

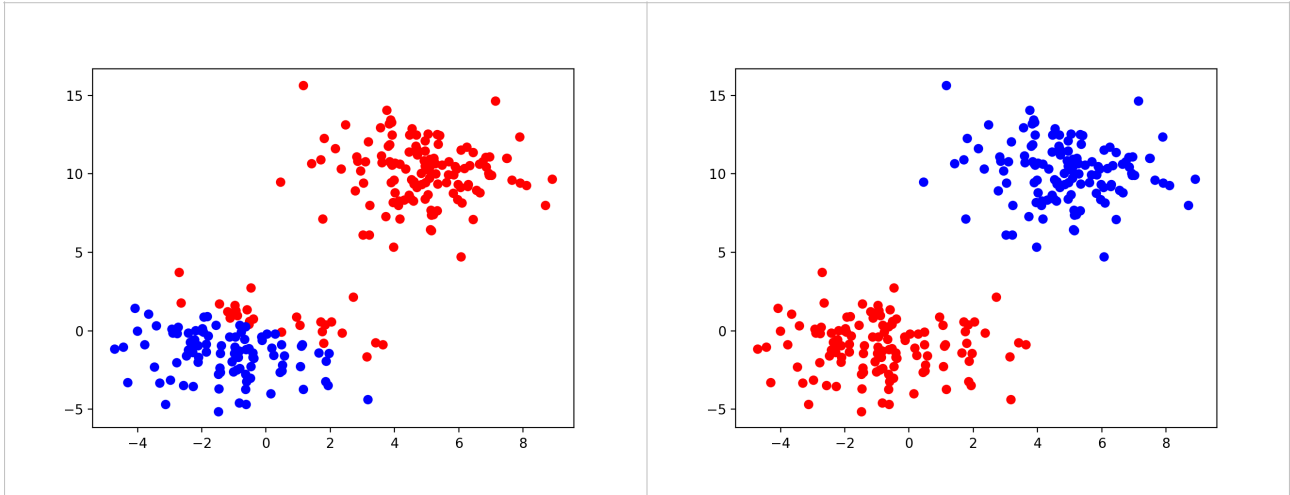
# Assignment2

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1.

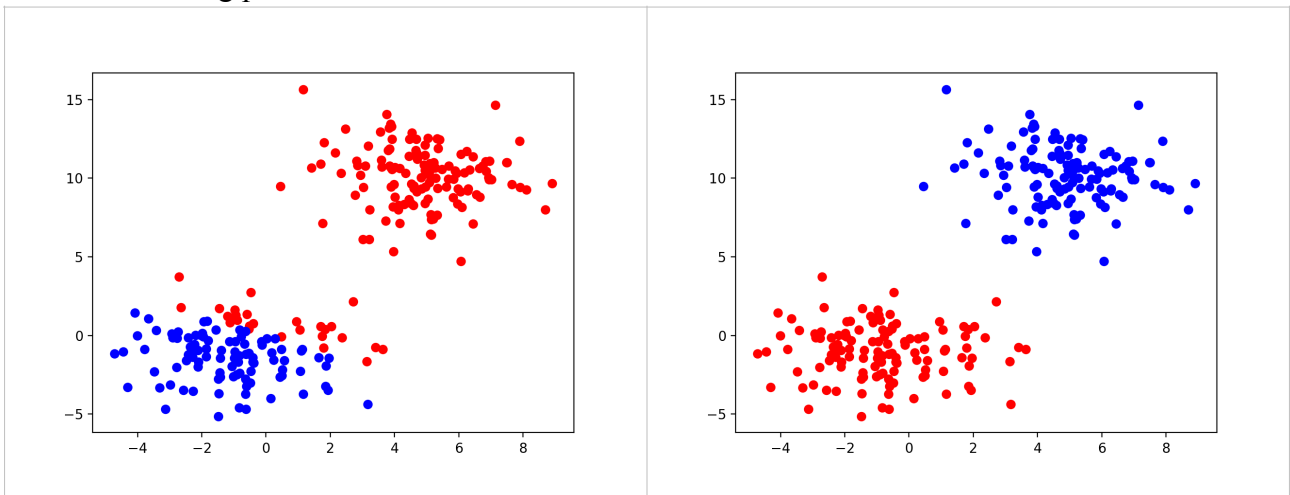
a. `train()`

Number of wrong predictions is: 27

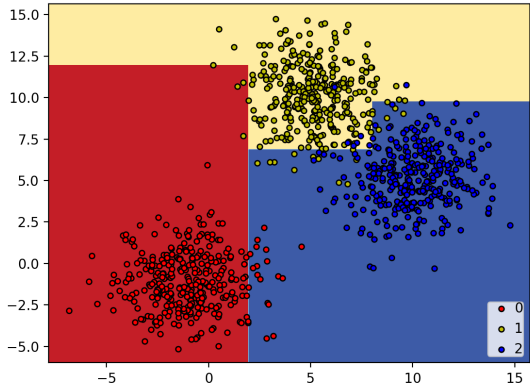
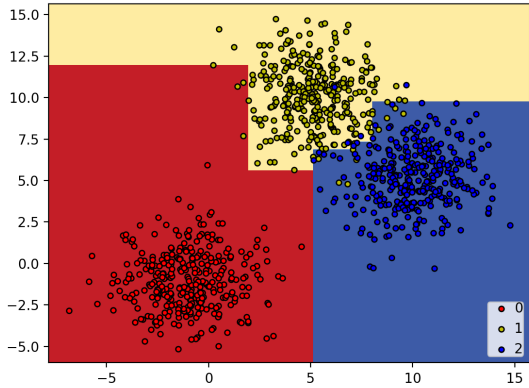
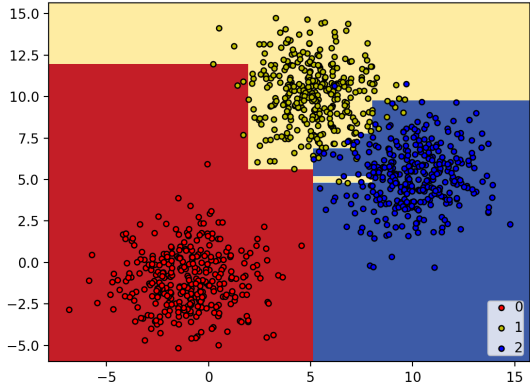


b. `train_matrix()`

Number of wrong predictions is: 27



2.

max_depth = 3	max_depth = 5
Number of wrong predictions is: 18	Number of wrong predictions is: 9
<p>Decision surface of a decision tree</p> 	<p>Decision surface of a decision tree</p> 
max_depth = 7	
Number of wrong predictions is: 10	
<p>Decision surface of a decision tree</p> 	

When the max\_depth is 5, it gives the minimum number of wrong predictions. Given max\_depth is 3, the error number is 18. And it drops to 9 when we add the max\_depth up to 5. But if we set the max\_depth higher, that is, to 7, the number of error growth to 10.

This may because when the decision tree that is too small (max\_depth is 3) is susceptible to underfitting, that is, the tree's complexity is smaller than the underlying of data. Here, as we need to increase the complexity of decision tree, the number of wrong prediction decreases. But if the decision tree is too complex (max\_depth is 7), the data is not enough to constrain it, which is overfitting. In such a case, we do not have a good generalization. In above examples, the case max\_depth = 5 makes a good balance between the complexity of decision tree and generalization of model.