

Homework #2

Azadeh Gilanpour

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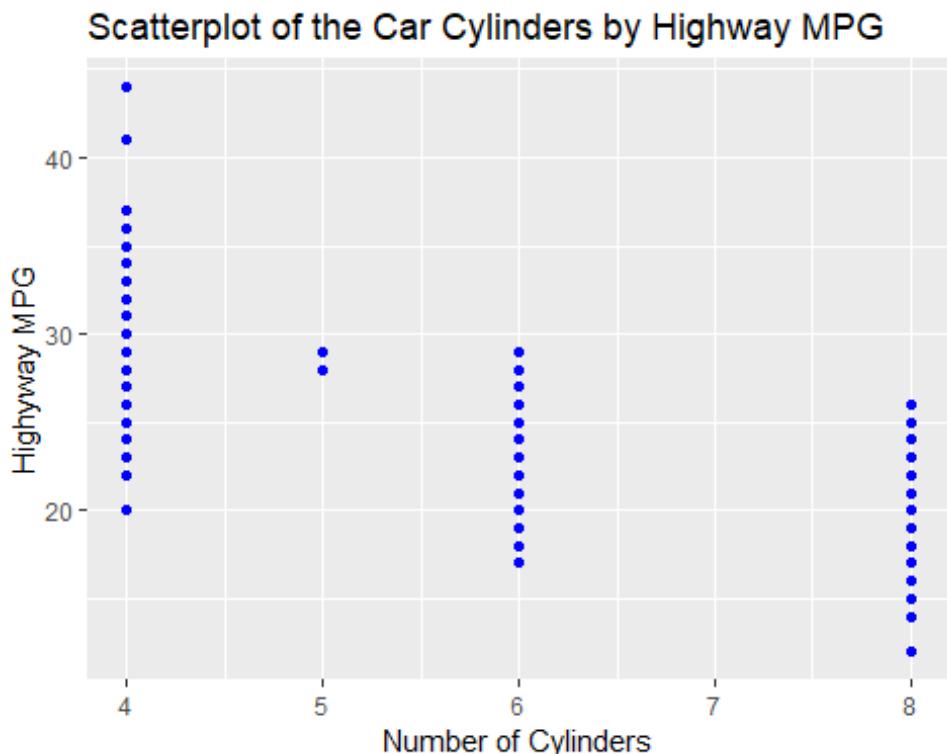
Problem 1: Learning ggplot2

(a)

- 3.2.4 Exercises 4:

Make a scatterplot of hwy vs cyl

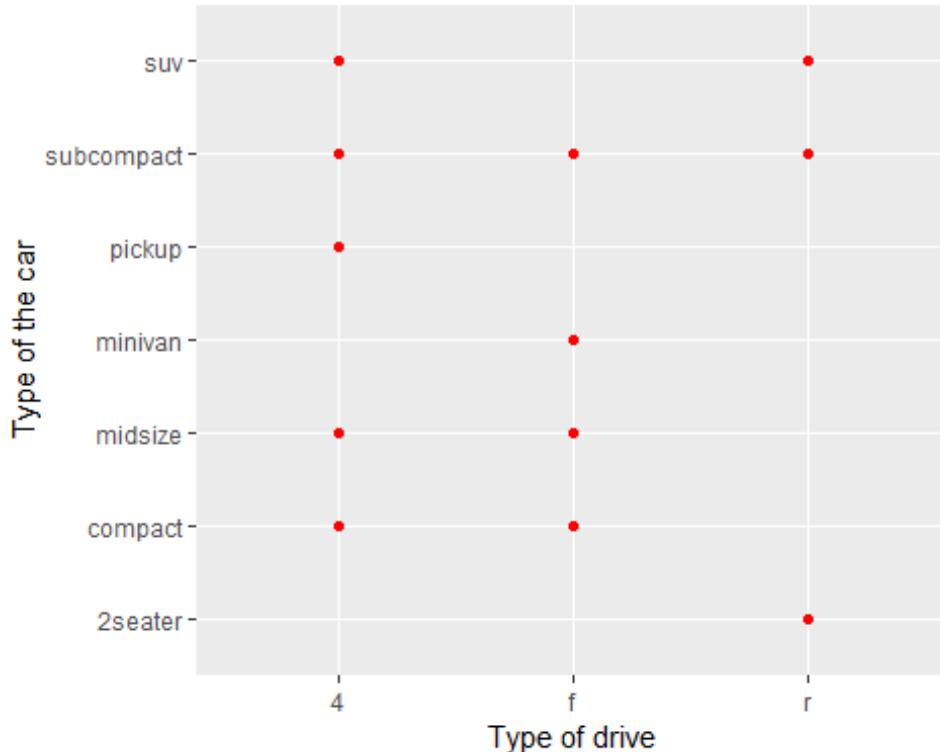
```
ggplot(data=mpg,mapping=aes(x=cyl,y=hwy))+
  geom_point(color="blue")+
  labs(x="Number of Cylinders",y="Highway MPG",title="Scatterplot of the Car Cylinders by Highway MPG")
```



- The plot shows a negative relationship between the number of cylinder (cyl) and fuel efficiency (hwy). That means the car has a bigger cylinder uses more fuel.
- 3.2.4 Exercises 5:

Make a scatterplot of class vs drv

```
ggplot(data = mpg, mapping = aes(x=drv,y=class))+
  geom_point(color="red")+
  labs(x="Type of drive",y="Type of the car")
```



- This scatterplot is useless because it compares two categorical attributes of "drv" and "class" with each other. Each point in the plot has overlap with the other.

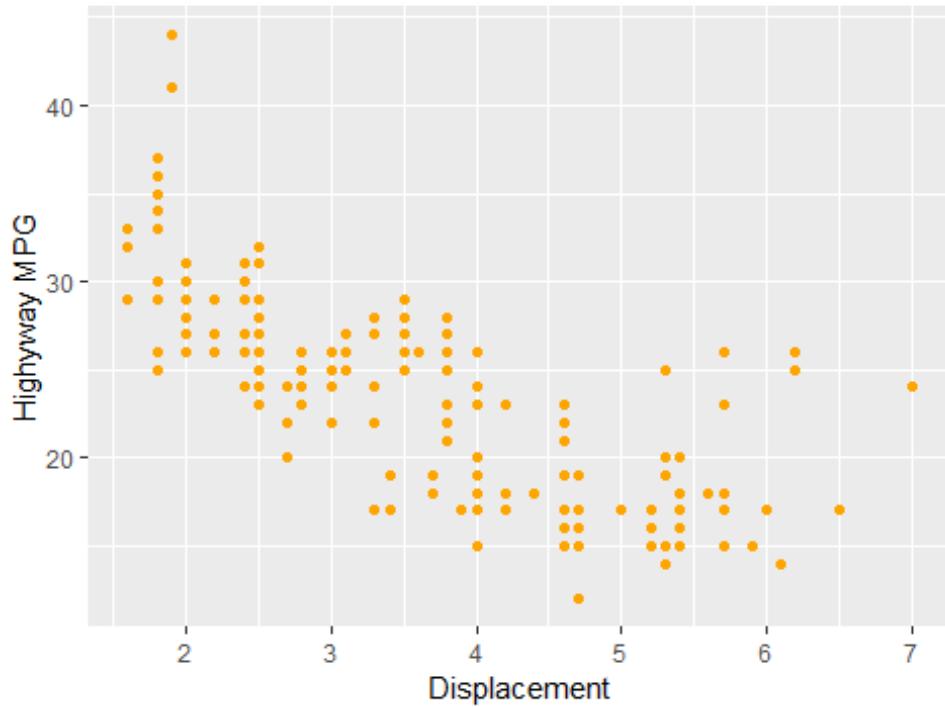
(b)

- 3.3.1 Exercises 3:

Map continuous variable to Color/Size/Shape

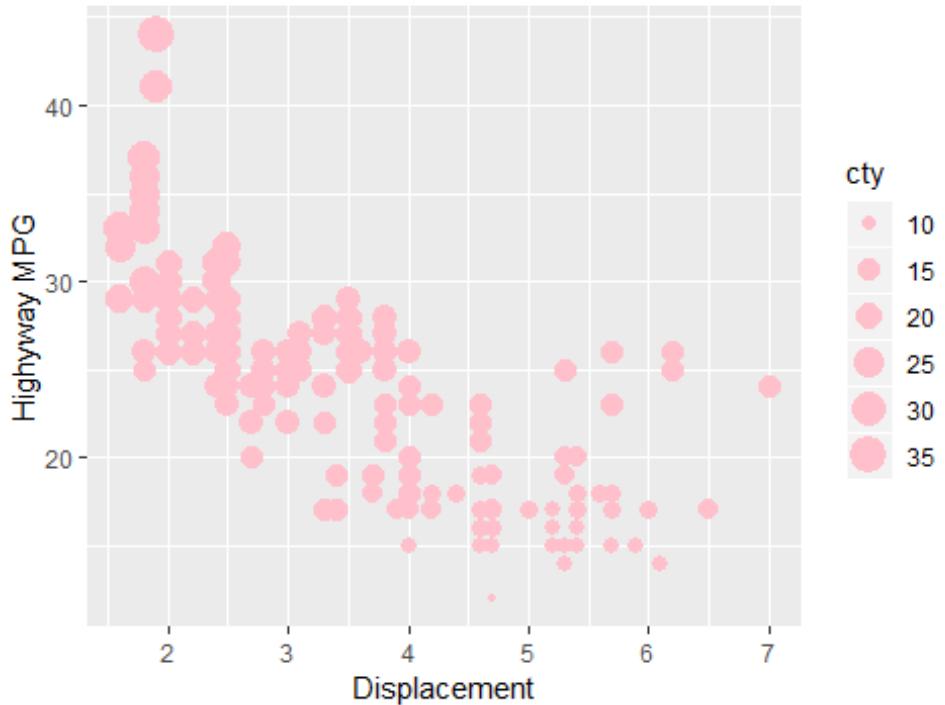
```
#continuous variable
ggplot(data=mpg)+
  geom_point(mapping = aes(x=displ,y=hwy,color= cyl),color='orange')+
  labs(x="Displacement",y="Highway MPG",title="Scatterplot of the displacement by Highway MPG")
```

Scatterplot of the displacement by Highway MPG



```
ggplot(data=mpg)+  
  geom_point(mapping = aes(x=displ,y=hwy,size= cty),color='pink')+  
  labs(x="Displacement",y="Highway MPG",title="Scatterplot of the displacement by Highway  
MPG")
```

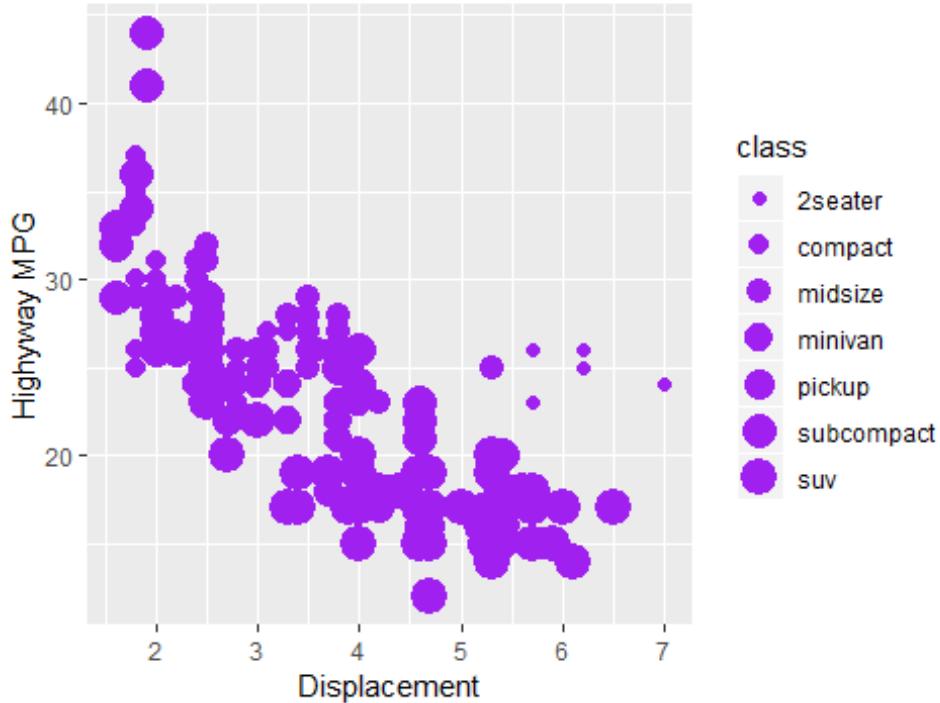
Scatterplot of the displacement by Highway MPG



```
#categorical variable  
ggplot(data=mpg)+  
  geom_point(mapping = aes(x=displ,y=hwy,size= class),color="purple") +  
  labs(x="Displacement",y="Highway MPG",title="Scatterplot of the displacement by Highway MPG")
```

Warning: Using size for a discrete variable is not advised.

Scatterplot of the displacement by Highway MPG



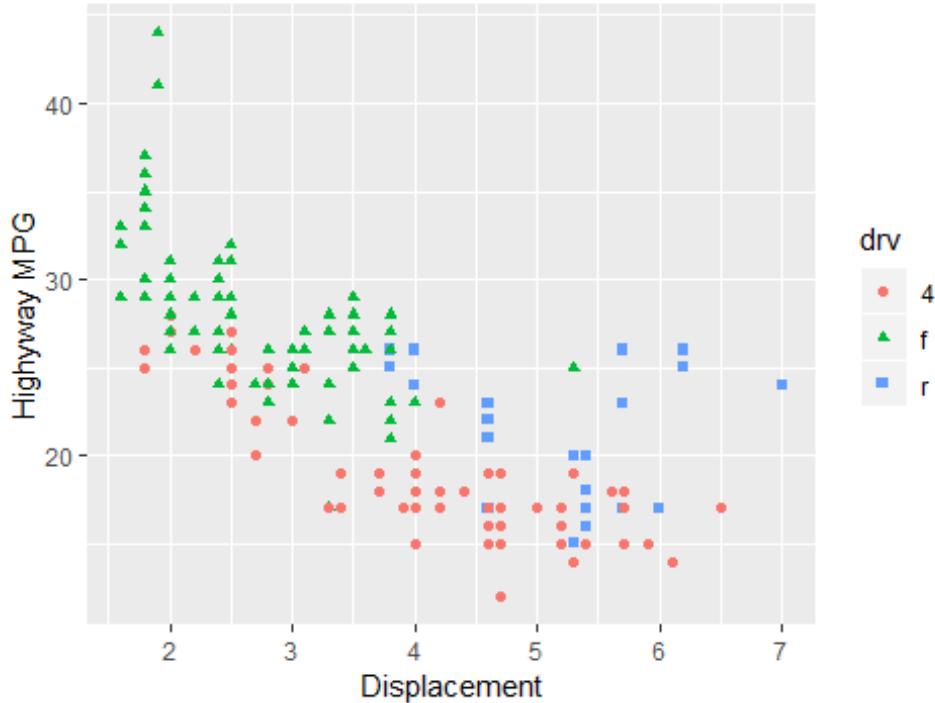
Behave of aesthetics for continous and categorical variables

- The result shows size and color aesthetic works for continuous variables while assigning shape aesthetic to continuous variables get error because shape are not ordered while continuous variables are ordered. Also, using size aesthetic for categorical variable does not make sense because of non ordinal of categorical variables.
- 3.3.1 Exercise 4:

The result of mapping same variables to multiple aesthetics

```
ggplot(data=mpg)+  
  geom_point(mapping = aes(x=displ,y=hwy,color=drv,shape=drv))+  
  labs(x="Displacement",y="Highway MPG",title="Scatterplot of the displacement by Highway MPG")
```

Scatterplot of the displacement by Highway MPG

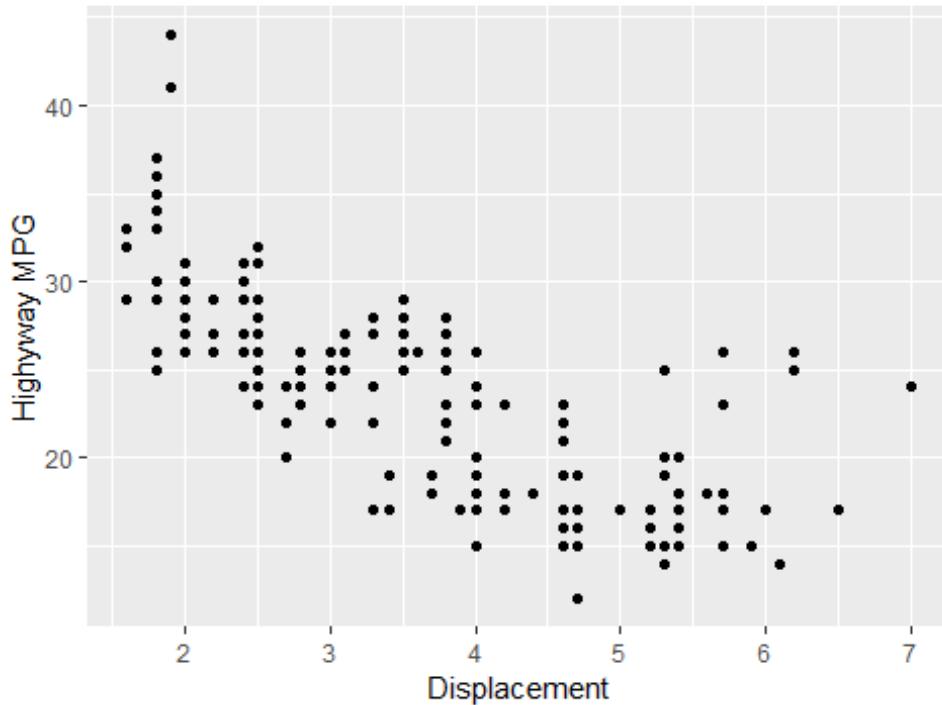


- We can map the same variable to multiple aesthetics. In some data even get us more insight to analyse data. But we just need consider that the aesthetics are compatible with the type of the variable.
- 3.3.1 Exercise 6:

The result of mapping aesthetic to sth other than a varibale name.

```
ggplot(data=mpg)+  
  geom_point(mapping = aes(x=displ,y=hwy,class= displ<5))+  
  labs(x="Displacement",y="Highway MPG",title="Scatterplot of the displasment by Highway  
MPG")  
## Warning: Ignoring unknown aesthetics: class
```

Scatterplot of the displacement by Highway MPG

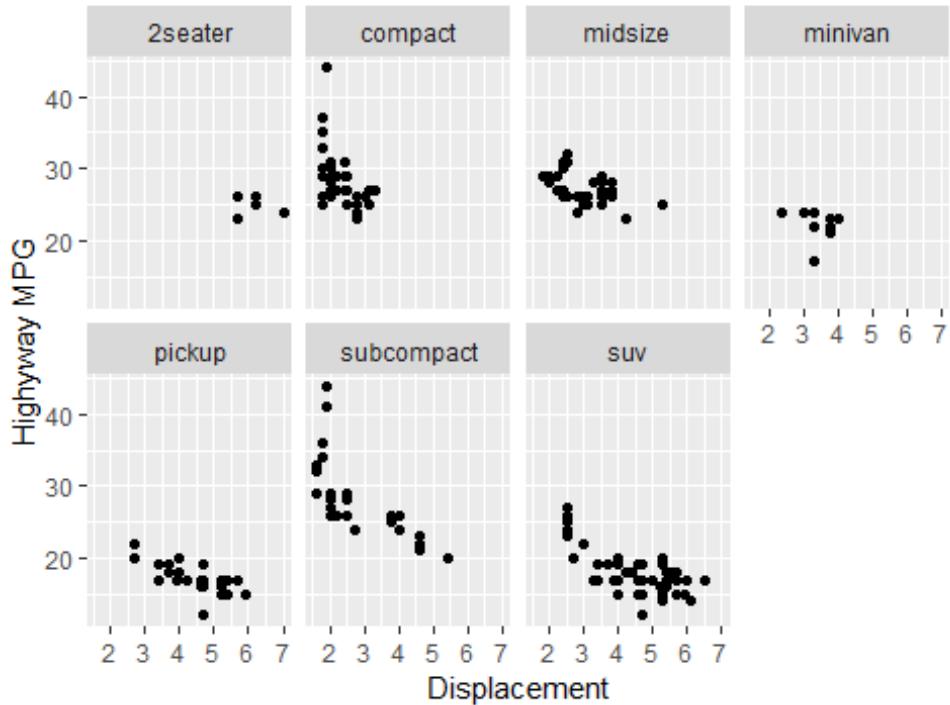


- The above plot shows that if aesthetics map to sth other than variable name divided the observation into boolean variables.
- 3.5.1 Exercise 4:

The visualtion of using “Facet” insted of “color”

```
ggplot(data = mpg) +
  geom_point(mapping = aes(x = displ, y = hwy)) +
  facet_wrap(~ class, nrow = 2) +
  labs(x = "Displacement", y = "Highway MPG", title = "Scatterplot of the displacement by Highway MPG")
```

Scatterplot of the displacement by Highway MPG



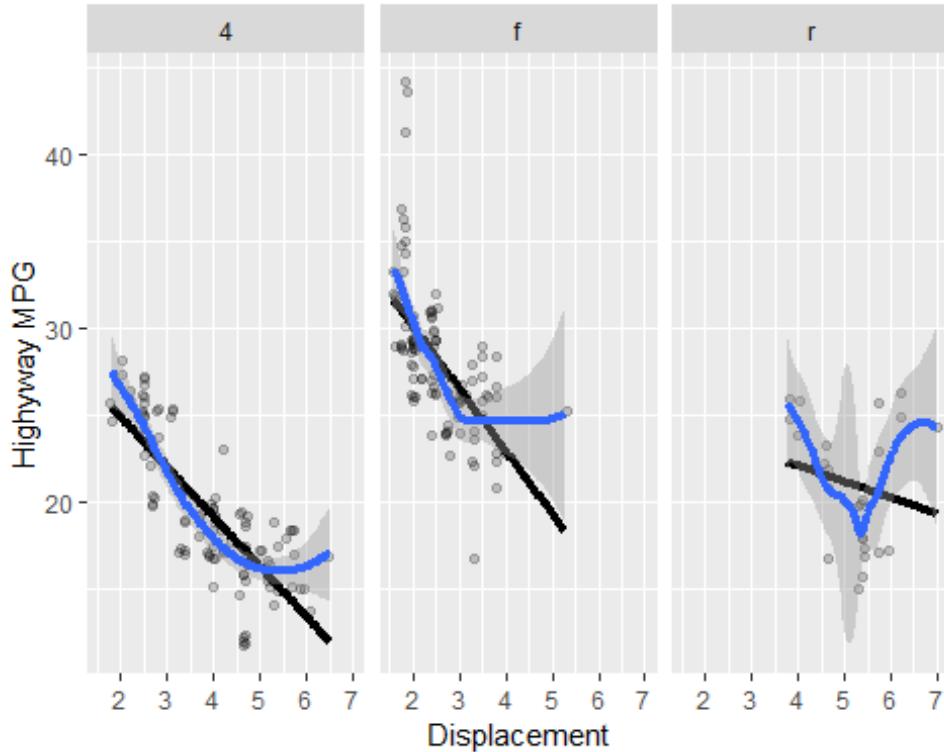
- Advantage: Advantages of using “facet” is allow us to encode more distinct categories insted of dealing with so several colors.
- Disadvantage: With “faceting” it is easier to examine the individual classes while “coloring” make easier to see how the classes are clustered overall.
- With large dataset: When we have large dataset using “Color”it may become difficult to clearly distinguish points based on color and using “facet” provide simple result.

(b)

Provide reproduce visualization for the “mpg” data

```
ggplot(mpg, aes(x=displ, y=hwy)) +
  geom_point(alpha=0.2, position = "jitter") +
  geom_smooth(method="lm", color="black", span = 0.89, se=F, size=1.5) +
  facet_wrap(~drv) +
  geom_smooth(size=1.5) +
  labs(x="Displacement", y="Highway MPG")

## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



The plot shows a negative relationship between engine size (displ) and fuel efficiency (hwy). Also we can see that front-wheel drive with the little engine can be more effecincy.

Problem 2: Generating data and advanced density plots

(a)

Generate random variable. Four sets of random numbers a-d have been generated in different distributions. The variales combined into a data frame “df” and then reshape data with “gather” function and create “df2”data frame.

melted into a new data frame (df2)

```
df data frame
df=data.frame(a,b,c,d)
head(df,2)

##      a      b      c      d
## 1 -0.5071735 0.3328360 5.127979 0.4846009
## 2 -1.7335064 0.3543254 2.322969 0.5682989
```

```
df2 data frame
df2<- gather(df, key="groupVar",value="value")
head(df2,4)
```

```

## groupVar    value
## 1      a -0.5071735
## 2      a -1.7335064
## 3      a  0.3623584
## 4      a  0.3280874

```

- The figure shows the density function for the random data sets. Different distributions have different behavior as shown in the

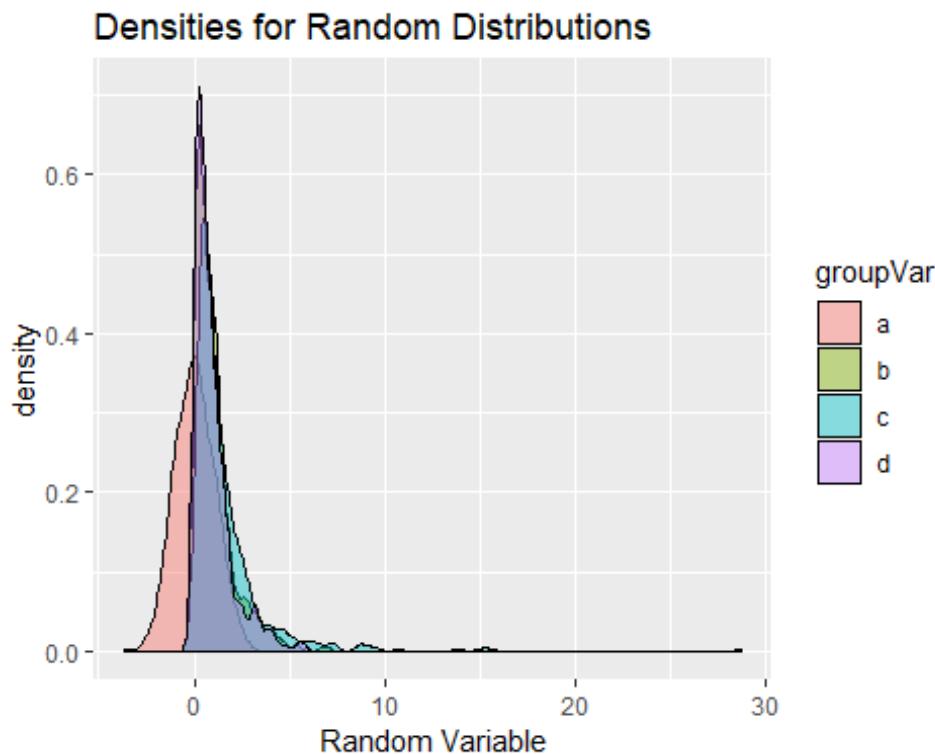
(b)

Using ggplot to visualize the densities

```

ggplot(df2, aes(x = value, fill = groupVar)) +
  geom_density(alpha = 0.45) +
  labs(x = "Random Variable",
       title = "Densities for Random Distributions")

```



problem 3: house prices data

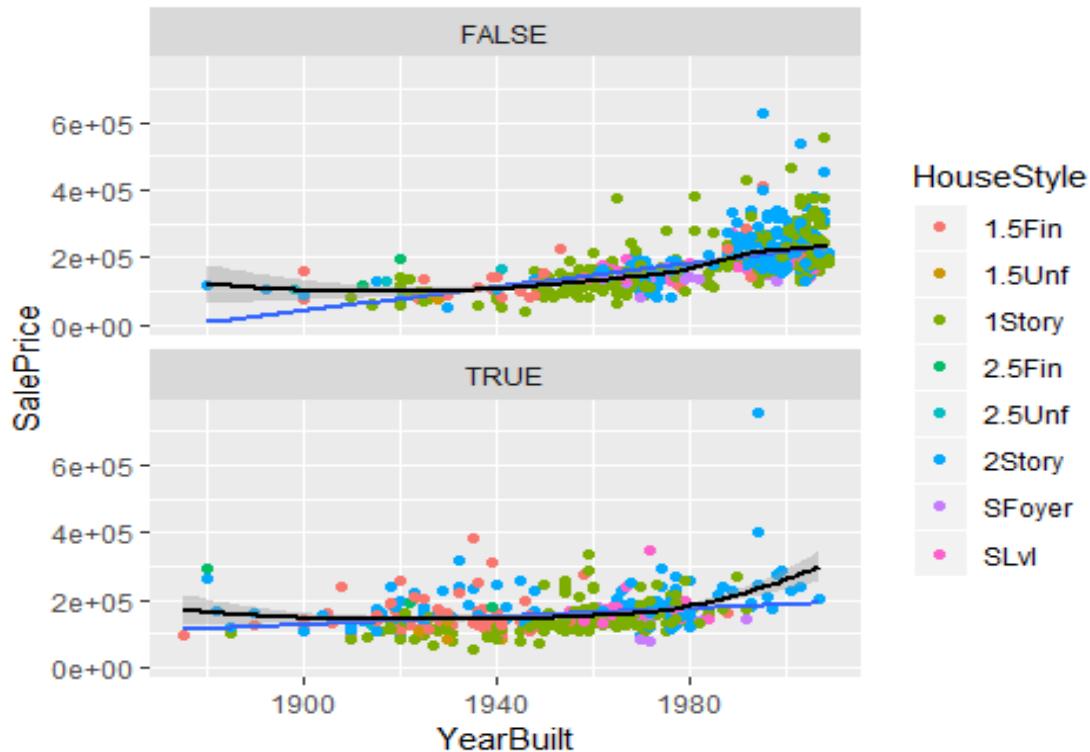
Read data from file “housingData.csv” and visualize the housing dataset
`housing<-read.csv("housingData.csv")`

```

ggplot(housing, aes(x=YearBuilt, y=SalePrice)) +
  geom_point(aes(color=HouseStyle)) +
  geom_smooth(method="lm", se=F) +
  facet_wrap(~OverallCond>5, nrow = 2) + geom_smooth(color="black")

```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



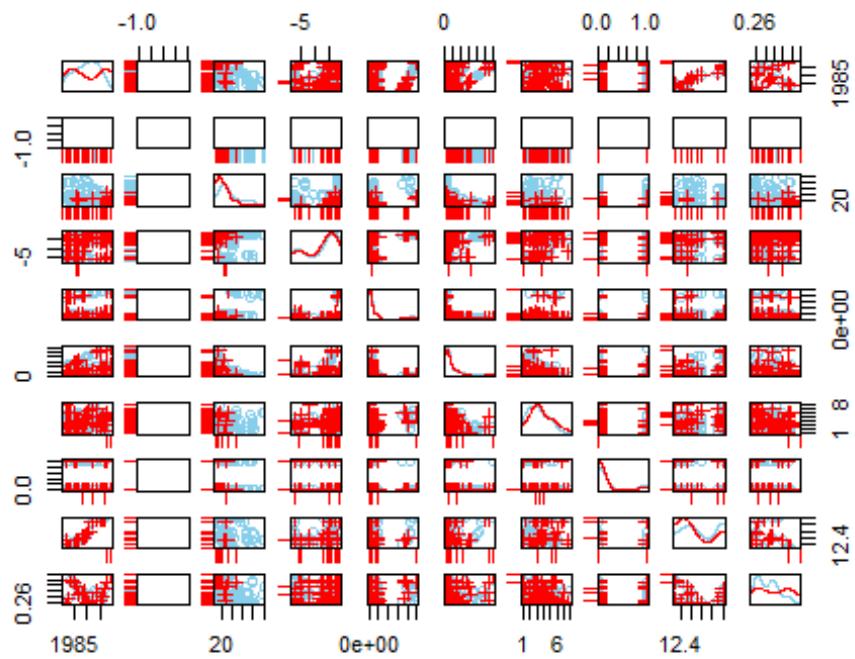
- The plot reveals that regardless of the houses condition during the 1950-1970 the 1story Style house with the min price of 200k were popular. This trend changes by the time over the first decade of 20 century. People pay more for same house style with the worse condition.

problem 4:Missing Data

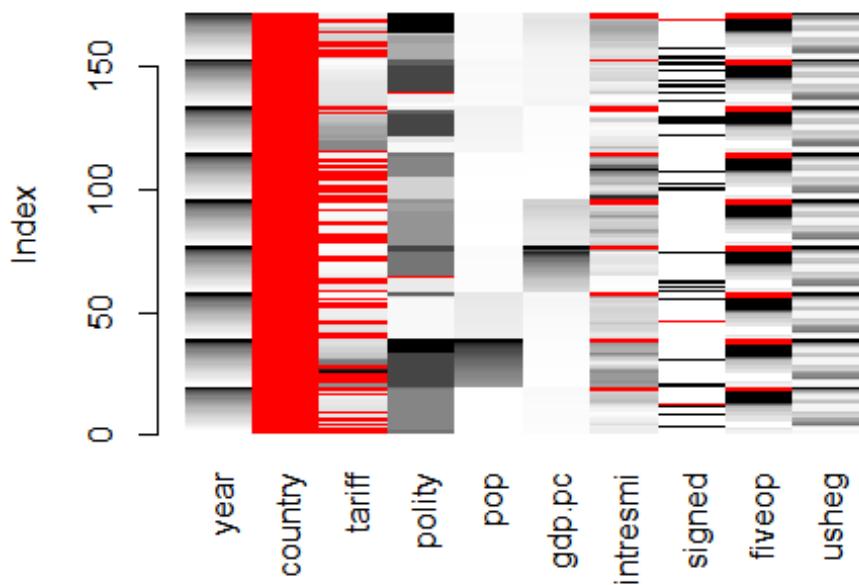
(a)

Explore the missingness on “freetrade” dataset

```
# scatterplot for missing values  
scattmatrixMiss(freetrade)
```



```
# Matrix plot for missing values
matrixplot(freetrade, interactive = F, sortby = "country")
```



- The Result of investigation on “freetrade” data set shows that the “tariff” variable contain more missing value among the other values and “fiveop” and “intresmi” are in the second and third place of missing values.

(b)

- I used Chi -squared test for this question and the result of it on original dataset and the dataset without “Nepal” and “Philippines” reveal that p-value is the same for all three conditions. Therefore we can assume that there is no significant change if we remove some data from dataset. So it can be concluded that the missing value in “tariff” is completely independent and there is no relationship between the “country” and “tariff” variable.