

Assignment 2

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Assignment 2

You are asked to submit both the R Markdown file and its pdf output.

Q1. Using R, compute the following

$$\frac{0.35 - 0.3}{\sqrt{\frac{0.2*(1-0.4)}{50}}}$$

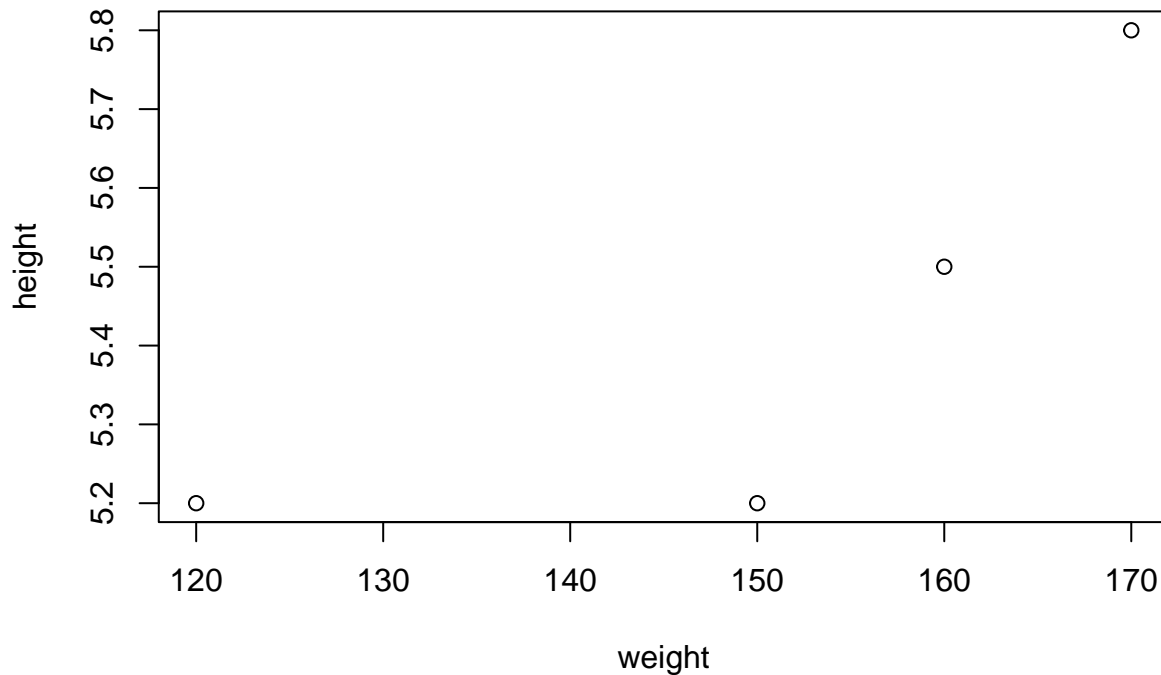
Answer

```
(0.35-0.3)/(sqrt((0.2*(1-0.4))/(50)))
```

```
## [1] 1.020621
```

Q2. Define two variables `weight = [150, 160, 170, 120]` and `height = [5.2, 5.5, 5.8, 5.2]` and plot weight vs height.

```
weight<-c(150,160,170,120)
height<-c(5.2,5.5,5.8,5.2)
plot(weight,height)
```



Q3. Without running any code, predict the outcome of each line:

```
x <- c(2, 3, 5, 7, 11, 13, 17, 19, 23, 29)
```

```
x[1:5]
```

```
x<-c(1,2,3,4,5)
```

```
x[c(1, 4)]
```

Select rows 1 and 4

```
x[-c(2, 5)]
```

Remove rows 2 and 5

Q4. Generate the matrix below

$$\begin{bmatrix} 1 & 4 & 7 & 3 \\ 2 & 4 & 3 & 8 \\ 3 & 2 & 1 & 5 \end{bmatrix}$$

```
G<-matrix(c(1,2,3,4,4,2,7,3,1,3,8,5),3,4)
G
```

```
##      [,1] [,2] [,3] [,4]
## [1,]    1    4    7    3
## [2,]    2    4    3    8
## [3,]    3    2    1    5
```

(a) Report the 2nd and the 3rd row.

```
G[c(2,3),]
```

```
##      [,1] [,2] [,3] [,4]
## [1,]    2    4    3    8
## [2,]    3    2    1    5
```

(b) Report all columns except the 2nd one.

```
G[,-2]
```

```
##      [,1] [,2] [,3]
## [1,]    1    7    3
## [2,]    2    3    8
## [3,]    3    1    5
```

(c) Rename row and column names to your names of choice.

```
rownames(G)<-c("lions","tigers","bears")
colnames(G)<-c("pink","purple","green","gold")
```

(d) Call the second and the third row using the names you defined.

```
G[c("tigers","bears"),]
```

```
##      pink purple green gold
## tigers    2     4     3     8
## bears     3     2     1     5
```

Q5. Create a dataframe with four features (columns), first is called `no_bedrooms`, `location`, `age`, `price`. Here is the info for five houses:

House 1: 4, 'Boston', 35, \$500K

House 2: 1, 'San Francisco', 55, \$900K

House 3: 4, 'Hartford', 87, \$300K

House 4: 3, 'Houston', 45, \$280K

House 5: 3, 'Seattle', 35, \$850K

```
no_bedrooms<-c(4,1,4,3,3)
location<-c("Boston","San Francisco","Hartford","Houston","Seattle")
age<-c(35,55,87,45,35)
price<-c(500000,900000,300000,280000,850000)
df<-data.frame(no_bedrooms,location,age,price)
df
```

```
##   no_bedrooms      location age  price
## 1           4         Boston  35 500000
## 2           1 San Francisco  55 900000
## 3           4         Hartford  87 300000
## 4           3          Houston  45 280000
## 5           3          Seattle  35 850000
```

Q6. Load the dataset that is already built-in data in R: `data(mtcars)`.

(a) How many observations are there in this dataset?

```
data(mtcars)
mtcars
```

```
##           mpg  cyl  disp  hp  drat    wt   qsec  vs  am  gear  carb
## Mazda RX4      21.0   6 160.0 110 3.90 2.620 16.46  0  1    4    4
## Mazda RX4 Wag  21.0   6 160.0 110 3.90 2.875 17.02  0  1    4    4
## Datsun 710      22.8   4 108.0  93 3.85 2.320 18.61  1  1    4    1
## Hornet 4 Drive  21.4   6 258.0 110 3.08 3.215 19.44  1  0    3    1
## Hornet Sportabout 18.7   8 360.0 175 3.15 3.440 17.02  0  0    3    2
## Valiant         18.1   6 225.0 105 2.76 3.460 20.22  1  0    3    1
## Duster 360      14.3   8 360.0 245 3.21 3.570 15.84  0  0    3    4
## Merc 240D       24.4   4 146.7  62 3.69 3.190 20.00  1  0    4    2
## Merc 230        22.8   4 140.8  95 3.92 3.150 22.90  1  0    4    2
## Merc 280        19.2   6 167.6 123 3.92 3.440 18.30  1  0    4    4
## Merc 280C       17.8   6 167.6 123 3.92 3.440 18.90  1  0    4    4
## Merc 450SE      16.4   8 275.8 180 3.07 4.070 17.40  0  0    3    3
## Merc 450SL      17.3   8 275.8 180 3.07 3.730 17.60  0  0    3    3
## Merc 450SLC     15.2   8 275.8 180 3.07 3.780 18.00  0  0    3    3
## Cadillac Fleetwood 10.4   8 472.0 205 2.93 5.250 17.98  0  0    3    4
## Lincoln Continental 10.4   8 460.0 215 3.00 5.424 17.82  0  0    3    4
## Chrysler Imperial 14.7   8 440.0 230 3.23 5.345 17.42  0  0    3    4
## Fiat 128        32.4   4  78.7  66 4.08 2.200 19.47  1  1    4    1
## Honda Civic     30.4   4  75.7  52 4.93 1.615 18.52  1  1    4    2
## Toyota Corolla  33.9   4  71.1  65 4.22 1.835 19.90  1  1    4    1
## Toyota Corona   21.5   4 120.1  97 3.70 2.465 20.01  1  0    3    1
## Dodge Challenger 15.5   8 318.0 150 2.76 3.520 16.87  0  0    3    2
## AMC Javelin     15.2   8 304.0 150 3.15 3.435 17.30  0  0    3    2
## Camaro Z28      13.3   8 350.0 245 3.73 3.840 15.41  0  0    3    4
## Pontiac Firebird 19.2   8 400.0 175 3.08 3.845 17.05  0  0    3    2
## Fiat X1-9       27.3   4  79.0  66 4.08 1.935 18.90  1  1    4    1
## Porsche 914-2   26.0   4 120.3  91 4.43 2.140 16.70  0  1    5    2
## Lotus Europa    30.4   4  95.1 113 3.77 1.513 16.90  1  1    5    2
## Ford Pantera L  15.8   8 351.0 264 4.22 3.170 14.50  0  1    5    4
## Ferrari Dino    19.7   6 145.0 175 3.62 2.770 15.50  0  1    5    6
## Maserati Bora   15.0   8 301.0 335 3.54 3.570 14.60  0  1    5    8
```

```
## Volvo 142E          21.4    4 121.0 109 4.11 2.780 18.60  1  1    4    2
```

```
str(mtcars)
```

```
## 'data.frame':    32 obs. of  11 variables:
## $ mpg : num  21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : num   6  6  4  6  8  6  8  4  4  6 ...
## $ disp: num  160 160 108 258 360 ...
## $ hp  : num  110 110  93 110 175 105 245  62  95 123 ...
## $ drat: num   3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
## $ wt  : num   2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num  16.5 17 18.6 19.4 17 ...
## $ vs  : num   0  0  1  1  0  1  0  1  1  1 ...
## $ am  : num   1  1  1  0  0  0  0  0  0  0 ...
## $ gear: num   4  4  4  3  3  3  3  4  4  4 ...
## $ carb: num   4  4  1  1  2  1  4  2  2  4 ...
```

32 observations (b) How many variables does this dataset have? What are the names of these variables?

11 variables: mpg,cyl,disp,hp,drat,wt,qsec,vs,am,gear,carb

(c) What are the classes of the variables?

```
class(mtcars$mpg)
```

```
## [1] "numeric"
```

```
class(mtcars$cyl)
```

```
## [1] "numeric"
```

```
class(mtcars$disp)
```

```
## [1] "numeric"
```

```
class(mtcars$hp)
```

```
## [1] "numeric"
```

```
class(mtcars$drat)
```

```
## [1] "numeric"
```

```
class(mtcars$wt)
```

```
## [1] "numeric"
```

```
class(mtcars$qsec)
```

```
## [1] "numeric"
```

(d) What are the statistics of mpg such as mean, median, standard deviation.

```
mean(mtcars$mpg)
```

```
## [1] 20.09062
```

```
median(mtcars$mpg)
```

```
## [1] 19.2
```

```
sd(mtcars$mpg)
```

```
## [1] 6.026948
```

(e) How many observations are with `drat > 3`?

```
log_drat <- (mtcars$drat > 3)
```

Q7. Start by loading the library and data:

```
library(dslabs)
data(murders)
murders
```

##		state	abb	region	population	total
## 1		Alabama	AL	South	4779736	135
## 2		Alaska	AK	West	710231	19
## 3		Arizona	AZ	West	6392017	232
## 4		Arkansas	AR	South	2915918	93
## 5		California	CA	West	37253956	1257
## 6		Colorado	CO	West	5029196	65
## 7		Connecticut	CT	Northeast	3574097	97
## 8		Delaware	DE	South	897934	38
## 9	District of	Columbia	DC	South	601723	99
## 10		Florida	FL	South	19687653	669
## 11		Georgia	GA	South	9920000	376
## 12		Hawaii	HI	West	1360301	7
## 13		Idaho	ID	West	1567582	12
## 14		Illinois	IL	North Central	12830632	364
## 15		Indiana	IN	North Central	6483802	142
## 16		Iowa	IA	North Central	3046355	21
## 17		Kansas	KS	North Central	2853118	63
## 18		Kentucky	KY	South	4339367	116
## 19		Louisiana	LA	South	4533372	351
## 20		Maine	ME	Northeast	1328361	11
## 21		Maryland	MD	South	5773552	293
## 22		Massachusetts	MA	Northeast	6547629	118
## 23		Michigan	MI	North Central	9883640	413
## 24		Minnesota	MN	North Central	5303925	53
## 25		Mississippi	MS	South	2967297	120
## 26		Missouri	MO	North Central	5988927	321
## 27		Montana	MT	West	989415	12
## 28		Nebraska	NE	North Central	1826341	32
## 29		Nevada	NV	West	2700551	84
## 30		New Hampshire	NH	Northeast	1316470	5
## 31		New Jersey	NJ	Northeast	8791894	246
## 32		New Mexico	NM	West	2059179	67
## 33		New York	NY	Northeast	19378102	517
## 34		North Carolina	NC	South	9535483	286
## 35		North Dakota	ND	North Central	672591	4
## 36		Ohio	OH	North Central	11536504	310
## 37		Oklahoma	OK	South	3751351	111
## 38		Oregon	OR	West	3831074	36
## 39		Pennsylvania	PA	Northeast	12702379	457
## 40		Rhode Island	RI	Northeast	1052567	16
## 41		South Carolina	SC	South	4625364	207
## 42		South Dakota	SD	North Central	814180	8
## 43		Tennessee	TN	South	6346105	219
## 44		Texas	TX	South	25145561	805
## 45		Utah	UT	West	2763885	22

```
## 46      Vermont VT      Northeast      625741      2
## 47      Virginia VA      South      8001024      250
## 48      Washington WA      West      6724540      93
## 49      West Virginia WV      South      1852994      27
## 50      Wisconsin WI      North Central      5686986      97
## 51      Wyoming WY      West      563626      5
```

- (a) Compute the per 100,000 murder rate for each state and store it in an object called `murder_rate`. Then use logical operators to create a logical vector named `low` that tells us which entries of `murder_rate` are lower than 1.

```
murder_rate<-(murders$total)/(murders$population)*100000
murder_rate

## [1] 2.8244238 2.6751860 3.6295273 3.1893901 3.3741383 1.2924531
## [7] 2.7139722 4.2319369 16.4527532 3.3980688 3.7903226 0.5145920
## [13] 0.7655102 2.8369608 2.1900730 0.6893484 2.2081106 2.6732010
## [19] 7.7425810 0.8280881 5.0748655 1.8021791 4.1786225 0.9992600
## [25] 4.0440846 5.3598917 1.2128379 1.7521372 3.1104763 0.3798036
## [31] 2.7980319 3.2537239 2.6679599 2.9993237 0.5947151 2.6871225
## [37] 2.9589340 0.9396843 3.5977513 1.5200933 4.4753235 0.9825837
## [43] 3.4509357 3.2013603 0.7959810 0.3196211 3.1246001 1.3829942
## [49] 1.4571013 1.7056487 0.8871131

low<-murder_rate<1
low
```

```
## [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE
## [13] TRUE FALSE FALSE TRUE FALSE FALSE FALSE TRUE FALSE FALSE FALSE TRUE
## [25] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE TRUE FALSE
## [37] FALSE TRUE FALSE FALSE FALSE TRUE FALSE FALSE TRUE TRUE FALSE FALSE
## [49] FALSE FALSE TRUE
```

- (b) Now use the results from the previous exercise and the function which to determine the indices of `murder_rate` associated with values lower than 1.

```
which(low)

## [1] 12 13 16 20 24 30 35 38 42 45 46 51
```

- (c) Use the results from the previous exercise to report the names of the states with murder rates lower than 1.

```
murders$state[low]

## [1] "Hawaii"      "Idaho"      "Iowa"      "Maine"
## [5] "Minnesota"   "New Hampshire" "North Dakota" "Oregon"
## [9] "South Dakota" "Utah"      "Vermont"   "Wyoming"
```

- (d) Now extend the code from exercises 2 and 3 to report the states in the Northeast with murder rates lower than 1. Hint: use the previously defined logical vector `low` and the logical operator `&`.

```
murders$state[low&(murders$region=="Northeast")]

## [1] "Maine"      "New Hampshire" "Vermont"
```

- (e) In a previous exercise we computed the murder rate for each state and the average of these numbers. How many states are below the average?

```
mean(murder_rate)
```

```
## [1] 2.779125
```

```
murders$state[mean(murder_rate)]
```

```
## [1] "Alaska"
```

- (f) Use the match function to identify the states with abbreviations AK, MI, and IA. Hint: start by defining an index of the entries of murders\$abb that match the three abbreviations, then use the [operator to extract the states.

```
murders$abb
```

```
## [1] "AL" "AK" "AZ" "AR" "CA" "CO" "CT" "DE" "DC" "FL" "GA" "HI" "ID" "IL" "IN"
## [16] "IA" "KS" "KY" "LA" "ME" "MD" "MA" "MI" "MN" "MS" "MO" "MT" "NE" "NV" "NH"
## [31] "NJ" "NM" "NY" "NC" "ND" "OH" "OK" "OR" "PA" "RI" "SC" "SD" "TN" "TX" "UT"
## [46] "VT" "VA" "WA" "WV" "WI" "WY"
```

```
match(c("AK", "MI", "IA"), murders$abb)
```

```
## [1] 2 23 16
```

- (g) Use the %in% operator to create a logical vector that answers the question: which of the following are actual abbreviations: MA, ME, MI, MO, MU?

```
murders$abb%in%c("MA", "ME", "MI", "MO", "MU")
```

```
## [1] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [13] FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE FALSE TRUE TRUE FALSE
## [25] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [37] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [49] FALSE FALSE FALSE
```

- (h) Extend the code you used in exercise 7 to report the one entry that is not an actual abbreviation. Hint: use the ! operator, which turns FALSE into TRUE and vice versa, then which to obtain an index.

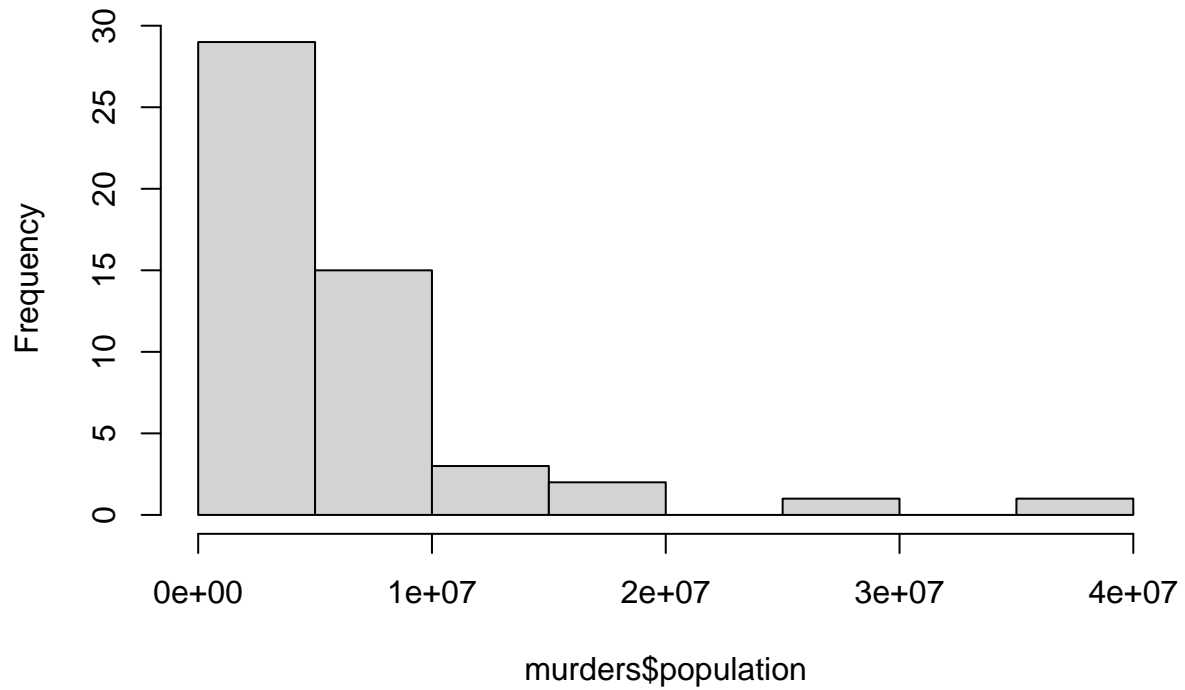
```
!murders$abb%in%c("MA", "ME", "MI", "MO", "MU")
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [13] TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE TRUE FALSE FALSE TRUE
## [25] TRUE FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [37] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [49] TRUE TRUE TRUE
```

- (i) Create a histogram of the state populations.

```
hist(murders$population)
```

Histogram of murders\$population



(j) Generate boxplots of the state populations by region.

```
boxplot(population~region,data=murders)
```

