Authorship Attribution



TECHNISCHE UNIVERSITÄT DARMSTADT

English for Computer Science II

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Overview



- Motivation
- How does it work?
- Take-Home-Messages
- Discussion
- References





• In this day and age there is an incredible amount of information worldwide

Quote [1]:

"The Amount of Digital Information Reached 281 Exabytes (281 Billion Gigabytes)".

• Rough estimate: ~ 85% of these are available in a textual representation





- As far as we know there are ~ 7.000.000.000 humans on earth
- From this it follows that there must be many (different) authors, who have produced these textual information !
- Sometimes it is not clear **which** author wrote a specific text
- In order to determine an author of an unknown text, so-called "authorship attribution methods" might be helpful
- These methods offer a wide range of applications...





Applications:

- Plagiarism detection (remember the Guttenberg Affair ?)
- Forensic evidence in court (verifing confessions)
- Unmasking pseudonymous authors (e.g. in terror extremist blogs)
- Finding additional material of the same author
- Categorization of texts by authors (e.g. in unstructured text collections)
- And many more...





- Authorship Attribution is an extensively researched topic
- Researchers claim: task is not far from being solved

(...for some scenarios)

Quote [2]:

"Trying to classify an unseen text as being written by one of two or of a few authors is a relatively simple task, which in most cases can be solved with high reliability and accuracies over 95%".





- But, how is it actually possible to recognize **who** wrote a given text?
- Dozens of techniques have been proposed to answer this question...
- Due to a lack of time, we will focus only on one 😳





Profile-based approach (proposed in: [5])

- Assume we have a document collection: $\mathbf{D} = \{ D_1, D_2, D_3, \dots \}$
- That have been produced by several authors: $A = \{ A_1, A_2, A_3, \dots \}$
- Merge both into clusters: $\mathbf{C} = \{ (A_1, D_3), (A_2, D_4, D_5), (A_3, D_7, D_8, D_{12}), \dots \}$







Profile-based approach (proposed in: [5])

• Build for each document cluster C_i one big textfile T_i







Profile-based approach:

- What is the reason we should do that?
- Imagine we collect from an author A₁ documents like: private e-Mails, scientific papers, blogs, reports, ...
- Merging these texts will lead to an abstraction of style variation
- Specific style patterns of A_1 remain in T_i (the big textfile)
- We need a modell to find these patterns...











- Sounds simple? Let's have a look behind the scenes...
- If we want to discriminate authors (represented thorugh ${\bf T}_{\rm i}$) we first must understand how to distinguish their style
- Bad news \rightarrow there is no definition for style \bigodot
- However, style can be approximated through a combination of various **features**
 - vocabulary richness
 - average word/sentence length
 - number of specific symbols (-.:,#?!'&)
 - number of: adjectives, nouns, verbs, …

...





 $\mathbf{F_{i}} = \begin{bmatrix} 7 \\ 54 \\ 109 \\ ... \end{bmatrix}$

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- What we have so far: $\mathbf{T}_1, \mathbf{T}_2, \mathbf{T}_3, \dots$
- Applying feature extraction on the $\mathbf{T}_{\mathbf{i}}$ will result in the so-called feature vectors:

- Task of the model: find the most similar ${\bf F}_{\rm i}$ in comparsion to the features of the unknown text document
- most similar = shortest distance (e.g. in a vector space...)



TECHNISCHE Recap: Linear Algebra UNIVERSITÄT DARMSTADT **F**₂ The unknown 3 text = U ß ×Х



Recap: Linear Algebra



- The angle $\boldsymbol{\beta}$ represents similarity between 2 vectors
- Can be computed as fallows:

$$\cos(\boldsymbol{\beta}) = \frac{\mathbf{F}_{i} * \mathbf{U}}{\|\mathbf{F}_{i}\| * \|\mathbf{U}\|}$$

• Standardized resulting number is between: [0;1]





Recap: Linear Algebra



• Besides the cosine similarity many other "metrics" are typically used



• All of these metrics share the same idea:

 \rightarrow Figure out if two vectors correlate with each other !



Take-Home-Messages



Authorship Attribution:

... is no longer a utopian vision !

... is useful in many scenarios beyond Computer Science !

... is something that could affect you too (hopefully not)



Thanks for your attention !

















Your turn ;-)



- A couple of features have been mentioned during the presentation, can you think of additional features?
- Do you think that symbol-related features (number of hyphens, commas, etc.) are useful to discriminate the style of authors?
- Imagine we have 1000 features to train the modell, how should we handle less useful features without discarding them?
- Name at least one application where you could use Authorship Attribution for your own purpose



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