

Unsupervised Models

Use the wine data to conduct Principal Component Analysis. The URL is <https://archive.ics.uci.edu/ml/machine-learning-databases/wine/wine.data>

- Load the dataset. What are the features?
- Create a DataFrame of given wine dataset. Standardize the features.
- Split the dataset into training and test sets (30%).
- Use the `linalg.eig` function from NumPy to obtain the eigenpairs of the Wine covariance matrix. Print the eigenvalues.
- Compute and plot the individual explained variance and the cumulative explained variance (range from 1 to 14). Using the NumPy `cumsum` function, calculate the cumulative sum of explained variances, which you will then plot via Matplotlib's `step` function. Explain the outcomes.
- Sort the eigenpairs by decreasing order of the eigenvalues. Make a list of (eigenvalue, eigenvector) tuples. Sort the (eigenvalue, eigenvector) tuples from high to low. Print the projection matrix `W`.
- Using the projection matrix, transform a sample `x` onto the PCA subspace (the principal components 1 and 2). Transform the entire 124 x 13-dimensional training dataset onto the two principal components by calculating the matrix dot product.
- Visualize the transformed Wine training set into a two-dimensional scatterplot.
- Method number 2. Import `ListedColormap`. Set up marker generator and color map. ---
- Plot the decision surface and plot the class samples.
- Import `LogisticRegression` and `PCA`. Visualize the decision regions (hint: `X_train_pca`, `y_train`, `classifier=lr`).