Name: Srinivasan JP Reg No: 21MIS1044

Linear Regression using Gradient descent method

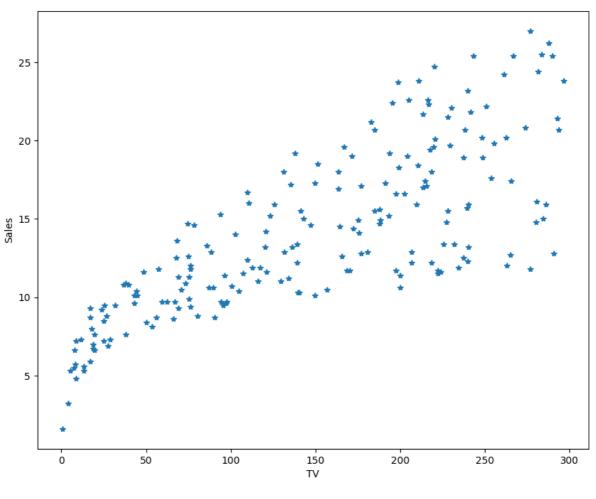
```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

Read the data set

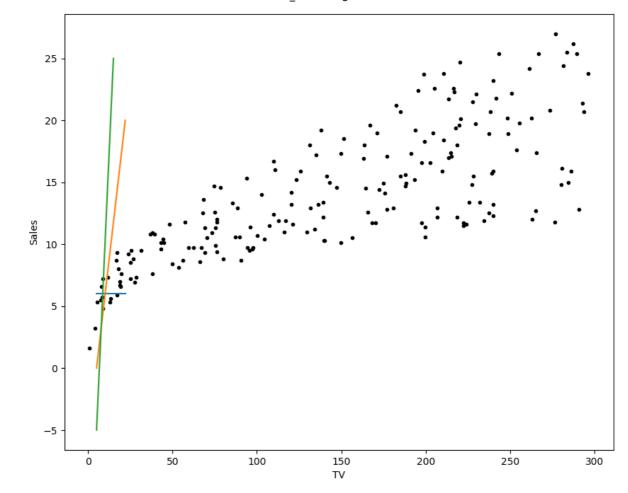
```
In [ ]: data = pd.read_csv('./tvmarketing.csv')
        print(data.head())
             TV Sales
          230.1
        0
                  22.1
                 10.4
        1
           44.5
                  9.3
        2
          17.2
        3 151.5
                 18.5
        4 180.8 12.9
In [ ]: X_df = pd.DataFrame(data.TV)
        y_df = pd.DataFrame(data.Sales)
        m = len(y df)
```

Ploting the inital scatter plot to view the data points

```
In []: plt.figure(figsize=(10,8))
    plt.plot(X_df, y_df, '*')
    plt.xlabel('TV')
    plt.ylabel('Sales')
Out[]: Text(0, 0.5, 'Sales')
```



```
In []: plt.figure(figsize=(10,8))
    plt.plot(X_df, y_df, 'k.')
    plt.plot([5, 22], [6,6], '-')
    plt.plot([5, 22], [0,20], '-')
    plt.plot([5, 15], [-5,25], '-')
    plt.xlabel('TV')
    plt.ylabel('Sales')
Out[]: Text(0, 0.5, 'Sales')
```



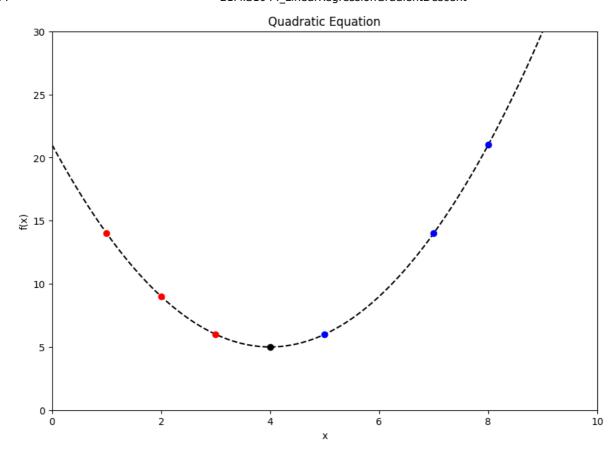
```
In []: x_{quad} = [n/10 \text{ for } n \text{ in } range(0, 100)]

y_{quad} = [(n-4)**2+5 \text{ for } n \text{ in } x_{quad}]
```

Ploting the quadratic equation where the Gradient point will move to find the best fit

```
In []: plt.figure(figsize = (10,7))
    plt.plot(x_quad, y_quad, 'k--')
    plt.axis([0,10,0,30])
    plt.plot([1, 2, 3], [14, 9, 6], 'ro')
    plt.plot([5, 7, 8],[6, 14, 21], 'bo')
    plt.plot(4, 5, 'ko')
    plt.xlabel('x')
    plt.ylabel('f(x)')
    plt.title('Quadratic Equation')
```

Out[]: Text(0.5, 1.0, 'Quadratic Equation')



```
In []: iterations = 100
alpha = 0.01

In []: ## Add a columns of 1s as intercept to X
X_df['intercept'] = 1

## Transform to Numpy arrays for easier matrix math and start theta at 0
X = np.array(X_df)
y = np.array(y_df).flatten()
theta = np.array([0, 0])
```

Function to calulate the cost function of the respective X,Y and theta

```
In []: def cost_function(X, y, theta):
    ## number of training examples
    m = len(y)

## Calculate the cost with the given parameters
    J = np.sum((X.dot(theta)-y)**2)/2/m

return J
```

```
In [ ]: cost_function(X, y, theta)
Out[ ]: 111.858125
```

Function to run the gradient iteration

```
hypothesis = X.dot(theta)
loss = hypothesis-y
gradient = X.T.dot(loss)/m
theta = theta - alpha*gradient
cost = cost_function(X, y, theta)
cost_history[iteration] = cost

return theta, cost_history
```

```
In []: (t, c) = gradient_descent(X,y,theta,alpha, iterations)

## Print theta parameters
print (t)
print (np.array([3.5, 1]).dot(t))
print (np.array([7, 1]).dot(t))

[-8.80025526e+244 -4.46899119e+242]
-3.084558330655545e+245
-6.164647670123577e+245
```

finding the best fit among the theta

```
In []: ## Plotting the best fit line
best_fit_x = np.linspace(0, 25, 20)
best_fit_y = [t[1] + t[0]*xx for xx in best_fit_x]
```

Ploting the best fit of the predicted item

```
In []: plt.figure(figsize=(10,6))
    plt.plot(X_df.TV, y_df, '.')
    plt.plot(best_fit_x, best_fit_y, '-')
    plt.axis([0,25,-5,25])
    plt.title('Sales vs. TV')
```

Out[]: Text(0.5, 1.0, 'Sales vs. TV')

