MLPRegressor_v1_stage_4

November 25, 2022

1 MLPRegressor

```
import numpy as np import pandas as pd

import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split, RandomizedSearchCV from sklearn.preprocessing import StandardScaler from sklearn.pipeline import Pipeline from sklearn.neural_network import MLPRegressor from sklearn.feature_selection import SelectFromModel from sklearn.metrics import r2_score, mean_absolute_percentage_error,u chean_absolute_error, mean_squared_error from statsmodels.tools.eval_measures import stde
```

1.1 Read the etl info results

1.2 Read the dataset

```
[]: df = pd.read_csv('../../dataset_clean/PlatteRiverWeir_features_v1_clean.csv')
df

[]: SensorTime CaptureTime Stage Discharge grayMean \
0 2012-06-09 13:15:00 2012-06-09T13:09:07 2.99 916.0 97.405096
```

```
2012-06-09 13:15:00 2012-06-09T13:10:29
                                                                        104.066757
     2
                                                       2.96
            2012-06-09 13:45:00 2012-06-09T13:44:01
                                                                 873.0
                                                                        105.636831
     3
            2012-06-09 14:45:00
                                 2012-06-09T14:44:30
                                                       2.94
                                                                 846.0
                                                                         104.418949
     4
            2012-06-09 15:45:00
                                 2012-06-09T15:44:59
                                                       2.94
                                                                 846.0
                                                                        106.763541
           2019-10-11 09:00:00
     42054
                                 2019-10-11T08:59:53
                                                       2.54
                                                                 434.0
                                                                         82.872720
     42055
           2019-10-11 10:00:00
                                                       2.54
                                                                 434.0
                                 2019-10-11T09:59:52
                                                                         89.028383
     42056
           2019-10-11 11:00:00 2019-10-11T10:59:52
                                                       2.54
                                                                 434.0
                                                                         94.722097
     42057
            2019-10-11 12:00:00 2019-10-11T11:59:53
                                                       2.54
                                                                 434.0
                                                                         96.693270
     42058
           2019-10-11 12:45:00 2019-10-11T12:59:52
                                                                 434.0
                                                                         98.738399
                                                       2.54
                            hMean
                                      hSigma
            graySigma
     0
            39.623303
                       105.368375
                                   41.572939
     1
            40.179745
                       112.399458
                                   41.795584
     2
            40.533218
                       114.021526
                                   42.145582
     3
            41.752678
                       112.612830
                                   43.575351
     4
            44.442097
                       114.839424
                                   46.302008
     42054
           57.702652
                        87.260572
                                   61.485334
     42055
           55.840861
                        94.175906
                                   59.006132
     42056
           54.355753
                       100.534577
                                   56.921028
                                   55.083532
     42057
                       102.891159
            52.787629
     42058 52.025453
                       105.292067
                                   53.994155
     [42059 rows x 8 columns]
[]: df['SensorTime'] = pd.to_datetime(df['SensorTime'])
     df['Year'] = df['SensorTime'].dt.year
[]: df.dtypes
[]: SensorTime
                    datetime64[ns]
     CaptureTime
                            object
     Stage
                           float64
     Discharge
                           float64
     grayMean
                           float64
     graySigma
                           float64
    hMean
                           float64
    hSigma
                           float64
     Year
                             int64
     dtype: object
[]: df = df[(df.Stage > 0) & (df.Discharge > 0)]
```

2.99

916.0

1

```
1.3 Divide dataset to X and Y
[]: np.random.seed(0)
    df_train = df[(df.Year >= 2012) & (df.Year <= 2017)]</pre>
    df_train = df_train.iloc[np.random.permutation(len(df_train))]
    df_test = df[(df.Year >= 2018) & (df.Year <= 2019)]</pre>
[]: df_train = df_train.drop(columns=["Year", "SensorTime", "CaptureTime"])
    df_test = df_test.drop(columns=["Year", "SensorTime", "CaptureTime"])
[]: y_train = df_train["Stage"]
    X_train = df_train.drop(columns=["Stage", "Discharge"])
    y test = df test["Stage"]
    X_test = df_test.drop(columns=["Stage", "Discharge"])
\rightarrow random_state=0)
    1.4 Train model
[]: pipeline = Pipeline([
        ('scaler', StandardScaler()),
        ('clf', MLPRegressor(shuffle=False, max_iter=2000))
```

```
])
#param_grid = {'clf_hidden_layer_sizes': [(10), (10, 20), (10, 5, 15), (20, ___
\rightarrow 30, 10, 15)], 'clf__alpha': np.arange(1e-3, 1, 0.001),
→ 'clf_ learning_rate_init': np.arange(1e-3, 0.1, 0.001), 'clf_ activation':
\hookrightarrow ['tanh', 'relu']}
param_grid = {'clf_hidden_layer_sizes': [(256, 256, 128, 128, 64), (512, 256), __
→(128, 64, 64, 32), (512, 256, 128, 128)], 'clf__alpha': np.arange(1e-3, 0.1, u
→0.001), 'clf_activation': ['tanh', 'relu']}
clf = RandomizedSearchCV(pipeline, param_distributions=param_grid, n_iter=10,_
```

```
[]: clf.fit(X_train, y_train)
```

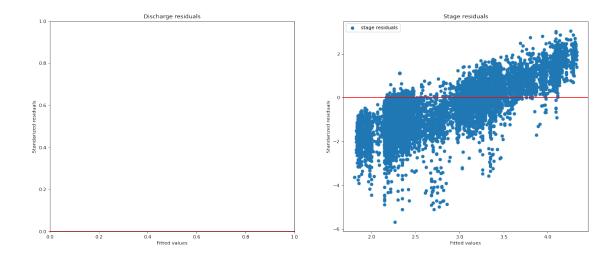
```
Fitting 5 folds for each of 10 candidates, totalling 50 fits
[CV 3/5] END clf_activation=tanh, clf_alpha=0.095,
clf hidden layer sizes=(128, 64, 64, 32);, score=-0.541 total time=
[CV 2/5] END clf_activation=tanh, clf_alpha=0.095,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.522 total time=
```

```
[CV 1/5] END clf_activation=tanh, clf_alpha=0.095,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.524 total time=
[CV 4/5] END clf_activation=tanh, clf_alpha=0.095,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.544 total time=
                                                                35.3s
[CV 5/5] END clf activation=tanh, clf alpha=0.095,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.537 total time=
[CV 4/5] END clf activation=tanh, clf alpha=0.019000000000000003,
clf_hidden_layer_sizes=(256, 256, 128, 128, 64);, score=-0.508 total time=
2.9min
[CV 3/5] END clf__activation=tanh, clf__alpha=0.019000000000000003,
clf_hidden_layer_sizes=(256, 256, 128, 128, 64);, score=-0.509 total time=
3.9min
[CV 3/5] END clf_activation=tanh, clf_alpha=0.064,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.539 total time= 2.7min
[CV 2/5] END clf__activation=tanh, clf__alpha=0.019000000000000003,
clf_hidden_layer_sizes=(256, 256, 128, 128, 64);, score=-0.487 total time=
4.3min
[CV 5/5] END clf_activation=tanh, clf_alpha=0.019000000000000003,
clf_hidden_layer_sizes=(256, 256, 128, 128, 64);, score=-0.514 total time=
4.6min
[CV 1/5] END clf activation=tanh, clf alpha=0.064,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.523 total time= 4.0min
[CV 1/5] END clf__activation=tanh, clf__alpha=0.019000000000000003,
clf_hidden_layer_sizes=(256, 256, 128, 128, 64);, score=-0.490 total time=
5.7min
clf_hidden_layer_sizes=(512, 256);, score=-0.562 total time= 2.4min
clf_hidden layer_sizes=(512, 256);, score=-0.594 total time= 2.2min
clf_hidden_layer_sizes=(512, 256);, score=-0.569 total time= 2.7min
[CV 2/5] END clf_activation=tanh, clf_alpha=0.064,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.510 total time= 6.5min
[CV 1/5] END clf_activation=tanh, clf_alpha=0.005,
clf hidden layer sizes=(128, 64, 64, 32);, score=-0.482 total time= 60.0s
[CV 2/5] END clf_activation=tanh, clf_alpha=0.005,
clf hidden layer sizes=(128, 64, 64, 32);, score=-0.485 total time= 1.1min
[CV 5/5] END clf__activation=tanh, clf__alpha=0.0260000000000000000,
clf_hidden_layer_sizes=(512, 256);, score=-0.581 total time= 2.4min
[CV 1/5] END clf__activation=tanh, clf__alpha=0.0260000000000000000,
clf_hidden_layer_sizes=(512, 256);, score=-0.557 total time= 4.3min
[CV 3/5] END clf_activation=tanh, clf_alpha=0.005,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.495 total time= 1.2min
[CV 4/5] END clf_activation=tanh, clf_alpha=0.005,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.510 total time= 51.4s
[CV 5/5] END clf_activation=tanh, clf_alpha=0.005,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.496 total time= 1.3min
[CV 4/5] END clf_activation=tanh, clf_alpha=0.064,
```

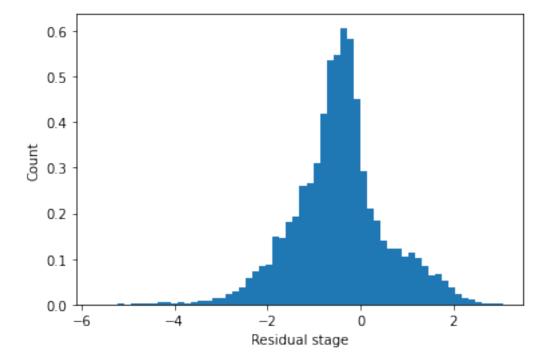
```
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.535 total time= 6.9min
[CV 5/5] END clf_activation=tanh, clf_alpha=0.064,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.536 total time= 6.4min
clf_hidden_layer_sizes=(512, 256);, score=-0.582 total time= 2.7min
[CV 2/5] END clf_activation=tanh, clf_alpha=0.035,
clf_hidden_layer_sizes=(512, 256);, score=-0.568 total time= 2.8min
[CV 3/5] END clf_activation=tanh, clf_alpha=0.035,
clf_hidden_layer_sizes=(512, 256);, score=-0.576 total time= 2.6min
[CV 4/5] END clf_activation=tanh, clf_alpha=0.035,
clf_hidden layer_sizes=(512, 256);, score=-0.594 total time= 2.4min
clf_hidden_layer_sizes=(512, 256);, score=-0.524 total time= 6.8min
[CV 1/5] END clf_activation=tanh, clf_alpha=0.035,
clf_hidden_layer_sizes=(512, 256);, score=-0.546 total time= 6.0min
[CV 5/5] END clf_activation=tanh, clf_alpha=0.035,
clf_hidden_layer_sizes=(512, 256);, score=-0.582 total time= 2.6min
clf_hidden_layer_sizes=(512, 256);, score=-0.506 total time= 7.3min
clf_hidden_layer_sizes=(512, 256);, score=-0.511 total time= 7.4min
clf_hidden_layer_sizes=(512, 256);, score=-0.536 total time= 6.9min
[CV 2/5] END clf_activation=relu, clf_alpha=0.079,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.527 total time=
[CV 1/5] END clf_activation=relu, clf_alpha=0.079,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.495 total time=
                                                             42.8s
[CV 4/5] END clf_activation=relu, clf_alpha=0.079,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.514 total time=
[CV 3/5] END clf_activation=relu, clf_alpha=0.079,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.500 total time=
                                                             59.3s
[CV 5/5] END clf_activation=relu, clf_alpha=0.079,
clf_hidden_layer_sizes=(128, 64, 64, 32);, score=-0.524 total time= 25.0s
[CV 2/5] END clf_activation=relu, clf_alpha=0.098,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.511 total time= 3.3min
[CV 1/5] END clf_activation=relu, clf_alpha=0.098,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.501 total time= 5.7min
[CV 4/5] END clf_activation=relu, clf_alpha=0.098,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.520 total time= 4.7min
[CV 4/5] END clf_activation=tanh, clf_alpha=0.068,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.545 total time= 2.8min
[CV 5/5] END clf_activation=relu, clf_alpha=0.098,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.521 total time= 4.9min
[CV 3/5] END clf_activation=relu, clf_alpha=0.098,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.511 total time= 5.4min
[CV 1/5] END clf_activation=tanh, clf_alpha=0.068,
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.521 total time= 5.0min
[CV 3/5] END clf_activation=tanh, clf_alpha=0.068,
```

```
clf_hidden_layer_sizes=(512, 256, 128, 128);, score=-0.524 total time= 5.4min
    [CV 2/5] END clf_activation=tanh, clf_alpha=0.068,
    clf_hidden layer_sizes=(512, 256, 128, 128);, score=-0.514 total time= 5.7min
    [CV 5/5] END clf_activation=tanh, clf_alpha=0.068,
    clf hidden layer sizes=(512, 256, 128, 128);, score=-0.536 total time= 4.6min
[]: RandomizedSearchCV(estimator=Pipeline(steps=[('scaler', StandardScaler()),
                                                  ('clf',
                                                   MLPRegressor(max_iter=2000,
                                                                shuffle=False))]),
                        n_jobs=8,
                        param_distributions={'clf__activation': ['tanh', 'relu'],
                                             'clf_alpha': array([0.001, 0.002,
    0.003, 0.004, 0.005, 0.006, 0.007, 0.008, 0.009,
           0.01, 0.011, 0.012, 0.013, 0.014, 0.015, 0.016, 0.017, 0.018,
           0.019, 0.02, 0.021, 0.022, 0.023, 0.024...
           0.064, 0.065, 0.066, 0.067, 0.068, 0.069, 0.07, 0.071, 0.072,
           0.073, 0.074, 0.075, 0.076, 0.077, 0.078, 0.079, 0.08, 0.081,
           0.082, 0.083, 0.084, 0.085, 0.086, 0.087, 0.088, 0.089, 0.09,
           0.091, 0.092, 0.093, 0.094, 0.095, 0.096, 0.097, 0.098, 0.099]),
                                             'clf_hidden_layer_sizes': [(256, 256,
                                                                          128, 128,
                                                                          64),
                                                                         (512, 256),
                                                                         (128, 64,
                                                                          64, 32),
                                                                         (512, 256,
                                                                          128,
                                                                          128)]},
                        scoring='neg mean squared error', verbose=3)
[]: clf.best_score_
[]: -0.4935050421898449
[]: clf.best_params_
[]: {'clf_hidden_layer_sizes': (128, 64, 64, 32),
      'clf__alpha': 0.005,
      'clf__activation': 'tanh'}
    1.5 Test model
[]: clf.score(X_test, y_test)
[]: -0.3542243112154066
```

```
[]: y_pred = clf.predict(X_test)
[]: print("R^2: ", r2 score(y test, y pred))
     print("mse: ", mean_squared_error(y_test, y_pred))
     print("rmse: ", mean_squared_error(y_test, y_pred, squared=False))
     print("mae: ", mean_absolute_error(y_test, y_pred))
     print("mape: ", mean_absolute_percentage_error(y_test, y_pred))
     print("Error estandar: ", stde(y_test.squeeze(),
           y_pred.squeeze(), ddof=2))
    R^2: 0.09298597613232973
    mse: 0.3542243112154066
    rmse: 0.5951674648495217
    mae: 0.46038871992222397
    mape: 0.1806500295655998
    Error estandar: 0.5416763198444101
[]: residuals = y_test - y_pred
     residuals_std = residuals/residuals.std()
     y_real_stage = y_test
     residual_stage = residuals
     #y_real_discharge = np.array([i[-1] for i in y_test])
     \#residual \ discharge = np.array([i[-1] \ for \ i \ in \ residuals])
     figure, ax = plt.subplots(ncols=2, figsize=(20, 8), dpi=80)
     ax[1].scatter(y_real_stage, residual_stage / residual_stage.std(), label="stage_u
     →residuals")
     #ax[0].scatter(y_real_discharge, residual_discharge / residual_discharge.std(), ا
     → label="discharge residuals")
     ax[1].axhline(y=0.0, color='r', linestyle='-')
     ax[0].axhline(y=0.0, color='r', linestyle='-')
     ax[1].set_title("Stage residuals")
     ax[0].set_title("Discharge residuals")
     ax[1].set_xlabel("Fitted values")
     ax[0].set_xlabel("Fitted values")
     ax[1].set ylabel("Standarized residuals")
     ax[0].set_ylabel("Standarized residuals")
     plt.legend()
     plt.show()
```



```
[]: plt.hist(residual_stage / residual_stage.std(), density=True, bins = 60)
plt.ylabel('Count')
plt.xlabel('Residual stage');
plt.show()
```

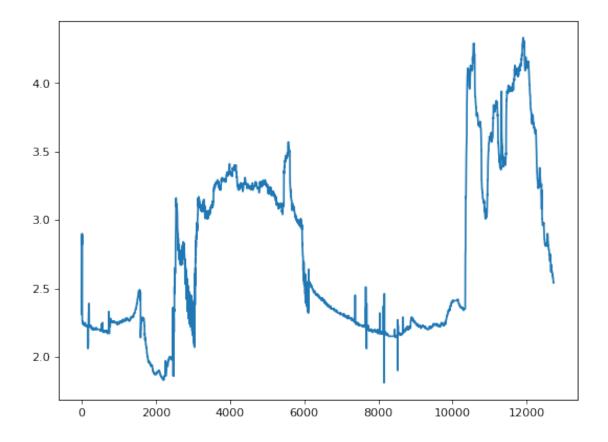


```
plt.xlabel('Residual discharge');
plt.show()"""
```

[]: "plt.hist(residual_discharge / residual_discharge.std(), density=True, bins =
60)\nplt.ylabel('Count')\nplt.xlabel('Residual discharge');\nplt.show()"

```
[]: plt.figure(figsize=(8, 6), dpi=80)
plt.plot(np.arange(len(y_test)), y_test, label="Stage real")
```

[]: [<matplotlib.lines.Line2D at 0x7fc7e4dce2c0>]



```
[]: figure, ax = plt.subplots(ncols=2, figsize=(20, 8), dpi=80)

ax[0].plot(np.arange(len(y_test)), y_test, label="Stage real")
ax[0].plot(np.arange(len(y_test)), y_pred, label="Stage pred")

ax[0].set_title("Stage predictions")
ax[1].set_title("Discharge predictions")

ax[1].set_ylabel("Values")
ax[0].set_ylabel("Values")
```

```
ax[1].set_xlabel("Time")
ax[0].set_xlabel("Time")

ax[0].legend()
ax[1].legend()
plt.show()
```

No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.

