# knn\_regression\_v1\_stage\_2

November 25, 2022

## 1 KNN regression

```
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.neighbors import KNeighborsRegressor 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

```
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
grayMean
                        float64
graySigma
                        float64
entropyMean
                        float64
entropySigma
                        float64
hMean
                        float64
hSigma
                        float64
sMean
                        float64
                        float64
sSigma
vMean
                        float64
vSigma
                        float64
Year
                          int64
dtype: object
```

```
[]: df = df[(df.Stage > 0) & (df.Discharge > 0)]
```

```
[]: df.isna().sum()
```

```
[]: SensorTime
                      0
     CaptureTime
                      0
     Stage
                      0
     Discharge
                      0
     grayMean
                      0
                      0
     graySigma
     entropyMean
                      0
     entropySigma
     hMean
     hSigma
                      0
     sMean
                      0
                      0
     sSigma
     vMean
                      0
                      0
     vSigma
     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"])</pre>
```

df\_test = df\_test.drop(columns=["Year", "SensorTime", "CaptureTime"])

```
[]: #y_train = df_train[["Stage", "Discharge"]]
    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"])
```

```
[]: \#X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.33, \_ \rightarrow random\_state=0)
```

#### 1.4 Train model

#### []: clf.fit(X\_train, y\_train)

```
Fitting 5 folds for each of 30 candidates, totalling 150 fits
[CV 1/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.284 total time=
0.4s[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.271 total
time= 0.3s

[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.269 total time=
0.4s[CV 4/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.291 total
time= 0.4s

[CV 5/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.272 total time=
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[CV 4/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.283 total time=
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[CV 3/5] END clf__leaf_size=50, clf__n_neighbors=5;, score=-0.282 total time=
0.4s
[CV 2/5] END clf__leaf_size=50, clf__n_neighbors=10;, score=-0.266 total time=
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0.5s
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- [CV 1/5] END clf\_\_leaf\_size=50, clf\_\_n\_neighbors=10;, score=-0.290 total time= 0.4s
- [CV 3/5] END clf\_leaf\_size=30, clf\_n\_neighbors=5;, score=-0.282 total time= 0.2s
- [CV 1/5] END clf\_leaf\_size=30, clf\_n\_neighbors=5;, score=-0.284 total time= 0.2s
- [CV 4/5] END clf\_leaf\_size=30, clf\_n\_neighbors=5;, score=-0.291 total time= 0.2s
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- [CV 2/5] END clf\_leaf\_size=30, clf\_n\_neighbors=5;, score=-0.269 total time= 0.3s
- [CV 1/5] END clf\_leaf\_size=45, clf\_n\_neighbors=60;, score=-0.326 total time= 0.5s
- [CV 2/5] END clf\_\_leaf\_size=45, clf\_\_n\_neighbors=60;, score=-0.312 total time= 0.5s
- [CV 4/5] END clf\_leaf\_size=45, clf\_\_n\_neighbors=60;, score=-0.330 total time= 0.5s
- [CV 3/5] END clf\_\_leaf\_size=45, clf\_\_n\_neighbors=60;, score=-0.312 total time= 0.5s
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- [CV 5/5] END clf\_leaf\_size=45, clf\_n\_neighbors=60;, score=-0.317 total time= 0.6s
- [CV 2/5] END clf\_leaf\_size=15, clf\_n\_neighbors=10;, score=-0.266 total time= 0.3s
- [CV 5/5] END clf\_leaf\_size=15, clf\_\_n\_neighbors=10;, score=-0.272 total time= 0.3s
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- [CV 1/5] END clf\_leaf\_size=50, clf\_n\_neighbors=40;, score=-0.313 total time= 0.4s
- [CV 2/5] END clf\_\_leaf\_size=50, clf\_\_n\_neighbors=40;, score=-0.298 total time= 0.4s
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- [CV 4/5] END clf\_\_leaf\_size=50, clf\_\_n\_neighbors=40;, score=-0.317 total time= 0.4s
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    0.3s
    [CV 3/5] END clf_leaf_size=60, clf_n neighbors=5;, score=-0.282 total time=
    0.3s
    [CV 5/5] END clf_leaf_size=30, clf_n_neighbors=20;, score=-0.285 total time=
    0.3s
    [CV 5/5] END clf__leaf_size=60, clf__n_neighbors=5;, score=-0.271 total time=
    0.2s
    [CV 1/5] END clf__leaf_size=15, clf__n_neighbors=5;, score=-0.284 total time=
    0.2s
    [CV 2/5] END clf__leaf_size=15, clf__n_neighbors=5;, score=-0.269 total time=
    [CV 3/5] END clf__leaf_size=15, clf__n_neighbors=5;, score=-0.282 total time=
    [CV 4/5] END clf__leaf_size=15, clf__n_neighbors=5;, score=-0.291 total time=
    [CV 5/5] END clf_leaf_size=15, clf_n neighbors=5;, score=-0.271 total time=
    0.2s
[]: RandomizedSearchCV(estimator=Pipeline(steps=[('scaler', StandardScaler()),
                                                 ('clf', KNeighborsRegressor())]),
                       n_iter=30, n_jobs=10,
```

0.6s

```
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_
[]: -0.276913137286918
[]: clf.best params
[]: {'clf_n_neighbors': 10, 'clf_leaf_size': 50}
    1.5 Test model
[]: clf.score(X_test, y_test)
[]: -0.3425465467117153
[]: 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: 0.12288763967408212
    mse: 0.3425465467117153
    rmse: 0.5852747617245385
    mae: 0.4101173882297478
    mape: 0.15431529814758768
    Error estandar: 0.5345684473064409
[]: 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(),u residual="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")

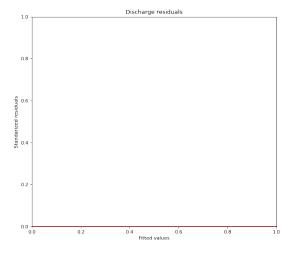
ax[0].set_title("Discharge residuals")

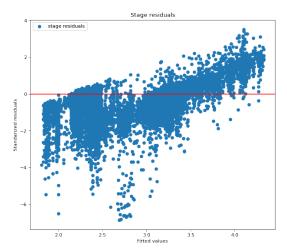
ax[1].set_xlabel("Fitted values")

ax[1].set_ylabel("Standarized residuals")

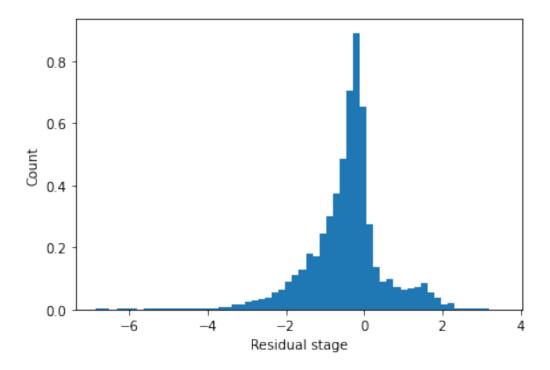
ax[0].set_ylabel("Standarized residuals")

plt.legend()
plt.show()
```



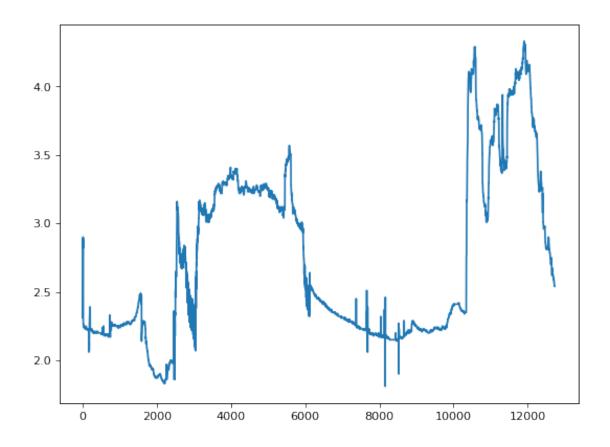


```
[]: plt.hist(residual_stage / residual_stage.std(), density=True, bins = 60)
    plt.ylabel('Count')
    plt.xlabel('Residual stage');
    plt.show()
```



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

[]: [<matplotlib.lines.Line2D at 0x7f9bd87e17b0>]



```
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].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.

