

In [1]: `#importing the main libraries that we need them for our regression model`

```
In [2]: import pandas as pd
import numpy as np
import warnings
warnings.simplefilter('ignore', FutureWarning)
```

In [3]: `#we will download our dataset`
`concrete_data=pd.read_csv('https://s3-api.us-geo.objectstorage.softlayer.`
`concrete_data.head() #to see the first 5 data line of the dataset`

Out[3]:

	Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate	Fine Aggregate	Age	St
0	540.0	0.0	0.0	162.0	2.5	1040.0	676.0	28	
1	540.0	0.0	0.0	162.0	2.5	1055.0	676.0	28	
2	332.5	142.5	0.0	228.0	0.0	932.0	594.0	270	
3	332.5	142.5	0.0	228.0	0.0	932.0	594.0	365	
4	198.6	132.4	0.0	192.0	0.0	978.4	825.5	360	

In [4]: `#to clean the dataset beofore using it , checking for any mising values`

In [5]: `concrete_data.shape`

Out[5]: (1030, 9)

In [6]: `concrete_data.describe()`

Out[6]:

	Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	A
count	1030.000000	1030.000000	1030.000000	1030.000000	1030.000000	103
mean	281.167864	73.895825	54.188350	181.567282	6.204660	97
std	104.506364	86.279342	63.997004	21.354219	5.973841	7
min	102.000000	0.000000	0.000000	121.800000	0.000000	80
25%	192.375000	0.000000	0.000000	164.900000	0.000000	93
50%	272.900000	22.000000	0.000000	185.000000	6.400000	96
75%	350.000000	142.950000	118.300000	192.000000	10.200000	102
max	540.000000	359.400000	200.100000	247.000000	32.200000	114

In [7]: `concrete_data.isnull().sum()`

```
Out [7]: Cement          0
Blast Furnace Slag    0
Fly Ash               0
Water                0
Superplasticizer     0
Coarse Aggregate     0
Fine Aggregate       0
Age                  0
Strength             0
dtype: int64
```

```
In [8]: concrete_data_columns=concrete_data.columns
predictors=concrete_data[concrete_data_columns[concrete_data_columns!='St
target=concrete_data['Strength'] #the Strenght cloumn is the Target
```

```
In [9]: #we can store the predictors number in a variable called pred_no
concrete_no=concrete_data.shape[1]
predictors.head(9)
```

```
Out [9]:
```

	Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate	Fine Aggregate	Age
0	540.0	0.0	0.0	162.0	2.5	1040.0	676.0	28
1	540.0	0.0	0.0	162.0	2.5	1055.0	676.0	28
2	332.5	142.5	0.0	228.0	0.0	932.0	594.0	270
3	332.5	142.5	0.0	228.0	0.0	932.0	594.0	365
4	198.6	132.4	0.0	192.0	0.0	978.4	825.5	360
5	266.0	114.0	0.0	228.0	0.0	932.0	670.0	90
6	380.0	95.0	0.0	228.0	0.0	932.0	594.0	365
7	380.0	95.0	0.0	228.0	0.0	932.0	594.0	28
8	266.0	114.0	0.0	228.0	0.0	932.0	670.0	28

```
In [10]: target.head(9)
```

```
Out [10]: 0    79.99
1    61.89
2    40.27
3    41.05
4    44.30
5    47.03
6    43.70
7    36.45
8    45.85
Name: Strength, dtype: float64
```

```
In [11]: #first normalize the data then use it withthe model
```

```
In [12]: predictors_normal=predictors-(predictors.mean())/predictors.std()
```

```
In [13]: predictors_normal.head(5)
```

Out [13]:

	Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate	
0	537.309562	-0.856472	-0.846733	153.497358	1.461362	1027.48721	6
1	537.309562	-0.856472	-0.846733	153.497358	1.461362	1042.48721	6
2	329.809562	141.643528	-0.846733	219.497358	-1.038638	919.48721	5
3	329.809562	141.643528	-0.846733	219.497358	-1.038638	919.48721	5
4	195.909562	131.543528	-0.846733	183.497358	-1.038638	965.88721	8

In [14]: `pred_no=predictors_normal.shape[1]`

B.Build use the normalized data

In [15]: `import keras #first import keras and sklearn using this code
import sklearn`

Using TensorFlow backend.
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/sklearn/uti
ls/validation.py:37: DeprecationWarning: distutils Version classes are dep
recated. Use packaging.version instead.
LARGE_SPARSE_SUPPORTED = LooseVersion(scipy_version) >= '0.14.0'

In [16]: `#import the important pakages from keras liabrary to build a regression m`In [17]: `from keras.models import Sequential`In [18]: `from keras.layers import Dense #to create our layers`In [19]: `#we split the data using sklearn`

```
In [20]: def regression_model():
#define our neural network with one hidden layers
model=Sequential()
model.add(Dense(10, activation='relu', input_shape=(pred_no,)))
model.add(Dense(10, activation='relu'))
model.add(Dense(10, activation='relu'))
model.add(Dense(1))
#sompling our model using adam optimizer and the mean squared error a
model.compile(optimizer='adam', loss='mean_squared_error')
```

```
In [21]: from sklearn.model_selection import train_test_split
x=predictors_normal
y=concrete_data['Strength']
y.head()
```

```
Out [21]: 0    79.99
1    61.89
2    40.27
3    41.05
4    44.30
Name: Strength, dtype: float64
```

In [22]: `x.head()`

Out[22]:

	Cement	Blast Furnace Slag	Fly Ash	Water	Superplasticizer	Coarse Aggregate	
0	537.309562	-0.856472	-0.846733	153.497358	1.461362	1027.48721	6
1	537.309562	-0.856472	-0.846733	153.497358	1.461362	1042.48721	6
2	329.809562	141.643528	-0.846733	219.497358	-1.038638	919.48721	5
3	329.809562	141.643528	-0.846733	219.497358	-1.038638	919.48721	5
4	195.909562	131.543528	-0.846733	183.497358	-1.038638	965.88721	8

In [23]: `x.shape`

Out[23]: (1030, 8)

In [24]: `#split the dataset to 30% for test and the rest for training`

In [25]: `x_train,x_test,y_train,y_test =train_test_split(x,y,random_state=10,test_`

In [26]: `x_train.shape`

Out[26]: (721, 8)

In [27]: `y_test.shape`

Out[27]: (309,)

In [28]: `#training the model using 50 epochs
#first we call the model function that we have defiened it above to build
model=regression_model()`

WARNING:tensorflow:From /home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/keras/backend/tensorflow_backend.py:68: The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From /home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/keras/backend/tensorflow_backend.py:508: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/keras/backend/tensorflow_backend.py:3837: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/keras/optimizers.py:757: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

In [29]: `#we will train our model on the training data using fit function`

In [30]: `model_fit=model.fit(x_train,y_train,epochs=50, verbose=0,validatio_data=(`

```
-----
-
AttributeError                                Traceback (most recent call las
t)
/tmp/ipykernel_1435/2163848.py in <module>
----> 1 model_fit=model.fit(x_train,y_train,epochs=50, verbose=0,validatio
_data=(x_test,y_test ))

AttributeError: 'NoneType' object has no attribute 'fit'
```

```
In [31]: #evaluate the model
scores=model.evaluate(x_test,y_test,verbose=0)
```

```
-----
-
AttributeError                                Traceback (most recent call las
t)
/tmp/ipykernel_1435/2593022917.py in <module>
      1 #evaluate the model
----> 2 scores=model.evaluate(x_test,y_test,verbose=0)

AttributeError: 'NoneType' object has no attribute 'evaluate'
```

```
In [32]: #computing the mean squarred error and print it
mean_squared_error=model_fit.history['val_loss'][-1]
print(mean_squared_error)
```

```
-----
-
NameError                                    Traceback (most recent call las
t)
/tmp/ipykernel_1435/830681345.py in <module>
      1 #computing the mean squarred error and print it
----> 2 mean_squared_error=model_fit.history['val_loss'][-1]
      3 print(mean_squared_error)

NameError: name 'model_fit' is not defined
```

```
In [33]: #we will repeat steps 1 to 3 50 times and calculate the error in each tim
```

```
In [34]: msq_list= [] #create the mean squared error list in each iteration
for train_cycle in range(50):
    x_train,x_test,y_train,y_test=train_test_split(x,y,random_state=10,te
    model_fit=model.fit(x_train,y_train,epochs=50,verbose=0,validation_da
    mean_squared_error=model_fit.history['val_loss'][-1] #calculate the m
    msq_list.append(mean_squared_error) #fill up the mean squared error l
    print('train_cycle {}: mean_squared_error{}'.format(train_cycle+1,mea
```

```

-----
-
AttributeError                                Traceback (most recent call las
t)
/tmp/ipykernel_1435/98300980.py in <module>
      2 for train_cycle in range(50):
      3     x_train,x_test,y_train,y_test=train_test_split(x,y,random_stat
e=10,test_size=0.30) #split the data to test_size 30%
----> 4     model_fit=model.fit(x_train,y_train,epochs=50,verbose=0,valida
tion_data=(x_test,y_test)) #train and validate the model
      5     mean_squared_error=model_fit.history['val_loss'][-1] #calculat
e the mean squared error
      6     msq_list.append(mean_squared_error) #fill up the mean squared
error list during the 50 cycles

AttributeError: 'NoneType' object has no attribute 'fit'

```

In [35]: `#report the mean and the standard deviation of the mean squared error`

In [36]: `print('the mean of the mean squared errors is : {}'.format(np.mean(msq_li
print('the standard deviation of the mean squared errors is : {}'.format(`

```

the mean of the mean squared errors is : nan
the standard deviation of the mean squared errors is : nan
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/numpy/core/
fromnumeric.py:3441: RuntimeWarning: Mean of empty slice.
  out=out, **kwargs)
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/numpy/core/
_methods.py:189: RuntimeWarning: invalid value encountered in double_scala
rs
  ret = ret.dtype.type(ret / rcount)
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/numpy/core/
_methods.py:263: RuntimeWarning: Degrees of freedom <= 0 for slice
  keepdims=keepdims, where=where)
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/numpy/core/
_methods.py:223: RuntimeWarning: invalid value encountered in true_divide
  subok=False)
/home/jupyterlab/conda/envs/python/lib/python3.7/site-packages/numpy/core/
_methods.py:254: RuntimeWarning: invalid value encountered in double_scala
rs
  ret = ret.dtype.type(ret / rcount)

```

The mean of the mean squared error when we use 100 epochs is a bit higher than the case B using the 50 epochs, which means even increasing the epochs does not help with the performance of the model

In []:

In []: