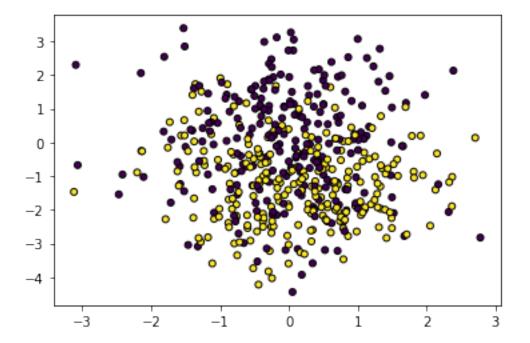
AungZarLin_lab4

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```
[1]: from sklearn.datasets import make_classification, make_blobs from sklearn.model_selection import train_test_split from sklearn. preprocessing import StandardScaler import matplotlib.pyplot as plt import numpy as np
```

[2]: <matplotlib.collections.PathCollection at 0x7fb0b246a910>



```
[3]: scaler = StandardScaler()
X = scaler.fit_transform(X)
```

```
[4]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
```

```
[5]: class Navie:
         def __init__(self,X_train, y_train):
             self.X_train = X_train
             self.y_train = y_train
             self.k = len(set(self.y_train))
         def mean_std(self):
             n = self.X_train.shape[1]
             self.mean = np.zeros((self.k, n))
             self.std = np.zeros((self.k, n))
             m = \{\}
             sum_m = 0
             for label in range(self.k):
                 self.mean[label, :] = self.X_train[self.y_train==label].mean(axis=0)
                 self.std[label, :] = self.X_train[self.y_train==label].std(axis=0)
                 m[label] = len(self.X_train[self.y_train == label])
                 sum_m += m[label]
             self.prior = {}
             for label in range(self.k):
                 self.prior[label] = m[label] / sum_m
             return self.mean, self.std
         def gaussian pdf(self,X, mean,std):
             left = 1 / (np.sqrt(2 * np.pi) * std)
             e = (X - mean) ** 2 / (2 * (std ** 2))
             right = np.exp(-e)
             self.pdf = left * right
             return self.pdf
         def fit(self, X_test, y_test):
             mean, std = self.mean_std()
             likelihood = {}
             total_likelihood = {}
             self.posterior = np.zeros((X_test.shape[0], self.k))
             for label in range(self.k):
                 likelihood[label] = self.gaussian_pdf(X_test, mean[label, :],__
      →std[label, :])
                 total_likelihood[label] = np.prod(likelihood[label], axis=1)
                 self.posterior[:,label] = self.prior[label] *__
      →total_likelihood[label]
             return self.posterior
         def accuracy(self):
             yhat = np.argmax(self.posterior, axis =1)
```

return yhat

```
[6]: model = Navie(X_train, y_train)
```

```
[7]: model.fit(X_test, y_test)
    yhat = model.accuracy()
    print(yhat)
    print(yhat.shape)
```

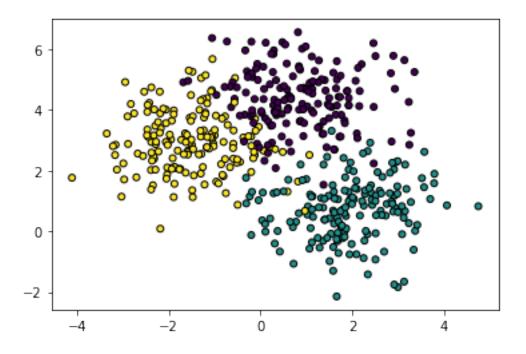
[8]: from sklearn.metrics import average_precision_score, classification_report print(average_precision_score(y_test, yhat))

0.8106740196078431

[9]: print(classification_report(y_test, yhat))

support	f1-score	recall	precision	
70	0.83	0.90	0.77	0
80	0.82	0.76	0.90	1
150	0.83			accuracy
150	0.83	0.83	0.83	macro avg
150	0.83	0.83	0.84	weighted avg

0.1 3 classes



```
model.fit(X_test, y_test)
     yhat = model.accuracy()
     print(yhat)
     print(yhat.shape)
    1 \; 2 \; 0 \; 0 \; 0 \; 2 \; 2 \; 1 \; 1 \; 1 \; 1 \; 0 \; 2 \; 2 \; 0 \; 2 \; 1 \; 0 \; 0 \; 1 \; 1 \; 2 \; 1 \; 0 \; 2 \; 2 \; 0 \; 1 \; 2 \; 1 \; 0 \; 1 \; 2 \; 0 \; 1 \; 1 \; 1
     2 2]
    (150,)
[12]: from sklearn.preprocessing import label_binarize
     from sklearn.metrics import average_precision_score
[13]: print("========Average precision score======")
     y_test_binarized = label_binarize(y_test, classes=[0, 1, 2])
     yhat binarized
                  = label_binarize(yhat, classes=[0, 1, 2])
     n_classes = len(np.unique(y_test))
     for i in range(n_classes):
        class_score = average_precision_score(y_test_binarized[:, i],__
      →yhat_binarized[:, i])
        print(f"Class {i} score: ", class_score)
```

[11]: model = Navie(X_train, y_train)

======Average precision score======

Class 0 score: 0.7981488801054017 Class 1 score: 0.9462820512820512 Class 2 score: 0.8381481481481481

[]: