Practical 1

AIM :- Implement Linear Regression (Diabetes Dataset)

THEORY : -

Logistic regression is a classification model in machine learning, extensively used in clinical analysis. It uses probabilistic estimations which helps in understanding the relationship between the dependent variable and one or more independent variables. Diabetes, being one of the most common diseases around the world, when detected early, may prevent the progression of the disease and avoid other complications. In this work, we design a prediction model, that predicts whether a patient has diabetes, based on certain diagnostic measurements included in the dataset, and explore various techniques to boost the performance and accuracy.

CODE / OUTPUT : -

```
# Import Dependencies

import numpy as np

import matplotlib.pyplot as plt

from sklearn import datasets,linear_model,metrics

from sklearn.model_selection import train_test_split

from sklearn.metrics import mean_squared_error, r2_score

import seaborn as sns


# Load the diabetes dataset

diabetes=datasets.load_diabetes()


# X - feature vectors

# y - Target values

X=diabetes.data

y=diabetes.target


# splitting X and y into training and testing sets

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4,
```

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random_state=1)

# Create linear regression object
lin_reg=linear_model.LinearRegression()

# Train the model using train and test data
lin_reg.fit(X_train,y_train)
```

```
Out[6]: LinearRegression()
```

```
# Predict values for X_test data
predicted = lin_reg.predict(X_test)

# Regression coefficients
print('\n Coefficients are:\n',lin_reg.coef_)

# Intercept
print('\nIntercept : ',lin_reg.intercept_)

# variance score: 1 means perfect prediction
print('Variance score: ',lin_reg.score(X_test, y_test))
```

```
Coefficients are:
[ -59.73663337 -215.62170919  599.92621335  291.96724002 -829.65206295
  544.63994617  164.85191153  224.2392528   768.94426062   70.84982207]

Intercept :  152.89009028286725
Variance score:  0.4160439011127657
```

```
# Mean Squared Error
 print("Mean squared error: %.2f\n"
       % mean_squared_error(y_test, predicted))
# Original data of X_test
expected = y_test
```

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Mean squared error: 2962.93
```

```
# Plot a graph for expected and predicted values

plt.title('Linear Regression ( DIABETES Dataset)')
plt.scatter(expected,predicted,c='b',marker='.',s=36)
plt.plot(np.linspace(0, 330, 100),np.linspace(0, 330, 100), '--r', linewidth=2)
plt.show()
```