SGD_For_Linear_Reg

June 3, 2020

```
[1]: import numpy as np
  import matplotlib.pyplot as plt
  import seaborn as sns
  from sklearn.preprocessing import StandardScaler
  from sklearn.datasets import load_boston
  from sklearn.linear_model import LinearRegression
  from sklearn.model_selection import train_test_split
  import pandas as pd
  import sklearn.metrics as metrics
  import matplotlib.pyplot as plt
  from numpy import random
  import math
  from statistics import mean
  from scipy.signal import argrelextrema
```

/usr/local/lib/python3.6/dist-packages/statsmodels/tools/_testing.py:19:
FutureWarning: pandas.util.testing is deprecated. Use the functions in the public API at pandas.testing instead.
import pandas.util.testing as tm

0.1 Loading boston dataset to use in SkLearn Linear regression and custom Linear regression

0.2 Defining MSE loss function

```
[0]: def sq_error(ys_orig, ys_line):
    return ((np.sum(np.subtract(ys_line,ys_orig)**2)/len(ys_orig)))
```

0.3 Using SKlearn Linear regression and getting MSE for it

```
[4]: lm = LinearRegression()
lm.fit(X_train, Y_train)

Y_pred = lm.predict(X_test)

print('error with skleran algorithm {}'.format(sq_error(Y_test,Y_pred)))
```

error with skleran algorithm 28.530458765974597

0.4 Formulating Custom Linear regression model

```
[0]: ''' Mathematical formulation of
linear regression'''

def add_1_to_start(mat):
    r,c=mat.shape
    mat=np.c_[ mat, np.ones(r) ]
    mat[:,[0,-1]]=mat[:,[-1,0]]
    return mat

def best_fit_parameter(xs,ys):
    xs=add_1_to_start(xs)
    beta= np.linalg.inv(xs.T.dot(xs)).dot(xs.T.dot(ys))

return beta
```

0.5 Getting MSE for Custom Linear Regression model

```
[6]: weights= best_fit_parameter(X_train,Y_train)

reg_lr =add_1_to_start(X_test).dot(weights)

print('error with devised algorithm {}'.format(sq_error(Y_test,reg_lr)))
```

0.6 Loading Anew the boston dataset to work with SGD

```
[0]: boston=load_boston()
bos = pd.DataFrame(boston.data)

bos['PRICE'] = load_boston().target

X = bos.drop('PRICE', axis = 1)
Y_B= bos['PRICE']
Y_B=Y_B.values.reshape(-1,1)
```

0.7 Finding the importance of each column

/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:4: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

after removing the cwd from sys.path.

```
feature
[8]:
                   score
   5
            5 0.349578
   12
           12 0.319401
   10
           10 0.059426
   9
             9 0.046548
   4
            4 0.043025
             2 0.036391
   2
   7
             7 0.033221
   0
            0 0.027991
   8
            8 0.022631
   6
            6 0.021163
   3
            3 0.018433
           11 0.017371
   11
            1 0.004821
   1
```

0.8 Making a new dataframe with columns in decreasing order of importance

```
[0]: df_feature = pd.DataFrame()
     count=0
     for i,feat in enumerate(series['feature']):
        df_feature[i] = X[feat]
[10]: df_feature.head(2)
[10]:
             0
                    1
                           2
                                   3
                                                   5
                                                                                       10
                                                                                               11
     12
     0 6.575 4.98 15.3 296.0 0.538 2.31 ... 0.00632 1.0 65.2 0.0
                                                                                           396.9
     18.0
     1 \quad 6.421 \quad 9.14 \quad 17.8 \quad 242.0 \quad 0.469 \quad 7.07 \quad \dots \quad 0.02731 \quad 2.0 \quad 78.9 \quad 0.0 \quad 396.9
     0.0
     [2 rows x 13 columns]
```

0.9 Using only top 8 columns of importance

```
[0]: X_New=df_feature.iloc[:,:8]
Y_New=Y_B

[12]: print(Y_New.shape)
print(X_New.shape)

(506, 1)
(506, 8)
```

0.10 Defining A function to determine Learning rate with iterations

0.11 Defining a function which will run a batch of data through it and return the weight update which has the least train error (trying to mimic checkpoints in keras)

```
[0]: def update(x_data, X, y_data, Y, beta,lr_func):
      it = 0
      x=x_{data}
      y=y_data
      b_{deriv} = np.full((len(x[0]),1),0)
      learning_rate = lr_func(it)
      error_tr=[]
      b0= beta[:x.shape[1]]
      b=[]
      while True:
        error_tr.append(sq_error(y,x.dot(b0)))
        b.append(b0)
       #adding 12 regularizer derivative too
        lam=b0.T.dot(b0)
        b_{deriv} = b_{deriv} + (-2*(x.T.dot(y-x.dot(b0)))) + lam*2*b0
        b1 = b0 - (b_deriv * learning_rate)
        #running itereations for 20k times
        if it==20000:
          if b0.all()==b1.all():
            return b0
            break
          else:
            return b[np.argmin(error_tr)]
            break
        else:
          it += 1
          b0 = b1
          learning_rate= lr_func(it)
```

0.12 Defining SGD to pass batchs of size 'n' to update function resulting in an epoch

```
[0]: def SGD(X1, Y1,beta,lr_func):
      error_tr = []
      error_ts = []
      global X, Y
     xtr=X1
     ytr = Y1
     N=len(xtr)
     minn=np.inf
     k=0
     n=100
     for i in range((N//n)+1):
        if N>k:
          k+=n
          beta=update(xtr[k-n:k],X,ytr[k-n:k],Y,beta,lr_func)
          # error_tr.extend(tr)
          # error_ts.extend(ts)
          tr=sq_error(ytr,xtr.dot(beta))
          ts=sq_error(Y,X.dot(beta))
          if minn>tr:
            minn=tr
            min2=ts
            beta_min=beta
          # error_ts.append(ts)
          # error_tr.append(tr)
        else:
          beta=update(xtr[k-n:N],X,ytr[k-n:N],Y,beta,lr_func)
          # error_tr.extend(tr)
          # error_ts.extend(ts)
          tr=sq_error(ytr,xtr.dot(beta))
          ts=sq_error(Y,X.dot(beta))
          if minn>tr:
            minn=tr
```

```
min2=ts
beta_min=beta

# error_ts.append(ts)
# error_tr.append(tr)

return beta_min
```

0.13 Data preprocessing

```
[0]: #preprocessing data
xtr, X, ytr, Y = train_test_split(X_New,Y_New, test_size = 0.33)

ss= StandardScaler(with_mean= False)
xtr=ss.fit_transform(xtr)
X=ss.transform(X)

xtr=add_1_to_start(xtr)
X=add_1_to_start(X)

[0]: #initializing stuff
tr=[]
ts=[]
lr_func=step_decay
epoch=0
flag_change = False
```

0.14 Running SGD on randomly initialized weights

```
flag_change= True
     print('learnig fuction changed to step_decay2')
c=SGD(xtr,ytr,c,lr_func)
ts.append(sq_error(Y,X.dot(c)))
tr.append(sq_error(ytr,xtr.dot(c)))
epoch+=1
print('for epoch {} the train score is {} and test score is {}'.
\rightarrow format(epoch, tr[-1], ts[-1]))
if ts[-1] == ts[-2] or ts[-1] < 24:
  zeros_sgd= X.dot(c)
  pass
else:
  c=SGD(xtr,ytr,c,lr_func)
  ts.append(sq_error(Y,X.dot(c)))
  tr.append(sq_error(ytr,xtr.dot(c)))
  epoch+=1
  print('for epoch {} the train score is {} and test score is {}'.
\rightarrowformat(epoch,tr[-1],ts[-1]))
```

```
for epoch 1 the train score is 1162.4983495834683 and test score is
1189.4192008053915
for epoch 2 the train score is 745.3830561680218 and test score is
767.6103366750067
for epoch 3 the train score is 545.0978496345091 and test score is
567.4903129862008
for epoch 4 the train score is 426.28218914504896 and test score is
448.1959498009902
for epoch 5 the train score is 347.48921603671437 and test score is
368.60461865693065
for epoch 6 the train score is 291.54891139064455 and test score is
311.7114311532534
for epoch 7 the train score is 249.96202585310667 and test score is
269.1042618635233
for epoch 8 the train score is 218.0018205163483 and test score is
236.10491893707874
```

- for epoch 9 the train score is 192.81382579845672 and test score is 209.88703787308785
- for epoch 10 the train score is 172.56524593533382 and test score is 188.6343888934054
- for epoch 11 the train score is 156.02235682973395 and test score is 171.12295537511915
- for epoch 12 the train score is 142.32308527825066 and test score is 156.4961279228402
- for epoch 13 the train score is 130.75709373021303 and test score is 148.04962284513206
- for epoch 14 the train score is 122.35287004971308 and test score is 138.62618830379833
- for epoch 15 the train score is 115.02257122512636 and test score is 130.33100243462565
- for epoch 16 the train score is 108.59347010956114 and test score is 122.99559945473138
- for epoch 17 the train score is 102.92478936244602 and test score is 116.47539414876323
- for epoch 18 the train score is 97.90153795324551 and test score is 110.65178727084398
- for epoch 19 the train score is 93.42911149093193 and test score is 105.4267795811548
- for epoch 20 the train score is 89.4291713042585 and test score is 100.71884564260974
- for epoch 21 the train score is 85.83646384459792 and test score is 96.45974619558316
- for epoch 22 the train score is 82.5963401930994 and test score is 92.59203950731931
- for epoch 23 the train score is 79.66280340807002 and test score is 89.06711985264887
- for epoch 24 the train score is 76.99695838852672 and test score is 85.84365811130132
- for epoch 25 the train score is 74.56577187012581 and test score is 82.88635232808943
- for epoch 26 the train score is 72.3410736252682 and test score is 80.16491947976772
- for epoch 27 the train score is 70.29874686753418 and test score is 77.6532765771343
- for epoch 28 the train score is 68.41806823118753 and test score is 75.32887156818985
- for epoch 29 the train score is 66.68116684125077 and test score is 73.17213362703377
- for epoch 30 the train score is 65.07257882236776 and test score is 71.1660192255958
- for epoch 31 the train score is 63.578878750676665 and test score is 69.29563552531137
- for epoch 32 the train score is 62.18837347955824 and test score is 67.54792653986557

for epoch 33 the train score is 60.89084678609971 and test score is 65.91141052615706

for epoch 34 the train score is 59.67734562001791 and test score is 64.37595938775351

for epoch 35 the train score is 58.54000055768741 and test score is 62.9326126899732

for epoch 36 the train score is 57.471874493712825 and test score is 61.57342031090227

for epoch 37 the train score is 56.466834732263365 and test score is 60.29130887897515

for epoch 38 the train score is 55.51944453837237 and test score is 59.0799680432072

for epoch 39 the train score is 54.624870927079584 and test score is 57.93375333808117

for epoch 40 the train score is 53.77880604476612 and test score is 56.84760298041905

for epoch 41 the train score is 52.97739996229549 and test score is 55.81696640016207

for epoch 42 the train score is 52.21720307674967 and test score is 54.8377426839031

for epoch 43 the train score is 51.49511662558284 and test score is 53.90622741707572

for epoch 44 the train score is 50.808350067910936 and test score is 53.01906666186671

for epoch 45 the train score is 50.1543842934332 and test score is 52.173217014160436

for epoch 46 the train score is 49.53093978885207 and test score is 51.36591085275124

for epoch 47 the train score is 48.935949031463444 and test score is 50.59462603456495

for epoch 48 the train score is 48.36753249539092 and test score is 49.85705940614667

for epoch 49 the train score is 47.82397775212011 and test score is 49.15110359864489

for epoch 50 the train score is 47.30372122708008 and test score is 48.47482665439835

for epoch 51 the train score is 46.80533224090001 and test score is 47.82645410091596

for epoch 52 the train score is 46.327499019964584 and test score is 47.20435314482036

for epoch 53 the train score is 45.86901640785895 and test score is 46.60701870606497

for epoch 54 the train score is 45.428775048815446 and test score is 46.033061053009

for epoch 55 the train score is 45.00575184756473 and test score is 45.481194832931195

for epoch 56 the train score is 44.59900153813334 and test score is 44.95022932140286

for epoch 57 the train score is 44.20764921793012 and test score is 44.439059738380415

for epoch 58 the train score is 43.83088372366152 and test score is 43.94665949969969

for epoch 59 the train score is 43.46795174277143 and test score is 43.47207329038454

for epoch 60 the train score is 43.11815256870774 and test score is 43.014410861336806

for epoch 61 the train score is 42.78083342077995 and test score is 42.57284146394618

for epoch 62 the train score is 42.45538526001236 and test score is 42.1465888482707

for epoch 63 the train score is 42.141239041516975 and test score is 41.734926760008776

for epoch 64 the train score is 41.837862351713625 and test score is 41.3371748796906

for epoch 65 the train score is 41.5447563854437 and test score is 40.95269515462521

for epoch 66 the train score is 41.26145322379513 and test score is 40.580888480265266

for epoch 67 the train score is 40.98751337842308 and test score is 40.221191692934845

for epoch 68 the train score is 40.72252357246218 and test score is 39.873074840499996

for epoch 69 the train score is 40.46609473181801 and test score is 39.536038701517576

for epoch 70 the train score is 40.21786016385684 and test score is 39.20961252689327

for epoch 71 the train score is 39.97747390329261 and test score is 38.89335198110008

for epoch 72 the train score is 39.744609207489944 and test score is 38.58683726264819

for epoch 73 the train score is 39.51895718551629 and test score is 38.289671385807544

for epoch 74 the train score is 39.3002255471033 and test score is 38.001478607605726

for epoch 75 the train score is 39.088137459285306 and test score is 37.72190298590063

for epoch 76 the train score is 38.882430499877046 and test score is 37.4506070558743

for epoch 77 the train score is 38.682855698180944 and test score is 37.18727061367485

for epoch 78 the train score is 38.489176654390555 and test score is 36.931589597140466

for epoch 79 the train score is 38.30116873008829 and test score is 36.68327505458784

for epoch 80 the train score is 38.11861830307714 and test score is 36.44205219361252

for epoch 81 the train score is 37.941322080506566 and test score is 36.20765950266343

for epoch 82 the train score is 37.76908646489708 and test score is 35.97984793889643

for epoch 83 the train score is 37.60172696823768 and test score is 35.75838017647077

for epoch 84 the train score is 37.439067669828894 and test score is 35.54302991002622

for epoch 85 the train score is 37.28094071398992 and test score is 35.33358120860308

for epoch 86 the train score is 37.127185844141806 and test score is 35.12982791572688

for epoch 87 the train score is 36.977649970127864 and test score is 34.9315730917886

for epoch 88 the train score is 36.832186765941984 and test score is 34.73862849522204

for epoch 89 the train score is 36.690656295313964 and test score is 34.55081409930677

for epoch 90 the train score is 36.55292466284464 and test score is 34.3679576417213

for epoch 91 the train score is 36.418863688605825 and test score is 34.189894204234115

for epoch 92 the train score is 36.28835060431777 and test score is 34.01646582016074

for epoch 93 the train score is 36.16126776938478 and test score is 33.84752110741956

for epoch 94 the train score is 36.037502405239415 and test score is 33.68291492522413

for epoch 95 the train score is 35.916946346574925 and test score is 33.5225080526102

for epoch 96 the train score is 35.799495808182186 and test score is 33.36616688716095

for epoch 97 the train score is 35.685051166214095 and test score is 33.21376316243033

for epoch 98 the train score is 35.5735167528056 and test score is 33.0651736826884

for epoch 99 the train score is 35.464800663075145 and test score is 32.92028007373902

for epoch 100 the train score is 35.358814573607276 and test score is 32.77896854865492

for epoch 101 the train score is 35.25547357159837 and test score is 32.64112968736823

for epoch 102 the train score is 35.154695993915816 and test score is 32.50665822915391

for epoch 103 the train score is 35.05640327537873 and test score is 32.375452877106156

for epoch 104 the train score is 34.96051980562893 and test score is 32.247416113789626

for epoch 105 the train score is 34.86697279400636 and test score is 32.1224540273065

for epoch 106 the train score is 34.77569214189697 and test score is 32.000476147082594

for epoch 107 the train score is 34.68661032205544 and test score is 31.881395288728864

for epoch 108 the train score is 34.59966226444717 and test score is 31.765127407380376

for epoch 109 the train score is 34.51478524819144 and test score is 31.65159145896887

for epoch 110 the train score is 34.43191879921255 and test score is 31.540709268914053

for epoch 111 the train score is 34.351004593241875 and test score is 31.432405407766055

for epoch 112 the train score is 34.27198636383617 and test score is 31.326607073362272

for epoch 113 the train score is 34.19480981510352 and test score is 31.22324397909398

for epoch 114 the train score is 34.11942253885082 and test score is 31.12224824790513

for epoch 115 the train score is 34.04577393588492 and test score is 31.02355431167843

for epoch 116 the train score is 33.97381514122248 and test score is 30.927098815681266

for epoch 117 the train score is 33.90349895297564 and test score is 30.83282052776884

for epoch 118 the train score is 33.83477976470224 and test score is 30.740660252068484

for epoch 119 the train score is 33.76761350101923 and test score is 30.65056074687951

for epoch 120 the train score is 33.701957556292754 and test score is 30.562466646544994

for epoch 121 the train score is 33.63777073623331 and test score is 30.476324387070992

for epoch 122 the train score is 33.57501320223172 and test score is 30.392082135276254

for epoch 123 the train score is 33.51364641828607 and test score is 30.309689721280563

for epoch 124 the train score is 33.45363310037535 and test score is 30.22909857413624

for epoch 125 the train score is 33.39493716815212 and test score is 30.150261660443526

for epoch 126 the train score is 33.33752369882218 and test score is 30.073133425770504

for epoch 127 the train score is 33.28135888310337 and test score is 29.997669738742232

for epoch 128 the train score is 33.22640998314706 and test score is 29.92382783764272

for epoch 129 the train score is 33.17264529232435 and test score is 29.851566279405635

for epoch 130 the train score is 33.12003409677907 and test score is 29.780844890863722

for epoch 131 the train score is 33.06854663865571 and test score is 29.71162472213864

for epoch 132 the train score is 33.018154080920986 and test score is 29.64386800206539

for epoch 133 the train score is 32.96882847369416 and test score is 29.57753809554067

for epoch 134 the train score is 32.92054272201474 and test score is 29.5125994627029

for epoch 135 the train score is 32.87327055497354 and test score is 29.449017619847726

for epoch 136 the train score is 32.826986496142645 and test score is 29.386759101997054

for epoch 137 the train score is 32.781665835238826 and test score is 29.325791427033266

for epoch 138 the train score is 32.737284600963406 and test score is 29.26608306133044

for epoch 139 the train score is 32.693819534958756 and test score is 29.207603386799114

for epoch 140 the train score is 32.65124806683202 and test score is 29.150322669286876

for epoch 141 the train score is 32.60954829019263 and test score is 29.094212028262326

for epoch 142 the train score is 32.568698939658994 and test score is 29.03924340772516

for epoch 143 the train score is 32.52867936878789 and test score is 28.985389548285596

for epoch 144 the train score is 32.48946952888145 and test score is 28.93262396035046

for epoch 145 the train score is 32.451049948637504 and test score is 28.880920898377205

for epoch 146 the train score is 32.413401714599765 and test score is 28.83025533613725

for epoch 147 the train score is 32.3765064523737 and test score is 28.780602942943947

for epoch 148 the train score is 32.340346308573785 and test score is 28.731940060804877

for epoch 149 the train score is 32.304903933469205 and test score is 28.68424368245351

for epoch 150 the train score is 32.27016246429737 and test score is 28.637491430222017

for epoch 151 the train score is 32.23610550921581 and test score is 28.591661535718547

for epoch 152 the train score is 32.20271713186435 and test score is 28.54673282027243

for epoch 153 the train score is 32.169981836511354 and test score is 28.502684676113102

for epoch 154 the train score is 32.137884553759065 and test score is 28.45949704825374

for epoch 155 the train score is 32.106410626782356 and test score is 28.417150417044663

for epoch 156 the train score is 32.07554579808229 and test score is 28.375625781373344

for epoch 157 the train score is 32.04527619672693 and test score is 28.334904642477788

for epoch 158 the train score is 32.015588326064 and test score is 28.294968988349964

for epoch 159 the train score is 31.986469051882903 and test score is 28.255801278707224

for epoch 160 the train score is 31.957905591008327 and test score is 28.21738443050219

for epoch 161 the train score is 31.929885500306277 and test score is 28.179701803951858

for epoch 162 the train score is 31.90239666608787 and test score is 28.142737189065752

for epoch 163 the train score is 31.875427293894045 and test score is 28.106474792651547

for epoch 164 the train score is 31.84896589864137 and test score is 28.070899225774347

for epoch 165 the train score is 31.82300129512274 and test score is 28.035995491660298

for epoch 166 the train score is 31.797522588840327 and test score is 28.001748974018188

for epoch 167 the train score is 31.772519167163285 and test score is 27.96814542576772

for epoch 168 the train score is 31.747980690793025 and test score is 27.935170958153105

for epoch 169 the train score is 31.723897085527238 and test score is 27.902812030232546

for epoch 170 the train score is 31.700258534309935 and test score is 27.871055438725868

for epoch 171 the train score is 31.677055469554087 and test score is 27.839888308204557

for epoch 172 the train score is 31.654278565729296 and test score is 27.809298081613644

for epoch 173 the train score is 31.63191873220274 and test score is 27.779272511112193

for epoch 174 the train score is 31.609967106324053 and test score is 27.749799649219558

for epoch 175 the train score is 31.5884150467423 and test score is 27.720867840252996

for epoch 176 the train score is 31.567254126952065 and test score is 27.692465712053572

for epoch 177 the train score is 31.546476129052888 and test score is 27.664582167979155

for epoch 178 the train score is 31.526073037718653 and test score is 27.637206379161306

for epoch 179 the train score is 31.50603703436673 and test score is 27.61032777701377

for epoch 180 the train score is 31.48636049152207 and test score is 27.583936045985883

for epoch 181 the train score is 31.4670359673654 and test score is 27.55802111654629

for epoch 182 the train score is 31.44805620045922 and test score is 27.532573158392314

for epoch 183 the train score is 31.429414104649634 and test score is 27.507582573878498

for epoch 184 the train score is 31.411102764130483 and test score is 27.48303999165094

for epoch 185 the train score is 31.393115428670573 and test score is 27.458936260486603

for epoch 186 the train score is 31.375445508991014 and test score is 27.435262443321374

for epoch 187 the train score is 31.35808657229516 and test score is 27.412009811470778

for epoch 188 the train score is 31.341032337937936 and test score is 27.3891698390258

for epoch 189 the train score is 31.32427667323478 and test score is 27.366734197426208

for epoch 190 the train score is 31.30781358940292 and test score is 27.34469475019945

for epoch 191 the train score is 31.291637237631914 and test score is 27.323043547863417

for epoch 192 the train score is 31.275741905274874 and test score is 27.30177282298139

for epoch 193 the train score is 31.260122012161876 and test score is 27.280874985372535

for epoch 194 the train score is 31.24477210702712 and test score is 27.260342617464705

for epoch 195 the train score is 31.229686864045878 and test score is 27.24016846978663

for epoch 196 the train score is 31.214861079479757 and test score is 27.220345456598217

for epoch 197 the train score is 31.200289668425587 and test score is 27.200866651650877

for epoch 198 the train score is 31.185967661661678 and test score is 27.18172528407221

for epoch 199 the train score is 31.17189020259312 and test score is 27.162914734375434

for epoch 200 the train score is 31.158052544287635 and test score is 27.144428530585184

for epoch 201 the train score is 31.144450046602522 and test score is 27.126260344477597

for epoch 202 the train score is 31.131078173397682 and test score is 27.10840398793002

for epoch 203 the train score is 31.11793248983211 and test score is 27.09085340937813

for epoch 204 the train score is 31.10500865974183 and test score is 27.07360269037498

for epoch 205 the train score is 31.092302443094336 and test score is 27.05664604224889

for epoch 206 the train score is 31.079809693521405 and test score is 27.03997780286056

for epoch 207 the train score is 31.06752635592111 and test score is 27.02359243344816

for epoch 208 the train score is 31.05544846413383 and test score is 27.00748451556805

for epoch 209 the train score is 31.0435721386836 and test score is 26.991648748118276

for epoch 210 the train score is 31.031893584587475 and test score is 26.976079944448134

for epoch 211 the train score is 31.02040908922627 and test score is 26.960773029547813

for epoch 212 the train score is 31.009115020279598 and test score is 26.945723037319503

for epoch 213 the train score is 30.998007823718265 and test score is 26.93092510792263

for epoch 214 the train score is 30.987084021854937 and test score is 26.91637448519267

for epoch 215 the train score is 30.976340211451486 and test score is 26.90206651413394

for epoch 216 the train score is 30.965773061878647 and test score is 26.887996638479056

for epoch 217 the train score is 30.955379313329463 and test score is 26.8741603983178

for epoch 218 the train score is 30.9451557750818 and test score is 26.860553427788677

for epoch 219 the train score is 30.935099323811098 and test score is 26.847171452835255

for epoch 220 the train score is 30.92520690194912 and test score is 26.834010289022373

for epoch 221 the train score is 30.915475516090076 and test score is 26.82106583941324

for epoch 222 the train score is 30.90590223543945 and test score is 26.808334092500978

for epoch 223 the train score is 30.896484190306964 and test score is 26.795811120198234

for epoch 224 the train score is 30.88721857064129 and test score is 26.783493075880035

for epoch 225 the train score is 30.878102624604182 and test score is 26.771376192478385

for epoch 226 the train score is 30.869133657183657 and test score is 26.759456780625857

for epoch 227 the train score is 30.860309028846103 and test score is 26.747731226853208

for epoch 228 the train score is 30.851626154224736 and test score is 26.73619599182876

for epoch 229 the train score is 30.843082500842296 and test score is 26.724847608647988

for epoch 230 the train score is 30.834675587870407 and test score is 26.71368268116693

for epoch 231 the train score is 30.826402984919824 and test score is 26.70269788237705

learnig fuction changed to step_decay2

for epoch 232 the train score is 30.826395942030263 and test score is 26.702687889493525

for epoch 233 the train score is 30.826388899748633 and test score is 26.702677897252045

for epoch 234 the train score is 30.826381858074917 and test score is 26.702667905652614

for epoch 235 the train score is 30.826374817009125 and test score is 26.702657914695223

for epoch 236 the train score is 30.82636777655119 and test score is 26.702647924379818

for epoch 237 the train score is 30.826360736701155 and test score is 26.702637934706452

for epoch 238 the train score is 30.826353697458945 and test score is 26.70262794567503

for epoch 239 the train score is 30.826346658824562 and test score is 26.702617957285575

for epoch 240 the train score is 30.82633962079798 and test score is 26.70260796953802

for epoch 241 the train score is 30.826332583379177 and test score is 26.702597982432373

for epoch 242 the train score is 30.826325546568146 and test score is 26.702587995968603

for epoch 243 the train score is 30.82631851036486 and test score is 26.702578010146723

for epoch 244 the train score is 30.8263114747693 and test score is 26.7025680249667

for epoch 245 the train score is 30.826304439781467 and test score is 26.702558040428503

for epoch 246 the train score is 30.826297405401302 and test score is 26.702548056532116

for epoch 247 the train score is 30.826290371628836 and test score is 26.702538073277527

for epoch 248 the train score is 30.82628333846403 and test score is

```
26.702528090664725
```

21 ---> 22

23

for epoch 249 the train score is 30.82627630590683 and test score is 26.702518108693667

for epoch 250 the train score is 30.826269273957244 and test score is 26.702508127364332

for epoch 251 the train score is 30.826262242615215 and test score is 26.70249814667669

for epoch 252 the train score is 30.826255211880767 and test score is 26.70248816663072

for epoch 253 the train score is 30.82624818175393 and test score is 26.702478187226486

for epoch 254 the train score is 30.826241152234672 and test score is 26.702468208464012

for epoch 255 the train score is 30.826234123322852 and test score is 26.70245823034304

for epoch 256 the train score is 30.82622709501852 and test score is 26.70244825286365

for epoch 257 the train score is 30.826220067321646 and test score is 26.702438276025866

for epoch 258 the train score is 30.826213040232275 and test score is 26.702428299829716

for epoch 259 the train score is 30.82620601375036 and test score is 26.702418324275108

for epoch 260 the train score is 30.82619898787587 and test score is 26.702408349362045

for epoch 261 the train score is 30.826191962608732 and test score is 26.702398375090453

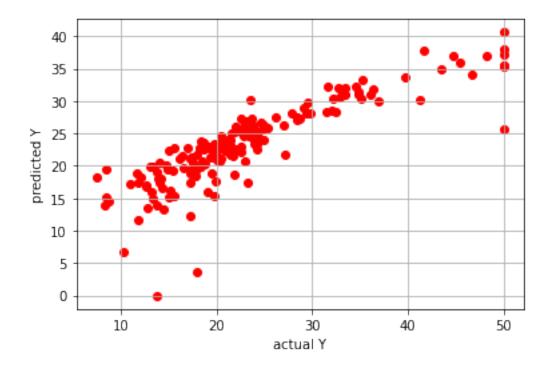
error_tr.extend(tr)

beta=update(xtr[k-n:k],X,ytr[k-n:k],Y,beta,lr_func)

KeyboardInterrupt:

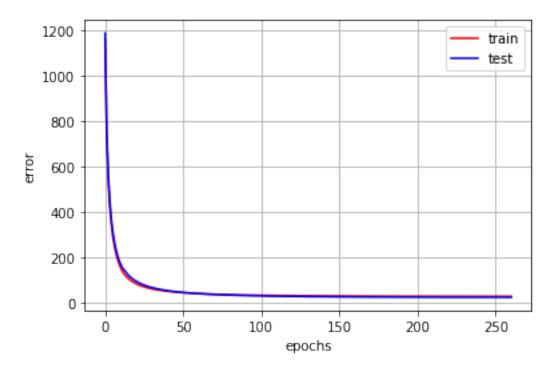
0.15 Forcefully interrupted the loop with keyboard interrupt as no progress can be seen

```
[98]: plt.scatter(Y,X.dot(c), color='r')
plt.xlabel('actual Y')
plt.ylabel('predicted Y')
plt.grid()
```



```
[104]: plt.plot(tr, color='r',label='train')
plt.plot(ts, color='b', label= 'test')
```

```
plt.ylabel('error')
plt.xlabel('epochs')
plt.legend()
plt.grid()
```



```
[86]: '''we see TRAIN ERROR HIGHER THAN TEST ERROR as model generalizes well enough but Test data size is too small hence these reults'''

print('Train error {}'.format(sq_error(ytr,xtr.dot(c))))

print('Test error {}'.format(sq_error(Y,X.dot(c))))
```

Train error 30.826191962608732 Test error 26.702398375090453

```
[0]: from prettytable import PrettyTable
    x = PrettyTable()
    x.field_names = ["S.NO.", "MODEL", "Error value"]
    x.add_row(["1", "Skleran Algorithm", sq_error(Y_test,Y_pred)])
    x.add_row(["2", "Formulated Algorithm", sq_error(Y_test,reg_lr)])
    x.add_row(["3", "SGD with random weights initialization",sq_error(Y,X.dot(c))])
[106]: print(x)
```

S.NO.		Error value
1 2 3	Skleran Algorithm Formulated Algorithm SGD with random weights initialization	28.530458765974597 28.53045876597665 26.702398375090453

[107]: | sudo apt-get install pandoc texlive-xetex !jupyter nbconvert --to pdf SGD_for_Linear_reg.ipynb

Reading package lists... Done Building dependency tree Reading state information... Done pandoc is already the newest version (1.19.2.4~dfsg-1build4). pandoc set to manually installed. The following additional packages will be installed: fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre javascript-common libcupsfilters1 libcupsimage2 libgs9 libgs9-common libijs-0.35 libjbig2dec0 libjs-jquery libkpathsea6 libpotrace0 libptexenc1 libruby2.5 libsynctex1 libtexlua52 libtexluajit2 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-did-you-mean ruby-minitest ruby-net-telnet ruby-power-assert ruby-test-unit ruby2.5 rubygems-integration t1utils tex-common tex-gyre texlive-base texlive-binaries texlive-fonts-recommended texlive-latex-base texlive-latex-extra texlive-latex-recommended texlive-pictures texlive-plain-generic tipa

Suggested packages:

fonts-noto apache2 | lighttpd | httpd poppler-utils ghostscript fonts-japanese-mincho | fonts-ipafont-mincho fonts-japanese-gothic | fonts-ipafont-gothic fonts-arphic-ukai fonts-arphic-uming fonts-nanum ri ruby-dev bundler debhelper gv | postscript-viewer perl-tk xpdf-reader | pdf-viewer texlive-fonts-recommended-doc texlive-latex-base-doc python-pygments icc-profiles libfile-which-perl libspreadsheet-parseexcel-perl texlive-latex-extra-doc texlive-latex-recommended-doc texlive-pstricks dot2tex prerex ruby-tcltk | libtcltk-ruby texlive-pictures-doc vprerex

The following NEW packages will be installed:

fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre javascript-common libcupsfilters1 libcupsimage2 libgs9 libgs9-common libijs-0.35 libjbig2dec0 libjs-jquery libkpathsea6 libpotrace0 libptexenc1 libruby2.5 libsynctex1 libtexlua52 libtexluajit2 libzzip-0-13 lmodern poppler-data preview-latex-style rake ruby ruby-did-you-mean ruby-minitest ruby-net-telnet ruby-power-assert ruby-test-unit ruby2.5 rubygems-integration t1utils tex-common tex-gyre texlive-base texlive-binaries texlive-fonts-recommended texlive-latex-base texlive-latex-extra texlive-latex-recommended texlive-pictures

```
texlive-plain-generic texlive-xetex tipa
```

O upgraded, 46 newly installed, O to remove and 32 not upgraded.

Need to get 146 MB of archives.

After this operation, 460 MB of additional disk space will be used.

Get:1 http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-droid-fallback all 1:6.0.1r16-1.1 [1,805 kB]

Get:2 http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-lato all 2.0-2
[2,698 kB]

Get:3 http://archive.ubuntu.com/ubuntu bionic/main amd64 poppler-data all 0.4.8-2 [1,479 kB]

Get:4 http://archive.ubuntu.com/ubuntu bionic/main amd64 tex-common all 6.09
[33.0 kB]

Get:5 http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-lmodern all 2.004.5-3 [4,551 kB]

Get:6 http://archive.ubuntu.com/ubuntu bionic/main amd64 fonts-noto-mono all 20171026-2 [75.5 kB]

Get:7 http://archive.ubuntu.com/ubuntu bionic/universe amd64 fonts-texgyre all 20160520-1 [8,761 kB]

Get:8 http://archive.ubuntu.com/ubuntu bionic/main amd64 javascript-common all
11 [6,066 B]

Get:9 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libcupsfilters1 amd64 1.20.2-Oubuntu3.1 [108 kB]

Get:10 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libcupsimage2 amd64 2.2.7-1ubuntu2.8 [18.6 kB]

Get:11 http://archive.ubuntu.com/ubuntu bionic/main amd64 libijs-0.35 amd64 0.35-13 [15.5 kB]

Get:12 http://archive.ubuntu.com/ubuntu bionic/main amd64 libjbig2dec0 amd64
0.13-6 [55.9 kB]

Get:13 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libgs9-common all 9.26~dfsg+0-Oubuntu0.18.04.12 [5,092 kB]

Get:14 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libgs9 amd64 9.26~dfsg+0-0ubuntu0.18.04.12 [2,264 kB]

Get:15 http://archive.ubuntu.com/ubuntu bionic/main amd64 libjs-jquery all 3.2.1-1 [152 kB]

Get:16 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libkpathsea6 amd64 2017.20170613.44572-8ubuntu0.1 [54.9 kB]

Get:17 http://archive.ubuntu.com/ubuntu bionic/main amd64 libpotrace0 amd64
1.14-2 [17.4 kB]

Get:18 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 libptexenc1 amd64 2017.20170613.44572-8ubuntu0.1 [34.5 kB]

Get:19 http://archive.ubuntu.com/ubuntu bionic/main amd64 rubygems-integration all 1.11 [4,994 B]

Get:20 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 ruby2.5 amd64 2.5.1-1ubuntu1.6 [48.6 kB]

Get:21 http://archive.ubuntu.com/ubuntu bionic/main amd64 ruby amd64 1:2.5.1
[5,712 B]

Get:22 http://archive.ubuntu.com/ubuntu bionic-updates/main amd64 rake all 12.3.1-1ubuntu0.1 [44.9 kB]

24% [22 rake 42.0 kB/44.9 kB 93%]^C

[NbConvertApp] WARNING | pattern u'SGD_for_Linear_reg.ipynb' matched no files This application is used to convert notebook files (*.ipynb) to various other formats.

WARNING: THE COMMANDLINE INTERFACE MAY CHANGE IN FUTURE RELEASES.

Options

Arguments that take values are actually convenience aliases to full Configurables, whose aliases are listed on the help line. For more information on full configurables, see '--help-all'.

--execute

Execute the notebook prior to export.

--allow-errors

Continue notebook execution even if one of the cells throws an error and include the error message in the cell output (the default behaviour is to abort conversion). This flag is only relevant if '--execute' was specified, too. --no-input

Exclude input cells and output prompts from converted document.

This mode is ideal for generating code-free reports.

--stdout

Write notebook output to stdout instead of files.

--stdin

read a single notebook file from stdin. Write the resulting notebook with default basename 'notebook.*'

--inplace

Run nbconvert in place, overwriting the existing notebook (only relevant when converting to notebook format)

-у

Answer yes to any questions instead of prompting.

--clear-output

Clear output of current file and save in place, overwriting the existing notebook.

--debug

set log level to logging.DEBUG (maximize logging output)

--no-prompt

Exclude input and output prompts from converted document.

--generate-config

generate default config file

--nbformat=<Enum> (NotebookExporter.nbformat_version)

Default: 4

Choices: [1, 2, 3, 4]

The nbformat version to write. Use this to downgrade notebooks.

--output-dir=<Unicode> (FilesWriter.build_directory)

Default: ''

```
Directory to write output(s) to. Defaults to output to the directory of each
   notebook. To recover previous default behaviour (outputting to the current
   working directory) use . as the flag value.
--writer=<DottedObjectName> (NbConvertApp.writer_class)
   Default: 'FilesWriter'
    Writer class used to write the results of the conversion
--log-level=<Enum> (Application.log level)
   Default: 30
    Choices: (0, 10, 20, 30, 40, 50, 'DEBUG', 'INFO', 'WARN', 'ERROR',
'CRITICAL')
    Set the log level by value or name.
--reveal-prefix=<Unicode> (SlidesExporter.reveal_url_prefix)
    Default: u''
    The URL prefix for reveal.js (version 3.x). This defaults to the reveal CDN,
   but can be any url pointing to a copy of reveal.js.
   For speaker notes to work, this must be a relative path to a local copy of
   reveal.js: e.g., "reveal.js".
    If a relative path is given, it must be a subdirectory of the current
   directory (from which the server is run).
   See the usage documentation
    (https://nbconvert.readthedocs.io/en/latest/usage.html#reveal-js-html-
    slideshow) for more details.
--to=<Unicode> (NbConvertApp.export_format)
   Default: 'html'
    The export format to be used, either one of the built-in formats
    ['asciidoc', 'custom', 'html', 'latex', 'markdown', 'notebook', 'pdf',
    'python', 'rst', 'script', 'slides'] or a dotted object name that represents
    the import path for an `Exporter` class
--template=<Unicode> (TemplateExporter.template_file)
   Default: u''
   Name of the template file to use
--output=<Unicode> (NbConvertApp.output_base)
   Default: ''
    overwrite base name use for output files. can only be used when converting
    one notebook at a time.
--post=<DottedOrNone> (NbConvertApp.postprocessor_class)
   Default: u''
    PostProcessor class used to write the results of the conversion
--config=<Unicode> (JupyterApp.config_file)
   Default: u''
   Full path of a config file.
To see all available configurables, use `--help-all`
Examples
```

The simplest way to use nbconvert is

```
> jupyter nbconvert mynotebook.ipynb
   which will convert mynotebook.ipynb to the default format (probably HTML).
   You can specify the export format with `--to`.
   Options include ['asciidoc', 'custom', 'html', 'latex', 'markdown',
'notebook', 'pdf', 'python', 'rst', 'script', 'slides'].
   > jupyter nbconvert --to latex mynotebook.ipynb
   Both HTML and LaTeX support multiple output templates. LaTeX includes
   'base', 'article' and 'report'. HTML includes 'basic' and 'full'. You
   can specify the flavor of the format used.
   > jupyter nbconvert --to html --template basic mynotebook.ipynb
   You can also pipe the output to stdout, rather than a file
   > jupyter nbconvert mynotebook.ipynb --stdout
   PDF is generated via latex
   > jupyter nbconvert mynotebook.ipynb --to pdf
   You can get (and serve) a Reveal.js-powered slideshow
   > jupyter nbconvert myslides.ipynb --to slides --post serve
   Multiple notebooks can be given at the command line in a couple of
   different ways:
   > jupyter nbconvert notebook*.ipynb
   > jupyter nbconvert notebook1.ipynb notebook2.ipynb
   or you can specify the notebooks list in a config file, containing::
       c.NbConvertApp.notebooks = ["my_notebook.ipynb"]
   > jupyter nbconvert --config mycfg.py
```

[0]: