



# VIT<sup>®</sup>

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## Vellore Institute of Technology

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**19BCE1311**

**CSE3506 – ESSENTIALS OF DATA ANALYTICS  
LAB-1**

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## Tasks for Week-1: Regression

Understand the following operations/functions on random dataset and perform similar operations on mtcars and 'data.csv' dataset based on given instructions.

**Aim:** To develop linear regression model for the given data using R programming and to verify the null hypothesis.

### Algorithm:

1. Set the working directory
2. Read data into a variable as a dataframe
3. Take 75% of the data for training the model
4. Take 25% of the data for testing the data
5. Find correlation between the 2 variables for additional statistics
6. Plot the data points
7. Train the linear model
8. Plot the linear model in the same graph
9. Print the summary of the model

### Statistics:

#### i) For mtcars:

#### Residuals:

Min	1Q	Median	3Q	Max
-4.6037	-2.6129	-0.1983	1.3715	6.5714

#### Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	38.2943	2.2919	16.71	5.50e-14
wt	-5.6437	0.7171	-7.87	7.73e-08

**Residual standard error:** 3.336 on 22 **degrees of freedom**

**Multiple R-squared:** 0.7379, **Adjusted R-squared:** 0.726

**F-statistic:** 61.94 on 1 and 22 DF, **p-value:** 7.733e-08

## ii) For data.csv:

### Residuals:

Min	1Q	Median	3Q	Max
-30.307	-13.598	1.082	13.168	28.924

### Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	170.562451	2.772873	61.511	<2e-16
Weight	-0.004918	0.025536	-0.193	0.847

**Residual standard error:** 16.22 on 373 **degrees of freedom**

**Multiple R-squared:** 9.944e-05, **Adjusted R-squared:** -0.002581

**F-statistic:** 0.03709 on 1 and 373 DF, **p-value:** 0.8474

### Inference:

**mtcars:** The linear model is accepted because p-value(7.733e-08) is less than 0.05

**Data.csv:** The linear model is rejected because p-value (0. 8474) is greater than 0.05

### Program:

#### i) For mtcars:

```
rm(list=ls())
```

```
library(dplyr)
```

```
library(Metrics)
```

```
data1 <- mtcars
```

```
## 75% of the sample size
smp_size <- floor(0.75 * nrow(mtcars))

#setting the seed to make your partition reproducible
set.seed(123)
train_ind <- sample(seq_len(nrow(mtcars)), size = smp_size)
train <- mtcars[train_ind, ]
test <- mtcars[-train_ind, ]

cr<-cor.test(train$wt,train$mpg)
print(cr)

plot(train$wt,train$mpg,xlab = "Wt",ylab = "mpg",main="mpg VS Wt")

## Linear model
lmodel<-lm(mpg~wt,data=train)
abline(lmodel,col="red")

summary(lmodel)

predicted<-predict(lmodel,data=test)
mae(test$mpg,predicted)
```

## **ii) For data.csv**

```
rm(list=ls())
library(dplyr)
library(Metrics)

setwd("C:/Users/Abhinav Vijayakumar/Desktop/VIT Academics/Sem 6/Essentials of
Data Analytics/LAB/LAB 1")
```

```
data<-read.csv('data.csv')

## 75% of the sample size
smp_size <- floor(0.75 * nrow(data))

#setting the seed to make your partition reproducible
set.seed(123)
train_ind <- sample(seq_len(nrow(data)), size = smp_size)
train <- data[train_ind, ]
test <- data[-train_ind, ]
cr<-cor.test(train$Height,train$Weight)
print(cr)

plot(train$Weight,train$Height,xlab = "Weight",ylab = "Height",main="Height vs
Weight")

##Linear model
lmodel<-lm(Height~Weight,data=train)
abline(lmodel,col="red")
summary(lmodel)

predicted<-predict(lmodel,data=test)
mae(test$Height,predicted)
```