Homework 7

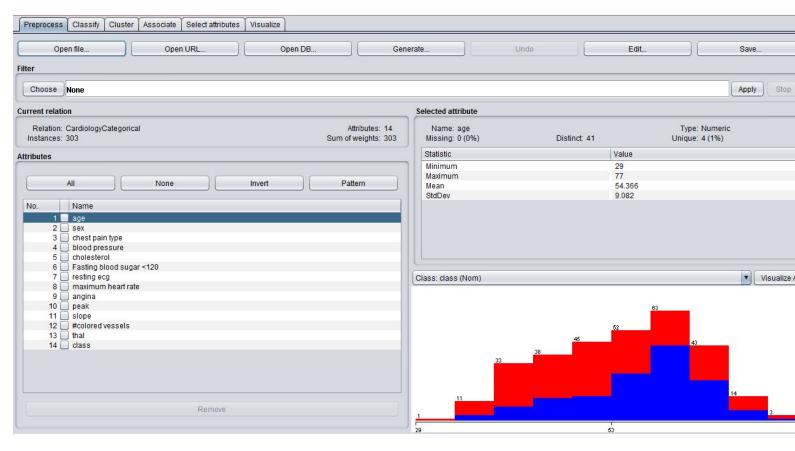
Clustering with WEKA

Ομάδα G

Ευαγγέλου Σωτήρης 2159 Καλαής Κωσταντίνος 2146 Χατζηευφραιμίδης Λευτέρης 2209



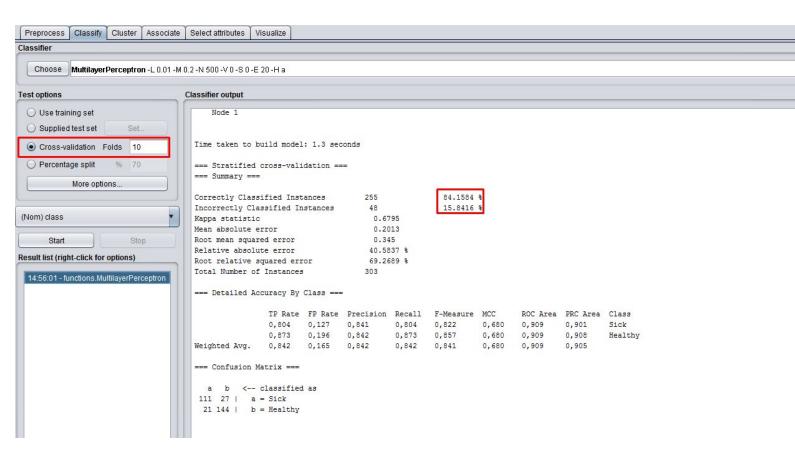
• We load the "CardiologyCategorical.csv" data file to build a neural network (using the MLP algorithm) that predicts whether a patient has a heart condition.



 We choose "True" value on "normalizeAttributes" to normalize our features and set learning rate for weights updates to 0.01

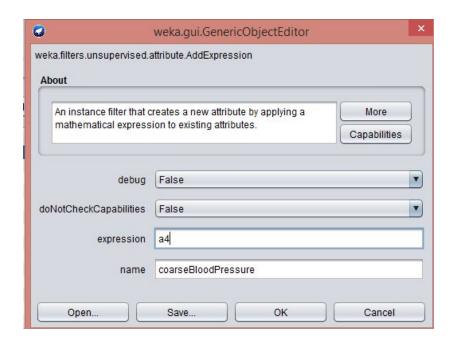


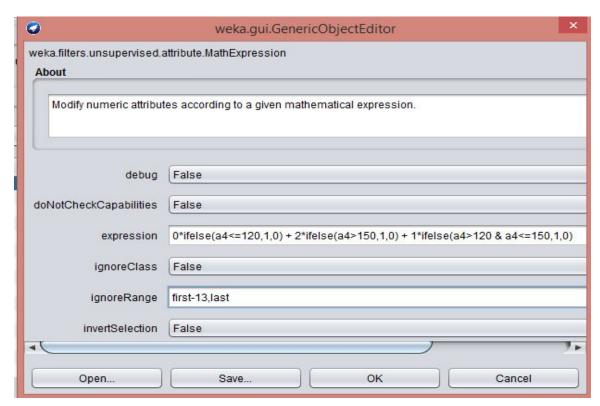
 The 10-fold cross-validation accuracy of your model (A1) is 84.1584%



(2.2)

- We create a new attribute coarseBloodPressure, with values: Low if blood pressure is less than or equal to 120, Normal if blood pressure is greater than 120 but less than or equal to 150, and High if blood pressure is greater than 150
- First we create a copy of *blood pressure* with the new name using filter *AddExpression* and then with *MathExpression* we transform *blood pressure's* values as follows:

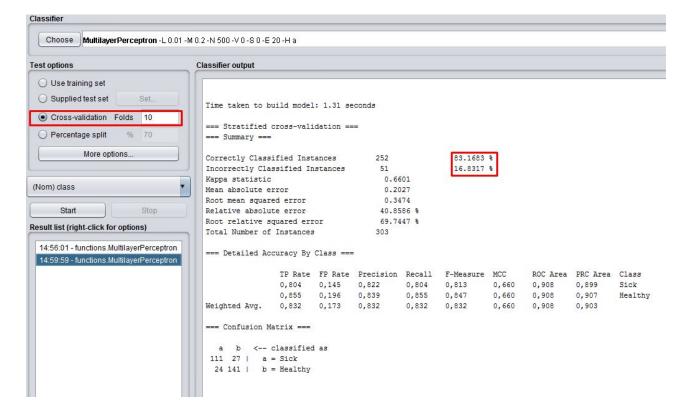




			4: blood pressure 5	5: cholesterol	6: Fastir	ng blood sugar 120 7: resting ecg		9: angina	10: peak	11: slope	12: #colored vessels	13: thal	14: coars		essure	
Numeric		Nominal	Numeric	Numeric		Nominal Nominal	Numeric		Numeric		Numeric	Nominal		Numeric	_	Nominal
	Male	Asymptomatic	130.0		FALSE	Hyp	132.0	TRUE		1 Idi	2.0	Rev		(Normal)		Sick
	Male	Abnormal Angi	130.0		FALSE	Normal	171.0	FALSE	0.6		0.0	Nor		22 - 51		Healthy
	Male	Angina	110.0 -		FALOE	Hyp		TRUE	1.0	Flat		Nor	_	(Low)		Healthy
	Male	Asymptomatic	130.0		FALSE	Нур		FALSE		Flat		Rev				Sick
	Male	Asymptomatic	140.0		TRUE	Нур		TRUE		Down		Rev				Sick
	Fem	Angina	150.0		TRUE	Нур		FALSE	1.0			Nor				Healthy
58.0	Male	Abnormal Angi	120.0		FALSE	Нур		FALSE		Flat		Nor				Sick
58.0	Male	NoTang	132.0		FALSE	Нур		FALSE		Up		Rev				Sick
63.0	Male	Angina	145.0		TRUE	Hyp	150.0	FALSE	2.3	Down	0.0	Fix		59031 230		Healthy
67.0	Male	Asymptomatic	160.0 -	206.0	FALCE	Hyp	108.0	TRUE	1.5	Flat	3.0	1401	_	(High)	2.0	Sick
67.0	Male	Asymptomatic	120.0	229.0	FALSE	Hyp	129.0	TRUE	2.6	Flat	2.0	Rev				Sick
37.0	Male	NoTang	130.0	250.0	FALSE	Normal	187.0	FALSE	3.5	Down	0.0	Nor			1.0	Healthy
41.0	Fem	Abnormal Angi	130.0	204.0	FALSE	Hyp	172.0	FALSE	1.4	Up	0.0	Nor			1.0	Healthy
56.0	Male	Abnormal Angi	120.0	236.0	FALSE	Normal	178.0	FALSE	0.8	Up	0.0	Nor			0.0	Healthy
62.0	Fem	Asymptomatic	140.0	268.0	FALSE	Hyp	160.0	FALSE	3.6	Down	2.0	Nor			1.0	Sick
57.0	Fem	Asymptomatic	120.0	354.0	FALSE	Normal	163.0	TRUE	0.6	Up	0.0	Nor			0.0	Healthy
57.0	Male	Asymptomatic	140.0	192.0	FALSE	Normal	148.0	FALSE	0.4	Flat	0.0	Fix			1.0	Healthy
56.0	Fem	Abnormal Angi	140.0	294.0	FALSE	Hyp	153.0	FALSE	1.3	Flat	0.0	Nor			1.0	Healthy
56.0	Male	NoTang	130.0	256.0	TRUE	Hyp	142.0	TRUE	0.6	Flat	1.0	Fix			1.0	Sick
44.0	Male	Abnormal Angi	120.0	263.0	FALSE	Normal	173.0	FALSE	0.0	Up	0.0	Rev			0.0	Healthy
50.0	Fem	NoTang	120.0	219.0	FALSE	Normal	158.0	FALSE	1.6	Flat	0.0	Nor			0.0	Healthy
58.0	Fem	NoTang	120.0	340.0	FALSE	Normal	172.0	FALSE	0.0	Up	0.0	Nor			0.0	Healthy
66.0	Fem	Angina	150.0	226.0	FALSE	Normal	114.0	FALSE	2.6	Down	0.0	Nor			1.0	Healthy
43.0	Male	Asymptomatic	150.0	247.0	FALSE	Normal	171.0	FALSE	1.5	Up	0.0	Nor			1.0	Healthy
40.0	Male	Asymptomatic	110.0	167.0	FALSE	Hyp	114.0	TRUE	2.0	Flat	0.0	Rev			0.0	Sick
69.0	Fem	Angina	140.0	239.0	FALSE	Normal	151.0	FALSE	1.8	Up	2.0	Nor			1.0	Healthy
60.0	Male	Asymptomatic	117.0	230.0	TRUE	Normal	160.0	TRUE	1.4	Up	2.0	Rev			0.0	Sick
64.0	Male	NoTang	140.0	335.0	FALSE	Normal	158.0	FALSE	0.0	Up	0.0	Nor			1.0	Sick
59.0	Male	Asymptomatic	135.0	234.0	FALSE	Normal	161.0	FALSE	0.5	Flat	0.0	Rev			1.0	Healthy
44.0	Male	NoTang	130.0	233.0	FALSE	Normal	179.0	TRUE	0.4	Up	0.0	Nor			1.0	Healthy
42.0	Male	Asymptomatic	140.0	226.0	FALSE	Normal	178.0	FALSE	0.0		0.0	Nor				Healthy
43.0	Male	Asymptomatic	120.0	177.0	FALSE	Нур	120.0	TRUE		Flat	0.0	Rev				Sick

 We build a neural network (using the MLP algorithm) that predicts whether a patient has a heart condition, using the new attribute coarseBloodPressure instead of the original blood pressure. The 10-fold cross-validation accuracy of your model (A2) is 83.1683%

No.	Name	
1 [age	
2 [sex	
3	chest pain type	
4 [cholesterol	
5	Fasting blood sugar <120	
6	resting ecg	
7	maximum heart rate	
8 [angina	
9 [peak	
10 [slope	
11 [#colored vessels	
12 [thal	
13 [coarseBloodPressure	
14	class	



(2.4)

 The accuracies we achieve from using blood pressure in the first occasion and coarse blood pressure in the second do not have a significant difference (1%). The generalisation of the blood pressure feature does not seem to work better than raw values, but it usually depends to the problem.

(2.5)

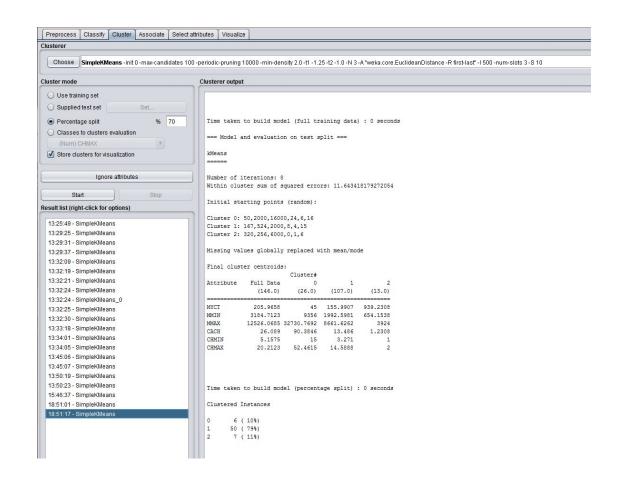
• We remove all records (4 in total) whose value of the attribute *resting ecg* is Abnormal.

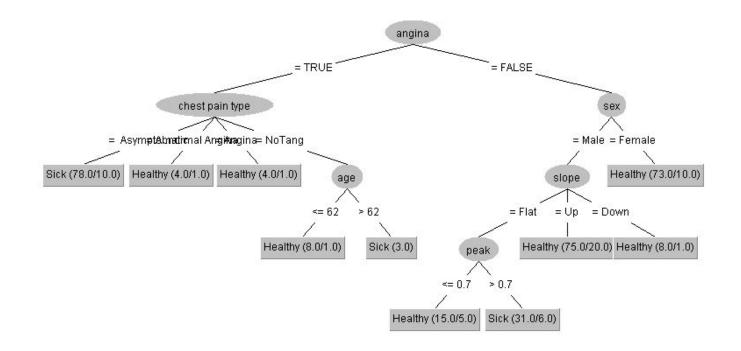
No.	1: age	2: sex	3: chest pain type	4: cholesterol	5: Fasting blood sugar 120	6: resting ecg	7: maximum heart rate	8: angina	9: peak	10: slope	11: #colored vessels	12
	Numeric	Nominal	Nominal	Numeric	Nominal	Nominal	Numeric	Nominal	Numeric	Nominal	Numeric	No
1	60.0	Male	Asymptomatic		FALSE	Нур		TRUE		Flat	2.0	R
2		Male	Abnormal Angi		FALSE	Normal		FALSE	0.6	100 m 2 m		N
3	64.0	Male	Angina	211.0	FALSE	Нур		TRUE	1.8	Flat	0.0	N
4	63.0	Male	Asymptomatic	254.0	FALSE	Нур	147.0	FALSE	1.4	Flat	1.0	R
5	53.0	Male	Asymptomatic	203.0	TRUE	Нур	155.0	TRUE	3.1	Down	0.0	R
3	58.0	Fem	Angina	283.0	TRUE	Нур	162.0	FALSE	1.0	Up	0.0	N
7	58.0	Male	Abnormal Angi	284.0	FALSE	Нур	160.0	FALSE	1.8	Flat	0.0	N
8	58.0	Male	NoTang	224.0	FALSE	Нур	173.0	FALSE	3.2	Up	2.0	R
9	63.0	Male	Angina		TRUE	Нур	150.0	FALSE		Down	0.0	Fi
10	67.0	Male	Asymptomatic		FALSE	Нур	108.0	TRUE	1.5	Flat	3.0	N
11		Male	Asymptomatic		FALSE	Нур	129.0	TRUE		- Contraction Comments	TOTAL	R
12	37.0	Male	NoTona	250.0	-FALSE	Abnormal	187.0	FALSE	3.5	Down	0.0	N
13	41.0	Fem	Undo		ALSE	Нур	172.0	FALSE	1.4	200		N
14	56.0	Contract of the Contract of th	Сору		ALSE	Normal	178.0	FALSE	8.0		0.0	N
15	62.0	Fem	ООРУ		ALSE	Нур		FALSE	3.6		2.0	N
16	57.0	Fem	Search		ALSE	Normal	163.0	TRUE	0.6	Up	0.0	N
17	57.0	Male	Clear search		ALSE	Normal		FALSE	0.4	Flat	0.0	Fi
18		Fem	Doloto colosted		ALSE	Нур		FALSE	1.3	(200,000)		N
19	56.0	Male	Delete selected	or other particular pa	RUE	Нур	142.0	TRUE	0.6	Flat	1.0	Fi
20	44.0	Male	Delete ALL selec	cted instances	LOC	Normal		FALSE			0.0	R
21	50.0	Fem	Insert new instar	nce	ALSE	Normal		FALSE	1.6	Flat	0.0	N
22	58.0	Fem	Set instance wei	ght	ALSE	Normal	172.0	FALSE	0.0	100	0.0	N
23	66.0	Fem	·		ALSE	Normal	114.0	FALSE	2.6	Down	0.0	N
24	43.0	Male	Asymptomatic	247.0	FALSE	Normal	171.0	FALSE	1.5		0.0	N
25	40.0	Male	Asymptomatic		FALSE	Нур		TRUE	2.0		0.0	R
26	69.0	Fem	Angina	239.0	FALSE	Normal	151.0	FALSE	1.8	Up	2.0	N
27	60.0	Male	Asymptomatic	230.0	TRUE	Normal	1000000	TRUE		N		R
82	64.0	Male	NoTang	335.0	FALSE	Normal	158.0	FALSE	0.0	Up	0.0	N
			THE RESERVE OF THE PARTY OF THE								22.3	

(2.6)

• We construct a decision tree (using the J48 Algorithm) that predicts whether a patient has a heart condition, given the attributes age, sex, chest pain type, coarseBloodPressure, angina, peak, and slope.

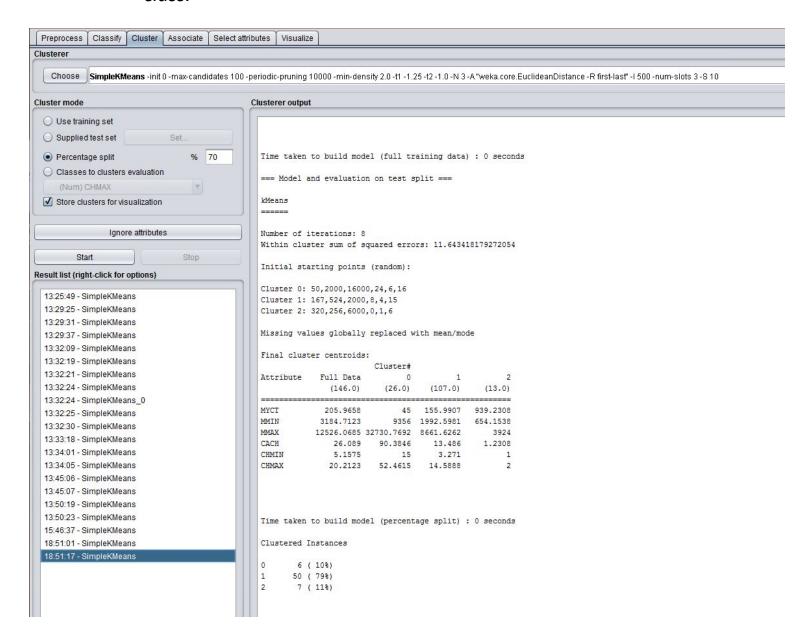
No.	Name
1	age
2	sex
3	chest pain type
4	angina angina
5	peak peak
6	slope
7	coarseBloodPressure
8	class





• Confusion matrix obtained with 10-fold cross-validation in our report.

• We load the "cpu.arff" data file, cluster the data (using the simple k-Means algorithm, with k=3) and delete the attribute class.



• We can observe, that the main mass of the data instances belongs to one cluster (79%). The other two clusters include

a very small percentage of the examples that we can categorise as outliers.

(3.2)

 We discretize the attributes MMIN, MMAX, CACH, CHMIN and CHMAX using 3 buckets in one step and converting numeric values into nominal.

😡 📵 weka.gui.GenericObjectEdi	tor							
weka.filters.unsupervised.attribute.Dis	cretize							
About								
An instance filter that discretizes a range of numeric More								
attributes in the dataset into nomi								
	Сорадиност							
attributeIndices	first-last							
binRangePrecision	6							
bins	3							
debug	False False							
desiredWeightOfInstancesPerInterval								
doNotCheckCapabilities								
findNumBins	False							
ignoreClass	False							
invertSelection	False							
makeBinary	False							
spreadAttributeWeight	False							
useBinNumbers	False							
useEqualFrequency	False							
Open Save	OK Cancel							

- Then, we apply the 'NumericToNominal' filter to every attribute in order to turn them to nominals.
- Then we find associations among these attributes using the Apriori algorithm, with support 0.2, confidence 0.95 and top 3 rules only being displayed. The result buffer:

```
=== Run information ===
Scheme:
            weka.associations.Apriori -N 3 -T 0 -C 0.95 -D 0.05 -U 1.0 -M 0.2 -S -1.0 -c -1
Relation:
       cpu-weka.filters.unsupervised.attribute.Remove-R7-weka.filters.unsupervised.att
       ribute.Remove-R1-weka.filters.unsupervised.attribute.Discretize-B3-M-1.0-Rfirst-
       last-precision1-weka.filters.unsupervised.attribute.NumericToNominal-Rfirst-last
Instances: 209
Attributes: 5
       MMIN
       MMAX
       CACH
       CHMIN
       CHMAX
=== Associator model (full training set) ===
Apriori
======
Minimum support: 0.9 (188 instances)
Minimum metric <confidence>: 0.95
Number of cycles performed: 2
Generated sets of large itemsets:
Size of set of large itemsets L(1): 3
Size of set of large itemsets L(2): 3
Size of set of large itemsets L(3): 1
Best rules found:
```

- 2. MMIN='(-inf-10709.3]' CACH='(-inf-85.3]' 189 ==> CHMIN='(-inf-17.3]' 188 <conf:(0.99)> lift:(1.02) lev:(0.02) [3] conv:(2.26)
- 3. MMIN='(-inf-10709.3]' 201 ==> CHMIN='(-inf-17.3]' 197 <conf:(0.98)> lift:(1) lev:(0) [0] conv:(0.96)

(3.3)

 Using the original data, we list the eigenvalues associated with the attributes selected by the Principal Components Analysis method, when the amount of variance covered by the subset of attributes is 75%.

```
=== Run information ===
Evaluator: weka.attributeSelection.PrincipalComponents -R 0.75 -A 5
          weka.attributeSelection.Ranker -T -1.7976931348623157E308 -N -1
Search:
Relation:
          cpu
Instances: 209
Attributes: 7
       MYCT
       MMIN
       MMAX
       CACH
       CHMIN
       CHMAX
       class
Evaluation mode: evaluate on all training data
=== Attribute Selection on all input data ===
Search Method:
```

Attribute ranking.

Attribute Evaluator (unsupervised):

Principal Components Attribute Transformer

Correlation matrix

1 -0.34 -0.38 -0.32 -0.3 -0.25 -0.34 1 0.76 0.53 0.52 0.27 -0.38 0.76 1 0.54 0.56 0.53 -0.32 0.53 0.54 1 0.58 0.49 -0.3 0.52 0.56 0.58 1 0.55 -0.25 0.27 0.53 0.49 0.55 1

eigenvalue	proportion	cumulative
3.35674	0.55946	0.55946
-0.469MMAX-0.435CH	MIN-0.429CACH-	0.427MMIN-0.374CHMAX
0.82936	0.13823	0.69768
0.682MYCT+0.559CHM	IAX-0.333MMIN-	+0.275CHMIN+0.152CACH
0.73923	0.1232	0.82089
0.669MYCT+0.548MMI	N-0.426CHMAX-	+0.264MMAX-0.03CHMIN

Eigenvectors

V1	V2	V3		
0.29	0.6822	0.6686	MYCT	
-0.4274	4	-0.333 0.547	7 MMIN	
-0.4691	1	-0.1141	0.2643	MMAX
-0.4286	5	0.1516	0.0199	CACH
-0.4353	3	0.2746	-0.0302	CHMIN
-0.3742	2	0.5588	-0.4264	CHMAX

Ranked attributes:

0.441 1 -0.469MMAX-0.435CHMIN-0.429CACH-0.427MMIN-0.374CHMAX...

0.302 2 0.682MYCT+0.559CHMAX-0.333MMIN+0.275CHMIN+0.152CACH...

0.179 3 0.669MYCT+0.548MMIN-0.426CHMAX+0.264MMAX-0.03CHMIN...

Selected attributes: 1,2,3:3