knn regression v1

October 21, 2022

1 KNN regression

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 \
```

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

```
1
            2012-06-09 13:15:00
                                 2012-06-09T13:10:29
                                                       2.99
                                                                 916.0
                                                                        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
                                                       2.54
     42055
           2019-10-11 10:00:00
                                                                 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
            2019-10-11 12:00:00 2019-10-11T11:59:53
     42057
                                                       2.54
                                                                 434.0
                                                                         96.693270
     42058
           2019-10-11 12:45:00 2019-10-11T12:59:52
                                                       2.54
                                                                 434.0
                                                                         98.738399
                       entropyMean
                                    entropySigma
            graySigma
                                                       hMean
                                                                 hSigma \
     0
            39.623303
                          0.203417
                                        0.979825
                                                  105.368375
                                                              41.572939
                          0.206835
     1
            40.179745
                                        1.002624
                                                  112.399458
                                                              41.795584
     2
            40.533218
                          0.204756
                                        0.994246
                                                  114.021526
                                                              42.145582
     3
            41.752678
                          0.202428
                                        0.983170
                                                  112.612830
                                                              43.575351
     4
            44.442097
                          0.202661
                                        0.989625
                                                  114.839424
                                                              46.302008
     42054
            57.702652
                          0.221708
                                        1.076393
                                                   87.260572 61.485334
     42055
            55.840861
                                                   94.175906
                                                              59.006132
                          0.233168
                                        1.124774
     42056
           54.355753
                          0.240722
                                        1.151833
                                                  100.534577
                                                              56.921028
     42057
                          0.244789
                                                  102.891159
            52.787629
                                        1.171987
                                                              55.083532
     42058
           52.025453
                          0.252812
                                        1.213278
                                                  105.292067
                                                              53.994155
                          sSigma
                                                 vSigma
                 sMean
                                       vMean
     0
            124.520218 4.111846 132.405971
                                              14.983367
                                  133.070221
     1
            124.317679 4.270429
                                              15.334166
     2
                        4.310293 133.294541
            124.304621
                                              15.502448
                                  133.458381
     3
            124.369736
                        4.120586
                                              15.190064
     4
            124.283191 4.088480
                                  133.573595
                                              14.801143
     42054
           127.807813
                        2.564157
                                  124.073149
                                              13.757842
     42055
            127.336000
                        2.585121
                                  124.882812
                                              13.234735
     42056
            126.958768
                        2.774867
                                  126.145409
                                              13.408480
     42057
                        2.998683
                                  127.508063
            126.679956
                                              13.863205
     42058
            126.328075
                        3.258103
                                  128.788256
                                              14.353808
     [42059 rows x 14 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
                        float64
grayMean
graySigma
                        float64
entropyMean
                        float64
entropySigma
                        float64
hMean
                        float64
hSigma
                        float64
sMean
                        float64
                        float64
sSigma
vMean
                        float64
                        float64
vSigma
Year
                          int64
dtype: object
```

1.3 Divide dataset to X and Y

```
[]: df_train = df[(df.Year >= 2012) & (df.Year <= 2017)]
df_test = df[(df.Year >= 2018) & (df.Year <= 2019)]

[]: df_train = df_train.drop(columns=["Year", "SensorTime", "CaptureTime"])
df_test = df_test.drop(columns=["Year", "SensorTime", "CaptureTime"])

[]: y_train = df_train[["Stage", "Discharge"]]
X_train = df_train.drop(columns=["Stage", "Discharge"])
y_test = df_test[["Stage", "Discharge"]]
X_test = df_test.drop(columns=["Stage", "Discharge"])

[]: #X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, □ → random_state=0)
```

1.4 Train model

[]: clf.fit(X_train, y_train)

```
Fitting 5 folds for each of 20 candidates, totalling 100 fits
[CV 2/5] END clf__leaf_size=15, clf__n_neighbors=20;, score=-262982.044 total
time=
      0.4s
[CV 5/5] END clf__leaf_size=15, clf__n_neighbors=20;, score=-254669.040 total
      0.4s
[CV 1/5] END clf__leaf_size=15, clf__n_neighbors=20;, score=-287950.474 total
time=
      0.4s
[CV 2/5] END clf_leaf_size=10, clf_n neighbors=15;, score=-274019.616 total
       0.4s
[CV 3/5] END clf_leaf_size=10, clf_n neighbors=15;, score=-631247.234 total
time=
      0.4s
[CV 3/5] END clf_leaf_size=15, clf_n neighbors=20;, score=-626108.878 total
       0.4s
[CV 5/5] END clf_leaf_size=10, clf_n neighbors=15;, score=-265557.194 total
      0.4s
[CV 1/5] END clf__leaf_size=10, clf__n_neighbors=15;, score=-298416.628 total
      0.5s
[CV 4/5] END clf leaf size=15, clf n neighbors=20;, score=-1583622.818 total
time=
      0.5s
[CV 4/5] END clf__leaf_size=10, clf__n_neighbors=15;, score=-1564682.906 total
time=
      0.5s
[CV 2/5] END clf_leaf_size=10, clf__n_neighbors=5;, score=-344968.835 total
       0.3s[CV 2/5] END clf_leaf_size=30, clf_n_neighbors=5;,
score=-344968.835 total time= 0.2s
[CV 5/5] END clf_leaf_size=30, clf_n_neighbors=5;, score=-340609.021 total
       0.2s
[CV 3/5] END clf__leaf_size=30, clf__n_neighbors=5;, score=-675004.507 total
time=
      0.2s
[CV 1/5] END clf_leaf_size=30, clf_n_neighbors=5;, score=-355831.746 total
[CV 1/5] END clf__leaf_size=10, clf__n_neighbors=5;, score=-355831.746 total
      0.3s
[CV 3/5] END clf__leaf_size=10, clf__n_neighbors=5;, score=-675004.507 total
time=
      0.3s
[CV 5/5] END clf__leaf_size=10, clf__n_neighbors=5;, score=-340609.021 total
time=
      0.3s
[CV 4/5] END clf_leaf_size=30, clf_n neighbors=5;, score=-1544870.773 total
      0.3s
[CV 4/5] END clf_leaf_size=10, clf_n neighbors=5;, score=-1544870.773 total
time=
       0.4s
[CV 3/5] END clf_leaf_size=60, clf_n neighbors=15;, score=-631247.234 total
```

```
time=
       0.3s
[CV 5/5] END clf__leaf_size=60, clf__n_neighbors=15;, score=-265557.194 total
       0.3s
[CV 2/5] END clf__leaf_size=60, clf__n_neighbors=15;, score=-274019.616 total
       0.3s
[CV 4/5] END clf__leaf_size=60, clf__n_neighbors=15;, score=-1564682.906 total
       0.3s
[CV 1/5] END clf__leaf_size=60, clf__n_neighbors=15;, score=-298416.628 total
       0.3s
[CV 2/5] END clf__leaf_size=20, clf__n_neighbors=20;, score=-262982.044 total
time=
       0.4s
[CV 3/5] END clf_leaf_size=20, clf_n neighbors=20;, score=-626108.878 total
      0.3s
[CV 1/5] END clf_leaf_size=20, clf_n neighbors=20;, score=-287950.474 total
       0.4s
[CV 5/5] END clf_leaf_size=20, clf_n neighbors=20;, score=-254669.040 total
time=
       0.4s
[CV 4/5] END clf_leaf_size=20, clf_n_neighbors=20;, score=-1583622.818 total
       0.5s
[CV 2/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-295919.184 total
       0.4s
[CV 3/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-642926.993 total
       0.4s
[CV 1/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-312122.763 total
time=
       0.4s
[CV 5/5] END clf_leaf_size=10, clf_n neighbors=10;, score=-289872.083 total
time=
      0.4s
[CV 4/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-1549846.239 total
       0.4s
[CV 2/5] END clf__leaf_size=45, clf__n_neighbors=40;, score=-239721.761 total
time=
       0.4s
[CV 1/5] END clf__leaf_size=45, clf__n_neighbors=40;, score=-273649.682 total
time=
       0.5s
[CV 3/5] END clf__leaf_size=45, clf__n_neighbors=40;, score=-624656.954 total
       0.4s
[CV 2/5] END clf_leaf_size=20, clf__n_neighbors=5;, score=-344968.835 total
       0.2s
[CV 4/5] END clf__leaf_size=45, clf__n_neighbors=40;, score=-1622677.623 total
time=
      0.5s
[CV 1/5] END clf__leaf_size=20, clf__n_neighbors=5;, score=-355831.746 total
time=
       0.3s
[CV 3/5] END clf_leaf_size=20, clf_n_neighbors=5;, score=-675004.507 total
[CV 5/5] END clf_leaf size=20, clf_neighbors=5;, score=-340609.021 total
time=
       0.3s
[CV 4/5] END clf_leaf_size=20, clf_n neighbors=5;, score=-1544870.773 total
time=
       0.3s
[CV 5/5] END clf_leaf_size=45, clf_n neighbors=40;, score=-240611.004 total
```

```
time=
       0.5s
[CV 3/5] END clf__leaf_size=45, clf__n_neighbors=20;, score=-626108.878 total
       0.3s
[CV 2/5] END clf__leaf_size=45, clf__n_neighbors=20;, score=-262982.044 total
       0.4s
[CV 1/5] END clf__leaf_size=45, clf__n_neighbors=20;, score=-287950.474 total
       0.4s
[CV 1/5] END clf__leaf_size=20, clf__n_neighbors=60;, score=-267550.136 total
       0.6s
[CV 5/5] END clf__leaf_size=45, clf__n_neighbors=20;, score=-254669.040 total
time=
       0.4s
[CV 4/5] END clf_leaf_size=45, clf_n_neighbors=20;, score=-1583622.818 total
       0.4s
[CV 3/5] END clf_leaf_size=20, clf_n neighbors=60;, score=-631250.151 total
       0.6s
[CV 5/5] END clf_leaf_size=20, clf_n neighbors=60;, score=-235553.162 total
time=
       0.6s
[CV 2/5] END clf_leaf_size=20, clf_n neighbors=60;, score=-227557.348 total
       0.7s
[CV 4/5] END clf__leaf_size=20, clf__n_neighbors=60;, score=-1645814.280 total
       0.6s
[CV 1/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-312122.763 total
       0.3s
[CV 2/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-295919.184 total
time=
       0.3s
[CV 3/5] END clf_leaf_size=30, clf_n neighbors=10;, score=-642926.993 total
      0.3s
[CV 5/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-289872.083 total
       0.3s
[CV 4/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-1549846.239 total
time=
       0.3s
[CV 3/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-626108.878 total
time=
       0.3s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-262982.044 total
      0.3s
[CV 1/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-287950.474 total
      0.4s
[CV 4/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-1583622.818 total
time=
      0.3s
[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-254669.040 total
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       0.3s
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[CV 2/5] END clf_leaf_size=60, clf_n neighbors=10;, score=-295919.184 total
time=
       0.3s
[CV 1/5] END clf_leaf_size=60, clf_n neighbors=10;, score=-312122.763 total
time=
       0.3s
[CV 5/5] END clf_leaf_size=60, clf_n neighbors=10;, score=-289872.083 total
```

```
time=
       0.3s
[CV 4/5] END clf__leaf_size=60, clf__n_neighbors=10;, score=-1549846.239 total
       0.3s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-274019.616 total
       0.3s
[CV 1/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-298416.628 total
       0.3s
[CV 3/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-631247.234 total
time=
       0.3s
[CV 4/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-1564682.906 total
time=
       0.3s
[CV 5/5] END clf_leaf_size=50, clf_n neighbors=15;, score=-265557.194 total
       0.3s
[CV 2/5] END clf_leaf_size=10, clf_n neighbors=60;, score=-227557.348 total
       0.7s
[CV 3/5] END clf_leaf_size=10, clf_n neighbors=60;, score=-631250.151 total
time=
       0.7s
[CV 5/5] END clf_leaf_size=10, clf_n neighbors=60;, score=-235553.162 total
       0.7s
[CV 1/5] END clf__leaf_size=10, clf__n_neighbors=60;, score=-267550.136 total
       0.8s
[CV 1/5] END clf__leaf_size=20, clf__n_neighbors=40;, score=-273649.682 total
       0.5s
[CV 4/5] END clf__leaf_size=10, clf__n_neighbors=60;, score=-1645814.280 total
time=
      0.8s
[CV 2/5] END clf_leaf_size=20, clf_n neighbors=40;, score=-239721.761 total
time=
      0.5s
[CV 3/5] END clf__leaf_size=20, clf__n_neighbors=40;, score=-624656.954 total
       0.5s
[CV 5/5] END clf__leaf_size=20, clf__n_neighbors=40;, score=-240611.004 total
time=
      0.5s
[CV 4/5] END clf__leaf_size=20, clf__n_neighbors=40;, score=-1622677.623 total
time=
       0.5s
[CV 2/5] END clf__leaf_size=15, clf__n_neighbors=60;, score=-227557.348 total
time=
      0.6s
[CV 1/5] END clf__leaf_size=15, clf__n_neighbors=60;, score=-267550.136 total
      0.6s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=40;, score=-239721.761 total
time=
      0.4s
[CV 1/5] END clf__leaf_size=50, clf__n_neighbors=40;, score=-273649.682 total
time=
       0.5s
[CV 3/5] END clf_leaf_size=15, clf_n neighbors=60;, score=-631250.151 total
[CV 3/5] END clf_leaf_size=50, clf_n neighbors=40;, score=-624656.954 total
time=
       0.4s
[CV 4/5] END clf_leaf_size=15, clf_n_neighbors=60;, score=-1645814.280 total
time=
       0.6s
[CV 5/5] END clf_leaf_size=15, clf_n neighbors=60;, score=-235553.162 total
```

```
0.6s
    time=
    [CV 5/5] END clf_leaf_size=50, clf_n_neighbors=40;, score=-240611.004 total
    time=
           0.4s
    [CV 4/5] END clf__leaf_size=50, clf__n_neighbors=40;, score=-1622677.623 total
           0.5s
    [CV 5/5] END clf__leaf_size=30, clf__n_neighbors=40;, score=-240611.004 total
    time=
           0.3s
    [CV 2/5] END clf__leaf_size=30, clf__n_neighbors=40;, score=-239721.761 total
           0.4s
    [CV 4/5] END clf__leaf_size=30, clf__n_neighbors=40;, score=-1622677.623 total
    time=
           0.3s
    [CV 3/5] END clf_leaf_size=30, clf_n neighbors=40;, score=-624656.954 total
          0.4s
    [CV 1/5] END clf_leaf_size=30, clf_n neighbors=40;, score=-273649.682 total
    time=
[]: RandomizedSearchCV(estimator=Pipeline(steps=[('scaler', StandardScaler()),
                                                 ('clf', KNeighborsRegressor())]),
                       n_iter=20, n_jobs=10,
                       param_distributions={'clf__leaf_size': [10, 15, 20, 30, 45,
                                                               50, 60],
                                            'clf_n_neighbors': [5, 10, 15, 20, 40,
                                                                 60]},
                       scoring='neg_mean_squared_error', verbose=3)
[]: clf.best_score_
[]: -600263.4049000422
[]: clf.best_params_
[]: {'clf_n_neighbors': 40, 'clf_leaf_size': 45}
         Test model
    1.5
[]: clf.score(X_test, y_test)
[]: -326231.6305243665
[ ]: 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=len(X_train.columns) + 1))

R^2: 0.03934311007092667 mse: 326231.6305243665 rmse: 404.15162808096187 mae: 264.95765391002413

mape: 2.1191357746924916e+16

Error estandar: [4.95443528e-01 7.49123513e+02]

[]: