```
In [1]: import numpy as np
        import seaborn as sns
        import pandas as pd
        import matplotlib.pyplot as plt
        import sklearn
        from sklearn import preprocessing
        from sklearn.model selection import train test split
        from sklearn.metrics import r2 score
        import xgboost as xgb
        from xgboost.sklearn import XGBRegressor
        from sklearn.model selection import GridSearchCV
        from sklearn.multioutput import MultiOutputRegressor
        import lightgbm as lgb #conda install lightqbm
        from sklearn.metrics import (roc_curve, auc, accuracy_score)
        from sklearn.metrics import r2 score, mean squared error
        import tensorflow as tf
        from keras.models import Sequential
        from keras.layers import Dense
        from keras.optimizers import Adam
        from keras.callbacks import EarlyStopping
        import pandas as p
        import sklearn
        from sklearn import preprocessing
        from sklearn.model selection import train test split
        from sklearn.metrics import r2 score
        from keras.optimizers import Adadelta,Adagrad,RMSprop,Adam,Adamax,Nadam
        from sklearn.model selection import GridSearchCV
        data = pd.read csv("D:\\javeed\\DS documents\\abid dataset\\electric motor tem
        p\\pmsm_temperature_data.csv")
```

Using TensorFlow backend.

```
In [16]: X= data.drop(['stator_yoke','pm'], axis = 1)
Y = data[['stator_yoke','pm']]
train_x,test_x,train_y,test_y = train_test_split(X,Y, test_size = 0.33, strat
ify=data.profile_id)
train_x = train_x.drop(['profile_id'], axis = 1)
test_x = test_x.drop(['profile_id'], axis = 1)
```

```
In [9]: from sklearn.preprocessing import MinMaxScaler
    sc = MinMaxScaler(feature_range = (0, 1))
    training_set_scaled = sc.fit_transform(train_x)
    testing_set_scaled = sc.fit_transform(test_x)
```

Neural Network

```
In [35]: from keras.layers import ELU
         from keras.layers import Dropout
         from keras.optimizers import Adadelta,Adagrad,RMSprop,Adam,Adamax,Nadam
         model3 = Sequential()
         model3.add(Dense(500,input shape=(10,)))
         model3.add(ELU())
         model3.add(Dense(400))
         model3.add(ELU())
         model3.add(Dense(300))
         model3.add(ELU())
         model3.add(Dense(200))
         model3.add(ELU())
         model3.add(Dense(100))
         model3.add(ELU())
         model3.add(Dense(2))
         import keras
         adam1 =Nadam(learning rate=0.001, beta 1=0.9, beta 2=0.99)
         model3.compile(loss='mean_squared_error', optimizer= adam1, metrics= ['accurac
         model3.fit(train x, train y, epochs=5, batch size=115, validation data=(test x
         , test y))
         WARNING:tensorflow:From D:\javeed\DS documents\anaconda install\lib\site-pack
         ages\keras\backend\tensorflow_backend.py:422: The name tf.global_variables is
         deprecated. Please use tf.compat.v1.global variables instead.
```

```
Train on 668706 samples, validate on 329364 samples
Epoch 1/5
63 - accuracy: 0.8318 - val_loss: 0.0523 - val_accuracy: 0.8624
Epoch 2/5
72 - accuracy: 0.8868 - val loss: 0.0332 - val accuracy: 0.8957
Epoch 3/5
51 - accuracy: 0.9073 - val_loss: 0.0221 - val_accuracy: 0.9158
Epoch 4/5
94 - accuracy: 0.9190 - val loss: 0.0177 - val accuracy: 0.9197
Epoch 5/5
59 - accuracy: 0.9269 - val loss: 0.0140 - val accuracy: 0.9317
```

```
Out[35]: <keras.callbacks.callbacks.History at 0x1e83daf15c0>
```

```
In [215]:
          nn100 train RMSE=np.mean((nn100 train pred - train y)**2, axis=0)
          nn100 train RMSE = np.sqrt(nn100 train RMSE)
          #stator yoke
                          0.053246
          #pm
                          0.406511
          nn100_test_RMSE=np.mean((nn100_test_pred - test_y)**2, axis=0)
          nn100_test_RMSE= np.sqrt(nn100_test_RMSE)
          from sklearn.metrics import r2 score
          nn100 train R2 = r2 score(train y, nn100 train pred)#0.9553449358644539
          nn100_test_R2 = r2_score(test_y, nn100_test_pred)#0.9552267680020445
In [216]:
          nn100 train = nn100 train RMSE
          nn100 train R2 = pd.Series(nn100 train R2)
          nn100 train = nn100 train.append(nn100 train R2)
          nn100 train = pd.DataFrame(nn100 train)
          nn_train=nn100_train.rename(columns={0: 'nn100_train'})
          nn100 test = nn100 test RMSE
          nn100 test R2 = pd.Series(nn100 test R2)
          nn100_test = nn100_test.append(nn100_test_R2)
          nn100 test = pd.DataFrame(nn100 test)
          nn_test=nn100_test.rename(columns={0: 'nn100_test'})
```

XGBRegressor

```
In [55]: import numpy as np
         import pandas as pd
         from sklearn import preprocessing
         import xgboost as xgb
         from xgboost.sklearn import XGBRegressor
         import datetime
         from sklearn.model selection import GridSearchCV
         from sklearn.multioutput import MultiOutputRegressor
         xgb multioutputregressor = MultiOutputRegressor(xgb.XGBRegressor()).fit(train
         x, train y)
         xgb train RMSE=np.mean((xgb multioutputregressor.predict(train x) - train y)**
         2, axis=0)
         xgb_test_RMSE=np.mean((xgb_multioutputregressor.predict(test_x) - test_y)**2,
         axis=0)
         xgb train RMSE = np.sqrt(xgb train RMSE)
         #stator_yoke
                         0.062031
                         0.401400
         #pm
         xgb_test_RMSE = np.sqrt(xgb_test_RMSE)
         #stator_yoke
                         0.062406
                         0.400764
         from sklearn.metrics import r2 score
         xgb_train_pred = xgb_multioutputregressor.predict(train_x)
         xgb test pred = xgb multioutputregressor.predict(test x)
         xgb_train_R2 = r2_score(train_y, xgb_train_pred)#0.9169291776206683
         xgb_test_R2 = r2_score(test_y, xgb_test_pred)#0.916830099751504
         xgb train score = xgb multioutputregressor.score(train x,train y)
         xgb test score = xgb multioutputregressor.score(test x,test y)
         xgb train RMSE=np.mean((xgb multioutputregressor.predict(train x) - train y)**
         2, axis=0)
         xgb test RMSE=np.mean((xgb multioutputregressor.predict(test x) - test y)**2,
         axis=0)
         xgb train RMSE = np.sqrt(xgb train RMSE)
```

[15:11:32] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:13:55] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

```
In [120]: xgb_train = xgb_train_RMSE
    xgb_train_R2 = pd.Series(xgb_train_R2)
    xgb_train = xgb_train.append(xgb_train_R2)
    xgb_train = pd.DataFrame(xgb_train)
    xgb_train=xgb_train.rename(columns={0: 'xgb_train'})

xgb_test = xgb_test_RMSE
    xgb_test_R2 = pd.Series(xgb_test_R2)
    xgb_test = xgb_test.append(xgb_test_R2)
    xgb_test = pd.DataFrame(xgb_test)
    xgb_test=xgb_test.rename(columns={0: 'xgb_test'})
```

LGBM

```
In [17]:
         from sklearn.multioutput import MultiOutputRegressor
         import matplotlib.pvplot as plt
         import lightgbm as lgb #conda install lightqbm
         from sklearn.metrics import (roc curve, auc, accuracy score)
         from sklearn.model selection import GridSearchCV
         lgb_multioutput = MultiOutputRegressor(lgb.LGBMRegressor(learning_rate=0.05,ma
         x depth=7,n jobs=1,n estimators=1000,nthread=-1))
         lgb multioutput.fit(train_x, train_y)
         lgb train RMSE=np.mean((lgb multioutput.predict(train x) - train y)**2, axis=0
         lgb train RMSE = np.sqrt(lgb train RMSE)
         #stator_yoke
                         0.026573
         #pm
                         0.164039
         lgb test RMSE=np.mean((lgb multioutput.predict(test x) - test y)**2, axis=0)
         lgb test RMSE= np.sqrt(lgb test RMSE)
         #stator_yoke
                         0.027163
         #pm
                         0.166146
         from sklearn.metrics import r2 score
         lgb_train_pred = lgb_multioutput.predict(train_x)
         lgb test pred = lgb multioutput.predict(test x)
         lgb train R2 = r2 score(train y, lgb train pred)#0.9861326172294375
         lgb_test_R2 = r2_score(test_y, lgb_test_pred)#0.985719127660748
         lgb train score = lgb multioutput.score(train x,train y)
         lgb test score =lgb multioutput.score(test x,test y)
In [61]: | lgb_train = lgb_train_RMSE
         a = np.float64(lgb_train_R2)
         a = pd.Series(a)
         lgb train = lgb train.append(a)
         lgb_train = pd.DataFrame(lgb_train)
         lgb train=lgb train.rename(columns={0: 'lgb train'})
         lgb_test= lgb_test_RMSE
         lgb test R2 = pd.Series(lgb test R2)
         lgb test = lgb test.append(lgb test R2)
         lgb test= pd.DataFrame(lgb test)
         lgb test=lgb test.rename(columns={0: 'lgb test'})
```

polynomial with 2 degree

```
In [211]: from sklearn.linear model import LinearRegression
          from sklearn.model selection import train test split
          from sklearn.preprocessing import PolynomialFeatures
          from sklearn.metrics import r2 score,mean squared error
          quad = PolynomialFeatures (degree = 2)
          train quad = quad.fit transform(train x)
          test quad=quad.fit transform(test x)
          plr = MultiOutputRegressor(LinearRegression()).fit(train quad,train y)
          Y train pred = plr.predict(train quad)
          Y_test_pred = plr.predict(test_quad)
          print('Polynomial Linear Regression:' ,plr.score(test quad,test y))#0.91164228
          53434928
          poly_R2 =plr.score(test_quad,test_y)
          #RMSE
          poly Train RMSE = np.sqrt(mean squared error(train y,Y train pred))#0.29693444
          794659535
          poly Test RMSE = np.sqrt(mean squared error(test y,Y test pred))#0.29665945037
          595576
          #R2
          poly train R2 = r2 score(train y, Y train pred)#0.9111167243895444
          poly test R2 = r2 score(test y, Y test pred)#0.911157317010483
```

Polynomial Linear Regression: 0.9115021337511484

```
In [212]: poly_test = {'poly_test':[poly_Test_RMSE,poly_test_R2]}
    poly_test = pd.DataFrame(poly_test)
    poly_test=poly_test.rename(index={0:"stator_yoke", 1: 0})
    poly_test

poly_train = {'poly_train':[poly_Train_RMSE,poly_train_R2]}
    poly_train = pd.DataFrame(poly_train)
    poly_train=poly_train.rename(index={0:"stator_yoke", 1: 0})
    poly_train
```

Out[212]:

```
        poly_train

        stator_yoke
        0.297187

        0
        0.910940
```

AdaBoostRegressor

```
In [19]: from sklearn.ensemble import AdaBoostRegressor
         from sklearn.multioutput import MultiOutputRegressor
         #ada multioutput = MultiOutputRegressor(AdaBoostRegressor(learning rate=0.01,
          n estimators=200))
         ada multioutput = MultiOutputRegressor(AdaBoostRegressor())
         ada_multioutput.fit(train_x, train_y)
         ada train RMSE=np.mean((ada multioutput.predict(train x) - train y)**2, axis=0
         ada train RMSE = np.sqrt(ada train RMSE)
         ada_test_RMSE=np.mean((ada_multioutput.predict(test_x) - test_y)**2, axis=0)
         ada test RMSE= np.sqrt(ada test RMSE)
         from sklearn.metrics import r2 score
         ada_train_pred = ada_multioutput.predict(train_x)
         ada test pred = ada multioutput.predict(test x)
         ada train R2 = r2 score(train y, ada train pred)\#0.8376333232840278
         ada_test_R2 = r2_score(test_y, ada_test_pred)#0.8376362009220268
         ada train score = ada multioutput.score(train x,train y)#0.8321852774400107
         ada_test_score =ada_multioutput.score(test_x,test_y)#0.8319736250251978
In [46]:
         ada train = ada train RMSE
         ada train R2 = pd.Series(ada train R2)
         ada train = ada train.append(ada train R2)
         ada train = pd.DataFrame(ada train)
         ada_train=ada_train.rename(columns={0: 'ada_train'})
         ada test = ada test RMSE
         ada test R2 = pd.Series(ada test R2)
         ada test =ada test.append(ada test R2)
         ada test = pd.DataFrame(ada test)
         ada_test=ada_test.rename(columns={0: 'ada_test'})
In [24]: | ada_train_score
```

RandomForestRegressor

Out[24]: 0.8321852774400107

```
In [20]:
         from sklearn import ensemble
         from sklearn.multioutput import MultiOutputRegressor
         import matplotlib.pyplot as plt
         import lightgbm as lgb #conda install lightqbm
         from sklearn.metrics import (roc curve, auc, accuracy score)
         from sklearn.model selection import GridSearchCV
         #rf multioutput = MultiOutputRegressor(ensemble.RandomForestRegressor(n estima
         tors=500, n jobs=1, verbose=1))
         rf multioutput = MultiOutputRegressor(ensemble.RandomForestRegressor())
         #lgb_multioutput = MultiOutputRegressor(lgb.LGBMRegressor(learning_rate=0.05,m
         ax depth=7,n jobs=1,n estimators=1000,nthread=-1))
         rf_multioutput.fit(train_x, train_y)
         rf_train_RMSE=np.mean((rf_multioutput.predict(train_x) - train_y)**2, axis=0)
         rf train RMSE = np.sqrt(rf train RMSE)
         rf test RMSE=np.mean((rf multioutput.predict(test x) - test y)**2, axis=0)
         rf_test_RMSE= np.sqrt(rf_test_RMSE)
         from sklearn.metrics import r2 score
         rf_train_pred = rf_multioutput.predict(train_x)
         rf_test_pred = rf_multioutput.predict(test_x)
         rf train R2 = r2 score(train y, rf train pred)#0.9997814045437312
         rf test R2 = r2 score(test y, rf test pred)#0.9990043465338014
         rf train score = rf multioutput.score(train x,train y)#0.9997838183035852
         rf_test_score =rf_multioutput.score(test_x,test_y)#0.9989816978585642
```

D:\javeed\DS documents\anaconda install\lib\site-packages\sklearn\ensemble\fo rest.py:246: FutureWarning: The default value of n_estimators will change fro m 10 in version 0.20 to 100 in 0.22.

"10 in version 0.20 to 100 in 0.22.", FutureWarning)

D:\javeed\DS documents\anaconda install\lib\site-packages\sklearn\ensemble\fo rest.py:246: FutureWarning: The default value of n_estimators will change fro m 10 in version 0.20 to 100 in 0.22.

"10 in version 0.20 to 100 in 0.22.", FutureWarning)

```
In [47]: rf_train = rf_train_RMSE
    rf_train_R2 = pd.Series(rf_train_R2)
    rf_train = rf_train.append(rf_train_R2)
    rf_train = pd.DataFrame(rf_train)
    rf_train=rf_train.rename(columns={0: 'rf_train'})

    rf_test = rf_test_RMSE
    rf_test_R2 = pd.Series(rf_test_R2)
    rf_test = rf_test.append(rf_test_R2)
    rf_test = pd.DataFrame(rf_test)
    rf_test=rf_test.rename(columns={0: 'rf_test'})
```

with profile_id

```
In [53]: X= data.drop(['stator yoke','pm'], axis = 1)
         Y = data[['stator_yoke','pm']]
         x_train,x_test,y_train,y_test = train_test_split(X,Y, test_size = 0.33, strat
         ify=data.profile id)
         import numpy as np
         import pandas as pd
         from sklearn import preprocessing
         import xgboost as xgb
         from xgboost.sklearn import XGBRegressor
         import datetime
         from sklearn.model selection import GridSearchCV
         from sklearn.multioutput import MultiOutputRegressor
         p_xgb_multioutputregressor = MultiOutputRegressor(xgb.XGBRegressor()).fit(x_tr
         ain, y train)
         p xgb train RMSE=np.mean((p xgb multioutputregressor.predict(x train) - y trai
         n)**2, axis=0)
         p xgb test RMSE=np.mean((p xgb multioutputregressor.predict(x test) - y test)*
         *2, axis=0)
         p_xgb_train_RMSE = np.sqrt(p_xgb_train_RMSE)
         #stator yoke
                        0.062031
         #pm
                         0.401400
         p_xgb_test_RMSE = np.sqrt(p_xgb_test_RMSE)
         #stator yoke
                        0.062406
                         0.400764
         #pm
         from sklearn.metrics import r2 score
         p xgb train pred = p xgb multioutputregressor.predict(x train)
         p xgb test pred = p xgb multioutputregressor.predict(x test)
         p_xgb_train_R2 = r2_score(y_train, p_xgb_train_pred)#0.9169291776206683
         p xgb test R2 = r2 score(y test, p xgb test pred)#0.916830099751504
```

[15:05:23] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

[15:08:05] WARNING: C:/Jenkins/workspace/xgboost-win64_release_0.90/src/objec tive/regression_obj.cu:152: reg:linear is now deprecated in favor of reg:squa rederror.

LGBMRegressor

```
In [28]:
         from sklearn.multioutput import MultiOutputRegressor
         import matplotlib.pyplot as plt
         import lightgbm as p lgb #conda install lightqbm
         from sklearn.metrics import (roc curve, auc, accuracy score)
         from sklearn.model selection import GridSearchCV
         p_lgb_multioutput = MultiOutputRegressor(p_lgb.LGBMRegressor(learning_rate=0.0
         5, max depth=7, n jobs=1, n estimators=1000, nthread=-1))
         p lgb multioutput.fit(x train, y train)
         p lgb train RMSE=np.mean((p lgb multioutput.predict(x train) - y train)**2, ax
         is=0)
         p lgb train RMSE = np.sqrt(p lgb train RMSE)
         #stator_yoke
                        0.025517
         #pm
                         0.115780
         p lgb test RMSE=np.mean((p lgb multioutput.predict(x test) - y test)**2, axis=
         0)
         p_lgb_test_RMSE= np.sqrt(p_lgb_test_RMSE)
         #stator yoke
                         0.025978
         #pm
                         0.117266
         from sklearn.metrics import r2_score
         p lgb train pred = p lgb multioutput.predict(x train)
         p lgb test pred = p lgb multioutput.predict(x test)
         p_lgb_train_R2 = r2_score(y_train, p_lgb_train_pred)#00.9929160367267138
         p lgb test R2 = r2 score(y test, p lgb test pred)\#0.9927245780698586
         p_lgb_train_score= p_lgb_multioutput.score(x_train,y_train)#0.9929160367267138
         p_lgb_test_score =p_lgb_multioutput.score(x_test,y_test)#0.9927245780698586
In [64]: p lgb train = p lgb train RMSE
         a = np.float64(p_lgb_train_R2)
         a = pd.Series(a)
         p lgb train = p lgb train.append(a)
         p lgb train = pd.DataFrame(p lgb train)
         p_lgb_train=p_lgb_train.rename(columns={0: 'p_lgb_train'})
         p_lgb_test= p_lgb_test_RMSE
         p_lgb_test_R2 = pd.Series(p_lgb_test_R2)
         p lgb test = p lgb test.append(p lgb test R2)
         p lgb test = pd.DataFrame(p lgb test)
```

p_lgb_test=p_lgb_test.rename(columns={0: 'p_lgb_test'})

polynomialFeatures

```
In [ ]: from sklearn.linear model import LinearRegression
          from sklearn.model selection import train_test_split
          from sklearn.preprocessing import polynomialFeatures
          from sklearn.metrics import r2 score,mean squared error
          quad = polynomialFeatures (degree = 2)
          train quad = quad.fit transform(x train)
          test quad=quad.fit transform(x test)
          plr = MultiOutputRegressor(LinearRegression().fit(train quad,y train))
          Y train pred = plr.predict(train quad)
          Y_test_pred = plr.predict(test_quad)
          print('P_polynomial Linear Regression:' ,plr.score(test_quad,y_test))#0.911642
          2853434928
          p_poly_R2 =plr.score(test_quad,y_test)
          #RMSE
          p poly Train RMSE = np.sqrt(mean squared error(y train,Y train pred))#0.296934
          44794659535
          p poly Test RMSE = np.sqrt(mean squared error(y test,Y test pred))#0.296659450
          37595576
          #R2
          p poly train R2 = r2 score(y train, Y train pred)#0.9111167243895444
          p poly test R2 = r2 score(y test, Y test pred)#0.911157317010483
In [187]: | p_poly_test = {'p_poly_test':[p_poly_Test_RMSE,p_poly_test_R2 ]}
          p_poly_test =pd.DataFrame(p_poly_test)
          p poly test=p poly test.rename(index={0:"stator yoke", 1: 0})
          p_poly_train = {'p_poly_train':[p_poly_Train_RMSE,p_poly_train_R2 ]}
          p poly train =pd.DataFrame(p poly train)
          p_poly_train=p_poly_train.rename(index={0:"stator_yoke", 1 : 0})
Out[187]:
                      p_poly_train
           stator_yoke
                        0.292664
```

AdaBoostRegressor

0

0.913636

```
In [31]:
         from sklearn.ensemble import AdaBoostRegressor
         from sklearn.multioutput import MultiOutputRegressor
         #p ada multioutput = MultiOutputRegressor(P adaBoostRegressor(learning rate=0.
         01, n estimators=200))
         p ada multioutput = MultiOutputRegressor(AdaBoostRegressor())
         p_ada_multioutput.fit(x_train, y_train)
         p_ada_train_RMSE=np.mean((p_ada_multioutput.predict(x_train) - y_train)**2, ax
         is=0)
         p ada train RMSE = np.sqrt(p ada train RMSE)
         #stator_yoke
                       0.150531
                         0.548006
         #pm
         p_ada_test_RMSE=np.mean((p_ada_multioutput.predict(x_test) - y_test)**2, axis=
         0)
         p ada test RMSE= np.sqrt(p ada test RMSE)
         #stator yoke
                       0.150474
         #pm
                         0.547884
         from sklearn.metrics import r2 score
         p_ada_train_pred = p_ada_multioutput.predict(x_train)
         p_ada_test_pred = p_ada_multioutput.predict(x_test)
         p ada train R2 = r2 score(y train, p ada train pred)#0.8372726936873622
         p_ada_test_R2 = r2_score(y_test, p_ada_test_pred)#0.8372322396789236
         p_ada_train_score= p_ada_multioutput.score(x_train,y_train)#0.9929160367267138
         p ada test score =p ada multioutput.score(x test,y test)
```

RandomForestRegressor

```
In [32]: from sklearn import ensemble
         from sklearn.multioutput import MultiOutputRegressor
         import matplotlib.pyplot as plt
         import lightgbm as lgb #conda install lightqbm
         from sklearn.metrics import (roc curve, auc, accuracy score)
         from sklearn.model selection import GridSearchCV
         #p_rf_multioutput = MultiOutputRegressor(ensemble.RandomForestRegressor(n_esti
         mators=500, n jobs=1, verbose=1))
         p rf multioutput = MultiOutputRegressor(ensemble.RandomForestRegressor())
         #lgb_multioutput = MultiOutputRegressor(lgb.LGBMRegressor(learning_rate=0.05,m
         ax depth=7,n jobs=1,n estimators=1000,nthread=-1))
         p_rf_multioutput.fit(x_train, y_train)
         p_rf_train_RMSE=np.mean((p_rf_multioutput.predict(x_train) - y_train)**2, axis
         =0)
         p rf train RMSE = np.sqrt(p rf train RMSE)
         p_rf_test_RMSE=np.mean((p_rf_multioutput.predict(x_test) - y_test)**2, axis=0)
         p rf test RMSE= np.sqrt(p rf test RMSE)
         from sklearn.metrics import r2 score
         p_rf_train_pred = p_rf_multioutput.predict(x_train)
         p_rf_test_pred = p_rf_multioutput.predict(x test)
         p rf train R2 = r2 score(y train, p rf train pred)#0.9997814045437312
         p_rf_test_R2 = r2_score(y_test, p_rf_test_pred)#0.9990043465338014
         p rf train score= p rf multioutput.score(x train,y train)#0.9929160367267138
         p_rf_test_score =p_rf_multioutput.score(x_test,y_test)
```

D:\javeed\DS documents\anaconda install\lib\site-packages\sklearn\ensemble\fo rest.py:246: FutureWarning: The default value of n_estimators will change fro m 10 in version 0.20 to 100 in 0.22.

"10 in version 0.20 to 100 in 0.22.", FutureWarning)

D:\javeed\DS documents\anaconda install\lib\site-packages\sklearn\ensemble\fo rest.py:246: FutureWarning: The default value of n_estimators will change fro m 10 in version 0.20 to 100 in 0.22.

"10 in version 0.20 to 100 in 0.22.", FutureWarning)

```
In [74]: q=data.loc[(data['profile_id']==20) & (data['profile_id'] ==6)]
```

In [219]: RMSE_R2_values = pd.concat([poly_test,poly_train,xgb_train, xgb_test,lgb_train
,lgb_test,ada_train,ada_test,rf_train,rf_test,p_poly_train,p_poly_test,p_xgb_t
rain, p_xgb_test,p_lgb_train,p_lgb_test,p_ada_train,p_ada_test,p_rf_train,p_rf
_test,nn_train, nn_test], axis=1, sort=False)
RMSE_R2_values = RMSE_R2_values.rename(index={'stator_yoke': 'stator_yoke RMS
E','pm':'pm RMSE',0:"R2"})
RMSE_R2_values

Out[219]:

	poly_test	poly_train	xgb_train	xgb_test	lgb_train	lgb_test	ada_train	ada_test
stator_yoke RMSE	0.296172	0.297187	0.060909	0.003707	0.026721	0.027337	0.147409	0.147518
R2	0.911502	0.910940	0.917367	0.917386	0.985995	0.985516	0.832185	0.831974
pm RMSE	NaN	NaN	0.400252	0.160081	0.164518	0.167241	0.557947	0.558125

3 rows × 22 columns

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In []:	
In []:	