

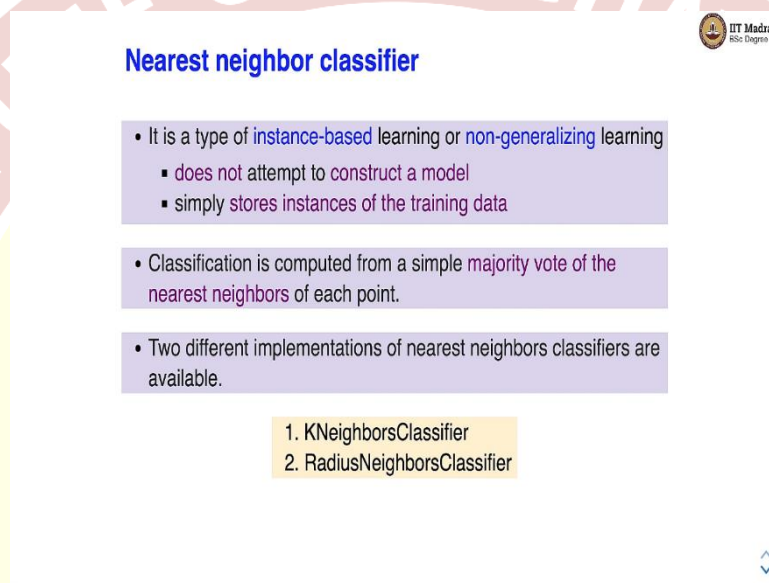
IIT Madras

ONLINE DEGREE

Machine Learning Program
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K - Nearest Neighbors

Namaste! welcome to the next video of Machine Learning Practice Course. In this video, we will discuss how to implement K-Nearest Neighbor with sklearn. We will mostly be doing this discussion in the context of classification.

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The slide is titled "Nearest neighbor classifier" and features the IIT Madras logo in the top right corner. It contains three bullet points describing the classifier, followed by a list of two sklearn implementations. The background of the slide is white, and the text is in a sans-serif font. The bullet points are highlighted with light blue and light orange backgrounds. The list of implementations is highlighted with a light orange background.

- It is a type of **instance-based** learning or **non-generalizing** learning
 - does not attempt to construct a model
 - simply stores instances of the training data
- Classification is computed from a simple **majority vote** of the **nearest neighbors** of each point.
- Two different implementations of nearest neighbors classifiers are available.

1. KNeighborsClassifier
2. RadiusNeighborsClassifier

So, as you know nearest neighbor classifier is a type of instance-based learning or non-generalizing learning. It does not attempt to construct a model, but it simply stores instances of the training data. Classification is computed from a simple majority vote of the nearest neighbor of each data point. There are two different implementations of nearest neighbor classifiers in sklearn. One is KNeighborsClassifier and second is RadiusNeighborsClassifier.

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How are KNeighborsClassifier and RadiusNeighborsClassifier different?

KNeighborsClassifier	RadiusNeighborsClassifier
<ul style="list-style-type: none">• learning based on the k nearest neighbors	<ul style="list-style-type: none">• learning based on the number of neighbors within a fixed radius r of each training point
<ul style="list-style-type: none">• most commonly used technique	<ul style="list-style-type: none">• used in cases where the data is not uniformly sampled
<ul style="list-style-type: none">• choice of the value k is highly data-dependent	<ul style="list-style-type: none">• fixed value of r is specified, such that points in sparser neighborhoods use fewer nearest neighbors for the classification

So, let us see, how KNeighborsClassifier and RadiusNeighborsClassifiers differ. So, KNeighborsClassifier learns based on k-nearest neighbors, whereas, the RadiusNeighborsClassifier learn based on the number of neighbors within a fixed radius r of each training point. KNeighborsClassifier is most commonly used techniques, whereas, RadiusNeighborsClassifier is used in cases where the data is not uniformly sampled.

In KNeighborsClassifier, the choice of the value of k is highly data-dependent, whereas, in RadiusNeighborsClassifier, we used fixed value of r , such that the points in sparser neighborhood use fewer nearest neighbors for the classification.

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How do you apply KNeighborsClassifier?

Step 1: Instantiate a KNeighborsClassifier estimator without passing any arguments to it to create a classifier object.

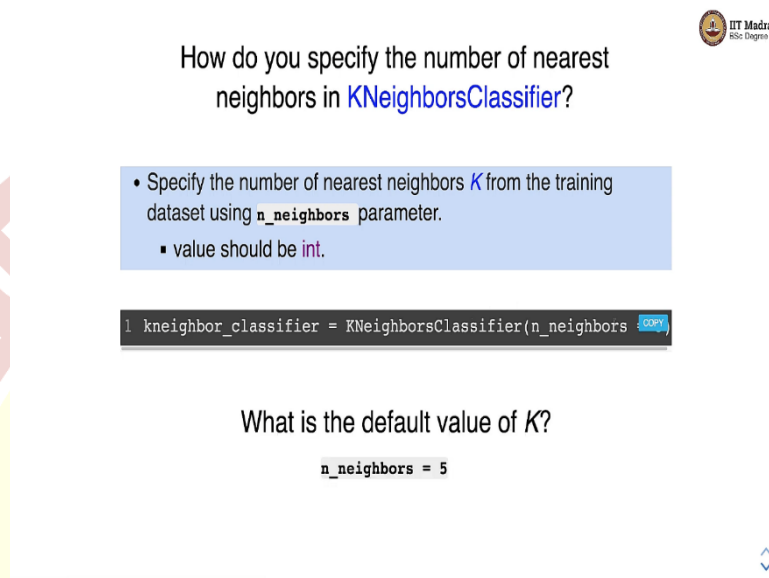
```
1 from sklearn.neighbors import KNeighborsClassifier
2 neighbor_classifier = KNeighborsClassifier()
```

Step 2: Call fit method on KNeighbors classifier object with training feature matrix and label vector as arguments.

```
1 # Model training with feature matrix X_train and
2 # label vector or matrix y_train
3 neighbor_classifier.fit(X_train, y_train)
```

Let us see, how to apply KNeighborsClassifier. We instantiate a KNeighborsClassifier without passing any argument to create a classifier object. So, we import KNeighborsClassifier from sklearn.neighbors module. And in the second step, we call fit method on the classifier object with training feature matrix and label vector as arguments.

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How do you specify the number of nearest neighbors in **KNeighborsClassifier**?

- Specify the number of nearest neighbors **K** from the training dataset using **n_neighbors** parameter.
 - value should be **int**.

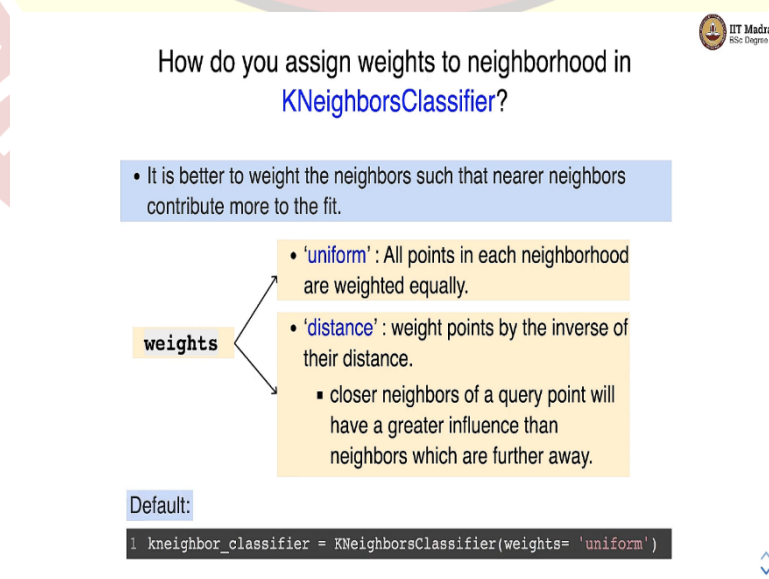
```
1 neighbor_classifier = KNeighborsClassifier(n_neighbors=5)
```

What is the default value of **K**?

```
n_neighbors = 5
```

Let us say, how to specify the number of nearest neighbor in KNeighborsClassifier. So, there is a parameter called **n_neighbors** that we need to basically set. So, here **k** while instantiating object of KNeighborsClassifier, we set **n_neighbors** to the desired value and the default value of nearest neighbor is 5.

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How do you assign weights to neighborhood in **KNeighborsClassifier**?

- It is better to weight the neighbors such that nearer neighbors contribute more to the fit.

weights

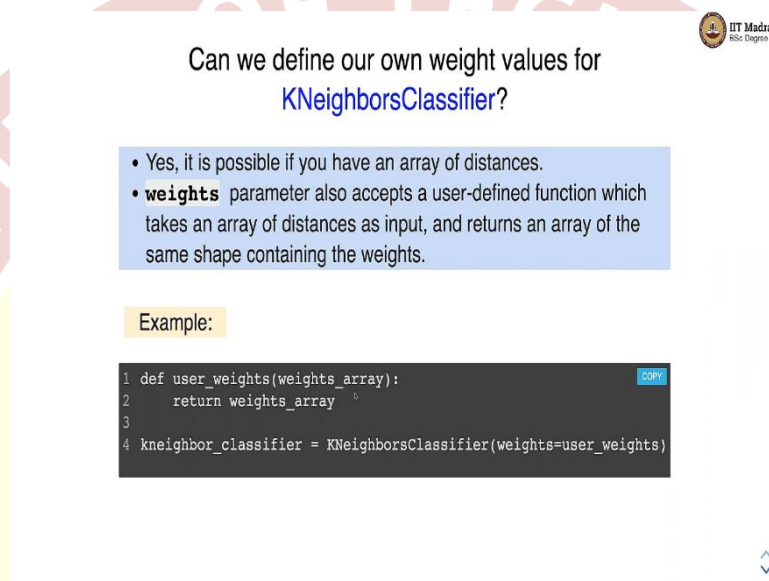
- 'uniform'**: All points in each neighborhood are weighted equally.
- 'distance'**: weight points by the inverse of their distance.
 - closer neighbors of a query point will have a greater influence than neighbors which are further away.

Default:

```
1 neighbor_classifier = KNeighborsClassifier(weights='uniform')
```

So, there is a possibility that we can assign weights to the neighborhood. So, the idea here is that the higher weights are assigned to the nearest neighbor, so that they contribute more towards the labelling of the new point. The weights can be uniform, where all points in the neighborhood are weighted equally or weight can be based on the distance. The points that are closer will have greater influence and the points that are further away. By default, KNeighborsClassifier uses uniform weight.

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Can we define our own weight values for KNeighborsClassifier?

- Yes, it is possible if you have an array of distances.
- **weights** parameter also accepts a user-defined function which takes an array of distances as input, and returns an array of the same shape containing the weights.

Example:

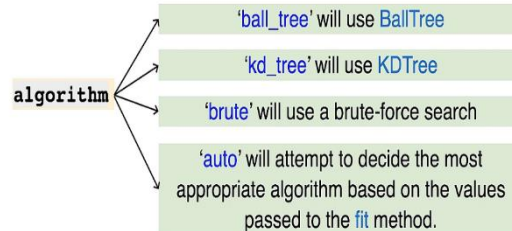
```
1 def user_weights(weights_array):  
2     return weights_array  
3  
4 neighbor_classifier = KNeighborsClassifier(weights=user_weights)
```

So, can we define our own weight? Yes, it is quite possible that we can define our own weights as a function of the distance. So, the weight parameter also accepts a user-defined function, which takes an array of distance as input, and returns an array of the same shape containing the weights.

So, here we have shown an example of a user-defined function. So, here in weight, we are specifying a user-defined function, which is `user_weights`, which we have defined over here. It takes an input, which is array of weights. As an example, we are just returning the weight array that we receive in the argument, but it is possible that you can perform some computation over here and return an array of the same shape as the argument.

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Which **algorithm** is used to compute the nearest neighbors in **KNeighborsClassifier**?



Default:

```
1 neighbor_classifier = KNeighborsClassifier(algorithm='auto')
```

So, let us see, which algorithms are available in sklearn to compute the nearest neighbor. Because nearest neighbor computation is the most expensive operation in the KNeighborsClassifier. We can use algorithm like ball tree or kd tree or we can use a brute-force search.

So, ball tree is specified with ball _tree, kd tree specified with kd _tree and brute-force is specified with brute in the parameter algorithm. There is also an additional value that algorithm parameter takes which is auto that decides the most appropriate algorithm based on the values passed in the fit method. So, by default, the algorithm parameter is set to auto.

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Some additional parameters for **tree algorithm** in **KNeighborsClassifier**?

For 'ball_tree' and 'kd_tree' algorithms, there are some other parameters to be set.

leaf_size

- can affect the speed of the construction and query, as well as the memory required to store the tree
- default = 30

metric

- Distance metric to use for the tree
- It is either string or callable function
 - some metrics are listed below:
 - "euclidean", "manhattan", "chebyshev", "minkowski", "wminkowski", "seuclidean", "mahalanobis"
- default = 'minkowski'

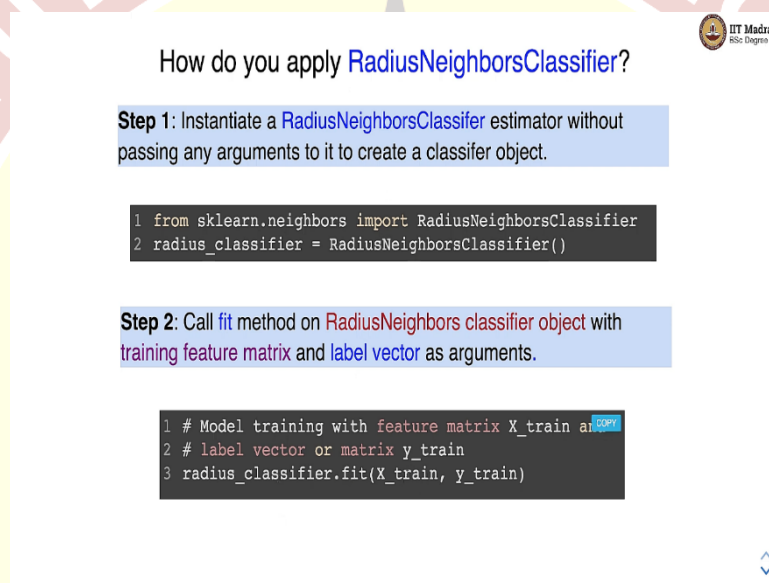
p

- Power parameter for the Minkowski metric.
- default = 2

There are some additional parameters for the tree algorithm. For `ball_tree` and `kd_tree` algorithms, there are some parameters like `leaf_size` that can affect the speed of the construction and querying the tree as well as the memory required to store the tree. The default leaf size is 30.

We also need to specify the distance metric for the tree. So, it is either string or a callable function. There are few options that are available here, which are Euclidean, Manhattan, Chebyshev, Minkowski, Mahalanobis, etc. The default one is Minkowski with the power parameter set to 2.

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How do you apply `RadiusNeighborsClassifier`?

Step 1: Instantiate a `RadiusNeighborsClassifier` estimator without passing any arguments to it to create a classifier object.

```
1 from sklearn.neighbors import RadiusNeighborsClassifier
2 radius_classifier = RadiusNeighborsClassifier()
```

Step 2: Call `fit` method on `RadiusNeighborsClassifier` object with training feature matrix and label vector as arguments.

```
1 # Model training with feature matrix X_train and
2 # label vector or matrix y_train
3 radius_classifier.fit(X_train, y_train)
```

So, that was about `KNeighborsClassifier`. Let us look at the other classifier, which is `RadiusNeighborsClassifier`. Here, if you want to use `RadiusNeighborsClassifier` in the first step, you need to instantiate radius neighbors estimator and `RadiusNeighborsClassifier` is implemented in `sklearn.neighbors` module. In the second step, we call the `fit` method on the classifier object with feature matrix and label vector as arguments.

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How do you specify the number of neighbors in
RadiusNeighborsClassifier?

- The number of neighbors is specified within a fixed radius r of each training point using **radius** parameter.
- r is a float value.

```
1 radius_classifier = RadiusNeighborsClassifier(radius=1.0)
```

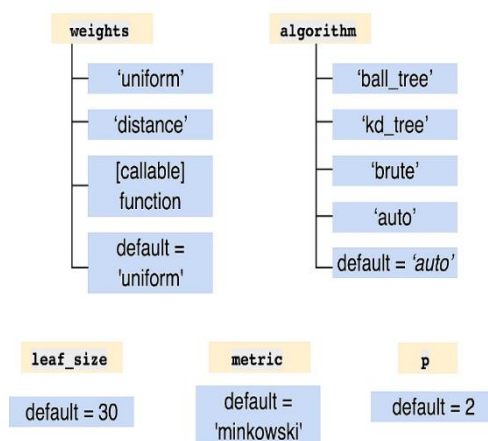
What is the default value of r ?

$r = 1.0$

Let us see, how to specify the number of neighbors in RadiusNeighborsClassifier. So, as we have seen earlier, we specify instead of specifying the neighbors, we specify the fixed radius r which is a float value. And that radius is set in the parameter radius in the constructor of RadiusNeighborsClassifier. The default value of r is 1.

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Parameters for **RadiusNeighborsClassifier**



There are some parameters for RadiusNeighborsClassifier like weights. Again, it could be uniform, distance or it could be callable function. The default weight is uniform. And again just like KNeighborsClassifier, we can use algorithms like ball_tree, kd_tree, brute-force search and auto. Auto decides the appropriate algorithm based on the values that are passed in the fit function. The default leaf size is 30 and the default matrix is Minkowski with the value

of p set to 2. So, in this video, we discussed couple of nearest neighbor classifiers implemented in sklearn.

