

Project

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Task 1: Analysis

The data is loaded using the following command.

```
cardata<-read.csv("~/Desktop/R-Programming/USA_cars_datasets.csv")
```

The following command calculates of Frequency of cars with its brand name

```
table(cardata$brand)
```

```
##          acura      audi      bmw      buick      cadillac
##            3         4        17        13         10
##      chevrolet     chrysler      dodge      ford       gmc
##      297           18        432      1235        42
## harley-davidson   heartland      honda      hyundai    infiniti
##            1         5        12        15         12
##      jaguar        jeep       kia       land      lexus
##            1        30        13         4          2
##      lincoln      maserati      mazda  mercedes-benz      nissan
##            2         1         2        10        312
##      peterbilt      ram      toyota
##            4         1         1
```

We can notice that car of the brand Ford is highly used by people(1235).

The following command calculates the number of observations having car price greater than 10000 dollars

```
table(cardata$price>10000.0)
```

```
##  
## FALSE TRUE  
## 615 1884
```

which says 1884 observations with price >10000.0 615 observations with price <=10000.0

The following command calculates the average price and average mileage(miles travelled) of cars

```
apply(cardata[2],2,mean)
```

```
## price  
## 18767.67
```

```
apply(cardata[7],2,mean)
```

```
## mileage  
## 52298.69
```

The following command calculates the standard deviation and quautiles of the car price

```
sd(cardata$price)
```

```
## [1] 12116.09
```

```
quantile(cardata$price)
```

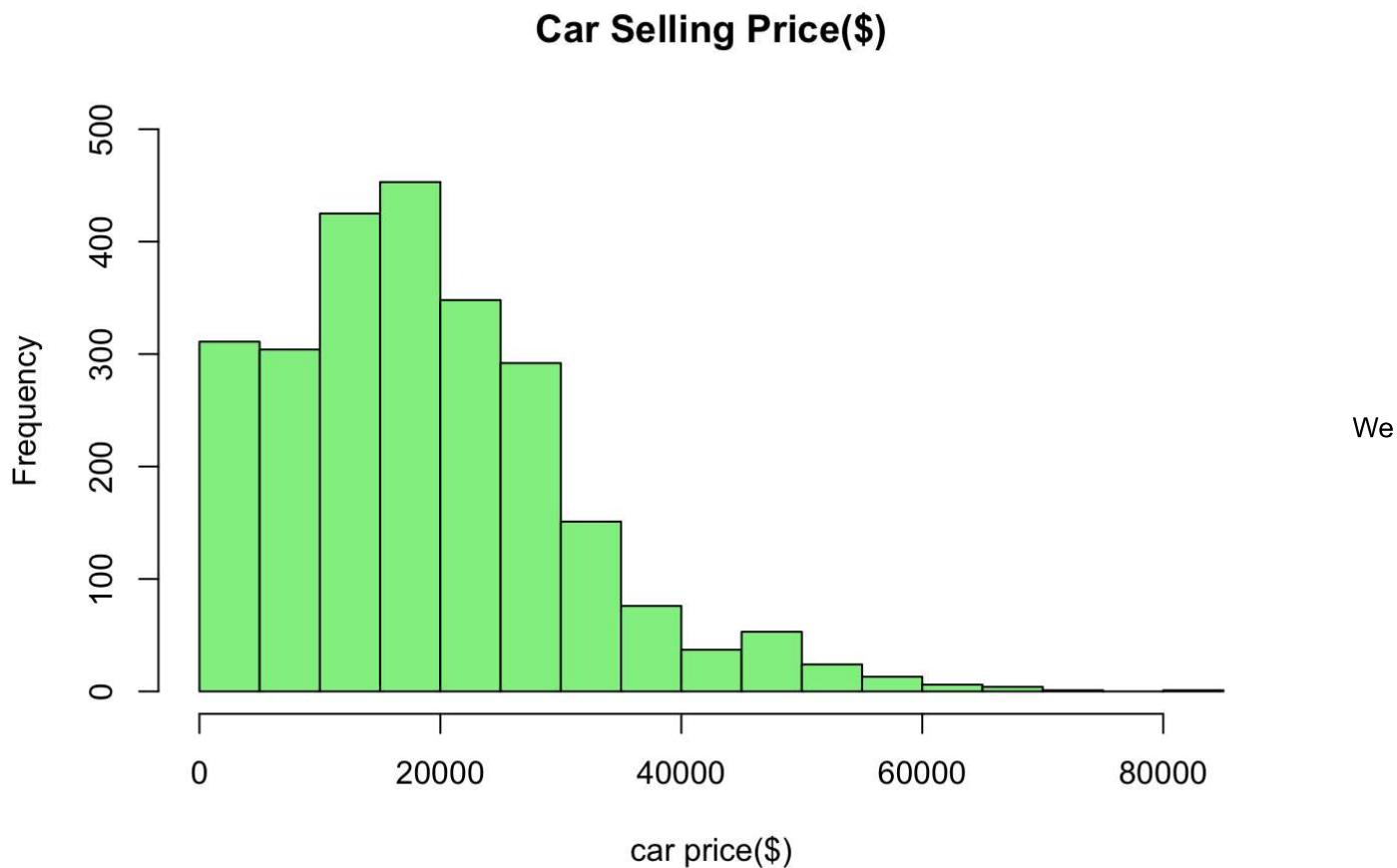
```
##      0%     25%     50%     75%    100%  
##     0.0 10200.0 16900.0 25555.5 84900.0
```

```
# Command showing 10th and 90th percentiles  
quantile(cardata$price,c(0.1,0.9))
```

```
## 10% 90%  
## 4140 33720
```

The following command produces a histogram of car selling price

```
hist(cardata$price,main="Car Selling Price($)",xlab="car price($)",col="light green",ylim = c(0, 500))
```



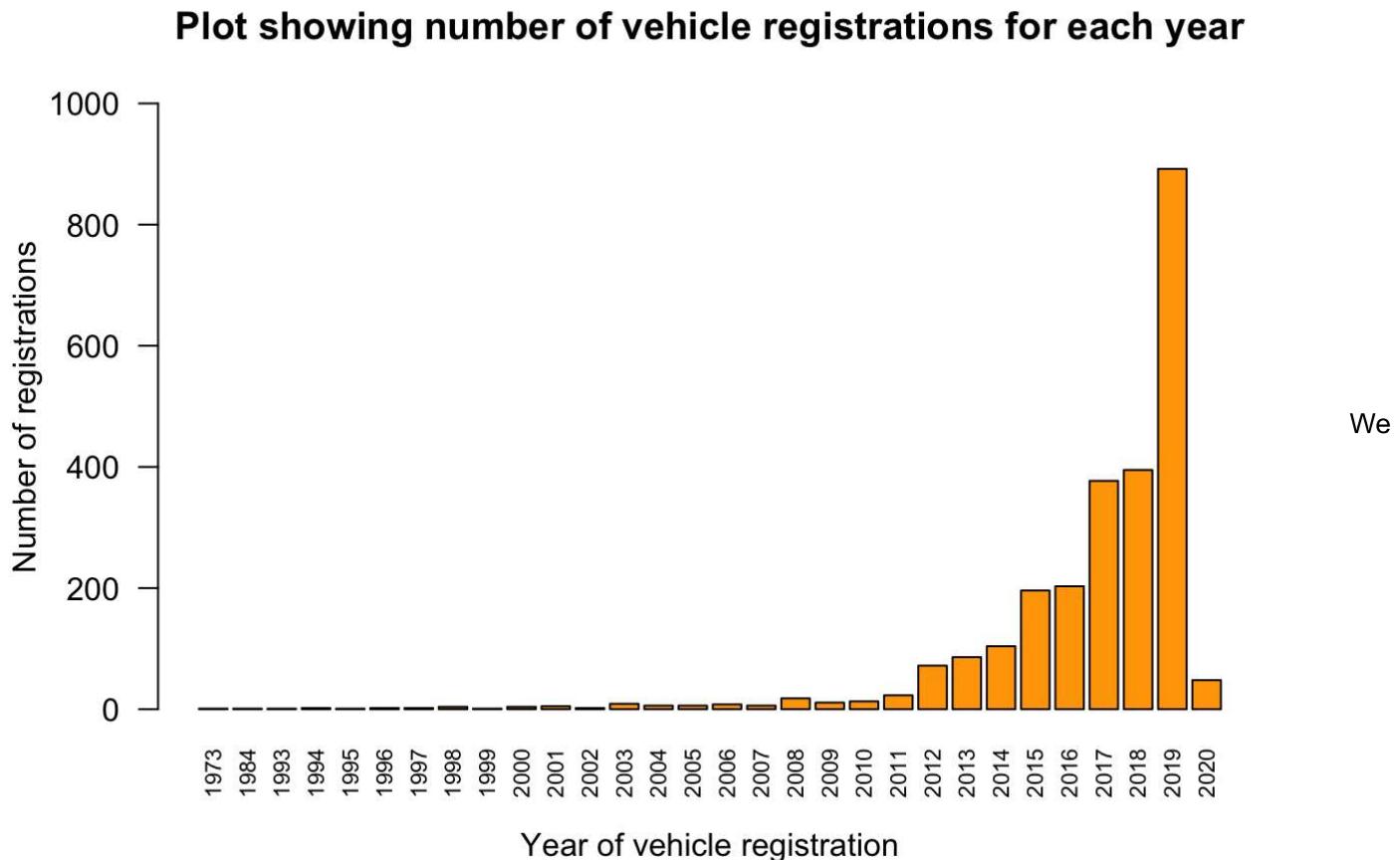
can see that Selling price of majority of the cars is less than 20000\$

The following command produces a table of the number of vehicle registrations for each year and a barplot.

```
tab1<-table(cardata$year)
tab1
```

```
##
## 1973 1984 1993 1994 1995 1996 1997 1998 1999 2000 2001 2002 2003 2004 2005 2006
##    1     1     1     2     1     2     2     4     1     4     5     2     9     6     6     8
## 2007 2008 2009 2010 2011 2012 2013 2014 2015 2016 2017 2018 2019 2020
##    6    18    11    13    23    72    86   104   196   203   377   395   892    48
```

```
barplot(tab1, xlab ="Year of vehicle registration",ylab="Number of registrations",las=2 ,cex.names=0.7,
       main ="Plot showing number of vehicle registrations for each year",ylim=c(0,1000),col="orange")
```

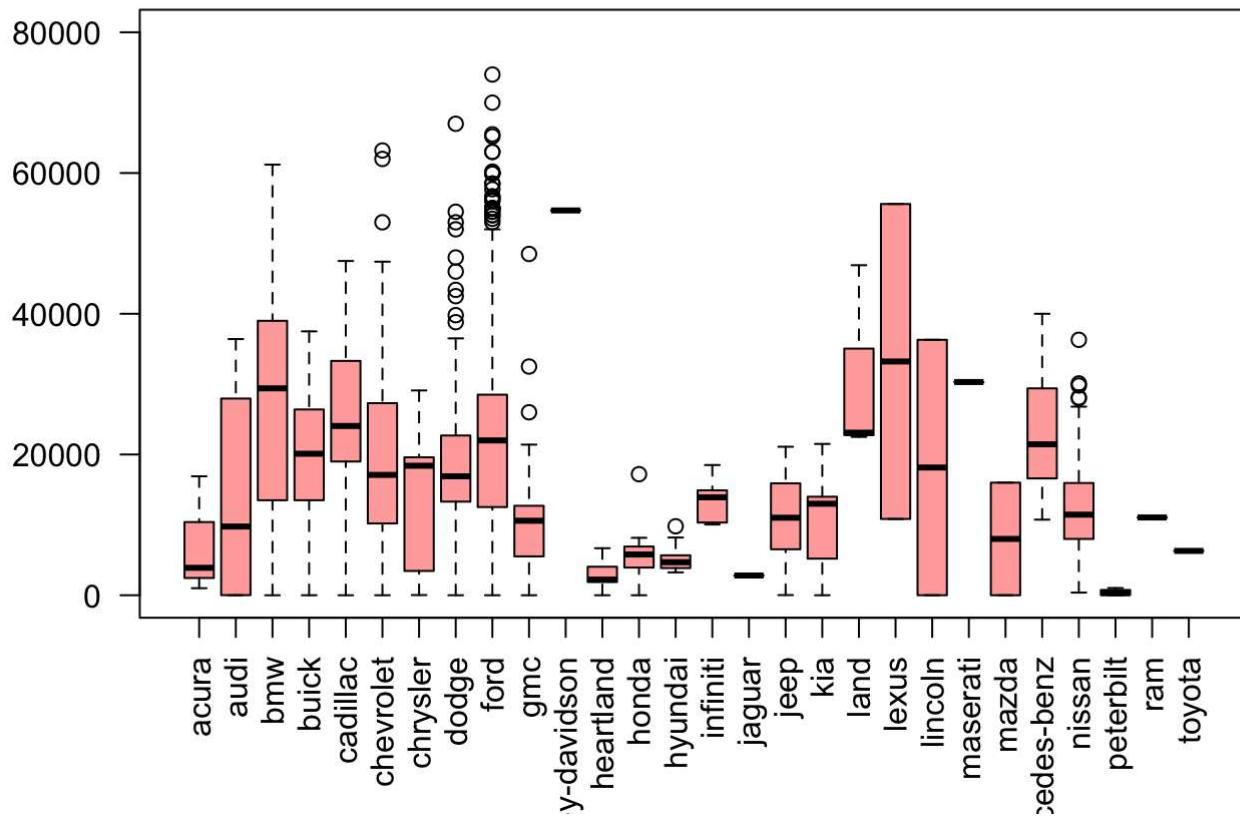


can see that majority of the vehicles are registered in the year 2019.

The following command produces a boxplot of car selling price for each car brand.

```
splitdat<-split(cardata$price,cardata$brand)
boxplot(splitdat,las=2 ,cex.names=1.0,
       main="Car selling price by brand",
       ylim=c(0,80000),col=rgb(1,0,0,alpha = 0.4))
```

Car selling price by brand



We install tidyverse library as ggplot2 is included in the tidyverse package. We can produce a ggplot to see car status.

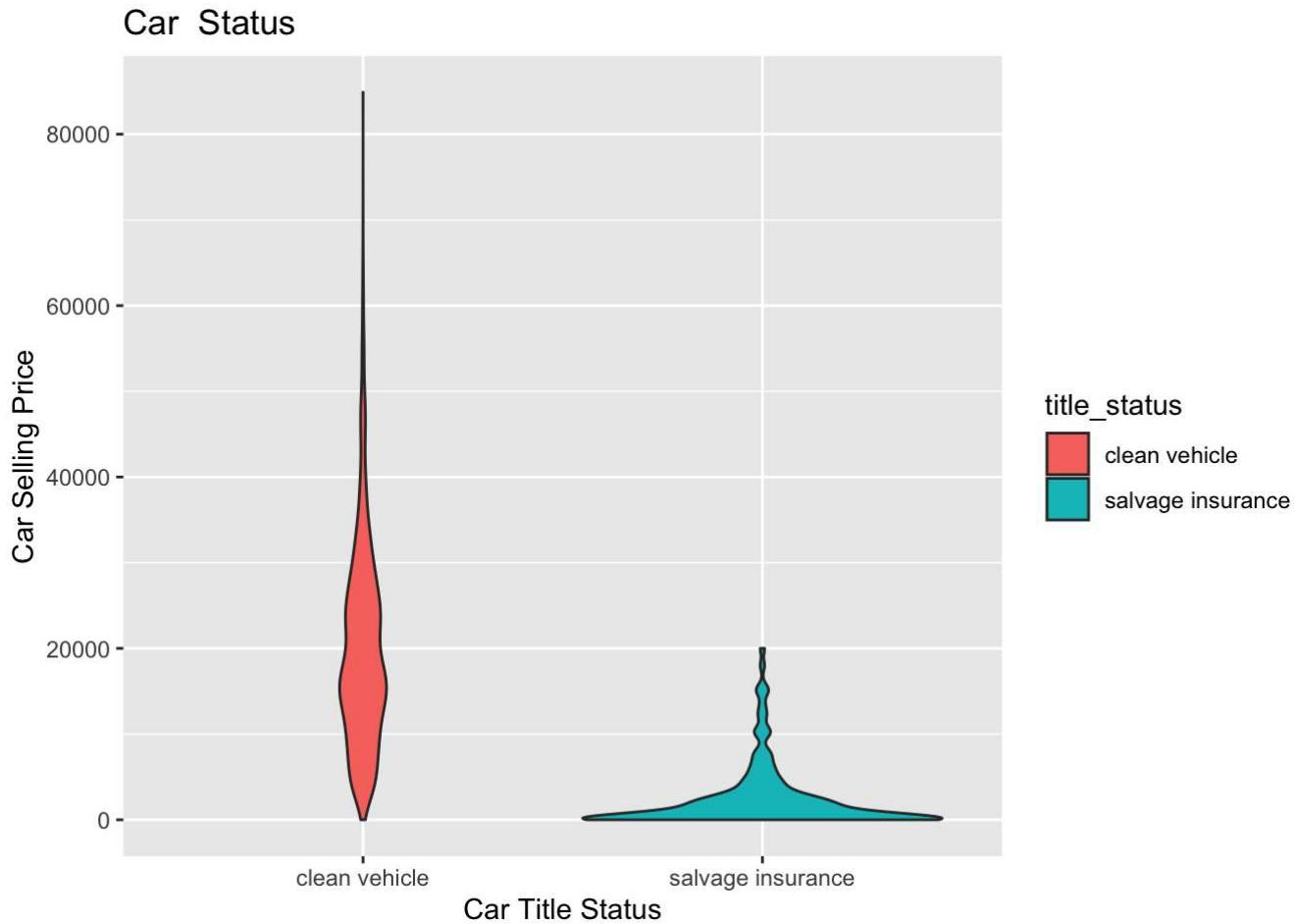
```
library(tidyverse)
```

```
## — Attaching packages ————— tidyverse 1.3.0 —
```

```
## ✓ ggplot2 3.3.2      ✓ purrr   0.3.4
## ✓ tibble  3.0.4      ✓ dplyr    1.0.2
## ✓ tidyr   1.1.2      ✓ stringr 1.4.0
## ✓ readr   1.4.0      ✓ forcats 0.5.0
```

```
## — Conflicts ————— tidyverse_conflicts() —
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()   masks stats::lag()
```

```
ggplot(data=cardata,aes(x=title_status,y=price,fill=title_status))+ggtitle("Car Status")+
  geom_violin(scale="area")+labs(x="Car Title Status",y="Car Selling Price")
```



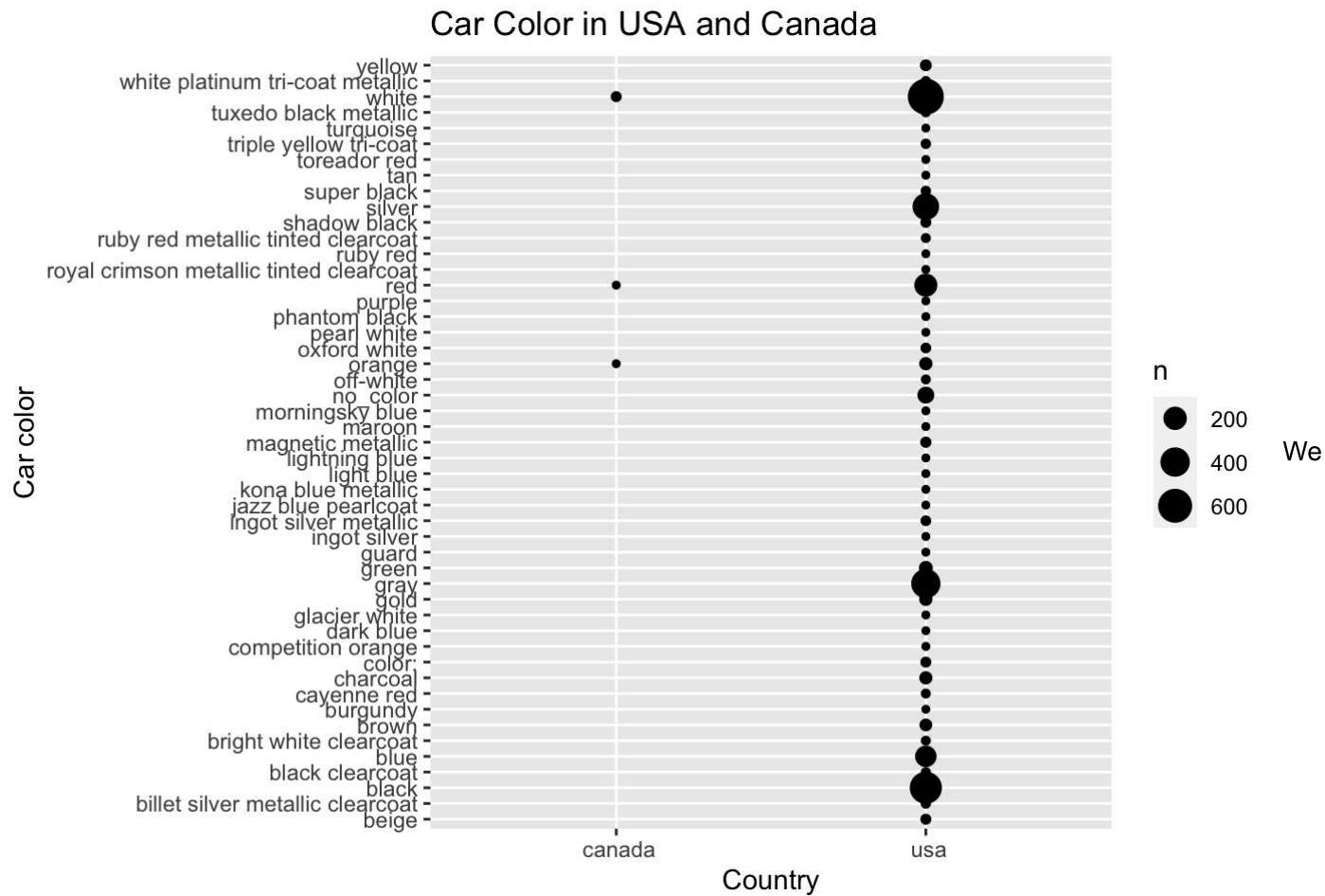
Clean Vehicle: A clean title is one you would receive in most cases when you purchase a vehicle. A brand new vehicle has a clean title and most pre-owned vehicles that can be driven safely and are insurable.

Salvage insurance: A salvage title is given when a vehicle is no longer drivable.

Here the salvage insurance is more when the price of the car is low(0-20000)but on the other hand we can see that the clean title is lesser when the price of the car is 40000(\$) or above.

The following command will tell us which car color is mostly preferred in USA and Canada.

```
ggplot(data=cardata,aes(country,color),width=500,height=4000)+geom_count()+ggtitle("Car Color in USA and Canada")+labs(x="Country",y="Car color")
```



can say that the most preferred car color is White in USA and Canada. And USA has more number of White Color cars than Canada.

The following command will give us summary of USA car dataset

```
summary(cardata)
```

```

##          X           price        brand        model
##  Min.   : 0.0   Min.   : 0   Length:2499      Length:2499
##  1st Qu.: 624.5 1st Qu.:10200  Class :character  Class :character
##  Median :1249.0 Median :16900  Mode  :character  Mode  :character
##  Mean   :1249.0  Mean   :18768
##  3rd Qu.:1873.5 3rd Qu.:25556
##  Max.   :2498.0  Max.   :84900
##          year       title_status        mileage        color
##  Min.   :1973   Length:2499      Min.   :     0   Length:2499
##  1st Qu.:2016   Class :character  1st Qu.: 21466  Class :character
##  Median :2018   Mode  :character  Median : 35365  Mode  :character
##  Mean   :2017
##  3rd Qu.:2019
##  Max.   :2020
##          vin          lot         state        country
##  Length:2499      Min.   :159348797  Length:2499      Length:2499
##  Class :character 1st Qu.:167625331  Class :character  Class :character
##  Mode  :character  Median :167745058  Mode  :character  Mode  :character
##                      Mean   :167691389
##                      3rd Qu.:167779772
##                      Max.   :167805500
##          condition
##  Length:2499
##  Class :character
##  Mode  :character
##          .
##          .
##          .

```

Task 2:

#I have used PURRR package in the following section.Purrr package provides an excellent complete and consistent set of tools for working with functions and vectors.I have focused on functions provided by the purrr package of Purr package.

loading the purr library

```
library(purrr)
```

#Functions:

map()

The map functions transform their input by applying a function to each element and returning a vector the same length as the input. Eg1:splits a data frame into pieces, fit a model to each piece and summarise the data

```

cardata %>%
  split(.brand) %>%
  map(~ lm(price ~ mileage, data = .)) %>%
  map(summary)

```

```
## $acura
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    375   391   596
## -6499  3070  3429
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.040e+04 1.256e+04 1.624   0.351
## mileage     -1.091e-01 9.713e-02 -1.123   0.463
##
## Residual standard error: 7964 on 1 degrees of freedom
## Multiple R-squared:  0.5578, Adjusted R-squared:  0.1155
## F-statistic: 1.261 on 1 and 1 DF,  p-value: 0.4631
##
##
## $audi
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    410     421     430     445
## -47.35  976.48 -6732.03  5802.90
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.167e+04 6.806e+03 6.122   0.0257 *
## mileage     -2.344e-01 5.104e-02 -4.593   0.0443 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6323 on 2 degrees of freedom
## Multiple R-squared:  0.9134, Adjusted R-squared:  0.8701
## F-statistic: 21.1 on 1 and 2 DF,  p-value: 0.04428
##
##
## $bmw
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -23896 -12136 -2281  9782  25240
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 3.774e+04 5.117e+03 7.375 2.31e-06 ***
## mileage     -2.371e-01 7.673e-02 -3.090  0.00747 **
```

```
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 14700 on 15 degrees of freedom  
## Multiple R-squared:  0.3889, Adjusted R-squared:  0.3482  
## F-statistic: 9.547 on 1 and 15 DF,  p-value: 0.007471  
##  
##  
## $buick  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
##      Min      1Q Median      3Q      Max  
## -9947.7 -2917.4   189.6  4132.6 12571.8  
##  
## Coefficients:  
##                 Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  2.746e+04  2.877e+03   9.545 1.17e-06 ***  
## mileage     -2.042e-01  5.796e-02  -3.524  0.00476 **  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 6693 on 11 degrees of freedom  
## Multiple R-squared:  0.5303, Adjusted R-squared:  0.4876  
## F-statistic: 12.42 on 1 and 11 DF,  p-value: 0.004763  
##  
##  
## $cadillac  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
##      Min      1Q Median      3Q      Max  
## -11014  -5383  -3232   2706  18050  
##  
## Coefficients:  
##                 Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 38236.0758  6028.3250   6.343 0.000222 ***  
## mileage      -0.3308      0.1278  -2.587 0.032258 *  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 9966 on 8 degrees of freedom  
## Multiple R-squared:  0.4555, Adjusted R-squared:  0.3875  
## F-statistic: 6.693 on 1 and 8 DF,  p-value: 0.03226  
##  
##  
## $chevrolet  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)
```

```

## 
## Residuals:
##   Min     1Q Median     3Q    Max
## -25765 -7312 -2318  6600 38349
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.577e+04 8.715e+02 29.56 <2e-16 ***
## mileage     -1.090e-01 9.782e-03 -11.14 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10250 on 295 degrees of freedom
## Multiple R-squared:  0.2961, Adjusted R-squared:  0.2937
## F-statistic: 124.1 on 1 and 295 DF,  p-value: < 2.2e-16
##
## 
## $chrysler
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##   Min     1Q Median     3Q    Max
## -14031.0 -2512.1   449.1  1753.1  9194.4
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.320e+04 2.114e+03 10.975 7.42e-09 ***
## mileage     -1.303e-01 2.207e-02 -5.905 2.22e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5807 on 16 degrees of freedom
## Multiple R-squared:  0.6855, Adjusted R-squared:  0.6658
## F-statistic: 34.87 on 1 and 16 DF,  p-value: 2.216e-05
##
## 
## $dodge
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##   Min     1Q Median     3Q    Max
## -19352 -4098 -2298  3802 44131
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.454e+04 5.528e+02 44.40 <2e-16 ***
## mileage     -1.530e-01 9.524e-03 -16.07 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##

```

```

## Residual standard error: 7451 on 430 degrees of freedom
## Multiple R-squared:  0.3752, Adjusted R-squared:  0.3737
## F-statistic: 258.2 on 1 and 430 DF,  p-value: < 2.2e-16
##
## $ford
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -27213  -7835  -1316   4546  82472
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.721e+04 4.530e+02 60.07 <2e-16 ***
## mileage     -1.065e-01 5.985e-03 -17.79 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11550 on 1233 degrees of freedom
## Multiple R-squared:  0.2043, Adjusted R-squared:  0.2036
## F-statistic: 316.6 on 1 and 1233 DF,  p-value: < 2.2e-16
##
## $gmc
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -15377  -3599  -408   541  34395
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.538e+04 1.856e+03 8.283 3.28e-10 ***
## mileage     -8.060e-02 2.389e-02 -3.374 0.00166 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7911 on 40 degrees of freedom
## Multiple R-squared:  0.2215, Adjusted R-squared:  0.2021
## F-statistic: 11.38 on 1 and 40 DF,  p-value: 0.001656
##
## $`harley-davidson`
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
## ALL 1 residuals are 0: no residual degrees of freedom!

```

```
##  
## Coefficients: (1 not defined because of singularities)  
## Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 54680 NA NA NA  
## mileage NA NA NA NA  
##  
## Residual standard error: NaN on 0 degrees of freedom  
##  
##  
## $heartland  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## 326 334 343 350 358  
## 1084 -1086 3714 -2966 -746  
##  
## Coefficients: (1 not defined because of singularities)  
## Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 2966 1129 2.627 0.0584 .  
## mileage NA NA NA NA  
## ---  
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 2525 on 4 degrees of freedom  
##  
##  
## $honda  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## Min 1Q Median 3Q Max  
## -5371.0 -1758.1 -1104.7 700.1 10836.0  
##  
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 7263.29815 2280.47169 3.185 0.00974 **  
## mileage -0.01240 0.02108 -0.588 0.56934  
## ---  
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 4206 on 10 degrees of freedom  
## Multiple R-squared: 0.03346, Adjusted R-squared: -0.0632  
## F-statistic: 0.3462 on 1 and 10 DF, p-value: 0.5693  
##  
##  
## $hyundai  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##
```

```

## Residuals:
##      Min     1Q Median     3Q    Max
## -1887 -1372   -541    503   4571
##
## Coefficients:
##                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.272e+03  9.507e+02   5.545 9.46e-05 ***
## mileage     -1.213e-03  1.394e-02  -0.087   0.932
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2049 on 13 degrees of freedom
## Multiple R-squared:  0.0005822, Adjusted R-squared:  -0.0763
## F-statistic: 0.007573 on 1 and 13 DF, p-value: 0.932
##
##
## $infiniti
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##      Min     1Q Median     3Q    Max
## -3190.6 -2576.0   853.7  2124.3  2914.5
##
## Coefficients:
##                  Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.126e+04  1.422e+03   7.919 1.29e-05 ***
## mileage     7.358e-02  4.799e-02   1.533   0.156
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2636 on 10 degrees of freedom
## Multiple R-squared:  0.1903, Adjusted R-squared:  0.1094
## F-statistic: 2.351 on 1 and 10 DF, p-value: 0.1562
##
##
## $jaguar
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
## ALL 1 residuals are 0: no residual degrees of freedom!
##
## Coefficients: (1 not defined because of singularities)
##                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)    2800        NA        NA        NA
## mileage        NA        NA        NA        NA
##
## Residual standard error: NaN on 0 degrees of freedom
##
##
## $jeep

```

```
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
##    Min     1Q Median     3Q    Max  
## -7876  -3110  -1611   4650   9517  
##  
## Coefficients:  
##                 Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 1.343e+04  1.208e+03   11.12 8.77e-12 ***  
## mileage      -6.446e-02  2.053e-02   -3.14  0.00396 **  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 4979 on 28 degrees of freedom  
## Multiple R-squared:  0.2604, Adjusted R-squared:  0.234  
## F-statistic:  9.86 on 1 and 28 DF,  p-value: 0.003961  
##  
##  
## $kia  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
##    Min     1Q Median     3Q    Max  
## -9220.2 -3631.8    16.8  3580.8  8952.2  
##  
## Coefficients:  
##                 Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 1.740e+04  2.999e+03   5.801 0.000119 ***  
## mileage      -1.108e-01  4.503e-02  -2.461 0.031608 *  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 5697 on 11 degrees of freedom  
## Multiple R-squared:  0.3551, Adjusted R-squared:  0.2965  
## F-statistic: 6.058 on 1 and 11 DF,  p-value: 0.03161  
##  
##  
## $land  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
##    452     453     463     476  
##   3206    9925   -1688  -11444  
##  
## Coefficients:  
##                 Estimate Std. Error t value Pr(>|t|)  
## (Intercept) -8322.669  30273.246  -0.275    0.809  
## mileage        1.399       1.119    1.250    0.338
```

```
##  
## Residual standard error: 11010 on 2 degrees of freedom  
## Multiple R-squared:  0.4388, Adjusted R-squared:  0.1581  
## F-statistic: 1.564 on 1 and 2 DF,  p-value: 0.3376  
##  
##  
## $lexus  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## ALL 2 residuals are 0: no residual degrees of freedom!  
##  
## Coefficients:  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 68497.056      NA      NA      NA  
## mileage     -1.576      NA      NA      NA  
##  
## Residual standard error: NaN on 0 degrees of freedom  
## Multiple R-squared:      1, Adjusted R-squared:      NaN  
## F-statistic:  NaN on 1 and 0 DF,  p-value: NA  
##  
##  
## $lincoln  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## ALL 2 residuals are 0: no residual degrees of freedom!  
##  
## Coefficients:  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 43322.4882      NA      NA      NA  
## mileage     -0.4829      NA      NA      NA  
##  
## Residual standard error: NaN on 0 degrees of freedom  
## Multiple R-squared:      1, Adjusted R-squared:      NaN  
## F-statistic:  NaN on 1 and 0 DF,  p-value: NA  
##  
##  
## $maserati  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## ALL 1 residuals are 0: no residual degrees of freedom!  
##  
## Coefficients: (1 not defined because of singularities)  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  30300       NA      NA      NA  
## mileage        NA       NA      NA      NA
```

```
##  
## Residual standard error: NaN on 0 degrees of freedom  
##  
##  
## $mazda  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## ALL 2 residuals are 0: no residual degrees of freedom!  
##  
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 26948.1286 NA NA NA  
## mileage -0.2293 NA NA NA  
##  
## Residual standard error: NaN on 0 degrees of freedom  
## Multiple R-squared: 1, Adjusted R-squared: NaN  
## F-statistic: NaN on 1 and 0 DF, p-value: NA  
##  
##  
## $`mercedes-benz`  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## Min 1Q Median 3Q Max  
## -20359 -10126 -3684 2892 45332  
##  
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 48950.9296 12603.9207 3.884 0.00465 **  
## mileage -0.3708 0.2026 -1.831 0.10453  
## ---  
## Signif. codes: 0 '****' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 19120 on 8 degrees of freedom  
## Multiple R-squared: 0.2952, Adjusted R-squared: 0.2071  
## F-statistic: 3.351 on 1 and 8 DF, p-value: 0.1045  
##  
##  
## $nissan  
##  
## Call:  
## lm(formula = price ~ mileage, data = .)  
##  
## Residuals:  
## Min 1Q Median 3Q Max  
## -15913.1 -3018.4 -530.5 2802.8 21626.5  
##  
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) 1.629e+04 5.252e+02 31.02 <2e-16 ***
## mileage      -9.952e-02 1.014e-02 -9.82 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5326 on 310 degrees of freedom
## Multiple R-squared: 0.2373, Adjusted R-squared: 0.2348
## F-statistic: 96.43 on 1 and 310 DF, p-value: < 2.2e-16
##
##
## $peterbilt
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
##    491     505     517     529
##   -2.347   18.119  -512.615  496.843
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 8.188e+01 5.031e+02 0.163 0.886
## mileage     4.384e-04 5.997e-04 0.731 0.541
##
## Residual standard error: 505 on 2 degrees of freedom
## Multiple R-squared: 0.2109, Adjusted R-squared: -0.1837
## F-statistic: 0.5345 on 1 and 2 DF, p-value: 0.5408
##
##
## $ram
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
## ALL 1 residuals are 0: no residual degrees of freedom!
##
## Coefficients: (1 not defined because of singularities)
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 11050        NA        NA        NA
## mileage       NA        NA        NA        NA
##
## Residual standard error: NaN on 0 degrees of freedom
##
##
## $toyota
##
## Call:
## lm(formula = price ~ mileage, data = .)
##
## Residuals:
## ALL 1 residuals are 0: no residual degrees of freedom!
##
## Coefficients: (1 not defined because of singularities)
```

```
##             Estimate Std. Error t value Pr(>|t|) 
## (Intercept)     6300        NA        NA        NA 
## mileage          NA        NA        NA        NA 
## 
## Residual standard error: NaN on 0 degrees of freedom
```

#map_lgl() The map_lgl() will return a logical vector. This map_lgl() checks according to the function given to it. In first example, we check whether cardata has any Double data type. In the second example, map_lgl() checks whether dataset has integer values or not, if present shows as TRUE else false. Map function always returns lists.

Eg1:

```
map_lgl(cardata, is.double)
```

	X	price	brand	model	year	title_status
##	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	mileage	color	vin	lot	state	country
##	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE
##	condition					
##	FALSE					

Eg2:

```
map_lgl(cardata, is.integer)
```

	X	price	brand	model	year	title_status
##	TRUE	TRUE	FALSE	FALSE	TRUE	FALSE
##	mileage	color	vin	lot	state	country
##	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE
##	condition					
##	FALSE					

map_chr()

map_chr() will return a character vector

```
map_chr(cardata, typeof)
```

	X	price	brand	model	year	title_status
##	"integer"	"integer"	"character"	"character"	"integer"	"character"
##	mileage	color	vin	lot	state	country
##	"double"	"character"	"character"	"integer"	"character"	"character"
##	condition					
##	"character"					

as_mapper()

The `as_mapper()` function arguments will allow you to refer to `.x` for one argument functions, `.x` and `.y` for two argument functions, and `..1`, `..2`, `..3`, etc, for functions with an arbitrary number of arguments.

```
as_mapper(~ length(unique(.cardata)))
```

```
## <lambda>
## function (... , .x = ..1, .y = ..2, . = ..1)
## length(unique(.cardata))
## attr(,"class")
## [1] "rlang_lambda_function" "function"
```

map_if

`map_if()` maps a function over the elements of dataset satisfying a predicate.

Eg1: Convert integers to characters

```
cardata %>%
  map_if(is.integer, as.character) %>%
  str()
```

```
## List of 13
## $ X          : chr [1:2499] "0" "1" "2" "3" ...
## $ price      : chr [1:2499] "6300" "2899" "5350" "25000" ...
## $ brand      : chr [1:2499] "toyota" "ford" "dodge" "ford" ...
## $ model      : chr [1:2499] "cruiser" "se" "mpv" "door" ...
## $ year       : chr [1:2499] "2008" "2011" "2018" "2014" ...
## $ title_status: chr [1:2499] "clean vehicle" "clean vehicle" "clean vehicle" "clean vehicle"
...
## $ mileage    : num [1:2499] 274117 190552 39590 64146 6654 ...
## $ color      : chr [1:2499] "black" "silver" "silver" "blue" ...
## $ vin         : chr [1:2499] "jtezu11f88k007763" "2fmdk3gc4bbb02217" "3c4pdccg5jt34641
3" "1ftfw1et4efc23745" ...
## $ lot         : chr [1:2499] "159348797" "166951262" "167655728" "167753855" ...
## $ state       : chr [1:2499] "new jersey" "tennessee" "georgia" "virginia" ...
## $ country     : chr [1:2499] "usa" "usa" "usa" "usa" ...
## $ condition   : chr [1:2499] "10 days left" "6 days left" "2 days left" "22 hours left" ...
```

map_at

This function will modify the elements corresponding to a character vector of names or a numeric vector of positions. Eg1: Specify which columns to map with a numeric vector of positions:

```
cardata %>%
  map_at(c(5, 9, 10), as.character) %>% #year,vin and Lot
  str()
```

```
## List of 13
## $ X           : int [1:2499] 0 1 2 3 4 5 6 7 8 9 ...
## $ price       : int [1:2499] 6300 2899 5350 25000 27700 5700 7300 13350 14600 5250 ...
## $ brand       : chr [1:2499] "toyota" "ford" "dodge" "ford" ...
## $ model       : chr [1:2499] "cruiser" "se" "mpv" "door" ...
## $ year        : chr [1:2499] "2008" "2011" "2018" "2014" ...
## $ title_status: chr [1:2499] "clean vehicle" "clean vehicle" "clean vehicle" "clean vehicle"
...
## $ mileage     : num [1:2499] 274117 190552 39590 64146 6654 ...
## $ color       : chr [1:2499] "black" "silver" "silver" "blue" ...
## $ vin         : chr [1:2499] "jtezu11f88k007763" "2fmdk3gc4bbb02217" "3c4pdccg5jt34641
3" "1ftfw1et4efc23745" ...
## $ lot          : chr [1:2499] "159348797" "166951262" "167655728" "167753855" ...
## $ state        : chr [1:2499] "new jersey" "tennessee" "georgia" "virginia" ...
## $ country      : chr [1:2499] "usa" "usa" "usa" "usa" ...
## $ condition    : chr [1:2499] "10 days left" "6 days left" "2 days left" "22 hours left" ...
```

Eg2: Specify which columns to map with a vector of names:

```
cardata %>% map_at(c("year", "condition"), as.character) %>% str()
```

```
## List of 13
## $ X           : int [1:2499] 0 1 2 3 4 5 6 7 8 9 ...
## $ price       : int [1:2499] 6300 2899 5350 25000 27700 5700 7300 13350 14600 5250 ...
## $ brand       : chr [1:2499] "toyota" "ford" "dodge" "ford" ...
## $ model       : chr [1:2499] "cruiser" "se" "mpv" "door" ...
## $ year        : chr [1:2499] "2008" "2011" "2018" "2014" ...
## $ title_status: chr [1:2499] "clean vehicle" "clean vehicle" "clean vehicle" "clean vehicle"
...
## $ mileage     : num [1:2499] 274117 190552 39590 64146 6654 ...
## $ color       : chr [1:2499] "black" "silver" "silver" "blue" ...
## $ vin         : chr [1:2499] "jtezu11f88k007763" "2fmdk3gc4bbb02217" "3c4pdccg5jt34641
3" "1ftfw1et4efc23745" ...
## $ lot          : int [1:2499] 159348797 166951262 167655728 167753855 167763266 167655771 167
753872 167692494 167763267 167656121 ...
## $ state        : chr [1:2499] "new jersey" "tennessee" "georgia" "virginia" ...
## $ country      : chr [1:2499] "usa" "usa" "usa" "usa" ...
## $ condition    : chr [1:2499] "10 days left" "6 days left" "2 days left" "22 hours left" ...
```

map_dfr

If we want each element of the output to be data frame, we can use `map_dfr` to row-bind them together:

```
cardata %>%
  split(.brand) %>%
  map(~ lm(price ~ mileage, data = .x)) %>%
  map_dfr(~ as.data.frame(t(as.matrix(coef(.)))))
```

```

##      (Intercept)      mileage
## 1  20398.23838 -0.1090846326
## 2   41665.19091 -0.2344288803
## 3   37740.13957 -0.2370727570
## 4   27462.09596 -0.2042438935
## 5   38236.07584 -0.3307570136
## 6   25765.35281 -0.1089513803
## 7   23200.29356 -0.1303241254
## 8   24544.20937 -0.1530438349
## 9   27213.10155 -0.1064853097
## 10  15376.57227 -0.0806027845
## 11  54680.00000          NA
## 12  2966.00000          NA
## 13  7263.29815 -0.0123996785
## 14  5271.94860 -0.0012128424
## 15 11257.14399  0.0735793901
## 16  2800.00000          NA
## 17 13430.06958 -0.0644634728
## 18 17397.65093 -0.1108260810
## 19 -8322.66920  1.3986254924
## 20 68497.05597 -1.5755015839
## 21 43322.48816 -0.4829439625
## 22 30300.00000          NA
## 23 26948.12862 -0.2292657764
## 24 48950.92956 -0.3708432616
## 25 16288.17633 -0.0995222941
## 26   81.88076  0.0004384128
## 27 11050.00000          NA
## 28  6300.00000          NA

```

Task 3:

Creating an S3 object

```

s <- list(brand = 'Ford', price = 10300, model = "fiesta", year='2019' , title_status = "clean vehicle" ,
          mileage = 18937 , color= "black", vin = "3fadp4bjxkm123238" , lot= 167759377, state="missouri")

class(s) <- 'cardat1'
s # or print(j)

```

```

## $brand
## [1] "Ford"
##
## $price
## [1] 10300
##
## $model
## [1] "fiesta"
##
## $year
## [1] "2019"
##
## $title_status
## [1] "clean vehicle"
##
## $mileage
## [1] 18937
##
## $color
## [1] "black"
##
## $vin
## [1] "3fadp4bjxkm123238"
##
## $lot
## [1] 167759377
##
## $state
## [1] " missouri"
##
## attr(,"class")
## [1] "cardat1"

```

attributes(s)

```

## $names
## [1] "brand"          "price"        "model"        "year"        "title_status"
## [6] "mileage"        "color"        "vin"          "lot"          "state"
##
## $class
## [1] "cardat1"

```

print method

```
print.cardat1 <- function(wkr) {
  cat(wkr$brand, '\n')
  cat('Price', wkr$price, '\n')
  cat('model', wkr$model, '\n')
  cat('year', wkr$year, '\n')
  cat('Car_status', wkr$title_status, '\n')
  cat('Car_mileage', wkr$mileage, '\n')
  cat('color', wkr$color, '\n')
  cat('vin', wkr$vin, '\n')
  cat('lot', wkr$lot, '\n')
  cat('state', wkr$state, '\n')
}
print.cardat1(s)
```

```
## Ford
## Price 10300
## model fiesta
## year 2019
## Car_status clean vehicle
## Car_mileage 18937
## color black
## vin 3fadp4bjxkm123238
## lot 167759377
## state missouri
```

#Function gives the list of available cars when given the price is \$4200

```
dat<-cardata
class(dat)<-"Car Details"
print.car<-function(p)
{
  {cat("Data about particular given price($)", p, "\n")
  print("\n")
  p1<-filter(cardata, cardata$price == p)
  print(p1)
  }
}
print.car(4200)
```

```
## Data about particular given price($) 4200
## [1] "\n"
##      X price   brand model year     title_status mileage color
## 1 193 4200    ford   door 2017    clean vehicle  60219  red
## 2 493 4200 chevrolet cutaway 2006    clean vehicle 118009 white
## 3 763 4200 chevrolet   door 2014    clean vehicle 102500 black
## 4 901 4200    dodge   door 2014    clean vehicle 144725 gray
## 5 1227 4200    ford   door 2016    clean vehicle 104383 blue
## 6 1236 4200    ford   door 2013    clean vehicle       0 no_color
## 7 1848 4200    ford cutaway 2015 salvage insurance  80032 color:
##                  vin      lot      state country condition
## 1 3fa6p0hd7hr273550 167656522    texas     usa 2 days left
## 2 1gbhg31u661113654 167781216  wisconsin     usa 21 hours left
## 3 1g1p75sz3e7332279 167799775    nevada     usa 2 days left
## 4 2c3cdxbg3eh170784 167519895    nevada     usa 22 hours left
## 5 3fa6p0hd5gr285811 167773291 pennsylvania     usa 2 days left
## 6 1fadp3j23dl155179 167773673 pennsylvania     usa 2 days left
## 7 1fdf4fs3fda25516 167529792    illinois     usa 16 hours left
```

#summary method

```
print.summary1<-function(pr)
{
  print(summary(print.car(pr)))
}
print.summary1(4200)
```

```

## Data about particular given price($) 4200
## [1] "\n"
##      X price   brand model year     title_status mileage color
## 1 193 4200    ford   door 2017    clean vehicle  60219  red
## 2 493 4200 chevrolet cutaway 2006    clean vehicle 118009 white
## 3 763 4200 chevrolet   door 2014    clean vehicle 102500 black
## 4 901 4200    dodge   door 2014    clean vehicle 144725 gray
## 5 1227 4200    ford   door 2016    clean vehicle 104383 blue
## 6 1236 4200    ford   door 2013    clean vehicle      0 no_color
## 7 1848 4200    ford cutaway 2015 salvage insurance  80032 color:
##                  vin      lot      state country condition
## 1 3fa6p0hd7hr273550 167656522      texas    usa 2 days left
## 2 1gbhg31u661113654 167781216      wisconsin    usa 21 hours left
## 3 1g1p75sz3e7332279 167799775      nevada    usa 2 days left
## 4 2c3cdxbg3eh170784 167519895      nevada    usa 22 hours left
## 5 3fa6p0hd5gr285811 167773291      pennsylvania    usa 2 days left
## 6 1fadp3j23dl155179 167773673      pennsylvania    usa 2 days left
## 7 1fdf4fs3fda25516 167529792      illinois    usa 16 hours left
##      X          price      brand      model
## Min. :193.0  Min. :4200  Length:7      Length:7
## 1st Qu.:628.0 1st Qu.:4200  Class :character  Class :character
## Median :901.0 Median :4200  Mode  :character  Mode  :character
## Mean   :951.6 Mean   :4200
## 3rd Qu.:1231.5 3rd Qu.:4200
## Max.  :1848.0  Max.  :4200
##      year     title_status     mileage      color
## Min. :2006  Length:7      Min. :     0  Length:7
## 1st Qu.:2014  Class :character 1st Qu.: 70126  Class :character
## Median :2014  Mode  :character  Median :102500  Mode  :character
## Mean   :2014
## 3rd Qu.:2016
## Max.  :2017      Mean   : 87124
##      vin      lot      state      country
## Length:7      Min. :167519895  Length:7      Length:7
## Class :character 1st Qu.:167593157  Class :character  Class :character
## Mode  :character  Median :167773291  Mode  :character  Mode  :character
## Mean   :167690595
## 3rd Qu.:167777444
## Max.  :167799775
##      condition
## Length:7
## Class :character
## Mode  :character
## 
## 
## 
```

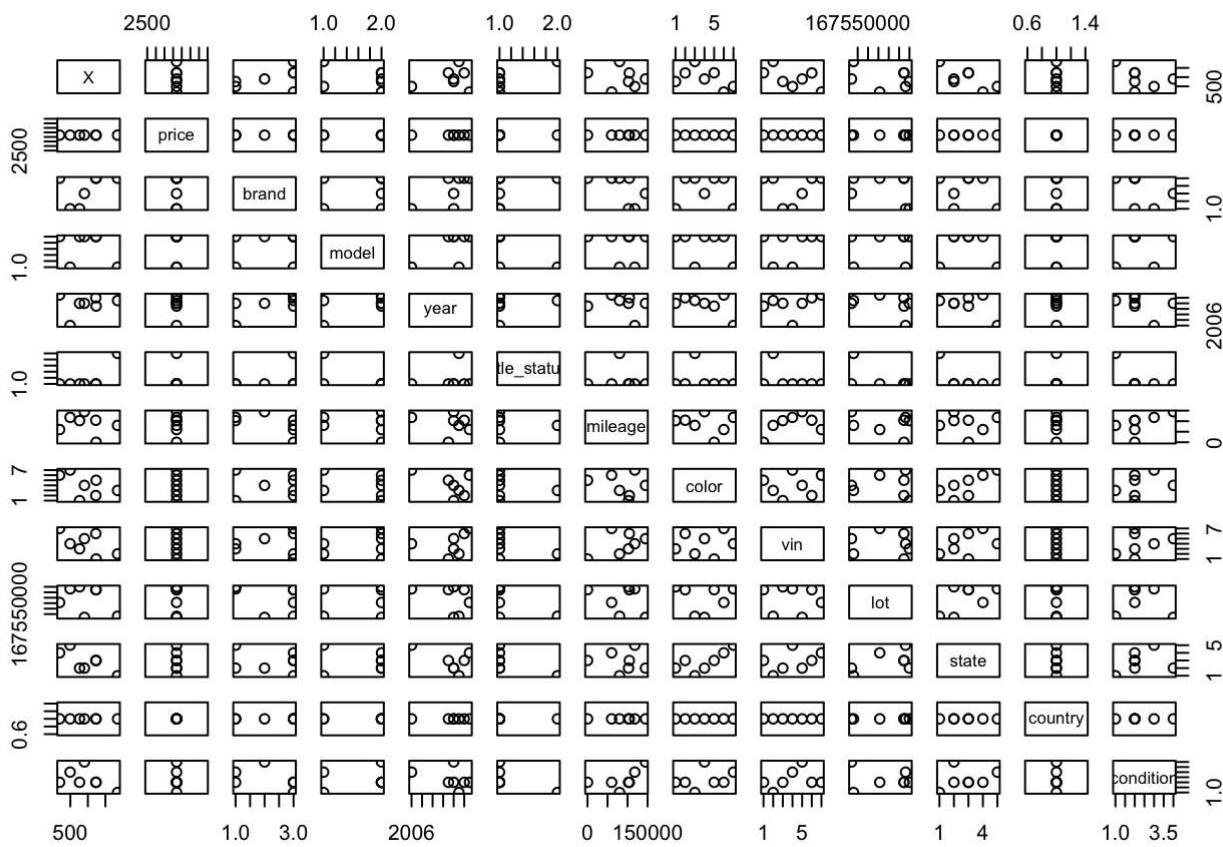
#plot method

```
plot(print.car(4200))
```

```

## Data about particular given price($) 4200
## [1] "\n"
##      X price   brand model year    title_status mileage color
## 1  193 4200     ford   door 2017    clean vehicle  60219  red
## 2  493 4200 chevrolet cutaway 2006    clean vehicle 118009 white
## 3  763 4200 chevrolet   door 2014    clean vehicle 102500 black
## 4  901 4200     dodge   door 2014    clean vehicle 144725 gray
## 5 1227 4200     ford   door 2016    clean vehicle 104383 blue
## 6 1236 4200     ford   door 2013    clean vehicle       0 no_color
## 7 1848 4200     ford cutaway 2015 salvage insurance  80032 color:
##                  vin     lot state country condition
## 1 3fa6p0hd7hr273550 167656522    texas   usa 2 days left
## 2 1gbhg31u661113654 167781216  wisconsin   usa 21 hours left
## 3 1g1p75sz3e7332279 167799775    nevada   usa 2 days left
## 4 2c3cdxbgb3eh170784 167519895    nevada   usa 22 hours left
## 5 3fa6p0hd5gr285811 167773291 pennsylvania   usa 2 days left
## 6 1fadp3j23dl155179 167773673 pennsylvania   usa 2 days left
## 7 1fdfef4fs3fda25516 167529792 illinois   usa 16 hours left

```



#help document

```
install.packages("devtools") install.packages("roxygen2") library(devtools) create("myRpackage")
```

```
#' @param p A number. #' @param p1 A vector #' @return Cars of specified price value #' @examples #' print.car(4200) print.car<-function(p) { p1<-filter(cardata,cardata$price == p) print(p1) }
```