

Simple Linear Regression

Prisma Erika Lopez Jimenez

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```
### Load your packages
```

```
library(MASS)
```

```
library(ISLR)
```

```
### Make a linear model on select variables of the `Boston` dataset
```

```
fix(Boston) # edit the `Boston` dataset in the `MASS` package
```

```
attach(Boston) # associate the variables `medv` and `lstat`
```

```
# with the dataset `Boston` without incl. in the model code below:
```

```
lm.fit=lm(medv~lstat) # linear model of the `Boston` data set where `medv`  
# is the response and `lstat` is the predictor
```

```
lm.fit # take a look at the linear model basics
```

```
##
```

```
## Call:
```

```
## lm(formula = medv ~ lstat)
```

```
##
```

```
## Coefficients:
```

```
## (Intercept)      lstat
```

```
##      34.55      -0.95
```

```
summary(lm.fit) # get a summary
```

```
##
```

```
## Call:
```

```
## lm(formula = medv ~ lstat)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -15.168  -3.990  -1.318   2.034  24.500
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) 34.55384    0.56263   61.41  <2e-16 ***
```

```
## lstat      -0.95005    0.03873  -24.53  <2e-16 ***
```

```
## ---
```

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

```
##
```

```
## Residual standard error: 6.216 on 504 degrees of freedom
```

```
## Multiple R-squared:  0.5441, Adjusted R-squared:  0.5432
```

```
## F-statistic: 601.6 on 1 and 504 DF, p-value: < 2.2e-16
```

```
names(lm.fit) # what you can extract from the model `lm.fit`
```

```
## [1] "coefficients" "residuals"      "effects"        "rank"
```

```
## [5] "fitted.values" "assign"          "qr"             "df.residual"
```

```
## [9] "xlevels"       "call"           "terms"          "model"
```

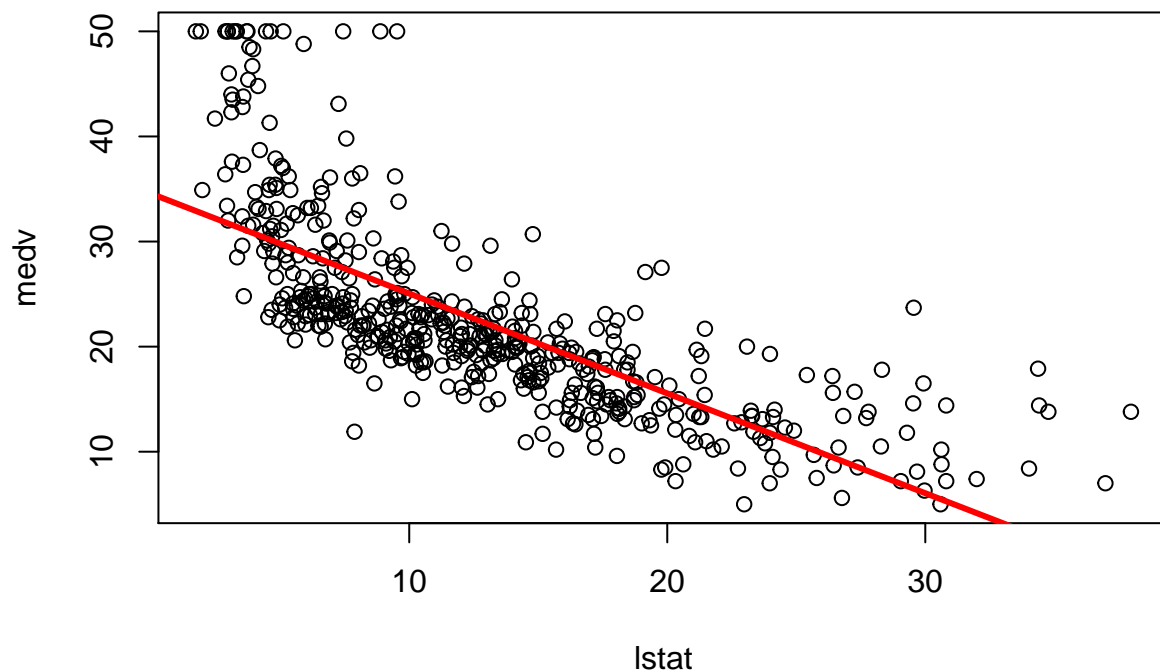
```
coef(lm.fit) # extract the coefficient values of the linear model
```

```
## (Intercept)      lstat  
## 34.5538409 -0.9500494
```

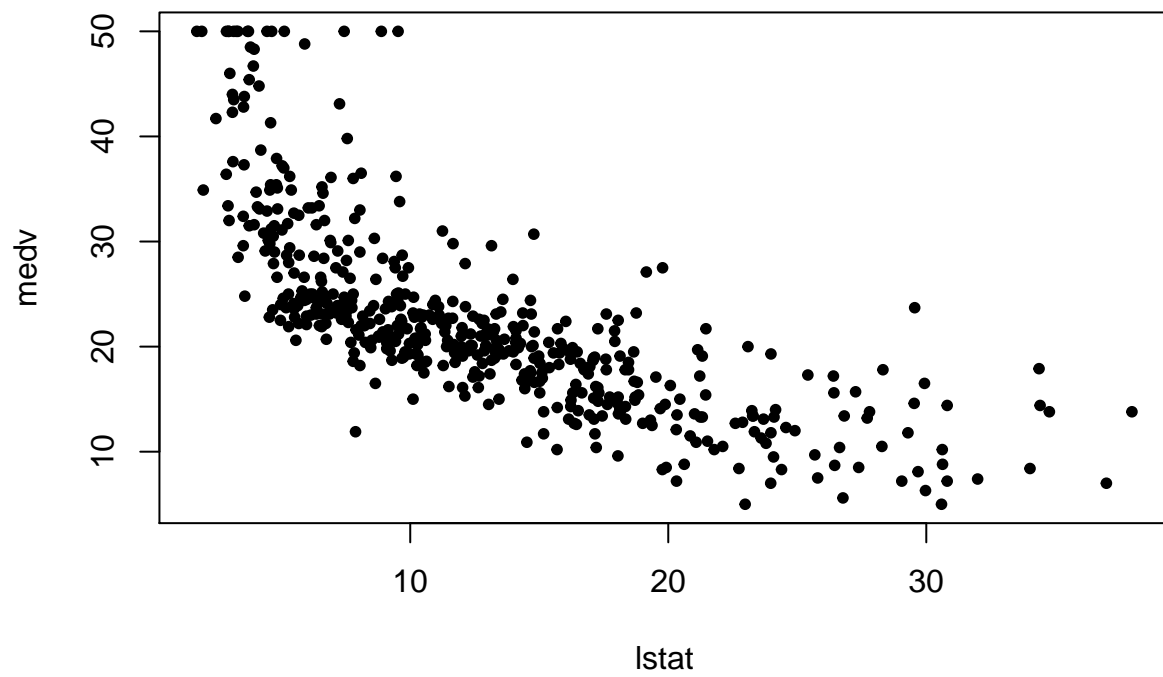
```
confint(lm.fit) # confidence intervals
```

```
##           2.5 %      97.5 %  
## (Intercept) 33.448457 35.6592247  
## lstat       -1.026148 -0.8739505
```

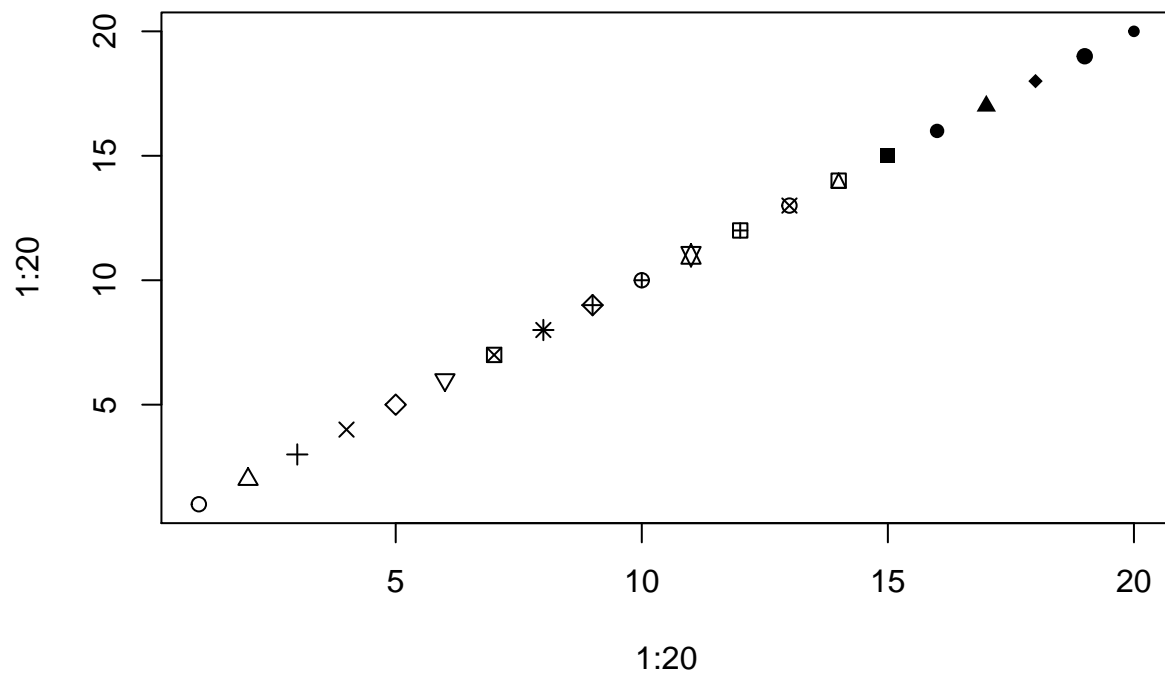
```
### Plot the data and linear regression line for the model  
plot(lstat,medv) # scatter plot of `lstat` vs. `medv`  
abline(lm.fit,lwd=3,col="red") # add linear regression line,
```



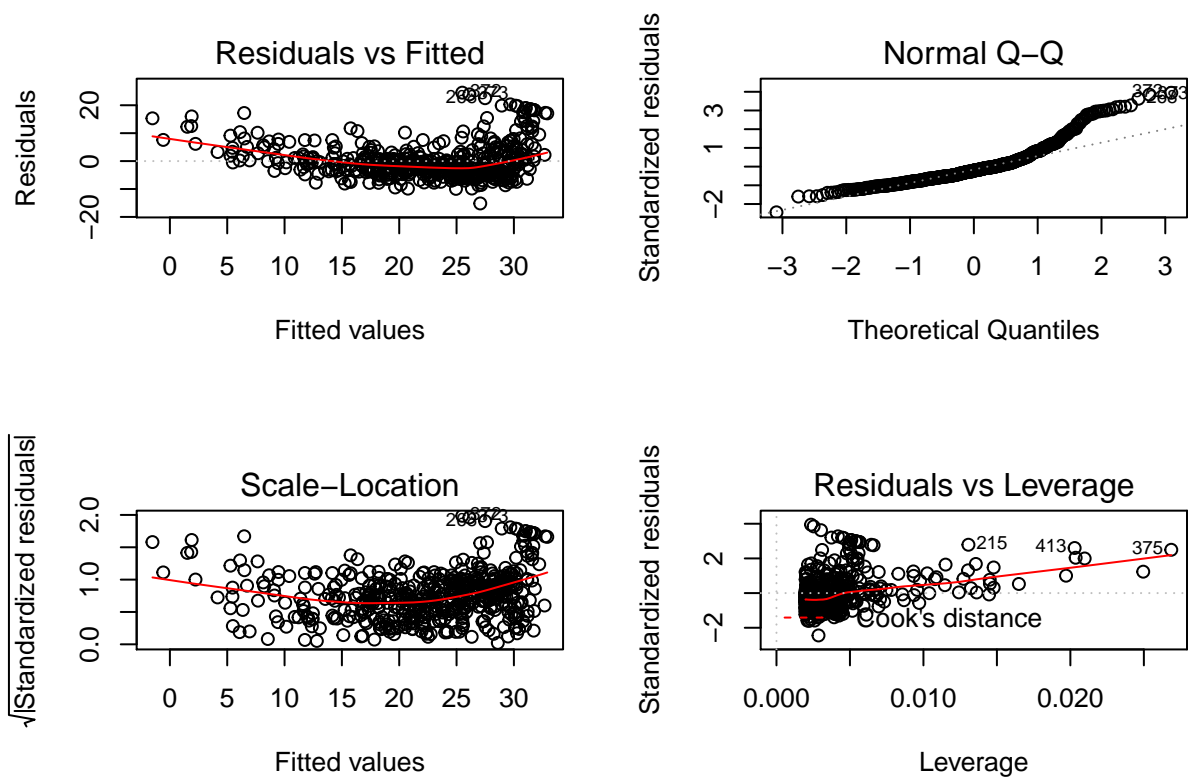
```
# color it red, and bold it 3x  
plot(lstat,medv,pch=20) # modify the dots on the scatter plot
```



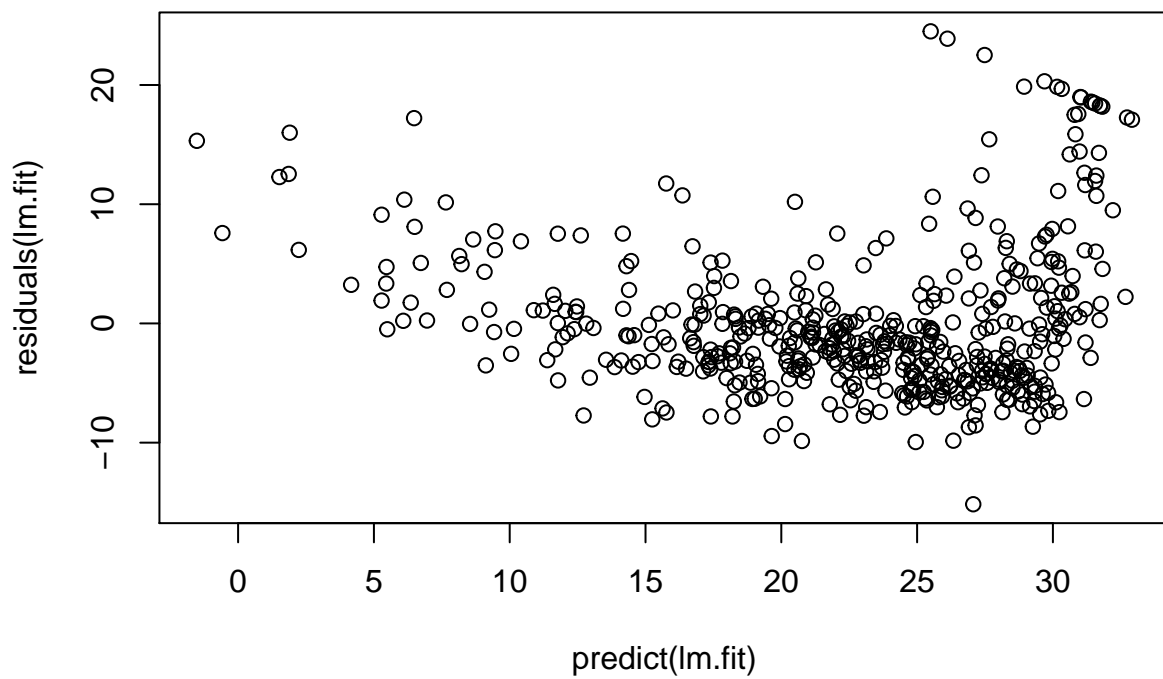
```
plot(1:20,1:20,pch=1:20)# example of plot() function implementation
```



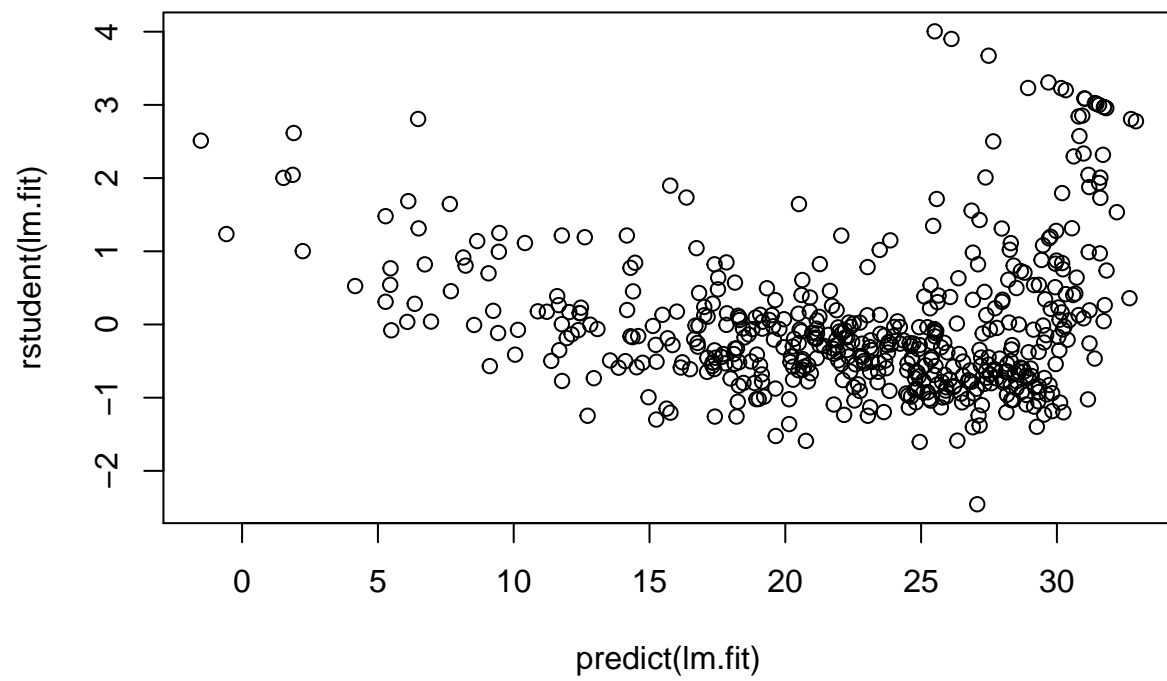
```
### Create Diagnostic plots on the complete model  
par(mfrow=c(2,2)) # separate the grid as a 2 x 2 fit  
plot(lm.fit) # plot 4 diagnostic plots on the `lm.fit` model
```



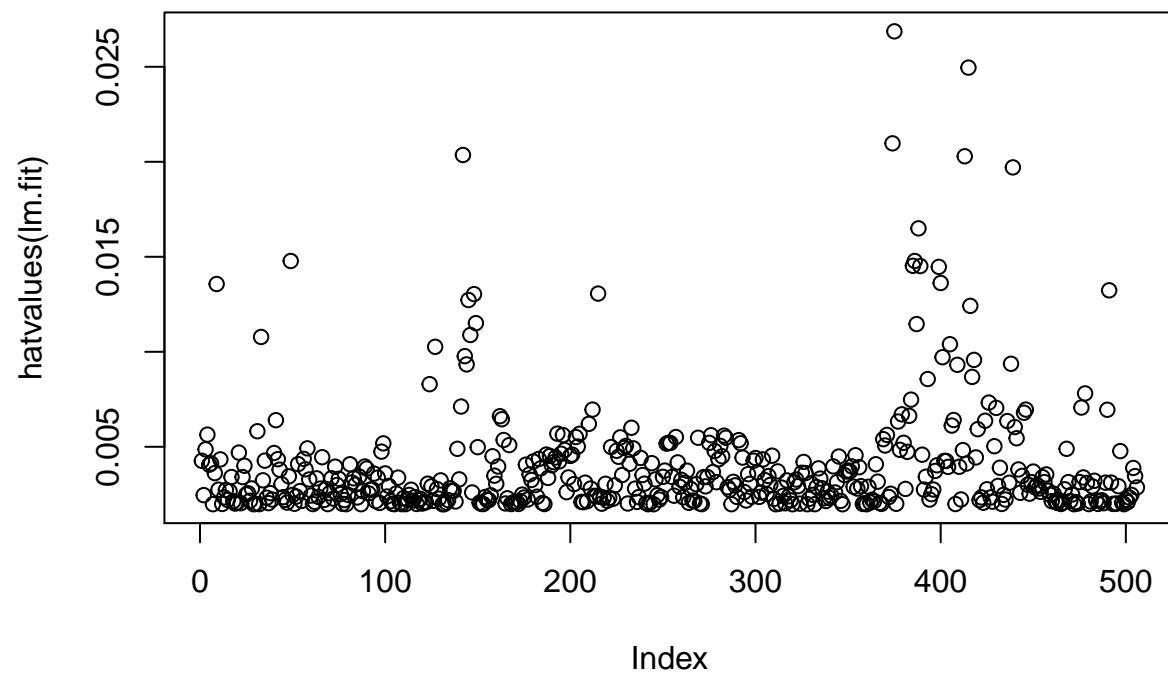
```
### Compute the residuals from a linear regression fit
plot(predict(lm.fit),residuals(lm.fit)) # compute residual method 1 note: residuals()
```



```
plot(predict(lm.fit),rstudent(lm.fit)) # compute residual method 2 note: rstudent( )
```



```
plot(hatvalues(lm.fit))
```



```
which.max(hatvalues(lm.fit))
```

```
## 375
```

```
## 375
```