Language detection using character n-gram profiles Inspiration from Cavnar and Trenkle (1994)

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Overview

- Introduction
- Methodology
- **3** Results
- 4 Discussion
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Motivation

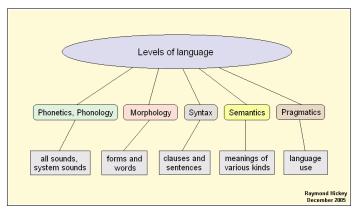


Figure 1: Levels/structures of languages; figure taken from Hickey (2005)

- Morphological profiling probably has lower data and compute requirements
- Makes sense given no external libraries are allowed

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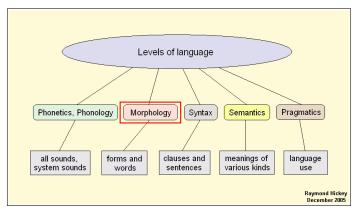


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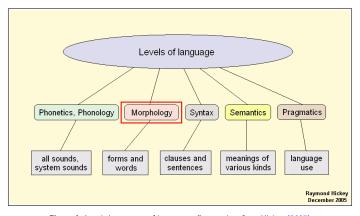


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Methodology

- Character n-gram profiling technique from Cavnar, Trenkle, et al. (1994)
- WiLI-2018 data set for 235 languages with 235,000 paragraphs (Thoma, 2018)
- Similarities: Data is lowercased and punctuations/special-tokens are removed
- Differences: Use vector-based difference norm instead of out-of-place distance
- Two hyperparameters: character n-gram length and ranked n-gram cutoff

N-Gram-Based Text Categorization

William B. Cavnar and John M. Trenkle Environmental Research Institute of Michigan P.O. Box 134001 Ann Arbor MI 48113-4001

Figure 2: Excerpt from Cavnar, Trenkle, et al. (1994)

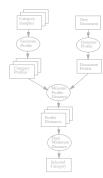


Figure 3: Flowchart from Cavnar, Trenkle, et al. (1994)

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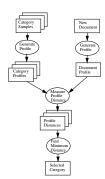


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Results

N-gram length	N-gram cutoff	Weighted Test F ₁	Best Language	Worst Language
2	100	0.865	Navajo	Konkani
2	300	0.893	Navajo	Pampanga
3	100	0.859	Dhivehi	Chavacano
3	300	0.898	Navajo	Chavacano

Table 1: Tabular summary of model performances; MLP from Thoma (2018) achieved an accuracy of 0.883

Language	N_1	N_2	N_3	\mathbb{N}_4	N_5
English	the	and			ent
Deutsch	der				
Italiano			ent		lla

Table 2: Tabular summary of top five character trigrams with highest relative frequency per language

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Discussion

Gold language	Utterance	Predicted language
English	What is this?	Cantonese
Deutsch	Was ist das?	Chavacano
Italiano	Cos'è questo?	Asturian

Table 3: Examples of erroneous language detection for short phrases

Conclusions

Bibliography I

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