

Statistical Learning Laboratory 1

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The dataset that we have used to solve the given task is part 12.

Part 2 :

1. Shapiro-wilk Normality test

data: df\$density

W = 0.75457, p-value = 9.117e-08

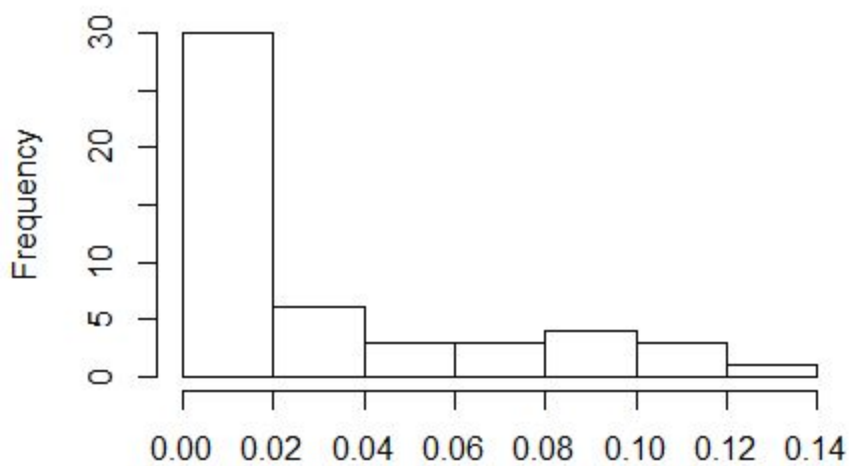
Comment-- it is not a normal dist as $p < 0.005$

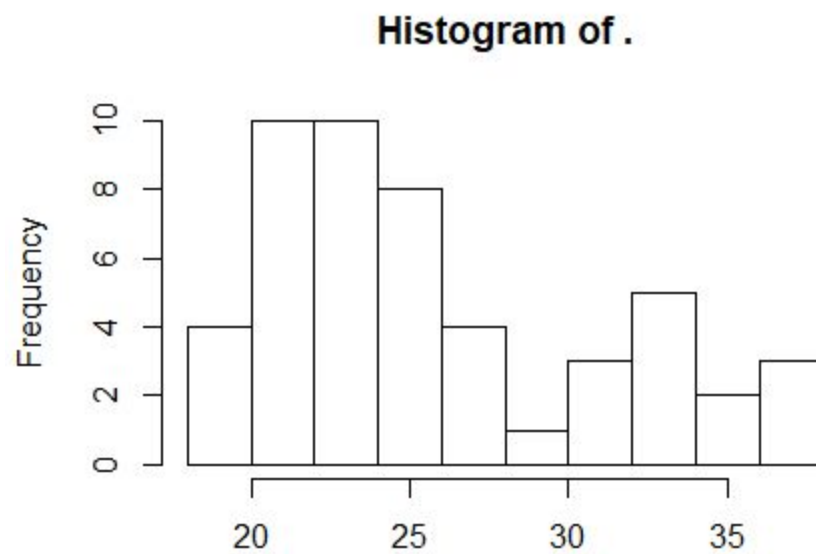
data: df\$concentration

W = 0.89393, p-value = 0.0003046

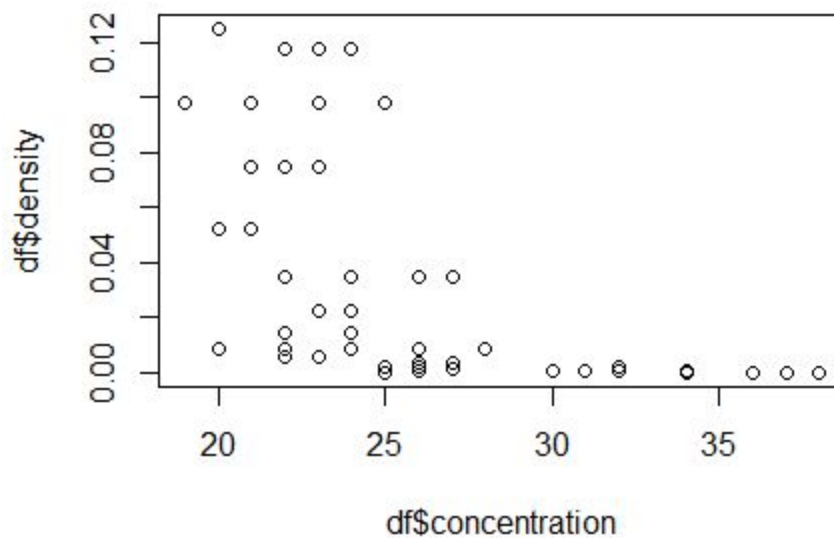
Comment--it is not a normal dist as $p < 0.005$

Histogram of .





2. Scatterplot



- because the output of p value is $7.737e-06 < 0.05$ we can say that correlation exist
correlation value is -0.5861335.
because the output of p value is $1.45e-12 < 0.05$ we can say that correlation exist
correlation value is 0.8071393.

4. comment on scatterplot and correlation
as it can be seen from the scatterplot and the correlation values the variables are negatively correlated to each other and can be written as a negative linear relationship between the two variables.

5. Coefficient of linear data

Coefficients:

(Intercept) df\$density

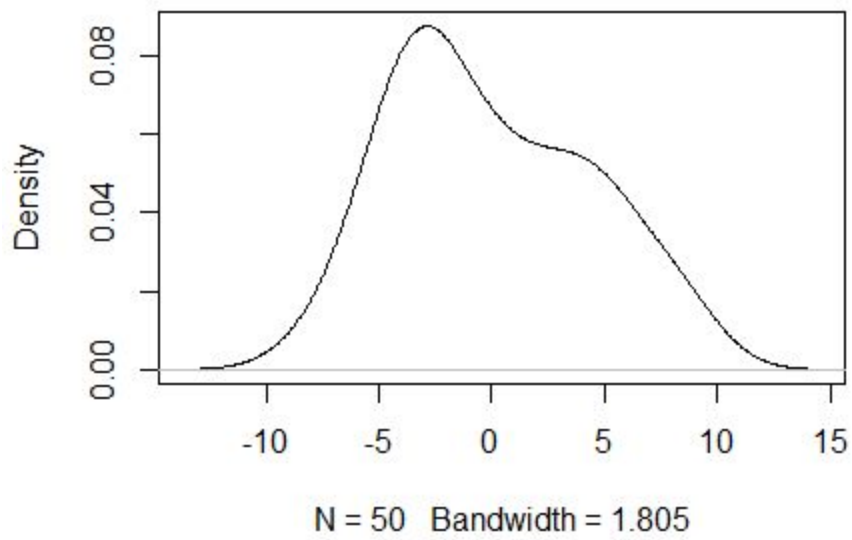
28.87 -78.47

coeff = -78.47

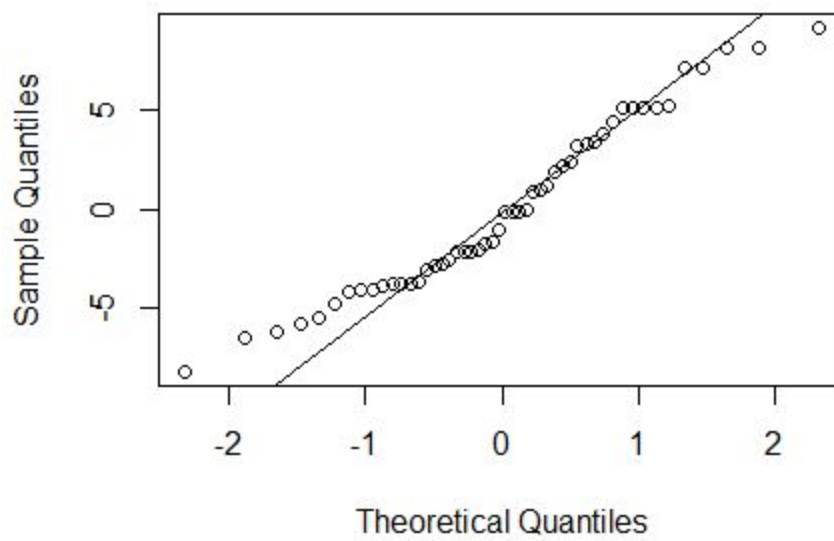
Residual Plot

```
> resid(model) #List of residuals
      1      2      3      4      5      6      7      8      9
-2.1238092 -1.0226672  5.1614262 -0.1603812 -4.1018370 -1.7775256 -2.7152218  5.1473804  1.8396188
     10     11     12     13     14     15     16     17     18
-3.7549314 -6.1813241 -3.1018370 -0.0226672 -4.7549314  4.3458945 -6.4479591 -4.1238092  7.1337269
     19     20     21     22     23     24     25     26     27
 8.1337269 -2.1813241  3.2847782  1.1473804  7.1337269 -5.4479591 -0.1238092 -3.8713735 -8.1813241
     28     29     30     31     32     33     34     35     36
 2.1844958  0.9342491 -3.7549314 -1.6134485 -2.8155042  3.3458945 -0.1813241  3.1614262 -4.1813241
     37     38     39     40     41     42     43     44     45
 5.1473804  3.8396188  5.1258016  0.8761908  5.1275279 -5.7579097 -3.7579097 -2.0226672  9.1258016
     46     47     48     49     50
 8.1268217  2.3458945 -3.7152218 -2.6134485 -2.1603812
```

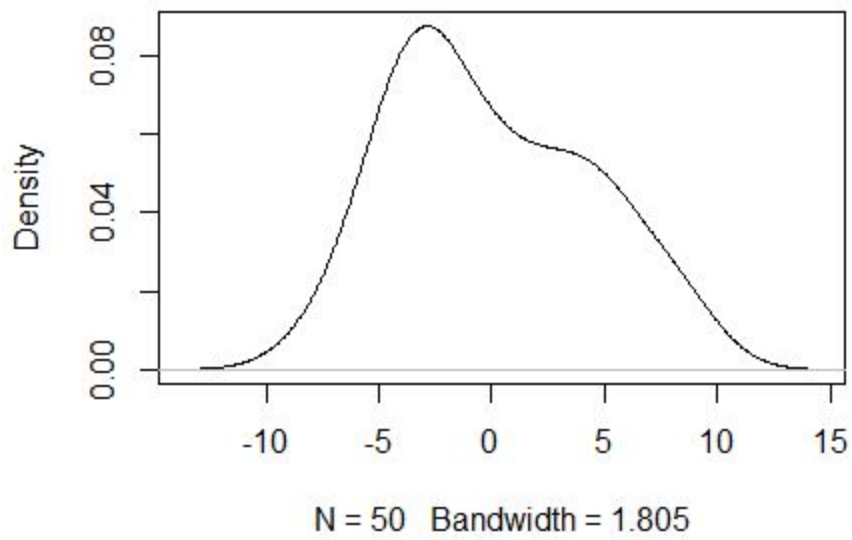
density.default(x = resid(model))



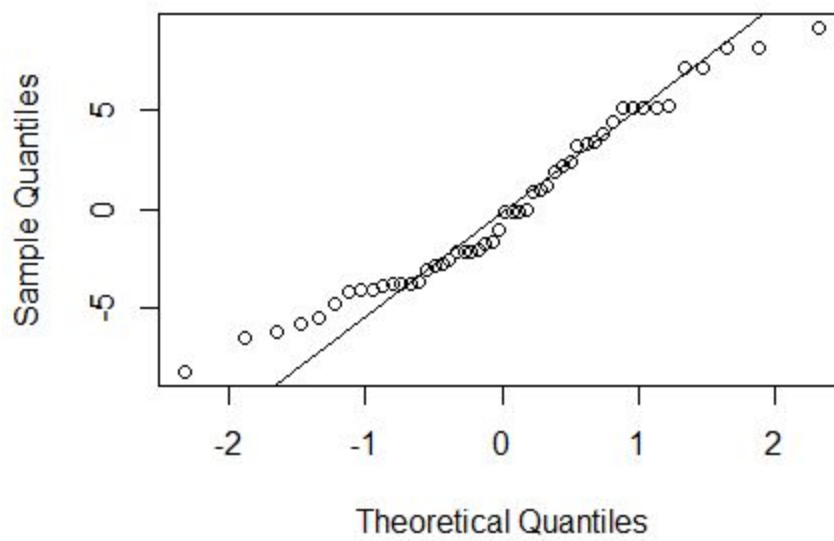
Normal Q-Q Plot



density.default(x = resid(model))



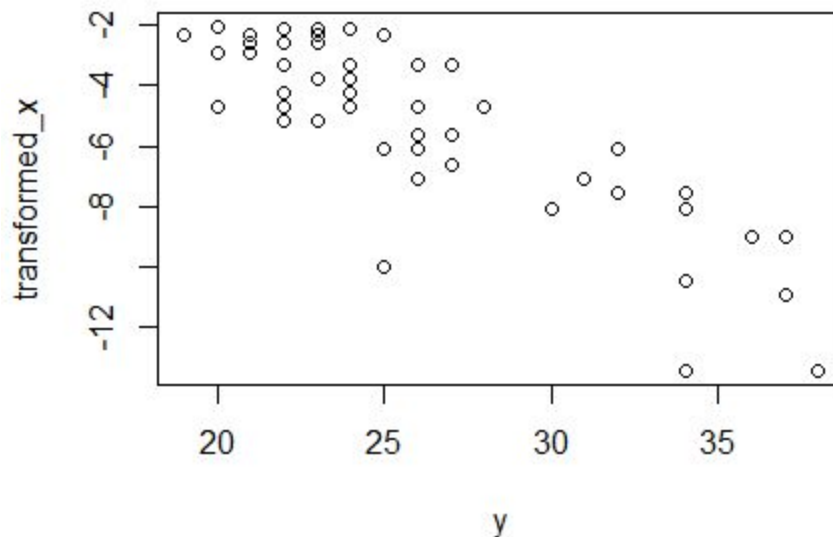
Normal Q-Q Plot



```

> y<-df$concentration
> x<-df$density
> x
[1] 0.035052 0.074573 0.000455 0.098306 0.022588 0.001233 0.002027 0.000276 0.098306 0.052497 0.008831
[12] 0.022588 0.074573 0.052497 0.117502 0.005433 0.035052 0.000102 0.000102 0.008831 0.002027 0.000276
[23] 0.000102 0.005433 0.035052 0.000037 0.008831 0.000749 0.125000 0.052497 0.003324 0.000749 0.117502
[34] 0.008831 0.000455 0.008831 0.000276 0.098306 0.000001 0.035052 0.000023 0.014227 0.014227 0.074573
[45] 0.000001 0.000014 0.117502 0.002027 0.003324 0.098306
> y
[1] 24 22 34 21 23 27 26 34 23 21 22 24 23 20 24 22 22 36 37 26 32 30 36 23 26 25 20 31 20 21 27 26 23
[34] 28 32 24 34 25 34 27 34 22 24 21 38 37 22 25 26 19
>

```



7. Pearson correlation coefficient

Because the output of p value is $6.695e-15 < 0.05$ we can say that correlation exist correlation value is -0.849047.

```
> pearson_corr
```

Pearson's product-moment correlation

data: y and transformed_x

t = -11.134, df = 48, p-value = $6.695e-15$

alternative hypothesis: true correlation is not equal to 0

95 percent confidence interval:

-0.9118880 -0.7473117

sample estimates:

cor

-0.849047

Spearman rank correlation coefficient

Because the output of p value is $1.45e-12 < 0.05$ we can say that correlation exist correlation value is -0.8071393.

```
> spearman_corr

Spearman's rank correlation rho

data: y and transformed_x
S = 37634, p-value = 1.45e-12
alternative hypothesis: true rho is not equal to 0
sample estimates:
      rho
-0.8071393
```

8.

a.

```
> cor.diff.test(x,transformed_x,y,y,method="pearson") #no sign difference
cor1: r=0.726, p=2.49e-09, n=50
cor2: r=1, p=0, n=50
diffence: p(one-sided)=0, p(two-sided)=0
$cor1
$cor1$estimate
[1] 0.7256493

$cor1$p.value
[1] 2.494812e-09

$cor1$n
[1] 50

$cor2
$cor2$estimate
[1] 1

$cor2$p.value
[1] 0

$cor2$n
[1] 50

$p.value.twosided
[1] 0

$p.value.onesided
[1] 0
```

B.

```

> cor.diff.test(x,transformed_x,y,y,method = 'spearman')
cor1: r=1, p=0, n=50
cor2: r=1, p=0, n=50
diffence: p(one-sided)=0, p(two-sided)=0
$cor1
$cor1$estimate
[1] 1

$cor1$p.value
[1] 0

$cor1$n
[1] 50

$cor2
$cor2$estimate
[1] 1

$cor2$p.value
[1] 0

$cor2$n
[1] 50

$p.value.twosided
[1] 0

$p.value.onesided
[1] 0

```

9.

Linear model after transformation

```

call:
lm(formula = y ~ transformed_x)

Coefficients:
(Intercept)  transformed_x
      18.068         -1.542

```



```

Call:
lm(formula = y ~ transformed_x)

Residuals:
    Min       1Q   Median       3Q      Max
-8.4661 -1.5631 -0.3267  2.2199  5.0346

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   18.0676     0.8574   21.07 < 2e-16 ***
transformed_x  -1.5417     0.1385  -11.13 6.7e-15 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.89 on 48 degrees of freedom
Multiple R-squared:  0.7209,    Adjusted R-squared:  0.7151
F-statistic: 124 on 1 and 48 DF,  p-value: 6.695e-15

```

```

Call:
lm(formula = df$concentration ~ df$density)

Residuals:
    Min       1Q   Median       3Q      Max
-8.181 -3.745 -0.602  3.331  9.126

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   28.8743     0.7905   36.526 < 2e-16 ***
df$density   -78.4682    15.6558   -5.012 7.74e-06 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

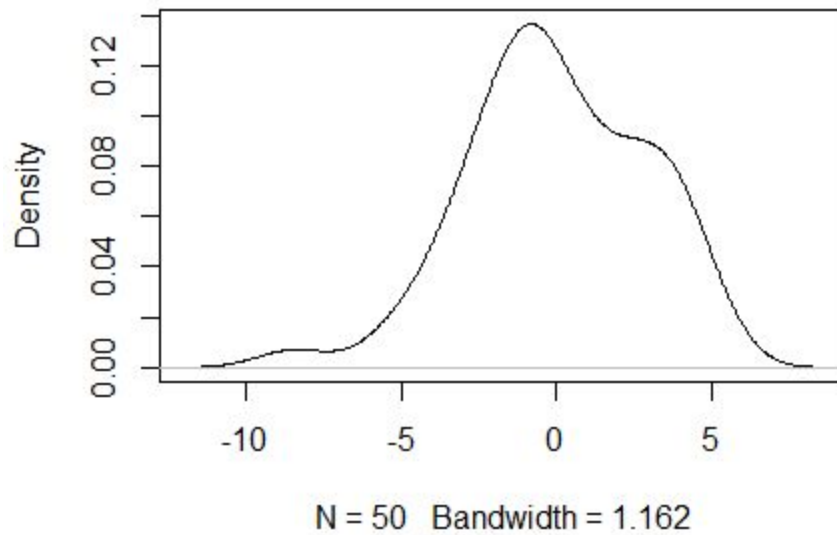
Residual standard error: 4.432 on 48 degrees of freedom
Multiple R-squared:  0.3436,    Adjusted R-squared:  0.3299
F-statistic: 25.12 on 1 and 48 DF,  p-value: 7.737e-06

```

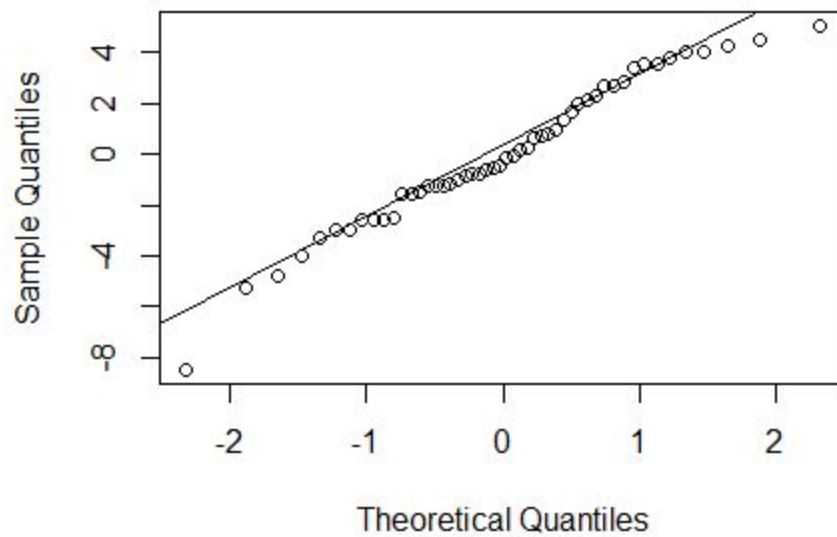
Comment: both p value and f value suggest the model have become better.

11.

density.default(x = resid(model_new))



Normal Q-Q Plot



A quantile normal plot - good for checking normality

12.

coefficient of determination ---- 0.7208808

coefficient of determination of old model --- 0.3435525

13.

model performance has become better after the transformation as can be seen from Q12.

Part 1 :

1. Shapiro-Wilk Test

```
> shapiro.test(df$dose)
```

```
shapiro-wilk normality test
```

```
data: df$dose
```

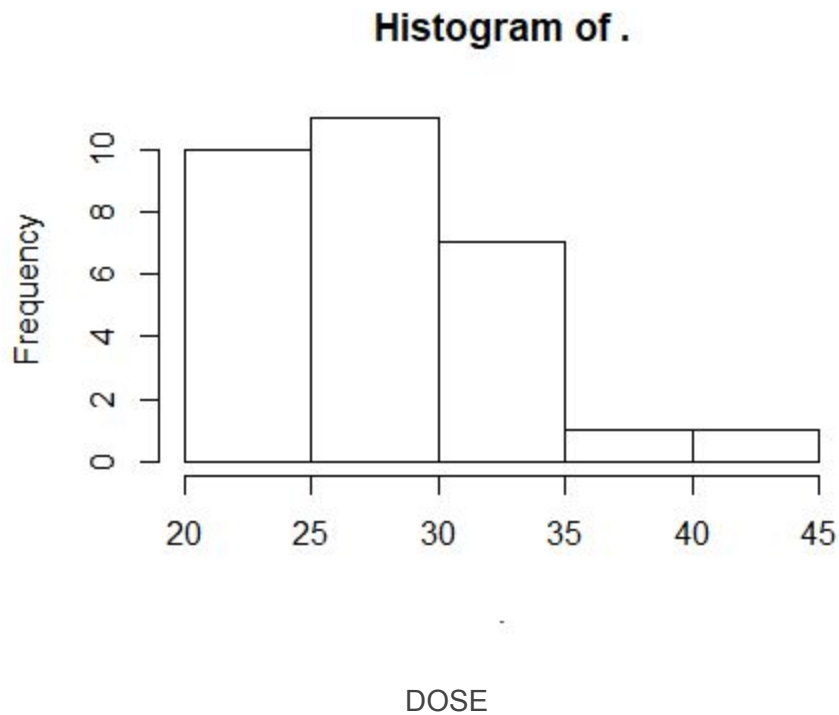
```
W = 0.93352, p-value = 0.06095
```

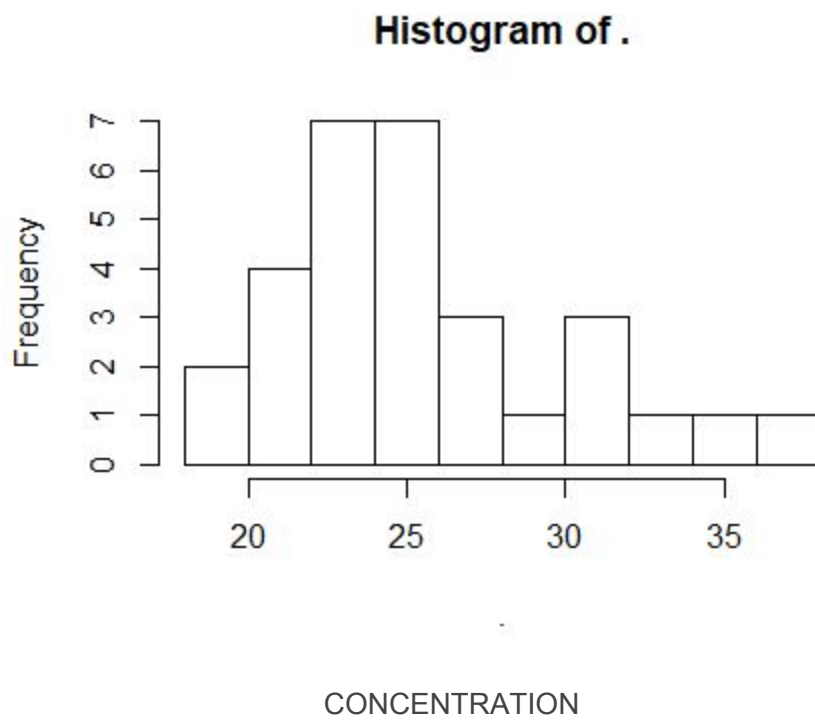
```
> shapiro.test(df$concentration)
```

```
shapiro-wilk normality test
```

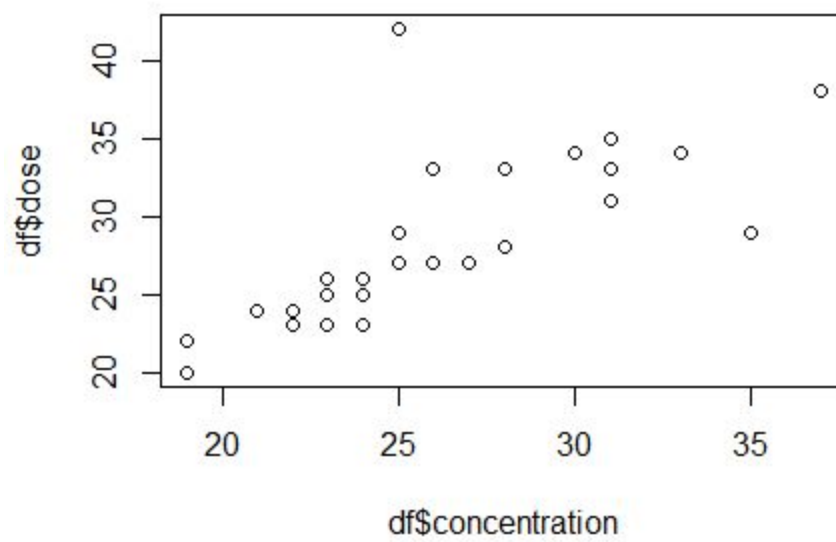
```
data: df$concentration
```

```
W = 0.94573, p-value = 0.1298
```





2. Scatterplot



3.

```
> pearson_corr
```

```
Pearson's product-moment correlation
```

```
data: df$dose and df$concentration  
t = 5.6646, df = 28, p-value = 4.53e-06  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 0.5028893 0.8636558  
sample estimates:  
      cor  
0.7307623
```

because the

output of p value is $4.53e-06 < 0.05$ we can say that correlation exist.

Cor value -- 0.7307623

```
> spearman_corr
```

```
spearman's rank correlation rho
```

```
data: df$dose and df$concentration  
S = 553.11, p-value = 2.074e-10  
alternative hypothesis: true rho is not equal to 0  
sample estimates:  
      rho  
0.876949
```

output of p value is $2.074e-10 < 0.05$ we can say that correlation exist.

Cor value -- 0.876949

4.

as it can be seen from the scatterplot and the correlation values the variables are negatively correlated to each other and can be written as a negative linear relationship between the two variables.

5.

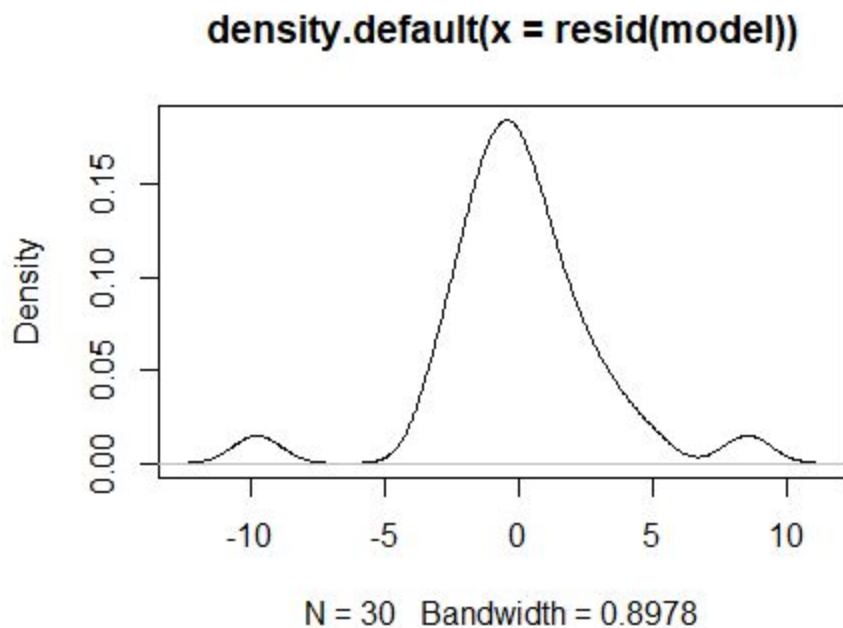
6.

7.

```
> summary(model_new)$r.squared
[1] 0.7208808
> mmary(model)$r.squared
Error in mmary(model) : could not
> summary(model)$r.squared
[1] 0.5340135
```

8.

```
> resid(model) #List of residuals
      1      2      3      4      5      6      7      8
3.26143390 -3.00429730 4.80144705 -0.19000219 -1.55286121 -0.64143828 -0.19000219 0.80999781
      9     10     11     12     13     14     15     16
-9.74711686 3.35001096 0.35001096 1.80999781 0.08427976 8.53571585 -0.55286121 -2.27857926
     17     18     19     20     21     22     23     24
-2.27857926 1.35856172 -1.46428415 -1.01284806 0.71286998 -0.91572024 0.35856172 2.17285683
     25     26     27     28     29     30
0.80999781 1.98715194 -0.55286121 -1.73001535 -3.01284806 -1.27857926
```



9.

10.

model performance has become better after the transformation.

