knn_regression_v1_seg_3

November 24, 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 \
```

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
       2012-06-09 13:45:00
                             2012-06-09T13:44:01
                                                     2.96
                                                               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
42054
       2019-10-11 09:00:00
                             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
       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
                                                               434.0
                                                                        98.738399
                                                     2.54
                        hMean
                                  hSigma
                                                            hMean0
                                                                    entropyMean1
       graySigma
                                            grayMean0
0
       39.623303
                   105.368375
                               41.572939
                                            97.084576
                                                        106.047217
                                                                         0.092532
1
       40.179745
                   112.399458
                               41.795584
                                           105.668610
                                                        114.886049
                                                                         0.090279
2
                               42.145582
       40.533218
                   114.021526
                                           106.786307
                                                        116.053131
                                                                         0.090561
3
       41.752678
                   112.612830
                               43.575351
                                           107.674299
                                                        117.005027
                                                                         0.095616
4
       44.442097
                   114.839424
                               46.302008
                                           114.858589
                                                        124.519271
                                                                         0.101601
42054
       57.702652
                    87.260572
                               61.485334
                                            43.737485
                                                         46.616662
                                                                         0.120668
                                            46.268458
42055
       55.840861
                    94.175906
                               59.006132
                                                         49.716207
                                                                         0.113951
42056
       54.355753
                   100.534577
                               56.921028
                                            49.841325
                                                         53.984763
                                                                         0.110346
                                            53.912185
42057
       52.787629
                   102.891159
                               55.083532
                                                         58.857575
                                                                         0.112571
42058
       52.025453
                   105.292067
                               53.994155
                                            59.611803
                                                         65.697745
                                                                         0.110247
       entropySigma1
                                   WwRawLineMean
                                                   WwRawLineSigma
                           hMean1
0
            0.632319
                       169.963345
                                         0.000000
                                                          0.00000
1
            0.620077
                       175.220945
                                         0.000000
                                                          0.000000
2
            0.620853
                       179.554842
                                         0.00000
                                                          0.00000
3
            0.651642
                       180.921521
                                         0.00000
                                                          0.00000
4
                                                          0.00000
            0.688024
                       183.131779
                                         0.00000
42054
            0.824195
                       126.181417
                                     38385.370066
                                                      15952.029728
42055
            0.783437
                       131.754200
                                     40162.989292
                                                      15467.708856
42056
            0.766074
                       138.014068
                                     42095.946590
                                                      16770.357949
42057
            0.777376
                       146.470365
                                     45345.490954
                                                      17498.432849
42058
            0.760248
                       156.957374
                                     47877.870782
                                                      19963.166359
       WwCurveLineMean
                         WwCurveLineSigma
0
              0.00000
                                 0.000000
1
              0.00000
                                  0.000000
2
              0.000000
                                  0.000000
3
              0.000000
                                  0.000000
4
                                  0.00000
              0.000000
                             16444.401209
42054
          37550.894823
          39397.339095
                             16009.008049
42055
42056
          41350.006568
                             17489.374617
```

```
42057
                                  18268.294896
               44553.920296
     42058
               47280.270559
                                  20559.358767
     [42059 rows x 17 columns]
[]: df['SensorTime'] = pd.to_datetime(df['SensorTime'])
     df['Year'] = df['SensorTime'].dt.year
[]: df.dtypes
[]: SensorTime
                         datetime64[ns]
     CaptureTime
                                  object
                                 float64
     Stage
     Discharge
                                 float64
     grayMean
                                 float64
     graySigma
                                float64
                                 float64
    hMean
     hSigma
                                float64
     grayMean0
                                 float64
     hMean0
                                 float64
     entropyMean1
                                float64
     entropySigma1
                                 float64
     hMean1
                                 float64
     WwRawLineMean
                                 float64
     WwRawLineSigma
                                 float64
     WwCurveLineMean
                                 float64
     WwCurveLineSigma
                                 float64
     Year
                                   int64
     dtype: object
[]: df = df[(df.Stage > 0) & (df.Discharge > 0)]
[]: df.isna().sum()
[]: SensorTime
                         0
     CaptureTime
                         0
                         0
     Stage
                         0
     Discharge
     grayMean
                         0
                         0
     graySigma
                         0
     hMean
     hSigma
                         0
                         0
     grayMean0
    hMean0
                         0
     entropyMean1
                         0
     entropySigma1
                         0
    hMean1
                         0
```

```
WwRawLineMean 0
WwRawLineSigma 0
WwCurveLineMean 0
WwCurveLineSigma 0
Year 0
dtype: int64
```

1.3 Divide dataset to X and Y

```
[]: np.random.seed(0)

df_train = df[(df.Year >= 2012) & (df.Year <= 2017)]
    df_train = df_train.iloc[np.random.permutation(len(df_train))]

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"]]
    y_train = df_train.drop(columns=["Stage", "Discharge"])
    y_test = df_test["Stage"]
    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, \( \) \(\) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \( \) \(
```

1.4 Train model

[]: clf.fit(X_train, y_train)

```
Fitting 5 folds for each of 30 candidates, totalling 150 fits
[CV 2/5] END clf_leaf_size=50, clf_n neighbors=5;, score=-0.177 total time=
0.3s
[CV 1/5] END clf_leaf_size=50, clf_n neighbors=5;, score=-0.162 total time=
0.3s
[CV 3/5] END clf_leaf_size=50, clf_n neighbors=5;, score=-0.171 total time=
0.3s
[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.169 total time=
0.3s
[CV 4/5] END clf_leaf_size=50, clf_n_neighbors=5;, score=-0.180 total time=
0.3s
[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.164 total time=
0.3s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.179 total time=
[CV 1/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.174 total time=
[CV 4/5] END clf_leaf_size=50, clf_n_neighbors=10;, score=-0.178 total time=
0.4s
[CV 3/5] END clf_leaf_size=50, clf_n neighbors=10;, score=-0.173 total time=
0.4s
[CV 1/5] END clf_leaf_size=30, clf_n neighbors=5;, score=-0.162 total time=
0.3s
[CV 2/5] END clf_leaf_size=30, clf_n neighbors=5;, score=-0.177 total time=
0.3s
[CV 3/5] END clf__leaf_size=30, clf__n_neighbors=5;, score=-0.171 total time=
[CV 5/5] END clf leaf size=30, clf n neighbors=5;, score=-0.164 total time=
0.3s
[CV 4/5] END clf_leaf_size=30, clf_n_neighbors=5;, score=-0.180 total time=
[CV 1/5] END clf_leaf_size=45, clf_n_neighbors=60;, score=-0.205 total time=
0.7s
[CV 3/5] END clf_leaf_size=45, clf__n_neighbors=60;, score=-0.220 total time=
0.7s
[CV 2/5] END clf_leaf_size=45, clf_n neighbors=60;, score=-0.224 total time=
0.7s
[CV 4/5] END clf__leaf_size=45, clf__n_neighbors=60;, score=-0.216 total time=
0.7s
[CV 5/5] END clf__leaf_size=45, clf__n_neighbors=60;, score=-0.211 total time=
0.8s
[CV 1/5] END clf__leaf_size=15, clf__n_neighbors=10;, score=-0.174 total time=
0.5s
[CV 2/5] END clf__leaf_size=15, clf__n_neighbors=10;, score=-0.179 total time=
0.5s
```

```
[CV 5/5] END clf__leaf_size=15, clf__n_neighbors=10;, score=-0.169 total time= 0.5s
```

- [CV 4/5] END clf__leaf_size=15, clf__n_neighbors=10;, score=-0.178 total time= 0.5s
- [CV 3/5] END clf__leaf_size=15, clf__n_neighbors=10;, score=-0.173 total time= 0.5s
- [CV 1/5] END clf_leaf_size=50, clf_n_neighbors=40;, score=-0.197 total time= 0.5s
- [CV 2/5] END clf__leaf_size=50, clf__n_neighbors=40;, score=-0.214 total time= 0.6s
- [CV 3/5] END clf_leaf_size=50, clf_n_neighbors=40;, score=-0.206 total time= 0.6s
- [CV 4/5] END clf__leaf_size=50, clf__n_neighbors=40;, score=-0.206 total time= 0.6s
- [CV 2/5] END clf_leaf_size=10, clf_n_neighbors=5;, score=-0.177 total time= 0.4s
- [CV 1/5] END clf_leaf_size=10, clf_n_neighbors=5;, score=-0.162 total time= 0.5s
- [CV 5/5] END clf_leaf_size=50, clf_n_neighbors=40;, score=-0.200 total time= 0.6s
- [CV 3/5] END clf_leaf_size=10, clf_n_neighbors=5;, score=-0.171 total time= 0.5s
- [CV 5/5] END clf_leaf_size=10, clf_n_neighbors=5;, score=-0.164 total time= 0.5s
- [CV 4/5] END clf_leaf_size=10, clf_n_neighbors=5;, score=-0.180 total time= 0.5s
- [CV 1/5] END clf_leaf_size=45, clf__n_neighbors=15;, score=-0.179 total time= 0.4s
- [CV 2/5] END clf__leaf_size=45, clf__n_neighbors=15;, score=-0.189 total time= 0.4s
- [CV 5/5] END clf__leaf_size=45, clf__n_neighbors=15;, score=-0.177 total time= 0.4s
- [CV 4/5] END clf__leaf_size=45, clf__n_neighbors=15;, score=-0.185 total time= 0.4s
- [CV 3/5] END clf__leaf_size=45, clf__n_neighbors=15;, score=-0.181 total time= 0.5s
- [CV 2/5] END clf_leaf_size=45, clf__n_neighbors=5;, score=-0.177 total time= 0.3s
- [CV 3/5] END clf_leaf_size=45, clf_n_neighbors=5;, score=-0.171 total time= 0.3s
- [CV 1/5] END clf_leaf_size=45, clf_n_neighbors=5;, score=-0.162 total time= 0.5s
- [CV 4/5] END clf_leaf_size=45, clf_n_neighbors=5;, score=-0.180 total time= 0.3s
- [CV 5/5] END clf_leaf_size=45, clf_n_neighbors=5;, score=-0.164 total time= 0.3s
- [CV 1/5] END clf_leaf_size=15, clf_n_neighbors=40;, score=-0.197 total time= 0.8s

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[CV 2/5] END clf leaf size=15, clf n neighbors=40;, score=-0.214 total time=
0.8s
[CV 3/5] END clf_leaf_size=15, clf_n neighbors=40;, score=-0.206 total time=
0.8s
[CV 4/5] END clf leaf size=15, clf n neighbors=40;, score=-0.206 total time=
0.8s
[CV 5/5] END clf leaf size=15, clf n neighbors=40;, score=-0.200 total time=
0.8s
[CV 1/5] END clf leaf size=50, clf n neighbors=20;, score=-0.185 total time=
0.4s
[CV 1/5] END clf_leaf_size=10, clf_n neighbors=10;, score=-0.174 total time=
0.6s
[CV 2/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-0.179 total time=
[CV 2/5] END clf_leaf_size=50, clf_n_neighbors=20;, score=-0.197 total time=
[CV 4/5] END clf_leaf_size=10, clf__n_neighbors=10;, score=-0.178 total time=
0.6s
[CV 3/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-0.187 total time=
0.4s
[CV 4/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-0.190 total time=
0.5s
[CV 5/5] END clf__leaf_size=10, clf__n_neighbors=10;, score=-0.169 total time=
0.6s
[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=20;, score=-0.184 total time=
0.4s
[CV 3/5] END clf_leaf_size=10, clf_n neighbors=10;, score=-0.173 total time=
0.7s
[CV 1/5] END clf_leaf_size=20, clf_n neighbors=10;, score=-0.174 total time=
[CV 1/5] END clf_leaf_size=30, clf_n neighbors=15;, score=-0.179 total time=
[CV 4/5] END clf_leaf_size=20, clf_n neighbors=10;, score=-0.178 total time=
0.5s
[CV 3/5] END clf leaf size=20, clf n neighbors=10;, score=-0.173 total time=
0.5s
[CV 2/5] END clf leaf size=30, clf n neighbors=15;, score=-0.189 total time=
0.5s
[CV 5/5] END clf__leaf_size=20, clf__n_neighbors=10;, score=-0.169 total time=
0.5s
[CV 2/5] END clf_leaf_size=20, clf__n_neighbors=10;, score=-0.179 total time=
0.6s
[CV 3/5] END clf_leaf_size=30, clf_n_neighbors=15;, score=-0.181 total time=
[CV 5/5] END clf_leaf_size=30, clf__n_neighbors=15;, score=-0.177 total time=
[CV 4/5] END clf_leaf_size=30, clf_n_neighbors=15;, score=-0.185 total time=
0.5s
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[CV 1/5] END clf leaf size=60, clf n neighbors=40;, score=-0.197 total time=
0.6s
[CV 3/5] END clf_leaf_size=60, clf_n neighbors=40;, score=-0.206 total time=
0.6s
[CV 4/5] END clf leaf size=60, clf n neighbors=40;, score=-0.206 total time=
0.6s
[CV 5/5] END clf leaf size=60, clf n neighbors=40;, score=-0.200 total time=
0.6s
[CV 2/5] END clf leaf size=60, clf n neighbors=40;, score=-0.214 total time=
0.7s
[CV 1/5] END clf_leaf_size=50, clf_n neighbors=15;, score=-0.179 total time=
0.4s
[CV 1/5] END clf_leaf_size=10, clf__n_neighbors=15;, score=-0.179 total time=
[CV 3/5] END clf_leaf_size=10, clf__n_neighbors=15;, score=-0.181 total time=
[CV 2/5] END clf__leaf_size=10, clf__n_neighbors=15;, score=-0.189 total time=
0.8s
[CV 4/5] END clf__leaf_size=10, clf__n_neighbors=15;, score=-0.185 total time=
0.8s
[CV 5/5] END clf__leaf_size=10, clf__n_neighbors=15;, score=-0.177 total time=
0.8s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-0.189 total time=
0.4s
[CV 3/5] END clf__leaf_size=50, clf__n_neighbors=15;, score=-0.181 total time=
0.4s
[CV 4/5] END clf_leaf_size=50, clf_n neighbors=15;, score=-0.185 total time=
0.4s
[CV 5/5] END clf_leaf_size=50, clf_n neighbors=15;, score=-0.177 total time=
[CV 1/5] END clf_leaf_size=45, clf_n neighbors=10;, score=-0.174 total time=
[CV 1/5] END clf_leaf_size=15, clf_n neighbors=20;, score=-0.185 total time=
0.6s
[CV 3/5] END clf leaf size=45, clf n neighbors=10;, score=-0.173 total time=
0.4s
[CV 2/5] END clf leaf size=45, clf n neighbors=10;, score=-0.179 total time=
0.4s
[CV 4/5] END clf__leaf_size=45, clf__n_neighbors=10;, score=-0.178 total time=
0.4s
[CV 3/5] END clf_leaf_size=15, clf__n_neighbors=20;, score=-0.187 total time=
0.6s
[CV 2/5] END clf__leaf_size=15, clf__n_neighbors=20;, score=-0.197 total time=
[CV 4/5] END clf_leaf_size=15, clf__n_neighbors=20;, score=-0.190 total time=
[CV 5/5] END clf_leaf_size=15, clf_n_neighbors=20;, score=-0.184 total time=
0.6s
```

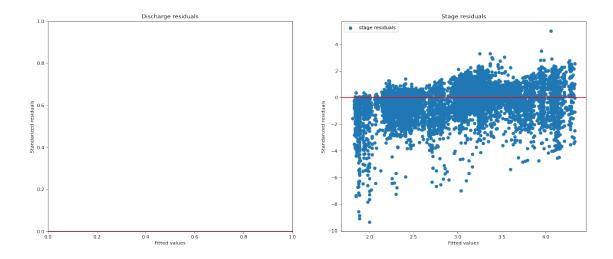
```
[CV 5/5] END clf leaf size=45, clf n neighbors=10;, score=-0.169 total time=
0.4s
[CV 2/5] END clf_leaf_size=60, clf_n neighbors=15;, score=-0.189 total time=
0.5s
[CV 1/5] END clf leaf size=60, clf n neighbors=15;, score=-0.179 total time=
0.5s
[CV 1/5] END clf leaf size=60, clf n neighbors=60;, score=-0.205 total time=
0.6s
[CV 3/5] END clf leaf size=60, clf n neighbors=15;, score=-0.181 total time=
0.5s
[CV 4/5] END clf_leaf_size=60, clf_n neighbors=15;, score=-0.185 total time=
0.5s
[CV 5/5] END clf__leaf_size=60, clf__n_neighbors=15;, score=-0.177 total time=
[CV 4/5] END clf__leaf_size=60, clf__n_neighbors=60;, score=-0.216 total time=
[CV 5/5] END clf_leaf_size=60, clf_n_neighbors=60;, score=-0.211 total time=
0.7s
[CV 2/5] END clf__leaf_size=60, clf__n_neighbors=60;, score=-0.224 total time=
0.7s
[CV 3/5] END clf__leaf_size=60, clf__n_neighbors=60;, score=-0.220 total time=
0.7s
[CV 1/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-0.174 total time=
0.4s
[CV 2/5] END clf__leaf_size=30, clf__n_neighbors=10;, score=-0.179 total time=
0.4s
[CV 3/5] END clf_leaf_size=30, clf_n neighbors=10;, score=-0.173 total time=
0.4s
[CV 4/5] END clf_leaf_size=30, clf_n neighbors=10;, score=-0.178 total time=
[CV 5/5] END clf_leaf_size=30, clf_n neighbors=10;, score=-0.169 total time=
[CV 1/5] END clf_leaf_size=30, clf_n neighbors=40;, score=-0.197 total time=
0.7s
[CV 2/5] END clf leaf size=30, clf n neighbors=40;, score=-0.214 total time=
0.7s
[CV 4/5] END clf leaf size=30, clf n neighbors=40;, score=-0.206 total time=
0.7s
[CV 3/5] END clf__leaf_size=30, clf__n_neighbors=40;, score=-0.206 total time=
0.7s
[CV 5/5] END clf_leaf_size=30, clf__n_neighbors=40;, score=-0.200 total time=
0.7s
[CV 1/5] END clf_leaf_size=15, clf__n_neighbors=15;, score=-0.179 total time=
[CV 3/5] END clf_leaf_size=15, clf__n_neighbors=15;, score=-0.181 total time=
[CV 4/5] END clf_leaf_size=15, clf_n_neighbors=15;, score=-0.185 total time=
0.6s
```

```
[CV 5/5] END clf leaf size=15, clf n neighbors=15;, score=-0.177 total time=
0.6s
[CV 2/5] END clf_leaf_size=15, clf_n neighbors=15;, score=-0.189 total time=
0.7s
[CV 1/5] END clf leaf size=20, clf n neighbors=40;, score=-0.197 total time=
0.8s
[CV 3/5] END clf leaf size=20, clf n neighbors=40;, score=-0.206 total time=
0.8s
[CV 2/5] END clf leaf size=20, clf n neighbors=40;, score=-0.214 total time=
0.9s
[CV 5/5] END clf_leaf_size=20, clf_n neighbors=40;, score=-0.200 total time=
0.8s
[CV 4/5] END clf_leaf_size=20, clf_n neighbors=40;, score=-0.206 total time=
[CV 1/5] END clf_leaf_size=60, clf_n_neighbors=20;, score=-0.185 total time=
[CV 1/5] END clf_leaf_size=15, clf__n_neighbors=60;, score=-0.205 total time=
1.0s
[CV 2/5] END clf__leaf_size=60, clf__n_neighbors=20;, score=-0.197 total time=
0.6s
[CV 3/5] END clf__leaf_size=15, clf__n_neighbors=60;, score=-0.220 total time=
1.0s
[CV 4/5] END clf__leaf_size=60, clf__n_neighbors=20;, score=-0.190 total time=
0.6s
[CV 3/5] END clf__leaf_size=60, clf__n_neighbors=20;, score=-0.187 total time=
0.6s
[CV 4/5] END clf_leaf_size=15, clf_n neighbors=60;, score=-0.216 total time=
1.0s
[CV 5/5] END clf_leaf_size=60, clf_n neighbors=20;, score=-0.184 total time=
[CV 2/5] END clf_leaf_size=15, clf_n neighbors=60;, score=-0.224 total time=
[CV 1/5] END clf_leaf_size=60, clf_n neighbors=5;, score=-0.162 total time=
0.3s
[CV 1/5] END clf leaf size=30, clf n neighbors=20;, score=-0.185 total time=
0.6s
[CV 2/5] END clf leaf size=60, clf n neighbors=5;, score=-0.177 total time=
0.3s
[CV 5/5] END clf__leaf_size=15, clf__n_neighbors=60;, score=-0.211 total time=
1.3s
[CV 2/5] END clf_leaf_size=30, clf__n_neighbors=20;, score=-0.197 total time=
0.5s
[CV 3/5] END clf__leaf_size=30, clf__n_neighbors=20;, score=-0.187 total time=
[CV 4/5] END clf_leaf_size=30, clf_n_neighbors=20;, score=-0.190 total time=
[CV 5/5] END clf_leaf_size=30, clf_n neighbors=20;, score=-0.184 total time=
```

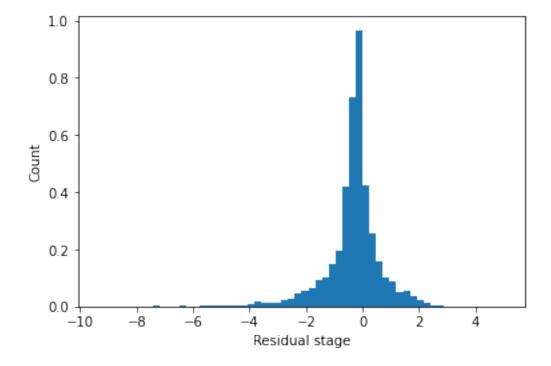
0.6s

```
[CV 5/5] END clf_leaf_size=60, clf_n neighbors=5;, score=-0.164 total time=
    0.3s
    [CV 3/5] END clf_leaf_size=60, clf_n neighbors=5;, score=-0.171 total time=
    0.3s
    [CV 4/5] END clf leaf size=60, clf n neighbors=5;, score=-0.180 total time=
    0.3s
    [CV 1/5] END clf leaf size=15, clf n neighbors=5;, score=-0.162 total time=
    0.3s
    [CV 3/5] END clf leaf size=15, clf n neighbors=5;, score=-0.171 total time=
    0.4s
    [CV 4/5] END clf_leaf_size=15, clf_n neighbors=5;, score=-0.180 total time=
    0.4s
    [CV 2/5] END clf__leaf_size=15, clf__n_neighbors=5;, score=-0.177 total time=
    [CV 5/5] END clf_leaf_size=15, clf_n neighbors=5;, score=-0.164 total time=
    0.3s
[]: RandomizedSearchCV(estimator=Pipeline(steps=[('scaler', StandardScaler()),
                                                 ('clf', KNeighborsRegressor())]),
                       n_iter=30, 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_
[]: -0.1705751897108685
[]: clf.best params
[]: {'clf_n_neighbors': 5, 'clf_leaf_size': 50}
    1.5 Test model
[]: clf.score(X_test, y_test)
[]: -0.19180460124145515
[]: 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: 0.5088720405117706
    mse: 0.19180460124145515
    rmse: 0.43795502193884606
    mae: 0.27828899190696943
    mape: 0.1032312787570824
    Error estandar: 0.41655566262825766
[]: 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(), __
     → 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()
```

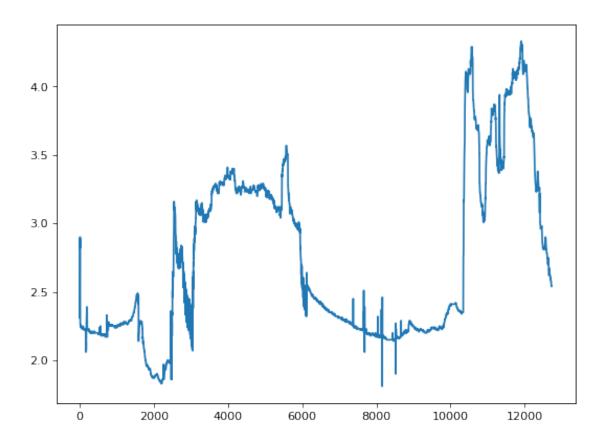


```
[]: plt.hist(residual_stage / residual_stage.std(), density=True, bins = 60)
   plt.ylabel('Count')
   plt.xlabel('Residual stage');
   plt.show()
```



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

[]: [<matplotlib.lines.Line2D at 0x7f174c5ff6d0>]



```
[]: 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")
```

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.

