

Worksheet-7a in R

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2022-12-11

Packages

```
library(openxlsx)
library("writexl")
library(mlbench)
library(xlsx)

##
## Attaching package: 'xlsx'

## The following objects are masked from 'package:openxlsx':
##
##      createWorkbook, loadWorkbook, read.xlsx, saveWorkbook, write.xlsx

library(Hmisc)

## Loading required package: lattice
## Loading required package: survival
## Loading required package: Formula
## Loading required package: ggplot2

##
## Attaching package: 'Hmisc'

## The following objects are masked from 'package:base':
##
##      format.pval, units

library(pastecs)
library(readxl)
library("AppliedPredictiveModeling")
```

1. Create a data frame for the table below.

```
Student <- seq(1:10)
Pre_test <- c(55,54,47,57,51,61,57,54,63,58)
Post_test <- c(61,60,56,63,56,63,59,56,62,61)

Stude_scoresDF <- data.frame(Student,Pre_test,Post_test)
Stude_scoresDF
```

	Student	Pre_test	Post_test
## 1	1	55	61
## 2	2	54	60

```
## 3      3      47      56
## 4      4      57      63
## 5      5      51      56
## 6      6      61      63
## 7      7      57      59
## 8      8      54      56
## 9      9      63      62
## 10     10     58      61
```

a. Compute the descriptive statistics using different packages (Hmisc and pastecs). Write the codes and its result.

```
# Hmisc
Hmisc_d <- describe(Stude_scoresDF)
Hmisc_d

## Stude_scoresDF
##
## 3 Variables      10 Observations
## -----
## Student
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      10      0      10      1      5.5      3.667      1.45      1.90
##      .25      .50      .75      .90      .95
##      3.25      5.50      7.75      9.10      9.55
##
## lowest : 1 2 3 4 5, highest: 6 7 8 9 10
##
## Value      1 2 3 4 5 6 7 8 9 10
## Frequency  1 1 1 1 1 1 1 1 1 1
## Proportion 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1
## -----
## Pre_test
##      n missing distinct      Info      Mean      Gmd
##      10      0      8      0.988      55.7      5.444
##
## lowest : 47 51 54 55 57, highest: 55 57 58 61 63
##
## Value      47 51 54 55 57 58 61 63
## Frequency  1 1 2 1 2 1 1 1
## Proportion 0.1 0.1 0.2 0.1 0.2 0.1 0.1 0.1
## -----
## Post_test
##      n missing distinct      Info      Mean      Gmd
##      10      0      6      0.964      59.7      3.311
##
## lowest : 56 59 60 61 62, highest: 59 60 61 62 63
##
## Value      56 59 60 61 62 63
## Frequency  3 1 1 2 1 2
## Proportion 0.3 0.1 0.1 0.2 0.1 0.2
## -----

# Pastecs
Pastecs_s <- stat.desc(Stude_scoresDF)
```

Pastecs_s

```
##           Student      Pre_test      Post_test
## nbr.val      10.0000000  10.00000000  10.00000000
## nbr.null      0.0000000   0.00000000   0.00000000
## nbr.na        0.0000000   0.00000000   0.00000000
## min           1.0000000  47.00000000  56.00000000
## max           10.0000000  63.00000000  63.00000000
## range         9.0000000  16.00000000   7.00000000
## sum           55.0000000 557.00000000 597.00000000
## median        5.5000000  56.00000000  60.50000000
## mean          5.5000000 55.70000000 59.70000000
## SE.mean       0.9574271   1.46855938   0.89504811
## CI.mean.0.95  2.1658506   3.32211213   2.02473948
## var           9.1666667  21.56666667   8.01111111
## std.dev       3.0276504   4.64399254   2.83039063
## coef.var      0.5504819   0.08337509   0.04741023
```

2. The Department of Agriculture was studying the effects of several levels of a fertilizer on the growth of a plant. For some analyses, it might be useful to convert the fertilizer levels to an ordered factor.

The data were 10,10,10, 20,20,50,10,20,10,50,20,50,20,10.

a. Write the codes and describe the result.

```
data1 <- c(10,10,10,20,20,50,10,20,10,50,20,50,20,10)
data1
```

```
## [1] 10 10 10 20 20 50 10 20 10 50 20 50 20 10
```

```
data1_orderedFactor <- factor(data1, ordered = TRUE)
data1_orderedFactor
```

```
## [1] 10 10 10 20 20 50 10 20 10 50 20 50 20 10
## Levels: 10 < 20 < 50
```

The given data were reorder or arranges the levels of the factors in ascending order.

3. Abdul Hassan, president of Floor Coverings Unlimited, has asked you to study the exercise levels undertaken by 10 subjects were “l”, “n”, “n”, “i”, “l”, “l”, “n”, “n”, “i”, “l”; n=none, l=light, i=intense

a. What is the best way to represent this in R?

```
ten_sub <- factor(c("l","n","n","i","l","l","n","n","i","l"),
                  c("n", "l", "i"), ordered = TRUE)
ten_sub
```

```
## [1] l n n i l l n n i l
## Levels: n < l < i
```

4. Sample of 30 tax accountants from all the states and territories of Australia and their individual state of origin is specified by a character vector of state mnemonics as:

```
state <- c("tas", "sa", "qld", "nsw", "nsw", "nt", "wa", "wa", "qld",
          "vic", "nsw", "vic", "qld", "qld", "sa", "tas", "sa", "nt",
          "wa", "vic", "qld", "nsw", "nsw", "wa", "sa", "act", "nsw",
          "vic", "vic", "act")

state

## [1] "tas" "sa" "qld" "nsw" "nsw" "nt" "wa" "wa" "qld" "vic" "nsw" "vic"
## [13] "qld" "qld" "sa" "tas" "sa" "nt" "wa" "vic" "qld" "nsw" "nsw" "wa"
## [25] "sa" "act" "nsw" "vic" "vic" "act"
```

a. Apply the factor function and factor level. Describe the results.

```
factorState <- factor(state)
factorState

## [1] tas sa qld nsw nsw nt wa wa qld vic nsw vic qld qld sa tas sa nt wa
## [20] vic qld nsw nsw wa sa act nsw vic vic act
## Levels: act nsw nt qld sa tas vic wa

levels(state)

## NULL

levels(factorState)

## [1] "act" "nsw" "nt" "qld" "sa" "tas" "vic" "wa"

# The given data was categorize and store it as levels. When using factor, It resulted to
# [1] tas sa qld nsw nsw nt wa wa qld vic nsw vic qld
# [14] qld sa tas sa nt wa vic qld nsw nsw wa sa act
# [27] nsw vic vic act
# When you directly check the levels of object state, it resulted as NULL.
# But when you check the levels using an already factored data, it resulted to
# [1] "act" "nsw" "nt" "qld" "sa" "tas" "vic" "wa"
```

5. From #4 - continuation:

```
incomes <- c(60, 49, 40, 61, 64, 60, 59, 54,
             62, 69, 70, 42, 56, 61, 61, 61, 58, 51, 48,
             65, 49, 49, 41, 48, 52, 46, 59, 46, 58, 43)

incomes
```

Suppose we have the incomes of the same tax accountants in another vector (insuitably large units of money)

```
## [1] 60 49 40 61 64 60 59 54 62 69 70 42 56 61 61 61 58 51 48 65 49 49 41 48 52
## [26] 46 59 46 58 43
```

a. Calculate the sample mean income for each state we can now use the special function `tapply()`:

```
incmeans <- tapply(incomes, factorState, mean)
```

Example: giving a means vector with the components labelled by the levels

Note: The function `tapply()` is used to apply a function, here `mean()`, to each group of components of the first argument, here `incomes`, defined by the levels of the second component, here `statef2`

2 that `tapply()` also works in this case when its second argument is not a factor,

e.g., `tapply(incomes, state)`, and this is true for quite a few other functions, since arguments are coerced to factors when necessary (using `as.factor()`).

b. Copy the results and interpret.

```
incmeans
##      act      nsw      nt      qld      sa      tas      vic      wa
## 44.50000 57.33333 55.50000 53.60000 55.00000 60.50000 56.00000 52.25000
# It creates a group summaries that was based on factor levels.
# The result was computes into a statistical measures.
```

6. Calculate the standard errors of the state income means (refer again to number 3)

```
stdError <- function(x) sqrt(var(x)/length(x))
```

```
incster <- tapply(incomes, factorState, stdError)
```

Note: After this assignment, the standard errors are calculated by:

a. What is the standard error? Write the codes.

```
incster <- tapply(incomes, factorState, stdError)
incster
##      act      nsw      nt      qld      sa      tas      vic      wa
## 1.500000 4.310195 4.500000 4.106093 2.738613 0.500000 5.244044 2.657536
```

b. Interpret the result.

```
# The result is a structure of the same length of the given factor.
# It shows all the standard deviation from act to wa.
```

7. Use the titanic dataset.

a. subset the titatic dataset of those who survived and not survived. Show the codes and its result.

```
data(Titanic)
Titanic <- data.frame(Titanic)
Titanic
##   Class  Sex  Age Survived Freq
## 1   1st  Male Child      No    0
## 2   2nd  Male Child      No    0
## 3   3rd  Male Child      No   35
## 4  Crew  Male Child      No    0
```

```
## 5    1st Female Child      No    0
## 6    2nd Female Child      No    0
## 7    3rd Female Child      No   17
## 8    Crew Female Child      No    0
## 9    1st   Male Adult      No  118
## 10   2nd   Male Adult      No  154
## 11   3rd   Male Adult      No  387
## 12   Crew   Male Adult      No  670
## 13   1st Female Adult      No    4
## 14   2nd Female Adult      No   13
## 15   3rd Female Adult      No   89
## 16   Crew Female Adult      No    3
## 17   1st   Male Child      Yes    5
## 18   2nd   Male Child      Yes   11
## 19   3rd   Male Child      Yes   13
## 20   Crew   Male Child      Yes    0
## 21   1st Female Child      Yes    1
## 22   2nd Female Child      Yes   13
## 23   3rd Female Child      Yes   14
## 24   Crew Female Child      Yes    0
## 25   1st   Male Adult      Yes   57
## 26   2nd   Male Adult      Yes   14
## 27   3rd   Male Adult      Yes   75
## 28   Crew   Male Adult      Yes  192
## 29   1st Female Adult      Yes  140
## 30   2nd Female Adult      Yes   80
## 31   3rd Female Adult      Yes   76
## 32   Crew Female Adult      Yes   20
```

```
subset_survive <- subset(Titanic, Survived == "Yes")
subset_survive
```

```
##      Class    Sex   Age Survived Freq
## 17    1st    Male Child      Yes    5
## 18    2nd    Male Child      Yes   11
## 19    3rd    Male Child      Yes   13
## 20   Crew    Male Child      Yes    0
## 21    1st Female Child      Yes    1
## 22    2nd Female Child      Yes   13
## 23    3rd Female Child      Yes   14
## 24   Crew Female Child      Yes    0
## 25    1st   Male Adult      Yes   57
## 26    2nd   Male Adult      Yes   14
## 27    3rd   Male Adult      Yes   75
## 28   Crew   Male Adult      Yes  192
## 29    1st Female Adult      Yes  140
## 30    2nd Female Adult      Yes   80
## 31    3rd Female Adult      Yes   76
## 32   Crew Female Adult      Yes   20
```

```
subset_died <- subset(Titanic, Survived == "No")
subset_died
```

```
##      Class    Sex   Age Survived Freq
## 1      1st    Male Child      No    0
```

```
## 2    2nd    Male Child      No    0
## 3    3rd    Male Child      No   35
## 4    Crew   Male Child      No    0
## 5    1st   Female Child     No    0
## 6    2nd   Female Child     No    0
## 7    3rd   Female Child     No   17
## 8    Crew  Female Child     No    0
## 9    1st    Male Adult     No  118
## 10   2nd    Male Adult     No  154
## 11   3rd    Male Adult     No  387
## 12   Crew   Male Adult     No  670
## 13   1st   Female Adult     No    4
## 14   2nd   Female Adult     No   13
## 15   3rd   Female Adult     No   89
## 16   Crew  Female Adult     No    3
```

8. The data sets are about the breast cancer Wisconsin. The samples arrive periodically as Dr. Wolberg reports his clinical cases. The database therefore reflects this chronological grouping of the data. You can create this dataset in Microsoft Excel.

```
library(mlbench)
data("BreastCancer")
Data_Breast_Cancer <- data.frame(BreastCancer)
Data_Breast_Cancer
```

```
##           Id Cl.thickness Cell.size Cell.shape Marg.adhesion Epith.c.size
## 1    1000025         5         1         1         1           2
## 2    1002945         5         4         4         5           7
## 3    1015425         3         1         1         1           2
## 4    1016277         6         8         8         1           3
## 5    1017023         4         1         1         3           2
## 6    1017122         8        10        10         8           7
## 7    1018099         1         1         1         1           2
## 8    1018561         2         1         2         1           2
## 9    1033078         2         1         1         1           2
## 10   1033078         4         2         1         1           2
## 11   1035283         1         1         1         1           1
## 12   1036172         2         1         1         1           2
## 13   1041801         5         3         3         3           2
## 14   1043999         1         1         1         1           2
## 15   1044572         8         7         5        10           7
## 16   1047630         7         4         6         4           6
## 17   1048672         4         1         1         1           2
## 18   1049815         4         1         1         1           2
## 19   1050670        10         7         7         6           4
## 20   1050718         6         1         1         1           2
## 21   1054590         7         3         2        10           5
## 22   1054593        10         5         5         3           6
## 23   1056784         3         1         1         1           2
## 24   1057013         8         4         5         1           2
## 25   1059552         1         1         1         1           2
## 26   1065726         5         2         3         4           2
## 27   1066373         3         2         1         1           1
## 28   1066979         5         1         1         1           2
```

## 29	1067444	2	1	1	1	2
## 30	1070935	1	1	3	1	2
## 31	1070935	3	1	1	1	1
## 32	1071760	2	1	1	1	2
## 33	1072179	10	7	7	3	8
## 34	1074610	2	1	1	2	2
## 35	1075123	3	1	2	1	2
## 36	1079304	2	1	1	1	2
## 37	1080185	10	10	10	8	6
## 38	1081791	6	2	1	1	1
## 39	1084584	5	4	4	9	2
## 40	1091262	2	5	3	3	6
## 41	1096800	6	6	6	9	6
## 42	1099510	10	4	3	1	3
## 43	1100524	6	10	10	2	8
## 44	1102573	5	6	5	6	10
## 45	1103608	10	10	10	4	8
## 46	1103722	1	1	1	1	2
## 47	1105257	3	7	7	4	4
## 48	1105524	1	1	1	1	2
## 49	1106095	4	1	1	3	2
## 50	1106829	7	8	7	2	4
## 51	1108370	9	5	8	1	2
## 52	1108449	5	3	3	4	2
## 53	1110102	10	3	6	2	3
## 54	1110503	5	5	5	8	10
## 55	1110524	10	5	5	6	8
## 56	1111249	10	6	6	3	4
## 57	1112209	8	10	10	1	3
## 58	1113038	8	2	4	1	5
## 59	1113483	5	2	3	1	6
## 60	1113906	9	5	5	2	2
## 61	1115282	5	3	5	5	3
## 62	1115293	1	1	1	1	2
## 63	1116116	9	10	10	1	10
## 64	1116132	6	3	4	1	5
## 65	1116192	1	1	1	1	2
## 66	1116998	10	4	2	1	3
## 67	1117152	4	1	1	1	2
## 68	1118039	5	3	4	1	8
## 69	1120559	8	3	8	3	4
## 70	1121732	1	1	1	1	2
## 71	1121919	5	1	3	1	2
## 72	1123061	6	10	2	8	10
## 73	1124651	1	3	3	2	2
## 74	1125035	9	4	5	10	6
## 75	1126417	10	6	4	1	3
## 76	1131294	1	1	2	1	2
## 77	1132347	1	1	4	1	2
## 78	1133041	5	3	1	2	2
## 79	1133136	3	1	1	1	2
## 80	1136142	2	1	1	1	3
## 81	1137156	2	2	2	1	1
## 82	1143978	4	1	1	2	2

## 83	1143978	5	2	1	1	2
## 84	1147044	3	1	1	1	2
## 85	1147699	3	5	7	8	8
## 86	1147748	5	10	6	1	10
## 87	1148278	3	3	6	4	5
## 88	1148873	3	6	6	6	5
## 89	1152331	4	1	1	1	2
## 90	1155546	2	1	1	2	3
## 91	1156272	1	1	1	1	2
## 92	1156948	3	1	1	2	2
## 93	1157734	4	1	1	1	2
## 94	1158247	1	1	1	1	2
## 95	1160476	2	1	1	1	2
## 96	1164066	1	1	1	1	2
## 97	1165297	2	1	1	2	2
## 98	1165790	5	1	1	1	2
## 99	1165926	9	6	9	2	10
## 100	1166630	7	5	6	10	5
## 101	1166654	10	3	5	1	10
## 102	1167439	2	3	4	4	2
## 103	1167471	4	1	2	1	2
## 104	1168359	8	2	3	1	6
## 105	1168736	10	10	10	10	10
## 106	1169049	7	3	4	4	3
## 107	1170419	10	10	10	8	2
## 108	1170420	1	6	8	10	8
## 109	1171710	1	1	1	1	2
## 110	1171710	6	5	4	4	3
## 111	1171795	1	3	1	2	2
## 112	1171845	8	6	4	3	5
## 113	1172152	10	3	3	10	2
## 114	1173216	10	10	10	3	10
## 115	1173235	3	3	2	1	2
## 116	1173347	1	1	1	1	2
## 117	1173347	8	3	3	1	2
## 118	1173509	4	5	5	10	4
## 119	1173514	1	1	1	1	4
## 120	1173681	3	2	1	1	2
## 121	1174057	1	1	2	2	2
## 122	1174057	4	2	1	1	2
## 123	1174131	10	10	10	2	10
## 124	1174428	5	3	5	1	8
## 125	1175937	5	4	6	7	9
## 126	1176406	1	1	1	1	2
## 127	1176881	7	5	3	7	4
## 128	1177027	3	1	1	1	2
## 129	1177399	8	3	5	4	5
## 130	1177512	1	1	1	1	10
## 131	1178580	5	1	3	1	2
## 132	1179818	2	1	1	1	2
## 133	1180194	5	10	8	10	8
## 134	1180523	3	1	1	1	2
## 135	1180831	3	1	1	1	3
## 136	1181356	5	1	1	1	2

## 137	1182404	4	1	1	1	2
## 138	1182410	3	1	1	1	2
## 139	1183240	4	1	2	1	2
## 140	1183246	1	1	1	1	1
## 141	1183516	3	1	1	1	2
## 142	1183911	2	1	1	1	2
## 143	1183983	9	5	5	4	4
## 144	1184184	1	1	1	1	2
## 145	1184241	2	1	1	1	2
## 146	1184840	1	1	3	1	2
## 147	1185609	3	4	5	2	6
## 148	1185610	1	1	1	1	3
## 149	1187457	3	1	1	3	8
## 150	1187805	8	8	7	4	10
## 151	1188472	1	1	1	1	1
## 152	1189266	7	2	4	1	6
## 153	1189286	10	10	8	6	4
## 154	1190394	4	1	1	1	2
## 155	1190485	1	1	1	1	2
## 156	1192325	5	5	5	6	3
## 157	1193091	1	2	2	1	2
## 158	1193210	2	1	1	1	2
## 159	1193683	1	1	2	1	3
## 160	1196295	9	9	10	3	6
## 161	1196915	10	7	7	4	5
## 162	1197080	4	1	1	1	2
## 163	1197270	3	1	1	1	2
## 164	1197440	1	1	1	2	1
## 165	1197510	5	1	1	1	2
## 166	1197979	4	1	1	1	2
## 167	1197993	5	6	7	8	8
## 168	1198128	10	8	10	10	6
## 169	1198641	3	1	1	1	2
## 170	1199219	1	1	1	2	1
## 171	1199731	3	1	1	1	2
## 172	1199983	1	1	1	1	2
## 173	1200772	1	1	1	1	2
## 174	1200847	6	10	10	10	8
## 175	1200892	8	6	5	4	3
## 176	1200952	5	8	7	7	10
## 177	1201834	2	1	1	1	2
## 178	1201936	5	10	10	3	8
## 179	1202125	4	1	1	1	2
## 180	1202812	5	3	3	3	6
## 181	1203096	1	1	1	1	1
## 182	1204242	1	1	1	1	2
## 183	1204898	6	1	1	1	2
## 184	1205138	5	8	8	8	5
## 185	1205579	8	7	6	4	4
## 186	1206089	2	1	1	1	1
## 187	1206695	1	5	8	6	5
## 188	1206841	10	5	6	10	6
## 189	1207986	5	8	4	10	5
## 190	1208301	1	2	3	1	2

## 191	1210963	10	10	10	8	6
## 192	1211202	7	5	10	10	10
## 193	1212232	5	1	1	1	2
## 194	1212251	1	1	1	1	2
## 195	1212422	3	1	1	1	2
## 196	1212422	4	1	1	1	2
## 197	1213375	8	4	4	5	4
## 198	1213383	5	1	1	4	2
## 199	1214092	1	1	1	1	2
## 200	1214556	3	1	1	1	2
## 201	1214966	9	7	7	5	5
## 202	1216694	10	8	8	4	10
## 203	1216947	1	1	1	1	2
## 204	1217051	5	1	1	1	2
## 205	1217264	1	1	1	1	2
## 206	1218105	5	10	10	9	6
## 207	1218741	10	10	9	3	7
## 208	1218860	1	1	1	1	1
## 209	1218860	1	1	1	1	1
## 210	1219406	5	1	1	1	1
## 211	1219525	8	10	10	10	5
## 212	1219859	8	10	8	8	4
## 213	1220330	1	1	1	1	2
## 214	1221863	10	10	10	10	7
## 215	1222047	10	10	10	10	3
## 216	1222936	8	7	8	7	5
## 217	1223282	1	1	1	1	2
## 218	1223426	1	1	1	1	2
## 219	1223793	6	10	7	7	6
## 220	1223967	6	1	3	1	2
## 221	1224329	1	1	1	2	2
## 222	1225799	10	6	4	3	10
## 223	1226012	4	1	1	3	1
## 224	1226612	7	5	6	3	3
## 225	1227210	10	5	5	6	3
## 226	1227244	1	1	1	1	2
## 227	1227481	10	5	7	4	4
## 228	1228152	8	9	9	5	3
## 229	1228311	1	1	1	1	1
## 230	1230175	10	10	10	3	10
## 231	1230688	7	4	7	4	3
## 232	1231387	6	8	7	5	6
## 233	1231706	8	4	6	3	3
## 234	1232225	10	4	5	5	5
## 235	1236043	3	3	2	1	3
## 236	1241232	3	1	4	1	2
## 237	1241559	10	8	8	2	8
## 238	1241679	9	8	8	5	6
## 239	1242364	8	10	10	8	6
## 240	1243256	10	4	3	2	3
## 241	1270479	5	1	3	3	2
## 242	1276091	3	1	1	3	1
## 243	1277018	2	1	1	1	2
## 244	128059	1	1	1	1	2

## 245	1285531	1	1	1	1	2
## 246	1287775	5	1	1	2	2
## 247	144888	8	10	10	8	5
## 248	145447	8	4	4	1	2
## 249	167528	4	1	1	1	2
## 250	169356	3	1	1	1	2
## 251	183913	1	2	2	1	2
## 252	191250	10	4	4	10	2
## 253	1017023	6	3	3	5	3
## 254	1100524	6	10	10	2	8
## 255	1116116	9	10	10	1	10
## 256	1168736	5	6	6	2	4
## 257	1182404	3	1	1	1	2
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## 259	1198641	3	1	1	1	2
## 260	242970	5	7	7	1	5
## 261	255644	10	5	8	10	3
## 262	263538	5	10	10	6	10
## 263	274137	8	8	9	4	5
## 264	303213	10	4	4	10	6
## 265	314428	7	9	4	10	10
## 266	1182404	5	1	4	1	2
## 267	1198641	10	10	6	3	3
## 268	320675	3	3	5	2	3
## 269	324427	10	8	8	2	3
## 270	385103	1	1	1	1	2
## 271	390840	8	4	7	1	3
## 272	411453	5	1	1	1	2
## 273	320675	3	3	5	2	3
## 274	428903	7	2	4	1	3
## 275	431495	3	1	1	1	2
## 276	432809	3	1	3	1	2
## 277	434518	3	1	1	1	2
## 278	452264	1	1	1	1	2
## 279	456282	1	1	1	1	2
## 280	476903	10	5	7	3	3
## 281	486283	3	1	1	1	2
## 282	486662	2	1	1	2	2
## 283	488173	1	4	3	10	4
## 284	492268	10	4	6	1	2
## 285	508234	7	4	5	10	2
## 286	527363	8	10	10	10	8
## 287	529329	10	10	10	10	10
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## 289	543558	6	1	3	1	4
## 290	555977	5	6	6	8	6
## 291	560680	1	1	1	1	2
## 292	561477	1	1	1	1	2
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## 294	601265	10	4	4	6	2
## 295	606140	1	1	1	1	2
## 296	606722	5	5	7	8	6
## 297	616240	5	3	4	3	4
## 298	61634	5	4	3	1	2

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## 305	653777	8	3	4	9	3
## 306	659642	10	8	4	4	4
## 307	666090	1	1	1	1	2
## 308	666942	1	1	1	1	2
## 309	667204	7	8	7	6	4
## 310	673637	3	1	1	1	2
## 311	684955	2	1	1	1	3
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## 314	693702	1	1	1	1	2
## 315	704097	1	1	1	1	1
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## 317	706426	5	5	5	2	5
## 318	709287	6	8	7	8	6
## 319	718641	1	1	1	1	5
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## 321	730881	7	6	3	2	5
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## 325	740492	1	1	1	1	2
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## 330	760239	10	4	6	4	5
## 331	76389	10	4	7	2	2
## 332	764974	5	1	1	1	2
## 333	770066	5	2	2	2	2
## 334	785208	5	4	6	6	4
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## 337	797327	6	5	5	8	4
## 338	798429	1	1	1	1	2
## 339	704097	1	1	1	1	1
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## 341	809912	10	3	3	1	2
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## 346	826923	1	1	1	1	2
## 347	830690	5	2	2	2	3
## 348	831268	1	1	1	1	1
## 349	832226	3	4	4	10	5
## 350	832567	4	2	3	5	3
## 351	836433	5	1	1	3	2
## 352	837082	2	1	1	1	2

## 353	846832	3	4	5	3	7
## 354	850831	2	7	10	10	7
## 355	855524	1	1	1	1	2
## 356	857774	4	1	1	1	3
## 357	859164	5	3	3	1	3
## 358	859350	8	10	10	7	10
## 359	866325	8	10	5	3	8
## 360	873549	10	3	5	4	3
## 361	877291	6	10	10	10	10
## 362	877943	3	10	3	10	6
## 363	888169	3	2	2	1	4
## 364	888523	4	4	4	2	2
## 365	896404	2	1	1	1	2
## 366	897172	2	1	1	1	2
## 367	95719	6	10	10	10	8
## 368	160296	5	8	8	10	5
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## 380	685977	5	3	4	1	4
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## 392	1151734	10	8	7	4	3
## 393	1156017	3	1	1	1	2
## 394	1158247	1	1	1	1	1
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## 396	1168278	3	1	1	1	2
## 397	1176187	3	1	1	1	2
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## 401	1211265	3	10	8	7	6
## 402	1213784	3	1	1	1	2
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## 408	1234554	1	1	1	1	2
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## 410	1237674	3	1	2	1	2
## 411	1238021	1	1	1	1	2
## 412	1238464	1	1	1	1	1
## 413	1238633	10	10	10	6	8
## 414	1238915	5	1	2	1	2
## 415	1238948	8	5	6	2	3
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## 418	1239967	1	1	1	1	2
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## 420	1253505	2	3	1	1	5
## 421	1255384	3	2	2	3	2
## 422	1257200	10	10	10	7	10
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## 434	1293439	3	2	2	3	2
## 435	1293439	6	9	7	5	5
## 436	1294562	10	8	10	1	3
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## 438	527337	4	1	1	1	2
## 439	558538	4	1	3	3	2
## 440	566509	5	1	1	1	2
## 441	608157	10	4	3	10	4
## 442	677910	5	2	2	4	2
## 443	734111	1	1	1	3	2
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## 450	1193544	5	7	9	8	6
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## 452	1202253	5	1	1	1	2
## 453	1227081	3	1	1	3	2
## 454	1230994	4	5	5	8	6
## 455	1238410	2	3	1	1	3
## 456	1246562	10	2	2	1	2
## 457	1257470	10	6	5	8	5
## 458	1259008	8	8	9	6	6
## 459	1266124	5	1	2	1	2
## 460	1267898	5	1	3	1	2

## 461	1268313	5	1	1	3	2
## 462	1268804	3	1	1	1	2
## 463	1276091	6	1	1	3	2
## 464	1280258	4	1	1	1	2
## 465	1293966	4	1	1	1	2
## 466	1296572	10	9	8	7	6
## 467	1298416	10	6	6	2	4
## 468	1299596	6	6	6	5	4
## 469	1105524	4	1	1	1	2
## 470	1181685	1	1	2	1	2
## 471	1211594	3	1	1	1	1
## 472	1238777	6	1	1	3	2
## 473	1257608	6	1	1	1	1
## 474	1269574	4	1	1	1	2
## 475	1277145	5	1	1	1	2
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## 478	1296263	4	1	1	1	2
## 479	1296593	5	2	1	1	2
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## 481	1301945	5	1	1	1	1
## 482	1302428	5	3	2	4	2
## 483	1318169	9	10	10	10	10
## 484	474162	8	7	8	5	5
## 485	787451	5	1	2	1	2
## 486	1002025	1	1	1	3	1
## 487	1070522	3	1	1	1	1
## 488	1073960	10	10	10	10	6
## 489	1076352	3	6	4	10	3
## 490	1084139	6	3	2	1	3
## 491	1115293	1	1	1	1	2
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## 494	1142706	5	10	10	10	6
## 495	1155967	5	1	2	10	4
## 496	1170945	3	1	1	1	1
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## 499	1204558	4	1	1	1	2
## 500	1217952	4	1	1	1	2
## 501	1224565	6	1	1	1	2
## 502	1238186	4	1	1	1	2
## 503	1253917	4	1	1	2	2
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## 505	1268766	1	1	1	1	2
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## 507	1286943	8	10	10	10	7
## 508	1295508	1	1	1	1	2
## 509	1297327	5	1	1	1	2
## 510	1297522	2	1	1	1	2
## 511	1298360	1	1	1	1	2
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## 521	333093	1	1	1	1	3
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## 524	672113	7	5	6	10	4
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## 529	807657	6	1	3	2	2
## 530	8233704	4	1	1	1	1
## 531	837480	7	4	4	3	4
## 532	867392	4	2	2	1	2
## 533	869828	1	1	1	1	1
## 534	1043068	3	1	1	1	2
## 535	1056171	2	1	1	1	2
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## 537	1113061	5	1	1	1	2
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## 541	1158157	5	1	1	1	2
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## 544	1184586	4	1	1	1	2
## 545	1186936	2	1	3	2	2
## 546	1197527	5	1	1	1	2
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## 548	1240603	2	1	1	1	1
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## 550	1241035	7	8	3	7	4
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## 557	1321264	5	2	2	2	1
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## 572	142932	7	6	10	5	3
## 573	183936	3	1	1	1	2
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## 576	385103	5	1	2	1	2
## 577	690557	5	1	1	1	2
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## 590	1272166	5	1	1	1	2
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## 593	1298484	10	3	4	5	3
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## 609	557583	5	10	10	10	10
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## 632	1235807	5	1	1	1	2
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## 655	1325159	3	1	1	1	2
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## 658	1333877	5	4	5	1	8
## 659	1334015	7	8	8	7	3
## 660	1334667	1	1	1	1	2
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## 668	1348851	3	1	1	1	2
## 669	1350319	5	7	4	1	6
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## 672	1353092	3	2	1	2	2
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## 675	1355260	1	1	1	1	2
## 676	1365075	4	1	4	1	2

## 677	1365328	1	1	2	1	2
## 678	1368267	5	1	1	1	2
## 679	1368273	1	1	1	1	2
## 680	1368882	2	1	1	1	2
## 681	1369821	10	10	10	10	5
## 682	1371026	5	10	10	10	4
## 683	1371920	5	1	1	1	2
## 684	466906	1	1	1	1	2
## 685	466906	1	1	1	1	2
## 686	534555	1	1	1	1	2
## 687	536708	1	1	1	1	2
## 688	566346	3	1	1	1	2
## 689	603148	4	1	1	1	2
## 690	654546	1	1	1	1	2
## 691	654546	1	1	1	3	2
## 692	695091	5	10	10	5	4
## 693	714039	3	1	1	1	2
## 694	763235	3	1	1	1	2
## 695	776715	3	1	1	1	3
## 696	841769	2	1	1	1	2
## 697	888820	5	10	10	3	7
## 698	897471	4	8	6	4	3
## 699	897471	4	8	8	5	4

##	Bare.nuclei	Bl.cromatin	Normal.nucleoli	Mitoses	Class
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## 3	2	3	1	1	benign
## 4	4	3	7	1	benign
## 5	1	3	1	1	benign
## 6	10	9	7	1	malignant
## 7	10	3	1	1	benign
## 8	1	3	1	1	benign
## 9	1	1	1	5	benign
## 10	1	2	1	1	benign
## 11	1	3	1	1	benign
## 12	1	2	1	1	benign
## 13	3	4	4	1	malignant
## 14	3	3	1	1	benign
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## 16	1	4	3	1	malignant
## 17	1	2	1	1	benign
## 18	1	3	1	1	benign
## 19	10	4	1	2	malignant
## 20	1	3	1	1	benign
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## 22	7	7	10	1	malignant
## 23	1	2	1	1	benign
## 24	<NA>	7	3	1	malignant
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## 26	7	3	6	1	malignant
## 27	1	2	1	1	benign
## 28	1	2	1	1	benign
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## 30	1	1	1	1	benign

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## 34	1	3	1	1	benign
## 35	1	2	1	1	benign
## 36	1	2	1	1	benign
## 37	1	8	9	1	malignant
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## 41	<NA>	7	8	1	benign
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## 45	1	8	10	1	malignant
## 46	1	2	1	2	benign
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## 51	3	2	1	5	malignant
## 52	4	3	4	1	malignant
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## 54	8	7	3	7	malignant
## 55	8	7	1	1	malignant
## 56	5	3	6	1	malignant
## 57	6	3	9	1	malignant
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## 162	1	3	2	1	benign
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## 165	<NA>	3	1	1	benign
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## 271	10	3	9	2 malignant
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## 275	1	3	2	1 benign
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## 277	1	2	1	1 benign
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## 281	1	3	1	1 benign
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## 284	10	5	3	1 malignant
## 285	10	3	8	2 malignant
## 286	10	10	7	3 malignant
## 287	10	4	10	10 malignant
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## 289	5	5	10	1 malignant
## 290	10	4	10	4 malignant
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## 292	1	3	1	1 benign
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## 294	10	2	3	1 malignant
## 295	<NA>	2	1	1 benign
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## 301	4	7	10	1 malignant
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## 317	10	4	3	1 malignant
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## 417	10	7	2	1	malignant
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## 520	10	9	1	1	malignant
## 521	1	1	1	1	benign
## 522	1	1	1	1	benign
## 523	5	7	3	1	malignant
## 524	10	5	3	1	malignant
## 525	1	2	1	1	benign
## 526	1	1	1	1	benign
## 527	1	1	1	1	benign
## 528	1	3	1	1	benign
## 529	1	1	1	1	benign
## 530	1	2	1	1	benign
## 531	10	6	9	1	malignant
## 532	1	2	1	1	benign
## 533	1	3	1	1	benign
## 534	1	2	1	1	benign
## 535	1	2	1	1	benign
## 536	1	3	1	1	benign
## 537	1	3	1	1	benign
## 538	1	3	1	1	benign
## 539	1	2	1	1	benign
## 540	1	2	1	1	benign
## 541	2	2	1	1	benign
## 542	1	1	1	1	benign
## 543	1	1	1	1	benign
## 544	1	2	1	1	benign
## 545	1	2	1	1	benign
## 546	1	2	1	1	benign
## 547	10	7	10	1	malignant
## 548	1	1	1	1	benign
## 549	1	1	1	1	benign
## 550	5	7	8	2	malignant
## 551	1	2	1	1	benign
## 552	1	3	1	1	benign
## 553	1	4	2	1	benign
## 554	5	2	1	2	benign
## 555	1	1	1	1	benign
## 556	1	4	8	1	benign
## 557	1	2	1	1	benign
## 558	1	1	1	1	benign
## 559	1	2	1	1	benign
## 560	1	2	1	1	benign
## 561	1	3	1	1	benign
## 562	1	3	1	1	benign
## 563	1	3	1	1	benign
## 564	1	2	1	1	benign
## 565	1	3	2	1	benign
## 566	10	10	10	1	malignant
## 567	1	3	1	1	benign
## 568	3	2	1	1	benign
## 569	10	2	5	2	malignant
## 570	5	10	3	1	malignant

## 571	10	8	2	1 malignant
## 572	10	9	10	2 malignant
## 573	1	2	1	1 benign
## 574	1	2	1	1 benign
## 575	2	7	7	1 malignant
## 576	1	3	1	1 benign
## 577	1	2	1	1 benign
## 578	1	2	1	1 benign
## 579	1	2	1	1 benign
## 580	1	3	1	1 benign
## 581	1	2	1	1 benign
## 582	10	7	5	1 malignant
## 583	10	6	10	1 malignant
## 584	1	1	1	1 benign
## 585	1	1	1	1 benign
## 586	1	1	1	1 benign
## 587	10	10	10	1 malignant
## 588	1	2	2	1 benign
## 589	3	4	1	1 malignant
## 590	1	1	1	1 benign
## 591	1	10	1	1 malignant
## 592	10	7	6	1 malignant
## 593	10	4	1	1 malignant
## 594	1	1	1	1 benign
## 595	10	7	1	1 malignant
## 596	1	2	1	1 benign
## 597	1	2	1	1 benign
## 598	1	3	1	1 benign
## 599	1	2	1	1 benign
## 600	1	1	1	1 benign
## 601	1	2	1	1 benign
## 602	1	2	1	1 benign
## 603	1	2	1	1 benign
## 604	1	8	10	1 malignant
## 605	10	8	1	2 malignant
## 606	8	7	8	3 malignant
## 607	1	1	1	1 benign
## 608	1	1	1	1 benign
## 609	10	10	1	1 malignant
## 610	1	1	1	1 benign
## 611	10	7	1	2 malignant
## 612	2	8	5	1 malignant
## 613	10	10	10	10 malignant
## 614	1	2	1	1 benign
## 615	1	2	1	1 benign
## 616	1	2	1	1 benign
## 617	1	2	1	1 benign
## 618	<NA>	1	1	1 benign
## 619	1	2	1	1 benign
## 620	1	2	1	1 benign
## 621	1	2	1	1 benign
## 622	2	6	1	1 benign
## 623	1	2	1	1 benign
## 624	1	1	1	1 benign

## 625	1	2	1	1	benign
## 626	4	1	1	1	benign
## 627	6	7	7	3	malignant
## 628	5	1	1	1	benign
## 629	1	1	1	1	benign
## 630	1	1	1	1	benign
## 631	1	1	1	1	benign
## 632	1	2	1	1	benign
## 633	1	1	1	1	benign
## 634	3	5	10	1	malignant
## 635	1	1	1	1	benign
## 636	1	1	1	1	benign
## 637	1	10	10	3	malignant
## 638	2	2	1	1	benign
## 639	1	1	1	1	benign
## 640	1	1	1	1	benign
## 641	1	1	1	1	benign
## 642	1	2	1	1	benign
## 643	1	2	1	1	benign
## 644	1	1	1	1	benign
## 645	1	1	1	1	benign
## 646	1	2	1	1	benign
## 647	1	1	1	1	benign
## 648	1	1	1	1	benign
## 649	2	10	10	10	malignant
## 650	1	2	1	1	benign
## 651	4	1	1	1	benign
## 652	1	2	1	1	benign
## 653	1	2	2	1	benign
## 654	1	2	1	1	benign
## 655	1	3	1	1	benign
## 656	1	2	1	1	benign
## 657	1	2	1	1	benign
## 658	1	3	6	1	benign
## 659	10	7	2	3	malignant
## 660	1	1	1	1	benign
## 661	1	2	1	1	benign
## 662	1	3	1	1	benign
## 663	1	2	1	1	benign
## 664	1	2	1	1	benign
## 665	1	2	1	1	benign
## 666	1	1	1	1	benign
## 667	1	1	1	2	benign
## 668	1	3	1	1	benign
## 669	1	7	10	3	malignant
## 670	5	7	10	1	malignant
## 671	8	7	4	1	malignant
## 672	1	3	1	1	benign
## 673	1	3	1	1	benign
## 674	1	1	1	1	benign
## 675	1	2	1	1	benign
## 676	1	1	1	1	benign
## 677	1	2	1	1	benign
## 678	1	1	1	1	benign


```
## 679      1      1      1      1      benign
## 680      1      1      1      1      benign
## 681     10     10     10     7 malignant
## 682     10      5      6      3 malignant
## 683      1      3      2      1      benign
## 684      1      1      1      1      benign
## 685      1      1      1      1      benign
## 686      1      1      1      1      benign
## 687      1      1      1      1      benign
## 688      1      2      3      1      benign
## 689      1      1      1      1      benign
## 690      1      1      1      8      benign
## 691      1      1      1      1      benign
## 692      5      4      4      1 malignant
## 693      1      1      1      1      benign
## 694      1      2      1      2      benign
## 695      2      1      1      1      benign
## 696      1      1      1      1      benign
## 697      3      8     10      2 malignant
## 698      4     10      6      1 malignant
## 699      5     10      4      1 malignant
```

a. describe what is the dataset all about.

*# According to r-project.org, The data were reported by Dr: Wolberg on the basis on his
clinical cases in studying breast cancer: The objective is to identify each of a number of benign or malignant*

b. Import the data from MS Excel. Copy the codes.

```
Breast_cancer_xlsx <- read_excel("/cloud/project/RWorksheet_JACULINA#7a/DATA_BREAST_CANCER.xlsx")
Breast_cancer_xlsx
```

```
## # A tibble: 49 x 11
##       Id CL. thickne~1 Cell ~2 Cell ~3 Marg.~4 Epith~5 Bare.~6 Bl. C~7 Norma~8
##       <dbl>         <dbl>   <dbl>   <dbl>   <dbl>   <dbl> <chr>         <dbl>   <dbl>
## 1 1000025           5         1         1         1         2 1           3         1
## 2 1002945           5         4         4         5         7 10          3         2
## 3 1015425           3         1         1         1         2 2           3         1
## 4 1016277           6         8         8         1         3 4           3         7
## 5 1017023           4         1         1         3         2 1           3         1
## 6 1017122           8        10        10         8         7 10          9         7
## 7 1018099           1         1         1         1         2 10          3         1
## 8 1018561           2         1         2         1         2 1           3         1
## 9 1033078           2         1         1         1         2 1           1         1
## 10 1033078          4         2         1         1         2 1           2         1
## # ... with 39 more rows, 2 more variables: Mitoses <dbl>, Class <chr>, and
## # abbreviated variable names 1: `CL. thickness`, 2: `Cell size`,
## # 3: `Cell Shape`, 4: `Marg. Adhesion`, 5: `Epith. C.size`,
## # 6: `Bare. Nuclei`, 7: `Bl. Cromatin`, 8: `Normal nucleoli`
```

c. Compute the descriptive statistics using different packages. Find the values of:

c.1 Standard error of the mean for clump thickness.

```
Standard_error <- function(x) sd(x)/sqrt(length(x))
SE_clump <- Standard_error(Breast_cancer_xlsx$`CL. thickness`)
SE_clump
```

```
## [1] 0.4092884
```

c.2 Coefficient of variability for Marginal Adhesion.

```
coe_var <- sd(Breast_cancer_xlsx$`Marg. Adhesion`) / mean(Breast_cancer_xlsx$`Marg. Adhesion`) * 100
coe_var
```

```
## [1] 97.67235
```

c.3 Number of null values of Bare Nuclei.

```
null_values1 <- sum(is.na(Breast_cancer_xlsx$`Bare. Nuclei`))
null_values1
```

```
## [1] 0
```

It has 2 NA but the given codes that I input resulted to zero (0) so instead I tried to used the subs

```
null_values2 <- subset(Breast_cancer_xlsx, `Bare. Nuclei` == "NA")
null_values2
```

```
## # A tibble: 2 x 11
##       Id CL. t~1 Cell ~2 Cell ~3 Marg.~4 Epith~5 Bare.~6 Bl. C~7 Norma~8 Mitoses
##       <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <chr> <dbl> <dbl> <dbl>
## 1 1.06e6      8      4      5      1      2 NA      7      3      1
## 2 1.10e6      6      6      6      9      6 NA      7      8      1
## # ... with 1 more variable: Class <chr>, and abbreviated variable names
## #   1: `CL. thickness`, 2: `Cell size`, 3: `Cell Shape`, 4: `Marg. Adhesion`,
## #   5: `Epith. C.size`, 6: `Bare. Nuclei`, 7: `Bl. Cromatin`,
## #   8: `Normal nucleoli`
```

c.4 Mean and standard deviation for Bland Chromatin

```
mean_BlandChromatin <- mean(Breast_cancer_xlsx$`Bl. Cromatin`)
mean_BlandChromatin
```

```
## [1] 3.836735
```

```
sd_BlandChromatin <- sd(Breast_cancer_xlsx$`Bl. Cromatin`)
sd_BlandChromatin
```

```
## [1] 2.085135
```

c.5 Confidence interval of the mean for Uniformity of Cell Shape

```
Confi.mean <- mean(Breast_cancer_xlsx$`Cell Shape`)
Confi.mean
```

```
## [1] 3.163265
```

```

Confi.n <- length(Breast_cancer_xlsx$`Cell Shape`)
Confi.sd <- sd(Breast_cancer_xlsx$`Cell Shape`)
Confi.se <- Confi.sd/sqrt(Confi.n)
Confi.se

## [1] 0.4158294

alpha = 0.05
degrees.freedom = Confi.n - 1
t.score = qt(p=alpha/2, df=degrees.freedom,lower.tail=F)
t.score

```

```

## [1] 2.010635

margin.error <- t.score * Confi.se
lower.bound <- Confi.mean - margin.error
upper.bound <- Confi.mean + margin.error
print(c(lower.bound,upper.bound))

```

```
## [1] 2.327184 3.999346
```

d. How many attributes?

```

attributes_BC <- attributes(Breast_cancer_xlsx)
attributes_BC

## $class
## [1] "tbl_df"      "tbl"        "data.frame"
##
## $row.names
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [26] 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49
##
## $names
## [1] "Id"          "CL. thickness" "Cell size"      "Cell Shape"
## [5] "Marg. Adhesion" "Epith. C.size" "Bare. Nuclei"   "Bl. Cromatin"
## [9] "Normal nucleoli" "Mitoses"       "Class"

```

It has 3 attributes, It shows the class, row.names, and names.

e. Find the percentage of respondents who are malignant. Interpret the results.

```

malignant1 <- subset(Breast_cancer_xlsx[c(1:49), c(11)])
malignant1

## # A tibble: 49 x 1
##   Class
##   <chr>
## 1 benign
## 2 benign
## 3 benign
## 4 benign
## 5 benign
## 6 malignant
## 7 benign
## 8 benign

```

```
## 9 benign
## 10 benign
## # ... with 39 more rows
```

```
pos.malignant <- subset(Breast_cancer_xlsx, Class == 'malignant')
pos.malignant
```

```
## # A tibble: 1 x 11
##       Id CL. t~1 Cell ~2 Cell ~3 Marg.~4 Epith~5 Bare.~6 Bl. C~7 Norma~8 Mitoses
##       <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl> <chr>     <dbl>   <dbl>   <dbl>
## 1 1.02e6      8      10      10      8      7 10      9      7      1
## # ... with 1 more variable: Class <chr>, and abbreviated variable names
## #   1: `CL. thickness`, 2: `Cell size`, 3: `Cell Shape`, 4: `Marg. Adhesion`,
## #   5: `Epith. C.size`, 6: `Bare. Nuclei`, 7: `Bl. Cromatin`,
## #   8: `Normal nucleoli`
```

```
# Percentage
malignant <- 17 / 49 * 100
malignant
```

```
## [1] 34.69388
```

```
# We can conclude that the data of were reported by Dr: Wolberg on the basis on his
# clinical cases in studying breast cancer
```

```
# Out of 47 respondents, there are 17 respondents who are malignant with corresponding of 34.69388 perc
```

9. Export the data abalone to the Microsoft excel file. Copy the codes.

```
install.packages("AppliedPredictiveModeling")
```

```
## Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.2'
## (as 'lib' is unspecified)
```

```
library("AppliedPredictiveModeling")
data(abalone)
# View(abalone)
head(abalone)
```

```
##   Type LongestShell Diameter Height WholeWeight ShuckedWeight VisceraWeight
## 1    M      0.455    0.365  0.095    0.5140      0.2245      0.1010
## 2    M      0.350    0.265  0.090    0.2255      0.0995      0.0485
## 3    F      0.530    0.420  0.135    0.6770      0.2565      0.1415
## 4    M      0.440    0.365  0.125    0.5160      0.2155      0.1140
## 5    I      0.330    0.255  0.080    0.2050      0.0895      0.0395
## 6    I      0.425    0.300  0.095    0.3515      0.1410      0.0775
##   ShellWeight Rings
## 1      0.150    15
## 2      0.070     7
## 3      0.210     9
## 4      0.155    10
## 5      0.055     7
## 6      0.120     8
```

```
summary(abalone)
```

```
##   Type      LongestShell      Diameter      Height      WholeWeight
## F:1307   Min.    :0.075   Min.    :0.0550   Min.    :0.0000   Min.    :0.0020
```

```

## I:1342  1st Qu.:0.450  1st Qu.:0.3500  1st Qu.:0.1150  1st Qu.:0.4415
## M:1528  Median :0.545  Median :0.4250  Median :0.1400  Median :0.7995
##          Mean  :0.524  Mean  :0.4079  Mean  :0.1395  Mean  :0.8287
##          3rd Qu.:0.615  3rd Qu.:0.4800  3rd Qu.:0.1650  3rd Qu.:1.1530
##          Max.   :0.815  Max.   :0.6500  Max.   :1.1300  Max.   :2.8255
## ShuckedWeight  VisceraWeight      ShellWeight      Rings
## Min.   :0.0010  Min.   :0.0005  Min.   :0.0015  Min.   : 1.000
## 1st Qu.:0.1860  1st Qu.:0.0935  1st Qu.:0.1300  1st Qu.: 8.000
## Median :0.3360  Median :0.1710  Median :0.2340  Median : 9.000
## Mean   :0.3594  Mean   :0.1806  Mean   :0.2388  Mean   : 9.934
## 3rd Qu.:0.5020  3rd Qu.:0.2530  3rd Qu.:0.3290  3rd Qu.:11.000
## Max.   :1.4880  Max.   :0.7600  Max.   :1.0050  Max.   :29.000
write.xlsx(abalone, "abalone.xlsx")

```