```
In [10]: import numpy as np # Importing NumPy library
import pandas as pd # Importing Pandas library
import matplotlib.pyplot as plt # Importing Matplotlib library's "pyplot" modu
import seaborn as sns # Importing Seaborn library

import os
```

In [11]: data = pd.read_csv("Regression_Dataset.csv")
 data

Out[11]:		S.No	Left- Lateral- Ventricle	Left- Inf- Lat- Vent	Left- Cerebellum- White- Matter	Left- Cerebellum- Cortex	Left- Thalamus	Left- Caudate	Left- Putamen	Le1 Pallidu
	0	1	22916.9	982.7	15196.7	55796.4	6855.5	2956.4	4240.7	2223
	1	2	22953.2	984.5	15289.7	55778.6	6835.1	3064.2	4498.6	2354
	2	3	23320.4	1062.1	15382.1	55551.2	7566.0	3231.7	4456.2	1995
	3	4	24360.0	1000.5	14805.4	54041.8	8004.6	3137.3	4262.2	1983
	4	5	25769.4	1124.4	16331.1	54108.6	6677.4	2964.4	4204.6	2409
	•••	•••								
	4221	4222	27065.6	532.4	12425.1	51042.9	6354.8	3822.6	4490.5	2019
	4222	4223	28408.8	912.7	14024.8	43103.5	6060.7	3114.2	3731.0	1937
	4223	4224	34467.9	1659.6	12744.5	54924.8	6256.7	3573.4	3526.6	2189
	4224	4225	31627.5	1334.4	15883.2	57148.2	6982.4	4475.8	4464.4	2317
	4225	4226	14879.4	704.2	11346.6	50468.5	6935.4	3258.5	3751.5	2226

4226 rows × 141 columns

```
In [12]: X = data.drop(["Age"], axis=1)
y = data['Age']
```

In [13]: from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test = train_test_split(X,y,test_size=0.2, random_state)

In [14]: from sklearn.ensemble import GradientBoostingRegressor
boost = GradientBoostingRegressor(n_estimators=100, learning_rate=0.1, max_dept
boost.fit(x_train, y_train)

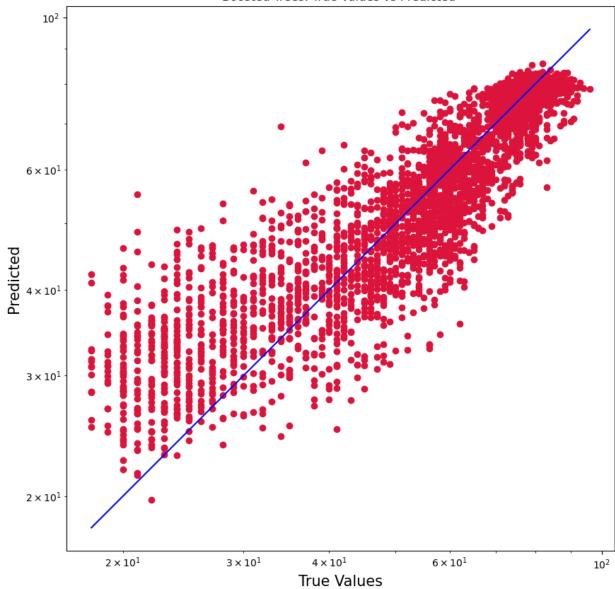
Out[14]: GradientBoostingRegressor(max_depth=1, random_state=0)

In [15]: from sklearn.model_selection import cross_val_predict # For K-Fold Cross Valid
from sklearn.metrics import r2_score # For find accuracy with R2 Score
from sklearn.metrics import mean_squared_error # For MSE
from math import sqrt # For squareroot operation

y_pred_boosted_train = boost.predict(x_train)

```
y pred boosted test = boost.predict(x test)
         r2_boosted_train = r2_score(y_train, y_pred_boosted_train)
         print("Training R^2 for Boosted Trees Model: ", r2_boosted_train)
         r2_boosted_test = r2_score(y_test, y_pred_boosted_test)
         print("Testing R^2 for Boosted Trees Model: ", r2 boosted test)
         RMSE_boosted_train = sqrt(mean_squared_error(y_train, y_pred_boosted_train))
         print("RMSE for Training Data: ", RMSE_boosted_train)
         RMSE boosted test = sqrt(mean squared error(y test, y pred boosted test))
         print("RMSE for Testing Data: ", RMSE_boosted_test)
         Training R<sup>2</sup> for Boosted Trees Model: 0.8440751927299814
         Testing R^2 for Boosted Trees Model: 0.8282934968877312
         RMSE for Training Data: 7.8774363396543565
         RMSE for Testing Data: 8.49144878829781
In [16]: true_value = y_train
         predicted_value = y_pred_boosted_train
In [17]: plt.figure(figsize=(10,10))
         plt.scatter(true value, predicted value, c='crimson')
         plt.yscale('log')
         plt.xscale('log')
         p1 = max(max(predicted value), max(true value))
         p2 = min(min(predicted value), min(true value))
         plt.plot([p1, p2], [p1, p2], 'b-')
         plt.xlabel('True Values', fontsize=15)
         plt.ylabel('Predicted', fontsize=15)
         plt.title("Boosted Trees: True Values vs Predicted ")
         plt.axis('equal')
         plt.show()
```

Boosted Trees: True Values vs Predicted



In []: