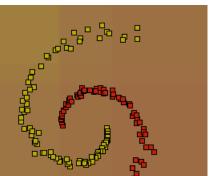
Classification problem for experiment-5:

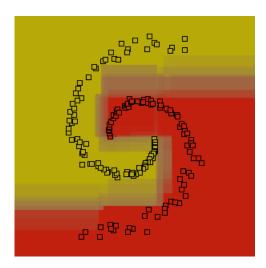
a) Varying tree depth(D).

Two classes:

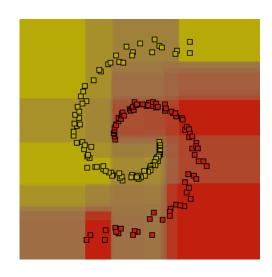
D = 2



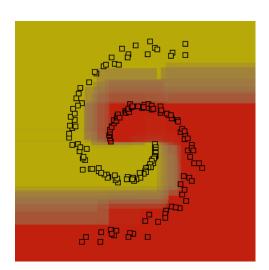
D = 2



D = 5



D = 20

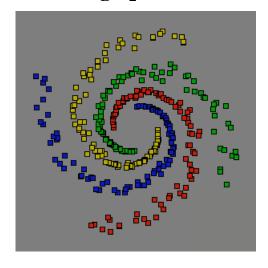


On increasing tree depth, the performance of the forest improves as shorter trees(with lesser D) give less precise information as compared to longer trees.

a)Varying tree depth(D).

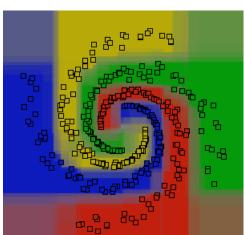
Four Classes:

D = 2

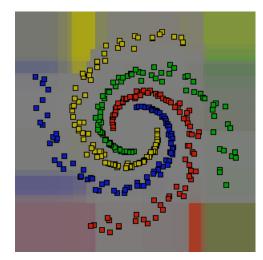


D = 10

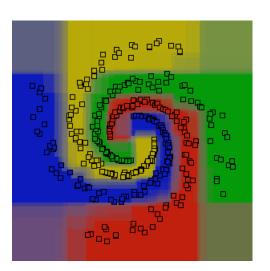




D = 5



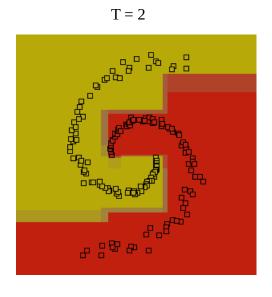
D = 20



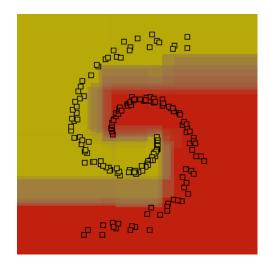
Similarly here, the performance of the model improves significantly upto D = 10 and then slowly reaches convergence.

b) Varying number of trees(T) in the forest.

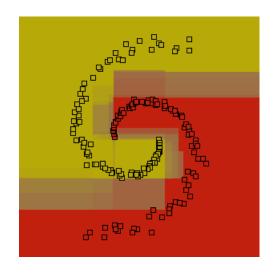
Two Classes:



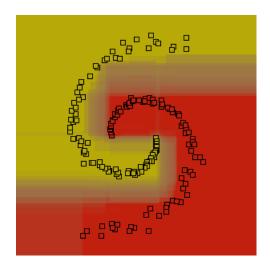
$$T = 10$$



T = 5



$$T = 20$$



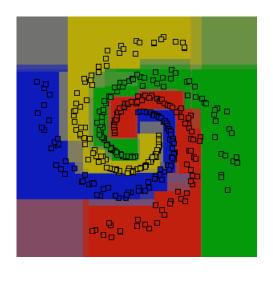
Increasing the number of trees seems to improve the accuracy of the model.

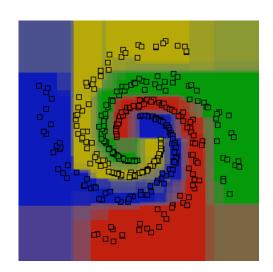
b) Varying number of trees (T) in the forest.

Four Classes:

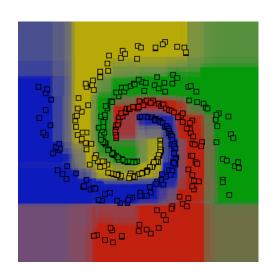
T = 2

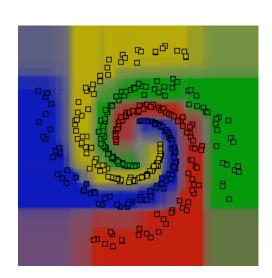
T = 5





T = 10 T = 20





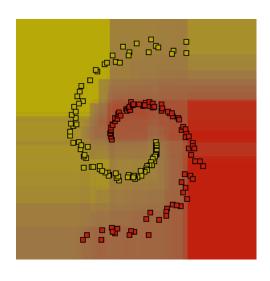
Increasing the number of trees has improved the accuracy of the model here as well.

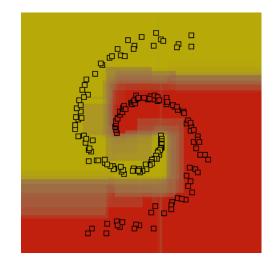
c)Varying number of candidate feature response functions(F) per split node.

Two Classes:

F = 2

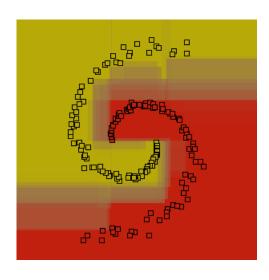


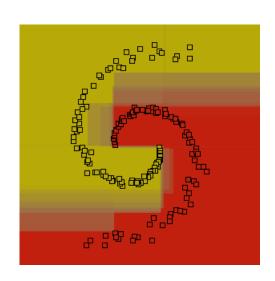




$$F = 10$$

$$F = 50$$

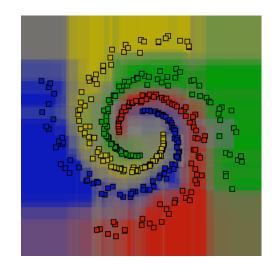


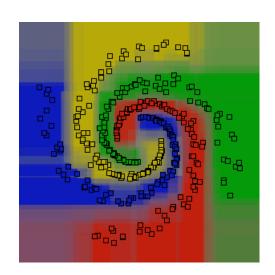


c)Varying number of candidate feature response functions(F) per split node.

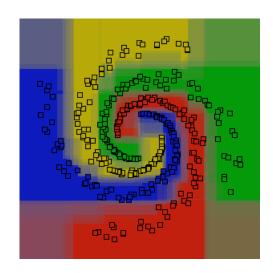
Four Classes:

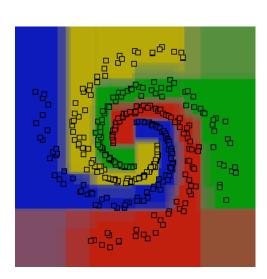
F = 2 F = 5





F = 20 F = 50





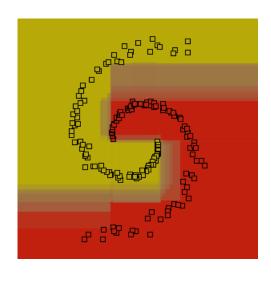
For both, 2 classes and 4 classes, the randomness of the model reduces.

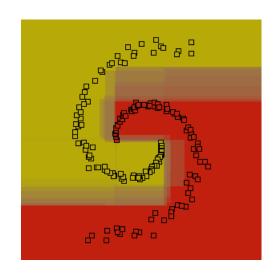
d)Varying number of candidate thresholds per feature response(L) function

Two Classes:

L = 2

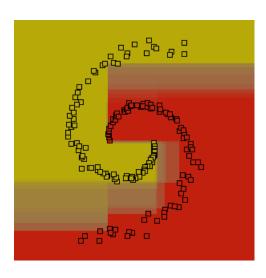
L = 5

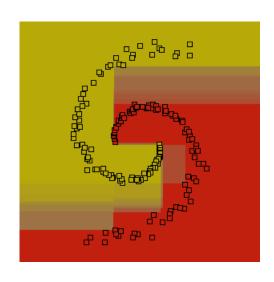




L = 20

L = 50



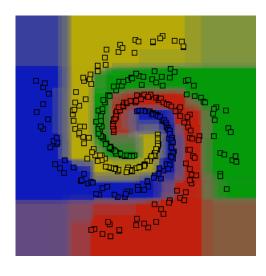


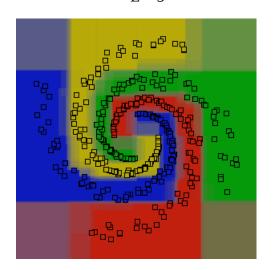
d)Varying number of candidate thresholds per feature response(L) function

Four Classes:

L = 2

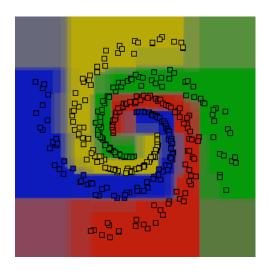


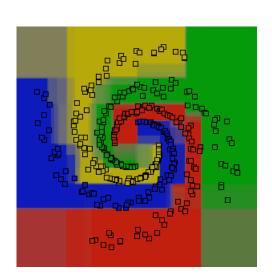




L = 20

$$L = 50$$





For both, 2 classes and 4 classes, the model confidence increases.