Worksheet #7a

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##R Markdown

1. Create a data frame for the table below.

```
Student_Scores <- data.frame(
  Student = c(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)
)
Student_Scores
```

```
##    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.

```
library(Hmisc)
```

```
## Warning: package 'Hmisc' was built under R version 4.2.2

## Loading required package: lattice

## Loading required package: survival

## Loading required package: Formula

## Loading required package: ggplot2

## Warning: package 'ggplot2' was built under R version 4.2.2
```

```
##
## Attaching package: 'Hmisc'

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

```
library(pastecs)
```

```
## Warning: package 'pastecs' was built under R version 4.2.2
```

```
describe(Student_Scores)
```

```
## Student_Scores
##
##  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
## ---------------------------------------------------------------------------
```

```
stat.desc(Student_Scores)
```

```
##                 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.

a. the data were 10,10,10, 20,20,50,10,20,10,50,20,50,20,10.

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

fertilizer_factor <- factor(fertilizer, ordered = TRUE)
fertilizer_factor
```

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

3. Abdul Hassan, president of Floor Coverings Unlimited, has asked you to study the exercise levels undertaken by 10 num3 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?

```
exercise_levels <- c("l","n","n","i","l","l","n","n","i","l")

data.frame(exercise_levels)
```

```
##    exercise_levels
## 1                l
## 2                n
## 3                n
## 4                i
## 5                l
## 6                l
```

```
## 7               n
## 8               n
## 9               i
## 10              l
```

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")
```

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

```
australia <- factor(state)
australia
```

```
##  [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
```

5.From #4 - continuation: Suppose we have the incomes of the same tax accountants in another vector (in suitably large units of money)

```
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)
```

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

```
sample_5 <- tapply(incomes, state, mean)
sample_5
```

```
##      act      nsw       nt      qld       sa      tas      vic       wa
## 44.50000 57.33333 55.50000 53.60000 55.00000 60.50000 56.00000 52.25000
```

b.Copy the results and interpret

```
#results
#act       nsw       nt      qld       sa      tas      vic       wa
#44.50000  57.33333 55.50000 53.60000 55.00000 60.50000 56.00000 52.25000
#the results shows the mean of every level of income for each state.
```

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

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

```
standard_err.n <- length(sample_5)
standard_err.sd <- sd(sample_5)
standard_err.se <- standard_err.sd/sqrt(standard_err.n)
standard_err.se
```

```
## [1] 1.653911
```

b. Interpret the result

```
#This is how I get the state income means by dividing the sd() to sqrt() or length()
#and that is how I get the standard errors of the state income means and this was
#the result.
```

7. Use the titanic dataset

a.Subset the titanic dataset of those who survived and not survived. Show the codes and its result.

```
data("Titanic")
Titanic <- data.frame(Titanic)

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

```
##    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
```

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

```
##    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.

a.describe what is the dataset all about.

```
#Answer: The dataset is all about the breast cancer Wisconsin.
```

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

```
library("readxl")
```

```
## Warning: package 'readxl' was built under R version 4.2.2
```

```
Breast_cancer <- read_excel("E:/CS 101/Worksheet 7a/Breast_Cancer.xlsx")
```

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

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

```
Breast_cancerc1.n <- length(Breast_cancer$'CL. thickness')
Breast_cancerc1.sd <- sd(Breast_cancer$'CL. thickness')
Breast_cancerc1.se <- Breast_cancerc1.sd/sqrt(Breast_cancer$'CL. thickness')
Breast_cancerc1.se
```

```
##  [1] 1.2812754 1.2812754 1.6541194 1.1696391 1.4325095 1.0129371 2.8650189
##  [8] 2.0258743 2.0258743 1.4325095 2.8650189 2.0258743 1.2812754 2.8650189
## [15] 1.0129371 1.0828754 1.4325095 1.4325095 0.9059985 1.1696391 1.0828754
## [22] 0.9059985 1.6541194 1.0129371 2.8650189 1.2812754 1.6541194 1.2812754
## [29] 2.0258743 2.8650189 1.6541194 2.0258743 0.9059985 2.0258743 1.6541194
## [36] 2.0258743 0.9059985 1.1696391 1.2812754 2.0258743 1.1696391 0.9059985
## [43] 1.1696391 1.2812754 0.9059985 2.8650189 1.6541194 2.8650189 1.4325095
```

c.2 Coefficient of variability for Marginal Adhesion.

```
sd(Breast_cancer$`Marg. Adhesion`) / mean(Breast_cancer$`Marg. Adhesion`) * 100
```

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

c.3 Number of null values of Bare Nuclei

```
Breast_cancerc3 <- subset(Breast_cancer,`Bare. Nuclei` == "NA")
Breast_cancerc3
```

```
## # 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(Breast_cancer$'Bl. Cromatin')
```

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

```
sd(Breast_cancer$'Bl. Cromatin')
```

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

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

```
breast_cancerc5 <- mean(Breast_cancer$`Cell Shape`)
breast_cancerc5
```

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

```
#Calculate the standard error of the mean
numA <- length(Breast_cancer$`Cell Shape`)
numB <- sd(Breast_cancer$`Cell Shape`)
numC <- numB/sqrt(numA)
numC
```

```
## [1] 0.4158294
```

```
#Find the t-score that corresponds to the confidence level
numD = 0.05
numE = numA - 1
numF = qt(p=numD/2, df=numE,lower.tail=F)
numF
```

```
## [1] 2.010635
```

```
#Constructing the confidence interval
numG <- numF * numC
numG
```

```
## [1] 0.836081
```

```r
#Lower
numH <- breast_cancerc5 - numG
numH
```

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

```r
#Upper
numI <- breast_cancerc5 + numG
numI
```

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

```r
c(numH,numI)
```

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

d. How many attributes?

```r
attributes(Breast_cancer)
```

```
## $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"
```

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

```r
cancer <- subset(Breast_cancer, Class == "malignant")
cancer
```

```
## # 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'
```

```r
17 / 49 * 100
```

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

```
#There are 34.69388 or 35% of respondents who are malignant.
```

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

```
library("AppliedPredictiveModeling")
```

```
## Warning: package 'AppliedPredictiveModeling' was built under R version 4.2.2
```

```
data(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
```

```
#exporting the data abalone

library(xlsx)
```

```
## Warning: package 'xlsx' was built under R version 4.2.2
```

```
write.xlsx("abalone","C:\\Users\\Quennie\\Documents\\abalone.xlsx")
```