

# Group\_descrip\_lin2.R

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Fri Jun 24 18:27:36 2016

```
## 2016年暑期课程设计
## 问题: Grupo Bimbo Inventory Demand
## 宾堡集团的库存需求
## 最大限度地提高销售和最大限度地减少烘焙食品的退回
## Daitu
## start:2016.06.21
## 参考借鉴kaggle上的公开程序
```

```
setwd("/Users/Daitu/数据分析/kaggle/Grupo Bimbo")
getwd()
```

```
## [1] "/Users/daitu/数据分析/kaggle/Grupo Bimbo"
```

```
## 加载包
```

```
library(data.table)
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.2.4
```

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

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

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

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

```
library(treemap)
```

```
## Warning: package 'treemap' was built under R version 3.2.4
```

```
## 读取数据####
## 1:读取训练集
system.time({
  traindata <- fread("train.csv",sep=",",header = TRUE)
})
```

```
##
Read 0.0% of 74180464 rows
Read 4.4% of 74180464 rows
Read 8.8% of 74180464 rows
Read 13.3% of 74180464 rows
Read 17.7% of 74180464 rows
Read 22.1% of 74180464 rows
Read 26.6% of 74180464 rows
Read 31.0% of 74180464 rows
Read 35.4% of 74180464 rows
Read 39.8% of 74180464 rows
Read 44.3% of 74180464 rows
Read 48.6% of 74180464 rows
Read 52.9% of 74180464 rows
Read 57.3% of 74180464 rows
Read 61.6% of 74180464 rows
Read 65.9% of 74180464 rows
Read 70.3% of 74180464 rows
Read 74.6% of 74180464 rows
Read 78.8% of 74180464 rows
Read 83.1% of 74180464 rows
Read 87.4% of 74180464 rows
Read 91.5% of 74180464 rows
Read 95.8% of 74180464 rows
Read 74180464 rows and 11 (of 11) columns from 2.980 GB file in 00:00:30
```

```
##      user      system elapsed
## 26.756      2.868      32.726
```

```
head(traindata)
```

```
##      Semana Agencia_ID Canal_ID Ruta_SAK Cliente_ID Producto_ID
## 1:         3         1110         7    3301       15766       1212
## 2:         3         1110         7    3301       15766       1216
## 3:         3         1110         7    3301       15766       1238
## 4:         3         1110         7    3301       15766       1240
## 5:         3         1110         7    3301       15766       1242
## 6:         3         1110         7    3301       15766       1250
##      Venta_uni_hoy Venta_hoy Dev_uni_proxima Dev_proxima Demanda_uni_equil
## 1:                3    25.14                0            0                3
## 2:                4    33.52                0            0                4
## 3:                4    39.32                0            0                4
## 4:                4    33.52                0            0                4
## 5:                3    22.92                0            0                3
## 6:                5    38.20                0            0                5
```

```
## 2:读取客户名单数据
```

```
cliente_tabla <- fread("cliente_tabla.csv",sep=",",header = TRUE)
head(cliente_tabla)
```

```
##      Cliente_ID      NombreCliente
## 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
```

```
## 3:读取产品名单数据
```

```
producto_tabla <- fread("producto_tabla.csv",sep=",",header = TRUE)
head(producto_tabla)
```

```
##      Producto_ID      NombreProducto
## 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
```

```
## 4:读取城镇和国家（州）数据
```

```
town_state <- fread("town_state.csv",sep=",",header = TRUE)
head(town_state)
```

```
##      Agencia_ID      Town      State
## 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.
```

```
## 数据的描述统计####
```

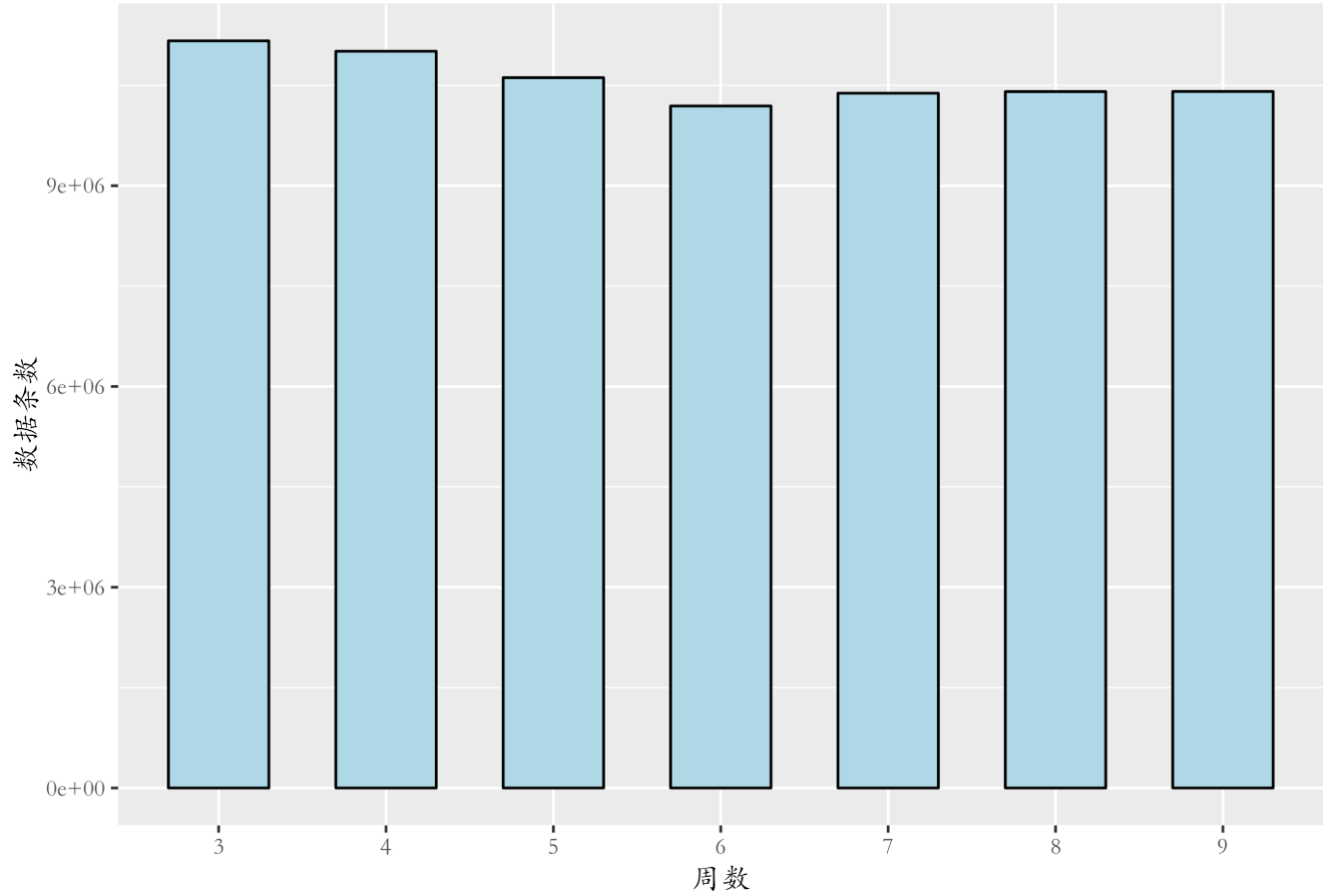
```
## 1:分析数据的周数: Semana
```

```
Semana <- data.frame(table(traindata$Semana))
colnames(Semana) <- c("Semana","Freq")
# 3      4      5      6      7      8      9
# 11165207 11009593 10615397 10191837 10382849 10406868 10408713
```

```
## 条形图
```

```
ggplot(data = Semana,aes(Semana,Freq)) +
  geom_bar(stat = "identity", width = 0.6,fill = "lightblue",colour = "black") +
  theme_grey(base_family = "STKaiti") +
  scale_y_continuous() +
  labs(x="周数",y="数据条数",title = "每周数据的记录数")
```

每周数据的记录数



```
Semana2 <- traindata %>%
  group_by(Semana) %>%
  summarise(Units = sum(Venta_uni_hoy),
            Return_Units = sum(Dev_uni_proxima),
            NetU = sum(Demanda_uni_equil)) %>%
  mutate(Retern_Rate = Return_Units / (Return_Units + Units))
Semana2
```

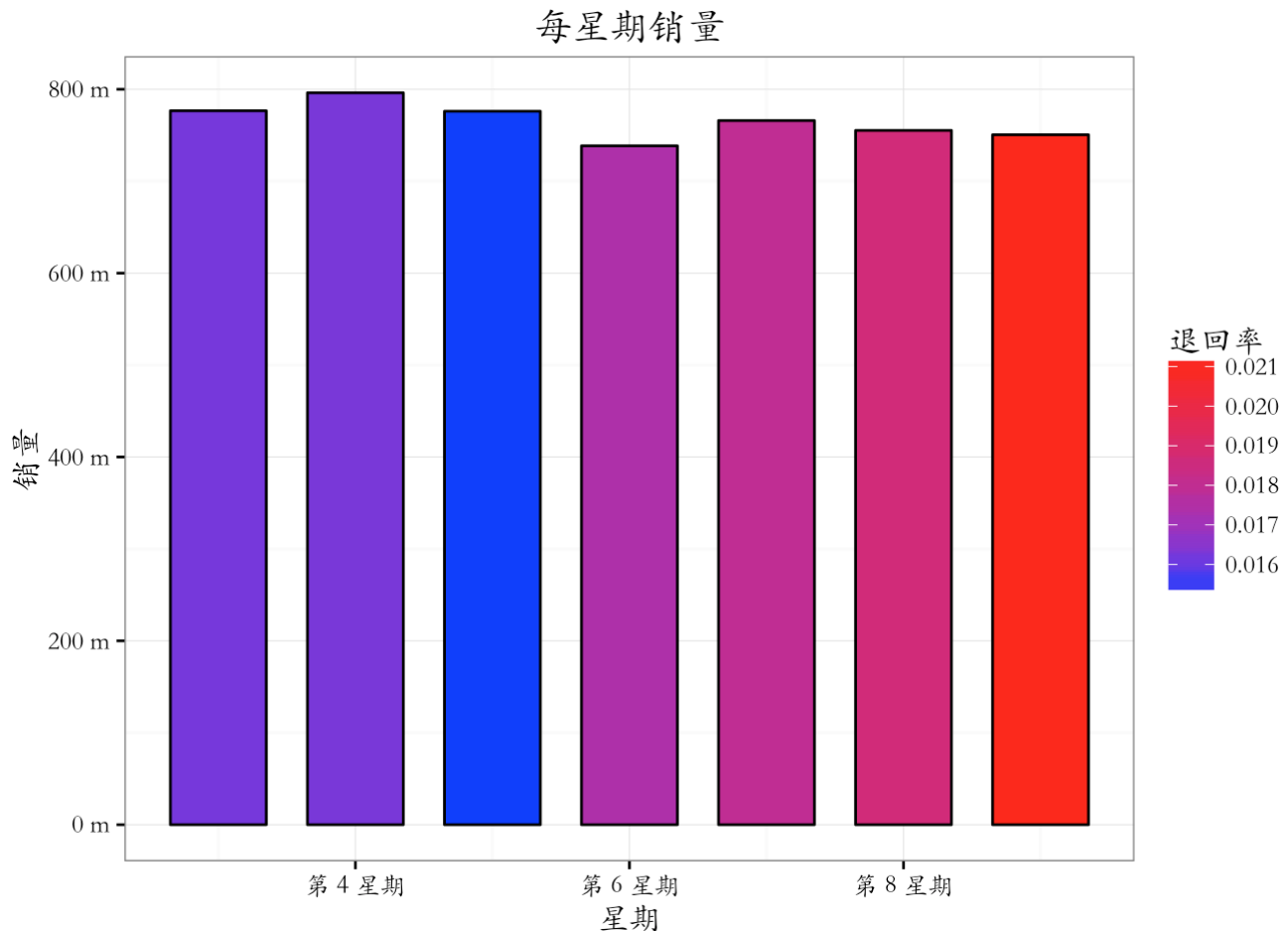
```
## Source: local data table [7 x 5]
##
##   Semana    Units Return_Units    NetU Retern_Rate
##   (int)    (int)      (int)      (int)      (dbl)
## 1      3 78519996    1286055 77664309 0.01611476
## 2      4 80509571    1322852 79618866 0.01616538
## 3      5 78450353    1220967 77610637 0.01532505
## 4      6 74753832    1321510 73851129 0.01737107
## 5      7 77548038    1417450 76597014 0.01795025
## 6      8 76497828    1452369 75525105 0.01863201
## 7      9 75991700    1641371 75054450 0.02114268
```

```
dim(Semana2) # 几个星期的数据
```

```
## [1] 7 5
```

```
## 可视化每周的情况
```

```
ggplot(data = Semana2,aes(Semana,NetU,fill = Return_Rate)) +
  geom_bar(stat = "identity",color = "black",width = 0.7) +
  theme_bw(base_family = "STKaiti") +
  scale_y_continuous(labels=function(x)paste(x/100000, "m"))+
  scale_x_continuous(labels = function(x)paste("第",x,"星期")) +
  scale_fill_gradient(name="退货率", low="blue", high="red")+
  labs(x = "星期",y = "销量",title = "每星期销量")
```



```
## 销售站的数据分析####
```

```
## 2 : 销售站Agencia 和 州State
```

```
agencias <- traindata %>%
  group_by(Agencia_ID) %>% # 数据按照销售战进行分组统计
  summarise(Units = sum(Venta_uni_hoy), # 总结多个值为一个值, units: 本销售站的销量和
            Pesos = sum(Venta_hoy), # 本周的销售量 (比索) 之和
            Return_Units = sum(Dev_uni_proxima), # 下星期的返回量之和
            Return_Pesos = sum(Dev_proxima), # 下星期的返回量 (比索) 之和
            Net = sum(Demanda_uni_equil)) %>% # 调整后的需求和
  mutate(Net_Pesos = Pesos - Return_Pesos, # mutate: 添加新的变量
         Return_Rate = Return_Units / (Units+Return_Units)) %>% # 添加变量退货比率
  arrange(desc(Units)) %>% # 将数据按照变量Units的降序排列
  inner_join(town_state, by="Agencia_ID") # 按照变量Agencia_ID, 连接两个表, return all
rows from x
head(agencias)
```

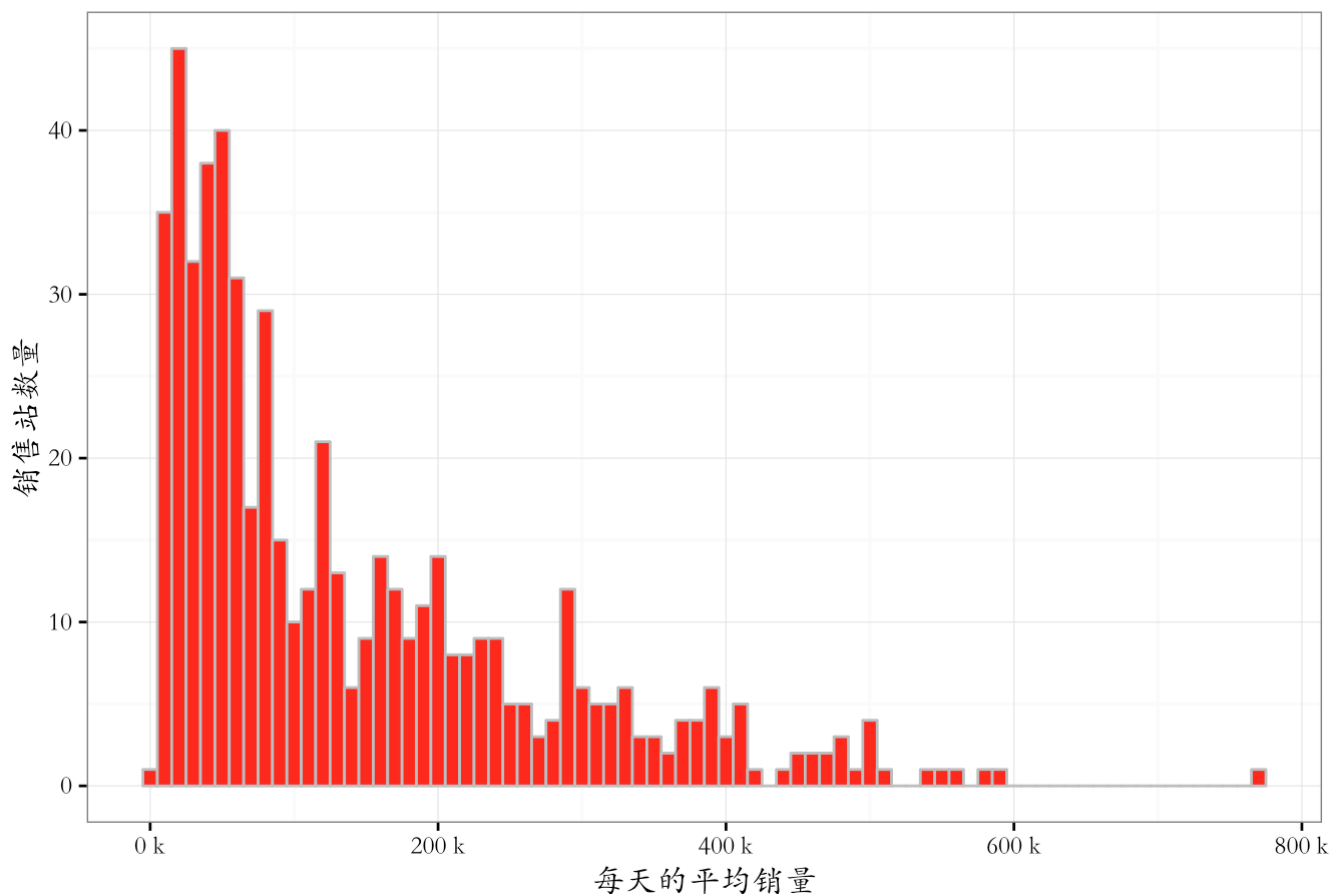
```
##      Agencia_ID    Units      Pesos Return_Units Return_Pesos      Net Net_Pesos
## 1      1110  877675  9274674      39900      214072.8  874523  9060601
## 2      1111 2720400 24070592      25231      264672.4 2701427 23805919
## 3      1112 1959534 16591688      23924      231897.4 1942114 16359791
## 4      1113 1442999 12094484      11865      117754.4 1434414 11976730
## 5      1114 3498170 62420320     150779     2480404.7 3363796 59939915
## 6      1116 3120201 27454358      37022      377100.6 3093985 27077257
##      Return_Rate      Town      State
## 1 0.043484184      2008 AG. LAGO FILT      MÉXICO, D.F.
## 2 0.009189509 2002 AG. AZCAPOTZALCO      MÉXICO, D.F.
## 3 0.012061763      2004 AG. CUAUTITLAN ESTADO DE MÉXICO
## 4 0.008155401      2008 AG. LAGO FILT      MÉXICO, D.F.
## 5 0.041321213 2029 AG. IZTAPALAPA 2      MÉXICO, D.F.
## 6 0.011726128 2011 AG. SAN ANTONIO      MÉXICO, D.F.
```

```
dim(agencias)  # 多少个销售站
```

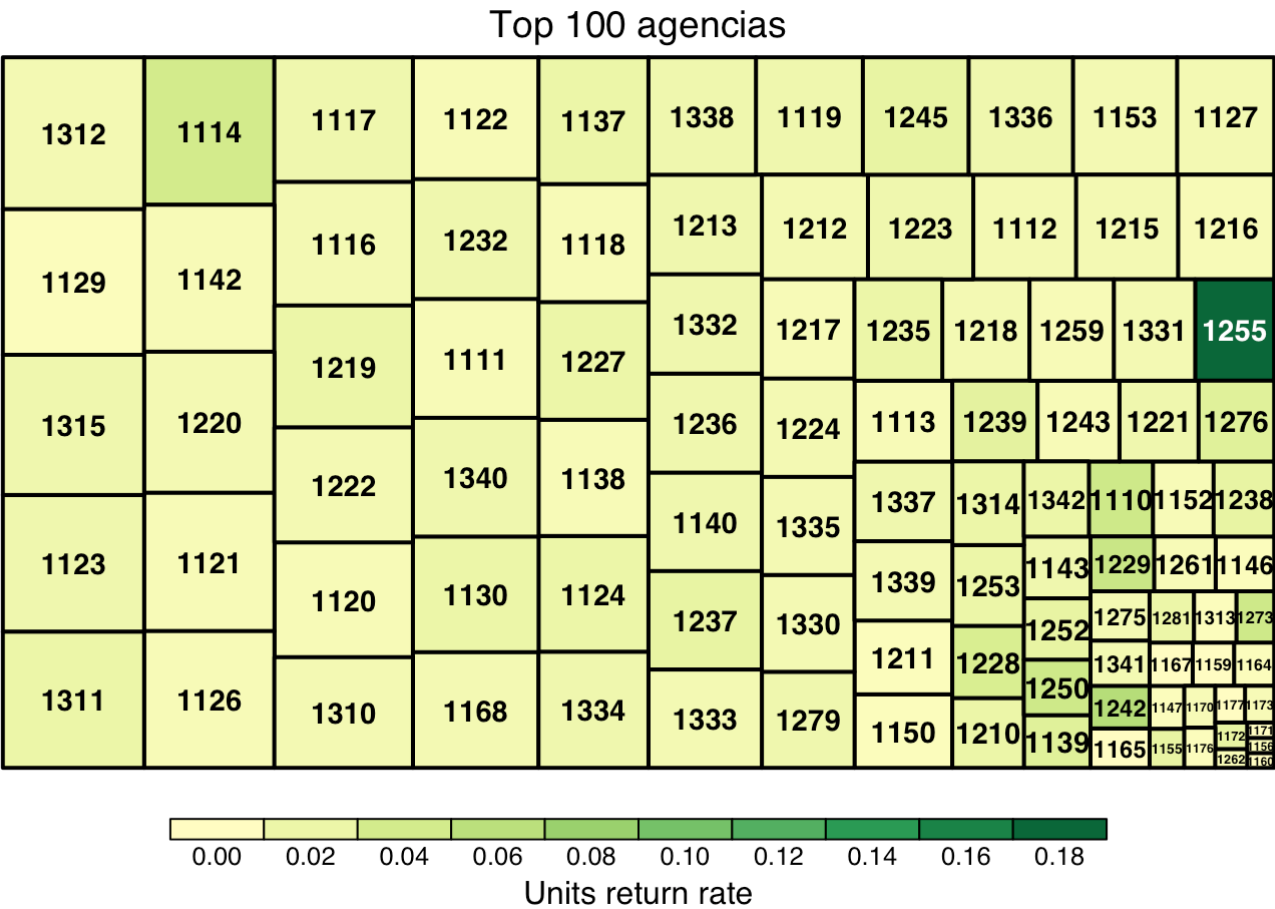
```
## [1] 552  10
```

```
## 可视化x: 每天销量, y: 销售站的数量
ggplot(agencias, aes(x=Units/7))+
  geom_histogram(fill="red", color="gray", binwidth=10000)+ #条形图的宽度为10000
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(labels=function(x)paste(x/1000, "k"))+
  labs(x = "每天的平均销量", y = "销售站数量", title = "销售站的销量")
```

销售站的销量

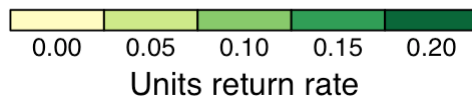
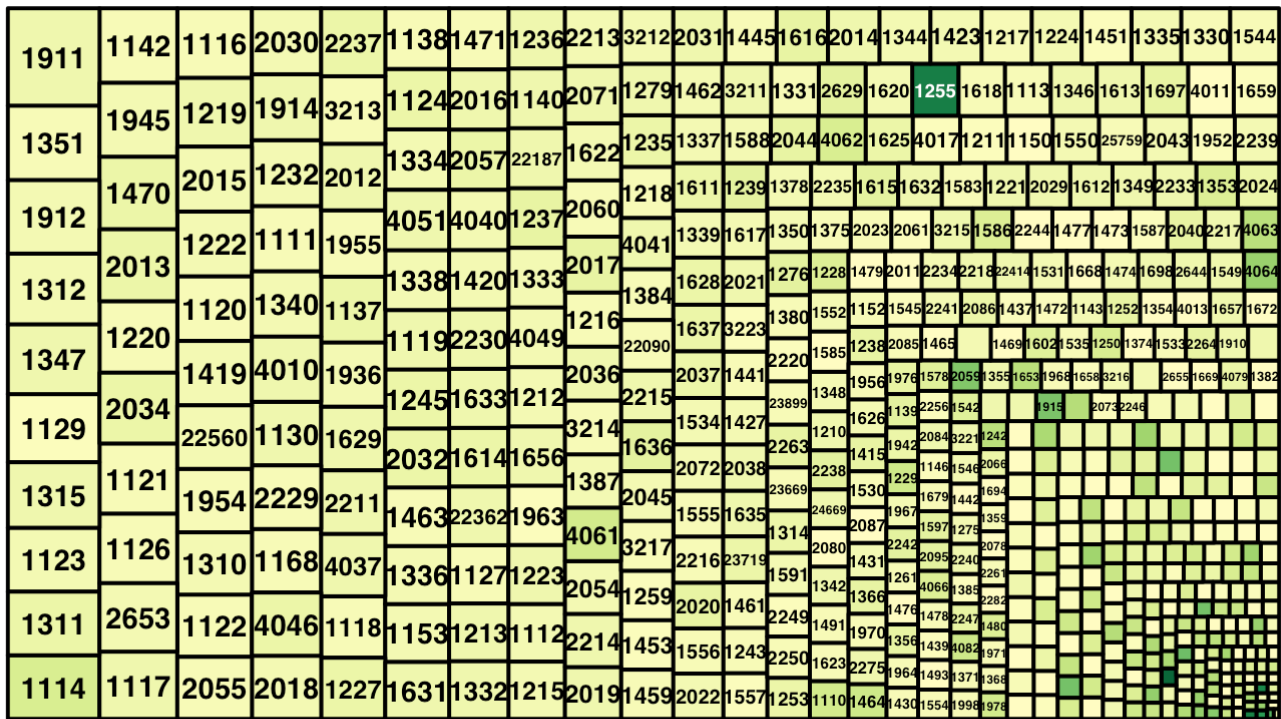


```
## 前100的销售站的销量树图
treemap(agencias[1:100, ],
        index=c("Agencia_ID"), vSize="Units", vColor="Return_Rate",
        type="value", title.legend="Units return rate", title="Top 100 agencias")
```



```
## 所有销售站的销量树图
treemap(agencias,
        index=c("Agencia_ID"), vSize="Units", vColor="Return_Rate",
        type="value", title.legend="Units return rate", title="All agencias")
```

## All agencias



## ## 销售站的历史数据分析

```

agencias_history <- traindata %>%
  group_by(Agencia_ID, Semana) %>% # 数据按照销售站、星期数，进行分组统计
  summarise(Units = sum(Venta_uni_hoy), # 总结多个值为一个值，units: 销量和
            Pesos = sum(Venta_hoy),
            Return_Units = sum(Dev_uni_proxima),
            Return_Pesos = sum(Dev_proxima),
            Net = sum(Demanda_uni_equil)) %>%
  mutate(Net_Pesos = Pesos - Return_Pesos,
         Avg_Pesos = Pesos / Units,
         Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(Agencia_ID, Semana) %>% # 将数据按照变量 销售站、星期数的降序排列
  inner_join(town_state, by="Agencia_ID")
head(agencias_history)

```



```
##      Agencia_ID Semana  Units   Pesos Return_Units Return_Pesos   Net
## 1      1110      3 120285 1296739      3577    29712.03 119951
## 2      1110      4 135788 1385112      5260    29932.82 135327
## 3      1110      5 127420 1345680      3342    26249.25 127077
## 4      1110      6 115255 1239051      5721    26595.87 114865
## 5      1110      7 122955 1297072      6950    34845.26 122513
## 6      1110      8 127277 1345696      8030    35157.98 126735
##      Net_Pesos Avg_Pesos Return_Rate      Town      State
## 1    1267027   10.78055   0.02887891 2008 AG. LAGO FILT MÉXICO, D.F.
## 2    1355179   10.20055   0.03729227 2008 AG. LAGO FILT MÉXICO, D.F.
## 3    1319431   10.56098   0.02555788 2008 AG. LAGO FILT MÉXICO, D.F.
## 4    1212456   10.75052   0.04729037 2008 AG. LAGO FILT MÉXICO, D.F.
## 5    1262227   10.54916   0.05350064 2008 AG. LAGO FILT MÉXICO, D.F.
## 6    1310538   10.57297   0.05934652 2008 AG. LAGO FILT MÉXICO, D.F.
```

```
dim(agencias_history)
```

```
## [1] 3863    12
```

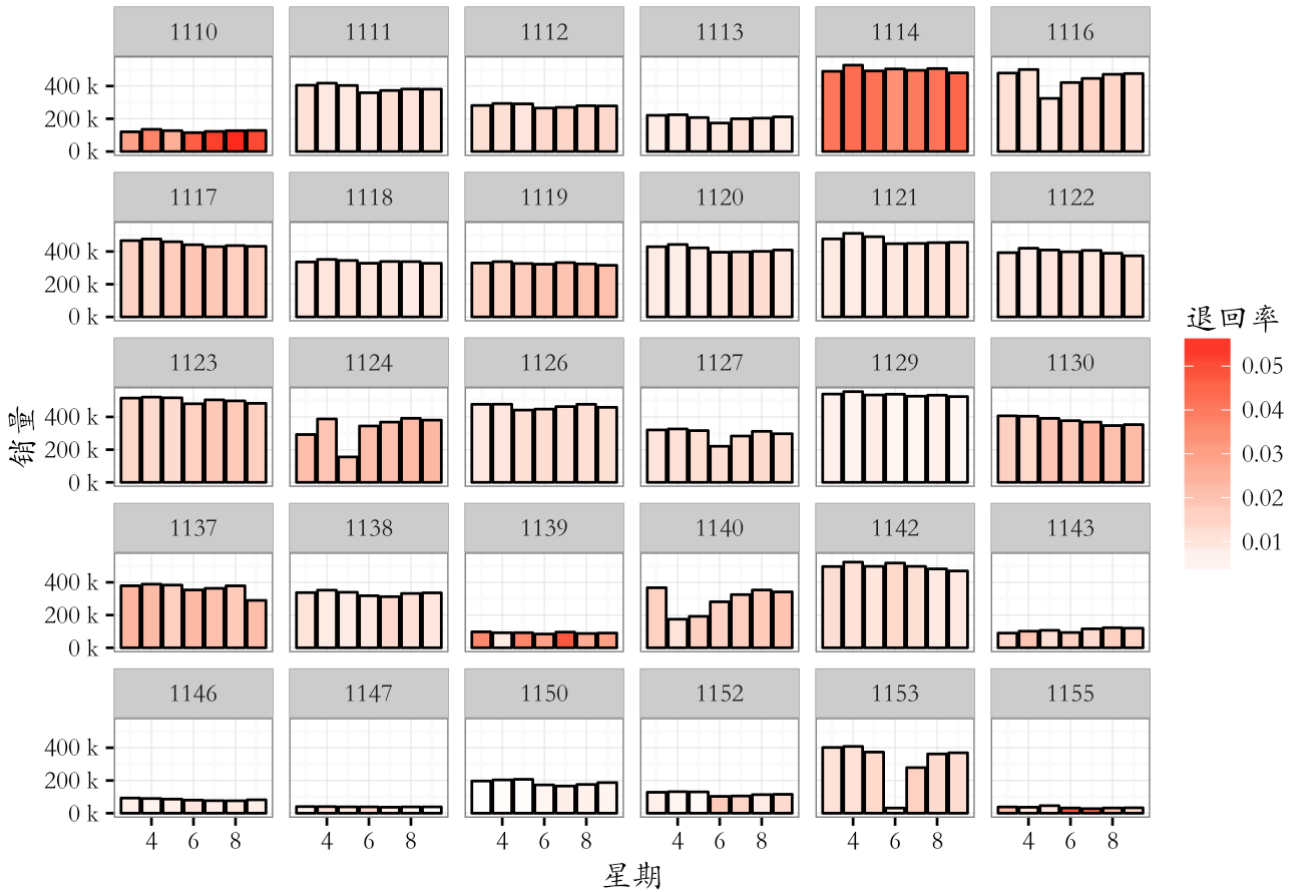
```
## 取出销售量前30的销售站ID
```

```
top30agencias <- agencias$Agencia_ID[1:30]
```

```
## 销量前30的销售站每周的销量和退货率图
```

```
ggplot(agencias_history %>% filter(Agencia_ID %in% top30agencias))+
  geom_bar(aes(x=Semana, y=Units, fill=Return_Rate), stat="identity", color="black")+
  theme_bw(base_family = "STKaiti") +
  facet_wrap(~Agencia_ID)+ # 按照销售站划分成子图
  scale_y_continuous(labels=function(x)paste(x/1000, "k"))+
  scale_fill_gradient(name="退货率", low="white", high="red")+
  ggtitle("销量前30的销售站") + ylab("销量") +xlab("星期")
```

## 销量前30的销售站



```
## 每个州的销售数据的分析
```

```
states <- agencias_history %>%
  group_by(State, Semana) %>%      #数据按照州和星期分组
  summarise(Units = sum(Units),
            Pesos = sum(Pesos),
            Return_Units = sum(Return_Units),
            Return_Pesos = sum(Return_Pesos),
            Net = sum(Net)) %>%
  mutate(Avg_Pesos = Pesos / Units,
         Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units))      # 数据按照销量排列
head(states)
```

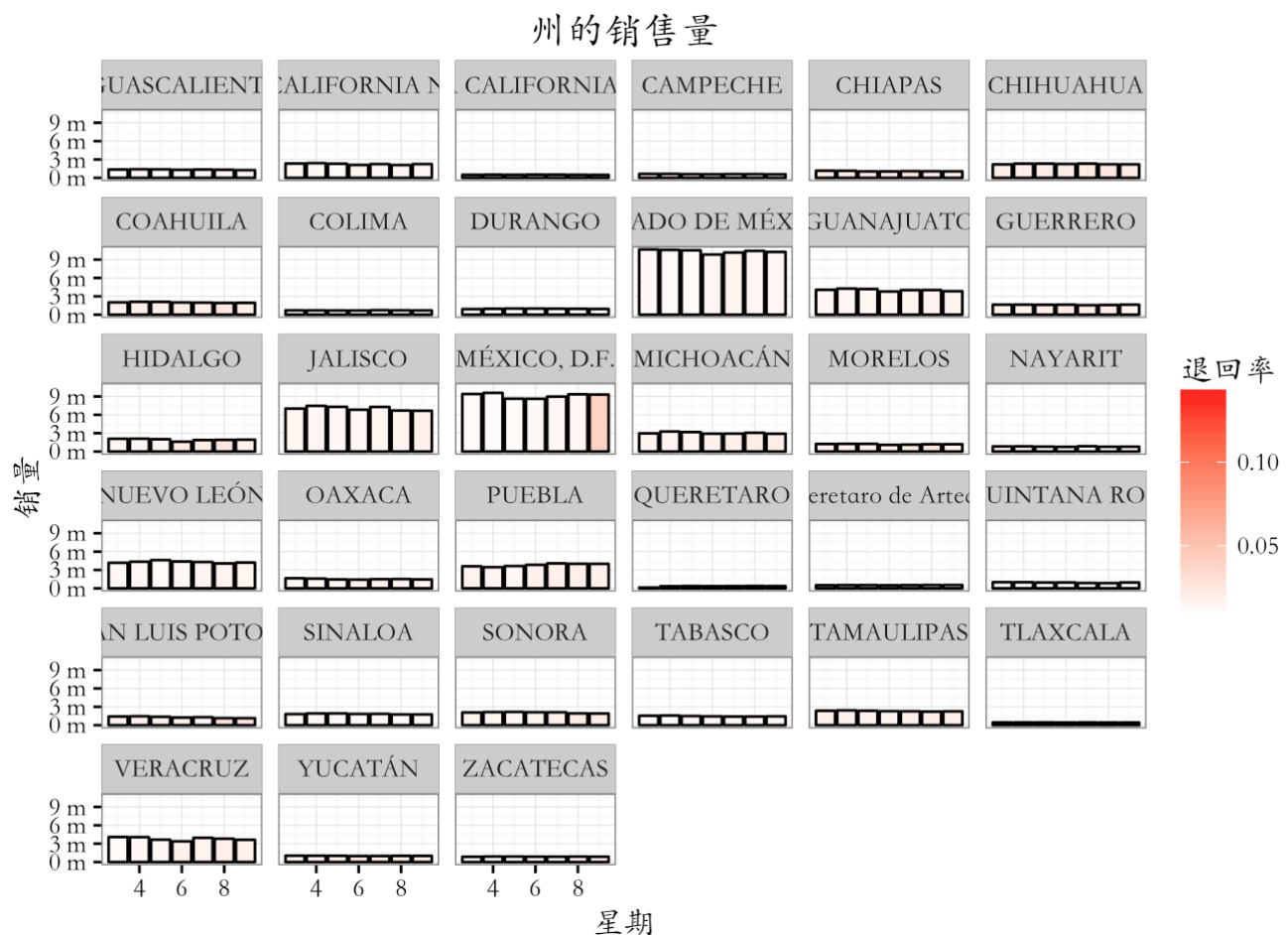
```
##           State Semana   Units   Pesos Return_Units Return_Pesos
## 1 ESTADO DE MÉXICO      3 10653345 103437148      154026      1510612
## 2 ESTADO DE MÉXICO      4 10584696 100758414      148409      1484139
## 3 ESTADO DE MÉXICO      5 10523973  99693511      144172      1449122
## 4 ESTADO DE MÉXICO      8 10422406  98640885      168741      1600855
## 5 ESTADO DE MÉXICO      9 10252045  98114659      160559      1523816
## 6 ESTADO DE MÉXICO      7 10156108  96457302      175260      1698870
##           Net Avg_Pesos Return_Rate
## 1 10552261  9.709359  0.01425194
## 2 10489671  9.519254  0.01382722
## 3 10424028  9.472992  0.01351425
## 4 10310433  9.464310  0.01593227
## 5 10144764  9.570252  0.01541968
## 6 10040746  9.497467  0.01696387
```

```
dim(states)[1] / 7 # 多少个州
```

```
## [1] 33
```

```
## 地点 -- 星期 -- 退回百分比 图像可视化
```

```
ggplot(states)+
  geom_bar(aes(x=Semana, y=Units, fill=Return_Rate), stat="identity", color="black")+
  theme_bw(base_family = "STKaiti") +
  facet_wrap(~State)+
  scale_y_continuous(labels=function(x)paste(x/1e6, "m"))+
  scale_fill_gradient(name="退回率", low="white", high="red")+
  ggtitle("州的销售量")+ ylab("销量") +xlab("星期")
```



```
## 州的销量树图
```

```
states2 <- states %>%
  group_by(State) %>%
  summarise(Units = sum(Units),
            Pesos = sum(Pesos),
            Return_Units = sum(Return_Units),
            Return_Pesos = sum(Return_Pesos),
            Net = sum(Net)) %>%
  mutate(Avg_Pesos = Pesos / Units,
         Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units)) # 数据按照销量排列

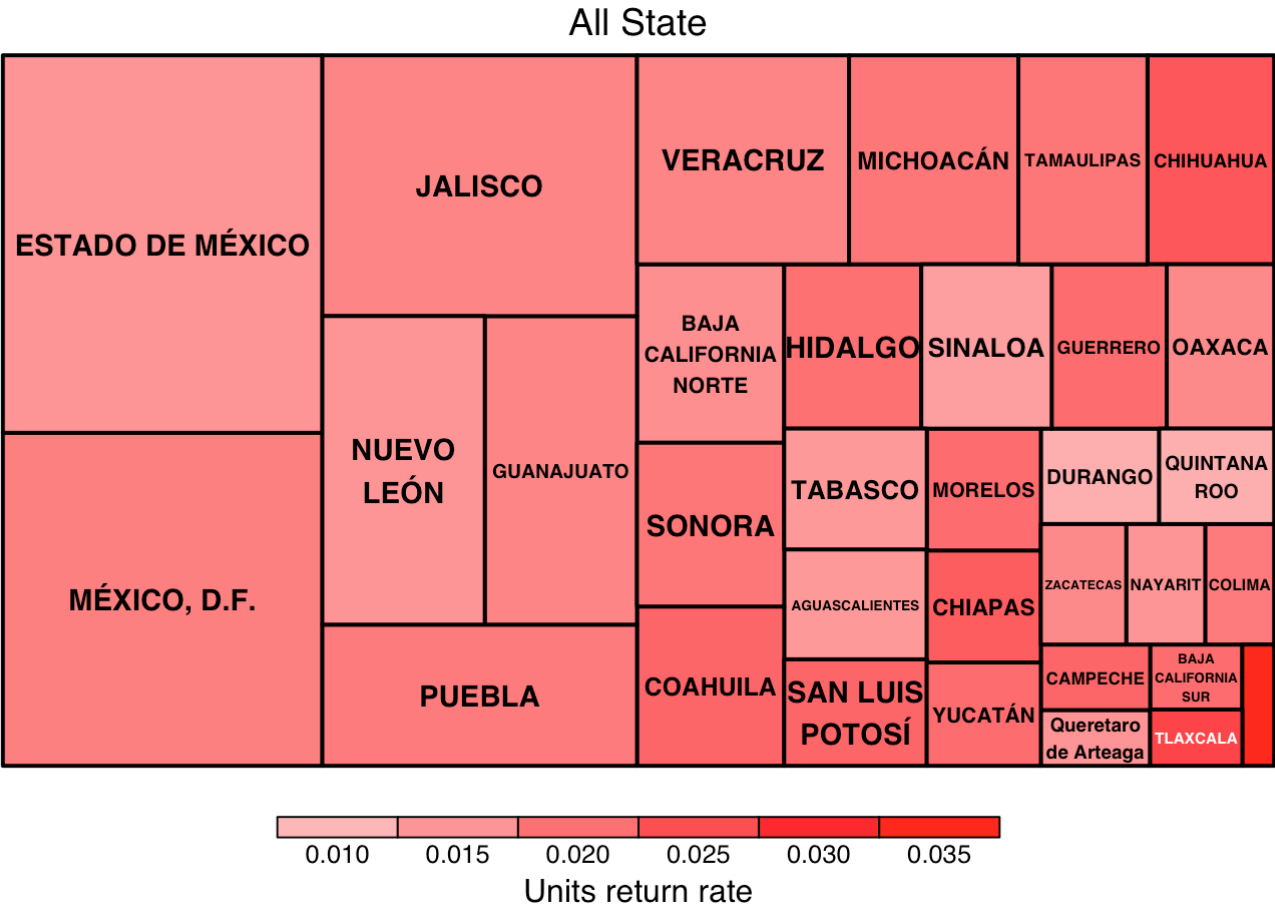
head(states2)
```

##	State	Units	Pesos	Return_Units	Return_Pesos	Net
## 1	ESTADO DE MÉXICO	72421491	692290237	1108790	10840260	71685681
## 2	MÉXICO, D.F.	63859463	641420811	1155641	9808764	63216897
## 3	JALISCO	49266415	429703973	837289	7840476	48701496
## 4	NUEVO LEÓN	30105293	286344224	466455	5252757	29756845
## 5	GUANAJUATO	28215533	241668605	481674	4447663	27884613
## 6	PUEBLA	26631917	226746531	502023	4189637	26293428
##	Avg_Pesos	Return_Rate				
## 1	9.559182	0.01507937				
## 2	10.044256	0.01777496				
## 3	8.722047	0.01671112				
## 4	9.511425	0.01525771				
## 5	8.565091	0.01678470				
## 6	8.514090	0.01850166				

```
dim(states2) # 多少个州
```

```
## [1] 33 8
```

```
treemap(states2,index = c("State"),vSize = "Units",vColor = "Return_Rate",
  type = "value",palette=c("#FFFFFF","#FFFFFF","#FF0000"),
  title.legend="Units return rate", title="All State")
```



## ## 销售渠道的分析####

```

canals2 <- traindata %>%
  group_by(Canal_ID) %>% #根据销售渠道和星期进行分组
  summarise(Units = sum(Venta_uni_hoy),
            Pesos = sum(Venta_hoy),
            Return_Units = sum(Dev_uni_proxima),
            Return_Pesos = sum(Dev_proxima),
            Net = sum(Demanda_uni_equil)) %>%
  mutate(Net_Pesos = Pesos - Return_Pesos,
         Avg_Pesos = Pesos / Units,
         Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units))
canals2

```

```

## Source: local data table [9 x 9]
##
##   Canal_ID      Units      Pesos Return_Units Return_Pesos      Net
##   (int)      (int)      (dbl)      (int)      (dbl)      (int)
## 1         1 390757290 3143719571      6177725   54588903.9 386646625
## 2         2  52402129  940084794      1226250   19762509.8 51384359
## 3         4  46663250  471003920       272376   2554061.4 46468113
## 4         5  20774423  172359549      1344109   9170591.8 19891458
## 5        11  14661539  165875147       252079   2379442.9 14589584
## 6         7  10282022  119925189       132817   1588764.4 10249097
## 7         6   3569312   28573289        54311    456051.2  3549277
## 8         8   3127856   41589248       202907   1724391.6  3109500
## 9         9    33497    1533784          0         0.0    33497
## Variables not shown: Net_Pesos (dbl), Avg_Pesos (dbl), Return_Rate (dbl)

```

```
dim(canals2)
```

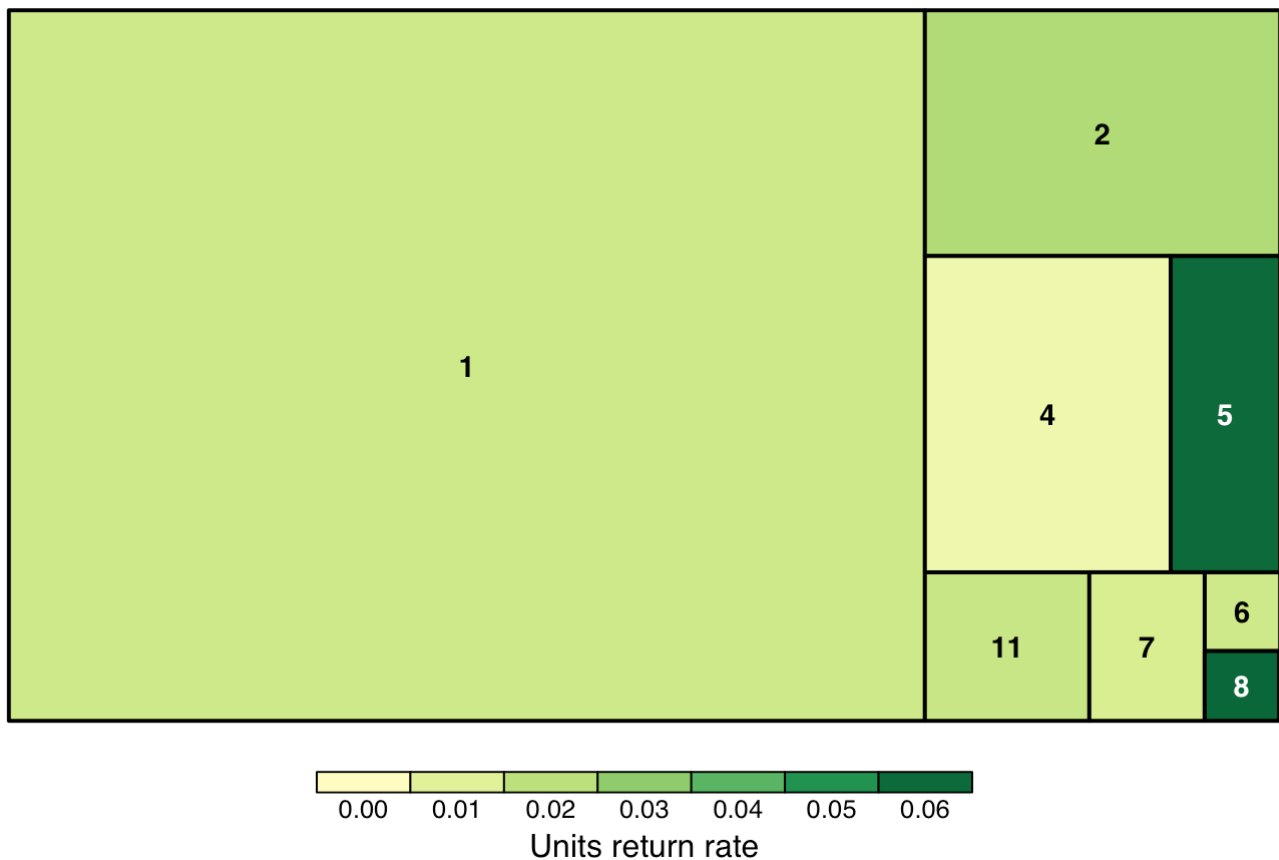
```
## [1] 9 9
```

```

# 销售渠道1占据主要的销量
treemap(canals2, index=c("Canal_ID"), vSize="Units", vColor = "Return_Rate", type="value",
        #palette=c("#FFFFFF", "#FFFFFF", "#FF0000"),
        title.legend="Units return rate",
        title="Canals repartition")

```

## Canals repartition

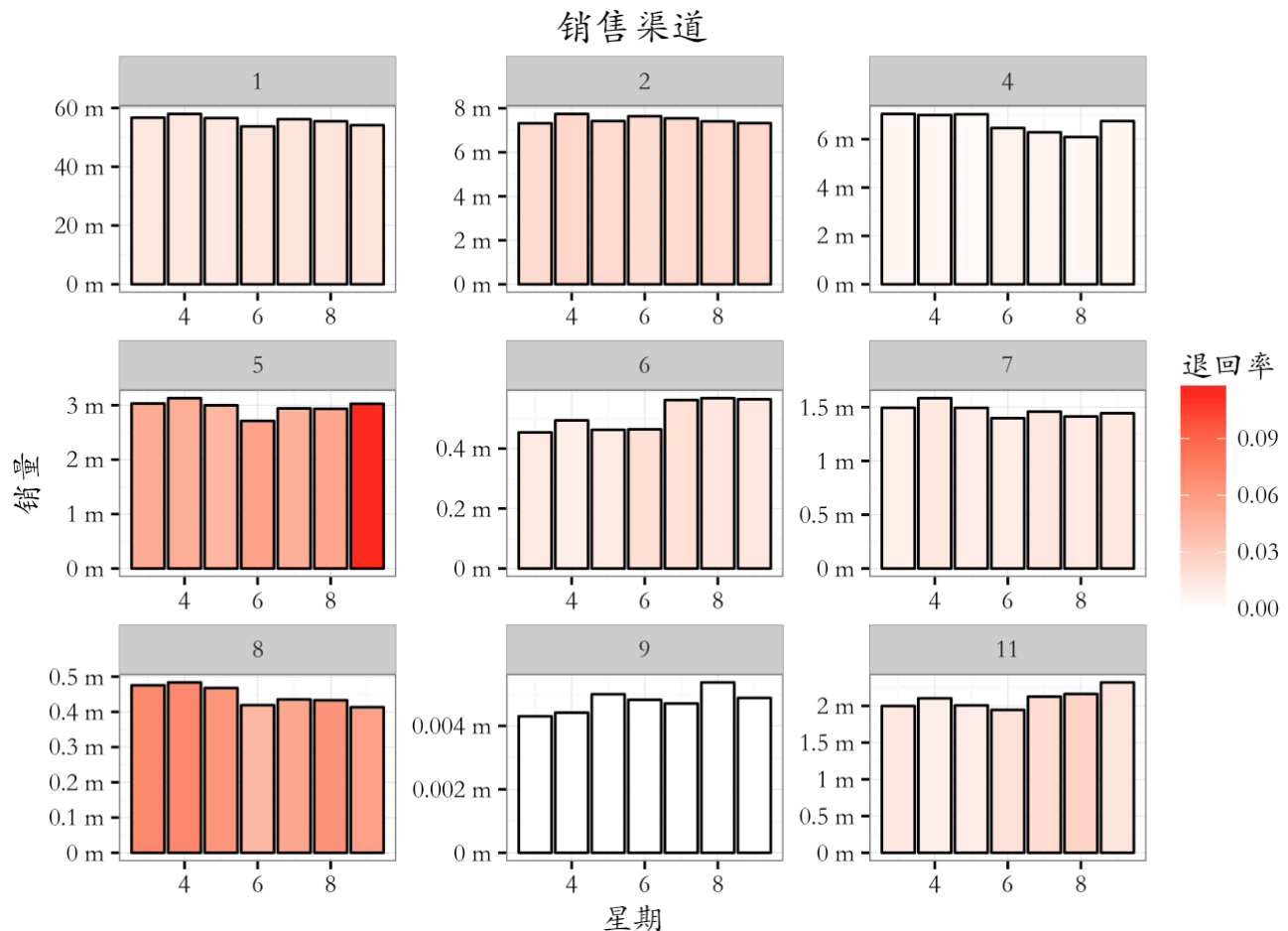


```
##
canals <- traindata %>%
  group_by(Canal_ID, Semana) %>% #根据销售渠道和星期进行分组
  summarise(Units = sum(Venta_uni_hoy),
            Pesos = sum(Venta_hoy),
            Return_Units = sum(Dev_uni_proxima),
            Return_Pesos = sum(Dev_proxima),
            Net = sum(Demanda_uni_equil)) %>%
  mutate(Net_Pesos = Pesos - Return_Pesos,
         Avg_Pesos = Pesos / Units,
         Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units))
head(canals)
```

##	Canal_ID	Semana	Units	Pesos	Return_Units	Return_Pesos	Net
## 1	1	4	57970962	469960421	839777	7536925	57415516
## 2	1	3	56697977	472277894	835051	7612406	56154862
## 3	1	5	56565253	450590664	798399	7198662	56026267
## 4	1	7	56188416	444780203	942701	8242810	55557236
## 5	1	8	55482753	441614833	957245	8294457	54845666
## 6	1	9	54142315	437679187	932279	7972306	53522411
##	Net_Pesos	Avg_Pesos	Return_Rate				
## 1	462423496	8.106825	0.01427931				
## 2	464665489	8.329713	0.01451429				
## 3	443392001	7.965856	0.01391820				
## 4	436537393	7.915870	0.01650066				
## 5	433320376	7.959497	0.01696040				
## 6	429706881	8.083865	0.01692757				

## 销售渠道的销量和星期和退货率

```
ggplot(canals)+
  geom_bar(aes(x=Semana, y=Units, fill=Return_Rate), stat="identity", color="black")+
  theme_bw(base_family = "STKaiti") +
  facet_wrap(~Canal_ID, scale="free")+
  scale_y_continuous(labels=function(x)paste(x/1e6, "m"))+
  scale_fill_gradient(name="退货率", low="white", high="red")+
  ggtitle("销售渠道")+ ylab("销量") +xlab("星期")
```



## 销售渠道和销售站分析####

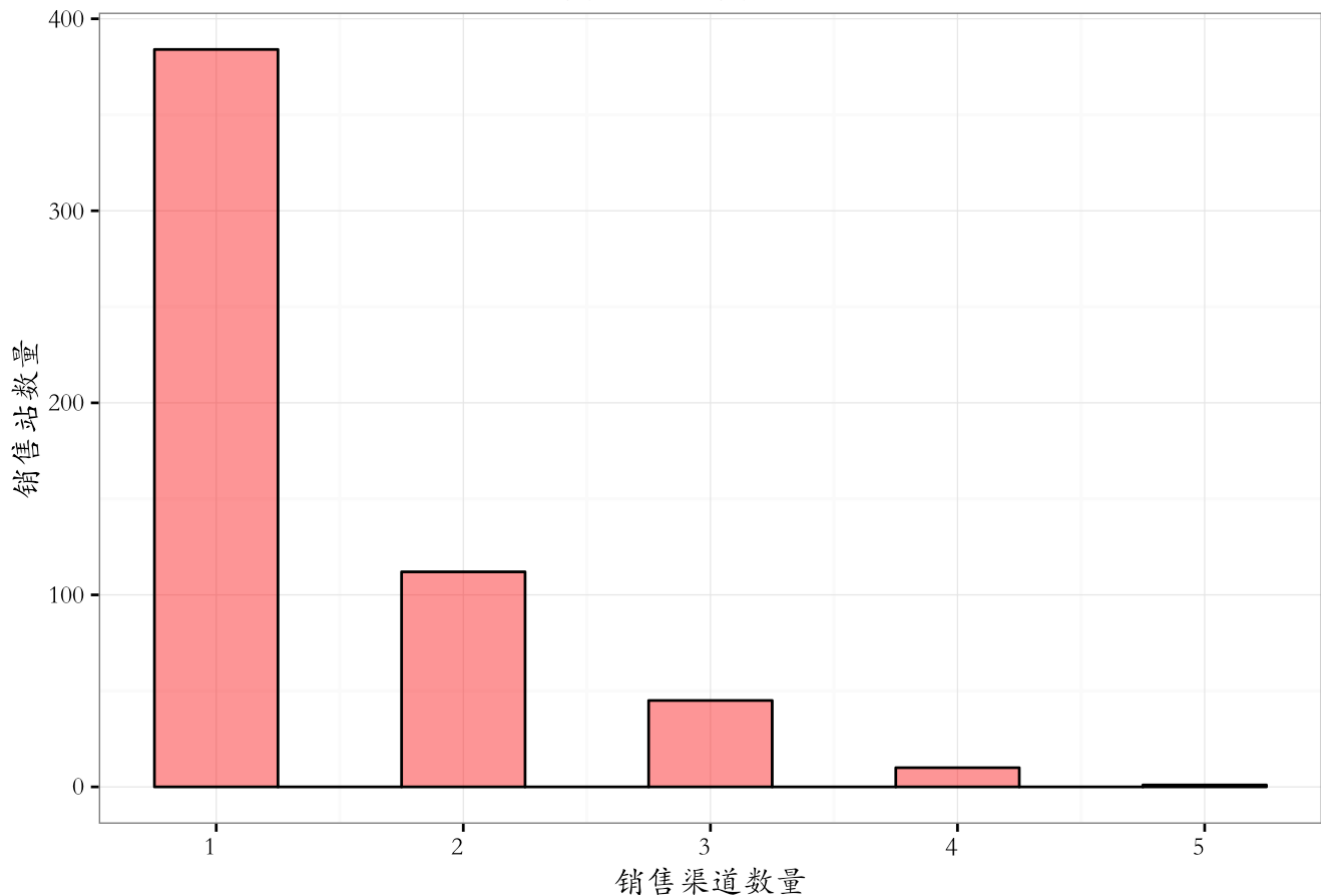
```
agencias_canals <- traindata %>%
  group_by(Agencia_ID) %>%
  summarise(n_canals = n_distinct(Canal_ID)) #添加该销售渠道有多少销售站
head(agencias)
```

```
##      Agencia_ID    Units      Pesos Return_Units Return_Pesos      Net Net_Pesos
## 1      1110  877675  9274674      39900      214072.8  874523  9060601
## 2      1111 2720400 24070592      25231      264672.4 2701427 23805919
## 3      1112 1959534 16591688      23924      231897.4 1942114 16359791
## 4      1113 1442999 12094484      11865      117754.4 1434414 11976730
## 5      1114 3498170 62420320     150779     2480404.7 3363796 59939915
## 6      1116 3120201 27454358      37022      377100.6 3093985 27077257
##      Return_Rate      Town      State
## 1 0.043484184      2008 AG. LAGO FILT      MÉXICO, D.F.
## 2 0.009189509 2002 AG. AZCAPOTZALCO      MÉXICO, D.F.
## 3 0.012061763      2004 AG. CUAUTITLAN ESTADO DE MÉXICO
## 4 0.008155401      2008 AG. LAGO FILT      MÉXICO, D.F.
## 5 0.041321213 2029 AG. IZTAPALAPA 2      MÉXICO, D.F.
## 6 0.011726128 2011 AG. SAN ANTONIO      MÉXICO, D.F.
```

## 销售渠道有多少销售站可视化

```
ggplot(agencias_canals)+
  geom_histogram(aes(x=n_canals), fill="red", color="black", alpha="0.5", binwidth=0.
5)+
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(breaks=1:5)+
  scale_y_continuous()+
  theme(axis.text.x=element_text(hjust=1)) +
  labs(x = "销售渠道数量", y = "销售站数量", title = "销售站的销售渠道量")
```

销售站的销售渠道量





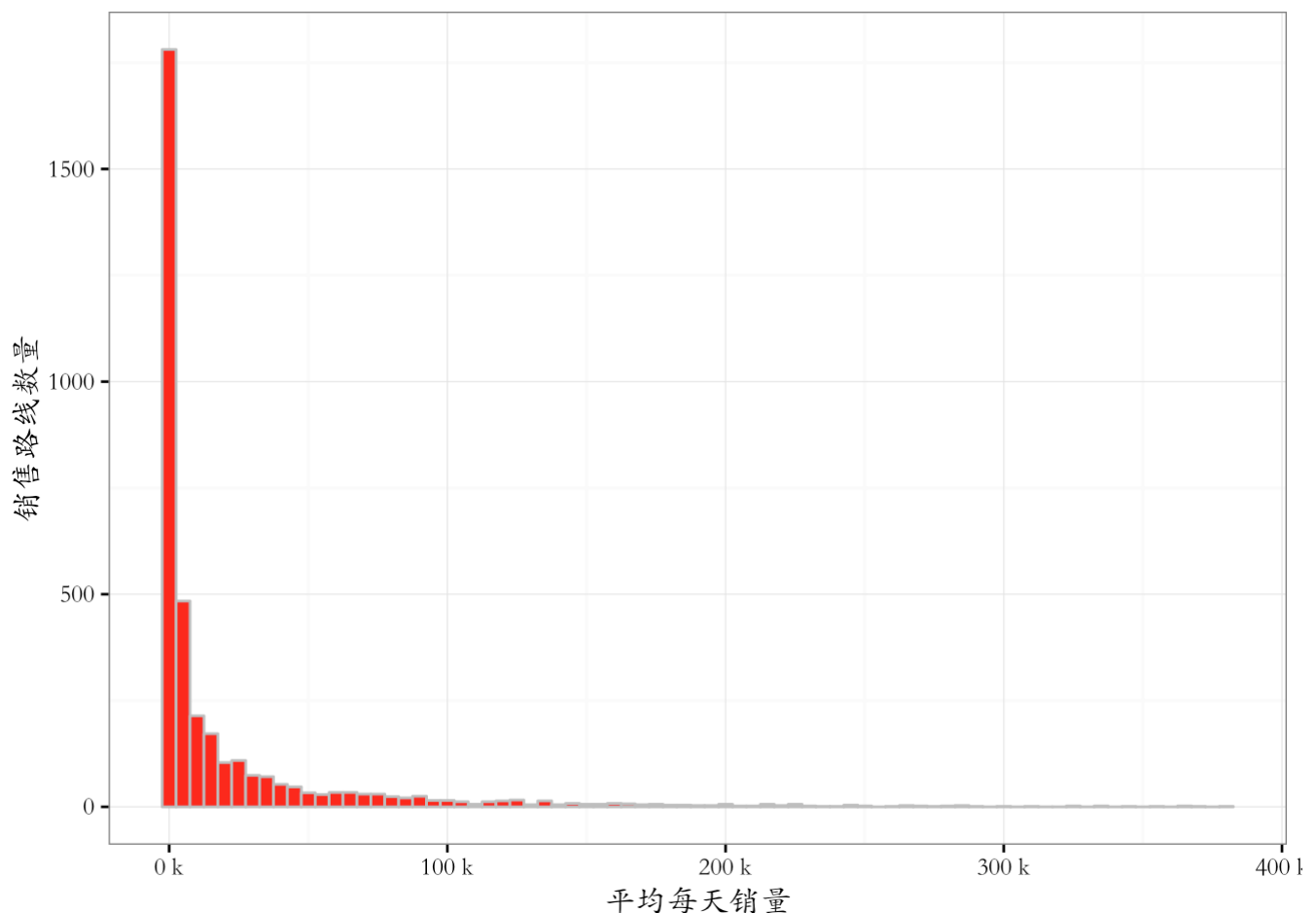
```
# 大部分销售站有1条销售渠道，只有很少的销售站有超过三条的销售渠道
```

```
## 销售路线的分析####
```

```
routes <- traindata %>% group_by(Ruta_SAK) %>%
  summarise(n_Agencias = n_distinct(Agencia_ID), #销售路线有多少销售站
            n_Clients = n_distinct(Cliente_ID), #销售路线有多少客户
            Units=sum(Venta_uni_hoy), #销售路线的销售量
            Return_Units = sum(Dev_uni_proxima)) %>% #销售路线的销售量退货量
  mutate(Return_Rate = Return_Units / (Units+Return_Units)) %>% # 添加退货率变量
  arrange(desc(Units)) # 按照销量排序
head(routes)
```

##	Ruta_SAK	n_Agencias	n_Clients	Units	Return_Units	Return_Rate
## 1	1101	82	5404	2645921	30981	0.011573453
## 2	6601	80	1104	2577239	10601	0.004096467
## 3	1102	82	5670	2571009	36771	0.014100499
## 4	1103	79	5434	2568287	37902	0.014543074
## 5	3001	49	20	2481974	402419	0.139516009
## 6	3002	49	25	2405887	125932	0.049739733

```
ggplot(routes, aes(x=Units/7))+
  geom_histogram(fill="red", color="gray", binwidth=5000)+
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(labels=function(x)paste(x/1000, "k"))+
  scale_y_continuous()+
  labs(x = "平均每天销量", y = "销售路线数量")
```



```
## 大部分的销售路线的销售量并不多, 超过2/3的销售路线每天的销售量不超过10千
```

```
## 销售路线和销售站####
```

```
routes_agencias <- traindata %>% group_by(Ruta_SAK, Agencia_ID) %>%
  summarise(count=n(),          #当前分组的观测数
            n_Clients = n_distinct(Cliente_ID), # 客户数量
            Units=sum(Venta_uni_hoy), #销售量求和
            Return_Units = sum(Dev_uni_proxima)) %>%
  mutate(Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units))
head(routes_agencias)
```

##	Ruta_SAK	Agencia_ID	count	n_Clients	Units	Return_Units	Return_Rate
## 1	900	22362	6953	9	201679	0	0.000000e+00
## 2	3	1142	773	1	163195	0	0.000000e+00
## 3	900	22560	2744	5	148973	0	0.000000e+00
## 4	1	1168	748	1	145380	8	5.502517e-05
## 5	8	1114	736	1	125192	174	1.387936e-03
## 6	900	22090	5043	6	111042	0	0.000000e+00

```
top100routes <- routes$Ruta_SAK[1:100] # 销量前100的路线
top100agencias <- agencias$Agencia_ID[1:100] # 销量前100的销售站
## 可视化
ggplot(routes_agencias %>%
  filter(Ruta_SAK %in% top100routes, Agencia_ID %in% top100agencias))+
  geom_point(aes(x=as.character(Ruta_SAK),
                y=as.character(Agencia_ID),
                size=Units, color=Return_Rate))+
  theme_bw(base_family = "STKaiti") +
  scale_color_gradient(name="退回率", low="blue", high="red")+
  scale_size_continuous(name = "销量", range = c(1,4)) +
  theme(axis.line=element_blank(),
        axis.text.x=element_blank(),
        axis.text.y=element_blank()) +
  labs(x = "销售路线", y = "销售站", title = "销量前100的销售站&销售路线")
```

## 销量前100的销售站&amp;销售路线

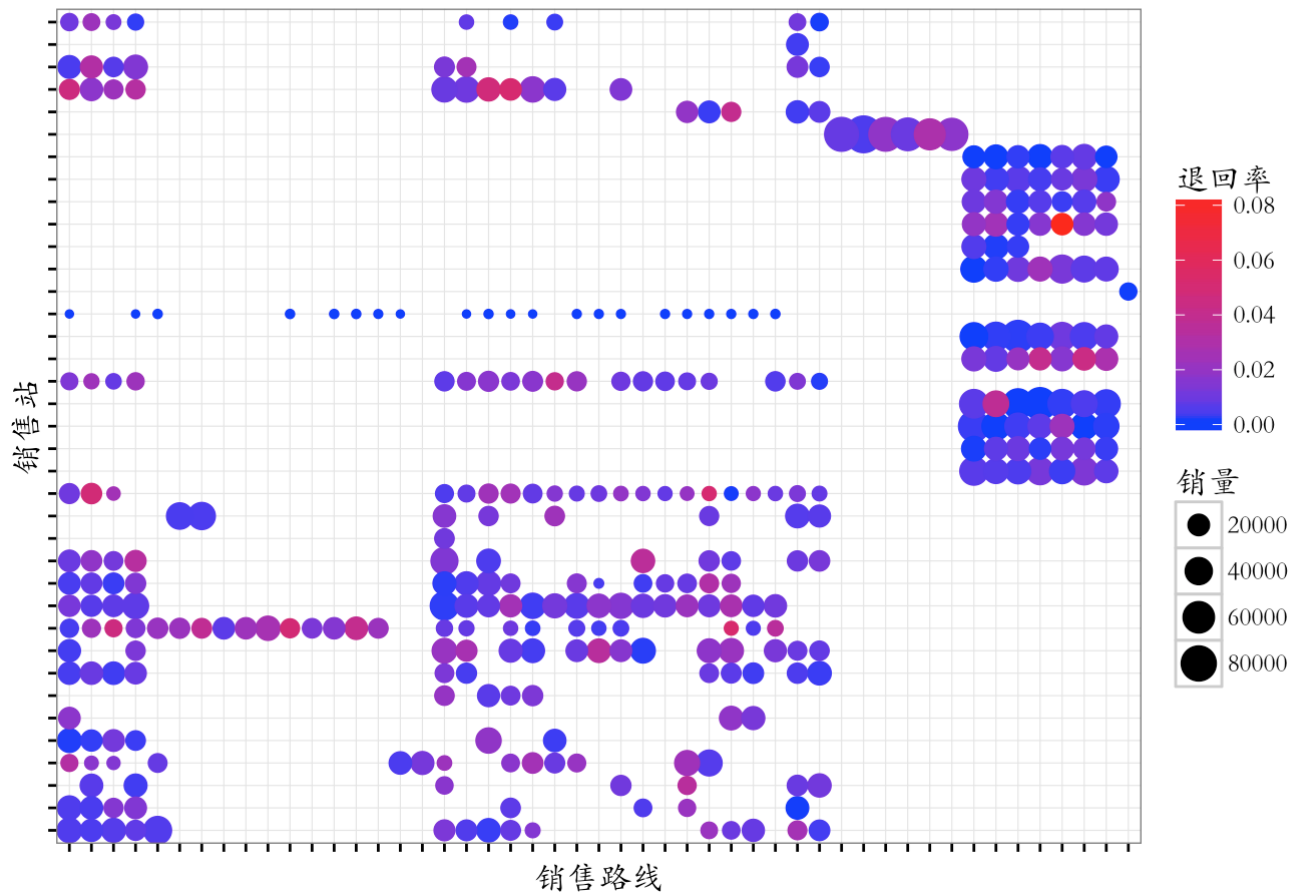


```

top50routes <- routes$Ruta_SAK[1:50] # 销量前50的路线
top50agencias <- agencias$Agencia_ID[1:50] # 销量前50的销售站
## 可视化
ggplot(routes_agencias %>%
  filter(Ruta_SAK %in% top50routes, Agencia_ID %in% top50agencias))+
  geom_point(aes(x=as.character(Ruta_SAK),
                 y=as.character(Agencia_ID),
                 size=Units, color=Return_Rate))+
  theme_bw(base_family = "STKaiti") +
  scale_color_gradient(name="退回率", low="blue", high="red")+
  scale_size_continuous(name = "销量", range = c(1,6)) +
  theme(axis.line=element_blank(),
        axis.text.x=element_blank(),
        axis.text.y=element_blank()) +
  labs(x = "销售路线", y = "销售站", title = "销量前50的销售站&销售路线")

```

## 销量前50的销售站&amp;销售路线



```
## 对客户数据进行分析####
```

```
sales <- traindata %>%      #客户数据
  group_by(Cliente_ID) %>%  # 按照客户id分组
  summarise(Units = sum(Venta_uni_hoy),
            Pesos = sum(Venta_hoy), # 本周销售金额
            Return_Units = sum(Dev_uni_proxima),
            Return_Pesos = sum(Dev_proxima), # 下星期的退回金额
            Net = sum(Demanda_uni_equil)) %>%
  mutate(Return_Rate = Return_Units / (Units+Return_Units),
         Avg_Pesos = Pesos / Units) %>% # 单价
  mutate(Net_Pesos = Pesos - Return_Pesos) %>% # 实际销售金额
  inner_join(cliente_tabla, by="Cliente_ID") %>%
  arrange(desc(Pesos)) # 本周销售金额排序
head(sales)
```

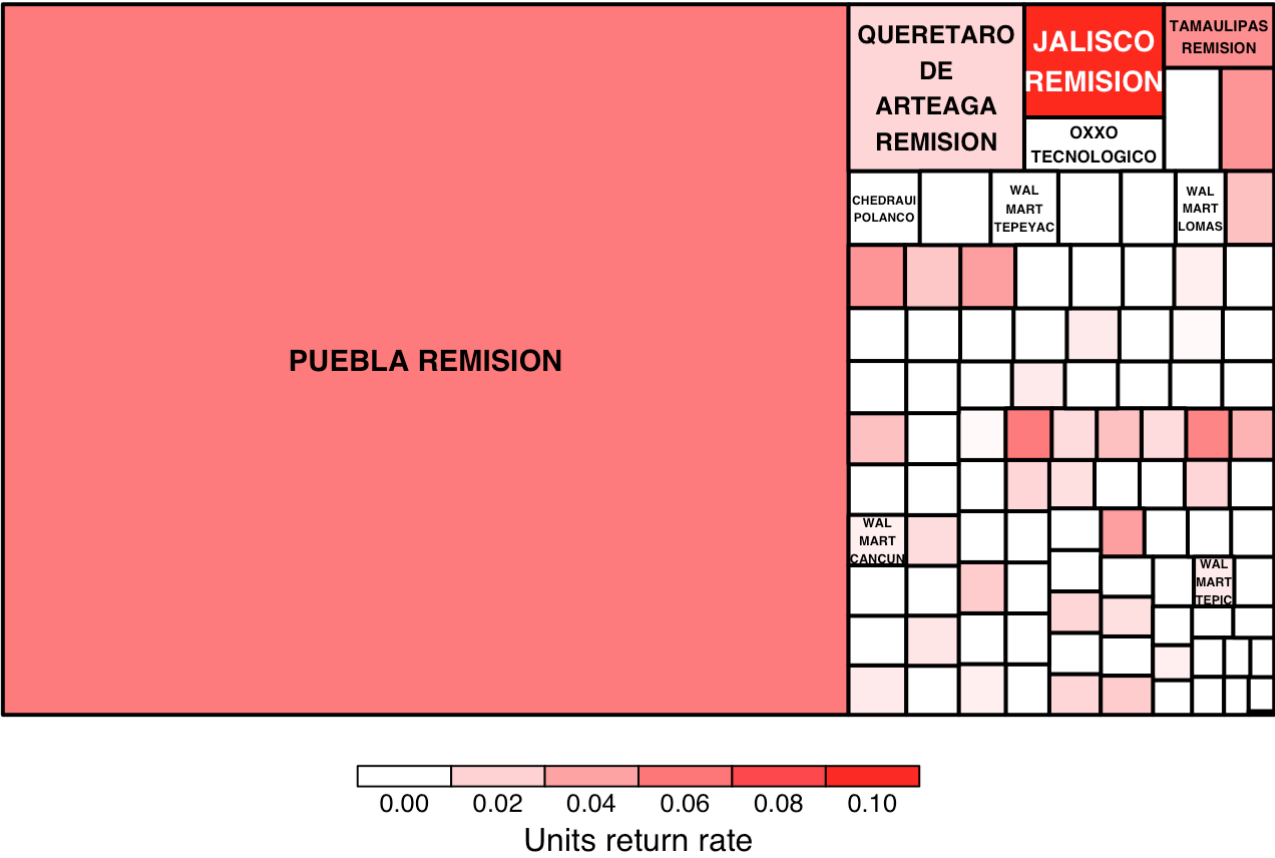
```
##      Cliente_ID      Units      Pesos Return_Units Return_Pesos      Net
## 1      653378 18650001 154662268      1131794   7367474.15 17866224
## 2      653039   909671   7697623        16642   143066.78  893756
## 3      827594    69264   4814696           0         0.00   69264
## 4      652850   490617   4018867        59664   495570.18  440039
## 5     1216931   232517   3325557          49       395.64  232472
## 6     5903732  162633   2931618           0         0.00  162633
##      Return_Rate Avg_Pesos Net_Pesos
## 1 0.0572139182   8.292883 147294794
## 2 0.0179658496   8.461986  7554557
## 3 0.0000000000  69.512248  4814696
## 4 0.1084246049   8.191454  3523296
## 5 0.0002106929  14.302425  3325161
## 6 0.0000000000  18.025974  2931618
##
##                               NombreCliente
## 1                               PUEBLA REMISION
## 2                QUERETARO DE ARTEAGA REMISION
## 3                MC DONALDS ANTONIO CUAUTITLAN
## 4                               JALISCO REMISION
## 5                               OXXO TECNOLOGICO
## 6 WAL MART SUPER CENTER DOMINGO DIEZ CUERNAVACA
```

```
dim(sales)
```

```
## [1] 885416      10
```

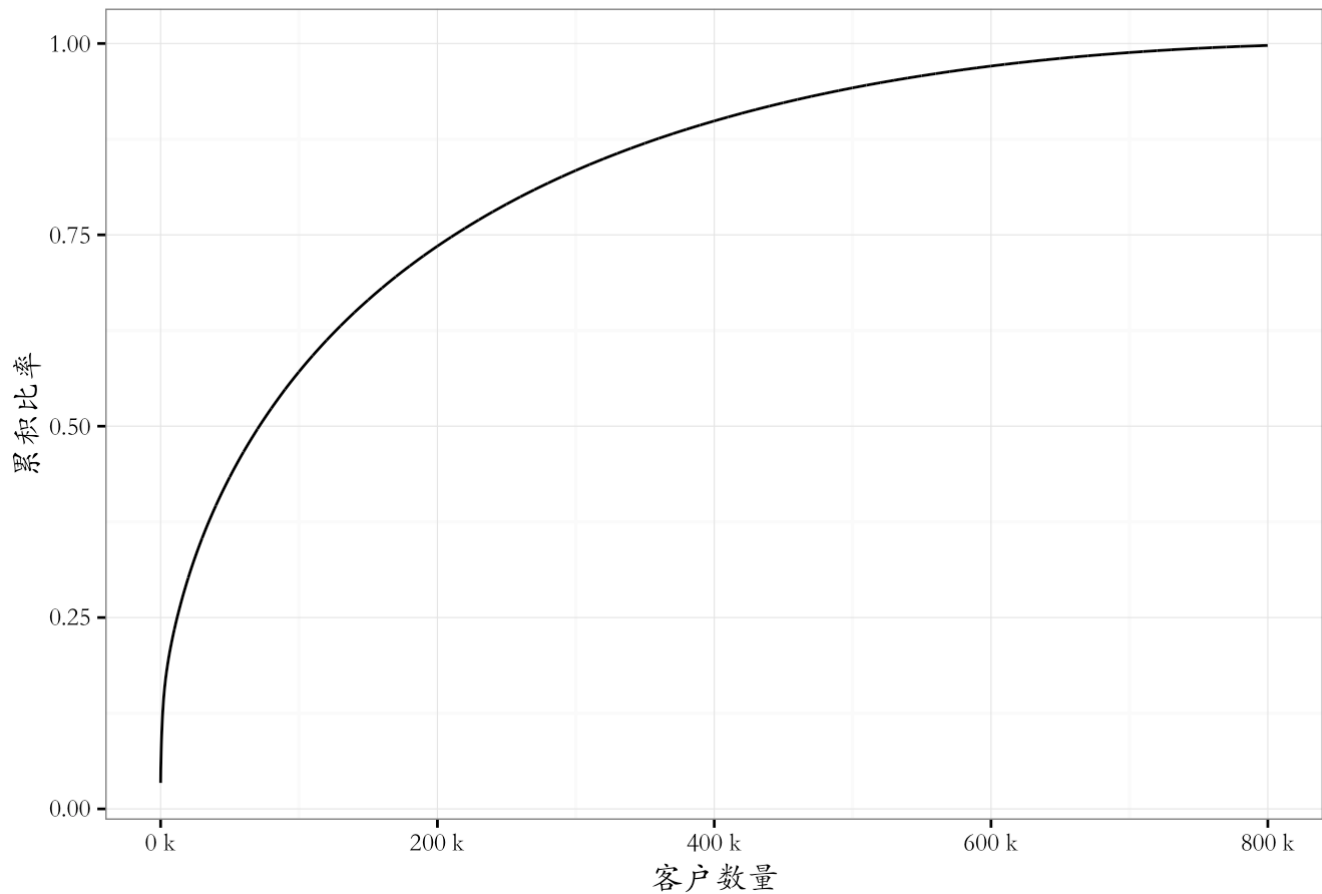
```
# 花费量前100个客户的树形图
# 可见有一个大客户: Puebla Remision
treemap(sales[1:100, ],
        index=c("NombreCliente"), vSize="Units", vColor="Return_Rate",
        palette=c("#FFFFFF", "#FFFFFF", "#FF0000"),
        type="value", title.legend="Units return rate", title="Top 100 clients")
```

Top 100 clients



```
## 客户的累积消耗量
sales$Cum_Units <- cumsum(sales$Units) / sum(sales$Units) # 累积百分比
s <- seq(1, 800000, 100) # 约有80万个客户
ggplot()+geom_line(aes(x=s, y=sales$Cum_Units[s]))+
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(labels=function(x) paste(x/1000, "k"))+
  ggtitle("客户分配")+ xlab("客户数量")+ylab("累积比率")
```

## 客户分配



```
## 前20万客户约贡献了75%的销售量
```

```
## 客户和销售站分析####
```

```
agencias_by_client <- traindata %>%
  group_by(Cliente_ID) %>% #按照客户id分组
  summarise(n_agencias = n_distinct(Agencia_ID)) %>% #多少个销售站
  inner_join(cliente_tabla, by="Cliente_ID")
head(agencias_by_client)
```

```
##   Cliente_ID n_agencias      NombreCliente
## 1         26          2 BODEGA COMERCIAL MEXICANA TOLUCA
## 2         60          2          SAMS CLUB TOLUCA
## 3         65          2        WAL MART METEPEC
## 4        101          1        WAL MART TOLUCA
## 5        105          1    SUPER KOMPRAS SAN BUENAVENTURA
## 6        106          1              ISSSTE 21
```

```
dim(agencias_by_client)
```

```
## [1] 885416      3
```

```
# 单个客户使用销售站的数量
# 大部分的客户只从一个销售站购买，只有几个客户购买狗的销售站 >= 5
table(agencias_by_client$n_agencias)
```

```
##
##      1      2      3      4      5      9      62
## 844113 37510 3771 19      1      1      1
```

## 使用销售站多的客户信息

```
agencias_by_client %>% filter(n_agencias %in% c(5, 9, 62)) #返回符合条件的行
```

```
## Source: local data table [3 x 3]
```

```
##
##   Cliente_ID n_agencias NombreCliente
##   (int)      (int)      (chr)
## 1    188391      9      DESAYUNOS ESCOLARES
## 2    653378     62      PUEBLA REMISION
## 3   1274327      5  COMERCIALIZADORA LA PUERTA DEL SOL
```

```
# Cliente_ID n_agencias NombreCliente
# (int)      (int)      (chr)
# 1    188391      9      DESAYUNOS ESCOLARES
# 2    653378     62      PUEBLA REMISION
# 3   1274327      5  COMERCIALIZADORA LA PUERTA DEL SOL
```

## 客户和购买渠道分析####

```
clients_canals <- traindata %>%
group_by(Cliente_ID) %>%
  summarise(n_canals = n_distinct(Canal_ID))
```

## 大多数客户只有一个购买渠道。不同的销售渠道可以为一个客户提供服务。

```
table(clients_canals$n_canals)
```

```
##
##      1      2      3      4
## 874022 6516 65      1
```

```
# 1      2      3      4
# 874022 6516 65      1
```

# 很少有销售站有同一个客户通过多个渠道。

```
clients_agencies_canals <- traindata %>%
  group_by(Cliente_ID, Agencia_ID) %>%
  summarise(n_canals = n_distinct(Canal_ID))
```

```
table(clients_agencies_canals$n_canals)
```

```
##
##      1      2      3
## 922108 3271 3
```



```
## 客户和路线分析####
clients_routes <- traindata %>%
  group_by(Cliente_ID) %>%
  summarise(n_routes = n_distinct(Ruta_SAK))
head(clients_routes)
```

```
##   Cliente_ID n_routes
## 1      15766         1
## 2      22926         2
## 3      24080         1
## 4      24695         1
## 5      50379         1
## 6      50395         1
```

```
dim(clients_routes)
```

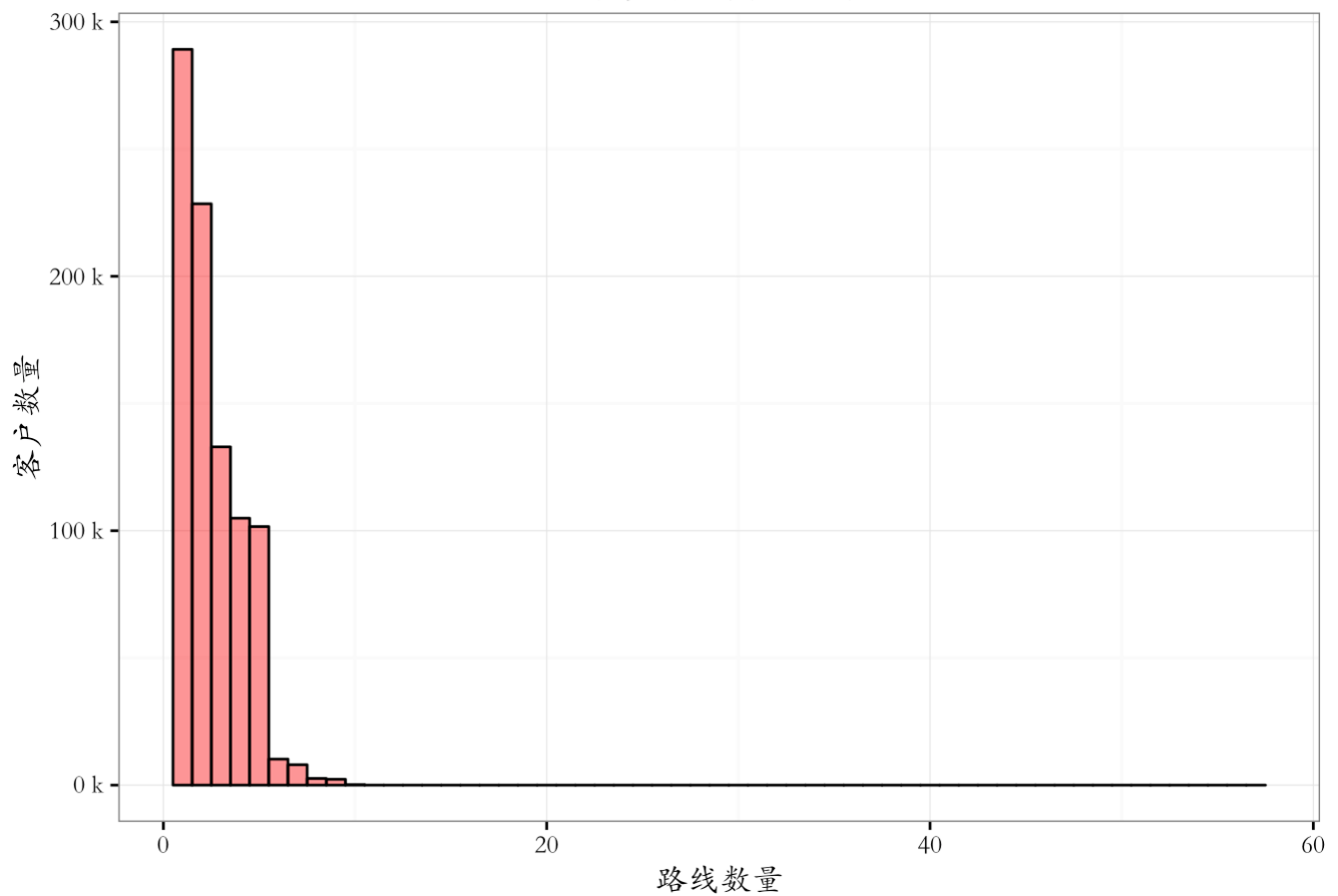
```
## [1] 880604      2
```

```
## 大多数客户只有不到5个仓库的交货，但超过240个客户的工作与10个仓库或更多。
sum(clients_routes$n_routes >= 10)
```

```
## [1] 242
```

```
ggplot(clients_routes)+
  geom_histogram(aes(x=n_routes), fill="red", color="black", alpha="0.5",
  binwidth=1)+
  theme_bw(base_family = "STKaiti") +
  scale_y_continuous(labels=function(x) paste(x/1000, "k"))+
  ggtitle("客户和销售路线")+ xlab("路线数量")+ylab("客户数量")
```

## 客户和销售路线



## 对集团销售的产品进行分析####

```
products <- traindata %>% group_by(Producto_ID) %>% #根据生产的产品进行分组
  summarise(Units = sum(Venta_uni_hoy), # 销量
            Pesos = sum(Venta_hoy),    # 卖出的总钱数
            Return_Units = sum(Dev_uni_proxima), # 被退回的总量
            Return_Pesos = sum(Dev_proxima), # 备退回的总钱数
            Net = sum(Demanda_uni_equil)) %>% #调整后的我总需求
  mutate(Avg_Pesos = Pesos / Units, # 每种产品的单价
         Return_Rate = Return_Units / (Units+Return_Units)) %>% # 退货率
  filter(!is.nan(Avg_Pesos)) %>% #剔除没有单价的商品
  inner_join(producto_tabla, by="Producto_ID") %>%
  arrange(desc(Units))
head(products)
```

##	Producto_ID	Units	Pesos	Return_Units	Return_Pesos	Net
## 1	2425	23860309	107365673	170005	765022.5	23728674
## 2	1278	19660667	88448180	178937	805123.9	19536596
## 3	1284	19625219	59297775	122273	387284.7	19539579
## 4	43285	15259454	80239869	158415	834036.1	15157951
## 5	36610	12836368	9884190	297745	229263.6	12680243
## 6	1240	12313013	104695281	195520	1800498.9	12167245
##	Avg_Pesos	Return_Rate	NombreProducto			
## 1	4.4997604	0.007074606	Nito 1p 62g Central BIM 2425			
## 2	4.4987375	0.009019182	Nito 1p 62g BIM 1278			
## 3	3.0215089	0.006191824	Rebanada 2p 55g BIM 1284			
## 4	5.2583709	0.010274766	Gansito 1p 50g MTB MLA 43285			
## 5	0.7700145	0.022669593	Bolsa Mini Rocko 40p 13g CU MLA 36610			
## 6	8.5028157	0.015630930	Mantecadas Vainilla 4p 125g BIM 1240			

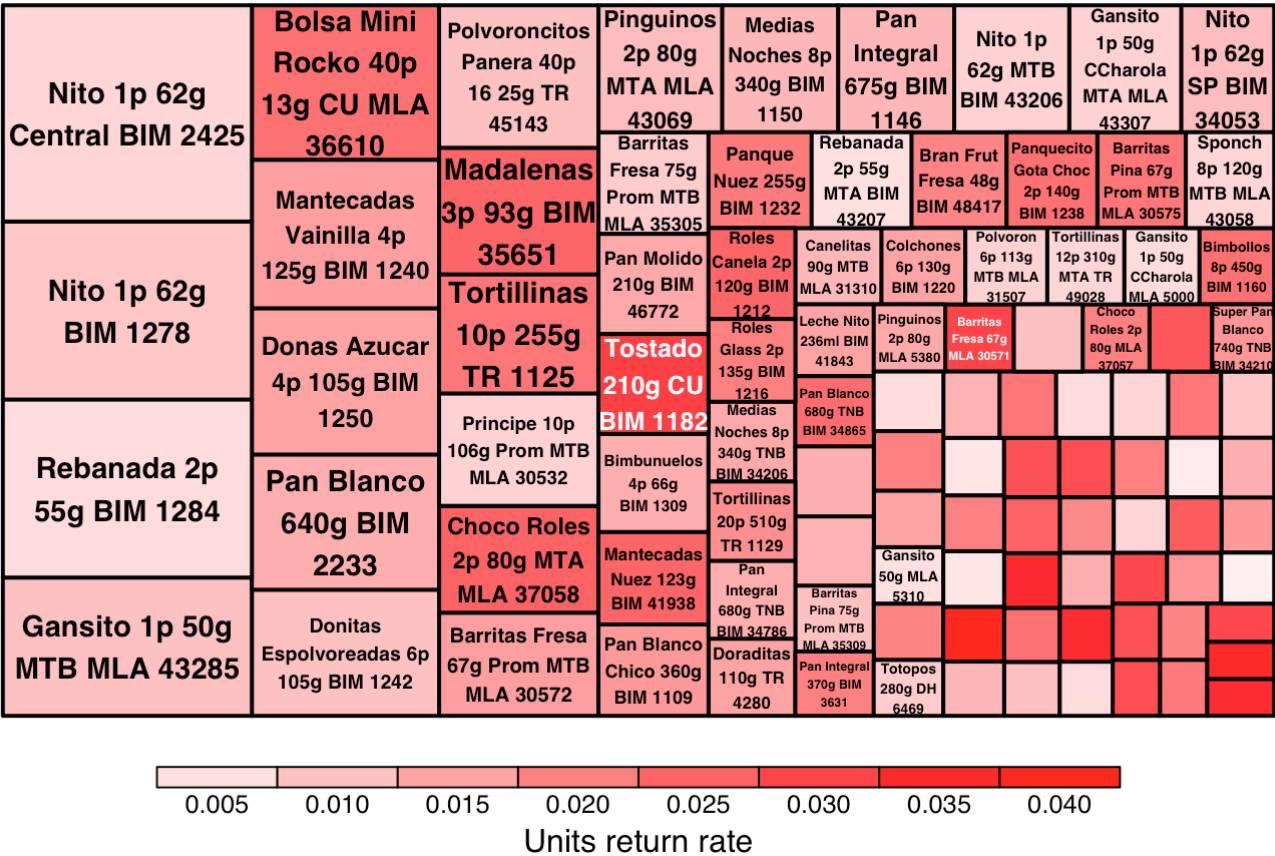
```
dim(products) # 产品数量
```

```
## [1] 1719      9
```

```
products$NombreProducto <- factor(as.character(products$NombreProducto), levels=products$NombreProducto)

# 销量前100的产品树图
treemap(products[1:100, ],
         index=c("NombreProducto"), vSize="Units", vColor="Return_Rate",
         palette=c("#FFFFFF", "#FFFFFF", "#FF0000"),
         type="value", title.legend="Units return rate", title="Top 100 products")
```

Top 100 products



```
## 产品的家的密度分布
```

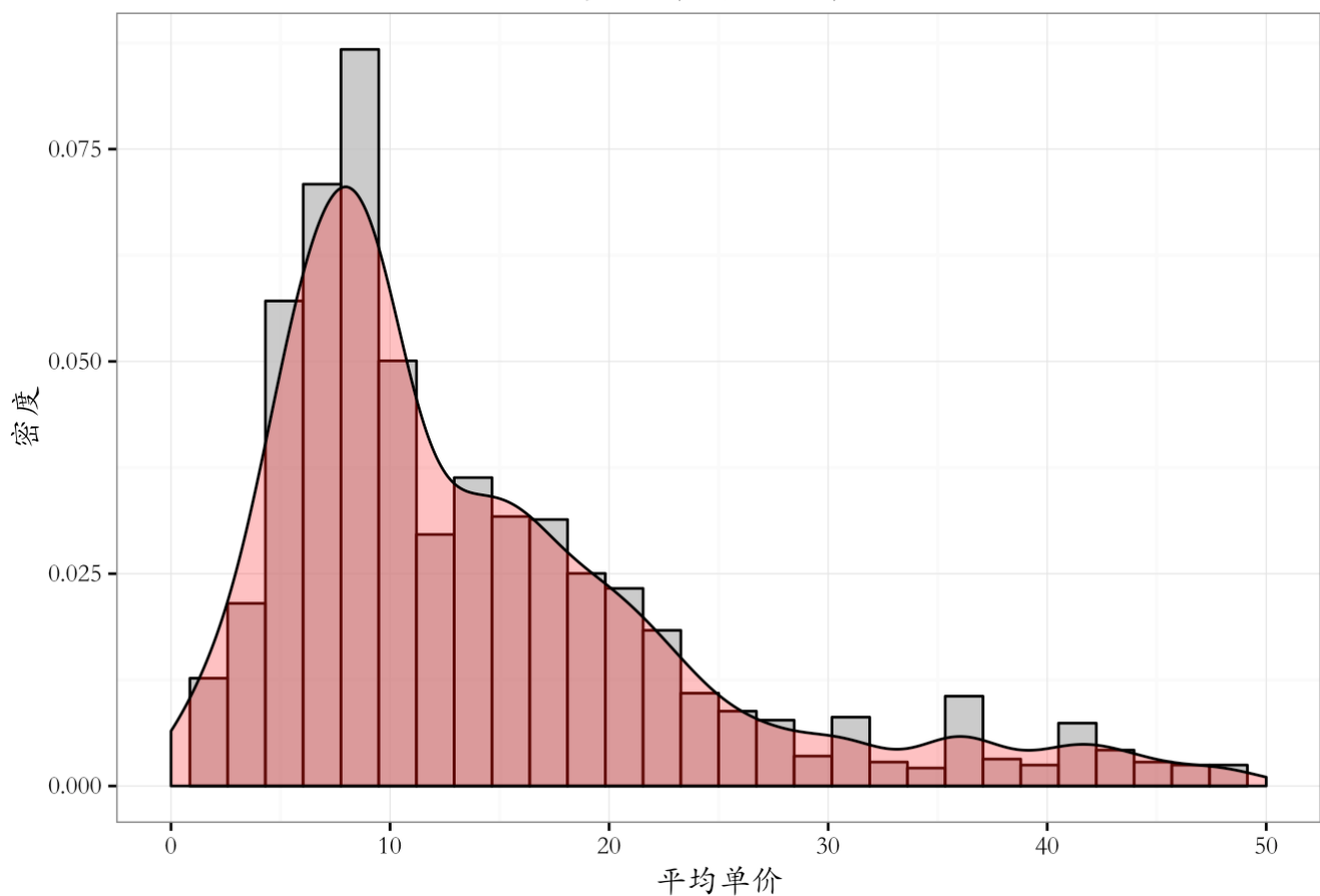
```
ggplot(products, aes(x=Avg_Pesos))+
  geom_histogram(aes(y=..density..), fill="gray", color="black", alpha="0.8")+
  geom_density(fill="red", alpha="0.3")+
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(lim=c(0, 50))+
  ggtitle("产品单价的分布")+ xlab("平均单价")+ylab("密度")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
## Warning: Removed 74 rows containing non-finite values (stat_bin).
```

```
## Warning: Removed 74 rows containing non-finite values (stat_density).
```

产品单价的分布



```
## 产品和销售站
```

```
products_agencies <- traindata %>% group_by(Agencia_ID) %>%
  summarise(n_products = n_distinct(Producto_ID))
head(products_agencies)
```

```
##   Agencia_ID n_products
## 1       1110         214
## 2       1111         208
## 3       1112         199
## 4       1113         202
## 5       1114         222
## 6       1116         200
```

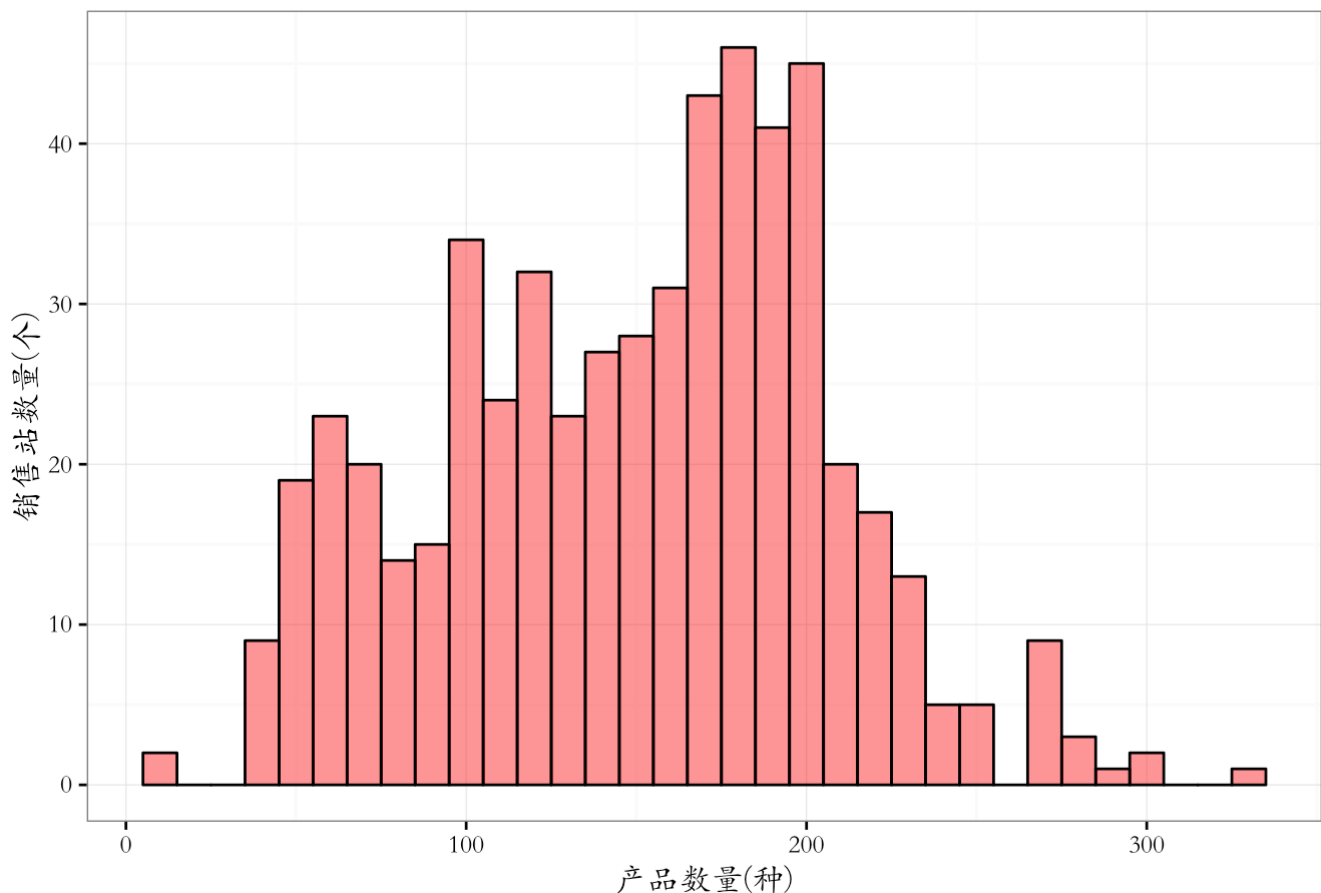
```
dim(products_agencies)
```

```
## [1] 552    2
```

```
## 大多数销售站会卖100~200种产品
```

```
ggplot(products_agencies)+  
  geom_histogram(aes(x = n_products), fill="red", color="black", alpha="0.5", binwidth=10)+  
  theme_bw(base_family = "STKaiti") +  
  ggtitle("销售站出售的产品数量")+ xlab("产品数量(种)") + ylab("销售站数量(个)")
```

销售站出售的产品数量



```
## 产品和销售路线
```

```
routes_products <- traindata %>% group_by(Producto_ID) %>%  
  summarise(n_routes = n_distinct(Ruta_SAK))  
table(routes_products$n_routes)
```

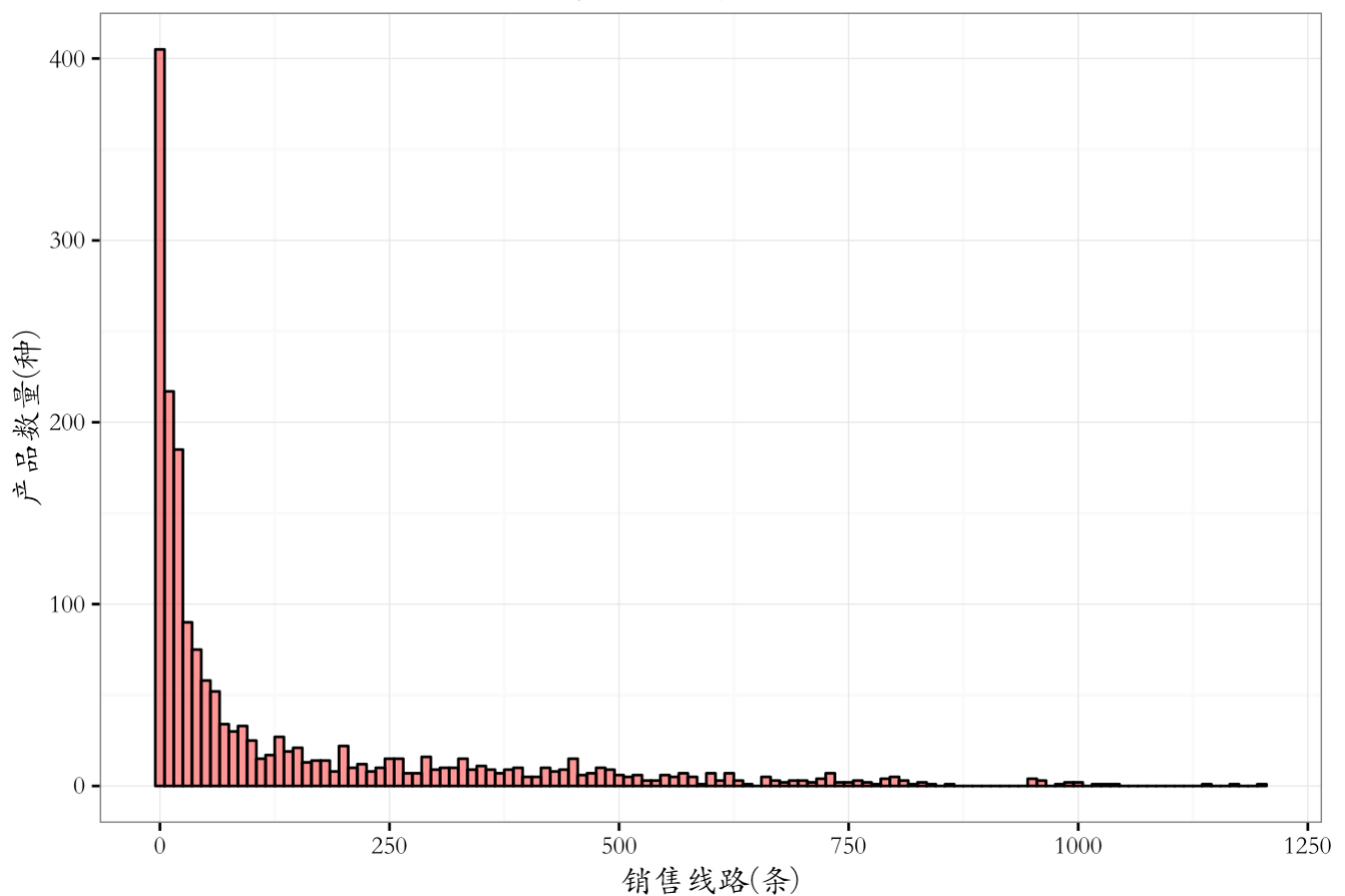
```
##
##      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15
## 155    87    67    40    56    36    26    19    16    15    26    14    16    34    15
## 16     17    18    19    20    21    22    23    24    25    26    27    28    29    30
## 18     21    46    14    23    13    13    12    14    11     6    12     5    13     8
## 31     32    33    34    35    36    37    38    39    40    41    42    43    44    45
## 10      9      6      9     12      4      8      5      7      5      7     11      7     11     10
## 46     47     48     49     50     51     52     53     54     55     56     57     58     59     60
##  9     14      4      5      3      5      5      4      6      3      6      8      7      3      5
## 61     63     64     65     66     67     68     69     70     71     72     73     74     75     76
##  7      7      4      5      1      5      2      2      8      4      3      2      4      3      4
## 77     78     79     80     82     83     84     85     86     87     88     89     90     91     92
##  1      2      3      2      2      4      5      7      2      5      2      3      4      5      5
## 93     94     95     96     97     98     99    100    101    102    103    104    105    106    108
##  1      2      4      5      3      2      2      2      1      2      2      3      3      2      1
## 109    111    113    114    115    116    117    118    119    122    123    124    125    126    127
##  2      3      2      3      2      4      1      2      2      1      1      1      5      3      2
## 128    129    130    131    132    133    134    135    136    137    138    139    140    141    142
##  2      1      4      3      3      5      1      3      1      2      1      2      1      1      4
## 143    144    145    146    147    148    149    150    151    152    153    154    155    156    157
##  1      5      1      2      2      2      1      2      3      1      3      2      3      2      1
## 158    160    161    162    164    167    168    169    170    171    173    174    177    178    179
##  4      2      1      1      2      3      1      2      3      2      2      1      1      2      4
## 181    182    183    184    185    186    190    192    193    194    195    196    197    198    200
##  2      2      1      1      1      1      1      2      2      1      1      2      4      4      1
## 201    202    203    204    205    206    208    209    213    214    215    216    217    218    219
##  2      2      1      4      2      1      2      3      2      1      1      1      2      1      3
## 221    223    225    226    227    230    231    235    237    238    239    240    243    245    247
##  1      2      2      1      3      1      1      2      2      3      1      1      2      1      1
## 248    249    251    252    253    254    255    256    257    258    260    261    262    263    264
##  1      1      2      1      2      3      4      1      2      2      2      1      4      2      1
## 266    269    270    273    274    275    276    277    279    283    284    286    287    288    289
##  1      2      1      1      1      1      1      3      1      1      1      1      1      4      2
## 291    292    293    294    297    298    299    300    303    305    306    307    308    312    313
##  2      2      1      3      2      2      1      1      2      1      1      1      1      1      5
## 315    318    320    321    322    325    326    327    328    330    331    332    334    336    337
##  1      2      3      1      1      3      2      3      4      1      1      3      1      1      2
## 340    341    343    345    348    349    351    352    353    354    356    357    358    360    361
##  2      1      2      1      1      1      5      2      1      1      2      1      1      1      1
## 364    365    369    370    372    373    374    375    376    380    381    382    383    385    387
##  1      2      1      1      1      2      1      1      1      2      1      1      2      2      5
## 388    391    392    397    398    400    402    405    408    410    413    415    418    420    421
##  2      2      1      1      1      1      1      1      2      1      1      1      2      2      1
## 423    424    425    428    431    432    433    434    435    437    438    439    442    443    444
##  1      2      2      1      1      2      2      1      1      1      2      1      1      1      1
## 445    446    449    450    451    452    453    454    455    456    459    460    462    463    467
##  2      1      1      1      1      2      2      4      3      1      1      1      1      2      1
## 468    470    471    474    478    479    482    483    485    486    488    490    492    493    494
##  1      1      3      1      2      2      3      2      1      2      1      1      1      2      1
## 495    497    498    499    500    503    504    509    511    515    516    520    521    522    525
##  1      1      1      1      1      1      1      2      1      2      1      2      1      1      1
## 529    535    539    540    541    546    547    549    553    564    565    567    572    574    576
##  1      2      1      1      1      2      2      1      1      2      3      4      2      1      1
## 579    580    582    593    596    598    600    601    602    604    608    615    616    619    621
##  2      1      1      1      1      1      1      2      1      1      2      1      2      1      2
## 622    624    632    635    637    656    658    662    665    669    671    674    680    682    689
##  1      1      1      2      1      1      1      1      2      1      1      1      1      1      1
```

```
## 690 691 696 697 702 707 708 716 719 722 724 729 731 733 735
## 1 1 1 1 1 1 1 1 1 1 1 2 2 1 2
## 736 744 748 756 765 771 773 782 789 791 793 798 799 803 804
## 1 1 2 2 1 1 1 1 2 1 1 2 1 1 1
## 807 811 818 830 832 843 861 947 955 956 957 964 979 991 994
## 1 2 1 1 1 1 1 3 1 1 1 1 1 1 1
## 997 1001 1020 1029 1036 1144 1166 1196
## 1 1 1 1 1 1 1 1
```

## 大部分的产品只有几条销售路线，只有几种产品的销售路线很多

```
ggplot(routes_products)+
  geom_histogram(aes(x=n_routes), fill="red", color="black", alpha="0.5",
  binwidth=10)+
  theme_bw(base_family = "STKaiti") +
  ggtitle("产品和销售线路")+ xlab("销售线路(条)")+ylab("产品数量(种)")
```

产品和销售线路



## 产品和销售路线 top100

```
routes.products <- traindata %>% group_by(Ruta_SAK, Producto_ID) %>%
  summarise(count=n(),
            n_Agencias = n_distinct(Agencia_ID),
            n_Clients = n_distinct(Cliente_ID),
            Units=sum(Venta_uni_hoy),
            Return_Units = sum(Dev_uni_proxima)) %>%
  mutate(Return_Rate = Return_Units / (Units+Return_Units)) %>%
  arrange(desc(Units))
head(routes.products)
```

```
## Ruta_SAK Producto_ID count n_Agencias n_Clients Units Return_Units
## 1 1154 2425 12062 27 1967 388887 2604
## 2 1151 2425 12759 27 2070 377794 2172
## 3 1155 2425 10914 24 1792 372039 2737
## 4 1152 2425 12196 26 2012 370659 2282
## 5 1156 2425 10844 24 1780 361474 2787
## 6 1153 2425 11219 24 1830 352817 2316
## Return_Rate
## 1 0.006651494
## 2 0.005716301
## 3 0.007303029
## 4 0.006118930
## 5 0.007651107
## 6 0.006521500
```

```
dim(routes.products)
```

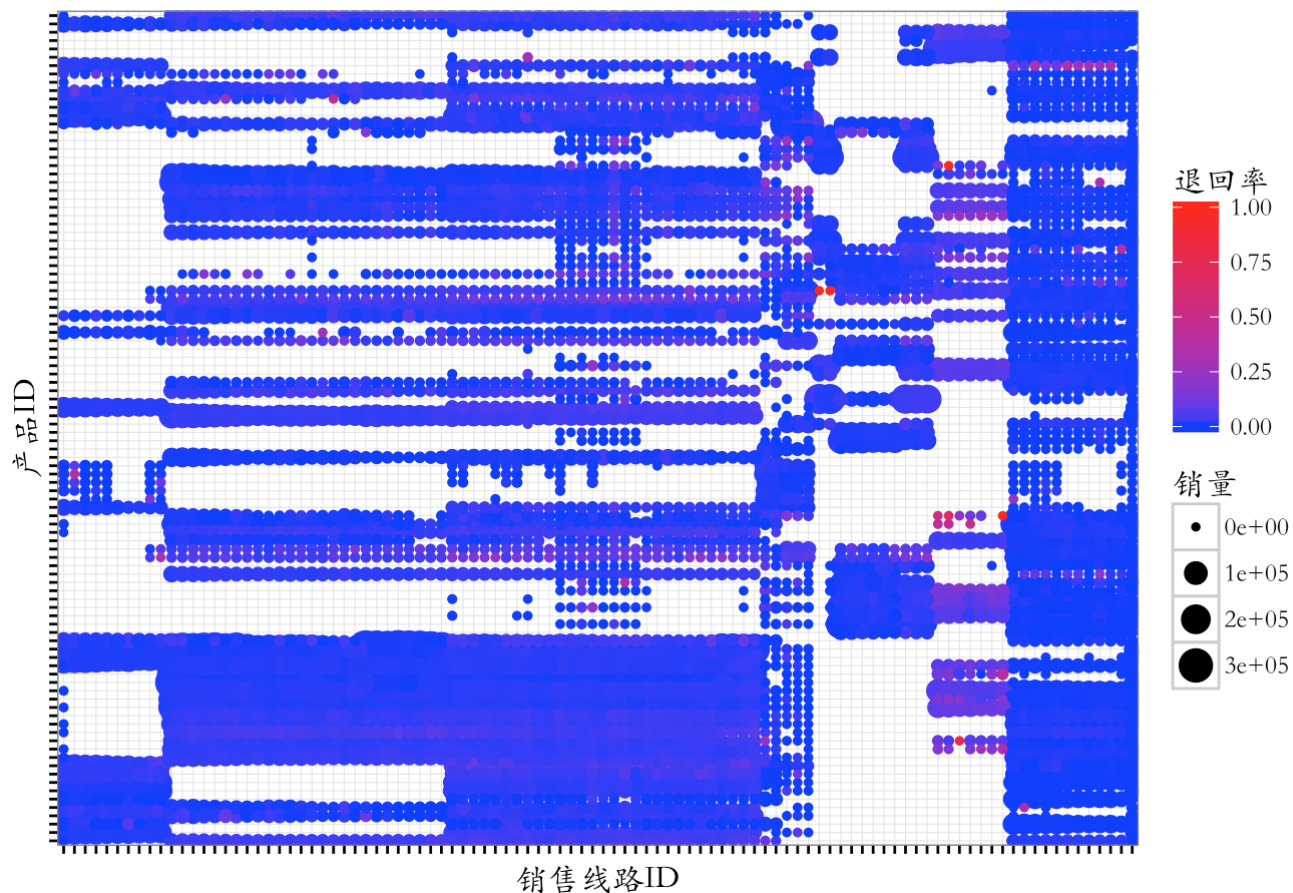
```
## [1] 250265      8
```

```
top100routes <- routes$Ruta_SAK[1:100]
top100products <- products$Producto_ID[1:100]

ggplot(routes.products %>%
  filter(Ruta_SAK %in% top100routes, Producto_ID %in% top100products))+
  geom_point(aes(x=as.character(Ruta_SAK),
    y=as.character(Producto_ID),
    size=Units, color=Return_Rate))+
  theme_bw(base_family = "STKaiti")+
  scale_color_gradient(name="退回率", low="blue", high="red")+
  scale_size_continuous(name = "销量", range = c(1,6)) +
  theme(axis.line=element_blank(),
    axis.text.x=element_blank(),
    axis.text.y=element_blank()) +
  ggtitle("销量前100产品&线路")+ xlab("销售线路ID")+ylab("产品ID")
```



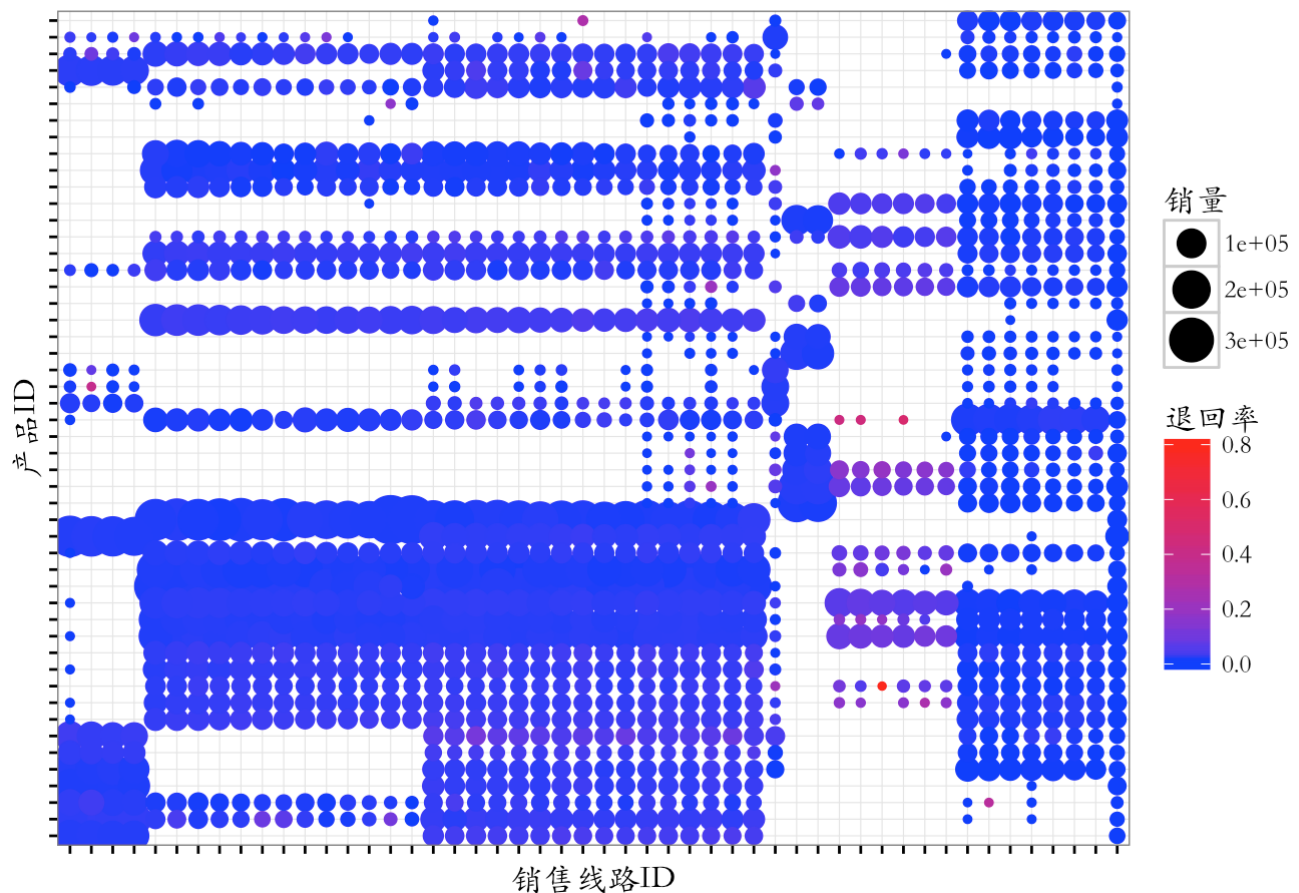
## 销量前100产品&amp;线路



```
## 产品和销售路线 top50
top50routes <- routes$Ruta_SAK[1:50]
top50products <- products$Producto_ID[1:50]

ggplot(routes.products %>%
  filter(Ruta_SAK %in% top50routes, Producto_ID %in% top50products))+
  geom_point(aes(x=as.character(Ruta_SAK),
    y=as.character(Producto_ID),
    size=Units, color=Return_Rate))+
  theme_bw(base_family = "STKaiti")+
  scale_color_gradient(name="退货率", low="blue", high="red")+
  scale_size_continuous(name = "销量", range = c(1,8)) +
  theme(axis.line=element_blank(),
    axis.text.x=element_blank(),
    axis.text.y=element_blank()) +
  ggtitle("销量前50产品&线路")+ xlab("销售线路ID")+ylab("产品ID")
```

销量前50产品&amp;线路



```
## 产品和客户
```

```
products_by_client <- traindata %>%
group_by(Cliente_ID) %>%
  summarise(n_products = n_distinct(Producto_ID)) %>%
  inner_join(cliente_tabla, by="Cliente_ID")
head(products_by_client)
```

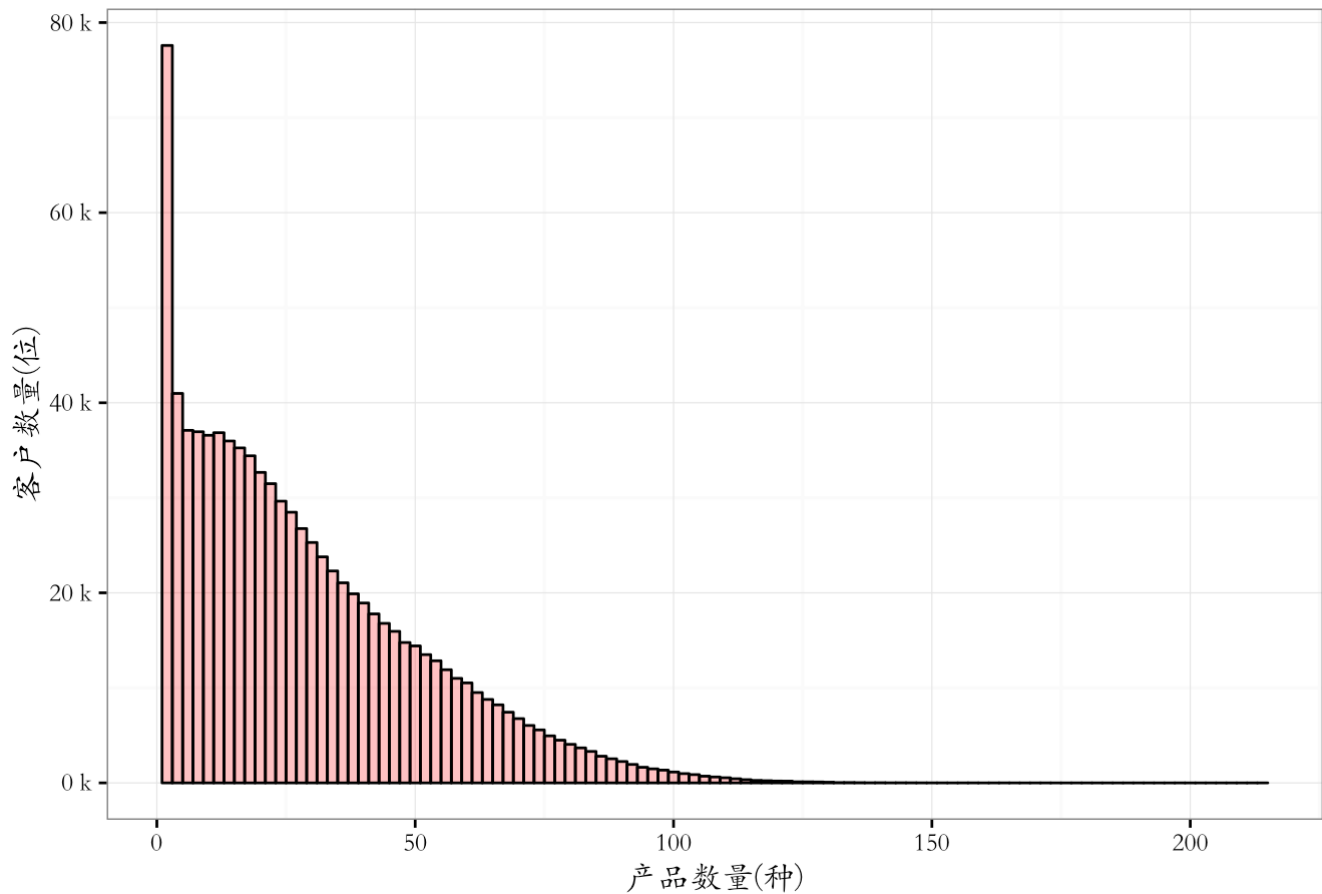
```
##   Cliente_ID n_products      NombreCliente
## 1         26         51 BODEGA COMERCIAL MEXICANA TOLUCA
## 2         60         34          SAMS CLUB TOLUCA
## 3         65        112        WAL MART METEPEC
## 4        101          4        WAL MART TOLUCA
## 5        105         72  SUPER KOMPRAS SAN BUENAVENTURA
## 6        106         21             ISSSTE 21
```

```
dim(products_by_client)
```

```
## [1] 885416      3
```

```
ggplot(products_by_client)+
  geom_histogram(aes(x=n_products), fill="red", color="black", alpha="0.3",
binwidth=2)+
  theme_bw(base_family = "STKaiti")+
  scale_y_continuous(labels=function(x)paste(x/1000, "k"))+
  ggtitle("产品量所对应的客户量")+ xlab("产品数量(种)") + ylab("客户数量(位)")
```

## 产品量所对应的客户量



```
## 需求量的数据分布####
```

```
Demanda_uni_equil <- traindata %>%
  group_by(Producto_ID, Semana, Agencia_ID, Canal_ID, Ruta_SAK) %>%
  summarise(Demanda = sum(Demanda_uni_equil),
            logDemanda = log(sum(Demanda_uni_equil+1))) #销量取对数, 均值
head(Demanda_uni_equil)
```

```
##   Producto_ID Semana Agencia_ID Canal_ID Ruta_SAK Demanda logDemanda
## 1         1212     3         1110      7      3301      55  4.290459
## 2         1216     3         1110      7      3301      27  3.610918
## 3         1238     3         1110      7      3301      49  4.127134
## 4         1240     3         1110      7      3301     109  4.867534
## 5         1242     3         1110      7      3301      37  3.912023
## 6         1250     3         1110      7      3301     165  5.225747
```

```
dim(Demanda_uni_equil)
```

```
## [1] 5363216      7
```

```
summary(Demanda_uni_equil$Demanda)
```

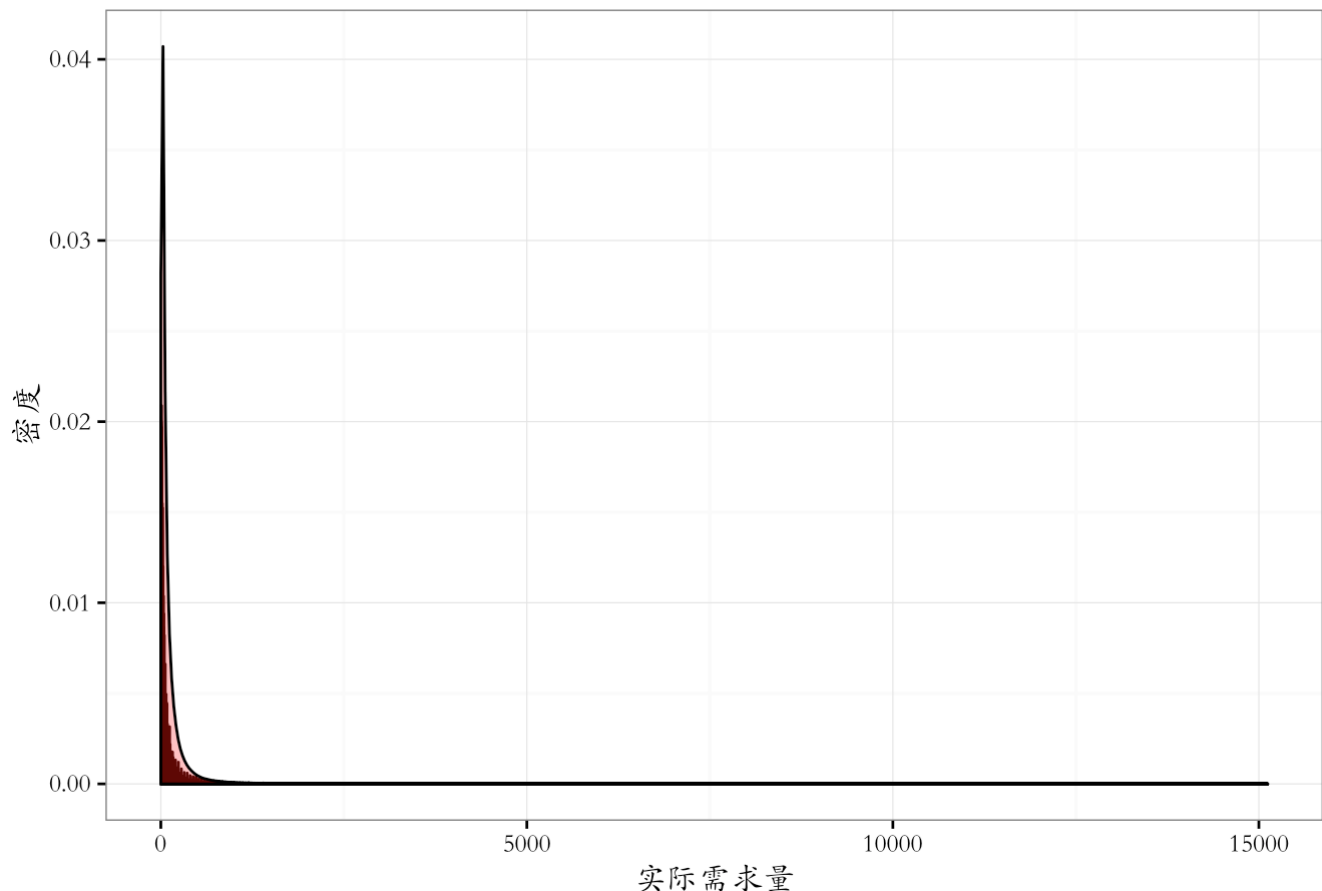
```
##      Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
##      0.00    14.00    40.00    99.93   104.00  15120.00
```

```
summary(Demanda_uni_equil$logDemanda)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.000   2.890   3.892   3.824   4.828   9.624
```

```
ggplot(Demanda_uni_equil,aes(x =Demanda)) +
  geom_histogram(aes(y = ..density..),fill="gray", color="black", alpha="0.8",binwidth
h = 1) +
  geom_density(fill="red", alpha="0.3")+
  theme_bw(base_family = "STKaiti") +
  #scale_x_continuous(lim=c(0, 100))+
  ggtitle("产品需求量的分布")+ xlab("实际需求量")+ylab("密度")
```

产品需求量的分布

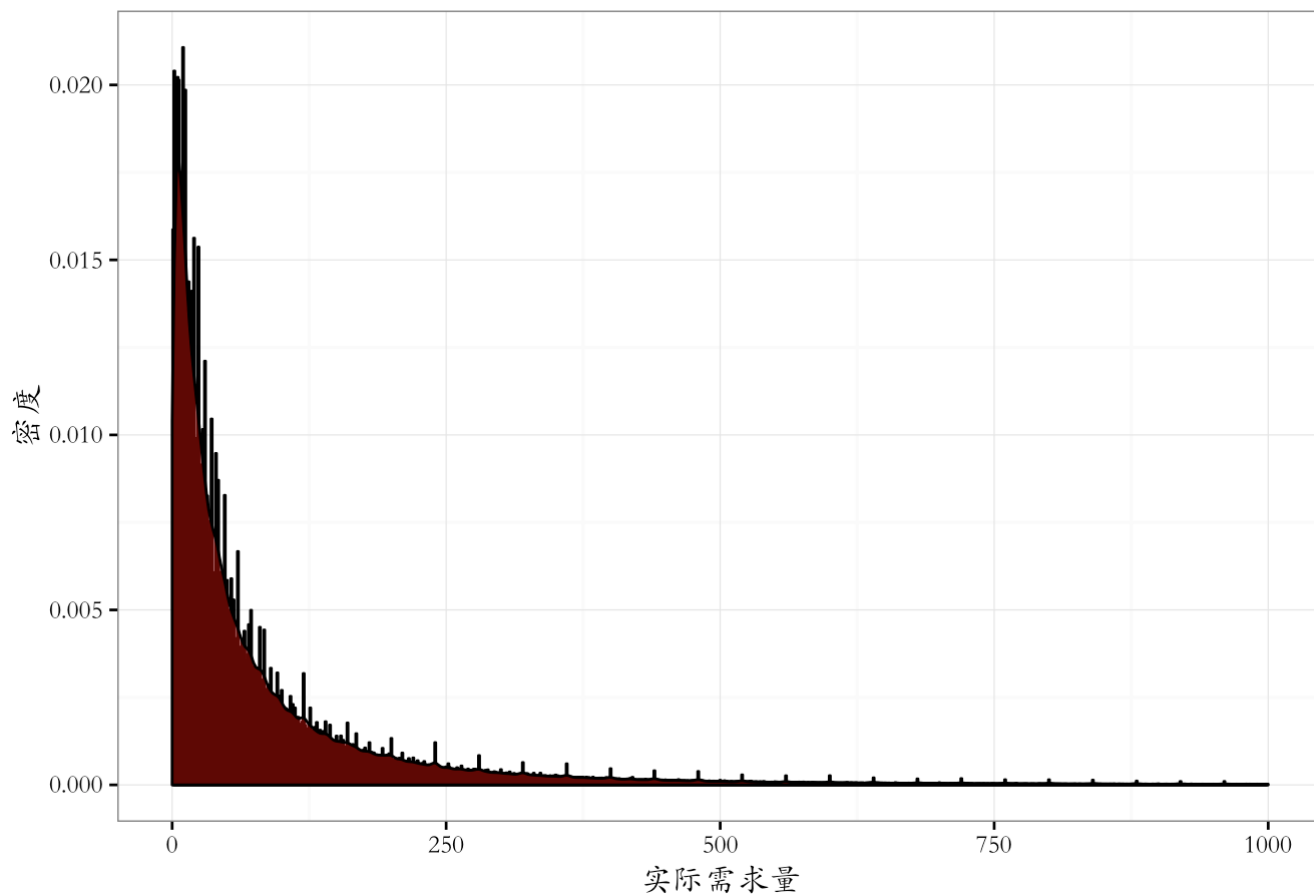


```
ggplot(Demanda_uni_equil,aes(x =Demanda)) +
  geom_histogram(aes(y = ..density..),fill="gray", color="black", alpha="0.8",binwidth
h = 1) +
  geom_density(fill="red", alpha="0.3")+
  theme_bw(base_family = "STKaiti") +
  scale_x_continuous(lim=c(0, 1000))+
  ggtitle("产品需求量的分布")+ xlab("实际需求量")+ylab("密度")
```

```
## Warning: Removed 47788 rows containing non-finite values (stat_bin).
```

```
## Warning: Removed 47788 rows containing non-finite values (stat_density).
```

## 产品需求量的分布



```
ggplot(Demanda_uni_equil,aes(x =logDemanda)) +  
  geom_histogram(aes(y = ..density..),fill="gray", color="black", alpha="0.8",binwidth  
h = 0.4) +  
  geom_density(fill="red", alpha="0.2")+  
  theme_bw(base_family = "STKaiti") +  
  #scale_x_continuous(lim=c(0, 50))+  
  ggtitle("产品需求量的分布")+ xlab("取对数后实际需求")+ylab("密度")
```

产品需求量的分布

