lab9 pca

November 15, 2022

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[]: import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import StandardScaler
     from sklearn.decomposition import PCA
     from sklearn.linear_model import LogisticRegression
     from sklearn.metrics import confusion matrix
     from sklearn import preprocessing
     from matplotlib.colors import ListedColormap
[]: data=pd.read_csv("C:\Coding\ML_python\machine-learning-lab-main\datasets\iris.
      ⇔csv")
     data.head()
[]:
       sepal_length sepal_width petal_length petal_width species
                5.1
                                            1.4
                              3.5
                                                         0.2 setosa
     0
                4.9
                                            1.4
                                                         0.2 setosa
     1
                              3.0
                4.7
                              3.2
                                           1.3
                                                         0.2 setosa
     2
     3
                4.6
                              3.1
                                           1.5
                                                         0.2 setosa
                5.0
                              3.6
                                            1.4
                                                         0.2 setosa
[]: x=data.iloc[:,0:4]
     y=data.iloc[:,4]
[]: xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size = 0.2,__
     →random_state = 0)
     sc = StandardScaler()
     xtrain = sc.fit_transform(xtrain)
     xtest = sc.transform(xtest)
[]: pca = PCA(n_components = 2)
     xtrain = pca.fit_transform(xtrain)
     xtest = pca.transform(xtest)
     explained_variance = pca.explained_variance_ratio_
```

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[]: classifier = LogisticRegression(random_state = 0)
    le = preprocessing.LabelEncoder()
    ytrain=le.fit_transform(ytrain)
    ytest=le.fit_transform(ytest)
    classifier.fit(xtrain, ytrain)
[]: LogisticRegression(random state=0)
[]: ypred = classifier.predict(xtest)
[]: cm = confusion_matrix(ytest, ypred)
    print(cm)
    [[11 0 0]
     [ 0 10 3]
     [ 0 1 5]]
[]: X_set, y_set = xtrain, ytrain
    X1,X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1,stop = X_set[:, 0].
      \rightarrowmax() + 1, step = 0.01),np.arange(start = X_{set}[:, 1].min() - 1, stop =
     \Delta X_{set}[:, 1].max() + 1, step = 0.01))
    plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).
      oreshape(X1.shape), alpha = 0.75,cmap = ListedColormap(('yellow', 'white', ∟

¬'aquamarine')))
    plt.xlim(X1.min(), X1.max())
    plt.ylim(X2.min(), X2.max())
    for i, j in enumerate(np.unique(y_set)):
        plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],c =__
      plt.title('Logistic Regression IRIS')
    plt.xlabel('PC1')
    plt.ylabel('PC2')
    plt.legend()
    plt.show()
```

c argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points. *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points.

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