

# Test #1

## CS5402 — Intro To Data Mining

Illya Starikov

Due Date: July 15<sup>th</sup>, 2018

### Multiple Choice

1. e. None of the above
2. c. Remove any attribute that has missing values.
3. b.  $\frac{1}{2}$
4. b. wt
5. d. Spearman's rank correlation coefficient
6. c. Healthland
7. b. slice for Time = Q1
8. d. roll up on Location = Beijing or Tokyo (i.e., from city to country)
9. c. drill down on Time = Q1 (i.e., from quarter to month)
10. a. dice for (location = Beijing or Tokyo) and (product = Chain or bracelet) and (time = Q1 or Q2)

### 11 Short Answer

Method #1 is the most accurate, because the true positive ( $y$ -axis) correctly identified the values, while the false positive ( $x$ -axis) incorrectly identified the values. Method #1 had the fastest growing function (with respect to  $y$ ).

## 12 1-R Method

Attribute	Attribute Value	# Rows With Attribute Value	Most Frequent Value For sportPref	Errors	Total Errors
ageGroup	youngAdult	3	football (2)	1	3
	middleAge	3	football/hockey/baseball (1/1/1)	2	
	senior	2	baseball (2)	0	
gender	M	5	baseball/football (2/2)	3	5
	F	3	football/hockey/baseball (1/1/1)	2	
petPreference	dog	5	football (3)	2	3
	cat	3	baseball (2)	1	

The rules are as follows:

ageGroup = **youngAdult**  $\implies$  football  
 ageGroup = **middleAge**  $\implies$  football  
 ageGroup = **senior**  $\implies$  baseball

## 13 Prism

For football, we get the following table:

gender	pet	drink	sport
M	dog	beer	football
F	dog	beer	football

For our P and T values:

	T	P	T/P
gender = M	3	1	1/3
gender = F	4	1	1/4
pet = dog	3	2	2/3
drink = beer	3	2	3/4

Seeing as not T/P values are 1, we must add a clause. We choose pet = dog as the base.

	T	P	T/P
gender = M	1	0	0
gender = F	1	0	0
drink = beer	2	2	1

pet = **dog** and drink = **beer**  $\implies$  football

## 14 Statistical Modeling

The likelihood would be as follows:

$$\text{likelihood} = 4/9 \times 2/9 \times 6/9 \times 3/9 \times 9/14$$

## 15 Entropy

(a) entropyBeforeSplit would be as follows:

$$-1/6 \log_2 (1/6) - 2/6 \log_2 (2/6) - 3/6 \log_2 (3/6)$$

(b) entropyPoor would be as follows:

$$-2/4 \log_2 (2/4) - 2/4 \log_2 (2/4)$$

(c) infoGain would be determined as follows:

$$\begin{aligned} \text{entropyAfterSplit} &= 3/6 \text{entropyShort} + 2/6 \text{entropyMed} + 1/6 \text{entropyLong} \\ \text{infoGain} &= \text{entropyBeforeSplit} - \text{entropyAfterSplit} \end{aligned}$$

## 16 Rule Induction

(a) The partitions would be as follows:

$$\begin{aligned} \{d\}^* &= \{\{x_1\}, \{x_2, x_3\}, \{x_5\}, \{x_5\}\} \\ \{e\}^* &= \{\{x_1, x_2, x_5\}, \{x_3, x_4\}\} \\ \{d, e\}^* &= \{\{x_1\}, \{x_2\}, \{x_3\}, \{x_4\}, \{x_5\}\} \end{aligned}$$

(b) The coverings are as follows:

- $\{d\}^*$  would not work, because every block in the partition is not a subset of a block in  $\{f\}^*$ .
- $\{d, e\}^*$  would work, because every block in the partition is a subset of a block in  $\{f\}^*$ .
- $\{a, d, e\}^*$  would not work, because although every block in the partition is a subset of a block in  $\{f\}^*$ , it is not minimal.

(c) The rules would be as follows:

$$d = X \text{ and } e = 4 \implies f = T$$

$$d = S \text{ and } e = 4 \implies f = T$$

$$d = S \text{ and } e = 3 \implies f = F$$

$$d = H \text{ and } e = 3 \implies f = F$$

$$d = M \text{ and } e = 4 \implies f = F$$

## 17 KD-Tree

Sorting, we get the following:  $[(2, 10), (4, 20), (6, 10), (8, 20), (10, 30)]$ .  
With a median of 6...

- $x < 6$  group:  $[(2, 10), (4, 20)]$
- $x \geq 6$  group:  $[(6, 10), (8, 20), (10, 30)]$

Sorting, we get the following:  $[(2, 10), (4, 20)] [(6, 10), (8, 20), (10, 30)]$   
With a median of 15 for the first group:

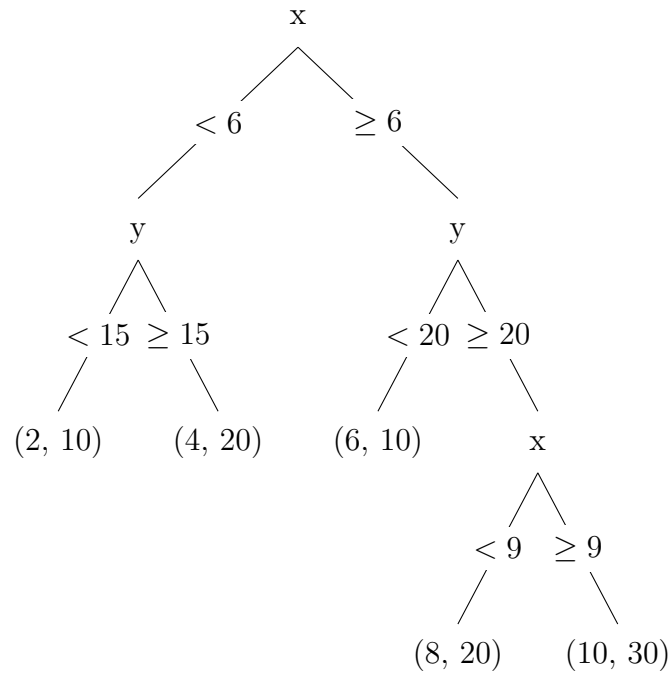
- $y < 15$  group:  $[(2, 10)]$
- $y \geq 15$  group:  $[(4, 20)]$

With a median of 20 for the second group:

- $y < 20$  group:  $(6, 10)$
- $y \geq 20$  group:  $[(8, 20), (10, 30)]$

(Using a shortcut for the final block), Sorting, and using a median of 9, our last block looks like as follows:

- $x < 9$  group:  $[(8, 20)]$
- $y \geq 9$  group:  $[(10, 30)]$



## 18 Clustering

x	y	distance to (2, 4)	distance to (5, 6)	distance to (8, 1)
2	4	0	5	9
5	6	5	0	8
8	1	9	8	0
7	3	6	5	3
4	10	8	5	13
3	0	5	8	6
9	8	11	6	8

Our clusters would be as follows:

**Cluster Center (2, 4)** (2, 4), (3, 0)

**Cluster Center (5, 6)** (5, 6), (4, 10), (9, 8)

**Cluster Center (8, 1)** (8, 1), (7, 3)

With means as follows:

**Cluster Mean of (2, 4), (3, 0)**  $(2.5, 2) \approx (3, 2)$

**Cluster Mean of (5, 6), (4, 10), (9, 8)**  $(6, 8)$

**Cluster Center of (8, 1), (7, 3)**  $(7.5, 2) \approx (8, 2)$

x	y	distance to (3, 2)	distance to (6, 8)	distance to (8, 2)
2	4	3	8	8
5	6	6	3	7
8	1	6	9	1
7	3	5	6	2
4	10	9	4	12
3	0	2	11	7
9	8	12	3	7

**Cluster Center (3, 2)** (2, 4), (3, 0)

**Cluster Center (6, 8)** (5, 6), (4, 10), (9, 8)

**Cluster Center (8, 2)** (8, 1), (7, 3)

*Clusters haven't changed!* Final cluster centers and instances are as follows:

**Cluster Center (3, 2)** (2, 4, 11, yes), (3, 0, 3, yes)

**Cluster Center (6, 8)** (5, 6, 5, no), (4, 10, 8, yes), (9, 8, 1, no)

**Cluster Center (8, 2)** (8, 1, 7, no), (7, 3, 4, yes)

## 19 Confusion Table

- (a) For a randomly produced results, there were 8 values that we predicted to be B, when they were actually G.
- (b) For a classifier produced results, there were 30 values that we predicted to be B, and were actually B.
- (c) The non-random classifier, 90 were predicted correctly. For the random classifier, 39 were predicted correctly. Therefore, 51 more were predicted correctly.
- (d) Kappa Statistic would be

$$\frac{\text{Non-Random Correct} - \text{Random Correct}}{\text{Total}}$$

Which would be as follows:

$$\frac{90 - 39}{100}$$