29. Data Science – Machine Learning – K Nearest Neighbor

Contents

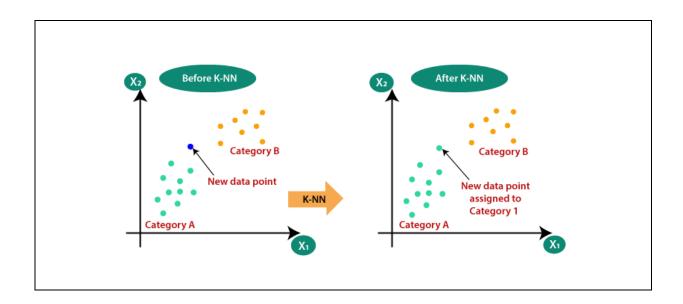
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1. K-Nearest Neighbor Algorithm

- ✓ K-Nearest Neighbor is a Supervised Learning technique.
- ✓ K-NN algorithm follow one basic rule that is, similar things are near to each other.
- ✓ It is also called a lazy learner algorithm because it does not learn from the training set immediately.
 - At the time of training phase this algorithm just stores the dataset
 - Whenever we get new data point then it classifies the category

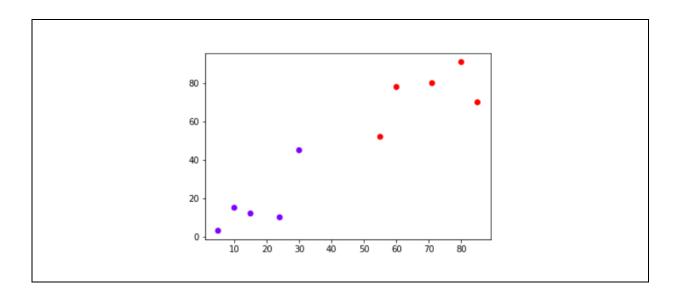
2. How it works?

- ✓ It simply calculates the distance of a new data point to all other training data points.
- ✓ The distance can be of any type e.g. Euclidean or Manhattan etc.
- ✓ It selects the K-nearest data points.
- ✓ Finally it assigns the data point to the class to which the majority of the K data points belong.

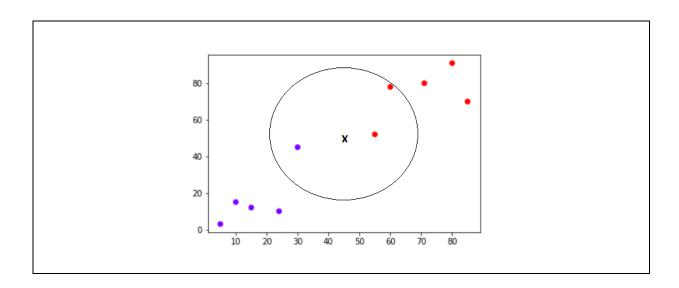


3. Scenario

✓ Suppose you have a dataset with two variables, which when plotted, looks like the one in the following figure.



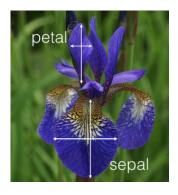
- ✓ Our task is to classify a new data point with 'X' into "Blue" class or "Red" class.
- ✓ Suppose the value of K is 3.
- ✓ The KNN algorithm starts by calculating the distance of point X from all the points.
- ✓ It then finds the 3 nearest points with least distance to point X.
- ✓ This is shown in the figure below; the three nearest points have been encircled.



- ✓ The final step of the KNN algorithm is to assign new point to the class to which majority of the three nearest points belong.
- ✓ From the above image we can see that the two of the three nearest points belong to the class "Red" while one belongs to the class "Blue".
- ✓ Therefore the new data point will be classified as "Red".

4. Use case

- ✓ Assuming that Abhi had a hobby which is interested in distinguishing the species of some iris flowers that he has found
- ✓ He has collected some measurements associated with each iris, which are:
 - The length and width of the petals
 - o The length and width of the sepals, all measured in centimetres.
- ✓ She also has the measurements of some irises that have been previously identified to the species
 - o setosa,
 - versicolor
 - o virginica
- ✓ The goal is to create a machine learning model that can learn from the measurements of these irises whose species are already known.
- ✓ So that we can predict the species for the new irises that she has found.









Iris Setosa



Iris Virginica

Flower codes

- ✓ Setosa 0
- ✓ Versicolor 1
- ✓ Virginica 2

Program Loading iris dataset

Name demo1.py

from sklearn.datasets import load_iris

iris = load_iris()

print(dir(iris))

Output

['DESCR', 'data', 'feature_names', 'filename', 'frame', 'target',

'target_names']

Program Name Displaying feature names

demo2.py

from sklearn.datasets import load_iris

iris = load_iris()

print(iris.feature_names)

Output

['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)', 'petal

width (cm)']

Program Name Displaying target names

demo3.py

from sklearn.datasets import load_iris

iris = load_iris()

print(iris.target_names)

Output

['setosa' 'versicolor' 'virginica']

Program Displaying data demo4.py Name from sklearn.datasets import load_iris iris = load_iris() print(iris.data) Output [[5.1 3.5 1.4 0.2] [4.9 3. 1.4 0.2] [4.7 3.2 1.3 0.2] [4.6 3.1 1.5 0.2] [5. 3.6 1.4 0.2] [5.4 3.9 1.7 0.4] [4.6 3.4 1.4 0.3] [5. 3.4 1.5 0.2] [4.4 2.9 1.4 0.2] [4.9 3.1 1.5 0.1] [4.8 3.4 1.6 0.2] 1.1 0.1

Program Length of the data

Name demo5.py

from sklearn.datasets import load_iris

iris = load_iris()

print(len(iris.data))

Output

Program Name

Create a Dataframe by using data and features demo6.py

import pandas as pd
from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)

print(df)

Output

```
      sepal length (cm)
      sepal length (cm)
      petal length (cm)
      petal width (cm)

      0
      5.1
      3.5
      1.4
      0.2

      1
      4.9
      3.0
      1.4
      0.2

      2
      4.7
      3.2
      1.3
      0.2

      3
      4.6
      3.1
      1.5
      0.2

      4
      5.0
      3.6
      1.4
      0.2

      ...
      ...
      ...
      ...
      ...
      ...

      145
      6.7
      3.0
      5.2
      2.3

      146
      6.3
      2.5
      5.0
      1.9

      147
      6.5
      3.0
      5.2
      2.0

      148
      6.2
      3.4
      5.4
      2.3

      149
      5.9
      3.0
      5.1
      1.8
```

```
Program
Name

Displaying target == 0 flowers

demo8.py

import pandas as pd
from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)

df['target'] = iris.target

print(df[df.target==0].head())

Output

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target
1.4 0.2 0
1.4 0.2 0
1.4 0.2 0
1.4 0.2 0
1.4 0.2 0
1.5 0.2 0
1.6 3.1 1.5 0.2 0
1.6 0.2 0
1.7 3.2 1.3 0.2 0
1.8 0.2 0
1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
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1.9 0.2 0
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1.9 0.2 0
1.9 0.2 0
1.9 0.2 0
```

```
Program
Displaying length of the target == 0 flowers
demo9.py

import pandas as pd
from sklearn.datasets import load_iris
```

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)
df['target'] = iris.target

print(len(df[df.target==0]))

Output

```
Program
Name

Displaying length of the target == 0 flowers

demo11.py

import pandas as pd
from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)

df['target'] = iris.target

print(len(df[df.target==1]))

Output
```

Program Name Displaying target == 2 flowers demo12.py import pandas as pd from sklearn.datasets import load_iris iris = load_iris() df = pd.DataFrame(iris.data, columns=iris.feature_names) df['target'] = iris.target print(df[df.target==2].head()) Output sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target 100 6.3 3.3 6.0 2.5 2.1 101 5.8 2.7 5.1 1.9 2 102 7.1 3.0 5.9 2.1 2 103 6.3 2.9 5.6 1.8 2 104 6.5 3.0 5.8 2.2 2

```
Program Displaying length of the target == 2 flowers

demo13.py

import pandas as pd
from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)
df['target'] = iris.target
```

print(len(df[df.target==2]))

Output

Program Name Displaying the flower names demo14.py import pandas as pd from sklearn.datasets import load_iris iris = load_iris() df = pd.DataFrame(iris.data, columns=iris.feature_names) df['target'] = iris.target df['flower_name'] = df.target.apply(lambda x: iris.target_names[x]) print(df) Output

```
    sepal length (cm)
    sepal width (cm)
    petal length (cm)
    petal width (cm)
    target flower_name

    0
    5.1
    3.5
    1.4
    0.2
    0
    setosa

    1
    4.9
    3.0
    1.4
    0.2
    0
    setosa

    2
    4.7
    3.2
    1.3
    0.2
    0
    setosa

    3
    4.6
    3.1
    1.5
    0.2
    0
    setosa

    4
    5.0
    3.6
    1.4
    0.2
    0
    setosa

    ...
    ...
    ...
    ...
    ...
    ...
    ...
    ...

    145
    6.7
    3.0
    5.2
    2.3
    2
    virginica

    146
    6.3
    2.5
    5.0
    1.9
    2
    virginica

    147
    6.5
    3.0
    5.2
    2.0
    2
    virginica

    148
    6.2
    3.4
    5.4
    2.3
    2
    virginica

    149
    5.9
    3.0
    5.1
    1.8
    2
    virginica

    [150 rows x 6 columns]
```

```
Program Name demo15.py

import pandas as pd
from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)

df['target'] = iris.target
df['flower_name'] = df.target.apply(lambda x:
iris.target_names[x])

setosa_50 = df[:50]
print(setosa_50.head())

Output

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target flower_name for the column petal length (cm) petal width (cm) target flower_name for the column petal length (cm) petal width (cm) target flower_name for the column petal length for the column petal le
```

```
Program All versicolor flowers demo16.py

import pandas as pd from sklearn.datasets import load_iris

iris = load_iris()

df = pd.DataFrame(iris.data, columns=iris.feature_names)

df['target'] = iris.target df['flower_name'] = df.target.apply(lambda x: iris.target_names[x])

versicolor_50 = df[50:100] print(versicolor_50.head())

Output

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target flower_name flower_na
```

```
Splitting the data
Program
            demo18.py
Name
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature names)
            df['target'] = iris.target
            df['flower_name'] = df.target.apply(lambda x:
            iris.target names[x])
            X = df.drop(['target', 'flower_name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test_size=0.2)
            print("Splitting the data")
Output
            Splitting the data
```

```
Program
            Model training
Name
            demo19.py
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            from sklearn.neighbors import KNeighborsClassifier
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature_names)
            df['target'] = iris.target
            df['flower name'] = df.target.apply(lambda x:
            iris.target_names[x])
            X = df.drop(['target', 'flower name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test_size=0.2)
            # Train Using K Neighbor classifier
            classifier = KNeighborsClassifier(n_neighbors = 5)
             classifier.fit(X train, y train)
            print('Model got trained')
Output
            Model got trained
```

```
Program
            Model score
            demo20.py
Name
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            from sklearn.neighbors import KNeighborsClassifier
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature_names)
            df['target'] = iris.target
            df['flower name'] = df.target.apply(lambda x:
            iris.target names[x])
            X = df.drop(['target', 'flower name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test_size=0.2)
            classifier = KNeighborsClassifier(n neighbors=5)
            classifier.fit(X_train, y_train)
            print(classifier.score(X_test, y_test))
Output
            0.966666666666666
```

```
Program
            Model prediction
            demo21.py
Name
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            from sklearn.neighbors import KNeighborsClassifier
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature_names)
            df['target'] = iris.target
            df['flower name'] = df.target.apply(lambda x:
            iris.target names[x])
            X = df.drop(['target', 'flower name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test_size=0.2)
            classifier = KNeighborsClassifier(n neighbors=5)
            classifier.fit(X_train, y_train)
            print(classifier.predict([[4.8, 3.0, 1.5, 0.3]]))
Output
            [0]
```

```
Program
            Model prediction
            demo22.py
Name
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            from sklearn.neighbors import KNeighborsClassifier
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature_names)
            df['target'] = iris.target
            df['flower name'] = df.target.apply(lambda x:
            iris.target names[x])
            X = df.drop(['target', 'flower name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test_size=0.2)
            classifier = KNeighborsClassifier(n neighbors=5)
            classifier.fit(X_train, y_train)
            y_pred = classifier.predict(X_test)
            print(y_pred)
Output
            [010121012002101000011022020010]
```

```
Program
            Model evaluation
Name
            demo22.py
            import pandas as pd
            from sklearn.datasets import load iris
            from sklearn.model_selection import train_test_split
            from sklearn.neighbors import KNeighborsClassifier
            from sklearn.metrics import classification report,
            confusion matrix
            iris = load_iris()
            df = pd.DataFrame(iris.data, columns=iris.feature names)
            df['target'] = iris.target
            df['flower name'] = df.target.apply(lambda x:
            iris.target names[x])
            X = df.drop(['target', 'flower name'], axis='columns')
            y = df.target
            X_train, X_test, y_train, y_test = train_test_split(X, y,
            test size=0.2)
            classifier = KNeighborsClassifier(n_neighbors=5)
            classifier.fit(X_train, y_train)
            y pred = classifier.predict(X test)
            print(confusion_matrix(y_test, y_pred))
            print(classification_report(y_test, y_pred))
```

Output

```
[[13 0 0]
[0 7 2]
[0 0 8]]
               precision
                             recall f1-score
                                                  support
                                           1.00
                                                        13
            0
                     1.00
                                1.00
            1
                     1.00
                                0.78
                                           0.88
                                                         9
            2
                                1.00
                                           0.89
                                                         8
                     0.80
                                           0.93
                                                        30
    accuracy
                                0.93
                                           0.92
                                                        30
   macro avg
                     0.93
                                           0.93
weighted avg
                     0.95
                                0.93
                                                        30
```