

First load the file "MR\_data.csv" into R:

Input:

```
MR <- read.csv(file.choose(), header = TRUE)
head(MR)
```

Output:

```
> MR <- read.csv(file.choose(), header = TRUE)
> head(MR)
  InvoiceNumber InvoiceDate Division      ProjectType HoursRequired ClientFeedbackScore InvoiceAmount RepeatClient
1      176563    4/21/2018   Mobile      Web_Design        100.7             100       12856.62             0
2      176835    7/10/2018 Portland Market_Research        174.4             95       22822.32             1
3      177176    6/14/2018 Portland Market_Research         94.9             98       13489.30             1
4      177565    5/12/2018 Portland Customer_Survey         80.1             99       10122.28             1
5      177958   10/5/2018 Portland Market_Research         72.1             83       10640.40             0
6      178262    8/28/2018 Atlanta Social_Media_Analytics         67.6             73        4194.47             1
> |
```

**1) Which of the firm's divisions offered web design services in 2018 for their clients?**

Answer: The firm's divisions '**Mobile**' and '**Denver**' offered web design services in 2018 to their clients.

Input:

```
div <- subset(MR, MR$ProjectType == "Web_Design", c("Division"))
div
unique(div)
```

Output:

```
<
> div <- subset(MR, MR$ProjectType == "Web_Design", c("Division"))
> div
  Division
1      Mobile
10     Denver
11      Mobile
12      Mobile
16      Mobile
>
> unique(div)
  Division
1      Mobile
10     Denver
> |
```

**2) What is the average Client Feedback Score in the Atlanta division for projects which required more than 100 hours?**

Answer: The average Client Feedback Score in the Atlanta division for projects that required more than 100 hours, is **61**.

Input:

```
AT <- MR[MR$Division == 'Atlanta', ]
AT
AT_100 <- AT[AT$HoursRequired > 100, ]
```

AT\_100

mean(AT\_100\$ClientFeedbackScore)

Output:

```
> AT <- MR[MR$Division == 'Atlanta', ]
> AT
  InvoiceNumber InvoiceDate Division ProjectType HoursRequired ClientFeedbackScore InvoiceAmount RepeatClient
6      178262    8/28/2018  Atlanta Social_Media_Analytics      67.6              73      4194.470           1
23     182249    5/3/2018  Atlanta Social_Media_Analytics      80.1              68      6417.504           0
25     182784    2/17/2018  Atlanta Social_Media_Analytics      91.8              70      8750.409           0
41     187241    8/11/2018  Atlanta Social_Media_Analytics      90.5              60     12625.969           0
42     187428    3/31/2018  Atlanta Social_Media_Analytics      70.8              57      7385.005           0
57     190434   12/16/2018  Atlanta Social_Media_Analytics      88.7              70     10170.297           0
63     191761    2/26/2018  Atlanta Social_Media_Analytics      89.2              52      8929.196           0
66     192353    3/9/2018  Atlanta Social_Media_Analytics      89.2              68      9891.326           0
71     193635    6/6/2018  Atlanta Social_Media_Analytics      67.2              56      6140.590           0
99     200460    6/10/2018  Atlanta Social_Media_Analytics      76.5              61      6947.698           0
105    202102    8/30/2018  Atlanta Social_Media_Analytics      85.8              48      9916.879           1
121    205698   11/17/2018  Atlanta Social_Media_Analytics      62.6              54      5257.010           0
134    208329   10/30/2018  Atlanta Social_Media_Analytics      82.7              40      9631.106           0
139    209699    8/26/2018  Atlanta Social_Media_Analytics      75.2              53     10404.764           0
149    211866   10/9/2018  Atlanta Social_Media_Analytics      86.0              63      9304.126           1
182    220214    8/13/2018  Atlanta Social_Media_Analytics      90.4              55      7500.974           0
197    223772    4/1/2018  Atlanta Market_Research      133.9             57     14224.613           0
202    224751   12/15/2018  Atlanta Social_Media_Analytics      79.9              55      7138.371           0

>
> AT_100 <- AT[AT$HoursRequired > 100,]
> AT_100
  InvoiceNumber InvoiceDate Division ProjectType HoursRequired ClientFeedbackScore InvoiceAmount RepeatClient
197    223772    4/1/2018  Atlanta Market_Research      133.9             57     14224.613           0
330    257284    7/4/2018  Atlanta Market_Research      125.0             87     15709.106           1
458    289850    3/16/2018  Atlanta Market_Research      103.5             53      7757.702           0
462    290487    5/23/2018  Atlanta Market_Research      216.7             58     24085.390           0
463    290814   12/9/2018  Atlanta Social_Media_Analytics      110.6             50      9987.136           0

>
>
> mean(AT_100$ClientFeedbackScore)
[1] 61
>
```

**3) What is the range (max and min) and average hourly rate for market research projects across all divisions?**

Answer: For the market research projects across all divisions, the **range** (min and max) of hourly rate is **65.75 to 163.89**, while the **average** hourly rate is **132.73**

Input:

```
MR$HR <- (MR$InvoiceAmount / MR$HoursRequired)
head(MR)
Mini <- min(MR$HR[MR$ProjectType=='Market_Research'])
Mini
Maxi <- max(MR$HR[MR$ProjectType=='Market_Research'])
Maxi
range(MR$HR[MR$ProjectType=='Market_Research'])
AvgHR <- mean(MR$HR[MR$ProjectType=='Market_Research'])
AvgHR
```

Output:

```

> MR$HR <- (MR$InvoiceAmount / MR$HoursRequired)
> head(MR)
  InvoiceNumber InvoiceDate Division      ProjectType HoursRequired ClientFeedbackScore InvoiceAmount RepeatClient      HR
1      176563   4/21/2018   Mobile      Web_Design      100.7             100          12856.62           0 127.67251
2      176835   7/10/2018 Portland      Market_Research  174.4             95          22822.32           1 130.86190
3      177176   6/14/2018 Portland      Market_Research   94.9             98          13489.30           1 142.14226
4      177565   5/12/2018 Portland      Customer_Survey  80.1             99          10122.28           1 126.37048
5      177958  10/5/2018 Portland      Market_Research  72.1             83          10640.40           0 147.57835
6      178262   8/28/2018 Atlanta Social_Media_Analytics  67.6             73           4194.47           1  62.04837
> Mini <- min(MR$HR[MR$ProjectType=='Market_Research'])
> Mini
[1] 65.74583
> Maxi <- max(MR$HR[MR$ProjectType=='Market_Research'])
> Maxi
[1] 163.8851
> range(MR$HR[MR$ProjectType=='Market_Research'])
[1] 65.74583 163.88515
> AvgHR <- mean(MR$HR[MR$ProjectType=='Market_Research'])
> AvgHR
[1] 132.7266
>

```

**4) What was the invoice number and the invoice date for the market research project that had the lowest hourly rate?**

Answer: For the market research project that had the lowest hourly rate, the **invoice number** is **313125** and the **invoice date** is **11/2/2018**

Input:

```

MR$InvoiceNumber[MR$ProjectType == 'Market_Research' & MR$HR == Mini]
MR$InvoiceDate[MR$ProjectType == 'Market_Research' & MR$HR == Mini]

```

Output:

```

>
> MR$InvoiceNumber[MR$ProjectType == 'Market_Research' & MR$HR == Mini]
[1] 313125
> MR$InvoiceDate[MR$ProjectType == 'Market_Research' & MR$HR == Mini]
[1] "11/2/2018"
>

```

**5) An analyst at corporate headquarters took the projects from 2018 across all divisions and divided them into quintiles (bottom 20%, ..., top 20%) by Client Feedback Score. Which division had more projects in the bottom 20% Client Feedback Scores than any other division?**

Answer: The division which had most projects in the bottom 20% of Client Feedback Scores is **Mobile**, with score of 61.

Input:

```

Q <- quantile(MR$ClientFeedbackScore, probs = seq(0, 1, 1/5))
QuintTable <- cut(MR$ClientFeedbackScore, quantile(MR$ClientFeedbackScore, probs = seq(0, 1, 1/5)),
include.lowest = TRUE)
A <- table(QuintTable, MR$Division)
A
max(A[,])

```

Output:

```

>
> Q <- quantile(MR$ClientFeedbackScore, probs = seq(0, 1, 1/5))
> QuintTable <- cut(MR$ClientFeedbackScore, quantile(MR$ClientFeedbackScore, probs = seq(0, 1, 1/5)), include.lowest = TRUE)
> A <- table(QuintTable, MR$Division)
> A

QuintTable  Atlanta  Denver  Mobile  Portland
[11,72]      44      22      61      0
(72,84.2]     9      56      36      20
(84.2,91]     2      24      31      75
(91,97]       1      20      14      82
(97,100]      0       9      38      75

>
> max(A[, ])
[1] 61
>

```

**6) The analyst also created the table below, showing the distribution of each division's project Client Feedback Scores. For example, 30% of Portland's projects had scores between 97 and 100, which was the top 20% score group across all divisions. Submit the R code to recreate this table.**

	[11,72]	(72,84.2]	(84.2,91]	(91,97]	(97,100]
Atlanta	0.79	0.16	0.04	0.02	0.00
Denver	0.17	0.43	0.18	0.15	0.07
Mobile	0.34	0.20	0.17	0.08	0.21
Portland	0.00	0.08	0.30	0.33	0.30

Answer/ Input:

```

percent_CFS <- function(CFS_data, val_low, val_high) {
  num <- length(CFS_data)
  num_in_range <- CFS_data[CFS_data > val_low & CFS_data <= val_high]
  percent <- length(num_in_range)/num
  return(percent)
}

```

```

atlanta_CFS = MR[MR$Division == "Atlanta", c("ClientFeedbackScore")]
a1 <- percent_CFS(atlanta_CFS, 11, 72)
a2 <- percent_CFS(atlanta_CFS, 72, 84.2)
a3 <- percent_CFS(atlanta_CFS, 84.2, 91)
a4 <- percent_CFS(atlanta_CFS, 91, 97)
a5 <- percent_CFS(atlanta_CFS, 97, 100)
A <- round(c(a1, a2, a3, a4, a5), digits = 2)
A

```

```

denver_CFS = MR[MR$Division == "Denver", c("ClientFeedbackScore")]
d1 <- percent_CFS(denver_CFS, 11, 72)
d2 <- percent_CFS(denver_CFS, 72, 84.2)
d3 <- percent_CFS(denver_CFS, 84.2, 91)
d4 <- percent_CFS(denver_CFS, 91, 97)
d5 <- percent_CFS(denver_CFS, 97, 100)
D <- round(c(d1, d2, d3, d4, d5), digits = 2)
D

```

```
mobile_CFS = MR[MR$Division == "Mobile", c("ClientFeedbackScore")]
m1 <- percent_CFS(mobile_CFS, 11, 72)
m2 <- percent_CFS(mobile_CFS, 72, 84.2)
m3 <- percent_CFS(mobile_CFS, 84.2, 91)
m4 <- percent_CFS(mobile_CFS, 91, 97)
m5 <- percent_CFS(mobile_CFS, 97, 100)
M <- round(c(m1, m2, m3, m4, m5), digits = 2)
M
```

```
portland_CFS <- MR[MR$Division == "Portland", c("ClientFeedbackScore")]
p1 <- percent_CFS(portland_CFS, 11, 72)
p2 <- percent_CFS(portland_CFS, 72, 84.2)
p3 <- percent_CFS(portland_CFS, 84.2, 91)
p4 <- percent_CFS(portland_CFS, 91, 97)
p5 <- percent_CFS(portland_CFS, 97, 100)
P <- round(c(p1, p2, p3, p4, p5), digits = 2)
P
```

```
CFStable <- matrix(c(A, D, M, P), nrow = 4, byrow = TRUE)
```

```
rname <- c("Atlanta", "Denver", "Mobile", "Portland")
rownames(CFStable) <- rname
cname <- c("[11,72]", "(72,84.2]", "(84.2,91]", "(91-97]", "(97-100]")
colnames(CFStable) <- cname
```

CFStable

### Output:

```
>
> percent_CFS <- function(CFS_data, val_low, val_high) {
+   num <- length(CFS_data)
+   num_in_range <- CFS_data[CFS_data > val_low & CFS_data <= val_high]
+   percent <- length(num_in_range)/num
+   return(percent)
+ }
> atlanta_CFS = MR[MR$Division == "Atlanta", c("ClientFeedbackScore")]
> a1 <- percent_CFS(atlanta_CFS, 11, 72)
> a2 <- percent_CFS(atlanta_CFS, 72, 84.2)
> a3 <- percent_CFS(atlanta_CFS, 84.2, 91)
> a4 <- percent_CFS(atlanta_CFS, 91, 97)
> a5 <- percent_CFS(atlanta_CFS, 97, 100)
> A <- round(c(a1, a2, a3, a4, a5), digits = 2)
> A
[1] 0.79 0.16 0.04 0.02 0.00
> denver_CFS = MR[MR$Division == "Denver", c("ClientFeedbackScore")]
```

```

> p2 <- percent_CFS(portland_CFS, 72, 84.2)
> p3 <- percent_CFS(portland_CFS, 84.2, 91)
> p4 <- percent_CFS(portland_CFS, 91, 97)
> p5 <- percent_CFS(portland_CFS, 97, 100)
> P <- round(c(p1, p2, p3, p4, p5), digits = 2)
> P
[1] 0.00 0.08 0.30 0.33 0.30
> CFStable <- matrix(c(A, D, M, P), nrow = 4, byrow = TRUE)
> rname <- c("Atlanta", "Denver", "Mobile", "Portland")
> rownames(CFStable) <- rname
> cname <- c("[11,72]", "(72,84.2]", "(84.2,91]", "(91-97]", "(97-100]")
> colnames(CFStable) <- cname
> CFStable
      [11,72] (72,84.2] (84.2,91] (91-97] (97-100]
Atlanta    0.79      0.16      0.04      0.02      0.00
Denver     0.17      0.43      0.18      0.15      0.07
Mobile     0.33      0.20      0.17      0.08      0.21
Portland   0.00      0.08      0.30      0.33      0.30
>

```

## 7) Create a conditional box plot of ClientFeedbackScore by RepeatClient. How does the customer satisfaction of new clients compare to repeat clients?

Answer: Client Feedback Score of repeat clients varies much less than that of new clients and is also concentrated towards the higher end, thus can be considered to be more consistent and dependable to understand customer satisfaction.

Input:

```

any(is.na(MR))
library(lattice)
bwplot(MR$RepeatClient ~ MR$ClientFeedbackScore, dataset = MR, xlab = " Customer Satisfaction",
       ylab = "1: New client    2: Repeat client", col = "red")

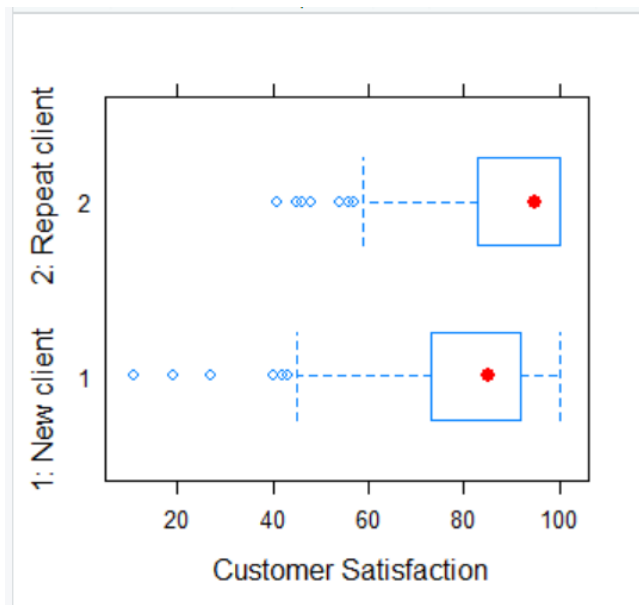
```

Output:

```

> any(is.na(MR))
[1] FALSE
> library(lattice)
> bwplot(MR$RepeatClient ~ MR$ClientFeedbackScore, dataset = MR, xlab = " Customer Satisfaction", ylab = "1:
New client    2: Repeat client", col = "blue")
>
> bwplot(MR$RepeatClient ~ MR$ClientFeedbackScore, dataset = MR, xlab = " Customer Satisfaction",
+       ylab = "1: New client    2: Repeat client", col = "red")
>

```



8) Create a visualization of Hours Required by Project Type. You may choose what type of chart to use, pick one that you think makes it easy to discern patterns in the data.

Answer/ Input:

```
any(is.na(MR))
```

```
library(ggplot2)
```

```
ggplot(MR, aes(x = HoursRequired, y = ProjectType)) + geom_boxplot(col = "red")
```

Output:

```
> any(is.na(MR))
[1] FALSE
> library(ggplot2)
> ggplot(MR, aes(x = HoursRequired, y = ProjectType)) + geom_boxplot(col = "red")
>
```

