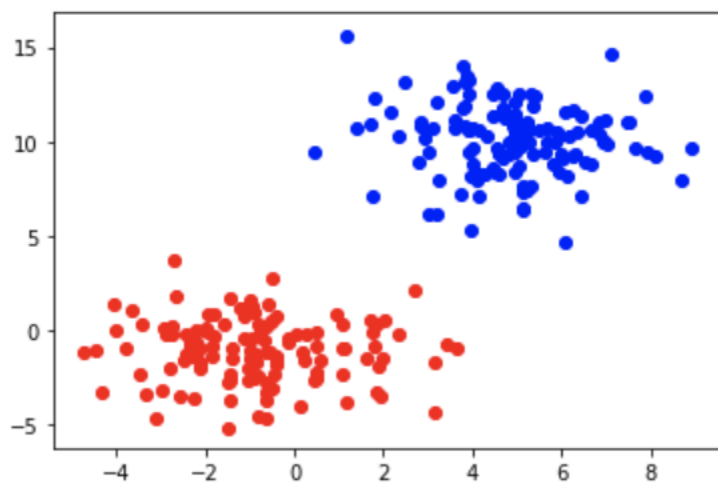
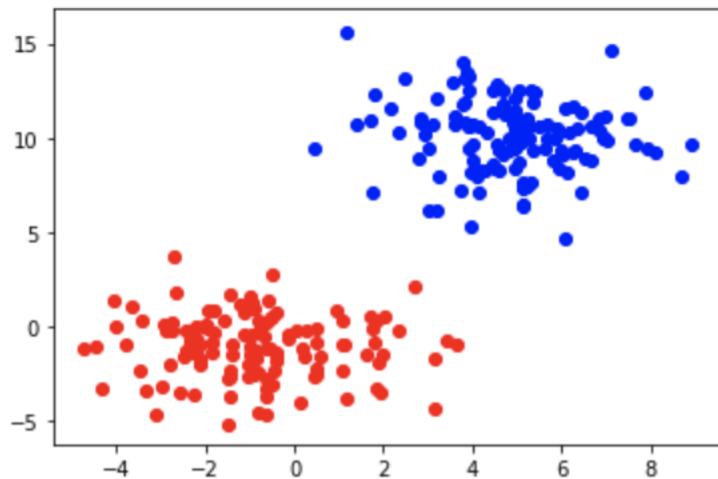


1)

Test with train()

```
w = train(X_train, y_train)
X_test_prediction = predict(X_test, w)
plot_prediction(X_test, X_test_prediction)
plot_prediction(X_test, y_test)

wrong = np.count_nonzero(y_test - X_test_prediction)
print ('Number of wrong predictions is: ' + str(wrong))
```

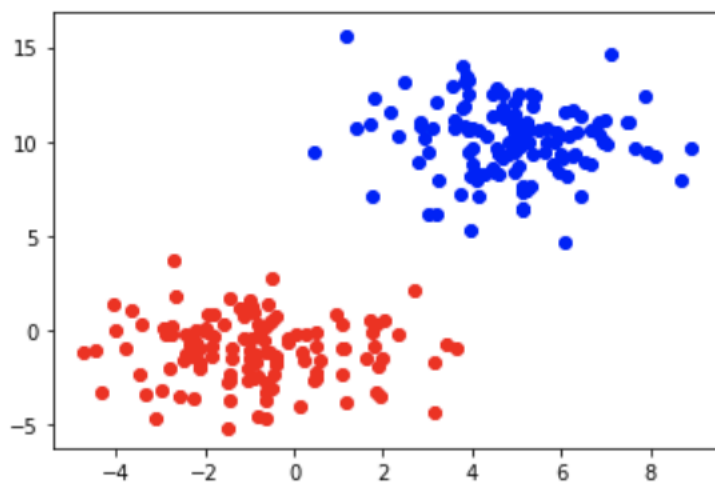
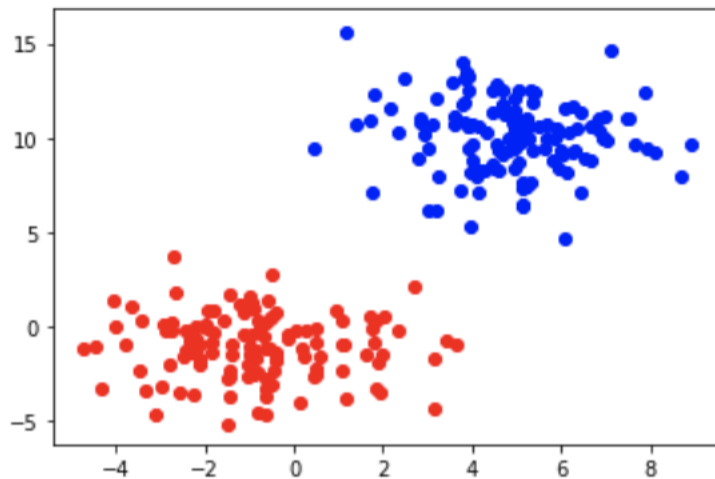


Number of wrong predictions is: 0

Test with train\_matrix()

```
w = train_matrix(X_train, y_train)
X_test_prediction = predict(X_test, w)
plot_prediction(X_test, X_test_prediction)
plot_prediction(X_test, y_test)

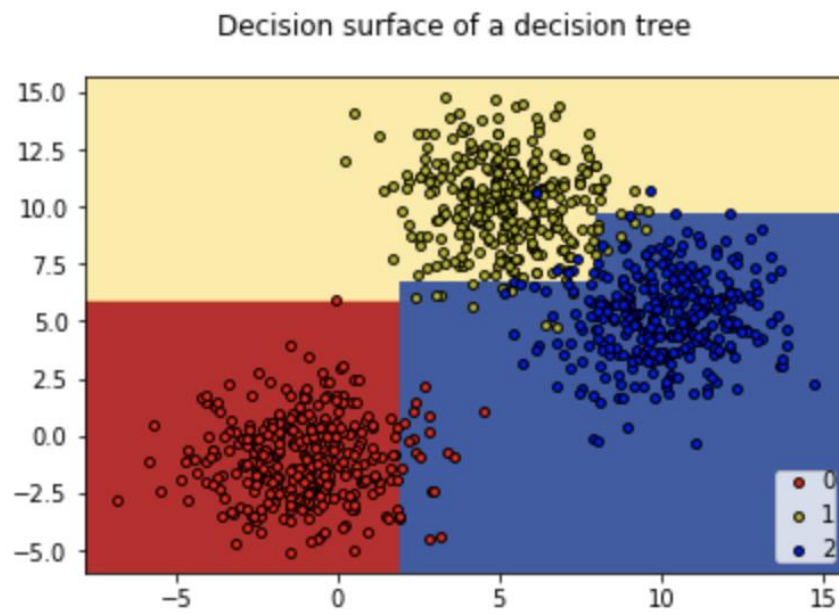
wrong = np.count_nonzero(y_test - X_test_prediction)
print ( 'Number of wrong predictions is: ' + str(wrong))
```



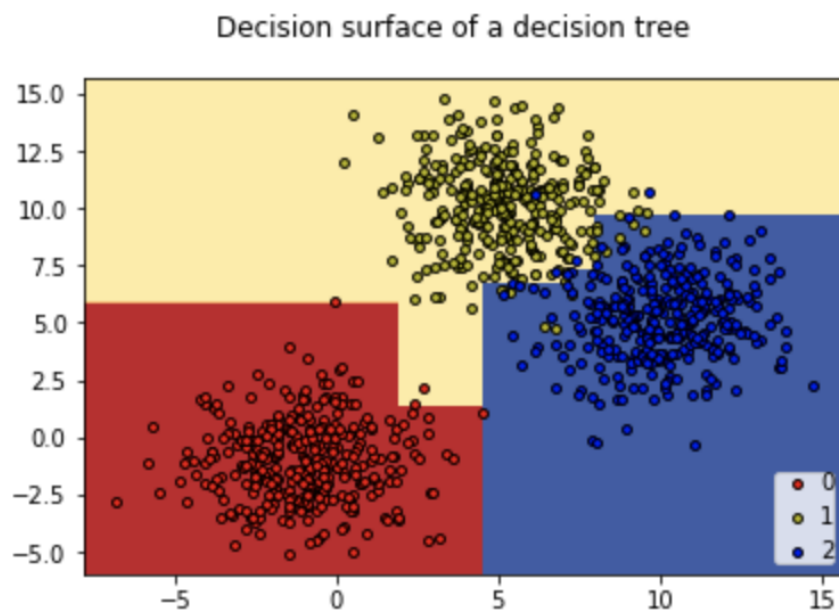
Number of wrong predictions is: 0

2)

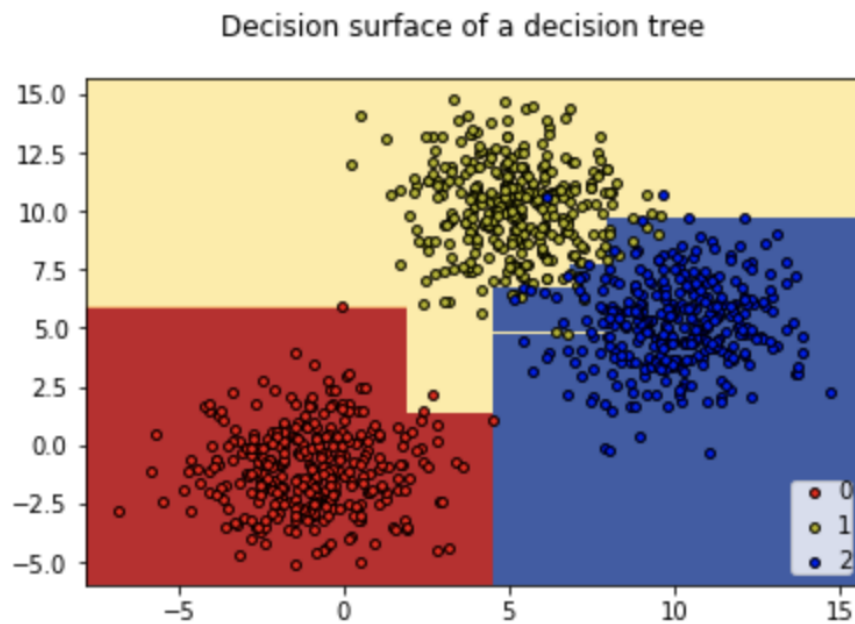
max\_dept set to 3:



max\_dept set to 5:



max\_dept set to 7:



Number of wrong predictions is: 8

Explanation:

The number of wrong predictions reduces when the value of max\_dept increases. This is because of the reduction in underfitting and reduce the bias while increase the variance for classification. Therefore, the number from max\_dept equal to 3 increasing to 5 can reduce the number of wrong predictions (decrease from 15 to 8). However, when the max\_dept allowed is further increased, the model start to overfit and we can see from the graph there is a yellow gap inside the blue region but that region which is initialized with center (5,10) should be classified as blue and the yellow instances are outliers. Thus, the starting of overfitting stop the accuracy from increasing and therefore there is no further improvement on the number of wrong predictions (i.e. still equal to 8).