

# Installation of Packages

First install packages like numpy, scikit-learn, matplotlib

```
In [ ]: !pip install numpy scikit-learn matplotlib
```

```
Requirement already satisfied: numpy in /home/john/contributions/.venv/lib/pytho
n3.10/site-packages (1.23.2)
Requirement already satisfied: scikit-learn in /home/john/contributions/.venv/li
b/python3.10/site-packages (1.1.2)
Requirement already satisfied: matplotlib in /home/john/contributions/.venv/lib/
python3.10/site-packages (3.6.1)
Requirement already satisfied: threadpoolctl>=2.0.0 in /home/john/contribution
s/.venv/lib/python3.10/site-packages (from scikit-learn) (3.1.0)
Requirement already satisfied: scipy>=1.3.2 in /home/john/contributions/.venv/li
b/python3.10/site-packages (from scikit-learn) (1.9.1)
Requirement already satisfied: joblib>=1.0.0 in /home/john/contributions/.venv/l
ib/python3.10/site-packages (from scikit-learn) (1.1.0)
Requirement already satisfied: python-dateutil>=2.7 in /home/john/contribution
s/.venv/lib/python3.10/site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: pyparsing>=2.2.1 in /home/john/contributions/.ven
v/lib/python3.10/site-packages (from matplotlib) (3.0.9)
Requirement already satisfied: fonttools>=4.22.0 in /home/john/contributions/.ve
nv/lib/python3.10/site-packages (from matplotlib) (4.37.4)
Requirement already satisfied: contourpy>=1.0.1 in /home/john/contributions/.ven
v/lib/python3.10/site-packages (from matplotlib) (1.0.5)
Requirement already satisfied: kiwisolver>=1.0.1 in /home/john/contributions/.ve
nv/lib/python3.10/site-packages (from matplotlib) (1.4.4)
Requirement already satisfied: pillow>=6.2.0 in /home/john/contributions/.venv/l
ib/python3.10/site-packages (from matplotlib) (9.2.0)
Requirement already satisfied: cycler>=0.10 in /home/john/contributions/.venv/li
b/python3.10/site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: packaging>=20.0 in /home/john/contributions/.ven
v/lib/python3.10/site-packages (from matplotlib) (21.3)
Requirement already satisfied: six>=1.5 in /home/john/contributions/.venv/lib/py
thon3.10/site-packages (from python-dateutil>=2.7->matplotlib) (1.16.0)
```

## Importation of packages

We import the necessary packages

```
In [ ]: import numpy as np
from sklearn.neural_network import MLPClassifier
from sklearn import datasets, metrics
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plot
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```

## Load Dataset

We load the necessary IRIS dataset.

```
In [ ]: wine = datasets.load_wine()
```

# Description of the Dataset

## Input features

```
In [ ]: wine.feature_names
```

```
Out[ ]: ['alcohol',  
        'malic_acid',  
        'ash',  
        'alcalinity_of_ash',  
        'magnesium',  
        'total_phenols',  
        'flavanoids',  
        'nonflavanoid_phenols',  
        'proanthocyanins',  
        'color_intensity',  
        'hue',  
        'od280/od315_of_diluted_wines',  
        'proline']
```

## Target feature

```
In [ ]: wine.target_names
```

```
Out[ ]: array(['class_0', 'class_1', 'class_2'], dtype='<U7')
```

## Verify number of records

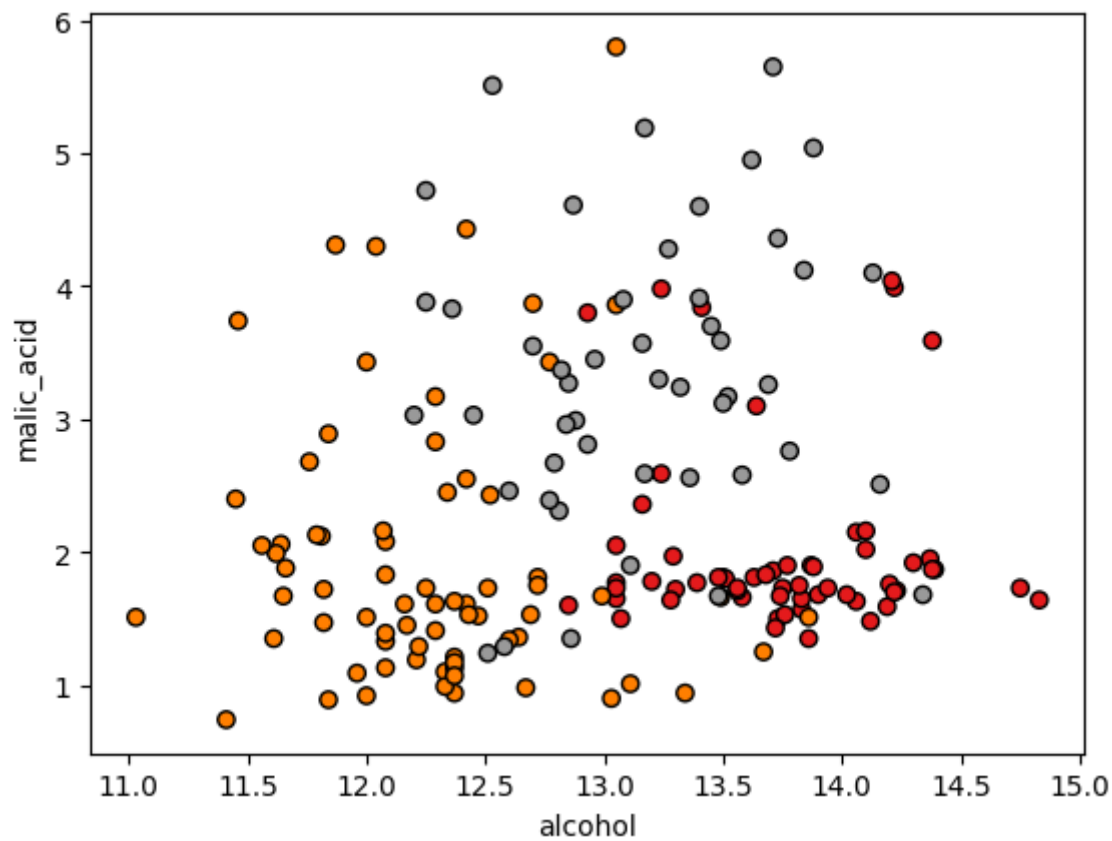
```
In [ ]: print(f"Number of Input Records: {len(wine.data)}")  
        print(f"Number of Target Records: {len(wine.target)}")
```

```
Number of Input Records: 178  
Number of Target Records: 178
```

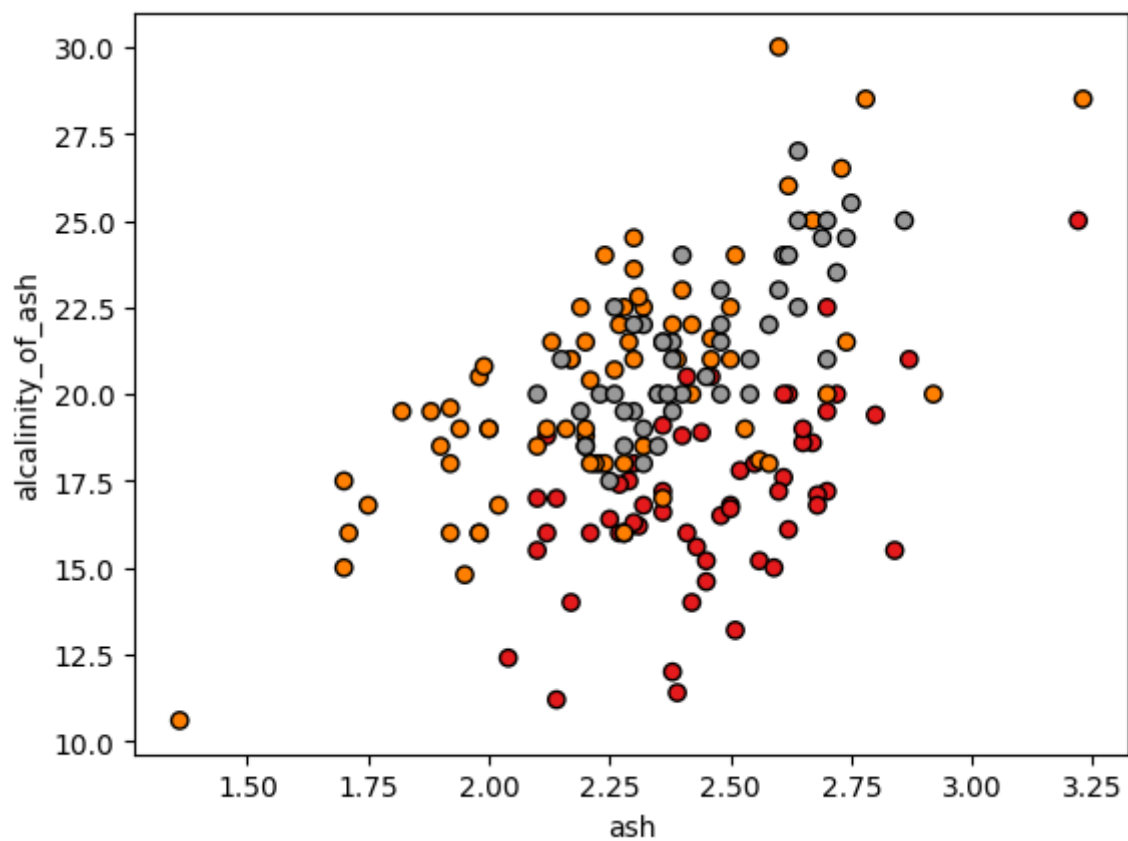
## Visulizing the dataset

```
In [ ]: x = wine.data  
        y = wine.target
```

```
In [ ]: plot.scatter(x[:, 0], x[:, 1], c=y, cmap=plot.cm.Set1, edgecolor="k")  
        plot.xlabel(wine.feature_names[0])  
        plot.ylabel(wine.feature_names[1])  
        plot.show()
```

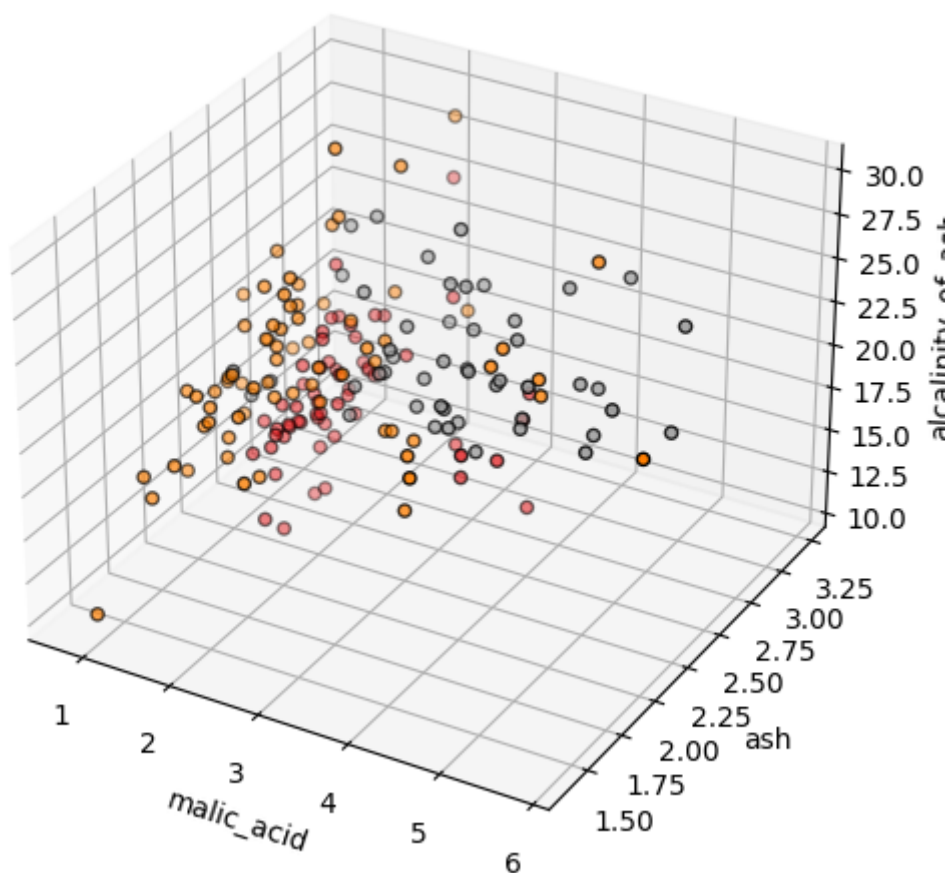


```
In [ ]: plot.scatter(x[:, 2], x[:, 3], c=y, cmap=plot.cm.Set1, edgecolor="k")
plot.xlabel(wine.feature_names[2])
plot.ylabel(wine.feature_names[3])
plot.show()
```



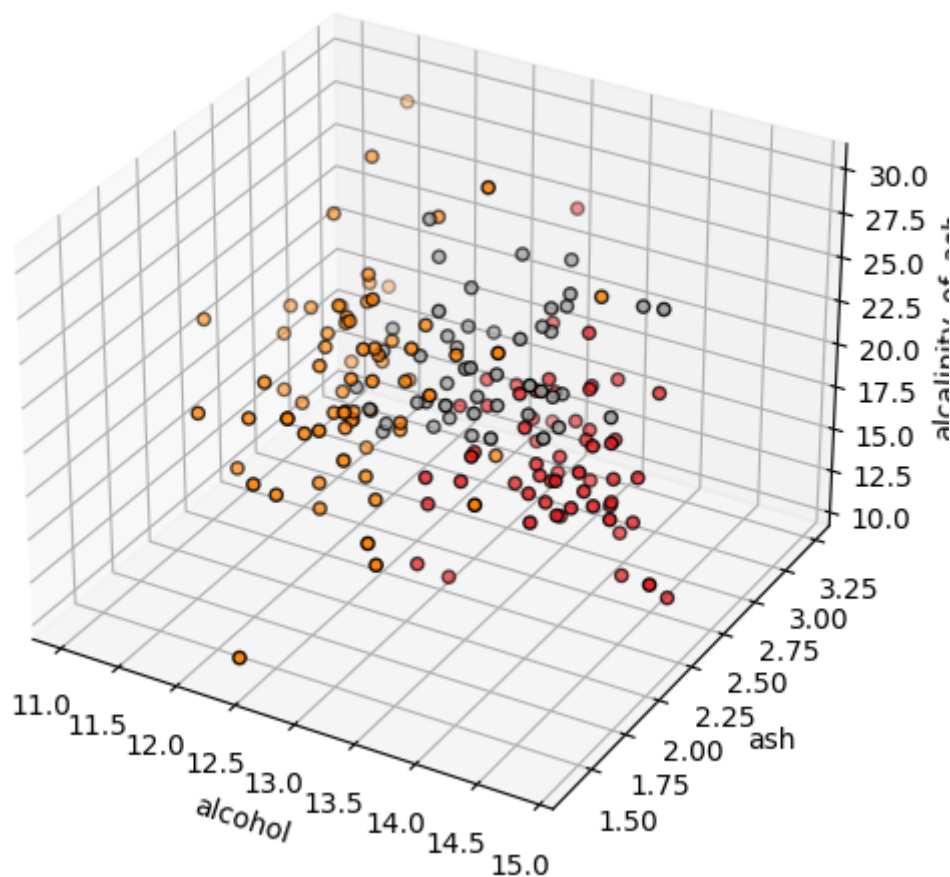
```
In [ ]: fig = plot.figure(figsize=(6, 6))
ax = fig.add_subplot(projection="3d")

ax.scatter(x[:, 1], x[:, 2], x[:, 3], c=y, cmap=plot.cm.Set1, edgecolor="k")
ax.set_xlabel(wine.feature_names[1])
ax.set_ylabel(wine.feature_names[2])
ax.set_zlabel(wine.feature_names[3])
plot.show()
```



```
In [ ]: fig = plot.figure(figsize=(6, 6))
ax = fig.add_subplot(projection="3d")

ax.scatter(x[:, 0], x[:, 2], x[:, 3], c=y, cmap=plot.cm.Set1, edgecolor="k")
ax.set_xlabel(wine.feature_names[0])
ax.set_ylabel(wine.feature_names[2])
ax.set_zlabel(wine.feature_names[3])
plot.show()
```



## Training

```
In [ ]: x = wine.data
        y = wine.target

        x_train, x_test, y_train, y_test = train_test_split(
            x, y, train_size=0.7, random_state=12, stratify=y
        )
```

```
In [ ]: print(f"Number of Training Records (input): {len(x_train)}")
        print(f"Number of Training Records (target): {len(y_train)}")

        print(f"Number of Test Records (input): {len(x_test)}")
        print(f"Number of Test Records (input): {len(x_test)}")
```

```
Number of Training Records (input): 124
Number of Training Records (target): 124
Number of Test Records (input): 54
Number of Test Records (input): 54
```

## Standardization of features

```
In [ ]: sc = StandardScaler()
        sc.fit(x_train)
        print(f"Mean: {sc.mean_} \nVariance={sc.var_}")
```

```
Mean: [1.30047581e+01 2.37379032e+00 2.35193548e+00 1.94088710e+01
       9.95161290e+01 2.29951613e+00 2.00870968e+00 3.52903226e-01
       1.58637097e+00 4.97782257e+00 9.67870968e-01 2.62653226e+00
       7.34419355e+02]
Variance=[6.14365264e-01 1.36032838e+00 7.36639958e-02 1.13516149e+01
          2.25798127e+02 3.83523959e-01 9.57067690e-01 1.61738293e-02
          3.23856991e-01 5.47239437e+00 5.11881769e-02 4.75337168e-01
          8.90159532e+04]
```

```
In [ ]: x_train_std = sc.transform(x_train)
        x_test_std = sc.transform(x_test)
```

```
In [ ]: classifier = MLPClassifier(max_iter=1000, random_state=12)

# training
classifier.fit(x_train_std, y_train)
```

```
Out[ ]: ▼                MLPClassifier
        MLPClassifier(max_iter=1000, random_state=12)
```

## Classification report

```
In [ ]: predicted_target = classifier.predict(x_test_std)

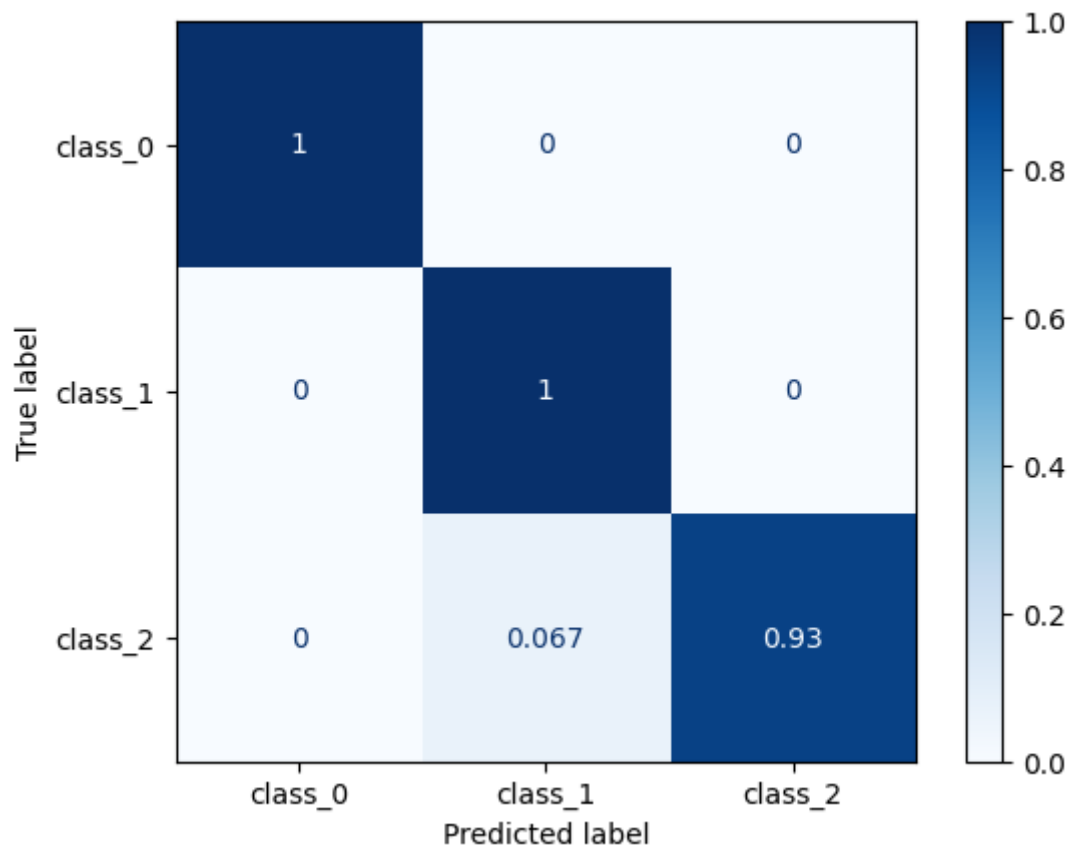
# classification report
print(metrics.classification_report(y_test, predicted_target))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	18
1	0.95	1.00	0.98	21
2	1.00	0.93	0.97	15
accuracy			0.98	54
macro avg	0.98	0.98	0.98	54
weighted avg	0.98	0.98	0.98	54

## Confusion matrix

```
In [ ]: cm = confusion_matrix(y_test, predicted_target, normalize="pred")
        disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=wine.target_names)
        disp.plot(cmap=plot.cm.Blues)
```

```
Out[ ]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x7f549f569420>
```



```
In [ ]: cm = confusion_matrix(y_test, predicted_target, normalize="true")
disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=wine.target_names)
disp.plot(cmap=plot.cm.Blues)
```

## References

- [The Iris Dataset](#)
- [3D scatterplot](#)
- [sklearn.preprocessing.StandardScaler](#)
- [sklearn.model\\_selection.train\\_test\\_split](#)
- [Iris classification with sklearn perceptron](#)
- [plot\\_confusion\\_matrix without estimator](#)
- [sklearn.neural\\_network.MLPClassifier](#)