Authorship Attribution in Greek Tweets Using Author's Multilevel N-gram Profiles

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A brief typology of authorship research

- Authorship attribution: Closed problem. We assume that one of 1, 2, 3... n candidates is the real author of a text.
- Author verification: Open problem. We assume an open set of authors and each text should be attributed to its real author without reference to any corpus from other authors.
- Author profiling: Closed problem. We assume that specific extralinguistic characteristics (gender, age, psychological profile etc.) of the author(s) can be traced in his/her texts.

Aims of the present research

- To perform authorship attribution experiments in tweets written in Modern Greek.
 - Create the first Modern Greek Tweets Corpus (GTC) in order to use it as a reference corpus for studying social media language including authorship attribution, sentiment analysis and linguistic variation.
 - Explore the effectiveness of a specific document representation called Author's Multilevel N-gram Profile (AMNP), which comprises of a combined vector of increasing size and different level n-grams
 - Investigate alternative ways to construct training sets for authorship attribution in Twitter data. More specifically:
 - Is it better to use single tweets for testing or do we need to merge tweets producing bigger text units?
 - In the case of merging tweets, what is the text size that produces the best attribution results?

The Greek Twitter Corpus

| Authors | No of Tweets | Total size (words) | Average size (words) | Standard Deviation |
|---------|-----------------|-----------------------|----------------------------|-----------------------|
| Α | 500 | 5,378 | 10.75 | 5.42 |
| В | 918 | 10,515 | 11.45 | 5.52 |
| С | 2,065 | 32,098 | 15.54 | 6.73 |
| D | 455 | 7,451 | 16.57 | 5.48 |
| E | 1,347 | 9,822 | 7.29 | 5.01 |
| F | 535 | 3,692 | 6.90 | 4.93 |
| G | 1,277 | 9,412 | 7.37 | 5.63 |
| Н | 2,306 | 26,212 | 11.36 | 5.86 |
| I | 2,986 | 18,720 | 6.26 | 4.28 |
| J | 584 | 7,618 | 13.06 | 6.74 |
| Total | 12,973 | 130,918 | | |

Author's Multilevel N-gram Profile - AMNP



Experimental methodology

- We used GTC to create 4 different datasets which contained merged tweets of increasing text sizes (25, 50, 75 and 100 words).
- As classification algorithm we used multiclass support vector classification (LIBLINEAR library).
- We tested the authorship attribution accuracy with each feature group separately and compared it with AMNP. Accuracy figures were calculated on two different conditions:
 - a) 10-fold cross-validation (cv) in the merged tweets text units
 - b) External test set which contained 500 single tweets not included in the training set (35-60 per author).



Accuracy vs. text size chunks using two different validation methods

Impact of text size and feature representation method in authorship attribution accuracy using cross-validation



Impact of text size and feature representation method in authorship attribution accuracy using external dataset



Conclusions

- Authorship attribution in tweets of Modern Greek is a feasible task. Our top performance (0.952 accuracy in 10-fold cv using 100-word text chunks) is a good indication that the tweet's linguistic structure is a significant carrier of authorship information.
- AMNP representation is based on a solid linguistic semiotic theoretical background and proved highly efficient compared to single n-gram feature groups in all text sizes.
- The obtained results indicated that optimal performance is achieved when **both** training and testing sets for authorship attribution contained merged tweets.

Thank you...