Small Dose, Big Effect: Leveraging Synthetic Data for Candy Speech Detection



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The "What & Why"

- Motivation

- While most research focuses on censoring negative speech, the active promotion of positive online discourse is often overlooked.
- Our goal is to detect "Candy Speech" to help foster more supportive and empathetic online communities.

What is 'Candy Speech'?

- Definition: An expression of positive attitudes on social media toward individuals or their output (videos, comments, etc.). It's the opposite of hate speech. Examples:
- Weiter so daumen hoch :)" (Keep it up, thumbs up:))
- "Ihr seid einfach der Hammer!" (You guys are just awesome!)
- "Echt geiles Lied " (Really cool song)

• Source: German-language YouTube

Key Challenge: The dataset is highly

The "How"

- Methods Explored

. Data Preprocessing:

- 1.Initial tests with aggressive cleaning (lemmatization, etc.) hurt performance.
- 2. Final models used raw text.

. Model Architectures:

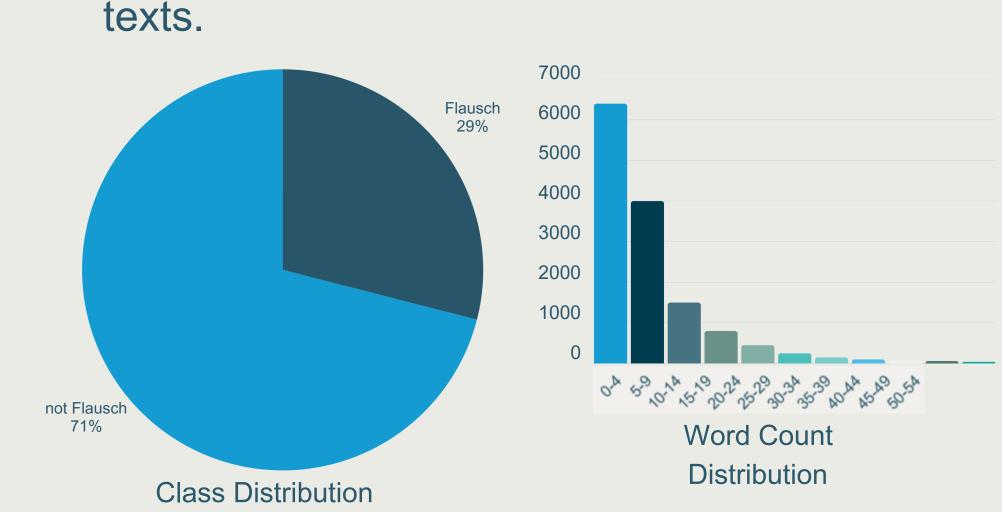
- 1. Classical Baselines: Logistic Regression, SVM, etc. (with TF-IDF, Word2Vec).
- 2.Full Fine-Tuning: mBERT, XLM-RoBERTa, mDeBERTa (Best).
- 3. Parameter-Efficient Fine-Tuning (PEFT): LoRA & Prompt Tuning were tested as computationally efficient alternatives.

The Winning Pipeline

Input: Raw German YouTube Comments

> Core Model: microsoft/mdeberta-v3-

base



imbalanced, with informal language and short

Key Insight

Dataset

comments

- 50% of comments have 6 words or fewer.
- "Noisy" text (emojis, slang, typos) is a critical predictivesignal, not just noise to be removed.

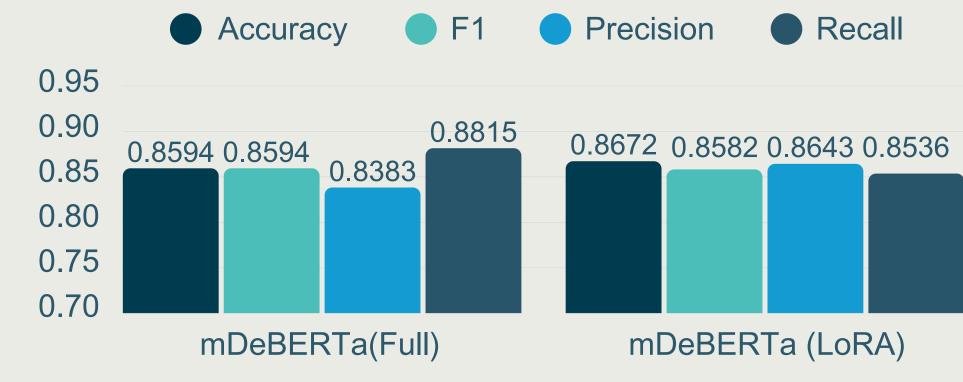
GitHub Repository



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PEFT vs. Full Fine-Tuning LoRA is an Efficient Alternative



Official Result

The "So What?"

0.8411 0.8370 0.8453

mBERT

0.8383

Base model

0.95

0.90

0.85

0.80

0.75

0.70

0.95

0.90

0.85

0.80

0.75

0.8763

- Model Comparison

AccuracyF1PrecisionRecall

0.8513 0.8482 0.8545

XLM-R

Recall

Augmentation Sweet Spot

0.8172

class distr. 70/30

Small Dose, Big Effect (mDeBERTa)

Precision

0.8383

mDeBERTa

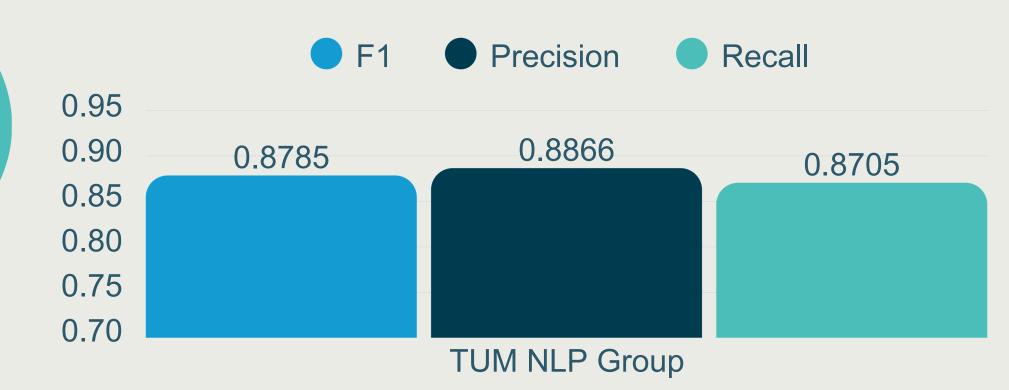
Accuracy

class distr. 65/35

0.8454

MDeBERTa Outperforms Other Transformers

Achieved Rating: 6th / 20



Takeaways & Conclusion

- Data > Model: A10% data augmentation with an LLM gave a bigger boost than changing model architecture.
- Efficiency Matters: PEFT (LoRA) achieves results comparable to full fine-tuning at a fraction of the computational cost.
- "Noise" is Signal: Informal language ("noisy" text) is a key feature, not a bug to be removed.
- Future Work: The most promising direction is expanding the high-quality dataset.

