assignment2

October 13, 2024

```
[15]: import numpy as np
    from sklearn.datasets import make_classification
    from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import StandardScaler
    from sklearn.cluster import KMeans
```

n_centers: Number of RBF neurons in the hidden layer (i.e., the number of centers).

sigma: The spread or width of the Gaussian function, which can be set manually or calculated automatically.

centers: The RBF centers, which will be selected from the input data.

weights: The weights connecting the RBF layer to the output.

```
[16]: class RBFNetwork:
          def __init__(self, n_centers, sigma=None):
              self.n_centers = n_centers
              self.centers = None
              self.sigma = sigma
              self.weights = None
              self.bias = None
          def _gaussian(self, x, c, s):
              return np.exp(-1 / (2 * s**2) * np.linalg.norm(x - c)**2)
          def _select_centers(self, X):
              kmeans = KMeans(n_clusters=self.n_centers, random_state=42)
              kmeans.fit(X)
              self.centers = kmeans.cluster_centers_
          def _calculate_sigma(self):
              if self.sigma is None:
                  distances = np.linalg.norm(self.centers[:, np.newaxis] - self.
       ⇔centers, axis=2)
                  self.sigma = np.mean(distances)
          def fit(self, X, y):
              self._select_centers(X)
              self._calculate_sigma()
```

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G = np.zeros((X.shape[0], self.n_centers))
              for i in range(X.shape[0]):
                  for j in range(self.n_centers):
                      G[i, j] = self._gaussian(X[i], self.centers[j], self.sigma)
              G = np.hstack([G, np.ones((G.shape[0], 1))])
              self.weights = np.linalg.pinv(G).dot(y)
          def predict(self, X):
              G = np.zeros((X.shape[0], self.n_centers))
              for i in range(X.shape[0]):
                  for j in range(self.n_centers):
                      G[i, j] = self._gaussian(X[i], self.centers[j], self.sigma)
              G = np.hstack([G, np.ones((G.shape[0], 1))])
              return np.dot(G, self.weights)
     Generate synthetic data
[17]: X, y = make_classification(n_samples=1000, n_features=3, n_informative=2,__
       →n_redundant=0, random_state=42)
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
[18]: X.shape, y.shape
[18]: ((1000, 3), (1000,))
     Sample X and y
[19]: print(X[:5])
     [[ 0.96422927    1.99566749    0.24414337]
      [-1.35806186 -0.25495579 0.50289028]
      [ 1.73205679  0.26125053 -2.21417748]
      [-1.51987766 1.02370955 -0.26269143]
      [ 4.02026158    1.38145408    -1.58214341]]
[25]: print(y[:5])
     [1 0 1 0 1]
[20]: scaler = StandardScaler()
      X_train = scaler.fit_transform(X_train)
      X test = scaler.transform(X test)
[21]: n centers = 10
      rbf = RBFNetwork(n_centers=n_centers)
```

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rbf.fit(X_train, y_train)
```

Test Accuracy

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[22]: y_pred = rbf.predict(X_test)

from sklearn.metrics import accuracy_score
y_pred_binary = (y_pred > 0.5).astype(int)
accuracy = accuracy_score(y_test, y_pred_binary)
print(f'Accuracy: {accuracy:.2f}')
```

Accuracy: 0.85

Train Accuracy

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[24]: y_pred = rbf.predict(X_train)

y_pred_binary = (y_pred > 0.5).astype(int)
accuracy = accuracy_score(y_train, y_pred_binary)
print(f'Accuracy: {accuracy:.2f}')
```

Accuracy: 0.87