

Week 2 Workshop

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Data Upload

```
library(MASS)  
library(ISLR)
```

```
## Warning: package 'ISLR' was built under R version 3.2.3
```

```
attach(Boston)
```

Simple Linear Regression

► Data Understanding

```
names(Boston)
```

```
## [1] "crim"      "zn"        "indus"     "chas"      "nox"      "rm"
## [8] "dis"       "rad"       "tax"       "ptratio"   "black"    "lstat"
```

```
head(Boston)
```

```
##      crim  zn  indus  chas   nox   rm   age   dis  rad  tax
## 1 0.00632 18  2.31    0 0.538 6.575 65.2 4.0900   1 296
## 2 0.02731  0  7.07    0 0.469 6.421 78.9 4.9671   2 242
## 3 0.02729  0  7.07    0 0.469 7.185 61.1 4.9671   2 242
## 4 0.03237  0  2.18    0 0.458 6.998 45.8 6.0622   3 222
## 5 0.06905  0  2.18    0 0.458 7.147 54.2 6.0622   3 222
## 6 0.02985  0  2.18    0 0.458 6.430 58.7 6.0622   3 222
##      lstat medv
## 1   4.98 24.0
## 2   2.14 21.6
## 3   2.14 21.6
## 4   2.14 21.6
## 5   2.14 21.6
## 6   2.14 21.6
```

Explore details of the Boston data set

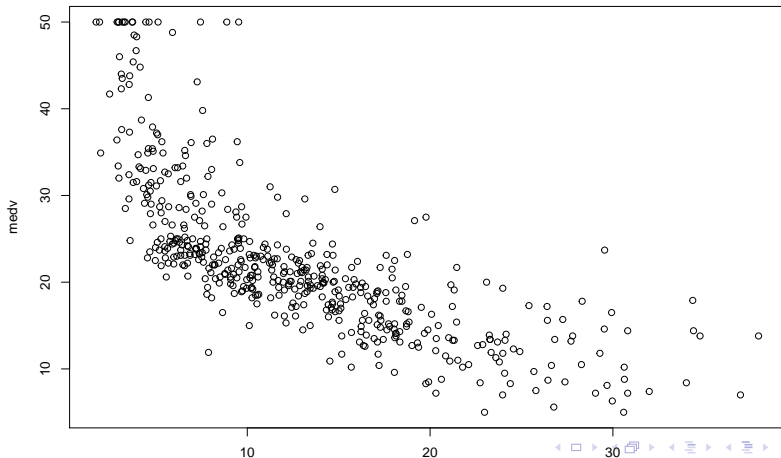
```
?Boston
```

```
## starting httpd help server ... done
```

What is the Target Variable?

- scatter plot

```
plot(medv~lstat,Boston)
```



Simple linear regression model

```
model1 = lm(medv~lstat)  
model1
```

```
##  
## Call:  
## lm(formula = medv ~ lstat)  
##  
## Coefficients:  
## (Intercept)          lstat  
##      34.55         -0.95
```

```
summary(model1)
```

```
##
```

```
## 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
```

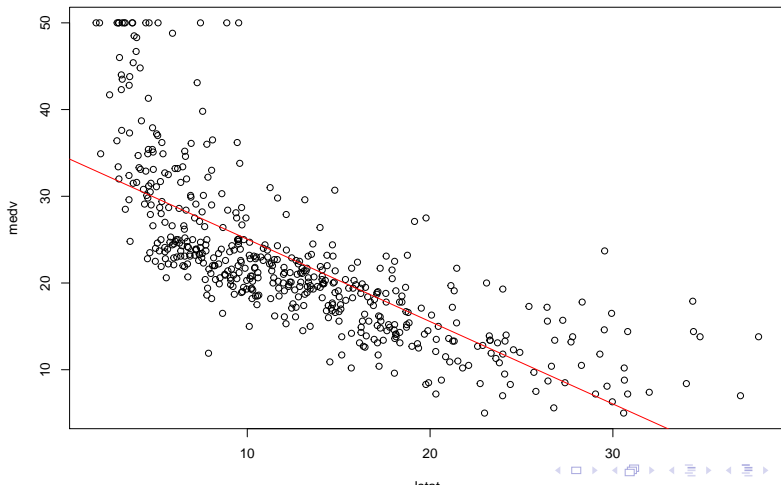
```
##
```

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

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

Plot the linear model within the scatterplot

```
plot(medv~lstat,Boston)
abline(model1,col="red")
```




```
names(model1)
```

```
## [1] "coefficients" "residuals"      "effects"        "ra"
## [5] "fitted.values" "assign"         "qr"            "df"
## [9] "xlevels"       "call"          "terms"         "mo"
```

Coefficients and Predicted values

```
confint(model1)
```

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

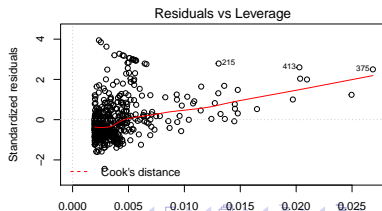
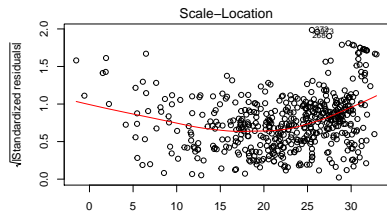
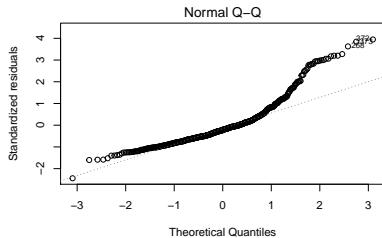
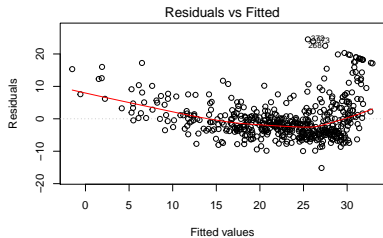
```
predict(model1,data.frame(lstat= c(5,10,15,20)),interval="confidence")
```

```
##      fit      lwr      upr  
## 1 29.80359 29.00741 30.59978  
## 2 25.05335 24.47413 25.63256  
## 3 20.30310 19.73159 20.87461  
## 4 15.55285 14.77355 16.33216
```

NOTE: We could also use list rather than data.frame such as -
`predict(model1,list(lstat= c(5,10,15,20)),interval="confidence")`

Model Checking

```
par(mfrow=c(2,2))  
plot(model1)
```



Exercises

Please upload the following via Turnitin link given in vUWS site.

Construct the linear Regression models for medv against all other numerical predictor variables.

- ▶ 1 Select all predictor variables having a significant linear relationship with the target variable medv.
- ▶ 2 For each linear model give the regression equation and its R squared value and discuss its strength.
- ▶ 3 plot the scatter plot with the abline for all significant linear models