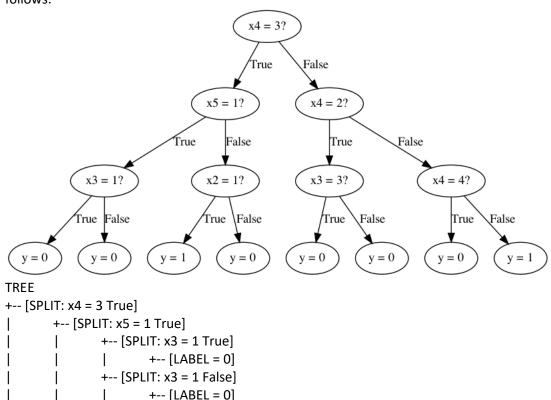
CS6375 Assignment 2 Report

Name: Haoda LE Netid: HXL180046

a. The python code is in another file $decision_tree.py$

the running result with default setting (dataset: monks-1.train, monks-1.test; max_depth = 3) is as follows:



+-- [LABEL = 0] +-- [SPLIT: x5 = 1 False] +-- [SPLIT: x2 = 1 True] +-- [LABEL = 1] +-- [SPLIT: x2 = 1 False] +-- [LABEL = 0] +-- [SPLIT: x4 = 3 False] +-- [SPLIT: x4 = 2 True] +-- [SPLIT: x3 = 3 True] +-- [LABEL = 0] +-- [SPLIT: x3 = 3 False] +-- [LABEL = 0] +-- [SPLIT: x4 = 2 False] +-- [SPLIT: x4 = 4 True] +-- [LABEL = 0] +-- [SPLIT: x4 = 4 False] +-- [LABEL = 1]

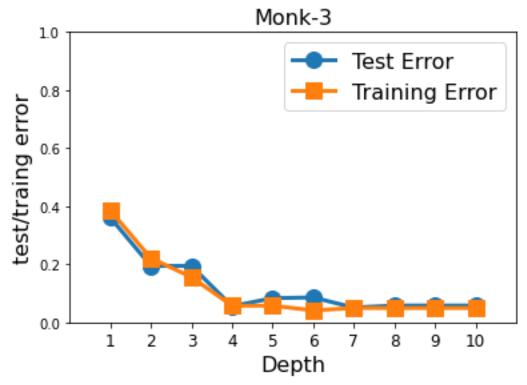
Test Error = 27.08%. Train Error = 24.19%.



for Monk-1 problem, training/test error shows above. Both training and test error rate declines to depth 3, then test error increase, which may have over fitting.

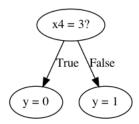


for Monk-2 problem, training/test error shows above. After depth 5, both training and test error remains low.



for Monk-3 problem, training/test error shows above. After depth 7, both training and test error remains low.

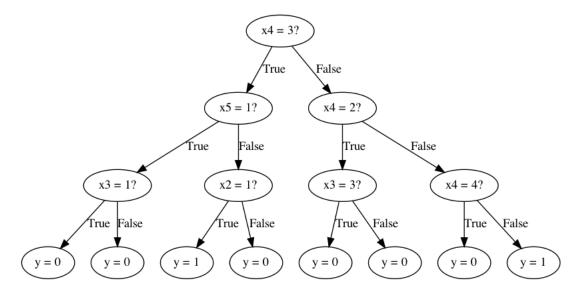
c. for monks-1 problem, id3 algorithm, max_depth 1:



depth:1 confusion matrix [[72 144]

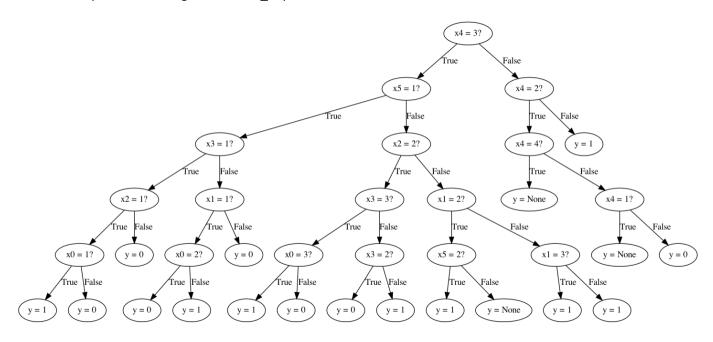
[36 180]]

for monks-1 problem, id3 algorithm, max_depth 3:



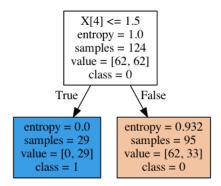
depth:3 confusion matrix [[198 18] [99 117]]

for monks-1 problem, id3 algorithm, max_depth 5:



depth:5 confusion matrix [[112 104] [54 162]]

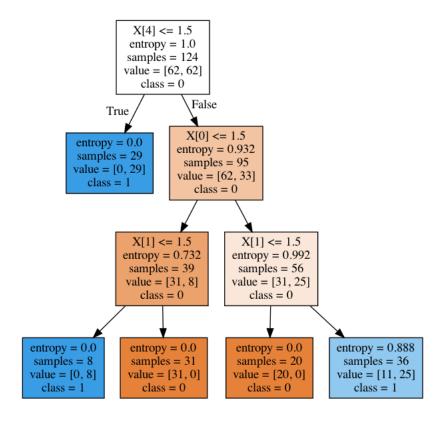
d. for monks-1 problem, scikit-learn's algorithm, max_depth 1:



depth:1 confusion matrix

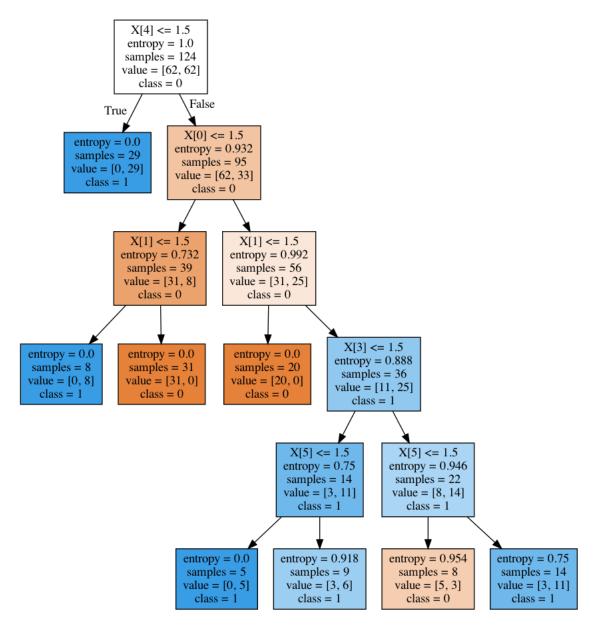
[[216 0] [108 108]]

for monks-1 problem, scikit-learn's algorithm, max_depth 3:



depth:3 confusion matrix

[[144 72] [0 216]] for monks-1 problem, scikit-learn's algorithm, max_depth 5:



depth:5 confusion matrix [[168 48] [24 192]] e.

Use other data sets in the UCI repository, "Iris Data Set". It has 4 features, totally 150 instances. so 120 instances used for training, and 30 instances for test.

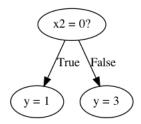
Totally 3 target classes.

class:

#-- 1 : Iris Setosa #-- 2 : Iris Versicolour #-- 3 : Iris Virginica

since it is continuous features, use a simple discretization strategy to pre-process them into binary features. x <= mean : 0; x > mean : 1

id3 algorithm, max_depth 1:



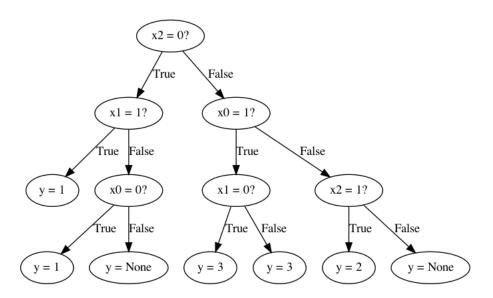
id3, depth:1 confusion matrix

[[10 0 0]

[208]

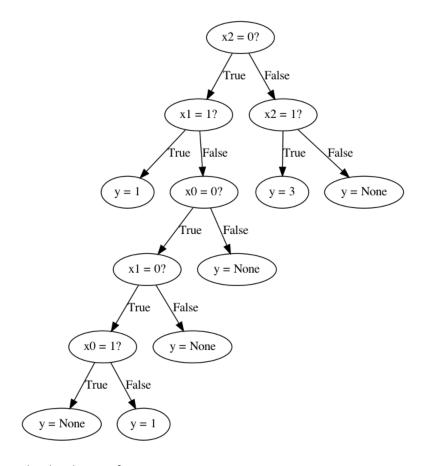
[0 0 10]]

id3 algorithm, max_depth 3:



id3, depth:3 confusion matrix [[10 0 0] [2 6 2] [0 1 9]]

id3 algorithm, max_depth 5:



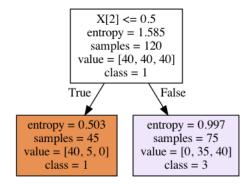
id3, depth:5 confusion matrix

[[10 0 0]

[208]

[0 0 10]]

scikit-learn's algorithm, max_depth 1:



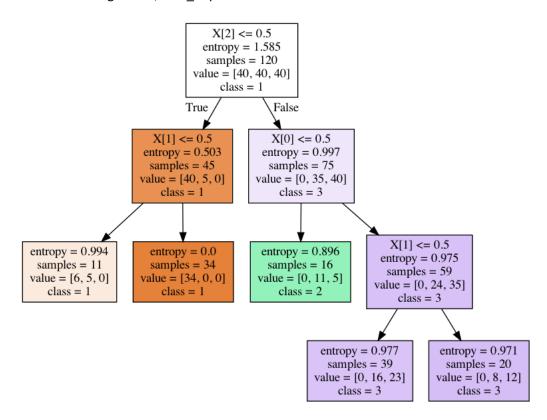
sk, depth:1 confusion matrix

[[10 0 0]

[208]

[0 0 10]]

scikit-learn's algorithm, max_depth 3:



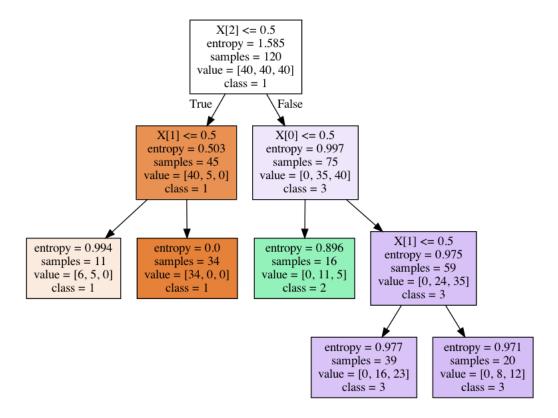
sk, depth:3 confusion matrix

[[10 0 0]

[2 6 2]

[0 1 9]]

scikit-learn's algorithm, max_depth 5:



sk, depth:5 confusion matrix

[[10 0 0]

[2 6 2]

[0 1 9]]

Confusion matrix comparison:

Max_depth	id3 confusion matrix	id3 accuracy	Scikit-learn confusion matrix	Sk accuracy
1	[[10 0 0]	0.667	[[10 0 0]	0.667
	[208]		[208]	
	[0 0 10]]		[0 0 10]]	
3	[[10 0 0]	0.833	[[10 0 0]	0.883
	[262]		[262]	
	[0 1 9]]		[0 1 9]]	
5	[[10 0 0]	0.667	[[10 0 0]	0.883
	[208]		[262]	
	[0 0 10]]		[0 1 9]]	

Based on the limited number of training and test data instance, as we can see from the result, on and before max_depth 3, my implemented id3 algorithm has the exact same confusion matrix and accuracy

as scikit-learn algorithm. When max_depth set to 5, my id3 algorithm will make the decision tree deeper, which reduces the accuracy.

While for scikit-learn algorithm, as we can see, the max_depth = 3 has the exact same decision tree as max_depth = 5, because decision tree with depth 3 already has high accuracy, make the tree deeper will reduce the accuracy.

So generally speaking, scikit-learn algorihtm is a little bit better than my implemented id3 algorithm.