

# Apuntes Data Cleaning II

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## Exercise 1\_Sales

Data\_Sales: [https://assets.datacamp.com/production/course\\_1294/datasets/sales.csv](https://assets.datacamp.com/production/course_1294/datasets/sales.csv)

### Importing

```
#Import sales.csv to the variable sales using the read.csv() function. Set the stringsAsFactors argument
sales<- read.csv("sales.csv",stringsAsFactors=FALSE)
```

### Examining the data

```
# View dimensions of sales
dim(sales)
```

```
## [1] 5000 46
```

```
# Inspect first 6 rows of sales
head(sales, n=6)
```

```
##      X          event_id      primary_act_id      secondary_act_id
## 1 1 abca1ad99a935fc661 43f0436b905bfa7c2eec b85143bf51323b72e53c
## 2 2 6c56d7f08c95f2aa453c 1a3e9aec0617706a794 f53529c5679ea6ca5a48
## 3 3 c7ab4524a121f9d687d2 4b677c3f5bec71eec8d1 b85143bf51323b72e53c
## 4 4 394cb493f893be9b9ed1 b1ccea01ad6ef8522796 b85143bf51323b72e53c
## 5 5 55b5f67e618557929f48 91c03a34b562436efa3c b85143bf51323b72e53c
## 6 6 4f10fd8b9f550352bd56 ac4b847b3fde66f2117e 63814f3d63317f1b56c4
##      purch_party_lkup_id
## 1 7dfa56dd7d5956b17587
## 2 4f9e6fc637eaf7b736c2
## 3 6c2545703bd527a7144d
## 4 527d6b1eaffc69ddd882
## 5 8bd62c394a35213bdf52
## 6 3b3a628f83135acd0676
##
##                                     event_name
## 1 Xfinity Center Mansfield Premier Parking: Florida Georgia Line
## 2                               Gorge Camping - dave matthews band - sept 3-7
```

```

## 3          Dodge Theatre Adams Street Parking - benise
## 4  Gexa Energy Pavilion Vip Parking : kid rock with sheryl crow
## 5          Premier Parking - motley crue
## 6          Fast Lane Access: Journey
##          primary_act_name secondary_act_name
## 1 XFINITY Center Mansfield Premier Parking          NULL
## 2          Gorge Camping Dave Matthews Band
## 3          Parking Event          NULL
## 4          Gexa Energy Pavilion VIP Parking          NULL
## 5 White River Amphitheatre Premier Parking          NULL
## 6          Fast Lane Access          Journey
##  major_cat_name      minor_cat_name la_event_type_cat
## 1          MISC          PARKING          PARKING
## 2          MISC          CAMPING          INVALID
## 3          MISC          PARKING          PARKING
## 4          MISC          PARKING          PARKING
## 5          MISC          PARKING          PARKING
## 6          MISC SPECIAL ENTRY (UPSELL)          UPSSELL
##          event_disp_name
## 1 Xfinity Center Mansfield Premier Parking: Florida Georgia Line
## 2          Gorge Camping - dave matthews band - sept 3-7
## 3          Dodge Theatre Adams Street Parking - benise
## 4  Gexa Energy Pavilion Vip Parking : kid rock with sheryl crow
## 5          Premier Parking - motley crue
## 6          Fast Lane Access: Journey
##
## 1  THIS TICKET IS VALID          FOR PARKING ONLY          GOOD THIS DAY ONLY          PREMIER PARKING PASS
## 2          %OVERNIGHT C A M P I N G%* * * * *
## 3          ADAMS STREET GARAGE%PARKING FOR 4/21/06 ONLY%DODGE THEATRE PARKING PASS
## 4  THIS TICKET IS VALID          FOR PARKING ONLY          GOOD FOR THIS DATE ONLY          VIP PARKING PASS
## 5          THIS TICKET IS VALID%FOR PARKING ONLY%GOOD THIS DATE ONLY%PREMIER PARKING PASS
## 6          FAST LANE          JOURNEY          FAST LANE EVENT          THIS IS NOT A TICKET
##  tickets_purchased_qty trans_face_val_amt delivery_type_cd
## 1          1          45          eTicket
## 2          1          75          TicketFast
## 3          1          5          TicketFast
## 4          1          20          Mail
## 5          1          20          Mail
## 6          2          10          TicketFast
##          event_date_time  event_dt presale_dt  onsale_dt
## 1 2015-09-12 23:30:00 2015-09-12          NULL 2015-05-15
## 2 2009-09-05 01:00:00 2009-09-04          NULL 2009-03-13
## 3 2006-04-22 01:30:00 2006-04-21          NULL 2006-02-25
## 4 2011-09-03 00:00:00 2011-09-02          NULL 2011-04-22
## 5 2005-07-31 01:00:00 2005-07-30 2005-03-02 2005-03-04
## 6 2012-07-22 02:00:00 2012-07-21          NULL 2012-04-11
##  sales_ord_create_dttm sales_ord_tran_dt  print_dt timezn_nm
## 1 2015-09-11 18:17:45          2015-09-11 2015-09-12          EST
## 2 2009-07-06 00:00:00          2009-07-05 2009-09-01          PST
## 3 2006-04-05 00:00:00          2006-04-05 2006-04-05          MST
## 4 2011-07-01 17:38:50          2011-07-01 2011-07-06          CST
## 5 2005-06-18 00:00:00          2005-06-18 2005-06-28          PST
## 6 2012-07-21 17:20:18          2012-07-21 2012-07-21          PST
##          venue_city  venue_state venue_postal_cd_sgmt_1

```

```

## 1      MANSFIELD MASSACHUSETTS      02048
## 2      QUINCY      WASHINGTON      98848
## 3      PHOENIX      ARIZONA      85003
## 4      DALLAS      TEXAS      75210
## 5      AUBURN      WASHINGTON      98092
## 6 SAN BERNARDINO      CALIFORNIA      92407
##      sales_platform_cd print_flg la_valid_tkt_event_flg fin_mkt_nm
## 1 www.concerts.livenation.com      T      N      Boston
## 2      NULL      T      N      Seattle
## 3      NULL      T      N      Arizona
## 4      NULL      T      N      Dallas
## 5      NULL      T      N      Seattle
## 6      www.livenation.com      T      N      Los Angeles
##      web_session_cookie_val gndr_cd age_yr income_amt edu_val
## 1      7dfa56dd7d5956b17587      <NA>      <NA>      <NA>      <NA>
## 2      4f9e6fc637eaf7b736c2      <NA>      <NA>      <NA>      <NA>
## 3      6c2545703bd527a7144d      <NA>      <NA>      <NA>      <NA>
## 4      527d6b1eaffc69ddd882      <NA>      <NA>      <NA>      <NA>
## 5      8bd62c394a35213bdf52      <NA>      <NA>      <NA>      <NA>
## 6      3b3a628f83135acd0676      <NA>      <NA>      <NA>      <NA>
##      edu_1st_indv_val edu_2nd_indv_val adults_in_hh_num married_ind
## 1      <NA>      <NA>      <NA>      <NA>
## 2      <NA>      <NA>      <NA>      <NA>
## 3      <NA>      <NA>      <NA>      <NA>
## 4      <NA>      <NA>      <NA>      <NA>
## 5      <NA>      <NA>      <NA>      <NA>
## 6      <NA>      <NA>      <NA>      <NA>
##      child_present_ind home_owner_ind occpn_val occpn_1st_val occpn_2nd_val
## 1      <NA>      <NA>      <NA>      <NA>      <NA>
## 2      <NA>      <NA>      <NA>      <NA>      <NA>
## 3      <NA>      <NA>      <NA>      <NA>      <NA>
## 4      <NA>      <NA>      <NA>      <NA>      <NA>
## 5      <NA>      <NA>      <NA>      <NA>      <NA>
## 6      <NA>      <NA>      <NA>      <NA>      <NA>
##      dist_to_ven
## 1      NA
## 2      59
## 3      NA
## 4      NA
## 5      NA
## 6      NA

```

```

# View column names of sales
names(sales)

```

```

## [1] "X"      "event_id"
## [3] "primary_act_id"      "secondary_act_id"
## [5] "purch_party_lkup_id"      "event_name"
## [7] "primary_act_name"      "secondary_act_name"
## [9] "major_cat_name"      "minor_cat_name"
## [11] "la_event_type_cat"      "event_disp_name"
## [13] "ticket_text"      "tickets_purchased_qty"
## [15] "trans_face_val_amt"      "delivery_type_cd"
## [17] "event_date_time"      "event_dt"
## [19] "presale_dt"      "onsale_dt"

```

```
## [21] "sales_ord_create_dttm" "sales_ord_tran_dt"
## [23] "print_dt"             "timezn_nm"
## [25] "venue_city"           "venue_state"
## [27] "venue_postal_cd_sgmt_1" "sales_platform_cd"
## [29] "print_flg"             "la_valid_tkt_event_flg"
## [31] "fin_mkt_nm"            "web_session_cookie_val"
## [33] "gndr_cd"               "age_yr"
## [35] "income_amt"            "edu_val"
## [37] "edu_1st_indv_val"      "edu_2nd_indv_val"
## [39] "adults_in_hh_num"      "married_ind"
## [41] "child_present_ind"     "home_owner_ind"
## [43] "occpn_val"             "occpn_1st_val"
## [45] "occpn_2nd_val"         "dist_to_ven"
```

```
#take a look with glimpse it's part of the dplyr package
library(dplyr)
```

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

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

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

```
glimpse(sales)
```

```
## Observations: 5,000
## Variables: 46
## $ X               <int> 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, ...
## $ event_id        <chr> "abcafladb99a935fc661", "6c56d7f08c95f2...
## $ primary_act_id   <chr> "43f0436b905bfa7c2eec", "1a3e9aec06177...
## $ secondary_act_id <chr> "b85143bf51323b72e53c", "f53529c5679ea6...
## $ purch_party_lkup_id <chr> "7dfa56dd7d5956b17587", "4f9e6fc637eaf7...
## $ event_name       <chr> "Xfinity Center Mansfield Premier Parki...
## $ primary_act_name  <chr> "XFINITY Center Mansfield Premier Parki...
## $ secondary_act_name <chr> "NULL", "Dave Matthews Band", "NULL", "...
## $ major_cat_name    <chr> "MISC", "MISC", "MISC", "MISC", "MISC",...
## $ minor_cat_name    <chr> "PARKING", "CAMPING", "PARKING", "PARKI...
## $ la_event_type_cat <chr> "PARKING", "INVALID", "PARKING", "PARKI...
## $ event_disp_name   <chr> "Xfinity Center Mansfield Premier Parki...
## $ ticket_text       <chr> "   THIS TICKET IS VALID           FOR PAR...
## $ tickets_purchased_qty <int> 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 2, 4, ...
## $ trans_face_val_amt <dbl> 45, 75, 5, 20, 20, 10, 30, 28, 20, 25, ...
## $ delivery_type_cd  <chr> "eTicket", "TicketFast", "TicketFast", ...
## $ event_date_time   <chr> "2015-09-12 23:30:00", "2009-09-05 01:0...
## $ event_dt          <chr> "2015-09-12", "2009-09-04", "2006-04-21...
## $ presale_dt        <chr> "NULL", "NULL", "NULL", "NULL", "2005-0...
## $ onsale_dt         <chr> "2015-05-15", "2009-03-13", "2006-02-25...
## $ sales_ord_create_dttm <chr> "2015-09-11 18:17:45", "2009-07-06 00:0...
## $ sales_ord_tran_dt  <chr> "2015-09-11", "2009-07-05", "2006-04-05...
## $ print_dt          <chr> "2015-09-12", "2009-09-01", "2006-04-05...
## $ timezn_nm         <chr> "EST", "PST", "MST", "CST", "PST", "PST..."
```

```
## $ venue_city          <chr> "MANSFIELD", "QUINCY", "PHOENIX", "DALL...
## $ venue_state         <chr> "MASSACHUSETTS", "WASHINGTON", "ARIZONA...
## $ venue_postal_cd_sgmt_1 <chr> "02048", "98848", "85003", "75210", "98...
## $ sales_platform_cd   <chr> "www.concerts.livenation.com", "NULL", ...
## $ print_flg          <chr> "T ", "T ", "T ", "T ", "T ", "T ", "T ...
## $ la_valid_tkt_event_flg <chr> "N ", "N ", "N ", "N ", "N ", "N ", "N ...
## $ fin_mkt_nm         <chr> "Boston", "Seattle", "Arizona", "Dallas...
## $ web_session_cookie_val <chr> "7dfa56dd7d5956b17587", "4f9e6fc637eaf7...
## $ gndr_cd            <chr> NA, NA, NA, NA, NA, NA, "M", NA, NA, NA...
## $ age_yr             <chr> NA, NA, NA, NA, NA, NA, "28", NA, NA, N...
## $ income_amt         <chr> NA, NA, NA, NA, NA, NA, "112500", NA, N...
## $ edu_val            <chr> NA, NA, NA, NA, NA, NA, "High School", ...
## $ edu_1st_indv_val    <chr> NA, NA, NA, NA, NA, NA, "High School", ...
## $ edu_2nd_indv_val    <chr> NA, NA, NA, NA, NA, NA, "NULL", NA, NA,...
## $ adults_in_hh_num    <chr> NA, NA, NA, NA, NA, NA, "4", NA, NA, NA...
## $ married_ind        <chr> NA, NA, NA, NA, NA, NA, "0", NA, NA, NA...
## $ child_present_ind   <chr> NA, NA, NA, NA, NA, NA, "1", NA, NA, NA...
## $ home_owner_ind      <chr> NA, NA, NA, NA, NA, NA, "0", NA, NA, NA...
## $ occpn_val          <chr> NA, NA, NA, NA, NA, NA, "NULL", NA, NA,...
## $ occpn_1st_val       <chr> NA, NA, NA, NA, NA, NA, "Craftsman Blue...
## $ occpn_2nd_val       <chr> NA, NA, NA, NA, NA, NA, "NULL", NA, NA,...
## $ dist_to_ven         <int> NA, 59, NA, NA, NA, NA, NA, NA, NA, NA,...
```

**Note:** Notice the first column, X, which appears to just be counting.

Next will be removing first column

*#Take a subset of sales to omit the first column. Assign the result to sales2*

```
sales2<- sales[,-1]
head(sales2,n=3)
```

```
##          event_id      primary_act_id      secondary_act_id
## 1 abcaf1adb99a935fc661 43f0436b905bfa7c2eec b85143bf51323b72e53c
## 2 6c56d7f08c95f2aa453c 1a3e9aec0617706a794 f53529c5679ea6ca5a48
## 3 c7ab4524a121f9d687d2 4b677c3f5bec71eec8d1 b85143bf51323b72e53c
##   purch_party_lkup_id
## 1 7dfa56dd7d5956b17587
## 2 4f9e6fc637eaf7b736c2
## 3 6c2545703bd527a7144d
##
##                                event_name
## 1 Xfinity Center Mansfield Premier Parking: Florida Georgia Line
## 2                                Gorge Camping - dave matthews band - sept 3-7
## 3                                Dodge Theatre Adams Street Parking - benise
##   primary_act_name secondary_act_name
## 1 XFINITY Center Mansfield Premier Parking      NULL
## 2                                Gorge Camping Dave Matthews Band
## 3                                Parking Event      NULL
##   major_cat_name minor_cat_name la_event_type_cat
## 1          MISC          PARKING          PARKING
## 2          MISC          CAMPING          INVALID
## 3          MISC          PARKING          PARKING
##
##                                event_disp_name
## 1 Xfinity Center Mansfield Premier Parking: Florida Georgia Line
## 2                                Gorge Camping - dave matthews band - sept 3-7
## 3                                Dodge Theatre Adams Street Parking - benise
```

```
##
## 1   THIS TICKET IS VALID           FOR PARKING ONLY           GOOD THIS DAY ONLY           PREMIER PARKING P
## 2                                     %OVERNIGHT C A M P I N G%* * * * *
## 3                                     ADAMS STREET GARAGE%PARKING FOR 4/21/06 ONLY%DODGE THEATRE PARKING P
##   tickets_purchased_qty trans_face_val_amt delivery_type_cd
## 1           1           45           eTicket
## 2           1           75           TicketFast
## 3           1           5           TicketFast
##   event_date_time   event_dt presale_dt   onsale_dt
## 1 2015-09-12 23:30:00 2015-09-12      NULL 2015-05-15
## 2 2009-09-05 01:00:00 2009-09-04      NULL 2009-03-13
## 3 2006-04-22 01:30:00 2006-04-21      NULL 2006-02-25
##   sales_ord_create_dttm sales_ord_tran_dt   print_dt timezn_nm venue_city
## 1   2015-09-11 18:17:45      2015-09-11 2015-09-12      EST  MANSFIELD
## 2   2009-07-06 00:00:00      2009-07-05 2009-09-01      PST   QUINCY
## 3   2006-04-05 00:00:00      2006-04-05 2006-04-05      MST   PHOENIX
##   venue_state venue_postal_cd_sgmt_1      sales_platform_cd
## 1 MASSACHUSETTS      02048 www.concerts.livenation.com
## 2   WASHINGTON      98848      NULL
## 3    ARIZONA      85003      NULL
##   print_flg la_valid_tkt_event_flg fin_mkt_nm web_session_cookie_val
## 1      T      N      Boston 7dfa56dd7d5956b17587
## 2      T      N      Seattle 4f9e6fc637eaf7b736c2
## 3      T      N      Arizona 6c2545703bd527a7144d
##   gndr_cd age_yr income_amt edu_val edu_1st_indv_val edu_2nd_indv_val
## 1   <NA> <NA>      <NA> <NA>      <NA>      <NA>
## 2   <NA> <NA>      <NA> <NA>      <NA>      <NA>
## 3   <NA> <NA>      <NA> <NA>      <NA>      <NA>
##   adults_in_hh_num married_ind child_present_ind home_owner_ind occpn_val
## 1      <NA>      <NA>      <NA>      <NA>      <NA>
## 2      <NA>      <NA>      <NA>      <NA>      <NA>
## 3      <NA>      <NA>      <NA>      <NA>      <NA>
##   occpn_1st_val occpn_2nd_val dist_to_ven
## 1      <NA>      <NA>      NA
## 2      <NA>      <NA>      59
## 3      <NA>      <NA>      NA
```

Many of the columns have information that's of no use to us. For example, the first four columns contain internal codes representing particular events. The last fifteen columns also aren't worth keeping; there are too many missing values to make them worthwhile.

An easy way to get rid of unnecessary columns is to create a vector containing the column indices you want to keep, then subset the data based on that vector using single bracket subsetting.

```
# Define a vector of column indices: keep
```

```
keep<-(5:(ncol(sales2)-15))
```

```
# Subset sales2 using keep: sales3
```

```
sales3<- sales2[keep]
```

We have a sales3 with 26 variables.

## Separating columns

Some of the columns in your data frame include multiple pieces of information that should be in separate columns. In this exercise, you will separate such a column into two: one for date and one for time. You will use the `separate()` function from the `tidyr` package (already installed for you).

Take a look at the `event_date_time` column by typing `head(sales3$event_date_time)` in the console. You'll notice that the date and time are separated by a space. Therefore, you'll use `sep = " "` as an argument to `separate()`.

```
# Load tidyr
library(tidyr)

# Split event_date_time in "event_dt", "event_time": sales4
sales4 <- separate(sales3, event_date_time,
                   c("event_dt", "event_time"), sep = " ")

# Split sales_ord_create_dttm in "ord_create_dt", "ord_create_time": sales5
sales5 <- separate(sales4, sales_ord_create_dttm, c("ord_create_dt", "ord_create_time"), sep = " ")

## Warning: Expected 2 pieces. Missing pieces filled with `NA` in 4 rows
## [2516, 3863, 4082, 4183].

;;Warning message!!
```

## Dealing with warnings

Looks like that second call to `separate()` threw a warning. Not to worry; warnings aren't as bad as error messages. It's not saying that the command didn't execute; it's just a heads-up that something unusual happened.

The warning says Too few values at 4 locations. You may be able to guess already what the issue is, but it's still good to take a look.

The locations (i.e. rows) given in the warning are 2516, 3863, 4082, and 4183. Have a look at the contents of the `sales_ord_create_dttm` column in those rows.

```
# Lets look at the warnings, we see 4 NA in $ord_create_time
# Define an issues vector
issues <- c(2516, 3863, 4082, 4183)

# Print values of sales_ord_create_dttm at these indices
sales3$sales_ord_create_dttm[issues]

## [1] "NULL" "NULL" "NULL" "NULL"

# Print a well-behaved value of sales_ord_create_dttm
sales3$sales_ord_create_dttm[2517]

## [1] "2013-08-04 23:07:19"
```

The warning was just because of four missing values. You'll ignore them for now, but if your analysis depended on complete date/time information, you would probably need to delete those rows.

## Identifying dates

Some of the columns in your dataset contain dates of different events. Right now, they are stored as character strings. That's fine if all you want to do is look up the date associated with an event, but if you want to do

any comparisons or math with the dates, it's MUCH easier to store them as Date objects.

Luckily, all of the date columns in this dataset have the substring “dt” in their name, so you can use the `str_detect()` function of the **stringr** package to find the date columns. Then you can coerce them to Date objects using a function from the **lubridate** package.

You'll use `lapply()` to apply the appropriate lubridate function to all of the columns that contain dates. Recall the following syntax for `lapply()` applied to some data frame columns of interest:

```
lapply(my_data_frame[, cols], function_name)
```

Also recall that function names in lubridate combine the letters y, m, d, h, m, and s depending on the format of the date/time string being read in.

```
# Load stringr
library(stringr)

# Find columns of sales5 containing "dt": date_cols
date_cols<-str_detect(names(sales5), pattern="dt")

# Load lubridate
library(lubridate)

##
## Attaching package: 'lubridate'
## The following object is masked from 'package:base':
##
##      date

# Coerce date columns into Date objects
sales5[, date_cols] <- lapply(sales5[, date_cols], ymd)
```

```
## Warning: 2892 failed to parse.
```

```
## Warning: 101 failed to parse.
```

```
## Warning: 4 failed to parse.
```

```
## Warning: 424 failed to parse.
```

**More warnings!** As you saw, some of the calls to `ymd()` caused a failure to parse warning. That's probably because of more missing data, but again, it's good to check to be sure.

The first two lines of code (provided for you here) create a list of logical vectors called `missing`. Each vector in the list indicates the presence (or absence) of missing values in the corresponding column of `sales5`. See if the number of missing values in each column is the same as the number of rows that failed to parse in the previous exercise.

```
# Find date columns (don't change)
date_cols <- str_detect(names(sales5), "dt")

# Create logical vectors indicating missing values (don't change)
missing<- lapply(sales5[,date_cols],is.na)

# Create a numerical vector that counts missing values: num_missing
num_missing<- sapply(missing,sum)

# Print num_missing
print(num_missing)
```



```
##          event_dt      presale_dt      onsale_dt      ord_create_dt
##              0          2892          101          4
## sales_ord_tran_dt      print_dt
##              0          424
```

## Combining columns

Sure enough, the number of NAs in each column match the numbers from the warning messages, so missing data is the culprit. How to proceed depends on your desired analysis. If you really need complete sets of date/time information, you might delete the rows or columns containing NAs.

As your last step, you'll use the tidyr function `unite()` to combine the `venue_city` and `venue_state` columns into one column with the two values separated by a comma and a space. For example, "PORTLAND" "MAINE" should become "PORTLAND, MAINE".

```
# Combine the venue_city and venue_state columns
sales6<- unite(sales5, "venue_city_state", "venue_city", "venue_state", sep=", ")
```

```
# View the head of sales6
glimpse(sales6$venue_city_state)
```

```
## chr [1:5000] "MANSFIELD, MASSACHUSETTS" "QUINCY, WASHINGTON" ...
```

## Exercise 2\_MBTransportation

Data\_Sales: <https://www.datacamp.com/courses/importing-cleaning-data-in-r-case-studies>

### Importing

The dataset is stored as an Excel spreadsheet called `mbta.xlsx` in your working directory. You'll use the `read_excel()` function from Hadley Wickham's `readxl` package to import it.

The first time you import a dataset, you might not know how many rows need to be skipped. In this case, the first row is a title, so you'll need to skip the first row.

```
library(readxl)

# Import mbta.xlsx and skip first row: mbta
mbta<-read_excel("mbta.xlsx", skip = 1)
```

### Examining the Data

```
# View the structure of mbta
str(mbta)

## Classes 'tbl_df', 'tbl' and 'data.frame': 11 obs. of 60 variables:
## $ X__1 : num 1 2 3 4 5 6 7 8 9 10 ...
## $ mode : chr "All Modes by Qtr" "Boat" "Bus" "Commuter Rail" ...
## $ 2007-01: chr "NA" "4" "335.819" "142.2" ...
## $ 2007-02: chr "NA" "3.6" "338.675" "138.5" ...
## $ 2007-03: num 1188 40 340 138 459 ...
## $ 2007-04: chr "NA" "4.3" "352.162" "139.5" ...
```

```

## $ 2007-05: chr "NA" "4.9" "354.367" "139" ...
## $ 2007-06: num 1246 5.8 350.5 143 477 ...
## $ 2007-07: chr "NA" "6.521" "357.519" "142.391" ...
## $ 2007-08: chr "NA" "6.572" "355.479" "142.364" ...
## $ 2007-09: num 1256.57 5.47 372.6 143.05 499.57 ...
## $ 2007-10: chr "NA" "5.145" "368.847" "146.542" ...
## $ 2007-11: chr "NA" "3.763" "330.826" "145.089" ...
## $ 2007-12: num 1216.89 2.98 312.92 141.59 448.27 ...
## $ 2008-01: chr "NA" "3.175" "340.324" "142.145" ...
## $ 2008-02: chr "NA" "3.111" "352.905" "142.607" ...
## $ 2008-03: num 1253.52 3.51 361.15 137.45 494.05 ...
## $ 2008-04: chr "NA" "4.164" "368.189" "140.389" ...
## $ 2008-05: chr "NA" "4.015" "363.903" "142.585" ...
## $ 2008-06: num 1314.82 5.19 362.96 142.06 518.35 ...
## $ 2008-07: chr "NA" "6.016" "370.921" "145.731" ...
## $ 2008-08: chr "NA" "5.8" "361.057" "144.565" ...
## $ 2008-09: num 1307.04 4.59 389.54 141.91 517.32 ...
## $ 2008-10: chr "NA" "4.285" "357.974" "151.957" ...
## $ 2008-11: chr "NA" "3.488" "345.423" "152.952" ...
## $ 2008-12: num 1232.65 3.01 325.77 140.81 446.74 ...
## $ 2009-01: chr "NA" "3.014" "338.532" "141.448" ...
## $ 2009-02: chr "NA" "3.196" "360.412" "143.529" ...
## $ 2009-03: num 1209.79 3.33 353.69 142.89 467.22 ...
## $ 2009-04: chr "NA" "4.049" "359.38" "142.34" ...
## $ 2009-05: chr "NA" "4.119" "354.75" "144.225" ...
## $ 2009-06: num 1233.1 4.9 347.9 142 473.1 ...
## $ 2009-07: chr "NA" "6.444" "339.477" "137.691" ...
## $ 2009-08: chr "NA" "5.903" "332.661" "139.158" ...
## $ 2009-09: num 1230.5 4.7 374.3 139.1 500.4 ...
## $ 2009-10: chr "NA" "4.212" "385.868" "137.104" ...
## $ 2009-11: chr "NA" "3.576" "366.98" "129.343" ...
## $ 2009-12: num 1207.85 3.11 332.39 126.07 440.93 ...
## $ 2010-01: chr "NA" "3.207" "362.226" "130.91" ...
## $ 2010-02: chr "NA" "3.195" "361.138" "131.918" ...
## $ 2010-03: num 1208.86 3.48 373.44 131.25 483.4 ...
## $ 2010-04: chr "NA" "4.452" "378.611" "131.722" ...
## $ 2010-05: chr "NA" "4.415" "380.171" "128.8" ...
## $ 2010-06: num 1244.41 5.41 363.27 129.14 490.26 ...
## $ 2010-07: chr "NA" "6.513" "353.04" "122.935" ...
## $ 2010-08: chr "NA" "6.269" "343.688" "129.732" ...
## $ 2010-09: num 1225.5 4.7 381.6 132.9 521.1 ...
## $ 2010-10: chr "NA" "4.402" "384.987" "131.033" ...
## $ 2010-11: chr "NA" "3.731" "367.955" "130.889" ...
## $ 2010-12: num 1216.26 3.16 326.34 121.42 450.43 ...
## $ 2011-01: chr "NA" "3.14" "334.958" "128.396" ...
## $ 2011-02: chr "NA" "3.284" "346.234" "125.463" ...
## $ 2011-03: num 1223.45 3.67 380.4 134.37 516.73 ...
## $ 2011-04: chr "NA" "4.251" "380.446" "134.169" ...
## $ 2011-05: chr "NA" "4.431" "385.289" "136.14" ...
## $ 2011-06: num 1302.41 5.47 376.32 135.58 529.53 ...
## $ 2011-07: chr "NA" "6.581" "361.585" "132.41" ...
## $ 2011-08: chr "NA" "6.733" "353.793" "130.616" ...
## $ 2011-09: num 1291 5 388 137 550 ...
## $ 2011-10: chr "NA" "4.484" "398.456" "128.72" ...

```

```
# View the first 6 rows of mbta
head(mbta, n=6)
```

```
## # A tibble: 6 x 60
##   X__1 mode `2007-01` `2007-02` `2007-03` `2007-04` `2007-05` `2007-06`
##   <dbl> <chr> <chr>      <chr>      <dbl> <chr>      <chr>      <dbl>
## 1     1 All M~ NA        NA        1188. NA        NA        1246.
## 2     2 Boat  4        3.6        40  4.3        4.9        5.8
## 3     3 Bus   335.819  338.675    340. 352.162    354.367    351.
## 4     4 Commu~ 142.2    138.5     138. 139.5     139        143
## 5     5 Heavy~ 435.294  448.271    459. 472.201    474.579    477.
## 6     6 Light~ 227.231  240.262    241. 255.557    248.262    246.
## # ... with 52 more variables: `2007-07` <chr>, `2007-08` <chr>,
## # `2007-09` <dbl>, `2007-10` <chr>, `2007-11` <chr>, `2007-12` <dbl>,
## # `2008-01` <chr>, `2008-02` <chr>, `2008-03` <dbl>, `2008-04` <chr>,
## # `2008-05` <chr>, `2008-06` <dbl>, `2008-07` <chr>, `2008-08` <chr>,
## # `2008-09` <dbl>, `2008-10` <chr>, `2008-11` <chr>, `2008-12` <dbl>,
## # `2009-01` <chr>, `2009-02` <chr>, `2009-03` <dbl>, `2009-04` <chr>,
## # `2009-05` <chr>, `2009-06` <dbl>, `2009-07` <chr>, `2009-08` <chr>,
## # `2009-09` <dbl>, `2009-10` <chr>, `2009-11` <chr>, `2009-12` <dbl>,
## # `2010-01` <chr>, `2010-02` <chr>, `2010-03` <dbl>, `2010-04` <chr>,
## # `2010-05` <chr>, `2010-06` <dbl>, `2010-07` <chr>, `2010-08` <chr>,
## # `2010-09` <dbl>, `2010-10` <chr>, `2010-11` <chr>, `2010-12` <dbl>,
## # `2011-01` <chr>, `2011-02` <chr>, `2011-03` <dbl>, `2011-04` <chr>,
## # `2011-05` <chr>, `2011-06` <dbl>, `2011-07` <chr>, `2011-08` <chr>,
## # `2011-09` <dbl>, `2011-10` <chr>
```

```
# View a summary of mbta
summary(mbta)
```

```
##      X__1      mode      2007-01      2007-02
## Min.   : 1.0    Length:11    Length:11    Length:11
## 1st Qu.: 3.5    Class :character Class :character Class :character
## Median : 6.0    Mode  :character Mode  :character Mode  :character
## Mean   : 6.0
## 3rd Qu.: 8.5
## Max.   :11.0
##      2007-03      2007-04      2007-05
## Min.   : 0.114   Length:11    Length:11
## 1st Qu.: 9.278   Class :character Class :character
## Median : 137.700 Mode  :character Mode  :character
## Mean   : 330.293
## 3rd Qu.: 399.225
## Max.   :1204.725
##      2007-06      2007-07      2007-08
## Min.   : 0.096   Length:11    Length:11
## 1st Qu.: 5.700   Class :character Class :character
## Median : 143.000 Mode  :character Mode  :character
## Mean   : 339.846
## 3rd Qu.: 413.788
## Max.   :1246.129
##      2007-09      2007-10      2007-11
## Min.   : -0.007   Length:11    Length:11
## 1st Qu.: 5.539   Class :character Class :character
## Median : 143.051 Mode  :character Mode  :character
```

```

## Mean      : 352.554
## 3rd Qu.: 436.082
## Max.      :1310.764
##      2007-12      2008-01      2008-02
## Min.      : -0.060 Length:11      Length:11
## 1st Qu.:  4.385 Class :character Class :character
## Median : 141.585 Mode  :character Mode  :character
## Mean      : 321.588
## 3rd Qu.: 380.594
## Max.      :1216.890
##      2008-03      2008-04      2008-05
## Min.      :  0.058 Length:11      Length:11
## 1st Qu.:  5.170 Class :character Class :character
## Median : 137.453 Mode  :character Mode  :character
## Mean      : 345.604
## 3rd Qu.: 427.601
## Max.      :1274.031
##      2008-06      2008-07      2008-08
## Min.      :  0.060 Length:11      Length:11
## 1st Qu.:  5.742 Class :character Class :character
## Median : 142.057 Mode  :character Mode  :character
## Mean      : 359.667
## 3rd Qu.: 440.656
## Max.      :1320.728
##      2008-09      2008-10      2008-11
## Min.      :  0.021 Length:11      Length:11
## 1st Qu.:  5.691 Class :character Class :character
## Median : 141.907 Mode  :character Mode  :character
## Mean      : 362.099
## 3rd Qu.: 453.430
## Max.      :1338.015
##      2008-12      2009-01      2009-02
## Min.      : -0.015 Length:11      Length:11
## 1st Qu.:  4.689 Class :character Class :character
## Median : 140.810 Mode  :character Mode  :character
## Mean      : 319.882
## 3rd Qu.: 386.255
## Max.      :1232.655
##      2009-03      2009-04      2009-05
## Min.      : -0.050 Length:11      Length:11
## 1st Qu.:  5.003 Class :character Class :character
## Median : 142.893 Mode  :character Mode  :character
## Mean      : 330.142
## 3rd Qu.: 410.455
## Max.      :1210.912
##      2009-06      2009-07      2009-08
## Min.      : -0.079 Length:11      Length:11
## 1st Qu.:  5.845 Class :character Class :character
## Median : 142.006 Mode  :character Mode  :character
## Mean      : 333.194
## 3rd Qu.: 410.482
## Max.      :1233.085
##      2009-09      2009-10      2009-11
## Min.      : -0.035 Length:11      Length:11

```

## 1st Qu.: 5.693	Class :character	Class :character
## Median : 139.087	Mode :character	Mode :character
## Mean : 346.687		
## 3rd Qu.: 437.332		
## Max. :1291.564		
## 2009-12	2010-01	2010-02
## Min. : -0.022	Length:11	Length:11
## 1st Qu.: 4.784	Class :character	Class :character
## Median : 126.066	Mode :character	Mode :character
## Mean : 312.962		
## 3rd Qu.: 386.659		
## Max. :1207.845		
## 2010-03	2010-04	2010-05
## Min. : 0.012	Length:11	Length:11
## 1st Qu.: 5.274	Class :character	Class :character
## Median : 131.252	Mode :character	Mode :character
## Mean : 332.726		
## 3rd Qu.: 428.420		
## Max. :1225.556		
## 2010-06	2010-07	2010-08
## Min. : 0.008	Length:11	Length:11
## 1st Qu.: 6.436	Class :character	Class :character
## Median : 129.144	Mode :character	Mode :character
## Mean : 335.964		
## 3rd Qu.: 426.769		
## Max. :1244.409		
## 2010-09	2010-10	2010-11
## Min. : 0.001	Length:11	Length:11
## 1st Qu.: 5.567	Class :character	Class :character
## Median : 132.892	Mode :character	Mode :character
## Mean : 346.524		
## 3rd Qu.: 451.361		
## Max. :1293.117		
## 2010-12	2011-01	2011-02
## Min. : -0.004	Length:11	Length:11
## 1st Qu.: 4.466	Class :character	Class :character
## Median : 121.422	Mode :character	Mode :character
## Mean : 312.917		
## 3rd Qu.: 388.385		
## Max. :1216.262		
## 2011-03	2011-04	2011-05
## Min. : 0.05	Length:11	Length:11
## 1st Qu.: 6.03	Class :character	Class :character
## Median : 134.37	Mode :character	Mode :character
## Mean : 345.17		
## 3rd Qu.: 448.56		
## Max. :1286.66		
## 2011-06	2011-07	2011-08
## Min. : 0.054	Length:11	Length:11
## 1st Qu.: 6.926	Class :character	Class :character
## Median : 135.581	Mode :character	Mode :character
## Mean : 353.331		
## 3rd Qu.: 452.923		
## Max. :1302.414		

```
##      2011-09      2011-10
## Min.   : 0.043   Length:11
## 1st Qu.: 6.660   Class :character
## Median :136.901   Mode  :character
## Mean   : 362.555
## 3rd Qu.: 469.204
## Max.   :1348.754
```

## Removing unnecessary rows and columns

```
# Remove the first, seventh, and eleventh rows of mbta (All Modes By Qtr, Pct Chg / Yr, and TOTAL). Name
mbta2<- mbta[-(c(1,7,11)),]
```

```
# Remove the first column of mbta2. Name the resulting data frame mbta3
```

```
mbta3<-mbta2[,-1]
head(mbta3)
```

```
## # A tibble: 6 x 59
##   mode      `2007-01` `2007-02` `2007-03` `2007-04` `2007-05` `2007-06`
##   <chr>      <chr>      <chr>      <dbl> <chr>      <chr>      <dbl>
## 1 Boat         4         3.6         40    4.3        4.9         5.8
## 2 Bus        335.819    338.675     340.   352.162    354.367     351.
## 3 Commuter Ra~ 142.2      138.5      138.    139.5     139         143
## 4 Heavy Rail   435.294    448.271     459.   472.201    474.579     477.
## 5 Light Rail   227.231    240.262     241.   255.557    248.262     246.
## 6 Private Bus   4.772      4.417       4.57    4.542     4.768        4.72
## # ... with 52 more variables: `2007-07` <chr>, `2007-08` <chr>,
## #   `2007-09` <dbl>, `2007-10` <chr>, `2007-11` <chr>, `2007-12` <dbl>,
## #   `2008-01` <chr>, `2008-02` <chr>, `2008-03` <dbl>, `2008-04` <chr>,
## #   `2008-05` <chr>, `2008-06` <dbl>, `2008-07` <chr>, `2008-08` <chr>,
## #   `2008-09` <dbl>, `2008-10` <chr>, `2008-11` <chr>, `2008-12` <dbl>,
## #   `2009-01` <chr>, `2009-02` <chr>, `2009-03` <dbl>, `2009-04` <chr>,
## #   `2009-05` <chr>, `2009-06` <dbl>, `2009-07` <chr>, `2009-08` <chr>,
## #   `2009-09` <dbl>, `2009-10` <chr>, `2009-11` <chr>, `2009-12` <dbl>,
## #   `2010-01` <chr>, `2010-02` <chr>, `2010-03` <dbl>, `2010-04` <chr>,
## #   `2010-05` <chr>, `2010-06` <dbl>, `2010-07` <chr>, `2010-08` <chr>,
## #   `2010-09` <dbl>, `2010-10` <chr>, `2010-11` <chr>, `2010-12` <dbl>,
## #   `2011-01` <chr>, `2011-02` <chr>, `2011-03` <dbl>, `2011-04` <chr>,
## #   `2011-05` <chr>, `2011-06` <dbl>, `2011-07` <chr>, `2011-08` <chr>,
## #   `2011-09` <dbl>, `2011-10` <chr>
```

## Observations are stored in columns

As is customary, you want to represent variables in columns rather than rows. The first step is to use the `gather()` function from the **tidyr** package, which will gather columns into key-value pairs.

```
library(tidyr)
```

```
# Gather columns of mbta3: mbta4
mbta4<- gather(data = mbta3, key = "month", value = "thou_riders", -mode)
```

```
# View the head of mbta4
head(mbta4)
```

```
## # A tibble: 6 x 3
##   mode      month thou_riders
##   <chr>    <chr>    <chr>
## 1 Boat      2007-01  4
## 2 Bus        2007-01 335.819
## 3 Commuter Rail 2007-01 142.2
## 4 Heavy Rail   2007-01 435.294
## 5 Light Rail   2007-01 227.231
## 6 Private Bus  2007-01  4.772
```

## Type conversions

But first, take this opportunity to change the average weekday ridership column, `thou_riders`, into numeric values rather than character strings. That way, you'll be able to do things like compare values and do math.

```
# Coerce thou_riders to numeric
```

```
mbta4$thou_riders<-as.numeric(mbta4$thou_riders)
head(mbta4,n=5)
```

```
## # A tibble: 5 x 3
##   mode      month thou_riders
##   <chr>    <chr>    <dbl>
## 1 Boat      2007-01      4
## 2 Bus        2007-01    336.
## 3 Commuter Rail 2007-01    142.
## 4 Heavy Rail   2007-01    435.
## 5 Light Rail   2007-01    227.
```

## Variables are stored in both rows and columns

Now, you can finish the job you started earlier: getting variables into columns. Right now, variables are stored as “keys” in the mode column. You'll use the tidy function `spread()` to make them into columns containing average weekday ridership for the given month and mode of transport.

```
# Spread the contents of mbta4: mbta5
mbta5<- spread(mbta4,mode,thou_riders)
```

```
# View the head of mbta5
head(mbta5)
```

```
## # A tibble: 6 x 9
##   month   Boat   Bus `Commuter Rail` `Heavy Rail` `Light Rail`
##   <chr> <dbl> <dbl>         <dbl>         <dbl>         <dbl>
## 1 2007-01     4   336.         142.         435.         227.
## 2 2007-02    3.6  339.         138.         448.         240.
## 3 2007-03   40   340.         138.         459.         241.
## 4 2007-04   4.3  352.         140.         472.         256.
## 5 2007-05   4.9  354.         139.         475.         248.
## 6 2007-06   5.8  351.         143.         477.         246.
```

```
## # ... with 3 more variables: `Private Bus` <dbl>, RIDE <dbl>, `Trackless
## #   Trolley` <dbl>
```

## Separating columns

In this exercise, you'll separate the month column into distinct month and year columns to make life easier.

```
# Split month column into month and year: mbta6
mbta6<- separate(mbta5, month, c("year","month"), sep="-")

# View the head of mbta6
head(mbta6,n=3)
```

```
## # A tibble: 3 x 10
##   year month Boat   Bus `Commuter Rail` `Heavy Rail` `Light Rail`
##   <chr> <chr> <dbl> <dbl>         <dbl>         <dbl>         <dbl>
## 1 2007  01     4   336.         142.         435.         227.
## 2 2007  02    3.6  339.         138.         448.         240.
## 3 2007  03    40   340.         138.         459.         241.
## # ... with 3 more variables: `Private Bus` <dbl>, RIDE <dbl>, `Trackless
## #   Trolley` <dbl>
```

## Outliers

Let's take a look of summary to see if there is any possible outlier. It may be sthng weird in Boat.

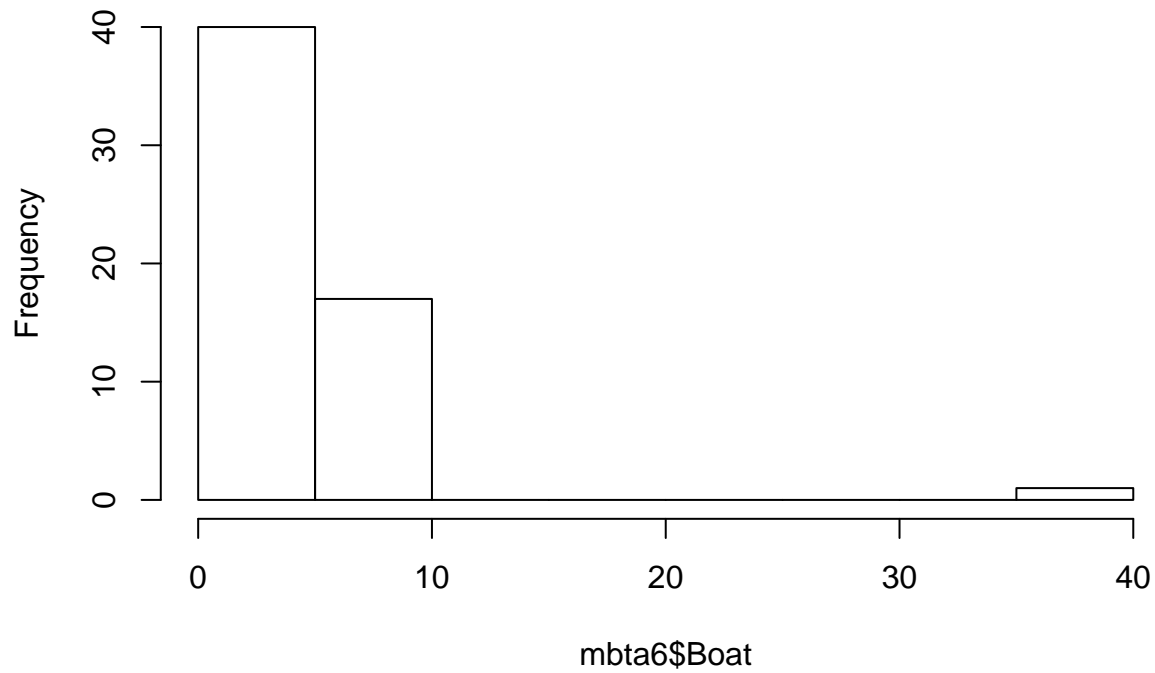
```
# View a summary of mbta6
summary(mbta6)
```

```
##      year      month      Boat      Bus
## Length:58      Length:58      Min.   : 2.985      Min.   :312.9
## Class :character Class :character 1st Qu.: 3.494      1st Qu.:345.6
## Mode  :character Mode  :character Median  : 4.293      Median :359.9
##                                     Mean   : 5.068      Mean   :358.6
##                                     3rd Qu.: 5.356      3rd Qu.:372.2
##                                     Max.    :40.000      Max.    :398.5
## Commuter Rail      Heavy Rail      Light Rail      Private Bus
## Min.   :121.4      Min.   :435.3      Min.   :194.4      Min.   :2.213
## 1st Qu.:131.4      1st Qu.:471.1      1st Qu.:220.6      1st Qu.:2.641
## Median :138.8      Median :487.3      Median :231.9      Median :2.820
## Mean   :137.4      Mean   :489.3      Mean   :233.0      Mean   :3.352
## 3rd Qu.:142.4      3rd Qu.:511.3      3rd Qu.:244.5      3rd Qu.:4.167
## Max.    :153.0      Max.    :554.9      Max.    :271.1      Max.    :4.878
##      RIDE      Trackless Trolley
## Min.   :4.900      Min.   : 5.777
## 1st Qu.:5.965      1st Qu.:11.679
## Median :6.615      Median :12.598
## Mean   :6.604      Mean   :12.125
## 3rd Qu.:7.149      3rd Qu.:13.320
## Max.    :8.598      Max.    :15.109
```

```
# Generate a histogram of Boat column
hist(mbta6$Boat)
```



## Histogram of mbta6\$Boat



Replace this 40 by a 4 to correct the outlier.

```
# Find the row number of the incorrect value: i
```

```
i<-which(mbta6$Boat==40)
```

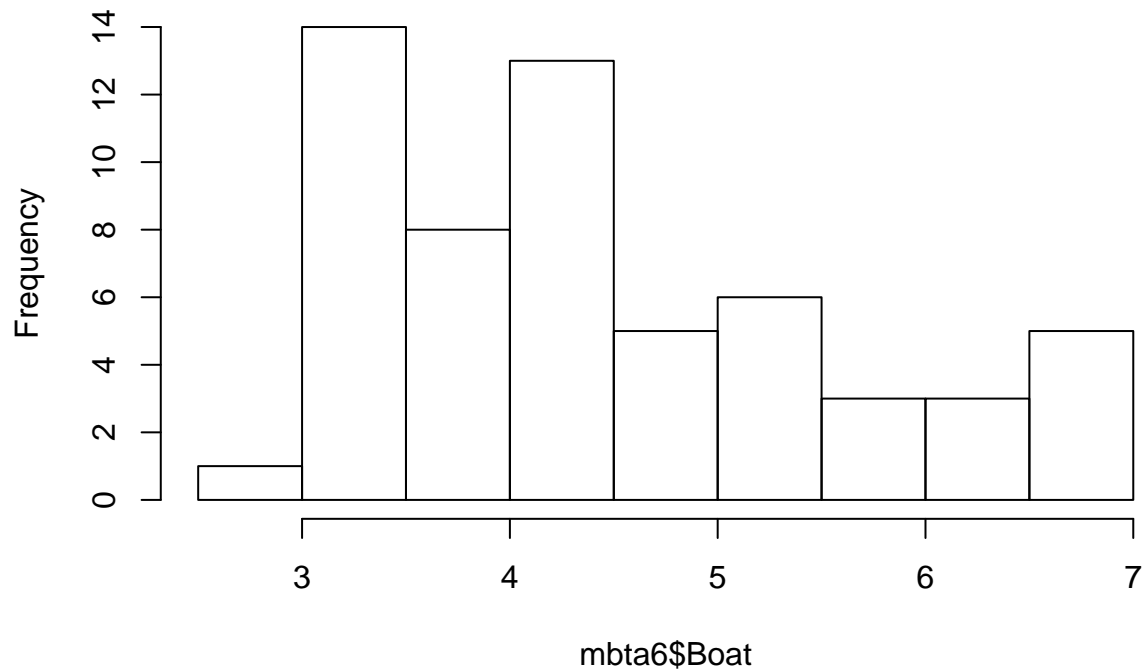
```
# Replace the incorrect value with 4
```

```
mbta6$Boat[i]<-4
```

```
# Generate a histogram of Boat column
```

```
hist(mbta6$Boat)
```

## Histogram of mbta6\$Boat



### Exercise 3\_Nutrition\_Food

Data\_Sales: <https://www.datacamp.com/courses/importing-cleaning-data-in-r-case-studies> Data\_file: food.csv

#### Importing

A large dataset called food.csv is ready for your use in the working directory. Instead of the usual `read.csv()`, however, you're going to use the faster **fread()** from the **data.table** package. The data will come in as a data table, but since you're used to working with data frames, you can just convert it.

```
library(data.table)

##
## Attaching package: 'data.table'

## The following objects are masked from 'package:lubridate':
##
##   hour, isoweek, mday, minute, month, quarter, second, wday,
##   week, yday, year

## The following objects are masked from 'package:dplyr':
##
##   between, first, last

# Import food.csv: dt_food
dt_food<- fread("food.csv")

# Convert dt_food to a data frame
df_food<- data.frame(dt_food)
```

## Examining

```
summary(df_food)
```

```
##          V1              code          url          creator
##  Min.   :   1.0    Min.   :100030  Length:1500      Length:1500
##  1st Qu.: 375.8    1st Qu.:124974  Class :character  Class :character
##  Median : 750.5    Median :149514  Mode  :character  Mode  :character
##  Mean   : 750.5    Mean   :149613
##  3rd Qu.:1125.2    3rd Qu.:174506
##  Max.   :1500.0    Max.   :199880
##
##  created_t      created_datetime  last_modified_t
##  Min.   :1.332e+09  Length:1500    Min.   :1.340e+09
##  1st Qu.:1.394e+09  Class :character  1st Qu.:1.424e+09
##  Median :1.425e+09  Mode  :character  Median :1.437e+09
##  Mean   :1.414e+09          Mean   :1.430e+09
##  3rd Qu.:1.436e+09          3rd Qu.:1.446e+09
##  Max.   :1.453e+09          Max.   :1.453e+09
##
##  last_modified_datetime  product_name      generic_name
##  Length:1500             Length:1500          Length:1500
##  Class :character        Class :character  Class :character
##  Mode  :character        Mode  :character  Mode  :character
##
##
##
##  quantity      packaging      packaging_tags
##  Length:1500    Length:1500    Length:1500
##  Class :character  Class :character  Class :character
##  Mode  :character  Mode  :character  Mode  :character
##
##
##
##  brands      brands_tags      categories
##  Length:1500  Length:1500    Length:1500
##  Class :character  Class :character  Class :character
##  Mode  :character  Mode  :character  Mode  :character
##
##
##
##  categories_tags  categories_en      origins
##  Length:1500      Length:1500    Length:1500
##  Class :character  Class :character  Class :character
##  Mode  :character  Mode  :character  Mode  :character
##
##
##
##  origins_tags      manufacturing_places  manufacturing_places_tags
##  Length:1500      Length:1500          Length:1500
```

```

## Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character
##
##
##
##      labels          labels_tags      labels_en
## Length:1500         Length:1500       Length:1500
## Class :character    Class :character   Class :character
## Mode  :character    Mode  :character   Mode  :character
##
##
##
##      emb_codes        emb_codes_tags   first_packaging_code_geo
## Length:1500          Length:1500       Length:1500
## Class :character     Class :character   Class :character
## Mode  :character     Mode  :character   Mode  :character
##
##
##
##      cities          cities_tags       purchase_places      stores
## Mode:logical         Length:1500        Length:1500           Length:1500
## NA's:1500            Class :character   Class :character      Class :character
##                      Mode  :character   Mode  :character      Mode  :character
##
##
##
##      countries        countries_tags    countries_en
## Length:1500           Length:1500       Length:1500
## Class :character      Class :character   Class :character
## Mode  :character      Mode  :character   Mode  :character
##
##
##
##      ingredients_text  allergens        allergens_en      traces
## Length:1500           Length:1500       Mode:logical      Length:1500
## Class :character      Class :character   NA's:1500          Class :character
## Mode  :character      Mode  :character   Mode  :character   Mode  :character
##
##
##
##      traces_tags      traces_en          serving_size      no_nutriments
## Length:1500           Length:1500        Length:1500        Mode:logical
## Class :character      Class :character   Class :character   NA's:1500
## Mode  :character      Mode  :character   Mode  :character
##
##
##
##

```

```

## additives_n additives additives_tags additives_en
## Min. : 0.000 Length:1500 Length:1500 Length:1500
## 1st Qu.: 0.000 Class :character Class :character Class :character
## Median : 1.000 Mode :character Mode :character Mode :character
## Mean : 1.846
## 3rd Qu.: 3.000
## Max. :17.000
## NA's :514
## ingredients_from_palm_oil_n ingredients_from_palm_oil
## Min. :0.0000 Mode:logical
## 1st Qu.:0.0000 NA's:1500
## Median :0.0000
## Mean :0.0487
## 3rd Qu.:0.0000
## Max. :1.0000
## NA's :514
## ingredients_from_palm_oil_tags ingredients_that_may_be_from_palm_oil_n
## Length:1500 Min. :0.0000
## Class :character 1st Qu.:0.0000
## Mode :character Median :0.0000
## Mean :0.1379
## 3rd Qu.:0.0000
## Max. :4.0000
## NA's :514
## ingredients_that_may_be_from_palm_oil
## Mode:logical
## NA's:1500
##
##
##
##
## ingredients_that_may_be_from_palm_oil_tags nutrition_grade_uk
## Length:1500 Mode:logical
## Class :character NA's:1500
## Mode :character
##
##
##
## nutrition_grade_fr pnns_groups_1 pnns_groups_2
## Length:1500 Length:1500 Length:1500
## Class :character Class :character Class :character
## Mode :character Mode :character Mode :character
##
##
##
## states states_tags states_en
## Length:1500 Length:1500 Length:1500
## Class :character Class :character Class :character
## Mode :character Mode :character Mode :character
##
##

```

```

##
##
## main_category      main_category_en      image_url
## Length:1500        Length:1500          Length:1500
## Class :character    Class :character    Class :character
## Mode :character     Mode :character    Mode :character
##
##
##
## image_small_url      energy_100g      energy_from_fat_100g      fat_100g
## Length:1500          Min. : 0.0      Min. : 0.00      Min. : 0.00
## Class :character      1st Qu.: 369.8      1st Qu.: 35.98      1st Qu.: 0.90
## Mode :character        Median : 966.5      Median : 237.00      Median : 6.00
##                        Mean :1083.2      Mean : 668.41      Mean : 13.39
##                        3rd Qu.:1641.5      3rd Qu.: 974.00      3rd Qu.: 20.00
##                        Max. :3700.0      Max. :2900.00      Max. :100.00
##                        NA's :700      NA's :1486      NA's :708
## saturated_fat_100g butyric_acid_100g caproic_acid_100g caprylic_acid_100g
## Min. : 0.000      Mode:logical      Mode:logical      Mode:logical
## 1st Qu.: 0.200      NA's:1500      NA's:1500      NA's:1500
## Median : 1.700
## Mean : 4.874
## 3rd Qu.: 6.500
## Max. :57.000
## NA's :797
## capric_acid_100g lauric_acid_100g myristic_acid_100g palmitic_acid_100g
## Mode:logical      Mode:logical      Mode:logical      Mode:logical
## NA's:1500      NA's:1500      NA's:1500      NA's:1500
##
##
##
##
## stearic_acid_100g arachidic_acid_100g behenic_acid_100g
## Mode:logical      Mode:logical      Mode:logical
## NA's:1500      NA's:1500      NA's:1500
##
##
##
##
## lignoceric_acid_100g cerotic_acid_100g montanic_acid_100g
## Mode:logical      Mode:logical      Mode:logical
## NA's:1500      NA's:1500      NA's:1500
##
##
##
##
## melissic_acid_100g monounsaturated_fat_100g polyunsaturated_fat_100g
## Mode:logical      Min. : 0.00      Min. : 0.400
## NA's:1500      1st Qu.: 3.87      1st Qu.: 1.653
##                        Median : 9.50      Median : 3.900

```

```

##           Mean    :19.77           Mean    : 9.986
##           3rd Qu.:29.00           3rd Qu.:12.700
##           Max.    :75.00           Max.    :46.200
##           NA's    :1465           NA's    :1464
## omega_3_fat_100g alpha_linolenic_acid_100g eicosapentaenoic_acid_100g
## Min.    : 0.033   Min.    :0.0800   Min.    :0.721
## 1st Qu.: 1.300   1st Qu.:0.0905   1st Qu.:0.721
## Median : 3.000   Median :0.1010   Median :0.721
## Mean    : 3.726   Mean    :0.1737   Mean    :0.721
## 3rd Qu.: 3.200   3rd Qu.:0.2205   3rd Qu.:0.721
## Max.    :12.400   Max.    :0.3400   Max.    :0.721
## NA's    :1491    NA's    :1497    NA's    :1499
## docosahexaenoic_acid_100g omega_6_fat_100g linoleic_acid_100g
## Min.    :1.09           Min.    :0.25   Min.    :0.5000
## 1st Qu.:1.09           1st Qu.:0.25   1st Qu.:0.5165
## Median :1.09           Median :0.25   Median :0.5330
## Mean    :1.09           Mean    :0.25   Mean    :0.5330
## 3rd Qu.:1.09           3rd Qu.:0.25   3rd Qu.:0.5495
## Max.    :1.09           Max.    :0.25   Max.    :0.5660
## NA's    :1499           NA's    :1499   NA's    :1498
## arachidonic_acid_100g gamma_linolenic_acid_100g
## Mode:logical           Mode:logical
## NA's:1500              NA's:1500
##
##
##
##
## dihomogamma_linolenic_acid_100g omega_9_fat_100g oleic_acid_100g
## Mode:logical           Mode:logical           Mode:logical
## NA's:1500              NA's:1500              NA's:1500
##
##
##
##
## elaidic_acid_100g gondoic_acid_100g mead_acid_100g erucic_acid_100g
## Mode:logical           Mode:logical           Mode:logical           Mode:logical
## NA's:1500              NA's:1500              NA's:1500              NA's:1500
##
##
##
##
## nervonic_acid_100g trans_fat_100g cholesterol_100g carbohydrates_100g
## Mode:logical           Min.    :0.0000   Min.    :0.0000   Min.    : 0.000
## NA's:1500              1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.: 3.792
##                      Median :0.0000   Median :0.0000   Median :13.500
##                      Mean    :0.0105   Mean    :0.0265   Mean    :27.958
##                      3rd Qu.:0.0000   3rd Qu.:0.0026   3rd Qu.:55.000
##                      Max.    :0.1000   Max.    :0.4300   Max.    :100.000
##                      NA's    :1481    NA's    :1477    NA's    :708
## sugars_100g sucrose_100g glucose_100g fructose_100g
## Min.    : 0.00   Mode:logical   Mode:logical   Min.    :100

```

## 1st Qu.: 1.00	NA's:1500	NA's:1500	1st Qu.:100
## Median : 4.05			Median :100
## Mean : 12.66			Mean :100
## 3rd Qu.: 14.70			3rd Qu.:100
## Max. :100.00			Max. :100
## NA's :788			NA's :1499
## lactose_100g	maltose_100g	maltodextrins_100g	starch_100g
## Min. :0.000	Mode:logical	Mode:logical	Min. : 0.00
## 1st Qu.:0.250	NA's:1500	NA's:1500	1st Qu.: 9.45
## Median :0.500			Median :39.50
## Mean :2.933			Mean :30.73
## 3rd Qu.:4.400			3rd Qu.:42.85
## Max. :8.300			Max. :71.00
## NA's :1497			NA's :1493
## polyols_100g	fiber_100g	proteins_100g	casein_100g
## Min. : 8.60	Min. : 0.000	Min. : 0.000	Min. :1.1
## 1st Qu.:59.10	1st Qu.: 0.500	1st Qu.: 1.500	1st Qu.:1.1
## Median :67.00	Median : 1.750	Median : 6.000	Median :1.1
## Mean :56.06	Mean : 2.823	Mean : 7.563	Mean :1.1
## 3rd Qu.:69.80	3rd Qu.: 3.500	3rd Qu.:10.675	3rd Qu.:1.1
## Max. :70.00	Max. :46.700	Max. :61.000	Max. :1.1
## NA's :1491	NA's :994	NA's :710	NA's :1499
## serum_proteins_100g	nucleotides_100g	salt_100g	sodium_100g
## Mode:logical	Mode:logical	Min. : 0.0000	Min. : 0.0000
## NA's:1500	NA's:1500	1st Qu.: 0.0438	1st Qu.: 0.0172
##		Median : 0.4498	Median : 0.1771
##		Mean : 1.1205	Mean : 0.4409
##		3rd Qu.: 1.1938	3rd Qu.: 0.4700
##		Max. :102.0000	Max. :40.0000
##		NA's :780	NA's :780
## alcohol_100g	vitamin_a_100g	beta_carotene_100g	vitamin_d_100g
## Min. : 0.00	Min. :0.0000	Mode:logical	Min. :0e+00
## 1st Qu.: 0.00	1st Qu.:0.0000	NA's:1500	1st Qu.:0e+00
## Median : 5.50	Median :0.0001		Median :0e+00
## Mean :10.07	Mean :0.0003		Mean :0e+00
## 3rd Qu.:13.00	3rd Qu.:0.0006		3rd Qu.:0e+00
## Max. :50.00	Max. :0.0013		Max. :1e-04
## NA's :1433	NA's :1477		NA's :1485
## vitamin_e_100g	vitamin_k_100g	vitamin_c_100g	vitamin_b1_100g
## Min. :0.0005	Min. :0	Min. :0.000	Min. :0.0001
## 1st Qu.:0.0021	1st Qu.:0	1st Qu.:0.002	1st Qu.:0.0003
## Median :0.0044	Median :0	Median :0.019	Median :0.0004
## Mean :0.0069	Mean :0	Mean :0.025	Mean :0.0006
## 3rd Qu.:0.0097	3rd Qu.:0	3rd Qu.:0.030	3rd Qu.:0.0010
## Max. :0.0320	Max. :0	Max. :0.217	Max. :0.0013
## NA's :1478	NA's :1498	NA's :1459	NA's :1478
## vitamin_b2_100g	vitamin_pp_100g	vitamin_b6_100g	vitamin_b9_100g
## Min. :0.0002	Min. :0.0006	Min. :0.0001	Min. :0e+00
## 1st Qu.:0.0003	1st Qu.:0.0033	1st Qu.:0.0002	1st Qu.:0e+00
## Median :0.0009	Median :0.0069	Median :0.0008	Median :1e-04
## Mean :0.0011	Mean :0.0086	Mean :0.0112	Mean :1e-04
## 3rd Qu.:0.0013	3rd Qu.:0.0140	3rd Qu.:0.0012	3rd Qu.:2e-04
## Max. :0.0066	Max. :0.0160	Max. :0.2000	Max. :2e-04
## NA's :1483	NA's :1484	NA's :1481	NA's :1483



```

## vitamin_b12_100g biotin_100g pantothenic_acid_100g silica_100g
## Min. :0 Min. :0 Min. :0.0000 Min. :8e-04
## 1st Qu.:0 1st Qu.:0 1st Qu.:0.0007 1st Qu.:8e-04
## Median :0 Median :0 Median :0.0020 Median :8e-04
## Mean :0 Mean :0 Mean :0.0027 Mean :8e-04
## 3rd Qu.:0 3rd Qu.:0 3rd Qu.:0.0051 3rd Qu.:8e-04
## Max. :0 Max. :0 Max. :0.0060 Max. :8e-04
## NA's :1489 NA's :1498 NA's :1486 NA's :1499
## bicarbonate_100g potassium_100g chloride_100g calcium_100g
## Min. :0.0006 Min. :0.0000 Min. :0.0003 Min. :0.0000
## 1st Qu.:0.0678 1st Qu.:0.0650 1st Qu.:0.0006 1st Qu.:0.0450
## Median :0.1350 Median :0.1940 Median :0.0009 Median :0.1200
## Mean :0.1692 Mean :0.3288 Mean :0.0144 Mean :0.2040
## 3rd Qu.:0.2535 3rd Qu.:0.3670 3rd Qu.:0.0214 3rd Qu.:0.1985
## Max. :0.3720 Max. :1.4300 Max. :0.0420 Max. :1.0000
## NA's :1497 NA's :1487 NA's :1497 NA's :1449
## phosphorus_100g iron_100g magnesium_100g zinc_100g
## Min. :0.0430 Min. :0.0000 Min. :0.0000 Min. :0.0005
## 1st Qu.:0.1938 1st Qu.:0.0012 1st Qu.:0.0670 1st Qu.:0.0009
## Median :0.3185 Median :0.0042 Median :0.1040 Median :0.0017
## Mean :0.3777 Mean :0.0045 Mean :0.1066 Mean :0.0016
## 3rd Qu.:0.4340 3rd Qu.:0.0077 3rd Qu.:0.1300 3rd Qu.:0.0022
## Max. :1.1550 Max. :0.0137 Max. :0.3330 Max. :0.0026
## NA's :1488 NA's :1463 NA's :1479 NA's :1493
## copper_100g manganese_100g fluoride_100g selenium_100g
## Min. :0e+00 Min. :0 Min. :0 Min. :0
## 1st Qu.:1e-04 1st Qu.:0 1st Qu.:0 1st Qu.:0
## Median :1e-04 Median :0 Median :0 Median :0
## Mean :1e-04 Mean :0 Mean :0 Mean :0
## 3rd Qu.:1e-04 3rd Qu.:0 3rd Qu.:0 3rd Qu.:0
## Max. :1e-04 Max. :0 Max. :0 Max. :0
## NA's :1498 NA's :1499 NA's :1498 NA's :1499
## chromium_100g molybdenum_100g iodine_100g caffeine_100g
## Mode:logical Mode:logical Min. :0 Mode:logical
## NA's:1500 NA's:1500 1st Qu.:0 NA's:1500
## Median :0
## Mean :0
## 3rd Qu.:0
## Max. :0
## NA's :1499
## taurine_100g ph_100g fruits_vegetables_nuts_100g
## Mode:logical Mode:logical Min. : 2.00
## NA's:1500 NA's:1500 1st Qu.:11.25
## Median :42.00
## Mean :36.88
## 3rd Qu.:52.25
## Max. :80.00
## NA's :1470
## collagen_meat_protein_ratio_100g cocoa_100g chlorophyll_100g
## Min. :12.00 Min. :30 Mode:logical
## 1st Qu.:13.50 1st Qu.:47 NA's:1500
## Median :15.00 Median :60
## Mean :15.67 Mean :57
## 3rd Qu.:17.50 3rd Qu.:70

```

```
## Max.      :20.00                      Max.      :81
## NA's      :1497                      NA's      :1491
## carbon_footprint_100g nutrition_score_fr_100g nutrition_score_uk_100g
## Min.      : 12.00          Min.      : -12.000          Min.      : -12.000
## 1st Qu.: 97.42           1st Qu.:  1.000           1st Qu.:  0.000
## Median :182.85           Median :  7.000           Median :  6.000
## Mean    :131.18           Mean    :  7.941           Mean    :  7.631
## 3rd Qu.:190.78           3rd Qu.: 15.000           3rd Qu.: 16.000
## Max.     :198.70           Max.     : 28.000           Max.     : 28.000
## NA's     :1497            NA's     : 825            NA's     : 825
```

```
library(dplyr)
glimpse(df_food)
```

```
## Observations: 1,500
## Variables: 160
## $ V1                <int> 1, 2, 3, 4, 5, 6, 7...
## $ code              <int> 100030, 100050, 100...
## $ url               <chr> "http://world-en.op...
## $ creator           <chr> "sebleouf", "foodor...
## $ created_t         <int> 1424747544, 1450316...
## $ created_datetime  <chr> "2015-02-24T03:12:2...
## $ last_modified_t   <int> 1438445887, 1450817...
## $ last_modified_datetime <chr> "2015-08-01T16:18:0...
## $ product_name      <chr> "Confiture de frais...
## $ generic_name       <chr> "", "", "Pâtes de f...
## $ quantity          <chr> "265 g", "375g", "1...
## $ packaging          <chr> "Bocal,Verre", "Pla...
## $ packaging_tags     <chr> "bocal,verre", "pla...
## $ brands             <chr> "Casino Délices", "...
## $ brands_tags        <chr> "casino-delices", "...
## $ categories         <chr> "Aliments et boisso...
## $ categories_tags    <chr> "en:plant-based-foo...
## $ categories_en      <chr> "Plant-based foods ...
## $ origins            <chr> "", "", "", "", "Ar...
## $ origins_tags       <chr> "", "", "", "", "ar...
## $ manufacturing_places <chr> "France", "Belgium"...
## $ manufacturing_places_tags <chr> "france", "belgium"...
## $ labels             <chr> "", "", "", "Vegeta...
## $ labels_tags        <chr> "", "", "", "en:veg...
## $ labels_en          <chr> "", "", "", "Vegeta...
## $ emb_codes          <chr> "EMB 78015", "", "...
## $ emb_codes_tags     <chr> "emb-78015", "", "...
## $ first_packaging_code_geo <chr> "48.983333,2.066667...
## $ cities             <lgl> NA, NA, NA, NA, NA,...
## $ cities_tags        <chr> "andresy-yvelines-f...
## $ purchase_places    <chr> "Lyon,France", "NSW...
## $ stores             <chr> "Casino", "", "", "...
## $ countries          <chr> "France", "Australi...
## $ countries_tags     <chr> "en:france", "en:au...
## $ countries_en       <chr> "France", "Australi...
## $ ingredients_text   <chr> "Sucre de canne, fr...
## $ allergens          <chr> "", "", "", "", "",...
## $ allergens_en       <lgl> NA, NA, NA, NA, NA,...
## $ traces            <chr> "Lait,Fruits à coqu..."
```

## \$ traces_tags	<chr> "en:milk,en:nuts", ...
## \$ traces_en	<chr> "Milk,Nuts", "", "..."
## \$ serving_size	<chr> "15 g", "", "", "",...
## \$ no_nutriments	<lg1> NA, NA, NA, NA, NA,...
## \$ additives_n	<int> 1, NA, 2, 5, 0, NA,...
## \$ additives	<chr> "[ sucre-de-canne -...
## \$ additives_tags	<chr> "en:e440", "", "en:...
## \$ additives_en	<chr> "E440 - Pectins", "..."
## \$ ingredients_from_palm_oil_n	<int> 0, NA, 0, 0, 0, NA,...
## \$ ingredients_from_palm_oil	<lg1> NA, NA, NA, NA, NA,...
## \$ ingredients_from_palm_oil_tags	<chr> "", "", "", "", "",...
## \$ ingredients_that_may_be_from_palm_oil_n	<int> 0, NA, 0, 1, 0, NA,...
## \$ ingredients_that_may_be_from_palm_oil	<lg1> NA, NA, NA, NA, NA,...
## \$ ingredients_that_may_be_from_palm_oil_tags	<chr> "", "", "", "e471-m...
## \$ nutrition_grade_uk	<lg1> NA, NA, NA, NA, NA,...
## \$ nutrition_grade_fr	<chr> "d", "", "", "d", "..."
## \$ pnns_groups_1	<chr> "Sugary snacks", "S...
## \$ pnns_groups_2	<chr> "Sweets", "Chocolat...
## \$ states	<chr> "en:to-be-checked, ..."
## \$ states_tags	<chr> "en:to-be-checked,e...
## \$ states_en	<chr> "To be checked,Comp...
## \$ main_category	<chr> "en:plant-based-foo...
## \$ main_category_en	<chr> "Plant-based foods ..."
## \$ image_url	<chr> "http://en.openfood...
## \$ image_small_url	<chr> "http://en.openfood...
## \$ energy_100g	<dbl> 918, NA, NA, 766, 2...
## \$ energy_from_fat_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ fat_100g	<dbl> 0.00, NA, NA, 16.70...
## \$ saturated_fat_100g	<dbl> 0.000, NA, NA, 9.90...
## \$ butyric_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ caproic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ caprylic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ capric_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ lauric_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ myristic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ palmitic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ stearic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ arachidic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ behenic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ lignoceric_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ cerotic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ montanic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ melissic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ monounsaturated_fat_100g	<dbl> NA, NA, NA, 2.9, 9...
## \$ polyunsaturated_fat_100g	<dbl> NA, NA, NA, 3.9, 32...
## \$ omega_3_fat_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ alpha_linolenic_acid_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ eicosapentaenoic_acid_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ docosahexaenoic_acid_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ omega_6_fat_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ linoleic_acid_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ arachidonic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ gamma_linolenic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ dihomogamma_linolenic_acid_100g	<lg1> NA, NA, NA, NA, NA,...

## \$ omega_9_fat_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ oleic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ elaidic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ gondoic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ mead_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ erucic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ nervonic_acid_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ trans_fat_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ cholesterol_100g	<dbl> NA, NA, NA, 0.00020...
## \$ carbohydrates_100g	<dbl> 54.00, NA, NA, 5.70...
## \$ sugars_100g	<dbl> 54.00, NA, NA, 4.20...
## \$ sucrose_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ glucose_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ fructose_100g	<int> NA, NA, NA, NA, NA,...
## \$ lactose_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ maltose_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ maltodextrins_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ starch_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ polyols_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ fiber_100g	<dbl> NA, NA, NA, 0.2, 9....
## \$ proteins_100g	<dbl> 0.00, NA, NA, 2.90,...
## \$ casein_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ serum_proteins_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ nucleotides_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ salt_100g	<dbl> 0.0000000, NA, NA, ...
## \$ sodium_100g	<dbl> 0.0000000, NA, NA, ...
## \$ alcohol_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_a_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ beta_carotene_100g	<lg1> NA, NA, NA, NA, NA,...
## \$ vitamin_d_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_e_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_k_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_c_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_b1_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_b2_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_pp_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_b6_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_b9_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ vitamin_b12_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ biotin_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ pantothenic_acid_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ silica_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ bicarbonate_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ potassium_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ chloride_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ calcium_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ phosphorus_100g	<dbl> NA, NA, NA, NA, 1.1...
## \$ iron_100g	<dbl> NA, NA, NA, NA, 0.0...
## \$ magnesium_100g	<dbl> NA, NA, NA, NA, 0.1...
## \$ zinc_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ copper_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ manganese_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ fluoride_100g	<dbl> NA, NA, NA, NA, NA,...
## \$ selenium_100g	<dbl> NA, NA, NA, NA, NA,...

```
## $ chromium_100g <lgl> NA, NA, NA, NA, NA,...
## $ molybdenum_100g <lgl> NA, NA, NA, NA, NA,...
## $ iodine_100g <dbl> NA, NA, NA, NA, NA,...
## $ caffeine_100g <lgl> NA, NA, NA, NA, NA,...
## $ taurine_100g <lgl> NA, NA, NA, NA, NA,...
## $ ph_100g <lgl> NA, NA, NA, NA, NA,...
## $ fruits_vegetables_nuts_100g <dbl> 54, NA, NA, NA, NA,...
## $ collagen_meat_protein_ratio_100g <int> NA, NA, NA, NA, NA,...
## $ cocoa_100g <int> NA, NA, NA, NA, NA,...
## $ chlorophyl_100g <lgl> NA, NA, NA, NA, NA,...
## $ carbon_footprint_100g <dbl> NA, NA, NA, NA, NA,...
## $ nutrition_score_fr_100g <int> 11, NA, NA, 11, 17,...
## $ nutrition_score_uk_100g <int> 11, NA, NA, 11, 17,...
```

```
# View column names of food
names(df_food)
```

```
## [1] "V1"
## [2] "code"
## [3] "url"
## [4] "creator"
## [5] "created_t"
## [6] "created_datetime"
## [7] "last_modified_t"
## [8] "last_modified_datetime"
## [9] "product_name"
## [10] "generic_name"
## [11] "quantity"
## [12] "packaging"
## [13] "packaging_tags"
## [14] "brands"
## [15] "brands_tags"
## [16] "categories"
## [17] "categories_tags"
## [18] "categories_en"
## [19] "origins"
## [20] "origins_tags"
## [21] "manufacturing_places"
## [22] "manufacturing_places_tags"
## [23] "labels"
## [24] "labels_tags"
## [25] "labels_en"
## [26] "emb_codes"
## [27] "emb_codes_tags"
## [28] "first_packaging_code_geo"
## [29] "cities"
## [30] "cities_tags"
## [31] "purchase_places"
## [32] "stores"
## [33] "countries"
## [34] "countries_tags"
## [35] "countries_en"
## [36] "ingredients_text"
## [37] "allergens"
## [38] "allergens_en"
```

```

## [39] "traces"
## [40] "traces_tags"
## [41] "traces_en"
## [42] "serving_size"
## [43] "no_nutriments"
## [44] "additives_n"
## [45] "additives"
## [46] "additives_tags"
## [47] "additives_en"
## [48] "ingredients_from_palm_oil_n"
## [49] "ingredients_from_palm_oil"
## [50] "ingredients_from_palm_oil_tags"
## [51] "ingredients_that_may_be_from_palm_oil_n"
## [52] "ingredients_that_may_be_from_palm_oil"
## [53] "ingredients_that_may_be_from_palm_oil_tags"
## [54] "nutrition_grade_uk"
## [55] "nutrition_grade_fr"
## [56] "pnns_groups_1"
## [57] "pnns_groups_2"
## [58] "states"
## [59] "states_tags"
## [60] "states_en"
## [61] "main_category"
## [62] "main_category_en"
## [63] "image_url"
## [64] "image_small_url"
## [65] "energy_100g"
## [66] "energy_from_fat_100g"
## [67] "fat_100g"
## [68] "saturated_fat_100g"
## [69] "butyric_acid_100g"
## [70] "caproic_acid_100g"
## [71] "caprylic_acid_100g"
## [72] "capric_acid_100g"
## [73] "lauric_acid_100g"
## [74] "myristic_acid_100g"
## [75] "palmitic_acid_100g"
## [76] "stearic_acid_100g"
## [77] "arachidic_acid_100g"
## [78] "behenic_acid_100g"
## [79] "lignoceric_acid_100g"
## [80] "cerotic_acid_100g"
## [81] "montanic_acid_100g"
## [82] "melissic_acid_100g"
## [83] "monounsaturated_fat_100g"
## [84] "polyunsaturated_fat_100g"
## [85] "omega_3_fat_100g"
## [86] "alpha_linolenic_acid_100g"
## [87] "eicosapentaenoic_acid_100g"
## [88] "docosahexaenoic_acid_100g"
## [89] "omega_6_fat_100g"
## [90] "linoleic_acid_100g"
## [91] "arachidonic_acid_100g"
## [92] "gamma_linolenic_acid_100g"

```

## [93] "dihomo\_gamma\_linolenic\_acid\_100g"  
## [94] "omega\_9\_fat\_100g"  
## [95] "oleic\_acid\_100g"  
## [96] "elaidic\_acid\_100g"  
## [97] "gondoic\_acid\_100g"  
## [98] "mead\_acid\_100g"  
## [99] "erucic\_acid\_100g"  
## [100] "nervonic\_acid\_100g"  
## [101] "trans\_fat\_100g"  
## [102] "cholesterol\_100g"  
## [103] "carbohydrates\_100g"  
## [104] "sugars\_100g"  
## [105] "sucrose\_100g"  
## [106] "glucose\_100g"  
## [107] "fructose\_100g"  
## [108] "lactose\_100g"  
## [109] "maltose\_100g"  
## [110] "maltodextrins\_100g"  
## [111] "starch\_100g"  
## [112] "polyols\_100g"  
## [113] "fiber\_100g"  
## [114] "proteins\_100g"  
## [115] "casein\_100g"  
## [116] "serum\_proteins\_100g"  
## [117] "nucleotides\_100g"  
## [118] "salt\_100g"  
## [119] "sodium\_100g"  
## [120] "alcohol\_100g"  
## [121] "vitamin\_a\_100g"  
## [122] "beta\_carotene\_100g"  
## [123] "vitamin\_d\_100g"  
## [124] "vitamin\_e\_100g"  
## [125] "vitamin\_k\_100g"  
## [126] "vitamin\_c\_100g"  
## [127] "vitamin\_b1\_100g"  
## [128] "vitamin\_b2\_100g"  
## [129] "vitamin\_pp\_100g"  
## [130] "vitamin\_b6\_100g"  
## [131] "vitamin\_b9\_100g"  
## [132] "vitamin\_b12\_100g"  
## [133] "biotin\_100g"  
## [134] "pantothenic\_acid\_100g"  
## [135] "silica\_100g"  
## [136] "bicarbonate\_100g"  
## [137] "potassium\_100g"  
## [138] "chloride\_100g"  
## [139] "calcium\_100g"  
## [140] "phosphorus\_100g"  
## [141] "iron\_100g"  
## [142] "magnesium\_100g"  
## [143] "zinc\_100g"  
## [144] "copper\_100g"  
## [145] "manganese\_100g"  
## [146] "fluoride\_100g"

```
## [147] "selenium_100g"
## [148] "chromium_100g"
## [149] "molybdenum_100g"
## [150] "iodine_100g"
## [151] "caffeine_100g"
## [152] "taurine_100g"
## [153] "ph_100g"
## [154] "fruits_vegetables_nuts_100g"
## [155] "collagen_meat_protein_ratio_100g"
## [156] "cocoa_100g"
## [157] "chlorophyll_100g"
## [158] "carbon_footprint_100g"
## [159] "nutrition_score_fr_100g"
## [160] "nutrition_score_uk_100g"
```

## Removing Duplicates

A vector has been created for you that lists out all of the duplicates; all you need to do is remove those columns from the dataset.

```
# Define vector of duplicate cols
duplicates <- c(4, 6, 11, 13, 15, 17, 18, 20, 22,
               24, 25, 28, 32, 34, 36, 38, 40,
               44, 46, 48, 51, 54, 65, 158)

# Remove duplicates from food: food2
food2<-df_food[,-duplicates]
```

## Removing Useless Info

```
# Define useless vector
useless <- c(1, 2, 3, 32:41)

# Remove useless columns from food2: food3
food3<-food2[,-useless]
```

## Finding columns

Looking much nicer! Recall from the first exercise that you are assuming you will be analyzing the sugar content of these foods. Therefore, your next step is to look at a summary of the nutrition information.

All of the columns with nutrition info contain the character string “100g” as part of their name, which makes it easy to identify them.

```
#Create a vector called nutrition containing the column indices of the nutrition data. To do this, use

library(stringr)
nutrition <- str_detect(names(food3), pattern = "100g")

#View a summary of the nutrition columns.
summary(food3[,nutrition])
```



```

## energy_from_fat_100g    fat_100g    saturated_fat_100g
## Min.    : 0.00      Min.    : 0.00    Min.    : 0.000
## 1st Qu.: 35.98      1st Qu.: 0.90    1st Qu.: 0.200
## Median : 237.00      Median : 6.00    Median : 1.700
## Mean    : 668.41      Mean    : 13.39   Mean    : 4.874
## 3rd Qu.: 974.00      3rd Qu.: 20.00   3rd Qu.: 6.500
## Max.    :2900.00      Max.    :100.00   Max.    :57.000
## NA's    :1486        NA's    :708     NA's    :797
## butyric_acid_100g caproic_acid_100g caprylic_acid_100g capric_acid_100g
## Mode:logical      Mode:logical      Mode:logical      Mode:logical
## NA's:1500          NA's:1500          NA's:1500          NA's:1500
##
##
##
##
## lauric_acid_100g myristic_acid_100g palmitic_acid_100g stearic_acid_100g
## Mode:logical      Mode:logical      Mode:logical      Mode:logical
## NA's:1500          NA's:1500          NA's:1500          NA's:1500
##
##
##
##
## arachidic_acid_100g behenic_acid_100g lignoceric_acid_100g
## Mode:logical      Mode:logical      Mode:logical
## NA's:1500          NA's:1500          NA's:1500
##
##
##
##
## cerotic_acid_100g montanic_acid_100g melissic_acid_100g
## Mode:logical      Mode:logical      Mode:logical
## NA's:1500          NA's:1500          NA's:1500
##
##
##
##
## monounsaturated_fat_100g polyunsaturated_fat_100g omega_3_fat_100g
## Min.    : 0.00      Min.    : 0.400    Min.    : 0.033
## 1st Qu.: 3.87      1st Qu.: 1.653     1st Qu.: 1.300
## Median : 9.50      Median : 3.900     Median : 3.000
## Mean    :19.77      Mean    : 9.986     Mean    : 3.726
## 3rd Qu.:29.00      3rd Qu.:12.700     3rd Qu.: 3.200
## Max.    :75.00      Max.    :46.200     Max.    :12.400
## NA's    :1465        NA's    :1464     NA's    :1491
## alpha_linolenic_acid_100g eicosapentaenoic_acid_100g
## Min.    :0.0800      Min.    :0.721
## 1st Qu.:0.0905      1st Qu.:0.721
## Median :0.1010      Median :0.721
## Mean    :0.1737      Mean    :0.721
## 3rd Qu.:0.2205      3rd Qu.:0.721

```

```

## Max. :0.3400 Max. :0.721
## NA's :1497 NA's :1499
## docosahexaenoic_acid_100g omega_6_fat_100g linoleic_acid_100g
## Min. :1.09 Min. :0.25 Min. :0.5000
## 1st Qu.:1.09 1st Qu.:0.25 1st Qu.:0.5165
## Median :1.09 Median :0.25 Median :0.5330
## Mean :1.09 Mean :0.25 Mean :0.5330
## 3rd Qu.:1.09 3rd Qu.:0.25 3rd Qu.:0.5495
## Max. :1.09 Max. :0.25 Max. :0.5660
## NA's :1499 NA's :1499 NA's :1498
## arachidonic_acid_100g gamma_linolenic_acid_100g
## Mode:logical Mode:logical
## NA's:1500 NA's:1500
##
##
##
##
## dihomogamma_linolenic_acid_100g omega_9_fat_100g oleic_acid_100g
## Mode:logical Mode:logical Mode:logical
## NA's:1500 NA's:1500 NA's:1500
##
##
##
##
## elaidic_acid_100g gondoic_acid_100g mead_acid_100g erucic_acid_100g
## Mode:logical Mode:logical Mode:logical Mode:logical
## NA's:1500 NA's:1500 NA's:1500 NA's:1500
##
##
##
##
## nervonic_acid_100g trans_fat_100g cholesterol_100g carbohydrates_100g
## Mode:logical Min. :0.0000 Min. :0.0000 Min. : 0.000
## NA's:1500 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.: 3.792
## Median :0.0000 Median :0.0000 Median : 13.500
## Mean :0.0105 Mean :0.0265 Mean : 27.958
## 3rd Qu.:0.0000 3rd Qu.:0.0026 3rd Qu.: 55.000
## Max. :0.1000 Max. :0.4300 Max. :100.000
## NA's :1481 NA's :1477 NA's :708
## sugars_100g sucrose_100g glucose_100g fructose_100g
## Min. : 0.00 Mode:logical Mode:logical Min. :100
## 1st Qu.: 1.00 NA's:1500 NA's:1500 1st Qu.:100
## Median : 4.05 Median :100
## Mean : 12.66 Mean :100
## 3rd Qu.: 14.70 3rd Qu.:100
## Max. :100.00 Max. :100
## NA's :788 NA's :1499
## lactose_100g maltose_100g maltodextrins_100g starch_100g
## Min. :0.000 Mode:logical Mode:logical Min. : 0.00
## 1st Qu.:0.250 NA's:1500 NA's:1500 1st Qu.: 9.45
## Median :0.500 Median :39.50

```

```

## Mean      :2.933                               Mean      :30.73
## 3rd Qu.:4.400                               3rd Qu.:42.85
## Max.      :8.300                               Max.      :71.00
## NA's      :1497                               NA's      :1493
## polyols_100g    fiber_100g    proteins_100g    casein_100g
## Min.      : 8.60    Min.      : 0.000    Min.      : 0.000    Min.      :1.1
## 1st Qu.:59.10    1st Qu.: 0.500    1st Qu.: 1.500    1st Qu.:1.1
## Median :67.00    Median : 1.750    Median : 6.000    Median :1.1
## Mean      :56.06    Mean      : 2.823    Mean      : 7.563    Mean      :1.1
## 3rd Qu.:69.80    3rd Qu.: 3.500    3rd Qu.:10.675    3rd Qu.:1.1
## Max.      :70.00    Max.      :46.700    Max.      :61.000    Max.      :1.1
## NA's      :1491    NA's      :994    NA's      :710    NA's      :1499
## serum_proteins_100g nucleotides_100g    salt_100g    sodium_100g
## Mode:logical    Mode:logical    Min.      : 0.0000    Min.      : 0.0000
## NA's:1500    NA's:1500    1st Qu.: 0.0438    1st Qu.: 0.0172
##                                     Median : 0.4498    Median : 0.1771
##                                     Mean      : 1.1205    Mean      : 0.4409
##                                     3rd Qu.: 1.1938    3rd Qu.: 0.4700
##                                     Max.      :102.0000    Max.      :40.0000
##                                     NA's      :780    NA's      :780
## alcohol_100g    vitamin_a_100g    beta_carotene_100g    vitamin_d_100g
## Min.      : 0.00    Min.      :0.0000    Mode:logical    Min.      :0e+00
## 1st Qu.: 0.00    1st Qu.:0.0000    NA's:1500    1st Qu.:0e+00
## Median : 5.50    Median :0.0001                                     Median :0e+00
## Mean      :10.07    Mean      :0.0003                                     Mean      :0e+00
## 3rd Qu.:13.00    3rd Qu.:0.0006                                     3rd Qu.:0e+00
## Max.      :50.00    Max.      :0.0013                                     Max.      :1e-04
## NA's      :1433    NA's      :1477                                     NA's      :1485
## vitamin_e_100g    vitamin_k_100g    vitamin_c_100g    vitamin_b1_100g
## Min.      :0.0005    Min.      :0    Min.      :0.000    Min.      :0.0001
## 1st Qu.:0.0021    1st Qu.:0    1st Qu.:0.002    1st Qu.:0.0003
## Median :0.0044    Median :0    Median :0.019    Median :0.0004
## Mean      :0.0069    Mean      :0    Mean      :0.025    Mean      :0.0006
## 3rd Qu.:0.0097    3rd Qu.:0    3rd Qu.:0.030    3rd Qu.:0.0010
## Max.      :0.0320    Max.      :0    Max.      :0.217    Max.      :0.0013
## NA's      :1478    NA's      :1498    NA's      :1459    NA's      :1478
## vitamin_b2_100g    vitamin_pp_100g    vitamin_b6_100g    vitamin_b9_100g
## Min.      :0.0002    Min.      :0.0006    Min.      :0.0001    Min.      :0e+00
## 1st Qu.:0.0003    1st Qu.:0.0033    1st Qu.:0.0002    1st Qu.:0e+00
## Median :0.0009    Median :0.0069    Median :0.0008    Median :1e-04
## Mean      :0.0011    Mean      :0.0086    Mean      :0.0112    Mean      :1e-04
## 3rd Qu.:0.0013    3rd Qu.:0.0140    3rd Qu.:0.0012    3rd Qu.:2e-04
## Max.      :0.0066    Max.      :0.0160    Max.      :0.2000    Max.      :2e-04
## NA's      :1483    NA's      :1484    NA's      :1481    NA's      :1483
## vitamin_b12_100g    biotin_100g    pantothenic_acid_100g    silica_100g
## Min.      :0    Min.      :0    Min.      :0.0000    Min.      :8e-04
## 1st Qu.:0    1st Qu.:0    1st Qu.:0.0007    1st Qu.:8e-04
## Median :0    Median :0    Median :0.0020    Median :8e-04
## Mean      :0    Mean      :0    Mean      :0.0027    Mean      :8e-04
## 3rd Qu.:0    3rd Qu.:0    3rd Qu.:0.0051    3rd Qu.:8e-04
## Max.      :0    Max.      :0    Max.      :0.0060    Max.      :8e-04
## NA's      :1489    NA's      :1498    NA's      :1486    NA's      :1499
## bicarbonate_100g    potassium_100g    chloride_100g    calcium_100g
## Min.      :0.0006    Min.      :0.0000    Min.      :0.0003    Min.      :0.0000

```

```

## 1st Qu.:0.0678 1st Qu.:0.0650 1st Qu.:0.0006 1st Qu.:0.0450
## Median :0.1350 Median :0.1940 Median :0.0009 Median :0.1200
## Mean :0.1692 Mean :0.3288 Mean :0.0144 Mean :0.2040
## 3rd Qu.:0.2535 3rd Qu.:0.3670 3rd Qu.:0.0214 3rd Qu.:0.1985
## Max. :0.3720 Max. :1.4300 Max. :0.0420 Max. :1.0000
## NA's :1497 NA's :1487 NA's :1497 NA's :1449
## phosphorus_100g iron_100g magnesium_100g zinc_100g
## Min. :0.0430 Min. :0.0000 Min. :0.0000 Min. :0.0005
## 1st Qu.:0.1938 1st Qu.:0.0012 1st Qu.:0.0670 1st Qu.:0.0009
## Median :0.3185 Median :0.0042 Median :0.1040 Median :0.0017
## Mean :0.3777 Mean :0.0045 Mean :0.1066 Mean :0.0016
## 3rd Qu.:0.4340 3rd Qu.:0.0077 3rd Qu.:0.1300 3rd Qu.:0.0022
## Max. :1.1550 Max. :0.0137 Max. :0.3330 Max. :0.0026
## NA's :1488 NA's :1463 NA's :1479 NA's :1493
## copper_100g manganese_100g fluoride_100g selenium_100g
## Min. :0e+00 Min. :0 Min. :0 Min. :0
## 1st Qu.:1e-04 1st Qu.:0 1st Qu.:0 1st Qu.:0
## Median :1e-04 Median :0 Median :0 Median :0
## Mean :1e-04 Mean :0 Mean :0 Mean :0
## 3rd Qu.:1e-04 3rd Qu.:0 3rd Qu.:0 3rd Qu.:0
## Max. :1e-04 Max. :0 Max. :0 Max. :0
## NA's :1498 NA's :1499 NA's :1498 NA's :1499
## chromium_100g molybdenum_100g iodine_100g caffeine_100g
## Mode:logical Mode:logical Min. :0 Mode:logical
## NA's:1500 NA's:1500 1st Qu.:0 NA's:1500
## Median :0
## Mean :0
## 3rd Qu.:0
## Max. :0
## NA's :1499
## taurine_100g ph_100g fruits_vegetables_nuts_100g
## Mode:logical Mode:logical Min. : 2.00
## NA's:1500 NA's:1500 1st Qu.:11.25
## Median :42.00
## Mean :36.88
## 3rd Qu.:52.25
## Max. :80.00
## NA's :1470
## collagen_meat_protein_ratio_100g cocoa_100g chlorophyl_100g
## Min. :12.00 Min. :30 Mode:logical
## 1st Qu.:13.50 1st Qu.:47 NA's:1500
## Median :15.00 Median :60
## Mean :15.67 Mean :57
## 3rd Qu.:17.50 3rd Qu.:70
## Max. :20.00 Max. :81
## NA's :1497 NA's :1491
## nutrition_score_fr_100g nutrition_score_uk_100g
## Min. : -12.000 Min. : -12.000
## 1st Qu.: 1.000 1st Qu.: 0.000
## Median : 7.000 Median : 6.000
## Mean : 7.941 Mean : 7.631
## 3rd Qu.: 15.000 3rd Qu.: 16.000
## Max. : 28.000 Max. : 28.000
## NA's :825 NA's :825

```

## Replacing missing values

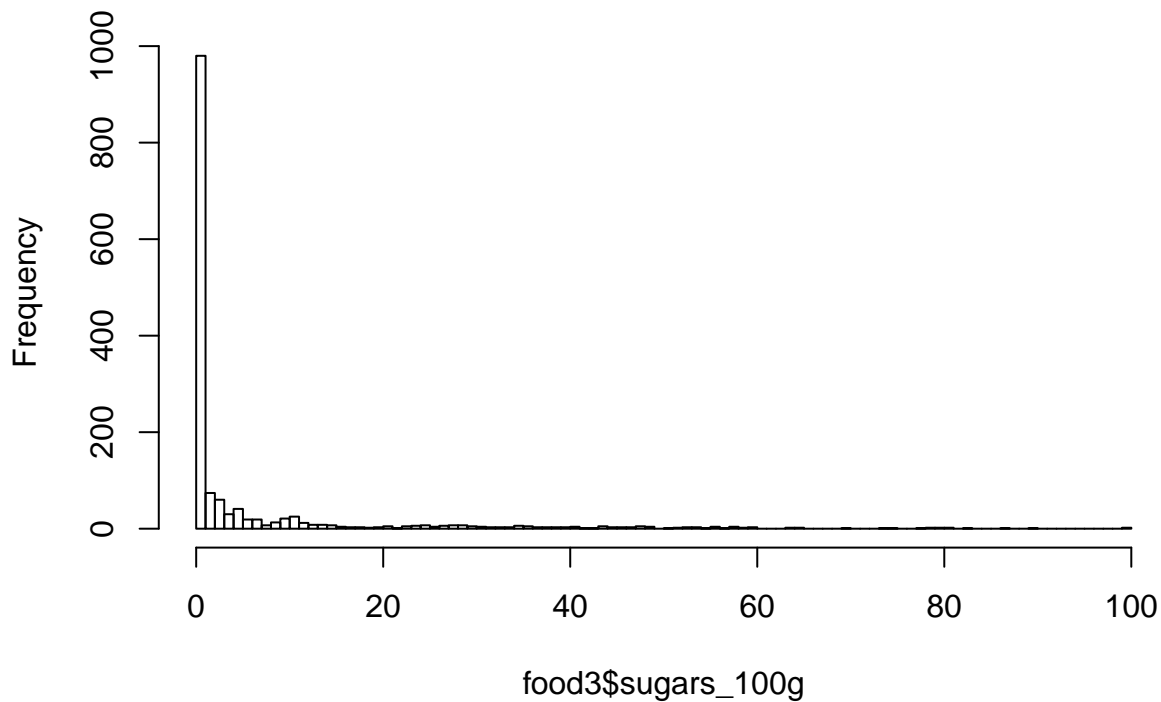
In this exercise, you'll replace all NA values with zeroes in the `sugars_100g` column and make histograms to visualize the result. Then, you will exclude the observations which have no sugar to see how the distribution changes.

```
# Find indices of sugar NA values: missing
missing <- is.na(food3$sugars_100g)

# Replace NA values with 0
food3$sugars_100g[missing] <- 0

# Create first histogram
hist(food3$sugars_100g, breaks = 100)
```

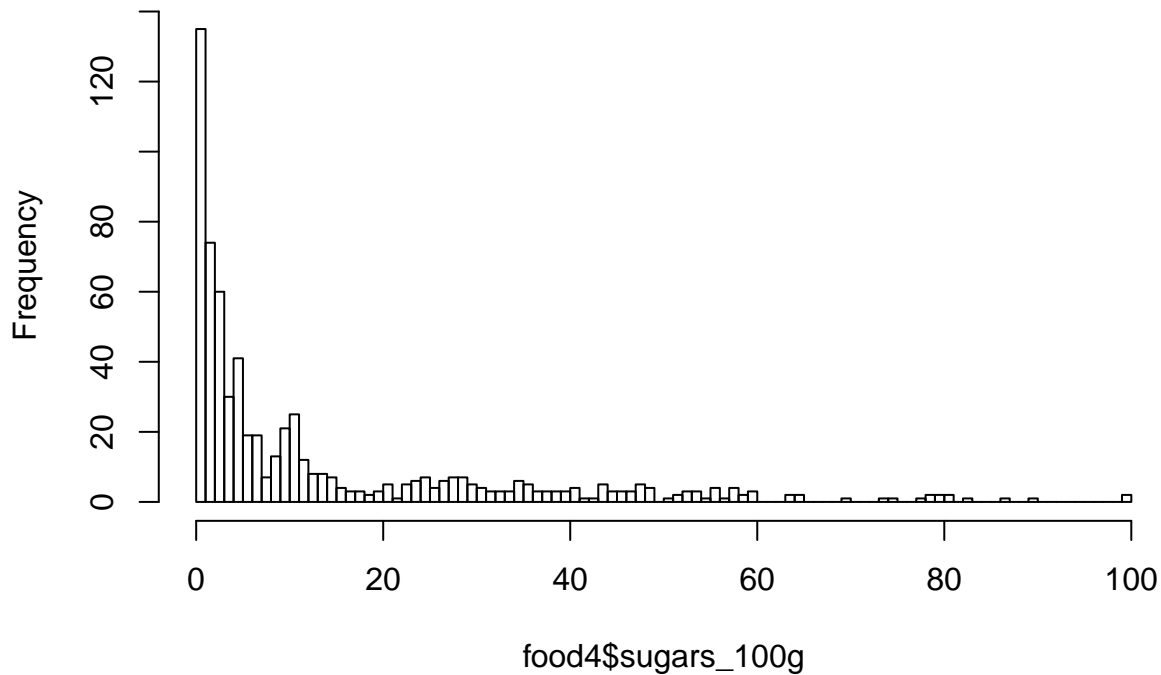
**Histogram of food3\$sugars\_100g**



```
# Create food4
food4 <- food3[food3$sugars_100g > 0, ]

# Create second histogram
hist(food4$sugars_100g, breaks = 100)
```

## Histogram of food4\$sugars\_100g



*#To get a general idea of how many of these foods are packaged in plastic, you can look through the pac*

```
# Find entries containing "plasti": plastic
plastic <- str_detect(food3$packaging, "plasti")

# Print the sum of plastic
sum(plastic)
```

```
## [1] 232
```

## Exercise 4\_PublicSchools\_Attendance

Data\_Sales: <https://www.datacamp.com/courses/importing-cleaning-data-in-r-case-studies> Data\_file: attendance.xls

In this chapter, you'll work with attendance data from public schools in the US, organized by school level and state, during the 2007-2008 academic year. The data contain information on average daily attendance (ADA) as a percentage of total enrollment, school day length, and school year length.

### Importing

```
# Load the gdata package
library(gdata)
```

```
## gdata: read.xls support for 'XLS' (Excel 97-2004) files ENABLED.
```

```
##
```

```
## gdata: read.xls support for 'XLSX' (Excel 2007+) files ENABLED.
```

```
##
## Attaching package: 'gdata'

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

## The following objects are masked from 'package:dplyr':
##
##     combine, first, last

## The following object is masked from 'package:stats':
##
##     nobs

## The following object is masked from 'package:utils':
##
##     object.size

## The following object is masked from 'package:base':
##
##     startsWith

# Import the spreadsheet: att
att<- read.xls("attendance.xls")
```

## Examining the data

```
str(att)

## 'data.frame':   59 obs. of  17 variables:
## $ Table.43..Average.daily.attendance..ADA..as.a.percentage.of.total.enrollment..school.day.length..
## $ X
## $ X.1
## $ X.2
## $ X.3
## $ X.4
## $ X.5
## $ X.6
## $ X.7
## $ X.8
## $ X.9
## $ X.10
## $ X.11
## $ X.12
## $ X.13
## $ X.14
## $ X.15
```

These are some messy data! The column names are mostly missing, there are irrelevant notes at the end of the data frame, and it looks like the numeric data were imported as factors. Let's start the cleaning process!

## Removing unnecessary rows

When you're importing a messy spreadsheet into R, it's good practice to compare the original spreadsheet with what you've imported. It turns out that, by default, the `read.xls()` function skips empty rows such as

the 11th and 17th.

After viewing your data frame, you realize you still need to get rid of the third row of att, as well as rows 56 through 59.

```
# Create remove
remove<-c(3,56:59)

# Create att2
att2<- att[!(remove),]
```

## Removing useless columns

Once more, for reference, here is an image of the first 22 rows of the original spreadsheet. You can see here that the columns 3, 5, 7, 9, 11, 13, 15, and 17 (or columns C, E, G, I, K, M, O, Q in Excel) don't contain the values of average daily attendance (ADA). You'll get rid of them in this exercise.

```
# Create remove
remove<-c(3, 5, 7, 9, 11, 13, 15, 17)

# Create att3
att3<- att2[,-remove]
```

## Splitting the data

In many cases, a single data frame stores multiple "tables" of information. You can often diagnose this problem by looking at the column names and noticing duplicate rows.

In this data frame, columns 1, 6, and 7 represent attendance data for US elementary schools, columns 1, 8, and 9 represent data for secondary schools, and columns 1 through 5 represent data for all schools in the US.

Each of these should be stored as its own separate data frame, so you'll split them up here.

```
#Subset att3 to include only data for elementary schools (columns 1, 6, and 7). Name the resulting data frame att_elem.
att_elem<- att3[,c(1,6,7)]

#Subset att3 to include only data for secondary schools (columns 1, 8, and 9). Name the resulting data frame att_sec.
att_sec<- att3[,c(1,8,9)]

#Subset att3 to include data for all schools (columns 1 through 5). Name the resulting data frame att4.
att4<- att3[, c(1:5)]
```

## Replacing the names

Since you went through so much trouble finding out which row stored the variable names, you should store that row as the actual column names of the data frame. We've modified the names a bit in order to be more stylistically sound; they're stored as cnames in the editor.

This will also allow you to remove the first two rows (currently storing variable names).

```
# Define cnames vector (don't change)
cnames <- c("state", "avg_attend_pct", "avg_hr_per_day",
            "avg_day_per_yr", "avg_hr_per_yr")
```



```

# Assign column names of att4
colnames(att4) <- cnames

# Remove first two rows of att4: att5
att5<- att4[-c(1,2),]

# View the names of att5
names(att5)

## [1] "state"          "avg_attend_pct" "avg_hr_per_day" "avg_day_per_yr"
## [5] "avg_hr_per_yr"

```

## Cleaning up extra characters

One of the most irritating things about this dataset is that the state names are all stored as the same number of characters, with periods padding the ends of the shorter states. That may be helpful for reading the spreadsheet, but it makes your life harder, so you'll deal with it in this exercise.

One pitfall to avoid: `.` is a special character in the language of regular expressions (a.k.a. regex). In order to specify that you actually want to remove periods and not their regex equivalent (which is “all characters”), use `\.` This is called an “escape” sequence.

```

library(stringr)

#Use the function str_replace_all() to replace all periods in the state column of att5 with "". Remember
att5$state<- str_replace_all(att5$state,pattern="\.", "")

#Remove white space around the state names, assigning the result back to att5$state once more. There's
att5$state <- str_trim(att5$state)
head(att5, n=20)

```

```

##           state avg_attend_pct avg_hr_per_day avg_day_per_yr
## 4      United States      93.1          6.6          180
## 5         Alabama      93.8          7.0          180
## 6          Alaska      89.9          6.5          180
## 7         Arizona      89.0          6.4          181
## 8        Arkansas      91.8          6.9          179
## 9       California      93.2          6.2          181
## 10      Colorado      93.9          7.0          171
## 11    Connecticut      87.9          6.5          181
## 12       Delaware      89.8          6.7          181
## 13 District of Columbia      91.2          6.9          181
## 14         Florida      92.7          6.4          184
## 15         Georgia      93.3          6.8          181
## 16         Hawaii      90.7          6.3          179
## 17         Idaho      92.4          6.6          173
## 18        Illinois      94.0          6.5          177
## 19         Indiana      95.7          6.8          180
## 20          Iowa      94.8          6.9          180
## 21         Kansas      95.4          7.0          178
## 22        Kentucky      93.1          6.7          180
## 23       Louisiana      90.3          7.1          178
##      avg_hr_per_yr

```

## 4	1,193
## 5	1,267
## 6	1,163
## 7	1,159
## 8	1,229
## 9	1,129
## 10	1,199
## 11	1,173
## 12	1,208
## 13	1,256
## 14	1,184
## 15	1,229
## 16	1,118
## 17	1,143
## 18	1,147
## 19	1,222
## 20	1,232
## 21	1,240
## 22	1,202
## 23	1,263

---

FIN