FIT3152 Mock eExam Questions

R Coding (10 Marks)

eExam Q1 (4 Marks)

The DunHumby (**DH**) data frame records the **Date** a **Customer** shops at a store, the number of **Days** since their last shopping visit, and amount **Spent** for 20 customers. The first 4 rows are shown below.

The DunHumby (**DH**) data frame records the **Date** a **Customer** shops at a store, the number of **Days** since their last shopping visit, and amount **Spent** for 20 customers. The first 4 rows are shown below.

> head(DH) customer_id visit_date visit_delta visit_spend <int> <chr> 40 04-04-10 <int> <db1> 44.8 NA 2 40 06-04-10 2 69.7 3 40 19-04-10 13 44.6 40 01-05-10 12 30.4

The following R code is run:

```
DHY = DH[as.Date(DH$visit_date,"%d-%m-%y") < as.Date("01-01-11","%d-%m-%y"),]
CustSpend = as.table(by(DHY$visit_spend, DHY$customer_id, sum))
CustSpend = sort(CustSpend, decreasing = TRUE)
CustSpend = head(CustSpend, 12)
CustSpend = as.data.frame(CustSpend)
colnames(CustSpend) = c("customer_id", "amtspent")
DHYZ = DHY[(DHY$customer_id %in% CustSpend$customer_id),]
write.csv(DHYZ, "DHYZ.csv", row.names = FALSE)
g = ggplot(data = DHYZ) + geom_histogram(mapping = aes(x = visit_spend)) +
facet_wrap(~ customer_id, nrow = 3)</pre>
```

Describe the action and output(s) of the R code.

Describe the action and output(s) of the R code.

| | _ | _ | • |
|---|-------|---|---|
| · | • | • | |

eExam Q2 (6 Marks)

| Desci | ribe the function performed by each line of code or code fragment. [1 Mark] |
|-------|--|
| (a) | DHY = DH[as.Date(DH\$visit_date,"%d-%m-%y") < as.Date("01-01-11","%d-%m-%y"),] |
| | |
| | |
| (b) | CustSpend = as.table(by(DHY\$visit_spend, DHY\$customer_id, sum)) [1 Mark] |
| | |
| | |
| (c) | CustSpend = sort(CustSpend, decreasing = TRUE) [1 Mark] |
| | |
| | |
| (d) | CustSpend = head(CustSpend, 12) [1 Mark] |
| | |
| | |
| (e) | DHYZ = DHY[(DHY\$customer_id %in% CustSpend\$customer_id),] [1 Mark] |
| | |
| | |
| (f) | + facet_wrap(~ customer_id, nrow = 3) [1 Mark] |
| | |
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Regression (10 Marks)

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A subset of the 'diamonds' data set from the R package 'ggplot2' was created. The data set reports price, size(carat) and quality (cut, color and clarity) information as well as specific measurements (x, y and z). The first 6 rows are printed below.

```
> head(dsmall)
                 cut color clarity depth table price
     carat
46434
      0.59 Very Good
                         н
                              VVS2
                                    61.1
                                            57 1771 5.39 5.48 3.32
35613
      0.30
                Good
                         Ι
                               VS1
                                    63.3
                                            59
                                                 473 4.20 4.23 2.67
43173
     0.42
             Premium
                         F
                                IF
                                    62.2
                                            56 1389 4.85 4.80 3.00
11200
     0.95
               Ideal
                         Н
                               SI1
                                    61.9
                                            56 4958 6.31 6.35 3.92
                                                 973 4.40 4.37 2.72
37189
     0.32
                         D
                              VVS1
                                    62.0
                                            60
             Premium
                                    60.7
                                            58 1689 5.17 5.21 3.15
45569 0.52
             Premium
                         Е
                               VS2
```

The least squares regression of log(price) on log(size) and color is given below. Note that 'log' in this

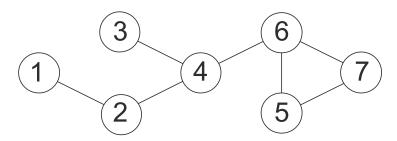
```
context means 'Log<sub>e</sub>(X).' Based on this output, answer the following questions.
 > library(ggplot2)
> set.seed(9999) # Random seed
 > dsmall <- diamonds[sample(nrow(diamonds), 1000), ] # sample of 1000 rows
 > attach(dsmall)
 > contrasts(color) = contr.treatment(7)
 > d.fit <- lm(log(price) ~ log(carat) + color)</pre>
> d.fit
> summary(d.fit)
 lm(formula = log(price) ~ log(carat) + color)
Residuals:
                10
                     Median
                                   3Q
                                           Max
 -0.97535 -0.16001 0.01106 0.15500
                                      0.99937
             Estimate Std. Error t value Pr(>|t|)
 (Intercept) 8.61356 0.02289 376.259 < 2e-16 ***
                         0.01365 127.529
                                          < 2e-16 ***
 log(carat)
              1.74075
 color2
             -0.06717
                         0.02833 -2.371
                                            0.0179 *
 color3
             -0.05469
                         0.02783 -1.965
                                            0.0496 *
             -0.07139
                         0.02770 -2.578
 color4
                                            0.0101 *
                         0.02973 -7.148 1.7e-12 ***
             -0.21255
 color5
                         0.03175 -10.393 < 2e-16 ***
 color6
             -0.32995
             -0.50842
                         0.04563 -11.143 < 2e-16 ***
 color7
Residual standard error: 0.2393 on 992 degrees of freedom
Multiple R-squared: 0.9446, Adjusted R-squared: 0.9443
F-statistic: 2418 on 7 and 992 DF, p-value: < 2.2e-16
 > contrasts(color)
   2 3 4 5 6 7
 D 0 0 0 0 0
E 1 0 0 0 0 0
 F 0 1 0 0 0 0
 G 0 0 1 0 0 0
 H 0 0 0 1 0 0
```

| eExa | m Q3 (4 Marks) |
|------|--|
| (a) | Write down the regression equation predicting log(price) as a function of size and color. [1 Mark] |
| (b) | Explain the different data types present in the variables: carat and color . What is the effect of this difference on the regression equation? [2 Marks] |
| (c) | What is the predicted price for a diamond of 1 carat of color H? [1 Mark] |
| | |
| eExa | m Q4 (6 Marks) |
| (a) | Which colour diamonds can be reliably assumed to have the highest value? Explain your reasoning. How sure can you be? [2 Marks] |
| (b) | Which colour diamonds have the lowest value? How reliable is the evidence? Explain your reasoning. [2 Marks] |
| (c) | Comment on the reliability of the model as a whole giving reasons. [2 Marks] |
| | |

Networks (10 Marks)

eExam Q5 (5 Marks)

The social network of a group of friends (numbered from 1-7) is drawn below.



Calculate the **betweenness centrality** for nodes 1 to 7. [2 Marks] (a)

1:0

2:5 3:0 4:11

5:0

- 6:8 7:0
- (b) Calculate the **closeness centrality** for nodes nodes 1 to 7. [2 Marks]

(c) Giving reasons based on your results in Parts a and b, which node is most central in the network? [1 Mark]

| eExar | fi Q6 (3 Marks) |
|-------|--|
| (a) | Calculate the density of the graph. [1 Mark] |
| | |
| (b) | Calculate the clustering coefficient of the graph. [1 Mark] |
| | |
| (c) | Calculate the diameter of the graph. [1 Mark] |
| | |
| | m Q7 (2 Marks) down the adjacency matrix for the network. [2 Marks] |
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Naïve Bayes (4 Marks)

eExam Q8 (3 Marks)

Use the data below and Naïve Bayes classification to predict whether the following test instance will be happy or not.

Test instance: (Age Range = young, Occupation = professor, Gender = F, Happy = ?)

| ID | Age Range | Occupation | Gender | Нарру |
|----|-------------|------------|--------|-------|
| 1 | Young | Tutor | F | Yes |
| 2 | Middle-aged | Professor | F | No |
| 3 | Old | Tutor | M | Yes |
| 4 | Middle-aged | professor | M | Yes |
| 5 | Old | Tutor | F | Yes |
| 6 | Young | Lecturer | M | No |
| 7 | Middle-aged | lecturer | F | No |
| 8 | Old | Tutor | F | No |

| | 8 | Old | Tutor | F | No | |
|---|------------|------------------|----------------|------------------|---------------------|------------------------|
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| e | Exam Q9 (| 1 Mark) | | | | |
| U | se the cor | nplete Naïve Bay | ves formula to | evaluate the cor | nfidence of predict | ing Happy = yes, based |
| | | | | | | pation = professor, |
| | ender = F | | 1 | (8 | <i>y y</i> 1 | 1 |
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Visualisation (6 Marks)

eExam Q10 (6 Marks)

A World Health study is examining how life expectancy varies between men and women in different countries and at different times in history. The table below shows a sample of the data that has been recorded. There are approximately 15,000 records in all.

| Country | Year of Birth | Gender | Age at Death |
|-------------|---------------|--------|--------------|
| Australia | 1818 | M | 9 |
| Afghanistan | 1944 | F | 40 |
| USA | 1846 | F | 12 |
| India | 1926 | F | 6 |
| China | 1860 | F | 32 |
| India | 1868 | M | 54 |
| Australia | 1900 | F | 37 |
| China | 1875 | F | 75 |
| England | 1807 | M | 15 |
| France | 1933 | M | 52 |
| Egypt | 1836 | M | 19 |
| USA | 1906 | M | 58 |

Using one of the graphic types from the Visualization Zoo (see formulae and references for a list of types) suggest a suitable graphic to help the researcher display as many variables as clearly as possible.

| Explain your decision. Which graph elements correspond to the variables you want to display? | |
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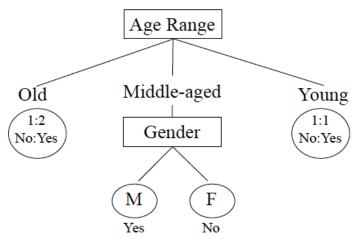
Decision Trees (10 Marks)

eExam Q11 (4 Marks)

Eight university staff completed a questionnaire on happiness. The results are given below.

| ID | Age Range | Occupation | Gender | Нарру |
|----|-------------|------------|--------|-------|
| 1 | Young | Tutor | F | Yes |
| 2 | Middle-aged | Professor | F | No |
| 3 | Old | Tutor | M | Yes |
| 4 | Middle-aged | Professor | M | Yes |
| 5 | Old | Tutor | F | Yes |
| 6 | Young | Lecturer | M | No |
| 7 | Middle-aged | Lecturer | F | No |
| 8 | Old | Tutor | F | No |

A decision tree was generated from the data.



(a) Using the decision tree generated from the data provided, assuming a required confidence level greater than 60% to classify as 'Happy', what is the predicted classification for the following instances: [2 Marks]

Instance 1: (Age Range = Young, Occupation = Professor, Gender = F, Happy = ?)

Instance 2: (Age Range = Old, Occupation = Professor, Gender = F, Happy = ?)

- (b) Is it possible to generate a 100% accurate decision tree using this data? Explain your answer. [1 Mark]
- (c) Explain how the concept of entropy is used in some decision tree algorithms. [1 Mark]

eExam Q12 (6 Marks) (a) Do you think entropy was used to generate the decision tree above? Explain your answer. [2

(b) What is the entropy of "Happy"? [1 Mark]

Marks]

- (c) What is information gain after the first node of the decision tree (Age Range) has been introduced? [2 Marks]
- (d) Explain why some decision tree algorithms are referred to as greedy algorithms. [1 Mark]

ROC and Lift (10 Marks)

eExam Q13 (4 Marks)

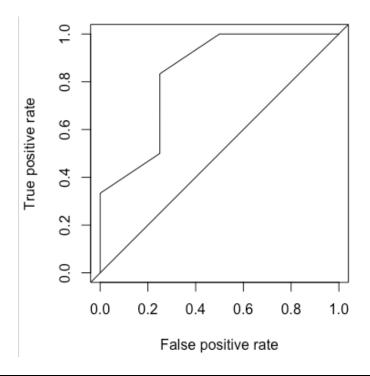
The following table shows the outcome of a classification model for customer data. The table lists customers by code and provides the following information: The model confidence of a customer buying/not buying a new product (confidence-buy); whether in fact the customer did not buy the product (buy = 1 if the customer purchased the model, buy = 0 if the customer did not buy the model).

| customer | confidence- | buy-not-buy |
|----------|-------------|-------------|
| | buy | |
| c1 | 0.9 | 1 |
| c2 | 0.8 | 1 |
| c3 | 0.7 | 0 |
| c4 | 0.7 | 1 |
| c5 | 0.6 | 1 |
| c6 | 0.5 | 1 |
| c7 | 0.4 | 0 |
| c8 | 0.4 | 1 |
| c9 | 0.2 | 0 |
| c10 | 0.1 | 0 |

| (a) | is required for a positive classification. [2 Marks] |
|-----|--|
| | |
| (b) | Calculate the True Positive Rate and the False Positive Rate when a confidence level of 80% is required for a positive classification. [2 Marks] |
| | |
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eExam Q14 (2 Marks)

The ROC chart for the previous question is shown below. Comment on the quality of the model overall. Give a single measure of classifier performance.



eExam Q15 (4 Marks)

(a) What is the lift value if you target the top 40% of customers that the classifier is most confident of? [2 Marks]

(b) Explain what the value of lift means in the previous question. [2 Marks]

Clustering (10 Marks)

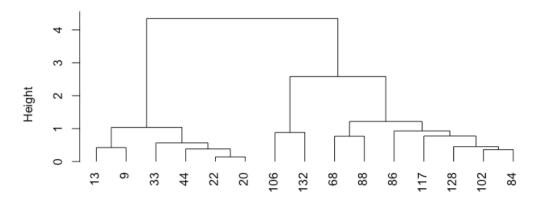
eExam Q16 (4 Marks)

A k-Means clustering algorithm is fitted to the iris data, as shown below.

```
rm(list = ls())
data("iris")
ikfit = kmeans(iris[,1:2], 4, nstart = 10)
ikfit
table(actual = niris$Species, fitted = ikfit$cluster)
Based on the R code and output below, answer the following questions.
> ikfit
K-means clustering with 4 clusters of sizes 24, 53, 41, 32
Cluster means:
  Sepal.Length Sepal.Width
      4.766667 2.891667
      5.924528 2.750943
2
3
      6.880488
                   3.097561
      5.187500 3.637500
Within cluster sum of squares by cluster:
[1] 4.451667 8.250566 10.634146 4.630000
  (between_SS / total_SS = 78.6 %)
> table(actual = niris$Species, fitted = ikfit$cluster)
         fitted
actual
             1 2
  tual 1 2 3 4 setosa 18 0 0 32
  versicolor 5 34 11 0
  virginica 1 19 30 0
       Comment on the quality of the clustering giving at least one quantitative measure. [2 Marks]
(a)
(b)
       What actions could be performed to improve the quality of the clustering? [2 Marks]
eExam Q17 (2 Marks)
For the previous question, if clustering was used to discriminate between the irises, what would be the
accuracy of the model? Explain your reasoning. [2 Marks]
```

eExam Q18 (4 Marks)

15 observations were sampled at random from the Iris data set. The dendrogram resulting from clustering, based on their sepal and petal measurements, is below.



(a) If you wanted just three clusters, which items would be in each cluster? [1 Mark]

(b) Based on the dendrogram, comment on the ease or difficulty of distinguishing between the three species of iris based on their sepal and petal measurements. Explain your reasoning with an example from the graph. [2 Marks]

(c) What does 'Height' mean in this context. [1 Mark]

eExam Q19 (2 Marks) Explain what is meant by the 'bag of words' approach to text mining. eExam Q20 (2 Marks) Apply the five main steps required to pre-process text documents for analysis to the corpus below. Write your processed documents in the space provided. Doc1 = { The church choir sang loudly. } $Doc2 = \{ \text{ The boys were singing in the church choir. } \}$ $Doc3 = \{ \text{ The boy asked to sing a song. } \}$ eExam Q21 (2 Marks) Construct the term document frequency matrix for the processed text documents above. [2 Marks]. eExam Q22 (2 Marks) Using the term document frequency matrix, calculate the Cosine Distance between each pair of documents. [2 Marks]

Text Analytics (8 Marks)

Ensemble Methods (7 Marks)

| ELXAIII UZO IZ IVIAI NO | eExam | Q23 | (2 Marl | ks |
|-------------------------|-------|-----|---------|----|
|-------------------------|-------|-----|---------|----|

(b)

| Exam Q2 | 4 (2 Marks) | | | | |
|--|--|--|-------------------------------------|----------------------------|--------------------------------|
| How do bo | posting and random f | orests differ fro | om bagging? | | |
| | | | | | |
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| Exam Q2 | 5 (3 Marks) | | | | |
| Exam Q2 | 5 (3 Marks) | | | | |
| | | NN) is to be us | ad to alossify who | than an nat ta | » Duv y a contain muchy |
| An artifici | al neural network (A | , | • | | o Buy a certain produ |
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| An artifici pased on I | al neural network (A Popularity, Sales and Popularity | Sales | . An extract of the Performance | data is below | • |
| An artifici pased on I ID 1 | al neural network (A Popularity, Sales and Popularity low | Sales 330000 | Performance 0.87 | Buy Maybe | • |
| An artificionased on I ID 1 2 | al neural network (A Popularity, Sales and Popularity low medium | Sales 330000 40000 | Performance 0.87 0.22 | Buy Maybe No | • |
| An artificionased on I ID 1 2 3 | al neural network (A Popularity, Sales and Popularity low medium low | Sales 330000 40000 50000 | Performance 0.87 0.22 NA | Buy Maybe No Yes | • |
| An artificionased on I ID 1 2 3 4 | al neural network (A Popularity, Sales and Popularity low medium low high | Sales 330000 40000 50000 30000 | Performance 0.87 0.22 NA 0 | Buy Maybe No Yes Yes | • |
| An artificionased on I ID 1 2 3 4 5 | al neural network (A Popularity, Sales and Popularity low medium low high low | Sales 330000 40000 50000 30000 100000 | Performance 0.87 0.22 NA 0 0.1 | Buy Maybe No Yes Yes No | • |
| An artifici based on I ID | al neural network (A Popularity, Sales and Popularity low medium low high low medium | Sales 330000 40000 50000 30000 100000 NA | Performance 0.87 0.22 NA 0 0.1 0.06 | Buy Maybe No Yes Yes No No | |

How many **output** nodes does the ANN require for this problem? [1 Mark]

Dirty and Tidy Data (7 Marks)

eExam Q26 (5 Marks)

The table below is an extract from the list of books in the British Library. Identify the instances of dirty data present, stating the way in which the data is dirty. One mark will be given for each correct instance up to a maximum of 6 marks.

| Identifier | Edition Statement | Place of Publication | Date of Publication | Publisher | Title | Author | Contributors |
|------------|-----------------------|----------------------|---------------------|----------------------|---|--------------------|------------------------|
| 206 | | London | 1879 [1878] | S. Tinsley & Co. | Walter Forbes. [A novel.] By A. A | A. A. | FORBES, Walter. |
| 216 | | London; Virtue & Yo | 1868 | Virtue & Co. | All for Greed. [A novel. The dedication s | A., A. A. | BLAZE DE BURY, Ma |
| 218 | | London | 1869 | Bradbury, Evans & 0 | Love the Avenger. By the author of â€o | A., A. A. | BLAZE DE BURY, Ma |
| 472 | | London | 1851 | James Darling | Welsh Sketches, chiefly ecclesiastical, to | A., E. S. | Appleyard, Ernest Si |
| 480 | A new edition, revis | London | 1857 | Wertheim & Macint | [The World in which Hive, and my place | A., E. S. | BROOME, John Hen |
| 481 | Fourth edition, revis | London | 1875 | William Macintosh | [The World in which Hive, and my place | A., E. S. | BROOME, John Hen |
| 519 | | London | 1872 | The Author | Lagonells. By the author of Darmayne (| A., F. E. | ASHLEY, Florence Er |
| 667 | | pp. 40. G. Bryan & C | o: Oxford, 1898 | | The Coming of Spring, and other poems | A., J. A., J. | ANDREWS, J Write |
| 874 | | London] | 1676 | | A Warning to the inhabitants of England | Remaʿ. | ADAMS, Mary. |
| 1143 | | London | 1679 | | A Satyr against Vertue. (A poem: suppo | A., T. | OLDHAM, John. |
| 1280 | | Coventry | 1802 | Printed by J. Turner | An Account of the many and great Loan | ¥ | CARTE, Samuel. JAC |
| 1808 | | Christiania | 1859 | | Erindringer som Bidrag til Norges Histor | AALL, Jacob. | AALL, J. C. LANGE, C |
| 1905 | | Firenze | 1888 | | Gli Studi storici in terra d'Otranto Fra | AAR, Ermanno - pse | S., L. G. D. SIMONE, |
| 1929 | | Amsterdam | 1839, 38-54 | | De Aardbol. Magazijn van hedendaagsc | t | WITKAMP, Pieter Ha |
| 2836 | | Savona | 1897 | | Cronache Savonesi dal 1500 al 1570 | ABATE, Giovanni Ag | ASSERETO, Giovann |
| 2854 | | London | 1865 | E. Moxon & Co. | See-Saw; a novel Edited [or rather, w | ABATI, Francesco. | READE, William Win |
| 2956 | | Paris | 1860-63 | | Geì odeì sie d'une partie de la Haute E | ABBADIE, Antoine T | RADAU, Rodolphe. |
| 2957 | | Paris | 1873 | | [With eleven maps.] | ABBADIE, Antoine T | |
| 3017 | Nueva edicion, anot | Puerto-Rico | 1866 | | [Historia geograì fica, civil y politica de | ABBAD Y LASIERRA, | ACOSTA Y CALBO, Jo |
| 3131 | | New York | 1899 | W. Abbatt | The Crisis of the Revolution, being the s | ABBATT, William. | ANDREÌ, John - Ma |
| 4598 | | Hull | 1814 | The Author | Peace: a lyric poem. [With prefatory ad | ABBOTT, Thomas Ea | WRANGHAM, Franc |
| 4884 | | London | 1820 | J. Hatchard & Son | Abdallah; or, The Arabian Martyr: a Chr | | BARHAM, Thomas F |
| 4976 | [Another edition.] A | Oxonii | 1800 | J. Cooke, etc. | [Abdollatiphi Historiæ Ægypti comper | • | WHITE, Joseph - Car |
| 5382 | | London | 1847, 48 [1846-48] | Punch Office | The Comic History of England With | A'BECKETT, Gilbert | LEECH, John - Artist |
| 5385 | [Another edition.] Il | London | [1897?] | Bradbury, Agnew & | [The comic history of England With to | A'BECKETT, Gilbert | LEECH, John - Artist |
| 5389 | [Another edition.] | London | [1897?] | Bradbury, Agnew & | [The Comic History of Rome Illustrate | A'BECKETT, Gilbert | LEECH, John - Artist |
| 5432 | | Milano | 1893 | | Signa: opera in tre atti [founded on the | A'BECKETT, Gilbert | MAZZUCATO, Giova |
| 6036 | | London | 1805 | C. & R. Baldwin | The Venetian Outlaw, a drama in three | | ELLISTON, Robert W |
| 6821 | | Aberdeen | 1837 | J. Davidson & Co. | Description of the Coast between Aber | d . | DUNCAN, William - |
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eExam Q27 (5 Marks)

The table below shows the English and Maths results for some students in Semester 1 and Semester 2. Rewrite the table as tidy data in the space provided.

| Student | English S1 | English S2 | Maths S1 | Maths S2 |
|---------|------------|------------|----------|----------|
| Alice | 80 | - | 56 | 78 |
| Billy | - | 54 | - | 55 |
| Carly | 6 | 77 | - | 13 |

A Tour Through the Visualization Zoo -**Summary of Graphic Types**

Time-Series Data

- Index Charts
- Stacked Graphs
- Small Multiples
- Horizon Graphs

Statistical Distributions

- Stem-and-Leaf Plots
- Q-Q Plots
- **SPLOM**
- Parallel Coordinates

Maps

- Flow Maps
- Choropleth Maps
- Graduated Symbol Maps
- Cartograms

Hierarchies

- Node-Link diagrams
- Adjacency Diagrams
- **Enclosure Diagrams**

Networks

- Force-Directed Layouts
- Arc Diagrams
- Matrix Views

Entropy

If S is an arbitrary collection of examples with a binary class attribute, then:

$$Entropy(S) = -P_{C1}log_2(P_{C1}) - P_{C2}log_2(P_{C2})$$

$$= -\frac{N_{C1}}{N} log_2\left(\frac{N_{C1}}{N}\right) - \frac{N_{C2}}{N} log_2\left(\frac{N_{C2}}{N}\right)$$

where C1 and C2 are the two classes. P_{C1} and P_{C2} are the probability of being in Class 1 or Class 2 respectively. N_{C1} and N_{C2} are the number of examples in each class. N is the total number of examples.

Note:
$$log_2 x = \frac{log_{10}x}{log_{10}2} = \frac{log_{10}x}{0.301}$$

Information gain

The Gain(S, A) of an attribute A relative to a collection of examples, S, with v groups having $|S_{\nu}|$ elements is:

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

Networking

Closeness Centrality: $C_{CL}(v) = \frac{1}{\sum_{u \in V} dist(u,v)}$

Betweenness Centrality: $C_B(v) = \sum_{s \neq t \neq v \in V} \frac{\sigma(s,t|v)}{\sigma(s,t)}$,

where (s,t) is the number of shortest paths between s

(s,t|v) is the number of shortest paths between s and t passing through v

Density: $den(g) = \frac{|E_g|}{|v_o|(|v_o|-1)/2}$,

where $|E_a|$ is number of edges, $|V_a|$ is number of vertices

Clustering coefficient: $clt(g) = \frac{3\tau_{\Delta}(g)}{\tau_{\Delta}(g)}$,

where $3\tau_{\Delta}(g)$ is number of triangles, $\tau 3(g)$ is number of connected triples

Naïve Bayes'

For events $A_1, A_2, ..., A_n$ and event C, classification probability is

$$P(C_j|A_1 \cap A_2 \dots \cap A_n) = \frac{P(C_j) \cdot P(A_1 \cap A_2 \dots \cap A_n|C_j)}{P(A_1 \cap A_2 \dots \cap A_n)}$$

For Bayesian classification, a new point is classified to C_i if $P(C_i) * P(A_1|C_i) * P(A_1|C_i) * ... * P(A_n|C_i)$ is maximised.

Naïve Bayes assumes $P(A \cap B) = P(A) * P(B)$ etc.

Cosine or normalised dot product

For documents i and j with terms w

$$Sim(D_i, D_j) = \frac{\sum_{t=1}^{N} w_{it} * w_{jt}}{\sqrt{\sum_{t=1}^{N} (w_{it})^2 * \sum_{t=1}^{N} (w_{jt})^2}}$$

$$ROC$$

$$TPR = \frac{TP}{TP + FN}, \quad FPR = \frac{FP}{FP + TN}$$

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