

In [107]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import numpy as np;
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.datasets import load_boston
from sklearn.metrics import mean_squared_error as MSE
#getting boston house price data from sklearn
boston = load_boston()
```

In [2]:

```
#finding no of datapoints in boston data

X = pd.DataFrame(data=boston['data'], columns=boston['feature_names']);
print("Shape of feature matrix: ", X.shape)
print(X.head(5))

print('*'*100)

y = pd.DataFrame(np.array(boston['target']))
print("Shape of output matrix", y.shape)
```

Shape of feature matrix: (506, 13)

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	\
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	

	PTRATIO	B	LSTAT
0	15.3	396.90	4.98
1	17.8	396.90	9.14
2	17.8	392.83	4.03
3	18.7	394.63	2.94
4	18.7	396.90	5.33

Shape of output matrix (506, 1)

In [183]:

```
class LinearRegression():
    def __init__(self, X, y, plot_error_per_iteration = False):
        self.X = X
        self.y = y #dimension: n*1
        self.data = pd.concat([X, y], axis=1);
        self.lr = 0.001 #defining learning rate
        self.b = np.random.normal(0,1,1);
        self.w = np.zeros((self.X.shape[1], 1)) #generating intial weight vector of dimension 1*d,
        where d is no of columns/features
        self.MSE_list = [];
        self.plot_error_per_iteration = plot_error_per_iteration;

    def computeDerivative(self, X, y):
        """
        function to obtain derivative value which is -2*x * (y - W_T * X)
        """
        wT_x = np.dot(X, self.w) #calculating x * w => multiplying X(n*d - dim) with weight(d*1), ou
        tput will be of n*1 - dim
        y_wT_x = y - wT_x #calculating y - (w_t * x) => output will be of dimension n*1
        x_prod = -2 * X;
        der = np.dot(y_wT_x.T, x_prod) #der will be of dimension 1 * d
        return self.w - self.lr * der.T;

    def computeIntercept(self, X, y):
        """
        function to compute new intercept
        """
```

```

    function to compute new intercept
    """
    wT_x = np.dot(X, self.w) #calculating x * w => multiplying X(n*d - dim) with weight(d*1), ou
tput will be of n*1 - dim
    y_wT_x = y - wT_x #calculating y - (w_t * x) => output will be of dimension n*1
    y_prod = -2 * y_wT_x;
    intercept = y_prod.sum(axis=0)
    intercept = intercept[0]
    return self.b - (self.lr * intercept)

def predict(self, X):
    """
    function to predict output given query points
    """
    wT_x = np.dot(self.w.T, X.T)
    y = wT_x + self.b;
    return y.T

def addMSE(self, ):
    """
    function to add mean squared error for each W, obtain in each iteration
    """
    y_pred = self.predict(self.X);
    self.MSE_list.append(MSE(y, y_pred));

def plotError(self, ):
    """
    function to plot error
    """
    plt.plot(np.arange(0, len(self.MSE_list)), self.MSE_list);
    plt.ylabel('MSE');
    plt.xlabel('iterations')
    plt.grid();
    plt.show();

def fit(self, ):
    itr = 1;
    while True:
        sample_data = self.data.sample(n=5)
        X = sample_data.loc[:, sample_data.columns != 0];
        y = sample_data.loc[:, [0]]
        new_w = self.computeDerivative(X, y);
        new_b = self.computeIntercept(X, y);
        if (np.all(new_w - self.w) < 0.01 and np.all(new_b - self.b) < 0.01):
            self.w = new_w
            self.b = new_b
            if self.plot_error_per_iteration:
                self.addMSE();
                self.plotError();
            break
        else:
            self.lr /= 2
            self.w = new_w
            self.b = new_b
            if self.plot_error_per_iteration:
                self.addMSE();

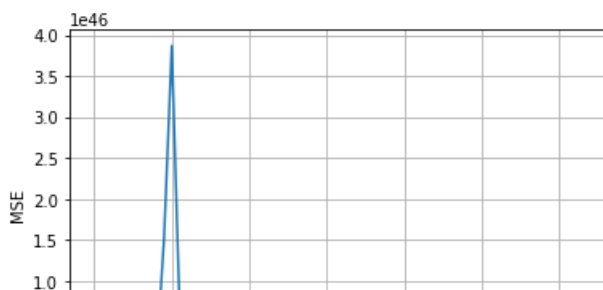
```

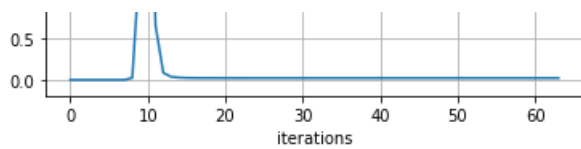
In [186]:

```

LR = LinearRegression(X, y, plot_error_per_iteration = True);
LR.fit()
y_pred = LR.predict(X);
self_model_error = MSE(y, y_pred)

```





In [177]:

```
from sklearn import linear_model
model = linear_model.SGDRegressor(penalty='none');
model.fit(X, y);
y_pred = model.predict(X);
sgd_model_error = MSE(y, y_pred)
```

In [178]:

```
from prettytable import PrettyTable

x = PrettyTable()
x.field_names = ["Model", "Mean squared error"]
x.add_row(["SGDRegressor", sgd_model_error])
x.add_row(["our created model", self_model_error ])
print(x)
```

Model	Mean squared error
SGDRegressor	1.216532166852742e+29
our created model	1.3757560711443702e+43