

Iris-flower

November 17, 2022

A machine learning system to predict the iris flower type based on the K nearest neighbors algorithm.

The independent variables are

- sepal length
- sepal width
- petal length
- petal width

The dependent variable is

- The iris flower type

Dataset

- The dataset used for this experiment can be downloaded from this [link](#)

Tools

To work with this project, multiple libraries and frameworks need to be installed. The following is a list of them. - [Pandas](#) - [NumPy](#) - [Matplotlib](#) - [Scikit-learn](#)

Import the necessary packets.

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

The dataset is unlabeled; an assigned label is needed

```
[4]: headers = ['sepal-length', 'sepal-width', 'petal-length', 'petal-width',
               → 'Class']
# read the iris.data as csv and assign the label
df = pd.read_csv('iris.data', names = headers)

df.head(3)
```

```
[4]:   sepal-length  sepal-width  petal-length  petal-width      Class
0         5.1         3.5         1.4         0.2  Iris-setosa
1         4.9         3.0         1.4         0.2  Iris-setosa
2         4.7         3.2         1.3         0.2  Iris-setosa
```

Data preprocessing Select the first four columns as independent (x) variables and the last column as dependent(y)

```
[5]: x = df.iloc[:, :-1].values
     y = df.iloc[:, 4].values
```

Normalize the data

we only have numeric data so we don't use onehotencoder

```
[6]: from sklearn.preprocessing import StandardScaler
     scaler = StandardScaler()
     scaler.fit(x)
     x = scaler.transform(x)
```

Split the data into train and test

```
[9]: from sklearn.model_selection import train_test_split
     X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.40)
```

import the k-nn model

```
[14]: from sklearn.neighbors import KNeighborsClassifier
     k = 10
     accuracy_result = []
     for i in range(1, k+1):

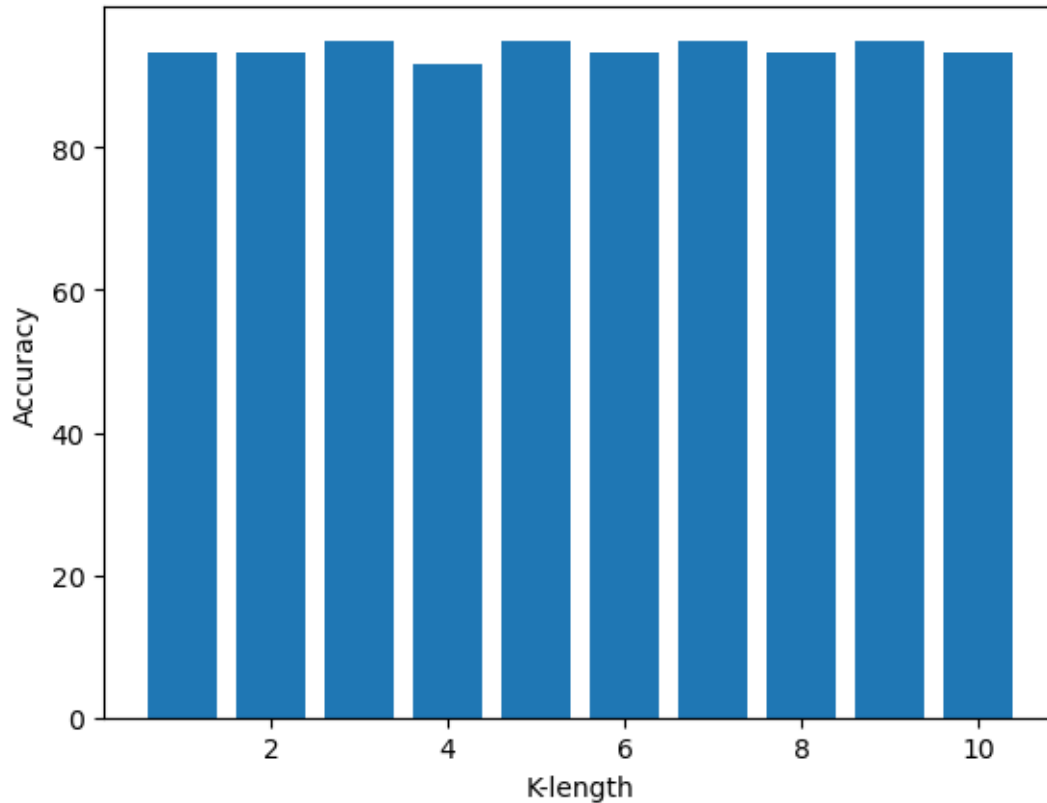
         model = KNeighborsClassifier(n_neighbors = i).fit(X_train, y_train)
         yhat = model.predict(X_test)

         from sklearn import metrics
         # print("Train set Accuracy: ", metrics.accuracy_score(y_train, model.
         ↪predict(X_train))*100)
         ac = metrics.accuracy_score(y_test, yhat)
         # print("Test set Accuracy: ", ac )

         accuracy_result.append(ac)

     plt.bar(np.arange(1, k+1), [i*100 for i in accuracy_result])
     plt.xlabel('K-length')
     plt.ylabel('Accuracy')
```

```
[14]: Text(0, 0.5, 'Accuracy')
```



```
[16]: print('The best accuracy achieved where k is equal to ', str(accuracy_result.
↪index(max(accuracy_result))), "The accuracy is_
↪",str(accuracy_result[accuracy_result.index(max(accuracy_result))]*100)+"%")
```

The best accuracy achieved where k is equal to 2 The accuracy is 95.0%

Deploying

```
[17]: import joblib
import os
if not os.path.exists('Model'):
    os.mkdir('Model')
if not os.path.exists('Scaler'):
    os.mkdir('Scaler')

joblib.dump(model, r'Model/model.pickle')
joblib.dump(scaler, r'Scaler/scaler.pickle')
```

```
[17]: ['Scaler/scaler.pickle']
```

To use the proposed model with real data create another file deploy.py and this lines

```
[19]: import pandas as pd

# 1- define the new data
new_data = pd.DataFrame([{'sepal-length':5.3, 'sepal-width':3.7, 'petal-length':
    ↳1.6, 'petal-width':0.22}])
new_data = new_data[['sepal-length','sepal-width','petal-length','petal-width']]

# 2- import the Scalar and KNN models

import joblib

model = joblib.load(r'Model/model.pickle')
scaler = joblib.load(r'Scaler/scaler.pickle')

# 3- Normalize the data
new_data = scaler.transform(new_data)
predict_calass = model.predict(new_data)

print(predict_calass)
```

```
['Iris-setosa']
```