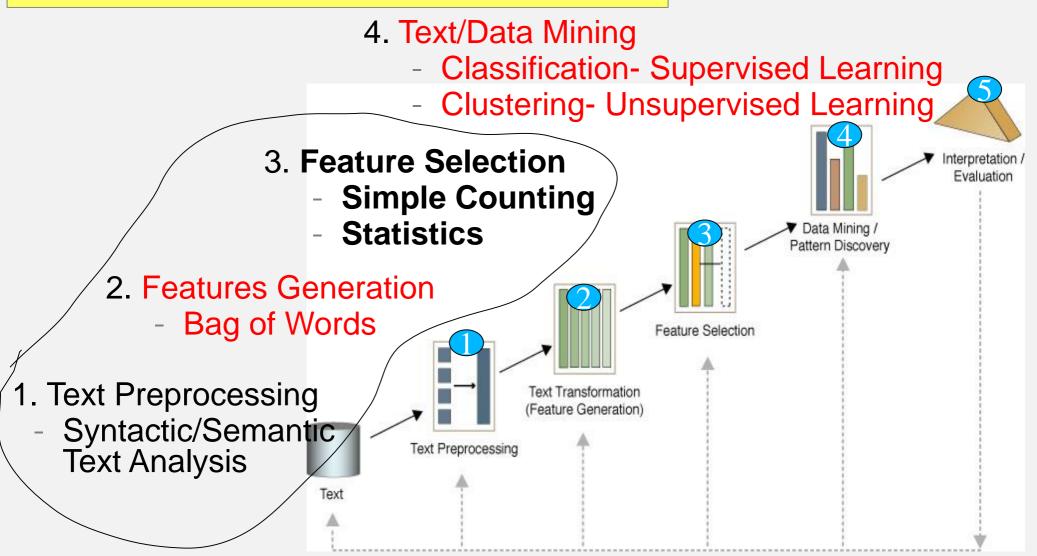
# Hons. Degree in Computing H4016 Text Analysis

Unit 4 – Generating test and training datasets

### Context

- To-date we have covered the theory relevant to steps 1, 2 and 3. This Unit looks at practical issues arising from the application of that theory.
- 5. Analyzing Results



## **Objectives**

Understand the relationship between a training process and applying that model to new document vectors

Cross validation splits a dataset into test and training datasets, both datasets have been prepared in the same way.

This Unit covers applying a model to a <u>new</u> sample of documents, i.e. documents that were not used in the training process.

G. Gray

## The following exercise appears in lab sheet #3:

Below are two processes, one to prepare data for cross validation / training; the other is to prepare new data that the model can be applied to (test process).

Where there are differences between the two processes, explain if that will effect the results when applying the model to new data? If the Dictionarystemmer was used in the test process rather than the training process, would that be OK?

#### Process for training dataset

- Process document
   (texts=../training;
   purnebelow=3;
   vector\_creation=TFIDF; )
  - ◆ StringTokeniser
  - ◆ EnglishStopWordFilter
  - ◆ DictionaryStemmer
  - ◆ PortersStemmer
- Output wordlist
- . . .generate model

#### Process for test datasets

- Root
  - Process document create wordlist from saved file
  - Process document

```
(texts=../test;
purnebelow=-1;
vector_creation=occurences;
word_list=lab3.txt)
```

- ◆ StringTokeniser
- ◆ StopWordFilterFile
- ◆ LovinsStemmer
- . . . test model

## Processes explained

- The training process uses the 15 texts from lab#2 to generate a decision tree that can predict if a text is about healthcare, crime or kenya based on keywords present in the 15 texts.
- ◆ The test process takes the model (e.g. decision tree) generated in the training process, and applies it to the 4 'unseen' documents. The decision tree will classify each one as being about either healthcare, crime or kenya.
- We can verify its accuracy based on how it classify's the 4 documents.

- Prunebelow=3 removes all term with less than three characters
- Prunebelow=-1 does not prune any terms

# Output word list / input word list

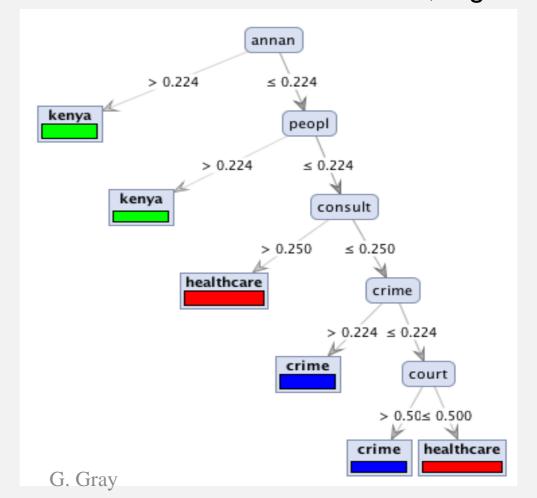
The final list of attributes resulting from the training process is as

follows:

word annan clash consult court crime former hospit hse kofi peopl rift right un vallei western

These are the attributes (column headings) that appear in the document vector.

Learners generated from this document vector will be based on some or all of these terms, e.g.



# Output word list / input word list

◆ If the model (e.g. Decision tree on the last slide) is to be tested on documents not yet processed, the document vector generated from the test documents MUST have the same column headings as the document vector generated by the training data.

Why?

• The decision tree on the last slide only used 5 of the 15 attributes. Other learners, such as k-NN or a neural network, would use all 15 attributes.

G. Grav

hospit Neural network for the western 15 documents. Darker Weights: 0.741 0.997 links refer to more <u>ke</u>nya 0.190 -3.205influential terms (high 0.185 positive or high negative -1.789-1.467values) -2.078-1.700

#### Document vector of training dataset

							/								
	annan	clash	consult	court	crime	former	hospit	hse	kofi	peopl	rift ▼	right	un	vallei	western
(	)	0	0	0	0	0	0	0	0	0.500	0.500	0	0	0.500	0.500
(	0.459	0.229	0	0	0	0.229	0	0	0.229	0.229	0.459	0.229	0.229	0.459	0.229
(	)	0.447	0	0	0	0	0	0	0	0.447	0.447	0	0	0.447	0.447
(	)	0	0	0	1	0	0	0	0	0	0	0	0	0	0
(	)	0	0	0	0.447	0	0	0	0	0	0	0.894	0	0	0

#### Document vector of new dataset, without using an input wordlist

due	effort	execut	feud	former	fridai	galwai	gardaì	gestur	hand	heroin	hse	huge	investig	issu
0	0	0.248	0	0	0	0	0	0	0	0	0.248	0.248	0	0.248
0	0.160	0	0.160	0.160	0	0	0	0.160	0.160	0	0	0	0	0
0	0	0	0	0	0	0.295	0	0	0	0	0	0	0.295	0
0.198	0	0	0	0	0.198	0	0.198	0	0	0.198	0	0	0	0

#### Document vector of new dataset using the wordlist from the training process

. (annan)	clash	consult	court	crime	former	hospit	hse	kofi	peopl	rift	right	un	vallei	western
0	0	0	0	0	0	0	1	0	0	0	0	0	0	0
0.500	0	0	0	0	0.500	0	0	0.500	0	0	0	0.500	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0	0	0	0	0	0	0

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#### Vector Creation. . .

Diferrences in a Decision tree's rules for Vector creation=TFIDF versus Vector creation=occurrences

## Occurrences

```
TFIDF
```

## Remaining operators

- Tokenisers: What would be the impact of one process using an n-gram tokeniser?
- Filters: Does it matter that different filters were used?
  - Filter Stopword (English) ?
  - Filter Stopword (Dictionary) ?
  - Filter Tokens by Length?
- Stemmers: Does it matter that different stemmers were used?
  - DictionaryStemmer?
  - Lovins Stemmer? (keny, peopl, valle, hospit)
  - Porters Stemmer? (kenya, peopl, vallei, hospit)

## Summary

- ◆ The document vector created during the training process MUST match the document vector created during the test process
  - Outputting the wordlist in the training process, and inputting the wordlist again in the test process will create identical columns headings
- Need to ensure <u>tokens</u>, and how they are <u>counted</u> also match in both preparation blocks.