



DWDM Assignment 03 Presentation

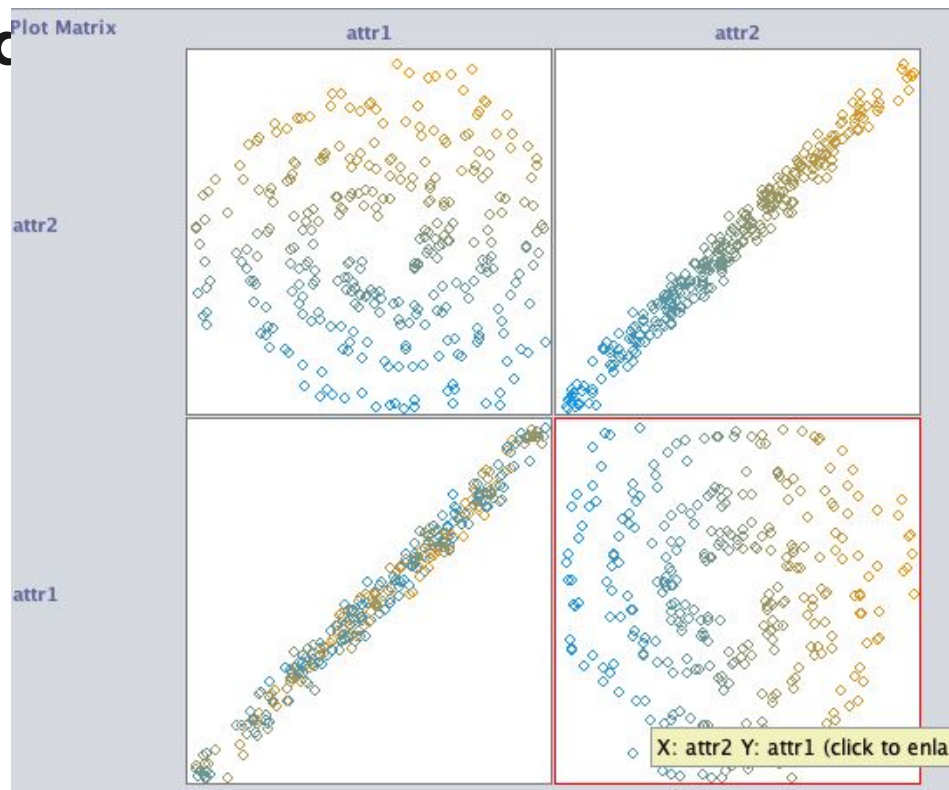
Qinyang Wu, st174540@stud.Uni-Stuttgart.de, 3519174

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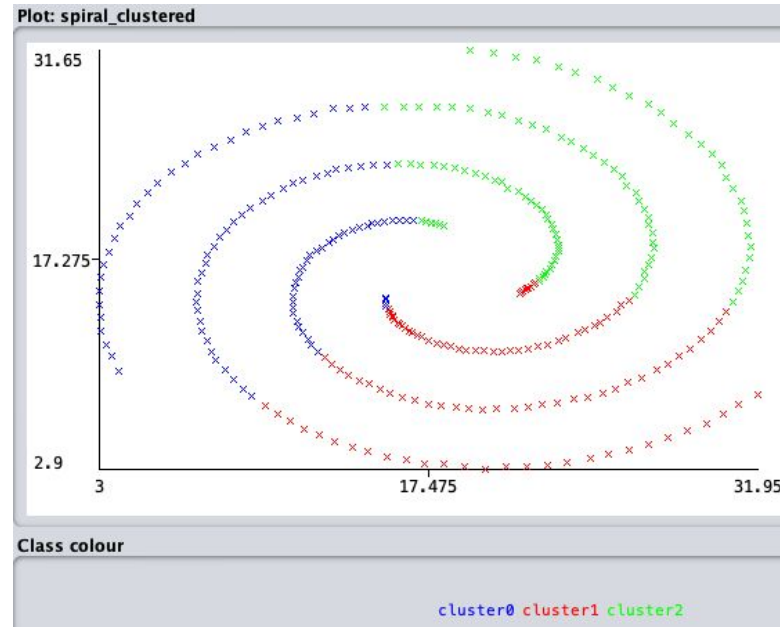
Kuang-Yu Li, st169971@stud.uni-stuttgart.de, 3440829

Task1.3: Explore Clustering Algorithms

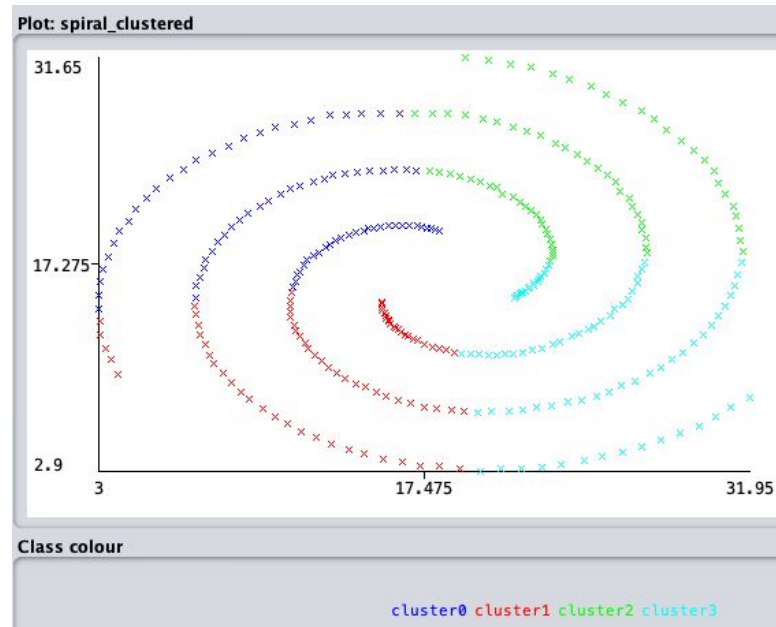
Spiral.csv



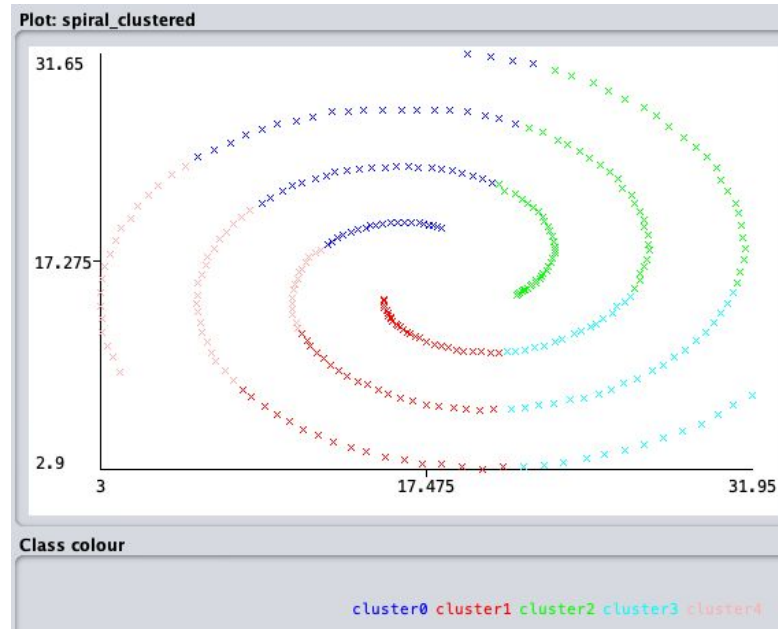
SimpleKMeans: values using k-means++ K = 3



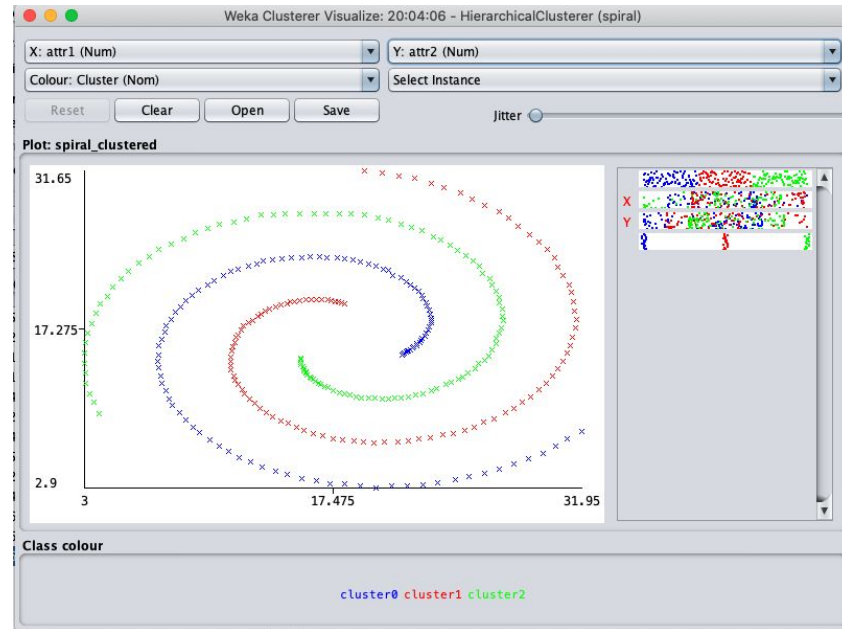
SimpleKMeans: values using k-means++ K = 4



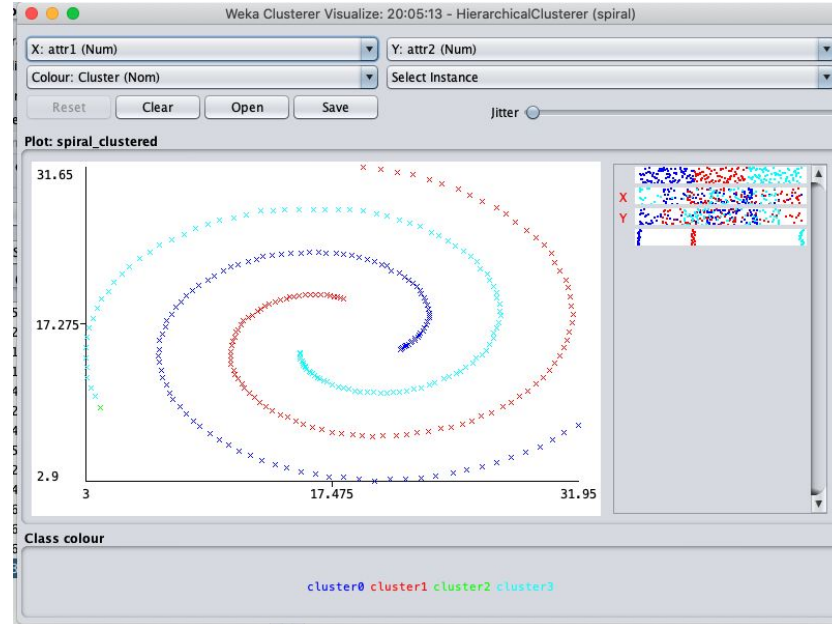
SimpleKMeans: values using k-means++ K = 5



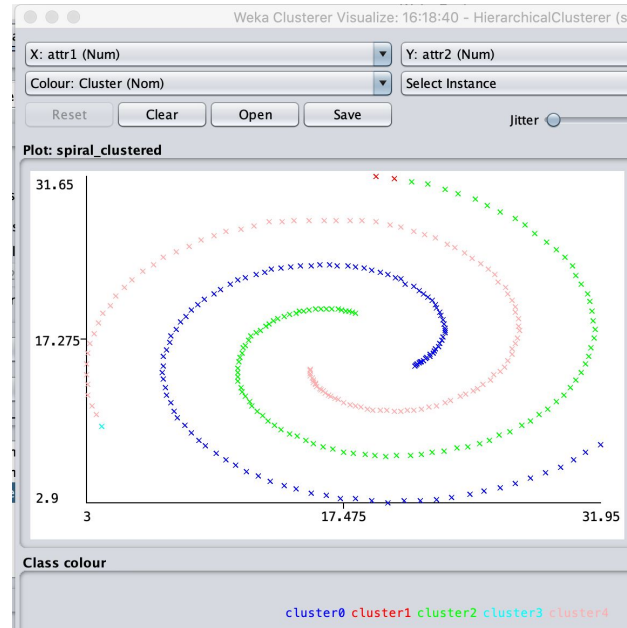
HierachyCluster:



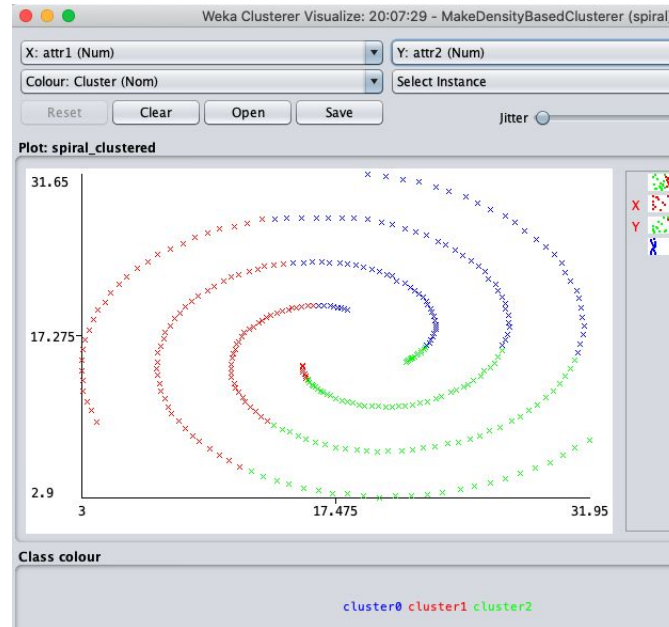
HierachyCluster:



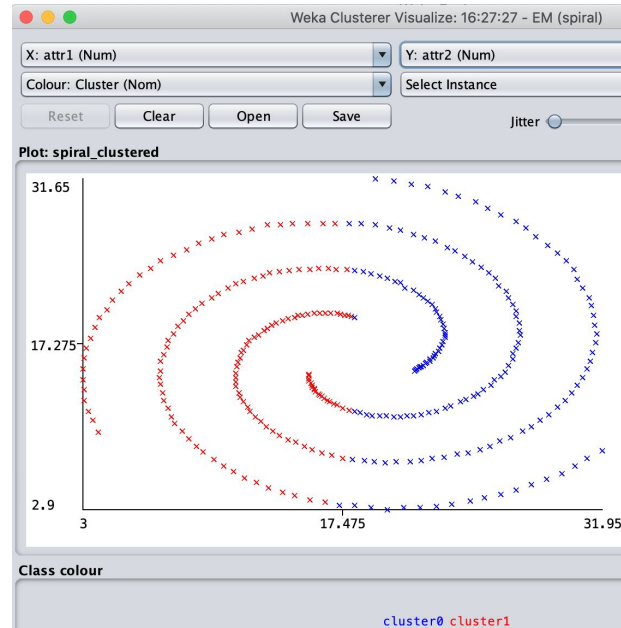
DensityCluster K = 5



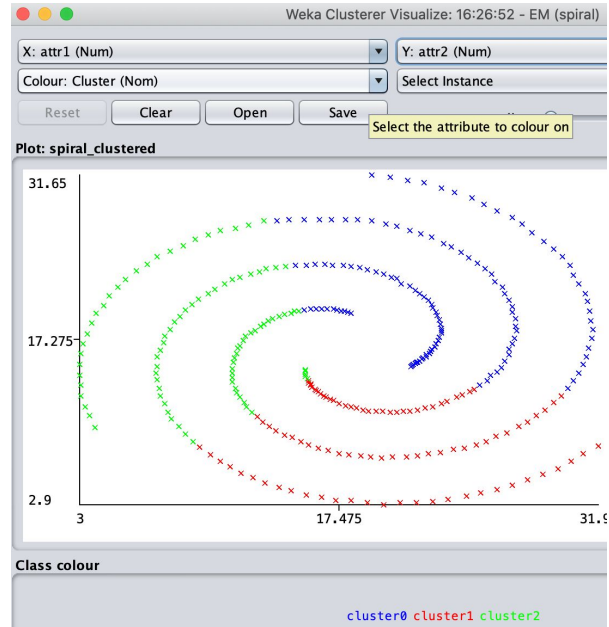
DensityCluster K = 3



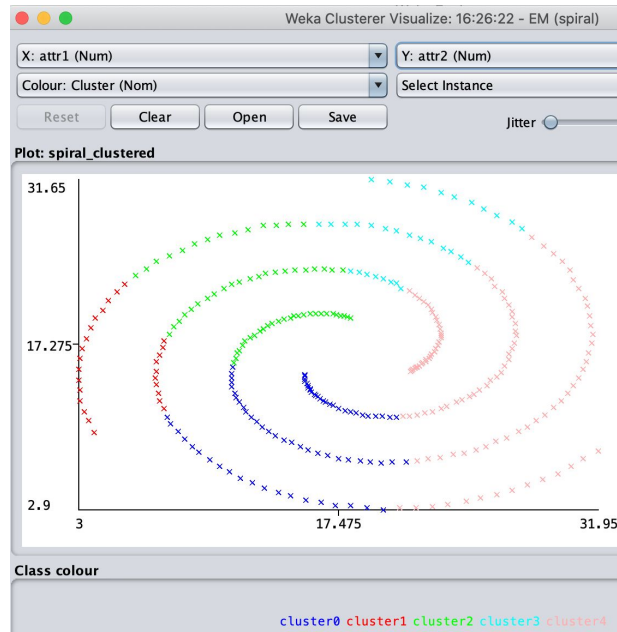
EM



EM



EM





Running HierachyCluster with Small Guasian takes longer and
memory explode



What is your conclusion: Is it a good idea to always run KMeans as clustering algorithm and with the same parameter setting for all datasets? Explain your answer.

Not Suitable to run KMeans on all datasets

Task2.3 Explore different algorithms and parameters



Observation: Apriori to upperBoundMinSupport to 0.1

Runtime: much longer


Frequency of best rules: every time

Cycles: 18 Minimum metric <confidence>: 0.9

Number of cycles performed: 18 vs 11 (prev.)

Best rules found:

1. water-project-cost-sharing=n physician-fee-freeze=y export-administration-act-south-africa=y 44 ==> crime=y 44 <conf:(1)> lift:(1.75) lev:(0.04)
2. water-project-cost-sharing=n physician-fee-freeze=y export-administration-act-south-africa=y 44 ==> Class=republican 44 <conf:(1)> lift:(2.59)
3. water-project-cost-sharing=y religious-groups-in-schools=n superfund-right-to-sue=n 44 ==> aid-to-nicaraguan-contras=y 44 <conf:(1)> lift:(1.8)
4. water-project-cost-sharing=y religious-groups-in-schools=n crime=n 44 ==> aid-to-nicaraguan-contras=y 44 <conf:(1)> lift:(1.8) lev:(0.04) [19]
5. water-project-cost-sharing=y religious-groups-in-schools=n crime=n 44 ==> Class=democrat 44 <conf:(1)> lift:(1.63) lev:(0.04) [16] conv:(16.99)
6. education-spending=y export-administration-act-south-africa=n Class=republican 44 ==> adoption-of-the-budget-resolution=n 44 <conf:(1)> lift:(2.46)
7. religious-groups-in-schools=n immigration=y crime=n 44 ==> physician-fee-freeze=n 44 <conf:(1)> lift:(1.76) lev:(0.04) [19] conv:(19.02)
8. physician-fee-freeze=n religious-groups-in-schools=y superfund-right-to-sue=y 44 ==> Class=democrat 44 <conf:(1)> lift:(1.63) lev:(0.04) [16] conv:(16.99)
9. education-spending=y duty-free-exports=n export-administration-act-south-africa=n 44 ==> physician-fee-freeze=y 44 <conf:(1)> lift:(2.46) lev:(0.04)
10. education-spending=y export-administration-act-south-africa=n Class=republican 44 ==> physician-fee-freeze=y 44 <conf:(1)> lift:(2.46) lev:(0.04)



Observation: FPGrowth upperBoundMinSupport and lowerBoundMinSupport to **0.1**.

Runtime: Much faster

40 rules found

FPGrowth found 40 rules (displaying top 10)

```
1. [crime=y, immigration=y, el-salvador-aid=y, water-project-cost-sharing=y]: 44 ==> [religious-groups-in-schools=y]: 44 <conf:(1)> lift:(1.6) lev:(0.04) conv:(16.49)
2. [adoption-of-the-budget-resolution=y, mx-missile=y, water-project-cost-sharing=y, handicapped-infants=y]: 44 ==> [aid-to-nicaraguan-contras=y]: 44 <conf:(1)> lift:(1.8) lev:(0.04) conv:(19.52)
3. [export-administration-act-south-africa=y, crime=y, immigration=y, education-spending=y, Class=republican]: 44 ==> [physician-fee-freeze=y]: 44 <conf:(1)> lift:(2.46) lev:(0.06) conv:(26.1)
4. [export-administration-act-south-africa=y, immigration=y, el-salvador-aid=y, education-spending=y, Class=republican]: 44 ==> [physician-fee-freeze=y]: 44 <conf:(1)> lift:(2.46) lev:(0.06) conv:(26.1)
5. [immigration=y, el-salvador-aid=y, superfund-right-to-sue=y, water-project-cost-sharing=y]: 45 ==> [religious-groups-in-schools=y]: 44 <conf:(0.98)> lift:(1.56) lev:(0.04) conv:(8.43)
6. [export-administration-act-south-africa=y, immigration=y, education-spending=y, Class=republican]: 45 ==> [crime=y]: 44 <conf:(0.98)> lift:(1.72) lev:(0.04) conv:(9.67)
7. [export-administration-act-south-africa=y, immigration=y, education-spending=y, Class=republican]: 45 ==> [el-salvador-aid=y]: 44 <conf:(0.98)> lift:(2.01) lev:(0.05) conv:(11.53)
8. [religious-groups-in-schools=y, export-administration-act-south-africa=y, crime=y, immigration=y, education-spending=y]: 45 ==> [el-salvador-aid=y]: 44 <conf:(0.98)> lift:(2.01) lev:(0.05) conv:(11.53)
9. [religious-groups-in-schools=y, export-administration-act-south-africa=y, immigration=y, el-salvador-aid=y, education-spending=y]: 45 ==> [crime=y]: 44 <conf:(0.98)> lift:(1.72) lev:(0.04) conv:(9.67)
10. [export-administration-act-south-africa=y, crime=y, immigration=y, physician-fee-freeze=y, education-spending=y]: 45 ==> [el-salvador-aid=y]: 44 <conf:(0.98)> lift:(2.01) lev:(0.05) conv:(11.53)
```



Observation: FPGrowth setting upperBoundMinSupport
and lowerBoundMinSupport to **1.0**

No rules found!



Three different settings of metricType and minMetric combinations.

lift / confidence ↑ (higher)

rules ↓ (fewer)

run-time ↓ (faster)

Aprior often crash with confidence < 0.9 and supportMin = Max = 0.1

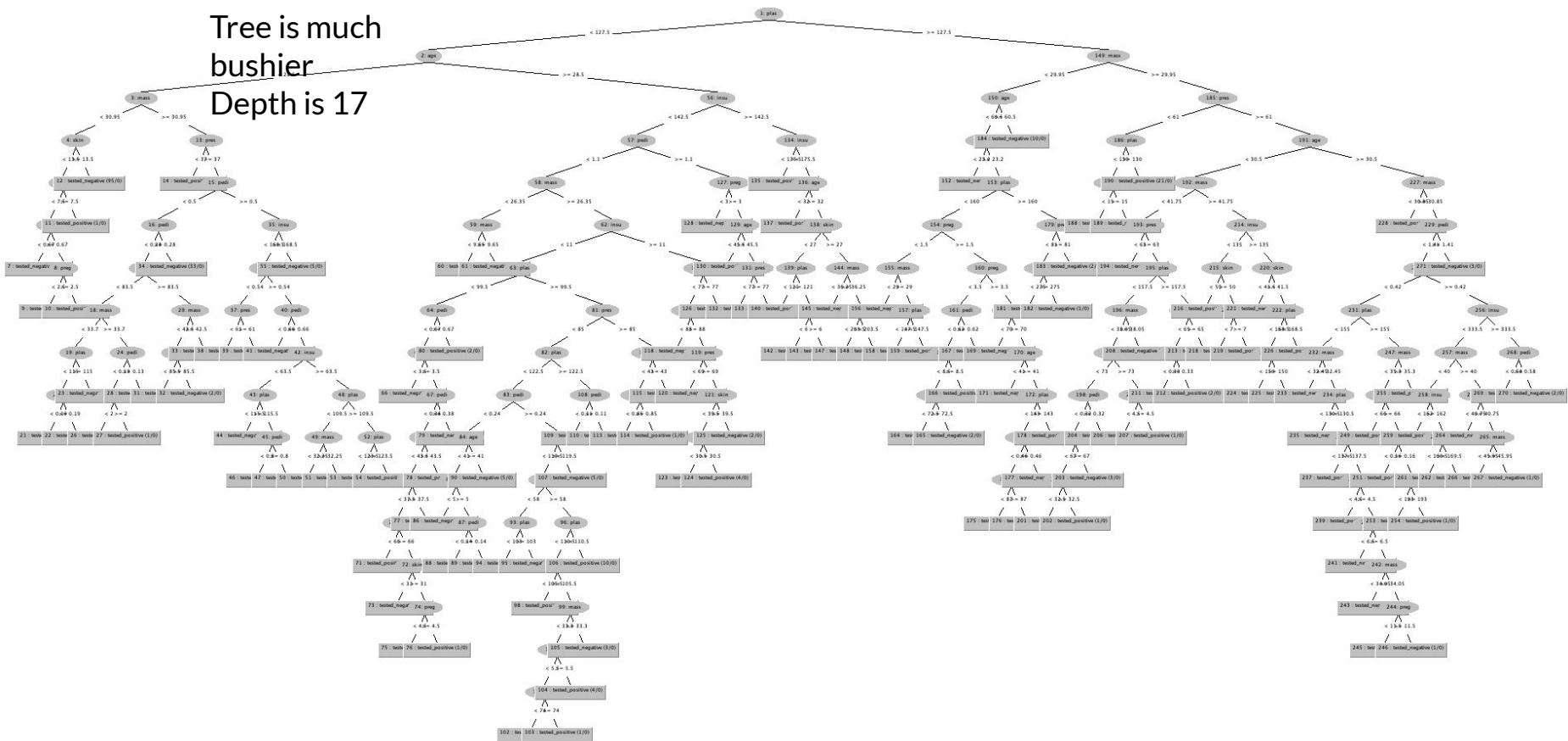
(Min, Max) = (0.1, 1.0) is too fast to tell difference

Task3.3: Explore different algorithms and parameters



	Precision	Recall	F-Measure
J48 Tree	0.735	0.738	0.736
Random Tree	0.684	0.681	0.682

Tree is much
bushier
Depth is 17



	Precision	Recall	F-Measure
J48 Tree	0.735	0.738	0.736
Random Tree D=17	0.684	0.681	0.682
Random Tree D=10	0.702	0.704	0.703
Random Tree D=7	0.723	0.727	0.724
Random Forest	0.754	0.758	0.755
NaiveBayes	0.759	0.763	0.760
Logistic	0.767	0.772	0.765
AdaBoostM1	0.737	0.743	0.738
IBk (Instance Based Learner)	0.696	0.702	0.698