3     132      72.80       9      18      4.044444  12
## 7:      4      86     848.20     105     254      3.339370  12
## 8:      4     132      22.00       3       6      3.666667  12
## 9:      5      86     889.30     104     259      3.433591  12
## 10:     5     132      75.10      11     23      3.265217  12
## 11:     6      86     838.00     98     238      3.521008  12
## 12:     6     132      38.30       5       9      4.255556  12
## 13:     7      86     892.20     99     251      3.554582  12
## 14:     7     132      52.30      11     17      3.076471  12
## 15:     8     132      24.20       4       7      3.457143  12
## 16:     8      86     764.05     94     215      3.553721  12
## 17:     9      86     914.60    103     258      3.544961  12
## 18:     9     132      42.40       5      12      3.533333  12
## 19:    10      86     948.40    109     276      3.436232  12
## 20:    10     132      57.80      10     18      3.211111  12
## 21:    11      86     918.00    100     254      3.614173  12
## 22:    11     132      39.30       6      11      3.572727  12
## 23:    12      86     841.20     98     240      3.505000  12
## 24:    12     132      26.20       4       8      3.275000  12
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
```

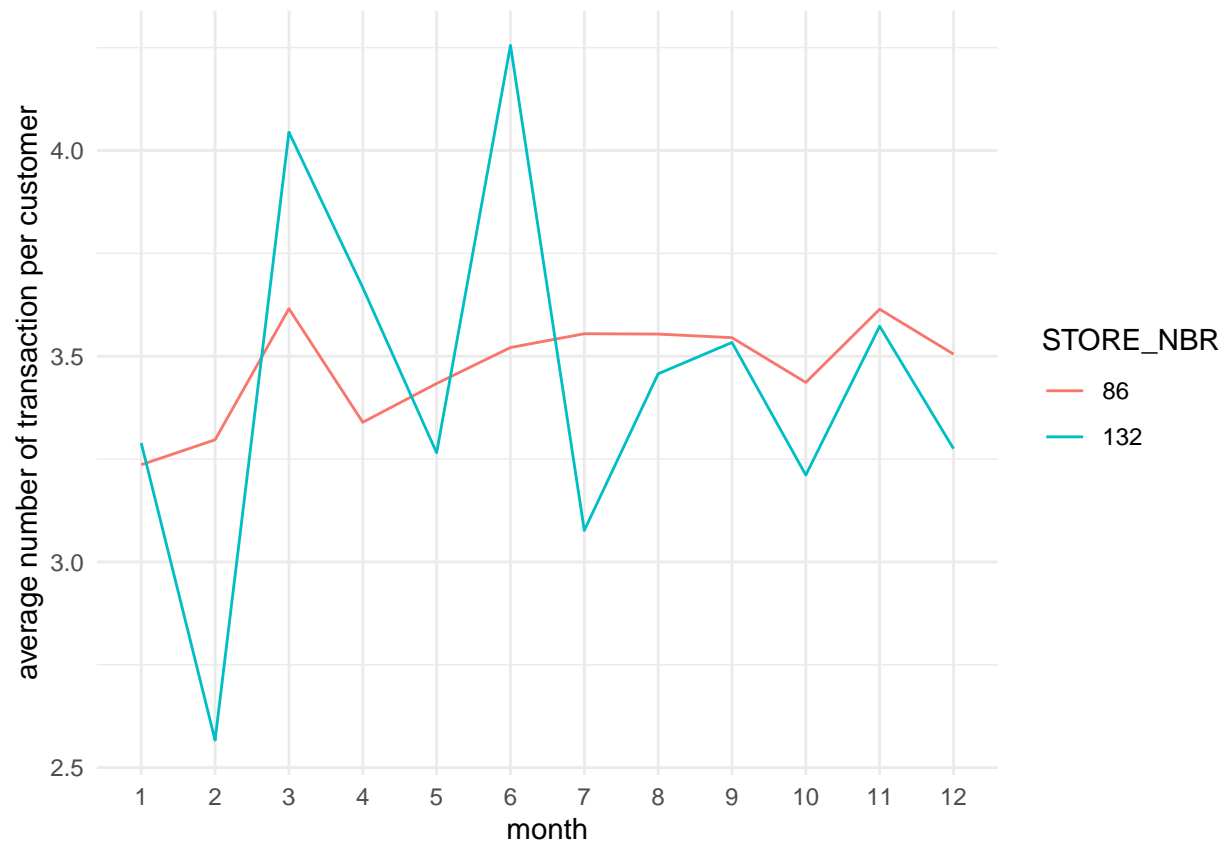
```
tibble(measureOverTime)
```

```
## # A tibble: 3,169 x 7
##      month STORE_NBR totalSales nCustomers nChipsPerTxn avgPricePerUnit      N
##      <dbl>      <int>      <dbl>      <int>      <int>      <dbl> <int>
## 1      1      1      155.       35       42      3.69    12
## 2      1      2      163.       43       49      3.32    12
## 3      1      3     1052.     102     236      4.46    12
## 4      1      4     1525      134     335      4.55    12
## 5      1      5      838       92     236      3.55    12
## 6      1      7      975.     102     221      4.41    12
## 7      1     10      879.     102     252      3.49    12
## 8      1     15      874.      99     248      3.52    12
## 9      1     17      365.      39      83      4.39    12
## 10     1     19      992     111     224      4.43    12
## # i 3,159 more rows
```

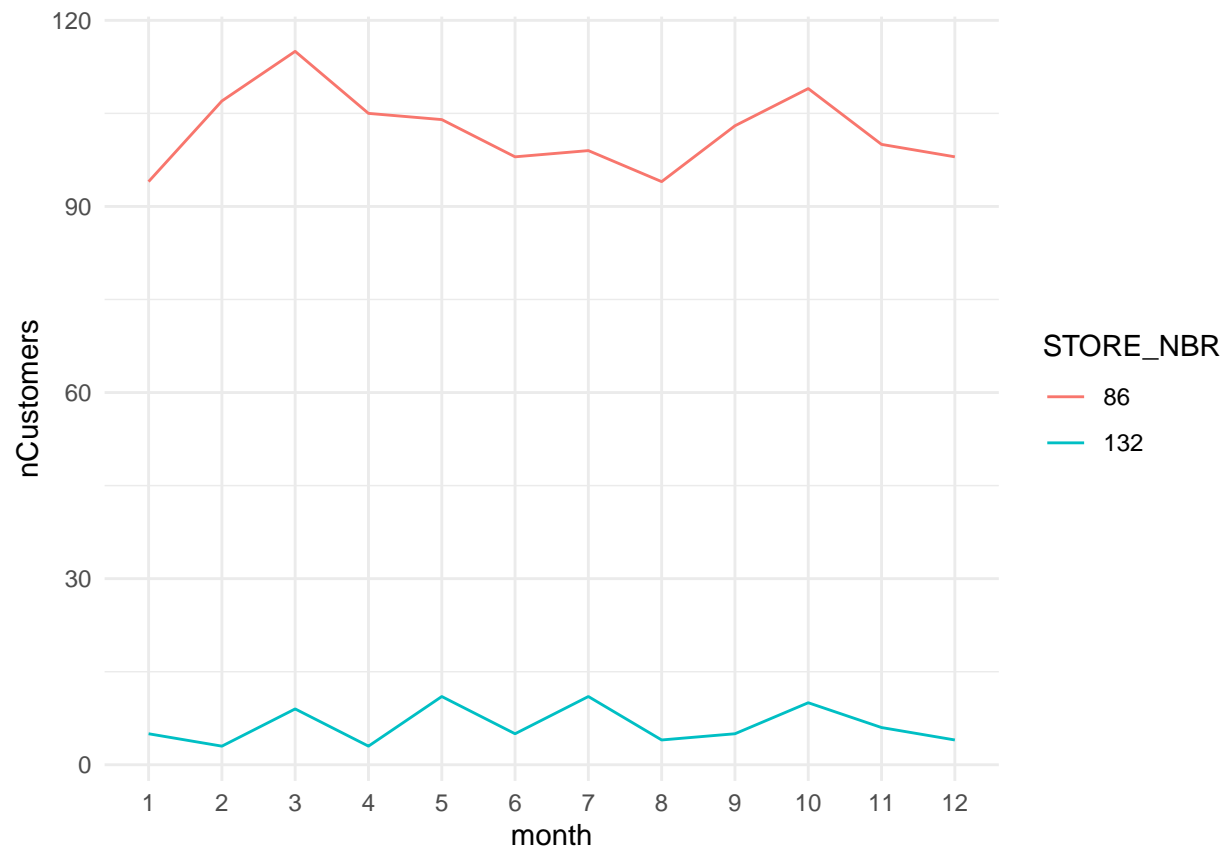
```
ggplot(data = measureOverTime[STORE_NBR %in% c(86,132),]) +geom_line(aes(x= as.factor(month), y = totalSales))
```



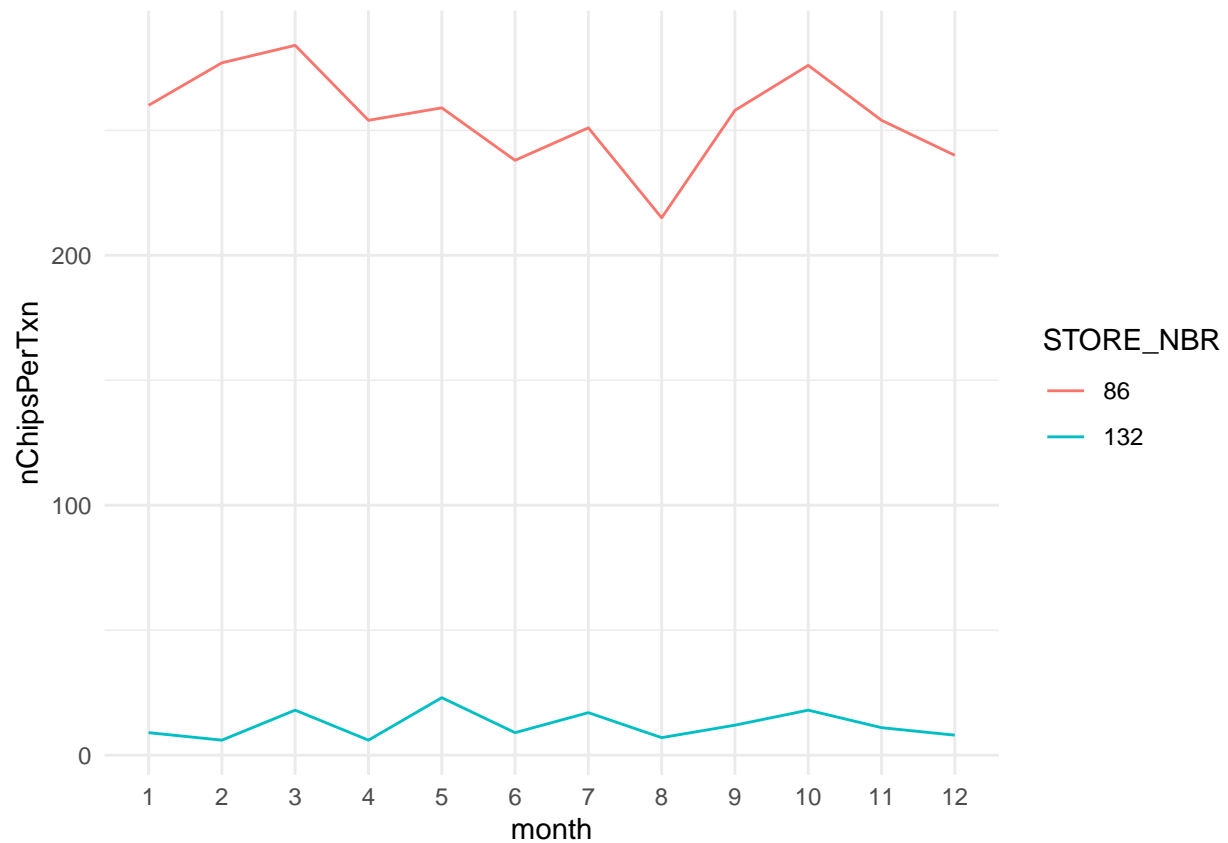
```
ggplot(data = measureOverTime[STORE_NBR %in% c(86,132),]) +geom_line(aes(x= as.factor(month), y = avgPr
```



```
ggplot(data = measureOverTime[STORE_NBR %in% c(86,132),]) +geom_line(aes(x= as.factor(month), y = nCust
```



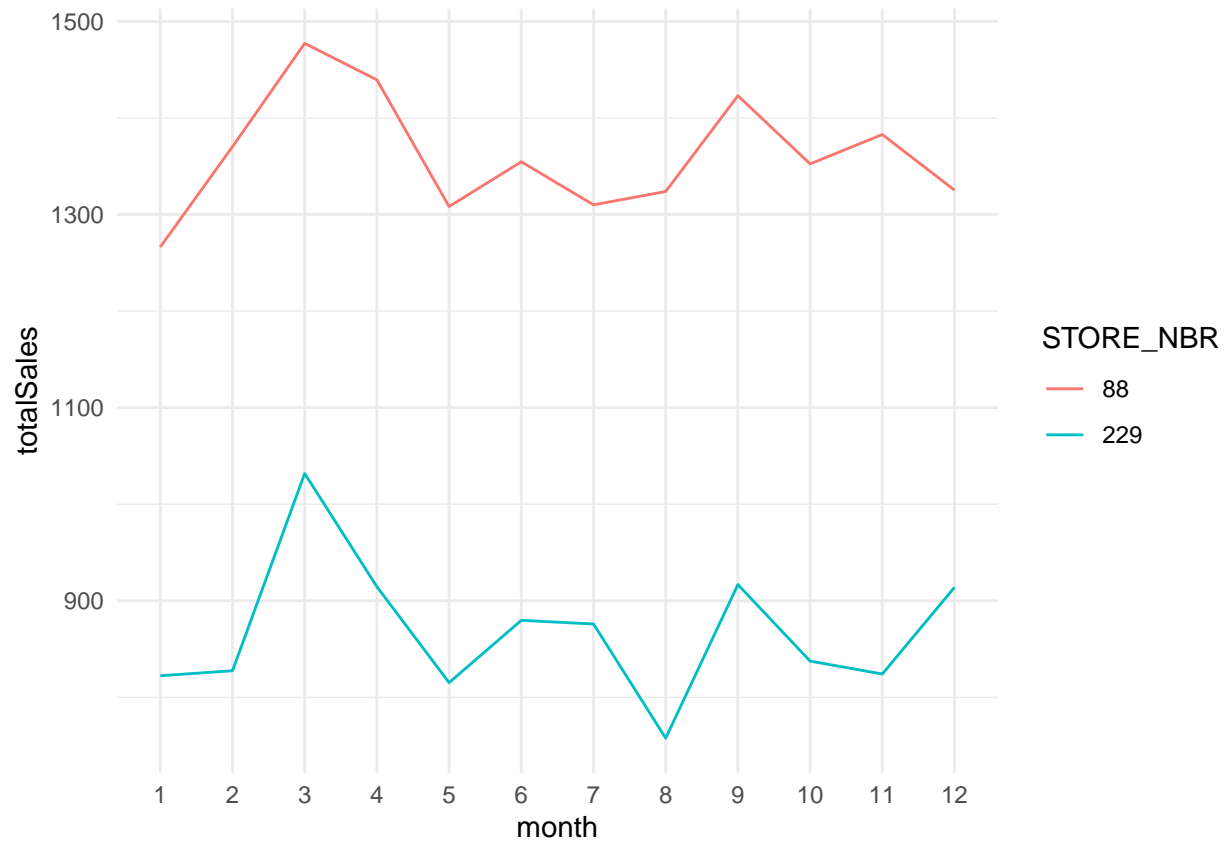
```
ggplot(data = measureOverTime[STORE_NBR %in% c(86,132),]) +geom_line(aes(x= as.factor(month), y = nChip
```



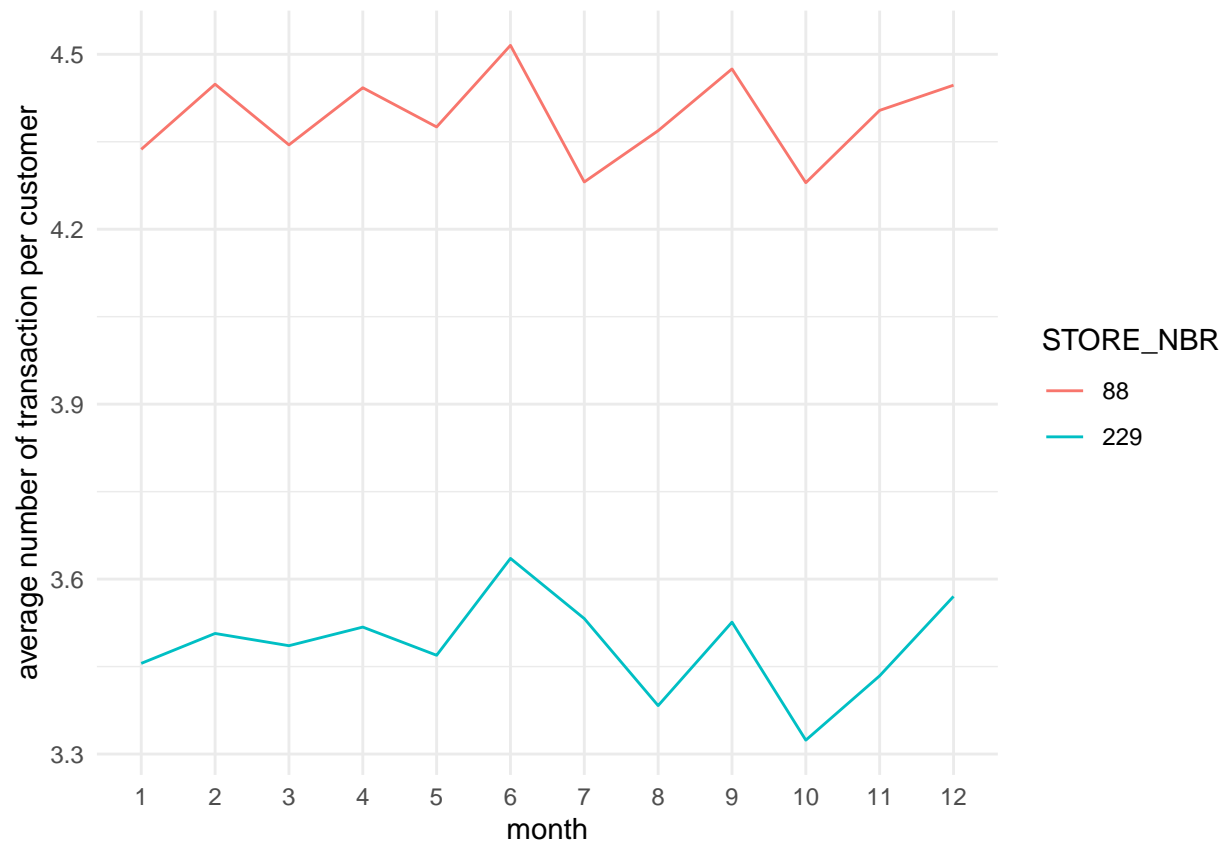
the differences between two stores are from the total number of customer and the number of chips purchased every time.

Finally for store 88 and 229

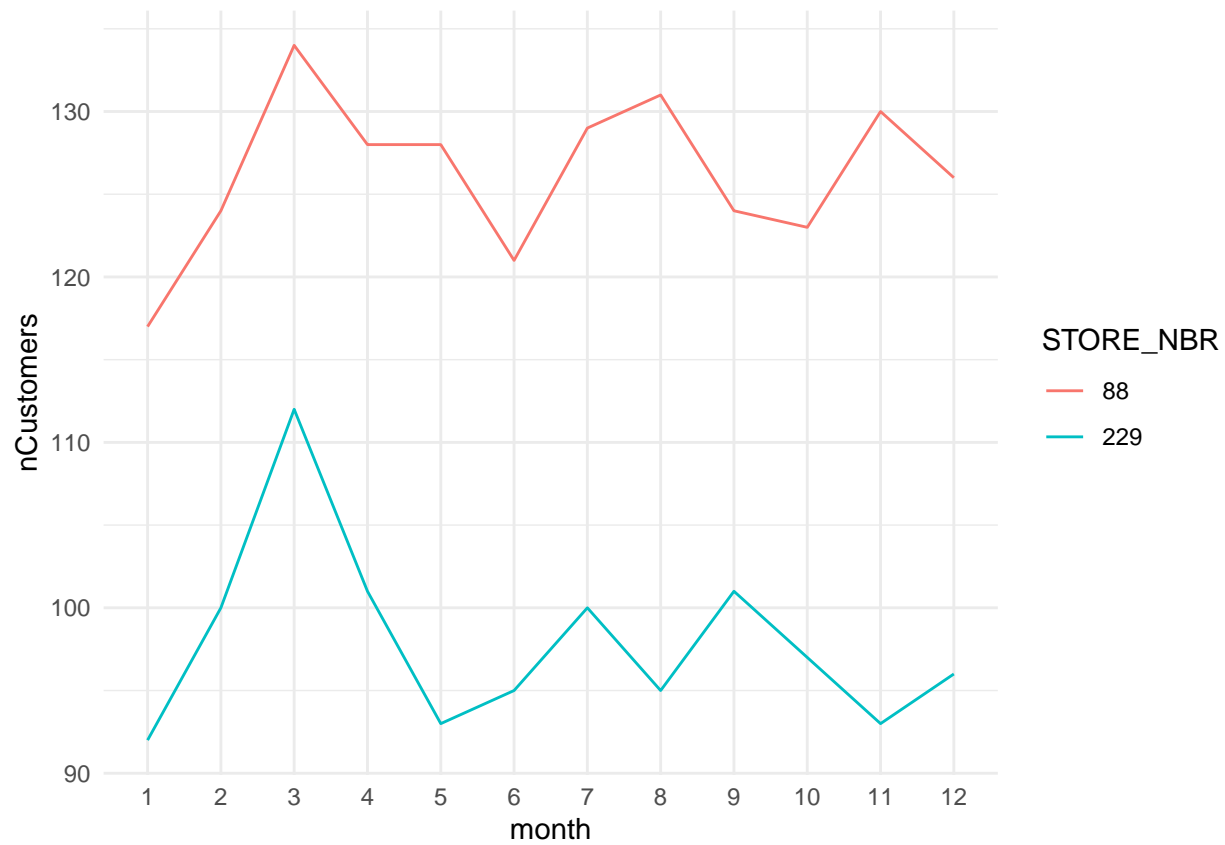
```
ggplot(data = measureOverTime[STORE_NBR %in% c(88,229),]) +geom_line(aes(x= as.factor(month), y = total
```

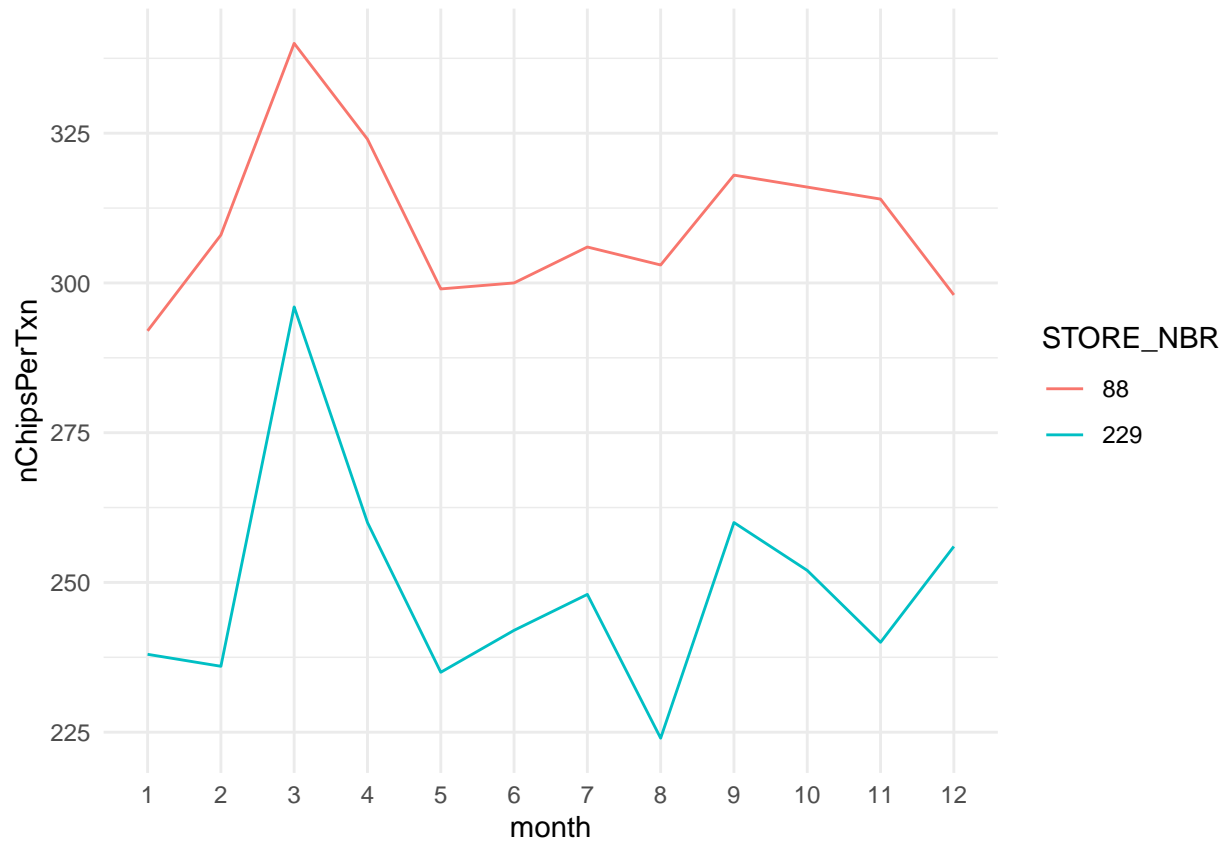
```
ggplot(data = measureOverTime[STORE_NBR %in% c(88,229),]) +geom_line(aes(x= as.factor(month), y = avgPr
```



```
ggplot(data = measureOverTime[STORE_NBR %in% c(88,229),]) +geom_line(aes(x= as.factor(month), y = nCust
```



```
ggplot(data = measureOverTime[STORE_NBR %in% c(88,229),]) +geom_line(aes(x= as.factor(month), y = nChip
```



the store metrics are much better than store 229.

Conclusion

- three stores have their own pros and cons compared to the correlated stores in the whole year.
- the most effective one is store 86, it shows this store has the highest improvement compared to similar stores.
- the store 77 performance is little better than store 167 and it is from the improvement of number of chips per transaction.
- the store 88 have a comprehensive improvement compared with store 229 in all metrics.