# Implementation-of-Linear-Regression-Using-Gradient-Descent

### AIM:

To write a program to predict the profit of a city using the linear regression model with gradient descent.

## **Equipments Required:**

- 1. Hardware PCs
- 2. Anaconda Python 3.7 Installation / Jupyter notebook

# **Algorithm**

- 1. Import required libraries in python for Gradient Design.
- 2. Upload the dataset and check any null value using .isnull() function.
- 3. Declare the default values for linear regression.
- 4. Calculate the loss usinng Mean Square Error.
- 5. Predict the value of y.
- 6. Plot the graph respect to hours and scores using scatter plot function.

# **Program:**

```
Program to implement the linear regression using gradient descent.
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import numpy as np
import pandas as pd
from sklearn.preprocessing import StandardScaler
def linear_regression(X1,y,learning_rate=0.01,num_iters=1000):
   X=np.c_[np.ones(len(X1)),X1]
   theta = np.zeros(X.shape[1]).reshape(-1,1)
   for _ in range(num_iters):
        predictions = (X).dot(theta).reshape(-1,1)
        errors = (predictions-y).reshape(-1,1)
        theta-=learning_rate*(1/len(X1))*X.T.dot(errors)
   return theta
data=pd.read_csv('50_Startups.csv',header=None)
data.head()
X = (data.iloc[1:,:-2].values)
print(X)
X1=X.astype(float)
scaler = StandardScaler()
y=(data.iloc[1:,-1].values).reshape(-1,1)
print(y)
X1_Scaled=scaler.fit_transform(X1)
Y1_Scaled=scaler.fit_transform(y)
print(X1_Scaled)
print(Y1 Scaled)
theta = linear_regression(X1_Scaled,Y1_Scaled)
new_data=np.array([165349.2,136897.8,471784.1]).reshape(-1,1)
new_Scaled = scaler.fit_transform(new_data)
prediction = np.dot(np.append(1,new_Scaled),theta)
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### **Output:**

prediction = prediction.reshape(-1,1)
pre=scaler.inverse transform(prediction)

print(f"Predicted Value:{pre}")

#### X values

# print(X) [['165349.2' '136897.8' '471784.1'] ['162597.7' '151377.59' '443898.53'] ['153441.51' '101145.55' '407934.54'] ['144372.41' '118671.85' '383199.62'] ['142107.34' '91391.77' '366168.42'] ['131876.9' '99814.71' '362861.36']

### y values

```
print(y)

[['192261.83']
  ['191792.06']
  ['191050.39']
  ['182901.99']
  ['166187.94']
  ['156991.12']
  ['156122.51']
```

### X Scaled values

```
print(X1_Scaled)

[[ 2.01641149e+00    5.60752915e-01    2.15394309e+00]
    [ 1.95586034e+00    1.08280658e+00    1.92360040e+00]
    [ 1.75436374e+00    -7.28257028e-01    1.62652767e+00]
    [ 1.55478369e+00    -9.63646307e-02    1.42221024e+00]
    [ 1.50493720e+00    -1.07991935e+00    1.28152771e+00]
    [ 1.27980001e+00    -7.76239071e-01    1.25421046e+00]
    [ 1.34006641e+00    9.32147208e-01    -6.88149930e-01]
```

### y Scaled values

### print(Y1\_Scaled)

```
[[ 2.01120333]
```

[ 1.99942997]

[ 1.98084225]

[ 1.77662724]

[ 1.35774012]

[ 1.12724963]

[ 1.10548055]

[ 1.09620987]

### **Predicted value**

Predicted value: [[203866.53076613]]

### **Result:**

Thus the program to implement the linear regression using gradient descent is written and verified using python programming.