Homework 1

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1. Decision Trees

- 1.(a) IF x_1 is TRUE then YES

 IF x_1 is FALSE then

 IF x_2 is FALSE then NO

 ELSE IF x_2 is TRUE then

 IF x_3 is TRUE the YES

 ELSE IF x_3 is FALSE then NO
- 1.(b) IF x_1 is TRUE then IF x_2 is TRUE then NO IF x_2 is FALSE then YES ELSE IF x_1 is FALSE then IF x_2 is TRUE then YES IF x_2 is FALSE then NO
- 1.(c)) IF x_1 is TRUE then IF x_2 is TRUE then YES ELSE IF x_2 is FALSE then IF x_3 is TRUE the YES ELSE IF x_3 is FALSE then NO IF x_1 is FALSE then IF x_2 is FALSE then NO ELSE IF x_2 is TRUE then IF x_3 is TRUE the YES ELSE IF x_3 is FALSE then NO
- 2.(a) The possible number of functions to map the four feature to Boolean functions is $2^{2.2.3.4} = 2^{48}$. The number of functions consistent with the given training dataset is 9.

2.(b)The ratio of positive labels is 5/9 and ratio of negative labels is 4/9

$$Entropy(S) = H(S) = -p_{+} \log_{2} p_{+} + p_{-} \log_{2} p_{-}$$
$$= -(5/9.log_{2}(5/9) + 4/9.log_{2}(4/9))$$
$$= 0.991$$

2.(c) The formula for information gain is:-

$$Gain(S, A) = Entropy(S) - \sum_{v \in A} |S_v|/|S|Entropy(S_v)$$

Using the Entropy Formula from 2(b)

Friday:NO

$$p_{+} = 4/6, p_{-} = 2/6$$

$$H_N = -(4/6\log_2(4/6) + 2/6\log_2(2/6)$$

 $H_N = 0.918$

Friday:Yes
$$p_{+} = 1/3, p_{-} = 2/3$$

$$H_N = -(1/3\log_2(1/3) + 2/3\log_2(2/3)$$

$$H_N = 0.918$$

Expected Entropy = 6/9.0.918 + 3/9.0.918 = 0.918

Information Gain = 0.991 - 0.918 = 0.073 (0.991 is entropy of labels calculated in 2b)

Similarly for all attributes of each feature vector we get:

Hungr:Yes

Entropy = 0.722

Hungry:No

Entropy = 0.811

Expected Entropy = 0.762

Information Gain = 0.991 - 0.762 = 0.229

Patrons:Some

Entropy = 0

 $\begin{array}{c} {\rm Patrons:Full} \\ {\rm Entropy} = 0.811 \end{array}$

Patrons:None

Entropy = 0

Expected Entropy = 0.360

Information Gain = 0.991 - 0.360 = 0.631

Type:French

Entropy = 1

Type:Thai

Entropy = 0.918

Type:Chinese

Entropy = 0.918

Type:Italian

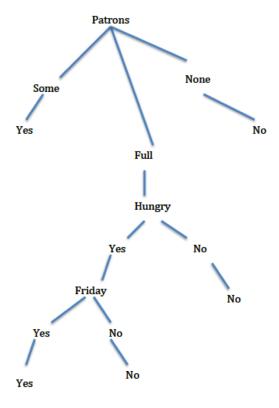
Entropy = 0

Expected Entropy = 0.834

Information Gain = 0.991 - 0.834 = 0.157

2.(d) As per the information gain calculated above in 2(c) the highest gain is of Patrons. Therefore Patrons will be the root of the ID3.

2(e)



2(f) Predicted labels:

ii)No

iii) Yes

Hence the accuracy is 2/3=0.6667 or 66.67%

3(a) Misclasification(S) = $1 - \max_i p_i$

i) Hence the

 $InformationGain = Misclassification(S) - \sum_{v \in A} |S_v| / |S| Misclassification(S_v)$

ii) Using the Misclassification Formula from 3(a) (i)

Missclassification of labels =
$$1 - \max 4/9, 5/9 = 1 - 5/9 = 4/9$$

Friday:NO

$$p_+ = 4/6, p_- = 2/6$$

$$Misclassification(No) = 1 - \max(4/6, 2/6) = 1 - 4/6 = 2/6$$

Friday:Yes
$$p_{+} = 1/3, p_{-} = 2/3$$

$$Misclassification(Yes) = 1 - \max 1/3, 2/3 = 1 - 2/3 = 1/3$$

Expected Misclassification =
$$6/9.2/6 + 3/9.1/3 = 1/3$$

Information Gain =
$$4/9 - 1/3 = 1/9 = 0.11$$

Similarly for all attributes of each feature vector we get:

Information Gain for Hungry =
$$4/9 - 2/9 = 0.222$$

Information Gain for Patrons =
$$4/9 - 1/9 = 3/9 = 0.333$$

Information Gain for Type =
$$4/9 - 1/3 = 0.111$$

The root is Patrons

3(b)
$$Gini(s) = \sum_{i} p_{i}(1 - p_{i})$$

Using the above equation we get:

$$Gini(labels) = 4/9.5/9 + 5/9.4/9 = 40/81$$

Friday:NO

$$p_+ = 4/6, p_- = 2/6$$

$$Gini(No) = 4/6.2/6 + 2/6.4/6 = 16/36$$

Friday: Yes
$$Gini(Yes) = 4/9$$

Expected Gini =
$$72/162$$

Information Gain =
$$40/81 - 72/162 = 0.049$$

Similarly for all attributes of each feature vector we get:

Information Gain for Hungry = 0.149

Information Gain for Patrons= 0.327

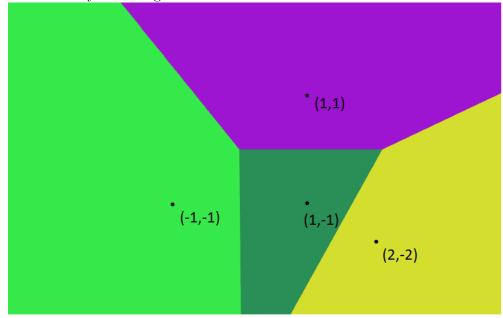
Information Gain for Type= 0.086

The root is Patrons

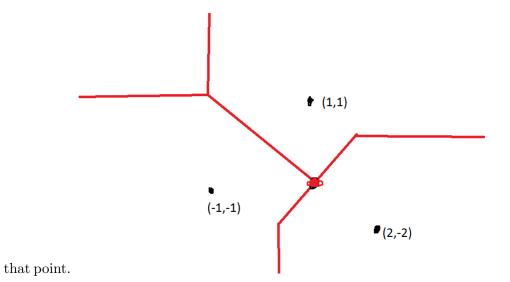
2. Neighbors

1. Euclidean Voronoi Map

In this using the euclidean distance we passed a perpendicular bisector through it each euclidean line. The intersection points of each bisector formed boundary of the regions.



2. Manhattan Distance Voronoi Using the manhattan distance formula we found regions from where the points were equidistant to regions in doin so for the three points common regions got absorbed as they fell being close to



3.taking the points from the above question we can use it to form our training set.

the point (2,0) is closer to c in manhattan distance but closer to A in the euclidean distance.

- 3.Experiment
- 1. The decision tree structure is like a tree generalized for multiple links and values for links. The structure stores if it is a leaf and also on which column the tree has split. The formation of the tree is part of ID3 algorithm where the subtree is added to the parent node. The tree node stores label data in case of leaf.
 - 2. The accuracy of my decision tree on the tic tac toe data is 85.7%
- 3. In KNN i have used comparison of characters to calculate distance. If the character is same distance is 0 else 1. This is done against columns of testdata with training data. Each row of test data against all of the training data.
 - 4. The average accuracy of each K value is as follows:

K value:1 80.315% K value:2 82.021% K value:3 89.1076% K value:4 88.7139% K value:5 91.8635%

Final K value is: 5 with avg accuracy 91.8635%

The Final Accuracy using K = 5 is 92.8571%