

Essential of Data Analytics

Tasks for Week-10: Gradient Descent algorithm

Aim: Implementing the Momentum Gradient Descent Algorithm

Algorithm:

- 1) Clear the space by clearing the list using rm() function.
- 2) Then declare function MGD using function() with $x_1, x_2, y, m_1, m_2, c, \text{learning rate}(\alpha), \gamma$, number of iterations as the parameters of the function.
- 3) Declare a variable iterations and assign a value 0.
- 4) Then declare variables LF, u_{m1} , u_c and assign value 0.
- 5) Run a while loop and check the condition if iterations are less than or equal to given number of iterations.
- 6) Then declare the variable y_{pred} and assign $m_1 * x_1 + m_2 * x_2 + c$ value to y_{pred} .
- 7) Then declare variable Lf_{new} and assign $0.5 * \text{sum of difference of } y_{pred} \text{ and } y \text{ power of } 2$.
- 8) Declare and assign variable nu_{m1} as $\gamma * u_{m1} + \alpha * \text{sum of difference of } y_{pred} \text{ and } y * x_1$.
- 9) Declare and assign variable nu_{m2} as $\gamma * u_{m2} + \alpha * \text{sum of difference of } y_{pred} \text{ and } y * x_2$.
- 10) Declare and assign variable nu_c as $\gamma * u_c + \alpha * \text{sum of difference of } y_{pred} \text{ and } y$.
- 11) Assign m_1 as difference of m_1 and nu_{m1} .
- 12) Assign m_2 as difference of m_2 and nu_{m2} .
- 13) Assign c as difference of c and nu_{m1} .
- 14) Assign u_{m1} as nu_{m1} .
- 15) Assign u_{m2} as nu_{m2} .
- 16) Assign u_c as nu_c .
- 17) Assign u_c as nu_c .
- 18) Assign Lf as Lf_{new} .
- 19) Then increase iterations to 1.
- 20) End the loop.
- 21) Then load the dataset.
- 22) Plot the two variables that chosen for applying momentum gradient descent using plot() function.

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- 23) Then call the function Mgd by giving the parameters.
- 24) Apply linear model to selected variables.
- 25) Then analyse the summary.

Inference:

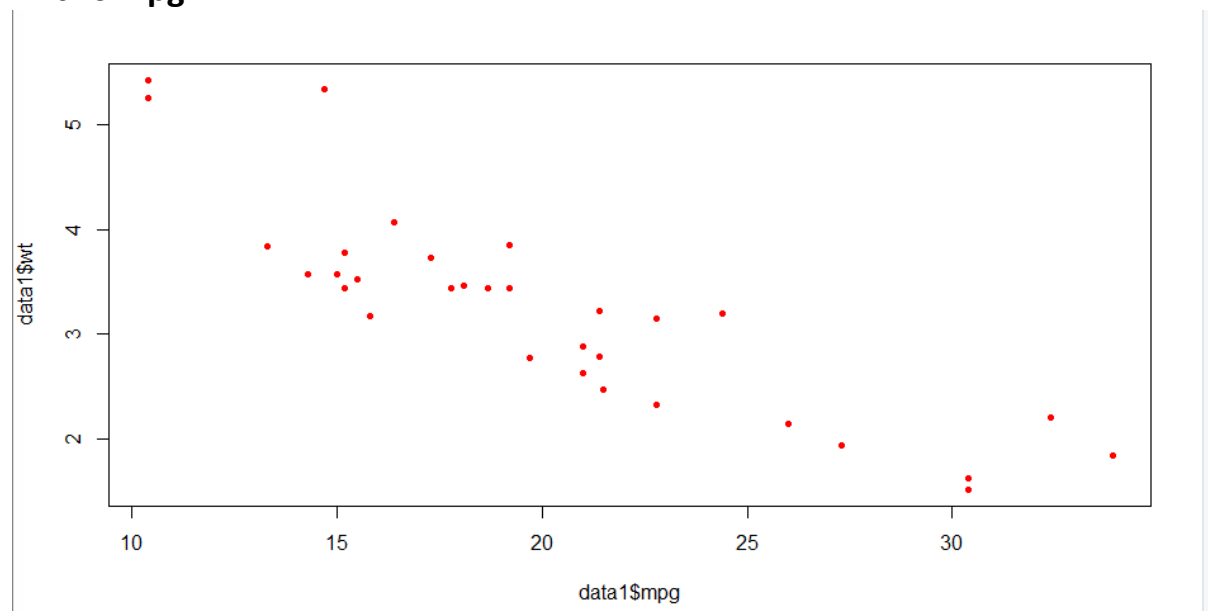
Dataset : mtcars X= hp and wt Y= mpg

Mgd(data1\$wt,data1\$hp,data1\$mpg,-0.2,-0.2,32,0.000002,0,1000000)

According to the observation we saw that intercept and slope of linear regression and momentum gradient descent algorithm doesn't have much difference but momentum gradient descent model performs a bit better than gradient descent model.

Result:

Wt vs mpg:



Linear Model summary:

```
Coefficients:
(Intercept)    data1$hp    data1$wt
  37.22727      -0.03177    -3.87783
```

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Code:

```
rm(list=ls())

Mgd<-function(x1,x2,y,m1,m2,c,alpha,gamma,iter){
  iterations=0
  Lf<-0
  u_m1<-0
  u_m2<-0
  u_c<-0
  while(iterations<=iter){
    y_pred<-m1*x1+m2*x2+c
    Lf_new<-0.5*sum((y_pred-y)^2)
    nu_m1<-gamma*u_m1+alpha*sum((y_pred-y)*x1)
    nu_m2<-gamma*u_m2+alpha*sum((y_pred-y)*x2)
    nu_c<-gamma*u_c+alpha*sum(y_pred-y)
    m1<-m1-nu_m1
    m2<-m2-nu_m2
    c<-c-nu_c
    u_m1<-nu_m1
    u_m2<-nu_m2
    u_c<-nu_c
    Lf<-Lf_new
    iterations=iterations+1
  }
```

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```
}  
  
  return(paste("optimal intercept:",c,"optimatl slope:",m1,m2,"Loss  
funciton:",Lf,"iterations:",iterations))  
  
}  
  
data1<-mtcars  
  
plot(data1$mpg,data1$wt,col="red",pch=20)  
  
Mgd(data1$wt,data1$hp,data1$mpg,-0.2,-0.2,32,0.000002,0,1000000)  
  
lr<-lm(data1$mpg~data1$hp+data1$wt)  
  
lr
```