

# Liqi Zhu's Assignment 5

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## 10.2

Suppose that the data mining task is to cluster points (with (x,y) representing location)

into three clusters, where the points are

A1(2,10),A2(2,5),A3(8,4),B1(5,8),B2(7,5),B3(6,4),C1(1,2),C2(4,9).

The distance function is Euclidean distance. Suppose initially we assign A1, B1, and C1

as the center of each cluster, respectively. Use the k-means algorithm to show only

(a) The three cluster centers after the first round of execution.

(b) The final three clusters

(a)

First Round:

Center/Distance	A2	A3	B2	B3	C2
A1	5.00	8.49	7.07	7.21	2.24
B1	4.24	5.00	3.61	4.12	1.41
C1	3.16	7.94	6.71	5.39	7.62

Clusters: {A1},{B1,A3,B2,B3,C2},{C1,A2}.

Centers:(2,10),(6,6),(1.5,3.5)

(b)

Final Clusters:{A1,B1,C2},{A3,B2,B3},{A2,C1}

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## 10.4

For the k-means algorithm, it is interesting to note that by choosing the initial cluster centers carefully, we may be able to not only speed up the algorithm's convergence, but also guarantee the quality of the final clustering. The k-means++ algorithm is a variant of k-means, which chooses the initial centers as follows. First, it selects one center uniformly at random from the objects in the data set. Iteratively, for each object  $p$  other than the chosen center, it chooses an object as the new center. This object is chosen at random with probability proportional to  $\text{dist}(p)^2$ , where  $\text{dist}(p)$  is the distance from  $p$  to the closest center that has already been chosen. The iteration continues until  $k$  centers are selected. Explain why this method will not only speed up the convergence of the k-means algorithm, but also guarantee the quality of the final clustering results.

Firstly, compared to K-means algorithm's randomly picking, k-means++ algorithm picks its initial center with a more smart and even concept, so as to lower the SSEs and thus lower the K-means cost. From this perspective, less iteration are needed. Additionally, K-means++ can lead to a better result since random picked centroids that are not distributed over the data set can cause the result to get stuck in local optimal. K-means++ is obviously prior to K-means considering the distribution of the centers.

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## 10.6

Both k-means and k-medoids algorithms can perform effective clustering.

(a) Illustrate the strength and weakness of k-means in comparison with k-medoids.

(b) Illustrate the strength and weakness of these schemes in comparison with a hierarchical clustering scheme (e.g., AGNES)

(a)

Medoid is less sensitive by outliers or noise in the data set than a mean, the K-medoids algorithm shows more tolerance with more robustness. While K-medoids algorithm costs more for processing.

(b)

Firstly, K-means and K-medoids are partitioning based methods, compared to AGNES, the first advantage of partitioning based method is that it's easy to undo a step by iterative relocation while it's impossible for AGNES to undo process. Hierarchical based clustering methods can not make any adjustments to the former split or merge, such weakness influences the quality of their result.

Secondly, partitioning based clustering shows better capability of finding spherical-shaped patterns for small to medium size databases. While compared to automatically determined number of clusters with AGNES or other hierarchical methods, partitioning based methods need the initialization of the number of clusters.

Hierarchical methods have its difficulties for scaling as each decision may require examination and evaluation of large amount of data. Integration with all kinds of clustering approaches should be considered in order to improve the clustering process.

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## 11.2

AllElectronics carries 1000 products,  $P_1, \dots, P_{1000}$ . Consider customers Ada, Bob, and Cathy such that Ada and Bob purchase three products in common,  $P_1, P_2$ , and  $P_3$ . For the other 997 products, Ada and Bob independently purchase seven of them randomly. Cathy purchases 10 products, randomly selected from the 1000 products. In Euclidean distance, what is the probability that  $\text{dist}(\text{Ada}, \text{Bob}) > \text{dist}(\text{Ada}, \text{Cathy})$ ? What if Jaccard similarity (Chapter 2) is used? What can you learn from this example?

Probability of Ada and Bob purchasing  $m$  same product:

$$P_{AB} = \frac{C_{990}^{10-m}}{C_{997}^7}$$

Probability of Ada and Cathy purchasing  $n$  same products:

$$P_{AC} = \frac{C_{990}^{10-n}}{C_{1000}^{10}}$$

With Euclidean distance,

$$P[\text{dist}(\text{Ada}, \text{Bob}) > \text{dist}(\text{Ada}, \text{Cathy})] = P(m < n) = \sum_{m=3}^{10} (P_{AB} \sum_{n=m+1}^{10} P_{AC})$$

With Jaccard similarity

$$\text{dist}(\text{Ada}, \text{Bob}) = \frac{(10 - m) + (10 - m)}{m + (10 - m) + (10 - m)} = \frac{20 - 2m}{20 - m}$$

$$\text{dist}(\text{Ada}, \text{Cathy}) = \frac{(10 - n) + (10 - n)}{n + (10 - n) + (10 - n)} = \frac{20 - 2n}{20 - n}$$

We can find that from Jaccard method we have the same result as

$P[\text{dist}(\text{Ada}, \text{Bob}) > \text{dist}(\text{Ada}, \text{Cathy})] = P(j < k)$ , leads to the same probability using Euclidean distance. As a conclusion, no matter what measurement method is applied to two binary attribute sets, the result would not be affected.

## 9.1

Using Weka, solve 9.1 with MLNN, SVM, and another classifier of your choice

Following results are generated with the settings of 10-fold cross-validation with all available evaluation metrics.

## MLNN

=== Run information ===

Scheme: weka.classifiers.functions.MultilayerPerceptron -L 0.3 -M 0.2 -N 500 -V 0 -S 0

-E 20 -H a

Relation: 9.1

Instances: 165

Attributes: 4

department

status

age

salary

Test mode: 10-fold cross-validation

=== Classifier model (full training set) ===

Sigmoid Node 0

Inputs Weights

Threshold 5.077849919304931

Node 2 -0.798870007989803

Node 3 -3.2553307300492724

Node 4 -1.6849103640762992

Node 5 -1.9722357346589903

Node 6 -0.6077028566801436

Node 7 0.6877457905152902

Node 8 -0.49825100408460005

Node 9 -2.614381465228299

Node 10 -2.1185097501440233

Sigmoid Node 1

Inputs Weights

Threshold -5.077776341999067

Node 2 0.8299328096639218

Node 3 3.2668069308453087

Node 4 1.6310381053303158

Node 5 1.9751693228544362

Node 6 0.5808522121372128

Node 7 -0.6837530207251162

Node 8 0.4846290073259015

Node 9 2.6150624262231497

Node 10 2.1484680707196433

Sigmoid Node 2

Inputs Weights

Threshold -0.08620349986676684

Attrib department=sales 0.012600993238492612

Attrib department=systems 0.3191193091224657

Attrib department=marketing 0.01847411578253666

Attrib department=secretary -0.19171693697429368

Attrib age=31;35 -0.3122482391062593

Attrib age=26;30 0.5835492730772195

Attrib age=21;25 0.7542810828120736

Attrib age=41;45 -0.25163364381064085

Attrib age=36;40 -0.38342085512708035

Attrib age=46;50 -0.20928257942558

Attrib salary=46K;50K -0.571945387696409

Attrib salary=26K;30K 0.3499069945630241

Attrib salary=31K;35K 0.8599736495383075

Attrib salary=66K;70K -0.6757140926272822

Attrib salary=41K;45K 0.47382264656480233

Attrib salary=36K;40K -0.23159494351598084

Sigmoid Node 3

Inputs Weights

Threshold 0.025346272016822595

Attrib department=sales -0.32864194581758754

Attrib department=systems 0.5920642195263259

Attrib department=marketing 0.20156508031761183

Attrib department=secretary -0.6007585551542152

Attrib age=31;35 -0.6062600733319657

Attrib age=26;30 1.2852738771258667  
Attrib age=21;25 1.5699313688961294  
Attrib age=41;45 -0.7509412765773987  
Attrib age=36;40 -1.0309086447642049  
Attrib age=46;50 -0.7302655318150391  
Attrib salary=46K;50K -1.2139777194307757  
Attrib salary=26K;30K 0.48350101620000774  
Attrib salary=31K;35K 1.8622117941871033  
Attrib salary=66K;70K -1.8354040575268293  
Attrib salary=41K;45K 1.170525103478124  
Attrib salary=36K;40K -0.7178887222234804  
Sigmoid Node 4

Inputs Weights

Threshold 0.0465277861983857  
Attrib department=sales -0.18370043339263006  
Attrib department=systems 0.38317269323577663  
Attrib department=marketing 0.06872941128984131  
Attrib department=secretary -0.3696472894376704  
Attrib age=31;35 -0.4057412390102155  
Attrib age=26;30 0.9276241611072675  
Attrib age=21;25 1.0970233668154243  
Attrib age=41;45 -0.44715060590159267  
Attrib age=36;40 -0.6263550280798215  
Attrib age=46;50 -0.42646235986262754  
Attrib salary=46K;50K -0.8282809963284237  
Attrib salary=26K;30K 0.3958174335925043  
Attrib salary=31K;35K 1.332524172687338  
Attrib salary=66K;70K -1.1400832034287853  
Attrib salary=41K;45K 0.6956100621626622  
Attrib salary=36K;40K -0.511068938201242

Sigmoid Node 5

Inputs Weights

Threshold -0.022426765433450343

Attrib department=sales -0.19306856622245014  
Attrib department=systems 0.4779361865593974  
Attrib department=marketing 0.16898916784784349  
Attrib department=secretary -0.4262141188821358  
Attrib age=31;35 -0.46731068585513974  
Attrib age=26;30 0.9470242446874014  
Attrib age=21;25 1.1610232481211777  
Attrib age=41;45 -0.56156635519177  
Attrib age=36;40 -0.7644067381327756  
Attrib age=46;50 -0.5476332961238081  
Attrib salary=46K;50K -0.9307277394440308  
Attrib salary=26K;30K 0.4105143519212113  
Attrib salary=31K;35K 1.4298906672631688  
Attrib salary=66K;70K -1.3146083724285471  
Attrib salary=41K;45K 0.870902790922854  
Attrib salary=36K;40K -0.5146412094931446

Sigmoid Node 6

Inputs Weights

Threshold -0.11246408558877943  
Attrib department=sales 0.030943733941151127  
Attrib department=systems 0.2707166034357742  
Attrib department=marketing 0.003653113313281002  
Attrib department=secretary -0.11375860909842499  
Attrib age=31;35 -0.22835596759234983  
Attrib age=26;30 0.5702668345109629  
Attrib age=21;25 0.6152697926035567  
Attrib age=41;45 -0.18140268271770174  
Attrib age=36;40 -0.22993512543217498  
Attrib age=46;50 -0.17777278513441389  
Attrib salary=46K;50K -0.5029538515802372  
Attrib salary=26K;30K 0.3585084705191638  
Attrib salary=31K;35K 0.7343350310882023  
Attrib salary=66K;70K -0.5033851892889031



Attrib salary=41K;45K 0.41705013418198206

Attrib salary=36K;40K -0.15310408987359056

Sigmoid Node 7

Inputs Weights

Threshold 0.14359968748690133

Attrib department=sales 0.0308844930321132

Attrib department=systems -0.038245001369986015

Attrib department=marketing -0.1722482354589961

Attrib department=secretary -0.1157825138492811

Attrib age=31;35 -0.0457880723340485

Attrib age=26;30 -0.0939867531181069

Attrib age=21;25 -0.10098280549550315

Attrib age=41;45 -0.07971900177279614

Attrib age=36;40 -0.1273617352323285

Attrib age=46;50 -0.09413156744851643

Attrib salary=46K;50K -0.16651353883117026

Attrib salary=26K;30K -0.035995972716067914

Attrib salary=31K;35K -0.0981135045576773

Attrib salary=66K;70K -0.07394911962561224

Attrib salary=41K;45K -0.21370187168690238

Attrib salary=36K;40K -0.06538005656135208

Sigmoid Node 8

Inputs Weights

Threshold -0.07037686547883158

Attrib department=sales 0.061932218451711145

Attrib department=systems 0.18996592605127247

Attrib department=marketing 0.04851228462146839

Attrib department=secretary -0.07753869898783228

Attrib age=31;35 -0.22139508044791156

Attrib age=26;30 0.4862724224154825

Attrib age=21;25 0.6094957485005332

Attrib age=41;45 -0.10289434094536438

Attrib age=36;40 -0.22298678451458484

Attrib age=46;50 -0.14035104663170997  
Attrib salary=46K;50K -0.4836930970715099  
Attrib salary=26K;30K 0.34113831771448133  
Attrib salary=31K;35K 0.7354872646431403  
Attrib salary=66K;70K -0.3849517106785908  
Attrib salary=41K;45K 0.37576312175178983  
Attrib salary=36K;40K -0.09983684200290624

Sigmoid Node 9

Inputs Weights

Threshold 0.01605893089995354  
Attrib department=sales -0.28192817133154097  
Attrib department=systems 0.5355771588258466  
Attrib department=marketing 0.1766652263286512  
Attrib department=secretary -0.493289269014781  
Attrib age=31;35 -0.5385435118436851  
Attrib age=26;30 1.1396325250733903  
Attrib age=21;25 1.3671122817007393  
Attrib age=41;45 -0.6635552565499636  
Attrib age=36;40 -0.9116240401660546  
Attrib age=46;50 -0.65155542422012  
Attrib salary=46K;50K -1.0697808162408653  
Attrib salary=26K;30K 0.4478045557254808  
Attrib salary=31K;35K 1.6680581330567885  
Attrib salary=66K;70K -1.5936574851889078  
Attrib salary=41K;45K 1.0555492088761849  
Attrib salary=36K;40K -0.6457927285160276

Sigmoid Node 10

Inputs Weights

Threshold 0.055144735855013755  
Attrib department=sales -0.1922017825107403  
Attrib department=systems 0.507332887135344  
Attrib department=marketing 0.13934348432502755  
Attrib department=secretary -0.45433208104598666

Attrib age=31;35 -0.49774153597825016  
Attrib age=26;30 1.0117806818794415  
Attrib age=21;25 1.2301685688075434  
Attrib age=41;45 -0.5648255767286017  
Attrib age=36;40 -0.7788934770237211  
Attrib age=46;50 -0.5608808550266506  
Attrib salary=46K;50K -0.9605860742985314  
Attrib salary=26K;30K 0.39450429042784996  
Attrib salary=31K;35K 1.4922512282746332  
Attrib salary=66K;70K -1.3967736509967557  
Attrib salary=41K;45K 0.9466130237580793  
Attrib salary=36K;40K -0.5329714314981593

Class senior

Input

Node 0

Class junior

Input

Node 1

Time taken to build model: 0.29 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 165 100 %

Incorrectly Classified Instances 0 0 %

Kappa statistic 1

Mean absolute error 0.0044

Root mean squared error 0.0055

Relative absolute error 1.0198 %

Root relative squared error 1.1821 %

Total Number of Instances 165

=== Detailed Accuracy By Class ===

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		TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC A
rea	PRC Area	Class						
		1.000	0.000	1.000	1.000	1.000	1.000	1.000
1.000		senior						
		1.000	0.000	1.000	1.000	1.000	1.000	1.000
1.000		junior						

Weighted Avg. 1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000

=== Confusion Matrix ===

a b <-- classified as

52 0 | a = senior

0 113 | b = junior

## SVM

=== Run information ===

Scheme: weka.classifiers.functions.VotedPerceptron -I 1 -E 1.0 -S 1 -M 10000

Relation: 9.1

Instances: 165

Attributes: 4

department

status

age

salary

Test mode: 10-fold cross-validation

=== Classifier model (full training set) ===

VotedPerceptron: Number of perceptrons=16

Time taken to build model: 0 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 163 98.7879 %

Incorrectly Classified Instances 2 1.2121 %

Kappa statistic 0.9716

Mean absolute error 0.0127

Root mean squared error 0.1095

Relative absolute error 2.9433 %

Root relative squared error 23.5649 %

Coverage of cases (0.95 level) 98.7879 %

Mean rel. region size (0.95 level) 50.303 %

Total Number of Instances 165

=== Detailed Accuracy By Class ===

		TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC A
rea	PRC Area	Class						
1.000	0.962	senior	0.000	1.000	0.962	0.980	0.972	1.000
1.000	1.000	junior	0.038	0.983	1.000	0.991	0.972	1.000

Weighted Avg. 0.988 0.026 0.988 0.988 0.988 0.972 1.000 1.000

=== Confusion Matrix ===

a b <-- classified as

50 2 | a = senior

0 113 | b = junior

## ClassificationViaRegression

=== Run information ===

Scheme: weka.classifiers.meta.ClassificationViaRegression -W

weka.classifiers.trees.M5P -- -M 4.0

Relation: 9.1

Instances: 165

Attributes: 4

department

status

age

salary

Test mode: 10-fold cross-validation

=== Classifier model (full training set) ===

Classification via Regression

Classifier for class with index 0:

M5 pruned model tree:

(using smoothed linear models)

salary=46K;50K,66K;70K,36K;40K <= 0.5 : LM1 (90/0%)

salary=46K;50K,66K;70K,36K;40K > 0.5 : LM2 (75/0%)

LM num: 1

status =

0.0593 \* age=31;35,41;45,36;40,46;50

+ 0.0869 \* salary=46K;50K,66K;70K,36K;40K

- 0.029

LM num: 2

status =

0.9025 \* age=31;35,41;45,36;40,46;50

+ 0.1014 \* salary=46K;50K,66K;70K,36K;40K

- 0.0338

Number of Rules : 2

Classifier for class with index 1:

M5 pruned model tree:

(using smoothed linear models)

salary=26K;30K,41K;45K,31K;35K <= 0.5 : LM1 (75/0%)

salary=26K;30K,41K;45K,31K;35K > 0.5 : LM2 (90/0%)

LM num: 1

status =

0.9025 \* age=26;30,21;25

+ 0.1014 \* salary=26K;30K,41K;45K,31K;35K

+ 0.0299

LM num: 2

status =

0.0593 \* age=26;30,21;25

+ 0.0869 \* salary=26K;30K,41K;45K,31K;35K

+ 0.8828

Number of Rules : 2

Time taken to build model: 0.01 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 165 100 %

Incorrectly Classified Instances 0 0 %

Kappa statistic 1

Mean absolute error 0.0301

Root mean squared error 0.0386

Relative absolute error 6.9589 %

Root relative squared error 8.3026 %

Coverage of cases (0.95 level) 100 %

Mean rel. region size (0.95 level) 56.9697 %

Total Number of Instances 165

=== Detailed Accuracy By Class ===

		TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC A
rea	PRC Area	Class						
		1.000	0.000	1.000	1.000	1.000	1.000	1.000
1.000		senior						
		1.000	0.000	1.000	1.000	1.000	1.000	1.000
1.000		junior						

Weighted Avg. 1.000 0.000 1.000 1.000 1.000 1.000 1.000 1.000

=== Confusion Matrix ===

a b <-- classified as  
52 0 | a = senior  
0 113 | b = junior