## Assignment 4

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# 1. Visual quantification: To visualize the following quantity, mention what plot you will generate

### 1. Association between two variables when

a) both are numeric

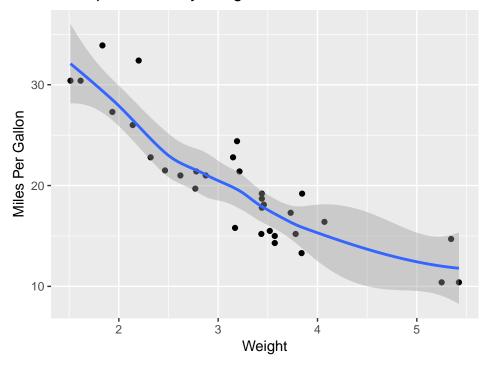
**Answer:** We would use a scatter plot.

#### Example:

```
data(mtcars)

ggplot(mtcars, aes(x=wt, y=mpg)) +
    geom_point() +
    geom_smooth(method="loess") +
    labs(title="Miles per Gallon by Weight", x="Weight", y="Miles Per Gallon")
```

## Miles per Gallon by Weight



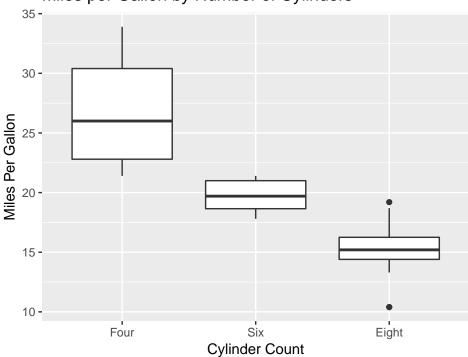
#### b) one numeric and one categorical

**Answer:** For this we could use a boxplot.

### Example:

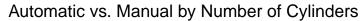
```
ggplot(mtcars, aes(x=factor(cyl), y=mpg)) +
  geom_boxplot() +
  labs(title="Miles per Gallon by Number of Cylinders", x="Weight", y="Miles Per Gallon") +
  scale_x_discrete("Cylinder Count", labels=c("Four", "Six", "Eight"))
```

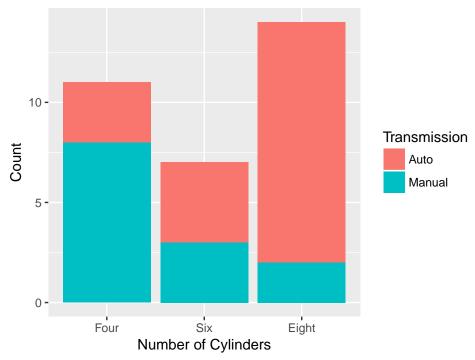
## Miles per Gallon by Number of Cylinders



#### c) both are categorical

Answer: A stacked barchart will work in this case.



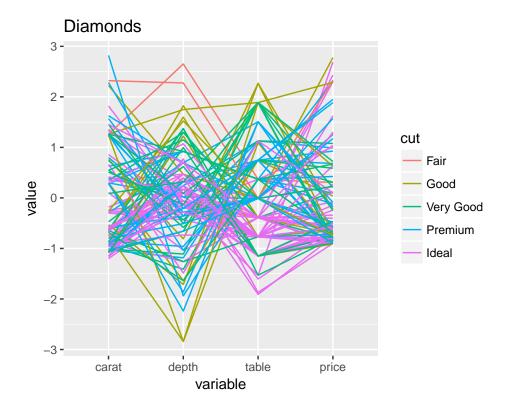


## Association of a categorical variable with many numerical variables together

**Answer:** For this we can use a parallel coordinate plot.

```
data(diamonds)
diamonds.samp <- diamonds[sample(1:dim(diamonds)[1], 100), ]

ggparcoord(data = diamonds.samp, columns = c(1, 5:7), groupColumn = 'cut') +
  labs(title='Diamonds')</pre>
```



## Difference of two variables in terms of

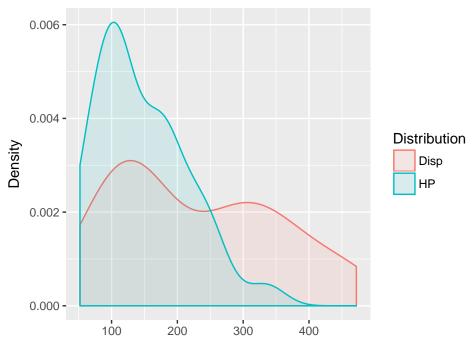
#### a) spread

**Answer:** Either a side-by-side boxplot or an overlayed density plot will work here.

## Example:

## Warning: Width not defined. Set with `position\_dodge(width = ?)`





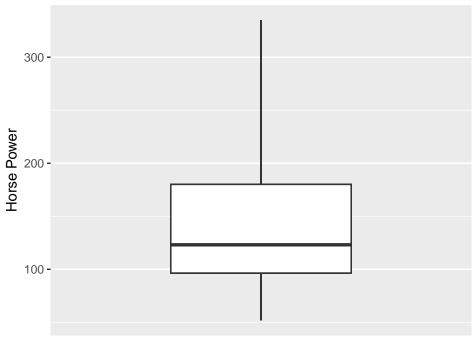
### b) center

**Answer:** A boxplot shows the median, or we can add a mean and median to a density plot.

**Example:** Here, we'll show both options.

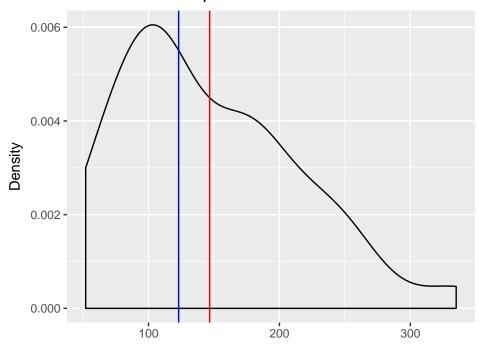
```
hp <- mtcars$hp
lower <- quantile(hp)[[2]]
upper <- quantile(hp)[[4]]
middle <- quantile(hp)[[3]]
ymin <- quantile(hp)[[1]]
ymax <- quantile(hp)[[5]]
ggplot(data.frame(hp), aes(hp, x=0)) +
    geom_boxplot(stat = "identity", aes(lower=lower, upper=upper, middle=middle, ymin=ymin, ymax=ymax)) +
    labs(title="Horse Power", x="", y="Horse Power") +
    scale_x_discrete(labels=(""))</pre>
```

## Horse Power



```
ggplot(mtcars, aes(hp, fill=hp)) +
  geom_density() +
  geom_vline(xintercept = mean(hp), col="red") +
  geom_vline(xintercept = median(hp), col="blue") +
  labs(title='Distribution of Displacement vs. Horse Power', x='', y='Density')
```

## Distribution of Displacement vs. Horse Power



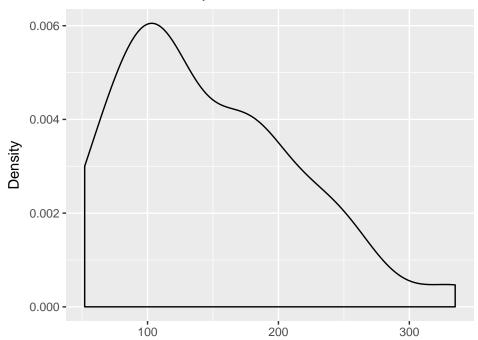
#### c) overall distribution

**Answer:** A distribution plot.

#### Example:

```
ggplot(mtcars, aes(hp, fill=hp)) +
  geom_density() +
  labs(title='Distribution of Displacement vs. Horse Power', x='', y='Density')
```

## Distribution of Displacement vs. Horse Power



## Proportion of categories of a variable

**Answer:** A mosaic plot can show this.

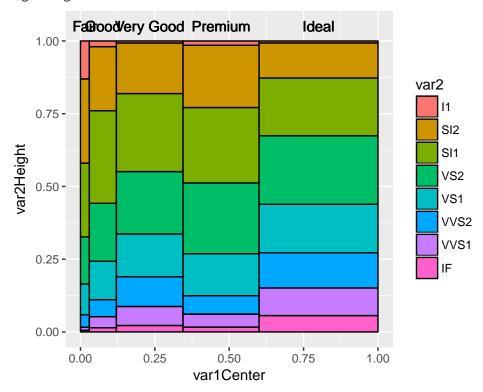
```
# Code from http://stackoverflow.com/questions/19233365/how-to-create-a-marimekko-mosaic-plot-in-ggplot
ggMMplot <- function(var1, var2){
    require(ggplot2)
    levVar1 <- length(levels(var1))
    levVar2 <- length(levels(var2))

jointTable <- prop.table(table(var1, var2))
    plotData <- as.data.frame(jointTable)
    plotData$marginVar1 <- prop.table(table(var1))
    plotData$var2Height <- plotData$Freq / plotData$marginVar1
    plotData$var1Center <- c(0, cumsum(plotData$marginVar1)[1:levVar1 -1]) +
        plotData$marginVar1 / 2

ggplot(plotData, aes(var1Center, var2Height)) +
        geom_bar(stat = "identity", aes(width = marginVar1, fill = var2), col = "Black") +
        geom_text(aes(label = as.character(var1), x = var1Center, y = 1.05))</pre>
```

```
}
ggMMplot(diamonds$cut, diamonds$clarity)
```

## Warning: Ignoring unknown aesthetics: width



## Spatial dependency of a variable

Answer: A map.

```
murder <- subset(crime, offense == "murder")
qmplot(lon, lat, data = murder, colour = I('red'), size = I(3), darken = .3)

## Using zoom = 11...

## Map from URL : http://tile.stamen.com/toner-lite/11/479/845.png

## Map from URL : http://tile.stamen.com/toner-lite/11/480/845.png

## Map from URL : http://tile.stamen.com/toner-lite/11/481/845.png

## Map from URL : http://tile.stamen.com/toner-lite/11/482/845.png

## Map from URL : http://tile.stamen.com/toner-lite/11/479/846.png

## Map from URL : http://tile.stamen.com/toner-lite/11/480/846.png

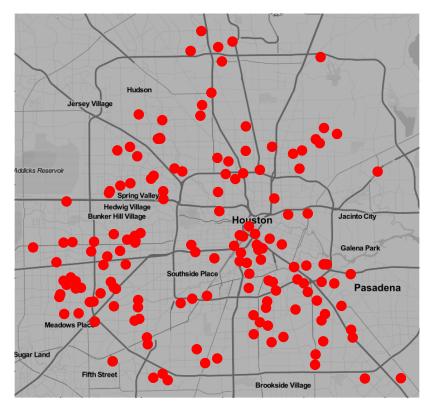
## Map from URL : http://tile.stamen.com/toner-lite/11/481/846.png

## Map from URL : http://tile.stamen.com/toner-lite/11/482/846.png

## Map from URL : http://tile.stamen.com/toner-lite/11/482/846.png

## Map from URL : http://tile.stamen.com/toner-lite/11/479/847.png</pre>
```

## Map from URL : http://tile.stamen.com/toner-lite/11/480/847.png
## Map from URL : http://tile.stamen.com/toner-lite/11/481/847.png
## Map from URL : http://tile.stamen.com/toner-lite/11/482/847.png
## Warning: `panel.margin` is deprecated. Please use `panel.spacing` property
## instead



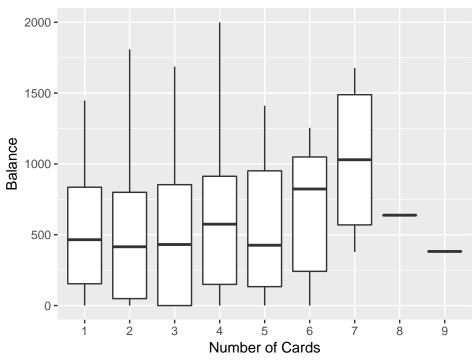
Credit card balances of 400 individuals are given in the data file Credit.csv. The goal is to find which explanatory variables are affecting the card balance. Based on this data, answer the following questions. Justify your answer by providing a suitable plot in each case.

```
credit <- read.csv('Credit.csv')</pre>
head(credit)
     id Income Limit Rating Cards Age Education Gender Student Married
                                2 34
## 1 1 14.891 3606
                        283
                                            11
                                                 Male
                        483
                                3 82
                                                                  Yes
## 2 2 106.025 6645
                                            15 Female
                                                          Yes
## 3 3 104.593 7075
                        514
                                4 71
                                            11
                                                 Male
                                                           No
                                                                   No
## 4 4 148.924 9504
                        681
                                3 36
                                            11 Female
                                                           No
                                                                   No
                                            16
## 5 5 55.882 4897
                               2 68
                        357
                                                 Male
                                                                  Yes
                                                           No
## 6 6 80.180 8047
                        569
                            4 77
                                                 Male
                                                           No
                                                                   No
    Ethnicity Balance
## 1 Caucasian
## 2
        Asian
                  903
## 3
        Asian
                  580
## 4
        Asian
                  964
## 5 Caucasian
                  331
## 6 Caucasian
                 1151
```

a) Do you think number of cards has anything to do with higher balance?

```
ggplot(credit, aes(x=factor(Cards), y=Balance)) +
  geom_boxplot() +
  labs(title="Balance by Number of Cards", x="Number of Cards")
```



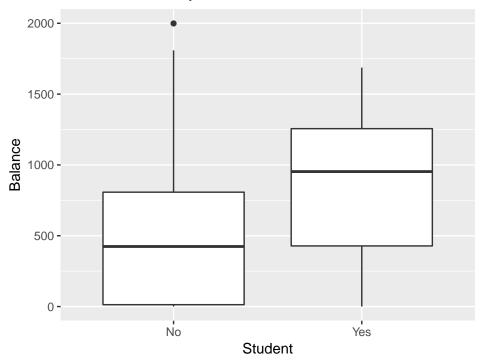


There does not appear to be a linear relationship between number of cards and balance. Although people with 6 and 7 cards appear to have the highest balance by median.

# b) Do you think that higher balances can be attributed to individual's studentship status?

```
ggplot(credit, aes(x=factor(Student), y=Balance)) +
  geom_boxplot() +
  labs(title="Student Status by Number of Cards", x="Student")
```

## Student Status by Number of Cards

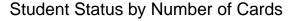


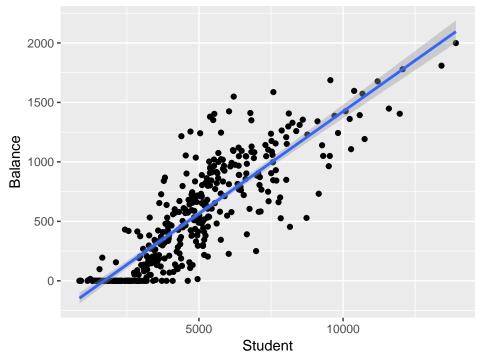
There appears to be clear visual evidence that being a student has a positive correlation to credit card balance. As most of us know from first hand experience.

## c) Does higher limit causes higher balance?

```
fit <- lm(credit$Balance ~ credit$Limit)
r.sq <- summary(fit)$adj.r.squared

ggplot(credit, aes(x=Limit, y=Balance)) +
   geom_point() +
   geom_smooth(method="lm") +
   labs(title="Student Status by Number of Cards", x="Student")</pre>
```





With an adjusted  $R^2$  of 0.7418753, there is a strong positive correlation between limit and balance.

## d) Is there any interaction effect of Student and limit on Balance?

```
credit$Student <- as.numeric(credit$Student) - 1
fit <- lm(credit$Balance ~ (credit$Student + credit$Limit)^2)
sum.fit <- summary(fit)
p.val <- sum.fit$coefficients[4,4]</pre>
```

With a p-value of 0.1830097, there is not enough evidence to say that the interaction effect is significantly different than zero.

## e) Is there any interaction effect of Income and Ethnicity on Balance?

```
credit$Ethnicity <- as.numeric(credit$Ethnicity) - 1
fit <- lm(credit$Balance ~ (credit$Income + credit$Ethnicity)^2)
sum.fit <- summary(fit)
eth.p.val <- sum.fit$coefficients[3,4]
inc.eth.p.val <- sum.fit$coefficients[4,4]</pre>
```

There doesn't appear to be any effect caused by Ethnicity, with a p-value of 0.124171, and the p-value for the interaction effect is 0.0651074 which is not less than 0.05, and thus we cannot say that the effect is significantly different than zero.

## f) Is there any 3-way interaction effect of Income, Ethnicity and Student on Balance?

```
fit <- lm(credit$Balance ~ (credit$Income + credit$Ethnicity + credit$Student)^3)
sum.fit <- summary(fit)</pre>
##
## Call:
## lm(formula = credit$Balance ~ (credit$Income + credit$Ethnicity +
       credit$Student)^3)
##
## Residuals:
     Min
             1Q Median
                            3Q
                                  Max
## -811.9 -332.3 -54.9 326.9 818.6
## Coefficients:
##
                                                 Estimate Std. Error t value
## (Intercept)
                                                 120.1304 59.1141 2.032
## credit$Income
                                                   7.5234
                                                              0.9916 7.587
## credit$Ethnicity
                                                  66.9405
                                                             39.9113 1.677
## credit$Student
                                                 613.1709
                                                            205.4414 2.985
## credit$Income:credit$Ethnicity
                                                  -1.1116
                                                              0.6806 -1.633
## credit$Income:credit$Student
                                                              3.1564 -0.756
                                                  -2.3873
## credit$Ethnicity:credit$Student
                                                 -95.7977
                                                            150.5243 -0.636
## credit$Income:credit$Ethnicity:credit$Student -0.1989
                                                              2.6686 -0.075
##
                                                 Pr(>|t|)
## (Intercept)
                                                  0.04281 *
## credit$Income
                                                 2.41e-13 ***
                                                  0.09429 .
## credit$Ethnicity
## credit$Student
                                                  0.00302 **
## credit$Income:credit$Ethnicity
                                                  0.10322
## credit$Income:credit$Student
                                                  0.44991
## credit$Ethnicity:credit$Student
                                                  0.52487
## credit$Income:credit$Ethnicity:credit$Student 0.94062
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 391.3 on 392 degrees of freedom
## Multiple R-squared: 0.2884, Adjusted R-squared: 0.2757
## F-statistic: 22.69 on 7 and 392 DF, p-value: < 2.2e-16
eth.p.val <- sum.fit$coefficients[3,4]</pre>
inc.eth.stu.p.val <- sum.fit$coefficients[8,4]</pre>
```

With a p-value of 0.9406179, there is no three-way interaction effect between Income, Ethnicity, and Student status on the balance.

- 3. Creating dashboard. For this problem you will start with dashboard.R as a template. Modify this template and create a new dashboard according to the instruction below;
- a) Inside sidebarMenu, create a new menuItem called "My state crime"
- b) Inside dashboardBody tabItems, create a new tabItem and make the header "My state crime map goes here"
- c) Inside the newly created tabItem create a fluidRow using column instead of box which has following features. For example codes, review the link https://github.com/mamajumder/usa-crime/blob/master/ui.R

Use first 5 columns to create a wellPanel where there will be one selectInput and one sliderInput. selectInput should allow to select a specific crime rate to display while sliderInput will use a specific year of crime rate. For this use usaCrimeDat.rds

Use 6 columns to show the state crime map of USA colored by crime rate.

d) Inside server add a new output of the state crime map so that map can be interactively generated based on crime rate and the year selected.