Assignment-EDA-2

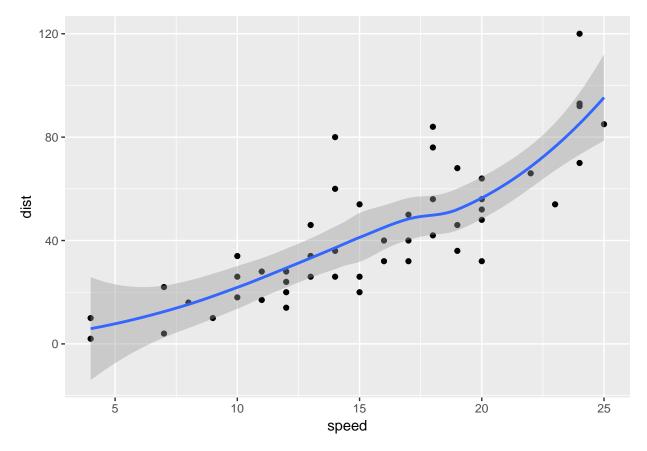
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Solution 1 First we are drawing the scatterplot and adding smoth curve to plot

```
## Warning: package 'ggplot2' was built under R version 3.3.3
cars.gg=ggplot(cars,aes(x=speed,y=dist))+geom_point()+geom_smooth()
cars.gg
```

`geom_smooth()` using method = 'loess'

library("ggplot2")



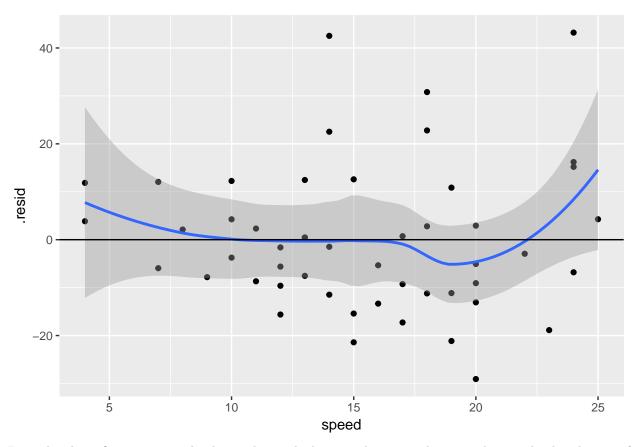
here the code which is used to model simple linear model

```
library(ggplot2)
library(broom)
```

```
## Warning: package 'broom' was built under R version 3.3.3
cars.lm=lm(dist~speed,data=cars)
#install.packages('broom')
car.lm.df=augment(cars.lm)
```

```
## Warning: Deprecated: please use `purrr::possibly()` instead
summary(car.lm.df)
##
         dist
                         speed
                                        .fitted
                                                          .se.fit
##
   Min.
          : 2.00
                     Min.
                            : 4.0
                                     Min.
                                            :-1.849
                                                      Min.
                                                              :2.181
    1st Qu.: 26.00
                     1st Qu.:12.0
                                     1st Qu.:29.610
                                                       1st Qu.:2.393
   Median : 36.00
                     Median :15.0
##
                                     Median :41.407
                                                       Median :2.640
##
    Mean
          : 42.98
                     Mean
                           :15.4
                                     Mean
                                           :42.980
                                                       Mean
                                                              :2.967
##
    3rd Qu.: 56.00
                     3rd Qu.:19.0
                                     3rd Qu.:57.137
                                                       3rd Qu.:3.358
##
    Max.
           :120.00
                             :25.0
                                            :80.731
                                                       Max.
                                                              :5.212
##
        .resid
                            .hat
                                                             .cooksd
                                              .sigma
           :-29.069
                              :0.02012
                                                                 :0.0000113
##
   Min.
                      Min.
                                         Min.
                                                :14.10
                                                         Min.
##
   1st Qu.: -9.525
                      1st Qu.:0.02420
                                         1st Qu.:15.37
                                                          1st Qu.:0.0017903
  Median : -2.272
                      Median :0.02946
                                         Median :15.47
                                                          Median: 0.0069914
          : 0.000
##
   Mean
                      Mean
                            :0.04000
                                         Mean :15.38
                                                          Mean
                                                                 :0.0210046
                                         3rd Qu.:15.53
                                                          3rd Qu.:0.0191012
##
    3rd Qu.: 9.215
                      3rd Qu.:0.04774
##
    Max.
           : 43.201
                      Max. :0.11486
                                               :15.54
                                                                 :0.3403959
                                         {\tt Max.}
                                                          Max.
##
      .std.resid
           :-1.924523
##
  Min.
##
  1st Qu.:-0.627833
## Median :-0.151050
          : 0.002765
##
  Mean
    3rd Qu.: 0.610374
##
  Max.
           : 2.919060
now we are plot stopping distance(response variable) the speed (explantory variable) then add a loess curve if
the confidence band contains the line y = 0, then maybe the model is fitting well.
ggplot(car.lm.df,aes(x=speed,y=.resid))+geom_point()+geom_smooth()+geom_abline(slope=0,intercept = 0)
## `geom_smooth()` using method = 'loess'
```

Warning: Deprecated: please use `purrr::possibly()` instead

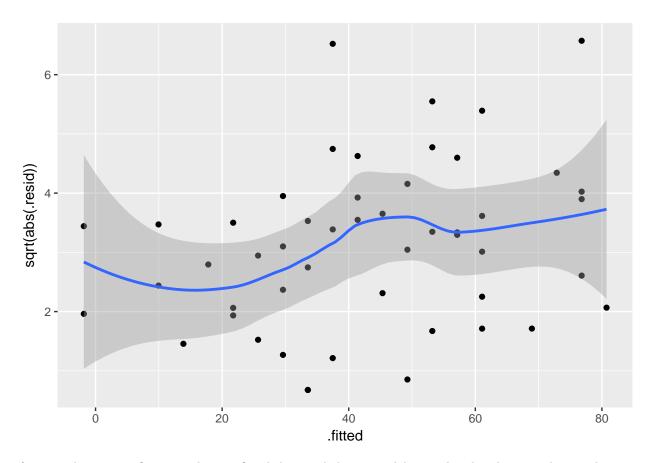


From the above figure, we can clearly see the residuals is wiggling around zero, and given the distribution of residual with respect to speed, we can say that the simple linear model fits.

more option we will check homosedasticity

```
ggplot(car.lm.df,aes(x=.fitted,y=sqrt(abs(.resid))))+geom_point()+geom_smooth()
```

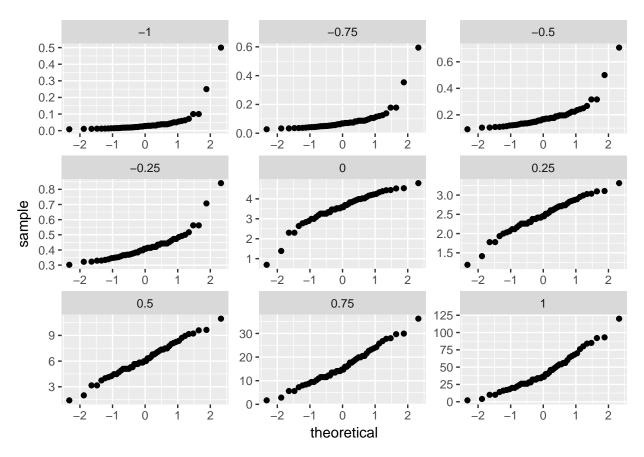
`geom_smooth()` using method = 'loess'



if we see above given figure its clear its fitted the simple linear model even clear by plot using homosedasticity check processor.

Solution 2 : Using ggplot() to reproduce the normal Q-Q plots on the next page for different power transformations of stopping distance.

```
dissort=sort(cars$dist)
n=length(dissort)
# here the length is equal
power=rep(seq(-1,1,0.25),each=n)
dsort=c(dissort^-1,dissort^-0.75,dissort^-0.5,dissort^-0.25,log(dissort),dissort^0.25,dissort^0.5,dissort
ggplot(data.frame(power,dsort), aes(sample = dsort)) + stat_qq() + facet_wrap(~power,scales = "free")
```



if we look above given qqplot and power transformation of Cars\$dist with vlaue 1,0.75,0.5,0.25 look a normal distrution.

more the 0.5 poewr transformation look so prefect as normal distribution.

q3 Based on the normal Q-Q plots of 0.5 power transformation that looks the best. and i am Use the transformed stopping distance to fit the linear model again and comparing on whether the model fits better after transformation or not.

```
library(ggplot2)
library(gridExtra)
cars.lm=lm(cars$dist~cars$speed)
car.lm.df=augment(cars.lm)

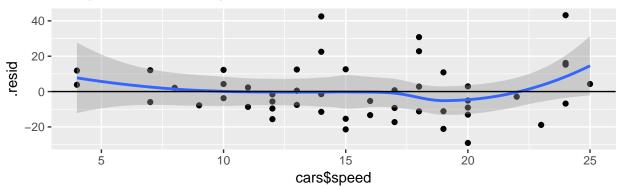
## Warning: Deprecated: please use `purrr::possibly()` instead

a=ggplot(car.lm.df, aes(x = cars$speed, y = .resid)) + geom_point() + geom_smooth() + geom_abline(slope = 0, intercept = 0)+ggtitle(" this plot show before pow transformatin")
car.p05.lm = lm((dissort)^0.5 ~ cars$speed)
car.p05.lm.df = augment(car.p05.lm)
```

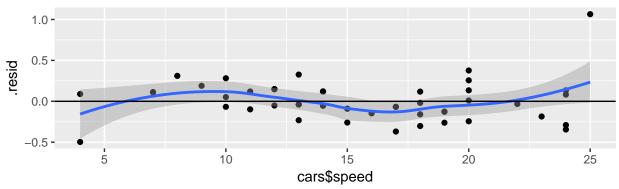
```
## Warning: Deprecated: please use `purrr::possibly()` instead
b=ggplot(car.p05.lm.df, aes(x = cars$speed, y = .resid)) + geom_point() + geom_smooth() +
geom_abline(slope = 0, intercept = 0)+ggtitle("this plot show After power transformation")
grid.arrange(a,b, nrow=2)
```

```
## `geom_smooth()` using method = 'loess'
## `geom_smooth()` using method = 'loess'
```

this plot show before pow transformatin



this plot show After power transformation

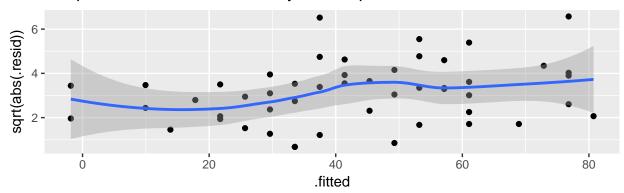


if we see the above given plot the residual wiggle around 0 so we can say it fitts linear model for both cases. now we are going to check with homoscedisity check

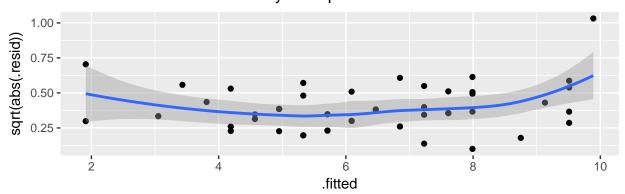
```
c <- ggplot(car.lm.df, aes(x = .fitted, y = sqrt(abs(.resid)))) + geom_point() +
geom_smooth()+ggtitle("this plot show the Homoscedisity before power transformation")
d<- ggplot(car.p05.lm.df, aes(x = .fitted, y = sqrt(abs(.resid)))) + geom_point() +
geom_smooth()+ggtitle(" this show the Homoscedisity after power transformation")
grid.arrange(c,d, nrow=2)</pre>
```

```
## `geom_smooth()` using method = 'loess'
## `geom_smooth()` using method = 'loess'
```

this plot show the Homoscedisity before power transformation



this show the Homoscedisity after power transformation



if we see the above given plot the plot without power transformation which is like horizontal line so we can this homoscedasicity is not correct method for this cases. for with power transformation its do not like horizontial line so the is ok for homoscadesity is correct for this. so we are say model fit linear and homescadesity for power transformation.

we are going to check residual fit plot for two cases

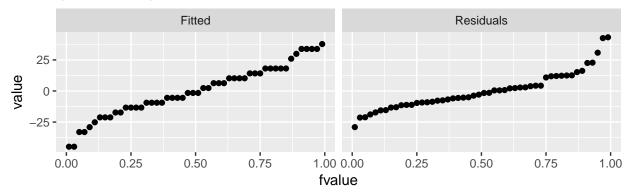
```
require(gridExtra)
l = nrow(car.lm.df)
fvalue = (0.5:(1 - 0.5))/l
car.ft = data.frame(fvalue, Fitted = sort(car.lm.df$.fitted) - mean(car.lm.df$.fitted),
Residuals = sort(car.lm.df$.resid))
library(tidyr)
```

```
## Warning: package 'tidyr' was built under R version 3.3.3

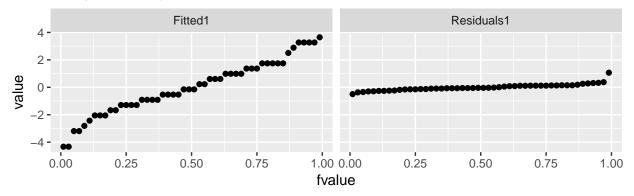
car.ft.lo = car.ft %>% gather(type, value, Fitted:Residuals)
p1 <- ggplot(car.ft.lo, aes(x = fvalue, y = value)) +
geom_point() + facet_wrap(~type) + ggtitle(" plot Before power transformation")
l1 = nrow(car.p05.lm.df)
fivalue = (0.5:(1 - 0.5))/l1
car.pt.ft = data.frame(fvalue, Fitted1 = sort(car.p05.lm.df$.fitted) - mean(car.p05.lm.df$.fitted),
Residuals1 = sort(car.p05.lm.df$.resid))
library(tidyr)
car.pt.ft.lo = car.pt.ft %>% gather(type, value, Fitted1:Residuals1)
p2 <- ggplot(car.pt.ft.lo, aes(x = fvalue, y = value)) +
geom_point() + facet_wrap(~type) +
ggtitle(" the plot After power transformation of 0.5")</pre>
```

grid.arrange(p1,p2,nrow=2)

plot Before power transformation



the plot After power transformation of 0.5



in with out the transformation both fitted and residual look normal in after transformation fitted and residual is different so linear model fit in befor transformation.