

Prompt 1

1A Calculate the number of missing values in each column?

Code

```
Source on Save | 🔎 | ⚙️ | ⌂
1 ######Peer graded assignment #####
2 dd <- read.csv('cities_short.csv',
3                   stringsAsFactors = F,
4                   header = T)
5 colsums(is.na(dd))|
```

Output

```
> colsums(is.na(dd))
      x          Time   OperationType     Barcode
      0           0            0            0
CashierName    LineItem     Department   Category
      0           0            0            0
CardholderName RegisterName StoreNumber TransactionNumber
      22467        0            0            0
CustomerCode       Cost        Price   Quantity
      0           0            0            0
Modifiers       Subtotal   Discounts NetTotal
      0           0            0            0
Tax           TotalDue  TNumber_LItem     Year
      2835         2835            0            0
Month        WeekDay MonthDay      Hour
      0           0            0            0
> |
```

This is the output that we get after running the code. As we can see, there are three columns that have missing values. They are Cardholder Name, Tax and TotalDue.

1B Explain what approach would you use to handle them? (you may ignore missing values in a categorical/ factor variable)

Code

```
6 library(tidyr)
7 tax <- replace_na(dd$Tax, mean(dd$Tax, na.rm = TRUE))
8 sum(is.na(tax))
9 total_due <- replace_na(dd$TotalDue, median(dd$TotalDue, na.rm = TRUE))
10 sum(is.na(total_due))
11 library(dplyr)
12 dd_miss2 <- dd %>% filter(!is.na(Tax)) %>% filter(!is.na(cardholderName))
13 colsums(is.na(dd_miss2))|
```

Output

```
> sum(is.na(tax))
[1] 0
> View(dd)
> total_due <- replace_na(dd$TotalDue, median(dd$TotalDue, na.rm = TRUE))
> sum(is.na(total_due))
[1] 0
```

```

> dd_miss2 <- dd %>% filter(!is.na(Tax)) %>% filter(!is.na(cardholderName))
> colSums(is.na(dd_miss2))
      X          Time   operationType        BarCode
0       0           0                 0             0
CashierName    LineItem     Department    Category
0            0                 0             0
CardholderName RegisterName  storeNumber transactionNumber
0            0                 0             0
CustomerCode      Cost        Price      Quantity
0            0                 0             0
Modifiers       Subtotal   Discounts   NetTotal
0            0                 0             0
Tax          TotalDue  TNumber_LItem        Year
0            0                 0             0
Month        WeekDay   MonthDay        Hour
0            0                 0             0
> |

```

This is the final dataframe after we eliminate the missing values. As we can see, the missing values are 0 in Cardholder Name, Tax and TotalDue.

Prompt 2

2A Create a new variable LineItem_LongName coded as 1 if the length of LineItem is greater than the mean, otherwise 0.

Code

```
dd <- dd %>% mutate(LineItem_LongName = ifelse(nchar(LineItem) >
mean(nchar(LineItem)), 1, 0))
```

Output

LineItem	Year	Month	WeekDay	MonthDay	Hour	LineItem_LongName
JIR153361_Salmon and Wheat Bran Salad	2014	4	2	7	0	1
J2B121842_Pork and Pepper Kabob	2016	1	5	14	0	1
3L74243_Rice	2016	7	7	2	0	0
VNB173227_Rice	2014	7	4	9	0	0
GK114869_Salmon and Wheat Bran Salad	2014	3	2	3	0	1
7439700_Naan	2014	6	5	26	0	0
DAR91009_Beef and Apple Burgers	2015	10	4	14	0	1
GC35978_Salmon and Wheat Bran Salad	2015	6	1	7	0	1
NTJ60666_Salmon and Wheat Bran Salad	2014	6	5	12	0	1

After executing the above code, our dataframe will look like this. A new variable LineItem_LongName has been created and is coded 1 if length of LineItem is greater than mean and 0 otherwise.

2B Visualize relationship between Department and LineItem_LongName using the number of observations.

Code

```
15 cross_tab <- xtabs(~Department +LineItem_LongName, data = dd)
16 department <- rownames(cross_tab)
17 lineitem_longname <- colnames(cross_tab)
18 dd1 <- expand.grid(department,lineitem_longname)
19 Count <- as.vector(cross_tab)
20 dd1$Count <- Count
21 library(ggplot2)
22 p1 <- ggplot(data = dd1, aes(x=var1, y = factor(var2), size = Count))
23 p2 <- p1 + geom_point(col = "red") + labs(x = "Department", y = "LineItem_LongName")
24 p2
```

Output



Prompt 3

3A Name top 10 CashierName based on the total quantity sold (hint – arrange function can help you sort a column)

Code

```
dd2 <- dd %>% select(CashierName,Quantity) %>% arrange(desc(Quantity)) %>%
group_by(CashierName)
```

dd2

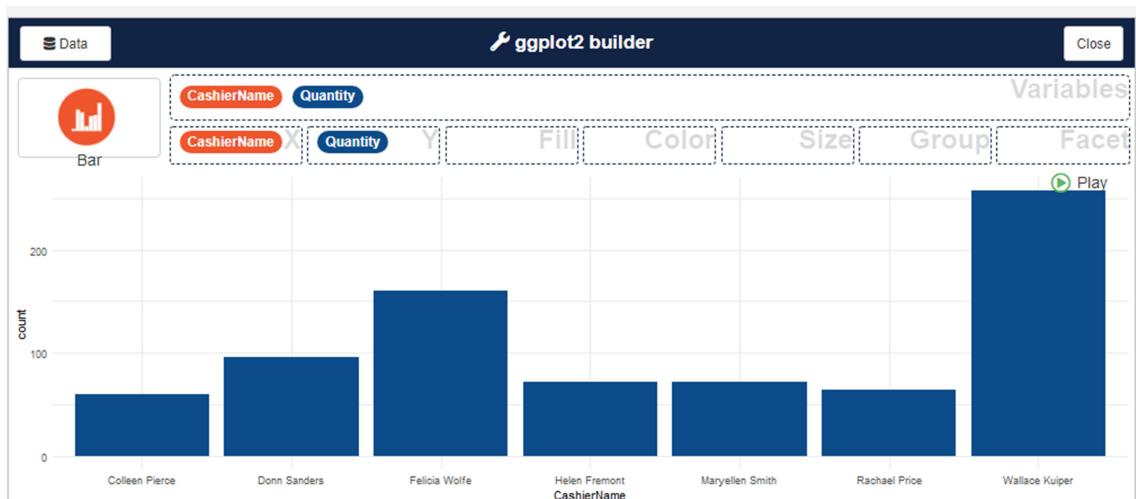
Output

```

# Groups.  CashierName [ ]
  CashierName      Quantity
  <chr>           <int>
  1 Wallace Kuiper    120
  2 Donn Sanders     96
  3 Felicia Wolfe   92
  4 Helen Fremont    72
  5 Wallace Kuiper    72
  6 Maryellen Smith   72
  7 Felicia Wolfe   68
  8 Wallace Kuiper    66
  9 Rachael Price    64
 10 Colleen Pierce   60

```

3B Visualize the 10 CashierNames and their total quantity sold



Prompt 4

4A Name bottom 5 LineItems based on average Price

Code

```
dd1 <- dd %>% filter(Price >= mean(Price)) %>% tail(dd, n = 5)
```

```
dd1$LineItem
```

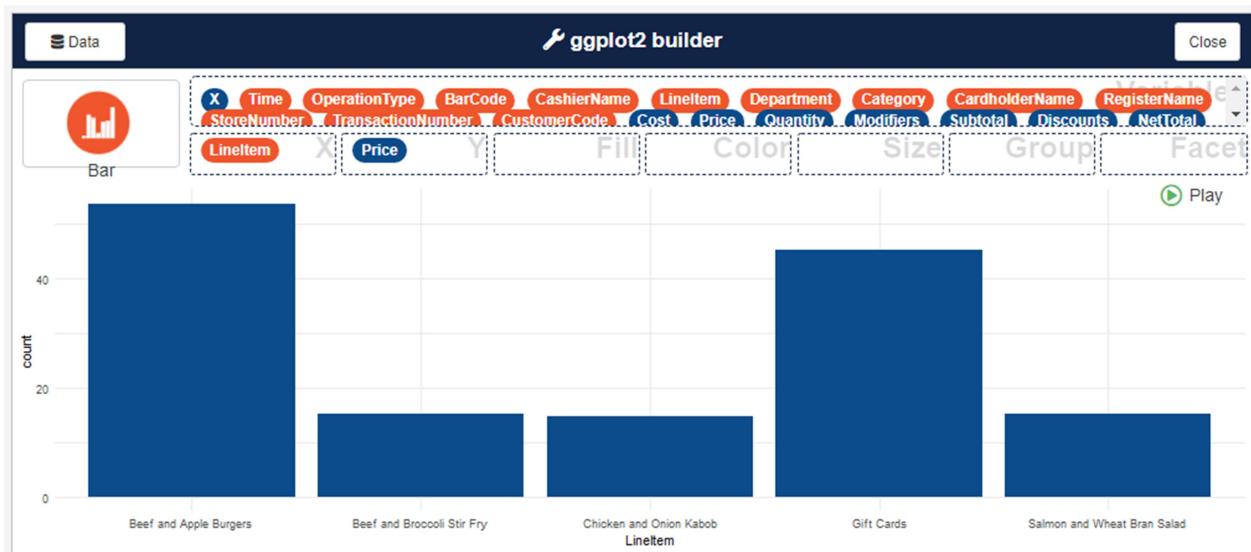
Output

```

> dd1$LineItem
[1] "Salmon and wheat Bran salad" "Beef and Broccoli stir Fry"
[3] "Beef and Apple Burgers"       "Chicken and Onion Kabob"
[5] "Gift Cards"
> |

```

4B Visualize the 5 LineItems and their average Price



Prompt 5

5A Transform the data into a structure where each unique row is a combination of Month and Department. The data must have columns on Average Price, Average Quantity and Average Cost.

Code

```

32 dd4 <- dd %>% select(Department, Month)
33 dd4$AvgPrice <- mean(dd$Price)
34 dd4$AvgQuantity <- mean(dd$Quantity)
35 dd4$AvgCost <- mean(dd$Cost)
36 dd4|
```

Output

	Department	Month	AvgPrice	AvgQuantity	AvgCost
1	Entrees	4	14.13681	1.182975	0.3416366
2	Kabobs	1	14.13681	1.182975	0.3416366
3	Sides	7	14.13681	1.182975	0.3416366
4	Sides	7	14.13681	1.182975	0.3416366
5	Entrees	3	14.13681	1.182975	0.3416366
6	Sides	6	14.13681	1.182975	0.3416366
7	Entrees	10	14.13681	1.182975	0.3416366
8	Entrees	6	14.13681	1.182975	0.3416366
9	Entrees	6	14.13681	1.182975	0.3416366
10	Entrees	10	14.13681	1.182975	0.3416366
11	Sides	9	14.13681	1.182975	0.3416366

Wing 1 to 12 of 80,000 entries, 5 total columns

5B In the transformed data frame, visualize the relationship between Average Price, Average Quantity and Average Cost

