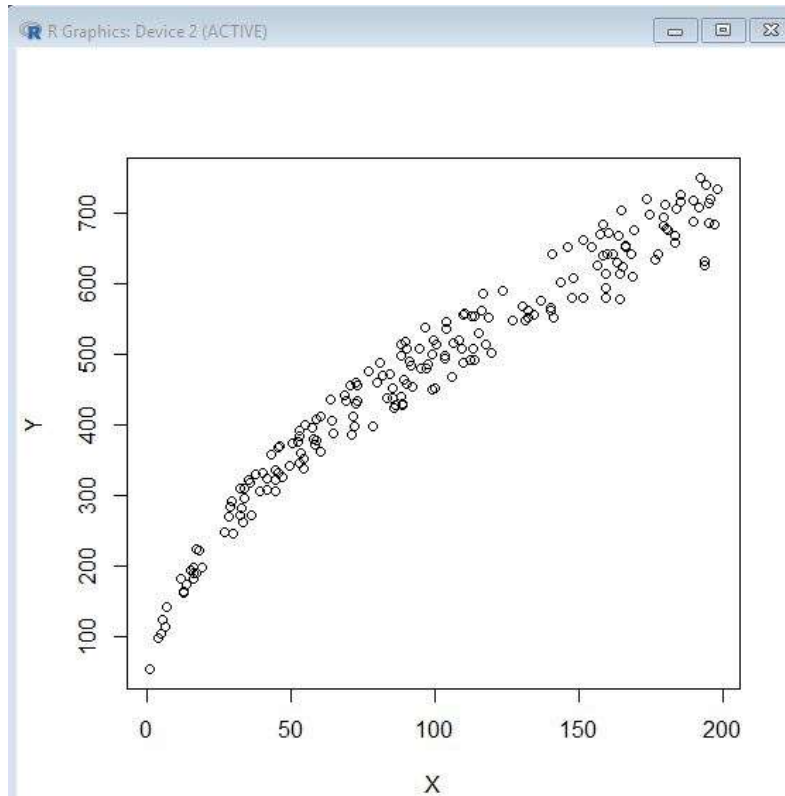


Question 2

i) Plot the dataset

```
>  
> plot(Y~X,data=ISLAS12)  
> |
```



II) summary lmfit original

```
> plot(Y~X,data=ISLAS12)  
> lmfit2<-lm(Y~X,data=ISLAS12)  
> summary(lmfit2);
```

Call:

```
lm(formula = Y ~ X, data = ISLAS12)
```

Residuals:

Min	1Q	Median	3Q	Max
-147.224	-22.183	1.705	29.137	71.576

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	199.5172	5.7231	34.86	<2e-16 ***
X	2.7689	0.0506	54.72	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 39.87 on 198 degrees of freedom

Multiple R-squared: 0.938, Adjusted R-squared: 0.9377

F-statistic: 2994 on 1 and 198 DF, p-value: < 2.2e-16

On transformed data

```
> lmdata4<-lm(Y~log(X,base=exp(1)),data=ISLAS12)
> summary(lmdata4)

Call:
lm(formula = Y ~ log(X, base = exp(1)), data = ISLAS12)

Residuals:
    Min       1Q   Median       3Q      Max
-77.93 -33.58 -12.30  23.00 330.00

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)   -289.257     19.427  -14.89  <2e-16 ***
log(X, base = exp(1))  175.574      4.396   39.94  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 53.2 on 198 degrees of freedom
Multiple R-squared:  0.8896,    Adjusted R-squared:  0.889
F-statistic: 1595 on 1 and 198 DF,  p-value: < 2.2e-16

> lmdata5<-lm(log(Y,base=exp(1))~log(X,base=exp(1)),data=ISLAS12)
> summary(lmdata5)

Call:
lm(formula = log(Y, base = exp(1)) ~ log(X, base = exp(1)), data = ISLAS12)

Residuals:
    Min       1Q   Median       3Q      Max
-0.107212 -0.046358 -0.000565  0.052449  0.092625

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)   3.903721    0.021283   183.4  <2e-16 ***
log(X, base = exp(1)) 0.501951    0.004816  104.2  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.05828 on 198 degrees of freedom
Multiple R-squared:  0.9821,    Adjusted R-squared:  0.982
F-statistic: 1.086e+04 on 1 and 198 DF,  p-value: < 2.2e-16

> lmdata5<-lm(log(Y,base=exp(1))~X,data=ISLAS12)
> summary(lmdata5)

Call:
lm(formula = log(Y, base = exp(1)) ~ X, data = ISLAS12)

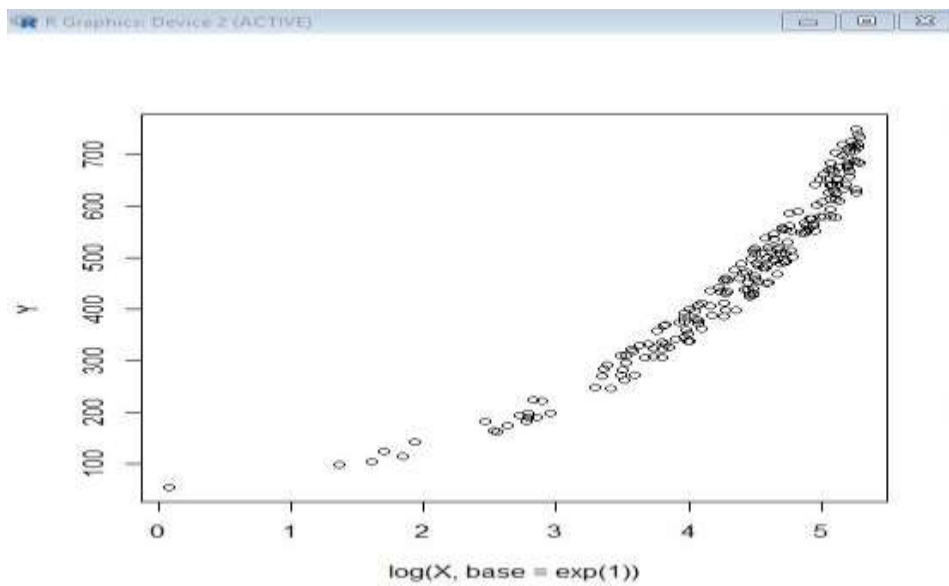
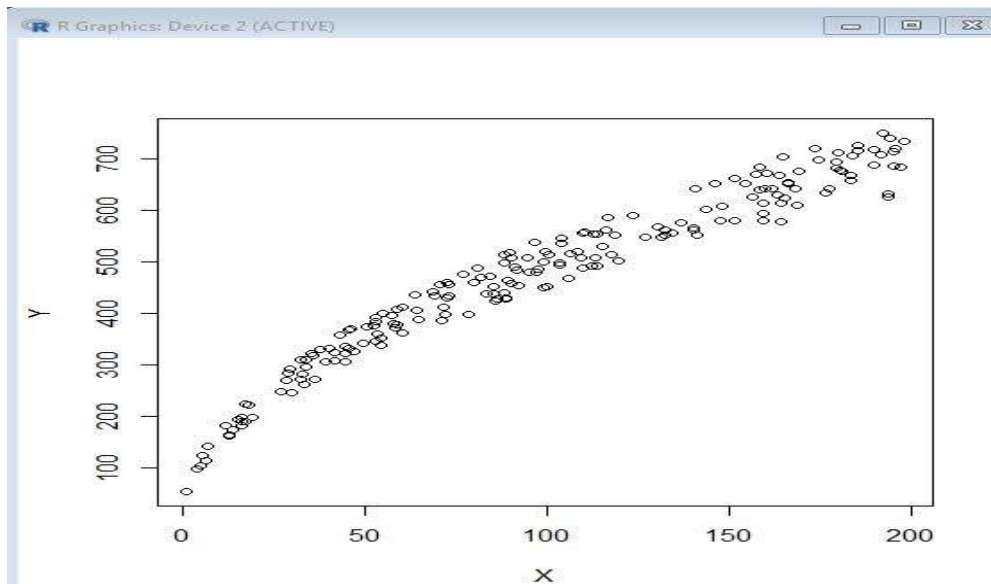
Residuals:
    Min       1Q   Median       3Q      Max
-1.39858 -0.08071  0.04548  0.12597  0.23619

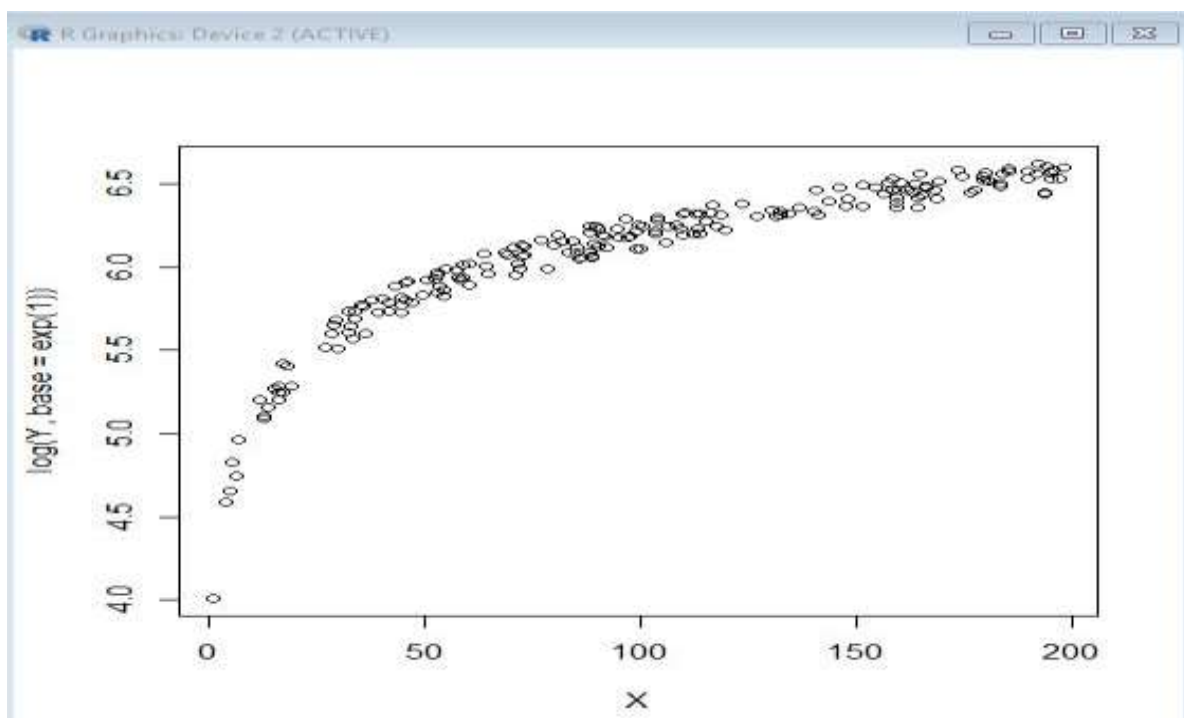
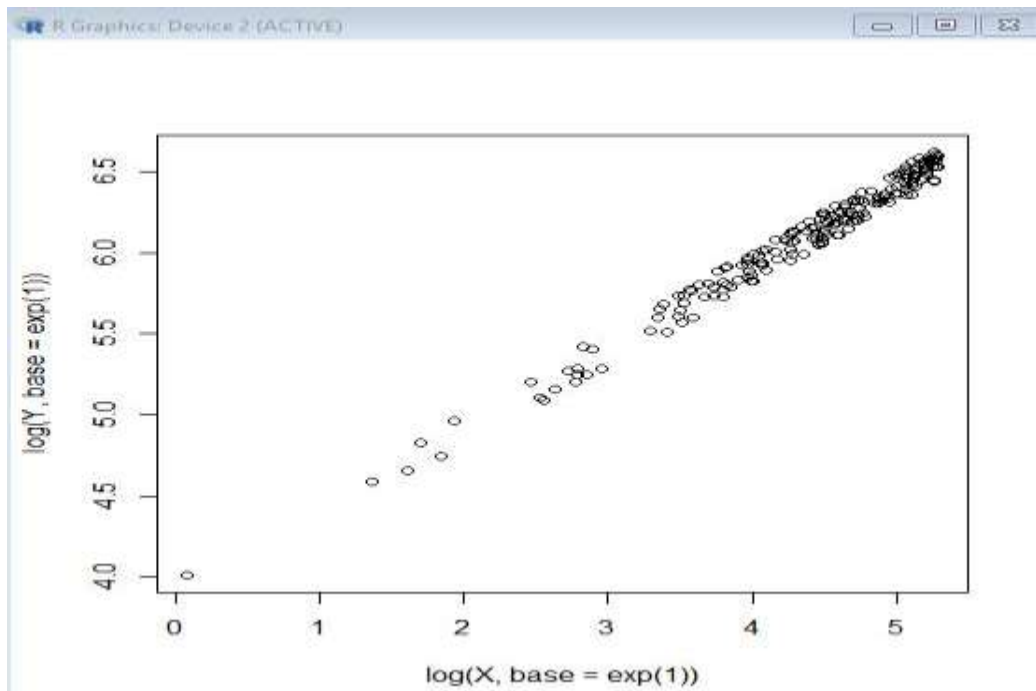
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  5.4039242    0.0293150   184.34  <2e-16 ***
X             0.0068710    0.0002592    26.51  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2042 on 198 degrees of freedom
Multiple R-squared:  0.7802,    Adjusted R-squared:  0.779
F-statistic: 702.6 on 1 and 198 DF,  p-value: < 2.2e-16
```

iv. Plot the linear and transformed graphs

```
> plot(Y~X,data=ISLAs12)
> plot(log(Y,base=exp(1))~X,data=ISLAs12)
> plot(log(Y,base=exp(1))~log(X,base=exp(1)),data=ISLAs12)
> plot(log(Y,base=exp(1))~X,data=ISLAs12)
> |
```





v. we need to compare

- Plots number is more for the transformed data when compared to that of the original data
- By observing p-values on f-statistics it can be said that transformed data is better fit than the original data set

Original data

Residual standard error: 39.87 on 198 degrees of freedom

Multiple R-squared: 0.938, Adjusted R-squared: 0.9377

F-statistic: 2994 on 1 and 198 DF, p-value: $< 2.2e-16$

Transforming data

Residual standard error: 0.05828 on 198 degrees of freedom

Multiple R-squared: 0.9821, Adjusted R-squared: 0.982

F-statistic: 1.086e+04 on 1 and 198 DF, p-value: < 2.2e-16

Conclusions: P is same in both base.

Transformed dataset has the low Residual standard error value and good F-statistic value which is a strong sign for better fit.

R-square value in the transformed data varies by 98% where as in original data it varies by 93%