

# FUNDAMENTALS OF MACHINE LEARNING IN DATA SCIENCE

CSIS 3290
SUPERVISED LEARNING 2 (SVM, KNN, NAÏVE BAYES)
IN SKLEARN
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#### **KNN**

```
In [13]: import pandas as pd
         from sklearn import datasets
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.model_selection import train_test_split
 In [2]: iris1=datasets.load_iris()
 In [4]: iris1.data.shape
 Out[4]: (150, 4)
         iris1.feature_names
 In [5]: iris1.feature_names
 Out[5]: ['sepal length (cm)',
          'sepal width (cm)',
          'petal length (cm)',
          'petal width (cm)']
 In [6]: iris1.target_names
 Out[6]: array(['setosa', 'versicolor', 'virginica'], dtype='<U10')
```

#### KNN

The parameter p is used to specify the **power parameter** for the Minkowski metric. When p is set to 1, this is equivalent to using manhattan\_distance (I1). When we set p=2, which is its default value, the Minkowski metric works as the euclidean distance metric.

```
In [7]: x=iris1.data
 In [8]: y=iris1.target
In [17]: x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.3, random_state=42, stratify=y)
         x train.shape
         x_test.shape
Out[17]: (45, 4)
In [15]: knn1=KNeighborsClassifier(n_neighbors=6, metric='minkowski', p=2)
In [18]: knn1.fit(x_train,y_train)
Out[18]:
                  KNeighborsClassifier
         KNeighborsClassifier(n_neighbors=6)
In [19]: y predict=knn1.predict(x test)
In [20]: print(knn1.score(x test, y test))
         0.95555555555556
```

### Naïve Bayes

```
In [27]: import pandas as pd
    from sklearn import datasets
    from sklearn.neighbors import KNeighborsClassifier
    from sklearn.model_selection import train_test_split
    from sklearn.naive_bayes import GaussianNB
    from sklearn.metrics import confusion_matrix, classification_report
```

### Naïve Bayes

```
In [23]: gnb1=GaussianNB()
In [24]: y_pred=gnb1.fit(x_train,y_train).predict(x_test)
In [25]: print(gnb1.score(x_test, y_test))
         0.9111111111111111
In [31]: print(confusion_matrix(y_test,y_pred))
         [[15 0 0]
          [ 0 14 1]
          [ 0 3 12]]
In [32]: print(classification_report(y_test,y_pred))
                       precision
                                    recall f1-score
                                                       support
                                                1.00
                            1.00
                                      1.00
                                                            15
                            0.82
                                      0.93
                                                0.87
                                                            15
                            0.92
                                      0.80
                                                0.86
                                                            15
                                                0.91
                                                            45
             accuracy
                            0.92
                                      0.91
                                                0.91
                                                            45
            macro avg
         weighted avg
                                                            45
                            0.92
                                      0.91
                                                0.91
```

## **SVM** (with Linear Kernel)

```
In [34]: import pandas as pd
        from sklearn import datasets
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.model_selection import train_test_split
         from sklearn.svm import SVC
         from sklearn.naive bayes import GaussianNB
         from sklearn.metrics import confusion_matrix, classification_report
 In [36]: svm1 = SVC(kernel='linear', random_state=0)
 In [37]: svm1.fit(x_train,y_train)
 Out[37]:
                               SVC
            SVC(kernel='linear', random_state=0)
 In [38]: pred2=svm1.predict(x_test)
 In [39]: print(svm1.score(x_test, y_test))
            1.0
```