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“Biometry  
for  
Clinical  
Research”  
author:  
“Kishore  
Puthezhath”  
date:  
“2022-  
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12”  
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---

## 1 About

“Scientific etiquette demands that a field be defined before its study is begun”-R  
R Sokal. *Bios* means “life” and *metron* means “measure”.

## 2 r Basics

```
# find out the working directory
getwd()
#> [1] "/Users/drkmemon/Sync/books/biometry"
# setwd in R

setwd("/Users/drkmemon/")

getwd()
#> [1] "/Users/drkmemon"

## change back

setwd("/Users/drkmemon/Sync/books/biometry")
getwd()
#> [1] "/Users/drkmemon/Sync/books/biometry"

# tab key will popout a list of things we may be looking for after entering
# first 1-2 alphabets

# save this as a project

# myrstats<-save.image("myrstats.rData")

# Essential packages

library(knitr)
library(tidyverse)
```

```

#> -- Attaching packages ----- tidyverse 1.3.1 --
#> v ggplot2 3.3.5      v purrr 0.3.4
#> v tibble 3.1.5       v dplyr 1.0.7
#> v tidyr 1.1.4        v stringr 1.4.0
#> v readr 2.0.2        v forcats 0.5.1
#> -- Conflicts ----- tidyverse_conflicts() --
#> x dplyr::filter() masks stats::filter()
#> x dplyr::lag()     masks stats::lag()
library(epiR)
#> Loading required package: survival
#> Package epiR 2.0.39 is loaded
#> Type help(epi.about) for summary information
#> Type browseVignettes(package = 'epiR') to learn how to use epiR for applied epidemiologic
#>
library(epiDisplay)
#> Loading required package: foreign
#> Loading required package: MASS
#>
#> Attaching package: 'MASS'
#> The following object is masked from 'package:dplyr':
#>
#>      select
#> Loading required package: nnet
#>
#> Attaching package: 'epiDisplay'
#> The following object is masked from 'package:ggplot2':
#>
#>      alpha
library(survival)
library(survminer)
#> Loading required package: ggpubr
library(randomizeR)
#> Loading required package: plotrix

```

```

x<- 2
x
#> [1] 2

```

x is an object containing variable 2

```

y<-"male"
y
#> [1] "male"

```

y is an object containing variable “male” x is numeric/integer and “male” is factor/character factors can be grouped for analysis, characters cannot be

```
class(x)
#> [1] "numeric"
class(y)
#> [1] "character"
```

Convert character to factor

```
z= as.factor(y)
class(z)
#> [1] "factor"
```

concatenation: it is a string of interconnected things

```
x1<- c(1,2,3,4,5)
x1
#> [1] 1 2 3 4 5
```

different methods of producing concatenated vectors

```
x1<-c(1:5)
x1<-seq(from=1, to=5, by=1)
x1
#> [1] 1 2 3 4 5
```

we can also create repeated sequence in R

```
x2<-rep(1, times=5)
x2
#> [1] 1 1 1 1 1

x3<-rep(seq(from=2, to=6, by=0.05), times=5)
x3
#> [1] 2.00 2.05 2.10 2.15 2.20 2.25 2.30 2.35 2.40 2.45 2.50
#> [12] 2.55 2.60 2.65 2.70 2.75 2.80 2.85 2.90 2.95 3.00 3.05
#> [23] 3.10 3.15 3.20 3.25 3.30 3.35 3.40 3.45 3.50 3.55 3.60
#> [34] 3.65 3.70 3.75 3.80 3.85 3.90 3.95 4.00 4.05 4.10 4.15
#> [45] 4.20 4.25 4.30 4.35 4.40 4.45 4.50 4.55 4.60 4.65 4.70
#> [56] 4.75 4.80 4.85 4.90 4.95 5.00 5.05 5.10 5.15 5.20 5.25
#> [67] 5.30 5.35 5.40 5.45 5.50 5.55 5.60 5.65 5.70 5.75 5.80
#> [78] 5.85 5.90 5.95 6.00 2.00 2.05 2.10 2.15 2.20 2.25 2.30
#> [89] 2.35 2.40 2.45 2.50 2.55 2.60 2.65 2.70 2.75 2.80 2.85
#> [100] 2.90 2.95 3.00 3.05 3.10 3.15 3.20 3.25 3.30 3.35 3.40
#> [111] 3.45 3.50 3.55 3.60 3.65 3.70 3.75 3.80 3.85 3.90 3.95
#> [122] 4.00 4.05 4.10 4.15 4.20 4.25 4.30 4.35 4.40 4.45 4.50
#> [133] 4.55 4.60 4.65 4.70 4.75 4.80 4.85 4.90 4.95 5.00 5.05
#> [144] 5.10 5.15 5.20 5.25 5.30 5.35 5.40 5.45 5.50 5.55 5.60
#> [155] 5.65 5.70 5.75 5.80 5.85 5.90 5.95 6.00 2.00 2.05 2.10
#> [166] 2.15 2.20 2.25 2.30 2.35 2.40 2.45 2.50 2.55 2.60 2.65
```

```

#> [177] 2.70 2.75 2.80 2.85 2.90 2.95 3.00 3.05 3.10 3.15 3.20
#> [188] 3.25 3.30 3.35 3.40 3.45 3.50 3.55 3.60 3.65 3.70 3.75
#> [199] 3.80 3.85 3.90 3.95 4.00 4.05 4.10 4.15 4.20 4.25 4.30
#> [210] 4.35 4.40 4.45 4.50 4.55 4.60 4.65 4.70 4.75 4.80 4.85
#> [221] 4.90 4.95 5.00 5.05 5.10 5.15 5.20 5.25 5.30 5.35 5.40
#> [232] 5.45 5.50 5.55 5.60 5.65 5.70 5.75 5.80 5.85 5.90 5.95
#> [243] 6.00 2.00 2.05 2.10 2.15 2.20 2.25 2.30 2.35 2.40 2.45
#> [254] 2.50 2.55 2.60 2.65 2.70 2.75 2.80 2.85 2.90 2.95 3.00
#> [265] 3.05 3.10 3.15 3.20 3.25 3.30 3.35 3.40 3.45 3.50 3.55
#> [276] 3.60 3.65 3.70 3.75 3.80 3.85 3.90 3.95 4.00 4.05 4.10
#> [287] 4.15 4.20 4.25 4.30 4.35 4.40 4.45 4.50 4.55 4.60 4.65
#> [298] 4.70 4.75 4.80 4.85 4.90 4.95 5.00 5.05 5.10 5.15 5.20
#> [309] 5.25 5.30 5.35 5.40 5.45 5.50 5.55 5.60 5.65 5.70 5.75
#> [320] 5.80 5.85 5.90 5.95 6.00 2.00 2.05 2.10 2.15 2.20 2.25
#> [331] 2.30 2.35 2.40 2.45 2.50 2.55 2.60 2.65 2.70 2.75 2.80
#> [342] 2.85 2.90 2.95 3.00 3.05 3.10 3.15 3.20 3.25 3.30 3.35
#> [353] 3.40 3.45 3.50 3.55 3.60 3.65 3.70 3.75 3.80 3.85 3.90
#> [364] 3.95 4.00 4.05 4.10 4.15 4.20 4.25 4.30 4.35 4.40 4.45
#> [375] 4.50 4.55 4.60 4.65 4.70 4.75 4.80 4.85 4.90 4.95 5.00
#> [386] 5.05 5.10 5.15 5.20 5.25 5.30 5.35 5.40 5.45 5.50 5.55
#> [397] 5.60 5.65 5.70 5.75 5.80 5.85 5.90 5.95 6.00

```

extracting elements from concatenated string

```

x1
#> [1] 1 2 3 4 5
x1[3]
#> [1] 3
x1[1:3]
#> [1] 1 2 3
x1[c(2,4)]
#> [1] 2 4
x1[-1]
#> [1] 2 3 4 5

```

Matrix of elements Vector is a list of numbers/characters Matrix is an array of numbers/characters in rows and columns

```

m1<-matrix(c(1:20),nrow=5, byrow=T)
m1
#>      [,1] [,2] [,3] [,4]
#> [1,]    1    2    3    4
#> [2,]    5    6    7    8
#> [3,]    9   10   11   12
#> [4,]   13   14   15   16
#> [5,]   17   18   19   20

```

similar to vectors, matrix can be subsetted

```
m1[c(2,3), 2]
#> [1] 6 10
```

data\_frame is a type of matrix we can create data\_frame in R or we can import

Creating data\_frame first we have to create concatenated strings of variable name<-c(letters[1:10]) age<-seq(from=63, to=82, by=2) type\_surg<-c(0,1,0,0,1,1,1,0,0,0)

### 3 R recognizes this as a number series, we have to covert this to factor

```
type_surg<-as.factor(type_surg)
pri_event<-c(0,0,0,0,0,1,1,0,0,1) pri_event<-as.factor(pri_event) time<-
c(24,24,24,24,24,3,2,24,24,7) test_data<- data.frame(name,age,type_surg,pri_event,time)
test_data
```

### 4 subsetting can be done. important to remember to specify the column as blank

### 5 after a coma

```
ageo70<-test_data[age>70,] ageo70
```

```
<!--chapter:end:1.1-rbasics.Rmd-->
```

```
# Data handling
```

```
```r
```

```
# create data_frame
```

```
# first we have to create concatenated strings of variable
name<-c(letters[1:10])
age<-seq(from=63, to=82, by=2)
type_surg<- c(0,1,0,0,1,1,1,0,0,0)
```

```
# R recognizes this as a number series, we have to covert this to factor
type_surg<-as.factor(type_surg)
```

```
pri_event<-c(0,0,0,0,0,1,1,0,0,1)
```

```

pri_event<-as.factor(pri_event)
time<-c(24,24,24,24,24,3,2,24,24,7)
test_data<- data.frame(name,age,type_surg,pri_event,time)
test_data
#>   name age type_surg pri_event time
#> 1    a  63         0         0   24
#> 2    b  65         1         0   24
#> 3    c  67         0         0   24
#> 4    d  69         0         0   24
#> 5    e  71         1         0   24
#> 6    f  73         1         1    3
#> 7    g  75         1         1    2
#> 8    h  77         0         0   24
#> 9    i  79         0         0   24
#> 10   j  81         0         1    7

#logic operations

# add new row to test_data2, cbind and rbind

new_data<-c("k", 63,0,0,24)
new_data
#> [1] "k"  "63" "0"  "0"  "24"

test_data2<-rbind(test_data,new_data)
test_data2
#>   name age type_surg pri_event time
#> 1    a  63         0         0   24
#> 2    b  65         1         0   24
#> 3    c  67         0         0   24
#> 4    d  69         0         0   24
#> 5    e  71         1         0   24
#> 6    f  73         1         1    3
#> 7    g  75         1         1    2
#> 8    h  77         0         0   24
#> 9    i  79         0         0   24
#> 10   j  81         0         1    7
#> 11   k  63         0         0   24

new_data2<-c(10:1)
test_data3<-cbind(test_data,new_data2)
test_data3
#>   name age type_surg pri_event time new_data2
#> 1    a  63         0         0   24         10
#> 2    b  65         1         0   24          9
#> 3    c  67         0         0   24          8

```



```

#> 4      d 69          0          0 24          7
#> 5      e 71          1          0 24          6
#> 6      f 73          1          1  3          5
#> 7      g 75          1          1  2          4
#> 8      h 77          0          0 24          3
#> 9      i 79          0          0 24          2
#> 10     j 81          0          1  7          1

# other logic operations, on the test_data in the r basics script

# is age >70?

typage<-age>70
typage[1:5]
#> [1] FALSE FALSE FALSE FALSE  TRUE

# get this answer as 0 and 1

typage2<-as.numeric(age>70)
typage2
#> [1] 0 0 0 0 1 1 1 1 1 1

# multiple logic operations

oldtha<-age>70 & type_surg=="1"
oldtha
#> [1] FALSE FALSE FALSE FALSE  TRUE  TRUE  TRUE FALSE FALSE
#> [10] FALSE

# add this to our data as a new column

test_data
#>   name age type_surg pri_event time
#> 1    a  63         0         0  24
#> 2    b  65         1         0  24
#> 3    c  67         0         0  24
#> 4    d  69         0         0  24
#> 5    e  71         1         0  24
#> 6    f  73         1         1   3
#> 7    g  75         1         1   2
#> 8    h  77         0         0  24
#> 9    i  79         0         0  24
#> 10   j  81         0         1   7
test_data5<-cbind(test_data,oldtha)
test_data5
#>   name age type_surg pri_event time oldtha

```

```

#> 1      a 63      0      0 24 FALSE
#> 2      b 65      1      0 24 FALSE
#> 3      c 67      0      0 24 FALSE
#> 4      d 69      0      0 24 FALSE
#> 5      e 71      1      0 24  TRUE
#> 6      f 73      1      1  3  TRUE
#> 7      g 75      1      1  2  TRUE
#> 8      h 77      0      0 24 FALSE
#> 9      i 79      0      0 24 FALSE
#> 10     j 81      0      1  7  FALSE

# Clearing workspace inr

rm(list=ls())

# remember how to import dataset.
# new_datax<-read.csv(file.choose(), header = T)

# create table
# table()

```

## 6 Data Cleaning

Main actions are `select()`, `filter()`, `group_by()`, `mutate()`, `summarise()`, `full_join()`, `pivot_wide()` and `pivot_long()`, `spread()`, `map()`, `strsplit()`

```

library(tidyverse)
#> -- Attaching packages ----- tidyverse 1.3.1 --
#> v ggplot2 3.3.5      v purrr  0.3.4
#> v tibble  3.1.5      v dplyr  1.0.7
#> v tidyr   1.1.4      v stringr 1.4.0
#> v readr   2.0.2      v forcats 0.5.1
#> -- Conflicts ----- tidyverse_conflicts() --
#> x dplyr::filter() masks stats::filter()
#> x dplyr::lag()     masks stats::lag()
# Use read_csv/read_tsv insted of read.csv
# This will create tibble insted of data frame
booking= read_csv('data/bookings.csv')
#> Rows: 10000 Columns: 8
#> -- Column specification -----
#> Delimiter: ","
#> chr (3): booker_id, checkin_day, status
#> dbl (4): property_id, room_nights, price_per_night, revi...
#> lgl (1): for_business
#>

```

```

#> i Use `spec()` to retrieve the full column specification for this data.
#> i Specify the column types or set `show_col_types = FALSE` to quiet this message.
property=read_csv('data/properties.csv')
#> Rows: 4178 Columns: 5
#> -- Column specification -----
#> Delimiter: ","
#> chr (3): destination, property_type, facilities
#> dbl (2): property_id, nr_rooms
#>
#> i Use `spec()` to retrieve the full column specification for this data.
#> i Specify the column types or set `show_col_types = FALSE` to quiet this message.

booking
#> # A tibble: 10,000 x 8
#>   booker_id      property_id room_nights price_per_night
#>   <chr>          <dbl>         <dbl>         <dbl>
#> 1 215934017ba98c09~      2668             4             91.5
#> 2 7f590fd6d318248a~      4656             5             107.
#> 3 10f0f138e8bb1015~      4563             6             87.0
#> 4 7b55021a4160dde6~      4088             7             92.4
#> 5 6694a79d158c7818~      2188             4             105.
#> 6 d0358740d5f15e85~      4171             2             110.
#> 7 944e568a0b511b91~      2907             4             116.
#> 8 95476c2ef6bb9e3c~      5141             4             111.
#> 9 df235631a4c281c0~      1696             1             106.
#> 10 ff610140227d40d2~      1901             7             82.3
#> # ... with 9,990 more rows, and 4 more variables:
#> #   checkin_day <chr>, for_business <lgl>, status <chr>,
#> #   review_score <dbl>

property
#> # A tibble: 4,178 x 5
#>   property_id destination property_type nr_rooms facilities
#>   <dbl> <chr>         <chr>         <dbl> <chr>
#> 1      2668 Brisbane   Hotel             32 airport s~
#> 2      4656 Brisbane   Hotel             39 on-site r~
#> 3      4563 Brisbane   Apartment          9 laundry
#> 4      4088 Brisbane   Apartment          9 kitchen,l~
#> 5      2188 Brisbane   Apartment          4 parking,k~
#> 6      4171 Brisbane   Apartment          5 kitchen,p~
#> 7      2907 Brisbane   Hotel             22 airport s~
#> 8      5141 Brisbane   Hotel             20 breakfast~
#> 9      1696 Brisbane   Apartment          5 free wifi~
#> 10     1901 Brisbane   Apartment         11 free wifi~
#> # ... with 4,168 more rows

```

```

#magrittr (pipe function, keyboard short cut: command+shift+m)
# %>%
# select() and filter() functions help to extract data and study it

x=booking %>%
  select(review_score)
x
#> # A tibble: 10,000 x 1
#>   review_score
#>   <dbl>
#> 1      NA
#> 2      NA
#> 3      6.26
#> 4      5.95
#> 5      6.43
#> 6      NA
#> 7      7.60
#> 8      NA
#> 9      6.97
#> 10     NA
#> # ... with 9,990 more rows

y=booking %>%
  filter(status=='stayed'&!is.na(review_score))
y
#> # A tibble: 6,183 x 8
#>   booker_id      property_id room_nights price_per_night
#>   <chr>          <dbl>         <dbl>         <dbl>
#> 1 10f0f138e8bb1015~      4563             6          87.0
#> 2 7b55021a4160dde6~      4088             7          92.4
#> 3 6694a79d158c7818~      2188             4          105.
#> 4 944e568a0b511b91~      2907             4          116.
#> 5 df235631a4c281c0~      1696             1          106.
#> 6 5a1442f4c7237ec5~      2307             9          84.2
#> 7 39804a2e3fb2e4c6~      2907             6          112.
#> 8 e150e559405ef29b~      2870             4          127.
#> 9 4e9c7c21dfcf2758~      1674             5          102.
#> 10 4a2b8eaf63613548~      2885             5          86.3
#> # ... with 6,173 more rows, and 4 more variables:
#> #   checkin_day <chr>, for_business <lgl>, status <chr>,
#> #   review_score <dbl>

cheap=booking %>%
  select(review_score,room_nights) %>%
  filter(booking$price_per_night<80)

```

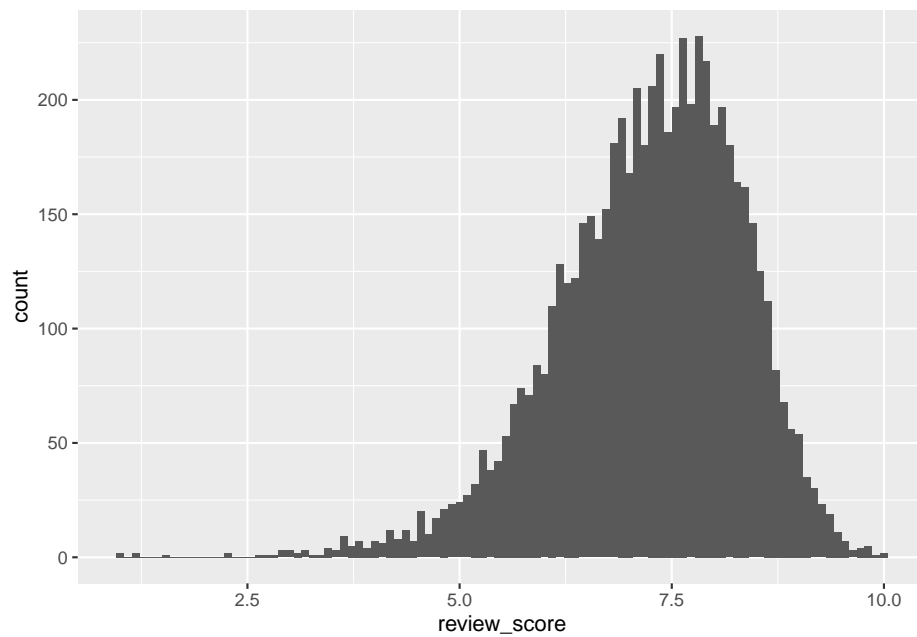
```

cheap
#> # A tibble: 434 x 2
#>   review_score room_nights
#>   <dbl>         <dbl>
#> 1      8.90           5
#> 2      5.87           6
#> 3      NA            4
#> 4      NA            7
#> 5      6.02           4
#> 6      9.64           6
#> 7      NA            3
#> 8      NA            4
#> 9      6.23           5
#> 10     NA            2
#> # ... with 424 more rows

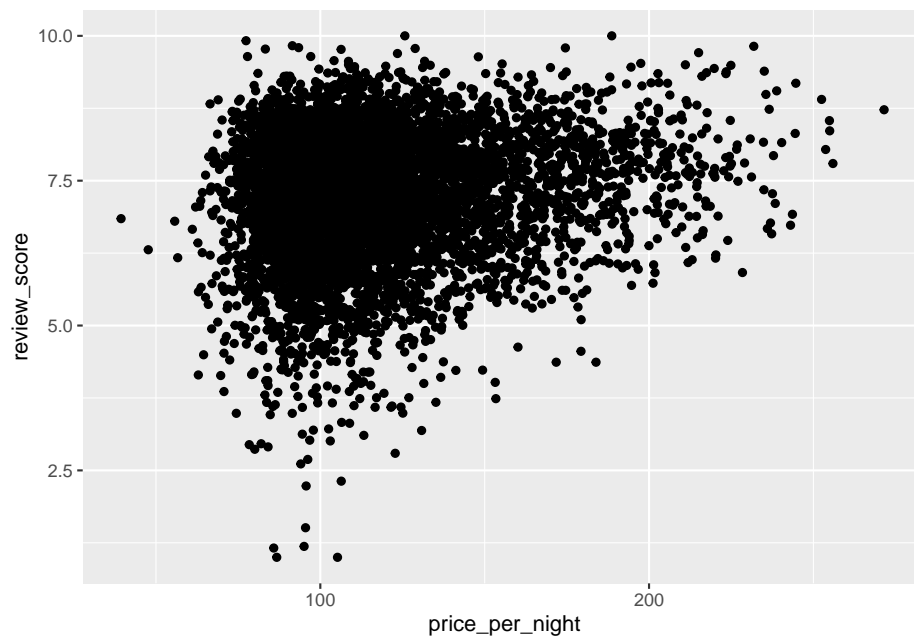
booking %>%
  filter(checkin_day=='wed') %>%
  select(property_id,status) %>%
  head(2)
#> # A tibble: 2 x 2
#>   property_id status
#>   <dbl> <chr>
#> 1    4563 stayed
#> 2    5141 cancelled

#ggplot2: for plotting
# ggplot(aes()+geom_histogram()/geom_point())....
booking %>%
ggplot(
  aes(review_score)
)+geom_histogram(bins = 100)
#> Warning: Removed 3817 rows containing non-finite values
#> (stat_bin).

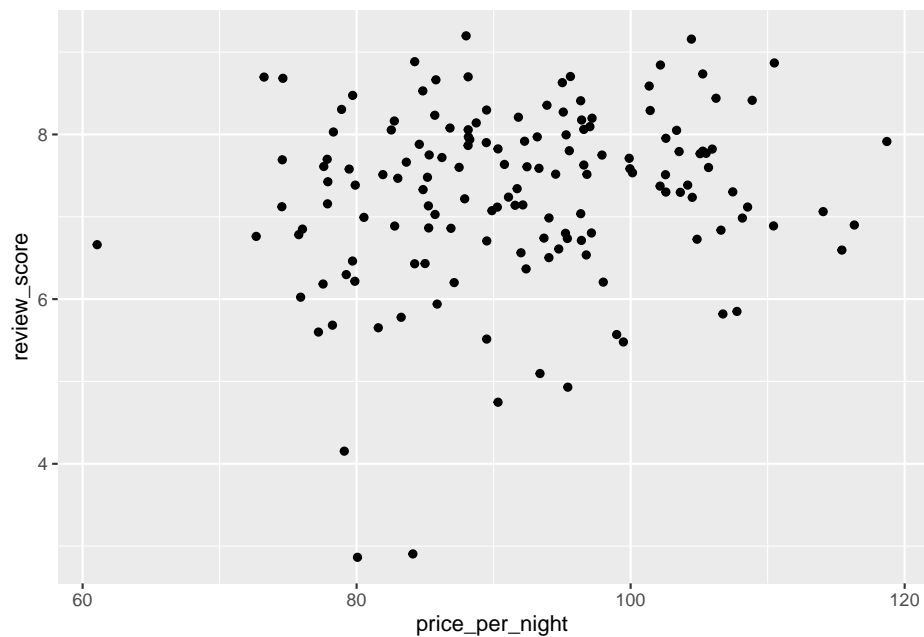
```



```
booking %>%  
  ggplot(  
    aes(price_per_night, review_score)  
  ) + geom_point()  
#> Warning: Removed 3817 rows containing missing values  
#> (geom_point).
```



```
booking %>%  
  filter(room_nights>7, status=='stayed') %>%  
  select(price_per_night,review_score) %>%  
  ggplot(  
    aes(price_per_night,review_score)  
  )+geom_point()  
#> Warning: Removed 41 rows containing missing values  
#> (geom_point).
```



```
#mutate

booking %>%
  mutate(centered_mean=price_per_night-mean(price_per_night)) %>%
  head(2)
#> # A tibble: 2 x 9
#>   booker_id      property_id room_nights price_per_night
#>   <chr>          <dbl>      <dbl>      <dbl>
#> 1 215934017ba98c09f~      2668         4        91.5
#> 2 7f590fd6d318248a4~      4656         5       107.
#> # ... with 5 more variables: checkin_day <chr>,
#> #   for_business <lgl>, status <chr>, review_score <dbl>,
#> #   centered_mean <dbl>

# summarise: extracts the number of variables

booking %>%
  summarise(
    n()
    , n_miss=sum(is.na(review_score))
    , mean_score=mean(review_score, na.rm=T))
#> # A tibble: 1 x 3
#>   `n()` n_miss mean_score
#>   <int> <int>    <dbl>
#> 1 10000  3817     7.22
```



```

booking %>%
  summarise(
    n()
    , stayed_booking=sum(status=='stayed')
    , mean_total=mean(price_per_night*room_nights)
  )
#> # A tibble: 1 x 3
#>   `n()` stayed_booking mean_total
#>   <int>      <int>      <dbl>
#> 1 10000      7775      348.
#group by
booking %>%
  group_by(
    for_business
  ) %>%
  summarise(
    n=n()
    , mean_review=mean(review_score,na.rm=T))
#> # A tibble: 2 x 3
#>   for_business      n mean_review
#>   <lgl>          <int>      <dbl>
#> 1 FALSE        6235      7.50
#> 2 TRUE         3715      6.85

mixed=booking %>%
  full_join(property) %>%
  count(destination,checkin_day) %>%
  pivot_wider(
    names_from = checkin_day,values_from = n
  )
#> Joining, by = "property_id"
mixed
#> # A tibble: 3 x 8
#>   destination  fri    mon    sat    sun    thu    tue    wed
#>   <chr>      <int> <int> <int> <int> <int> <int> <int>
#> 1 Amsterdam   1074   517   889   813   667   498   542
#> 2 Brisbane    162   133   114   153   162   148   128
#> 3 Tokyo       451   718   322   576   718   655   560

# make a long data form

long=mixed%>%
  pivot_longer(cols = 2:8, names_to = 'day',values_to = 'count')
long

```

```

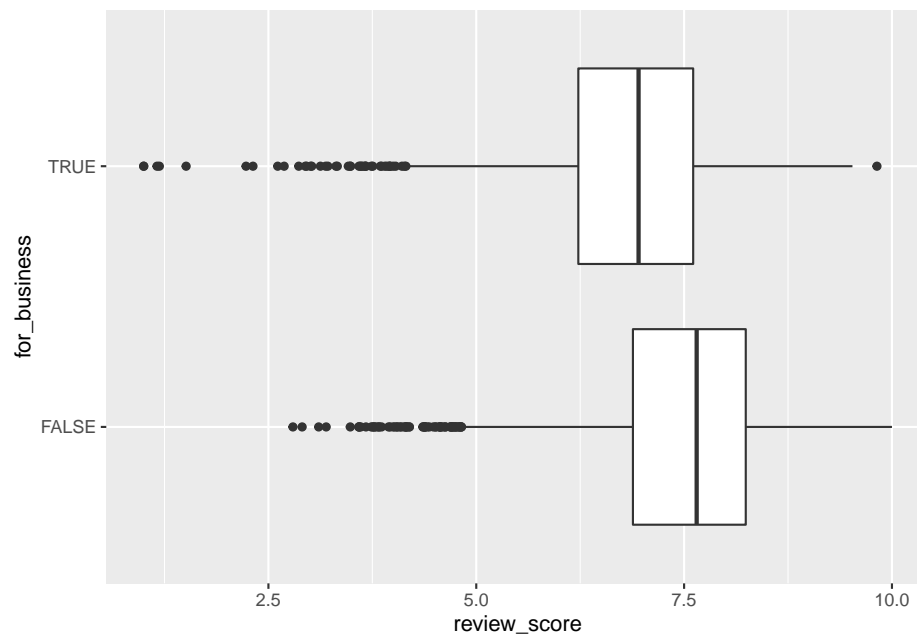
#> # A tibble: 21 x 3
#>   destination day    count
#>   <chr>      <chr> <int>
#> 1 Amsterdam  fri    1074
#> 2 Amsterdam  mon     517
#> 3 Amsterdam  sat     889
#> 4 Amsterdam  sun     813
#> 5 Amsterdam  thu     667
#> 6 Amsterdam  tue     498
#> 7 Amsterdam  wed     542
#> 8 Brisbane  fri     162
#> 9 Brisbane  mon     133
#> 10 Brisbane  sat     114
#> # ... with 11 more rows

# make long data form

wide= long %>%
  pivot_wider(names_from = "day", values_from = "count")
wide
#> # A tibble: 3 x 8
#>   destination  fri    mon    sat    sun    thu    tue    wed
#>   <chr>      <int> <int> <int> <int> <int> <int> <int>
#> 1 Amsterdam   1074   517   889   813   667   498   542
#> 2 Brisbane    162   133   114   153   162   148   128
#> 3 Tokyo       451   718   322   576   718   655   560

# Boxplot with ggplot2
booking %>%
  ggplot(
    aes(
      review_score, for_business
    )
  ) + geom_boxplot()
#> Warning: Removed 3817 rows containing non-finite values
#> (stat_boxplot).

```



```
# hash the property id
# we need to know map() function to do this. map(x,~.), where x = object and ~. is a function
library(digest)

property %>%
  mutate(property_id=map_chr(property_id,digest))
#> # A tibble: 4,178 x 5
#>   property_id destination property_type nr_rooms facilities
#>   <chr>          <chr>          <chr>          <dbl> <chr>
#> 1 c5fe5a36c3~ Brisbane      Hotel              32 airport s~
#> 2 6abfc65c14~ Brisbane      Hotel              39 on-site r~
#> 3 8740143b90~ Brisbane      Apartment          9 laundry
#> 4 e30b95c1ec~ Brisbane      Apartment          9 kitchen,l~
#> 5 ab19240af8~ Brisbane      Apartment          4 parking,k~
#> 6 b2efd881c3~ Brisbane      Apartment          5 kitchen,p~
#> 7 d49c23b12c~ Brisbane      Hotel              22 airport s~
#> 8 1fd9f14595~ Brisbane      Hotel              20 breakfast~
#> 9 7319c32a43~ Brisbane      Apartment          5 free wifi~
#> 10 a38cc66d5f~ Brisbane      Apartment         11 free wifi~
#> # ... with 4,168 more rows

# list in data frame
# If we have a column with multiple strings, we can split it in to vectors using strsplit()
```

```

property %>%
  mutate(facilities=strsplit(facilities","))
#> # A tibble: 4,178 x 5
#>   property_id destination property_type nr_rooms facilities
#>   <dbl> <chr> <chr> <dbl> <list>
#> 1      2668 Brisbane Hotel      32 <chr [6]>
#> 2      4656 Brisbane Hotel      39 <chr [7]>
#> 3      4563 Brisbane Apartment    9 <chr [1]>
#> 4      4088 Brisbane Apartment    9 <chr [3]>
#> 5      2188 Brisbane Apartment    4 <chr [5]>
#> 6      4171 Brisbane Apartment    5 <chr [6]>
#> 7      2907 Brisbane Hotel      22 <chr [8]>
#> 8      5141 Brisbane Hotel      20 <chr [8]>
#> 9      1696 Brisbane Apartment    5 <chr [6]>
#> 10     1901 Brisbane Apartment   11 <chr [8]>
#> # ... with 4,168 more rows
property$facilities[1]
#> [1] "airport shuttle,free wifi,garden,breakfast,pool,on-site restaurant"

# add a column with the number of facilities

property %>%
  mutate(facilities=strsplit(facilities",")) %>%
  mutate(n_facility=map_int(facilities,length))
#> # A tibble: 4,178 x 6
#>   property_id destination property_type nr_rooms facilities
#>   <dbl> <chr> <chr> <dbl> <list>
#> 1      2668 Brisbane Hotel      32 <chr [6]>
#> 2      4656 Brisbane Hotel      39 <chr [7]>
#> 3      4563 Brisbane Apartment    9 <chr [1]>
#> 4      4088 Brisbane Apartment    9 <chr [3]>
#> 5      2188 Brisbane Apartment    4 <chr [5]>
#> 6      4171 Brisbane Apartment    5 <chr [6]>
#> 7      2907 Brisbane Hotel      22 <chr [8]>
#> 8      5141 Brisbane Hotel      20 <chr [8]>
#> 9      1696 Brisbane Apartment    5 <chr [6]>
#> 10     1901 Brisbane Apartment   11 <chr [8]>
#> # ... with 4,168 more rows, and 1 more variable:
#> #   n_facility <int>

```