Project Stage 1 CS 839, Spring 2019

Team members:

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Entity type:

Person Names like: "Norman Mineta" --full name, "Boyd"--surname and "Jamelia" --name.

More specifically, we consider as person names: full names, surnames or even only the name without the surname.

Dataset: http://mlg.ucd.ie/datasets/bbc.html

It includes articles from 5 different categories: News, Sports, Entertainment, Politics, Business and Tech.

Total Number of Files we extracted:

300 files

Total number of mentions:

3891 in 300 files

Number of files in Set I and mentions:

200 files 2652 mentions

Number of files in Set J and mentions:

100 files

1239 mentions

Cross Validation Results:

Recall: 0.787704130644

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-----SVM results-----
True Positives: 757
True Negatives: 2443
False Positives: 232
False Negatives: 284
Total number of data: 3716
Accuracy: 0.861141011841
Precision: 0.765419615774
Recall: 0.727185398655
F1: 0.745812807882
-----RF results-----
True Positives: 570
True Negatives: 2568
False Positives: 107
False Negatives: 471
Total number of data: 3716
Accuracy: 0.844456404736
Precision: 0.841949778434
Recall: 0.547550432277
F1: 0.663562281723
-----DT results-----
True Positives: 820
True Negatives: 2478
False Positives: 197
False Negatives: 221
Total number of data: 3716
Accuracy: 0.887513455328
Precision: 0.806293018682
Recall: 0.787704130644
F1: 0.796890184645
-----LogR results-----
True Positives: 743
True Negatives: 2465
False Positives: 210
False Negatives: 298
Total number of data: 3716
Accuracy: 0.86329386437
Precision: 0.779643231899
Recall: 0.713736791547
F1: 0.745235707121
-----LinearR results-----
True Positives: 1031
True Negatives: 596
False Positives: 2079
False Negatives: 10
Total number of data: 3716
Accuracy: 0.437836383208
Precision: 0.331511254019
Recall: 0.990393852065
F1: 0.496747771621
Best classifier M --> Decision Tree
F1: 0.796890184645
Precision: 0.806293018682
```

More organised:

Classifier	Precision	Recall	F1
Decision Tree(DT)	0.8062	0.7877	0.7968
Random Forest(RF)	0.8419	0.5475	0.6635
SVM	0.7654	0.7271	0.7458
Linear Regression	0.3315	0.9903	0.4967
Logistic Regression	0.7796	0.7137	0.7452

The best classifier (M) after the cross validation is the Decision Tree:

Classifier M	Precision	Recall	F1
Decision Tree	0.8062	0.7877	0.7968

Debugging Results:

Before debugging:

-----DT results-----

True Positives: 452 True Negatives: 1746 False Positives: 99 False Negatives: 342

Total number of data: 2639 Accuracy: 0.832891246684 Precision: 0.820326678766 Recall: 0.569269521411

F1: 0.672118959108

After debugging:

-----DT results-----

True Positives: 601 True Negatives: 1732 False Positives: 113 False Negatives: 193

Total number of data: 2639 Accuracy: 0.884046987495 Precision: 0.841736694678 Recall: 0.756926952141

F1: 0.797082228117

More organised:

Debugging	Precision	Recall	F1
Before	0.8203	0.5692	0.6721
After	0.8417	0.7569	0.7970

Final classifier X (after debugging):

Precision	Recall	F1
0.8417	0.7569	0.7970

Note: After each debugging stage the Decision Tree classifier remained the best classifier in the Cross Validation process.

Rule-Based post-processing:

In order to create our post-processing rules we used the False Positives examples. We created a list with 56 words which can be found in directory "/code" in file "rules.csv". These words are common words that were falsely classified as names (ex: Bridge, Indoor, Professional, White, Five, Six etc.)

Final Results:

Decision Tree results after rule-based post-processing:

-----DT results-----

True Positives: 927
True Negatives: 2853
False Positives: 157
False Negatives: 312

Total number of data: 4249 Accuracy: 0.889621087315 Precision: 0.855166051661 Recall: 0.74818401937 F1: 0.798105897546

More organised:

Precision	Recall	F1
0.8551	0.7481	0.7981

Why we didn't reach the required Precision?

Almost all our False Positives are person names that were mistakenly **NOT** tagged from us, during the labeling process, as person names. However, the classifier predicts them correctly as person names but we cannot accept them as True Positives since they are not tagged with the person>.../> tag during the labeling process. This means that we need to be very careful during the labeling process.

Features we used:

Our candidates are 1-grams, 2-grams and 3-grams with capitals in the beginning which are also not stop-words.

The features we extracted are:

- 1. Is candidate a 1-gram? (0 or 1)
- 2. Is candidate a 2-gram? (0 or 1)
- 3. Is candidate a 3-gram? (0 or 1)
- 4. Is candidate a Noun? (0 or 1)
- 5. Is candidate a Verb? (0 or 1)
- 6. Is candidate an Adjective? (0 or 1)
- 7. Is candidate the first word of a sentence? (0 or 1)
- 8. Is candidate followed by "'s"? (0 or 1)
- 9. Is candidate followed by a verb? (0 or 1)
- 10. Is there a prefix (ex: Mr, Mrs etc.) before the candidate? (0 or 1)
- 11. Does candidate include countries or cities? (0 or 1)
- 12. Does candidate include dates? (0 or 1)
- 13. Does candidate include words associated with sports (ex: Dynamo Kiev, Manchester United etc..)? (0 or 1)
- 14. Number of capitals in candidate Numeric
- 15. Number of vowels in candidate Numeric
- 16. Number of consonants in candidate Numeric
- 17. Has the candidate symbols? (ex: -, \$, ! etc..) (0 or 1)
- 18. Has the candidate numbers?- (0 or 1)