

Did the Names I Used within My Essay Affect My Score? Diagnosing Name Biases in Automated Essay Scoring

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Onomastics – the study of names

Names and (human) essay grading

Names and (automated) essay grading

Results and conclusions



What are Onomastics?

Onomastics is the study of proper names

- This can be in a wide variety of contexts
 - Etymological
 - Historical
 - Social

Names carry social and cultural context



Names Have Power

- We know that proper names affect how people are perceived
 - In job applications (Åslund and Skans 2012)
 - During grading (Anderson-Clark et al. 2008)
 - When looking to rent (Carpusor and Loges 2006)
 - Among many others...
- This can be an issue when dealing with highstakes situations



Checking for Human Biases

 We know that humans have implicit biases (Greenwald et al. 1998)

 These can reflect on how we perform on our day-to-day tasks

 On high-stakes situations, this can lead to undesirable results

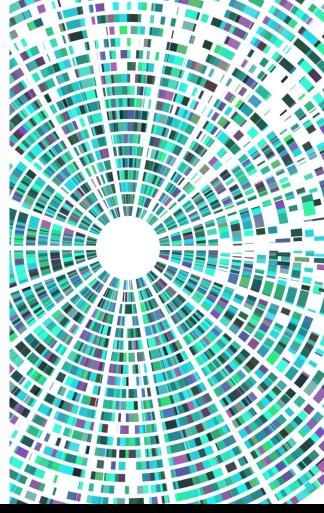


Biases in Essay Grading

Essay grading can be a high-stakes situation

 Students should be graded based on their knowledge and skills

 Discovering and acknowledging biases can reduce the impact they have



Assessing Names? (Aldrin 2017) – Design

Take an essay where a given name appears once

- The topic was "my childhood"
- The language of the essay was Swedish

Select three names with different sociocultural implications

- Carl, commonly associated with higher economic status
- Kevin, commonly associated with lower economic status
- Mohammed, an ethnically marked Muslim name

Substitute the names on the original essay

• This leads to three different versions of the essays

Randomly give a professional grader one of these three versions

• 113 high school teachers across Sweden graded the essays

Assessing Names? (Aldrin 2017) – Results

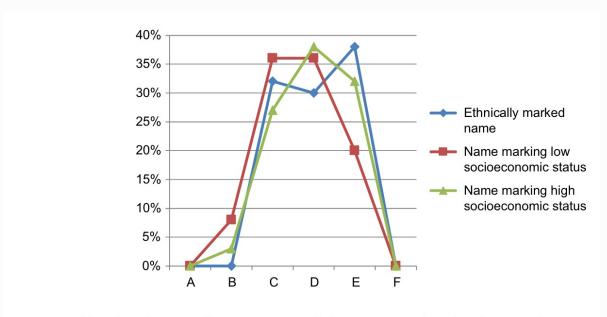
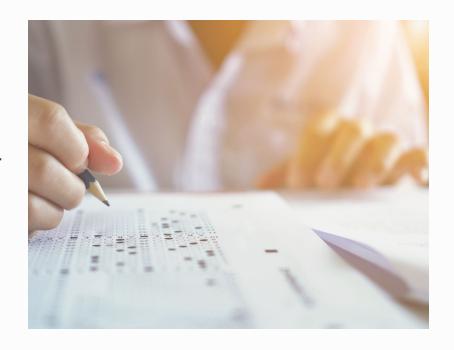


FIGURE 1 Teachers' general assessment of the text correlated to inserted name.

From "Assessing Names? Effects of Name-Based Stereotypes on Teachers' Evaluations of Pupils' Texts" by Aldrin (2017) [Link]

Assessing Names? (Aldrin 2017) – Conclusions

- The quantitative differences were small and not statistically significant
- The essay version with the Muslimmarked name
 - Tended to get lower grading across all rubrics
 - It also got the most comments on its deficiencies across three dimensions



PART 3 I Have As Many Names As There Are Winds

Getting computers involved in essay grading

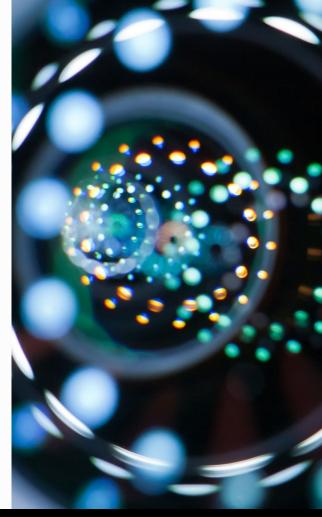
Bias in Machine Learning

- Al looks at insane amounts of data to learn
- It does so by looking at patterns and exploiting them
- However, human biases are reflected as patterns in the data
- This can affect the fairness of Al models



Al for Second Language Evaluation

- The task
 - Given a second-language learner's essay, determine the CEFR level it belongs to
- Several ways to do it
 - Extract linguistic features + classical ML
 - Language models (BERT or GPT)
 - Smaller models to test different things



Measuring Fairness

- A model is fair if it performs equally for different subgroups
- An essay with a Swedish name in its text should be graded the same as the same essay with an Arabic name in its text

 If we find biases in one or more models, we can explore where they come from



Our Models

- Feature-based model (Pilán et al. 2016; Volodina et al. 2016b)
 - Uses lexical, morphological, syntactic and semantic features
 - We expected to find little to no bias at all
- Swedish BERT (Malmsten et al. 2020)
 - Learns syntax and semantics through context
 - This means it might have picked up biases during any stage of its training process



Checking for Human-Like Biases

- Take 9 essays
 - Two for each CEFR level except for C1 (one essay) and C2 (no essays)
 - From the Swell-Pilot corpus of L2 Swedish learner essays



Checking for Human-Like Biases

Take 9 essays

- Generate a list of 20 names, for each of four ethnic groups
 - Swedish
 - Finnish
 - Anglo-American
 - Arabic



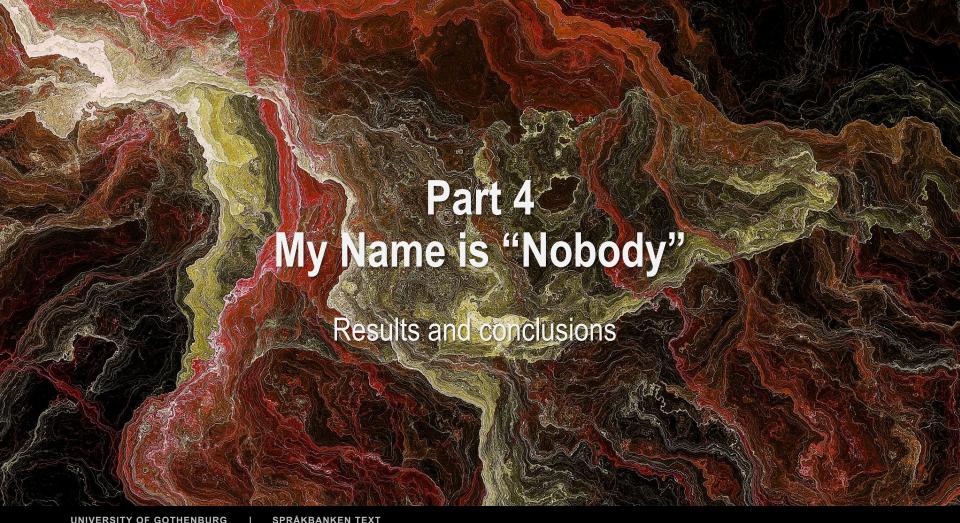
Checking for Human-Like Biases

Take 9 essays

 Generate a list of 20 names, for each of four ethnic groups

 Substitute a given name in the original essay for one on the list





Performance on the Test Set

Model	Accuracy	F1 Macro	F1 Weighted
Feature-Based	0.25	0.08	0.1
BERT	0.66	0.65	0.65

Performance on the Diagnostic Set

	Feature-Based		BERT	
Name Groups	Accuracy	Recall	Accuracy	Recall
Swedish	0.14	0.20	0.86	0.60
Finnish	0.14	0.20	0.86	0.60
Anglo-American	0.14	0.20	0.86	0.60
Arabic	0.14	0.20	0.86	0.60

What Does this Mean?

 Changing a single name within an essay did not change the models' performance

 The performance does not change either when taking (binary) gender into account

 This is what we would expect from a fair automated essay assessment system



Things to Keep in Mind

• The sample size from which the diagnostic set was generated is small

 We did not account whether the names were in BERT's dictionary or not

 This does not mean that neither the models nor the data are free of biases



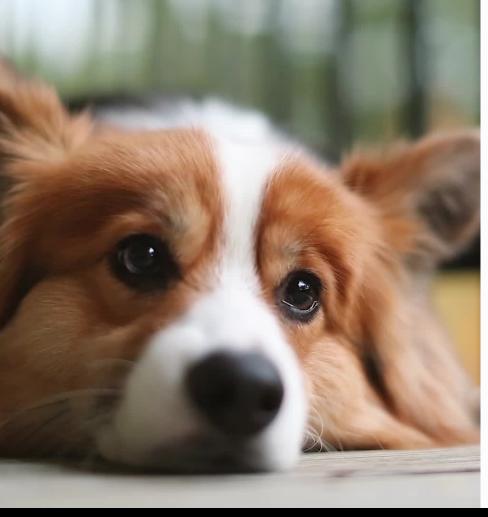
Changing Names with Fairness in Mind

 We need to make sure these names don't affect the outcome of automated systems

 There is a trade-off between privacy, fairness, and performance

In the end we're doing this to support people





Dogs have human names. It's what keeps them from being wolves.

- T. Kingfisher, Nettle & Bone





SPRÅKBANKENTEXT

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Section Titles

- Section 1 most famously from Romeo and Juliet, a play by Shakespeare
- Section 2 is title of a book by André Aciman, later adapted into film by Luca Guadagnino
- Section 3 is from American Gods, a book from Neil Gaiman
- Conclusion is the Pseudonym taken by Odysseus when talking to the cyclops Polyphemus in the Odyssey, an epic poem by Homer