# principal component analysis

November 7, 2024

# 1 Principal Component Analysis (PCA)

#### 1.1 Importing the libraries

```
[]: import numpy as np import matplotlib.pyplot as plt import pandas as pd
```

## 1.2 Importing the dataset

```
[]: dataset = pd.read_csv('Wine.csv')
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
```

### 1.3 Splitting the dataset into the Training set and Test set

## 1.4 Feature Scaling

```
[]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

# 1.5 Applying PCA

```
[]: from sklearn.decomposition import PCA
pca = PCA(n_components = 2)
X_train = pca.fit_transform(X_train)
X_test = pca.transform(X_test)
```

# 1.6 Training the Logistic Regression model on the Training set

```
[]: from sklearn.linear_model import LogisticRegression classifier = LogisticRegression(random_state = 0) classifier.fit(X_train, y_train)
```

```
[]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True, intercept_scaling=1, l1_ratio=None, max_iter=100, multi_class='auto', n_jobs=None, penalty='l2', random_state=0, solver='lbfgs', tol=0.0001, verbose=0, warm_start=False)
```

# 1.7 Making the Confusion Matrix

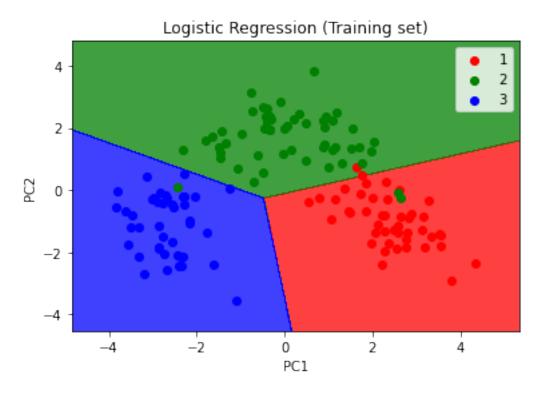
#### 1.8 Visualising the Training set results

```
[]: from matplotlib.colors import ListedColormap
     X_set, y_set = X_train, y_train
     X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, __
      0].max() + 1, step = 0.01),
                          np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, __
      41].max() + 1, step = 0.01))
     plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).
      ⇒reshape(X1.shape),
                  alpha = 0.75, cmap = ListedColormap(('red', 'green', 'blue')))
     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 = ListedColormap(('red', 'green', 'blue'))(i), label = j)
     plt.title('Logistic Regression (Training set)')
     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 a 2-D array with a single row if you really want to specify the same RGB or RGBA value for all points.

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#### 1.9 Visualising the Test set results

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