



Automatic Misogyny **Identification**

with ALBERTo, UmBERTo and XLM-T



Task **description**



EVALITA 2020 task: Automatic Misogyny Identification (AMI) in Italian tweets





Recognize if a tweet is misogynous or not, and if misogynous, whether it's aggressive






Task importance: address growing problem of online misogyny and develop NLP solutions for diverse languages and cultures






Problem modeling






  **nonmisogynous** ...

@

Mi sono svegliata ora   ho dormito tutto il pomeriggio

  **misogynous** ...

@

@ Hai ragione, per esempio modestamente questo tweet è meraviglioso e nessuno  , l'avesse scritto una   avrebbe 10.000 like, ti pare giusto?

  **aggressive** ...

@

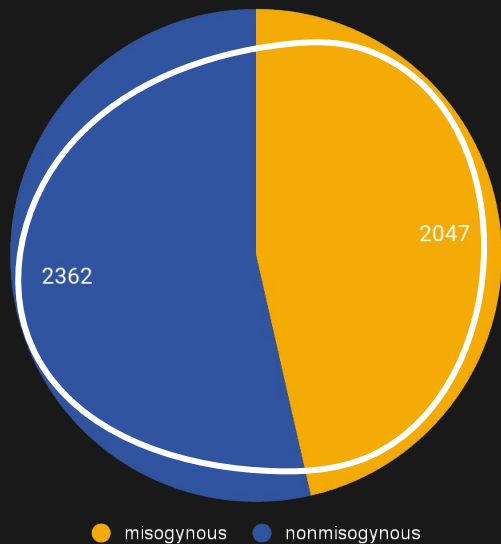
@ Taci che fai più bella figura

**binary
or
ternary?**

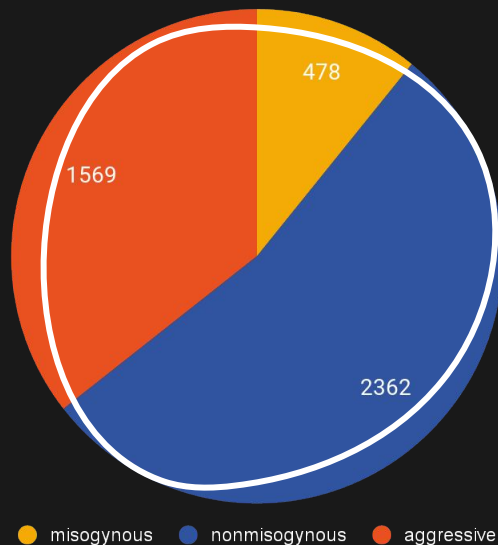


Data distribution

Binary modeling



Ternary modeling



Related work: **ALBERTo**

Outperformed
top submitted model
in 2020 AMI task (Muti et al.)

**Ternary
classification**

**Hyperparameter
tuning**

A. Muti
A. Barrón-Cedeño

Related work: **UmBERTo**

Comparison of three
BERT models
**ALBERTo, UmBERTo,
GiLBERTo**
in 2018 AMI task (Santini)

**Binary
classification**

UmBERTo

emerged as the best
performer

All models
outperformed

systems evaluated in AMI
Evalita 2018 campaign

Project goals

Evaluate performance of **ALBERTo**, **UmBERTo**, and **XLM-T** in the Automatic Misogyny Identification task

01

02

Approach task with two problem modeling strategies: **binary** and **ternary** classification

Examine impact of **data augmentation** on model performance

03

04

Model comparison in different settings:

- binary classification
- ternary classification
- binary classification with data augmentation
- ternary classification with data augmentation

Language **models**



ALBERTo

Italian BERT
model for Twitter
language
understanding



UmBERTo

RoBERTa-based
Italian Language
Model trained on
large Italian
Corpora



XLM-T

Multilingual
Language Model
Toolkit for Twitter

Data augmentation



Binary classification: **slight** imbalance
Ternary classification: **significant** imbalance



Techniques applied:

- Random character **swap**
- Random character **insert**
- Random character **deletion**



Performed using the **NLPAug** library

Training



ALBERTO, UmBERTo

Epochs: 8

Batch size: 16



XLM-T

Epochs: 5

Batch size: 8

Evaluation **metrics**

Accuracy

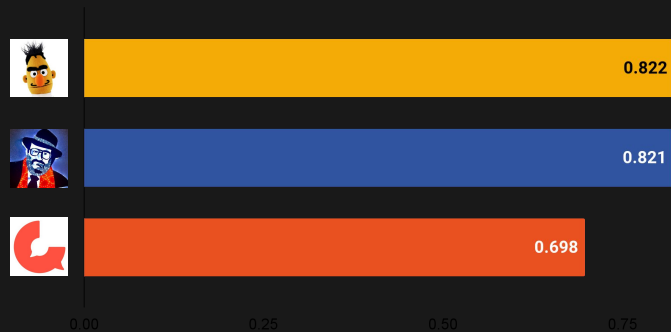
Binary classification
Santini (UmBERTo), AMI 2018

Weighed F1-score

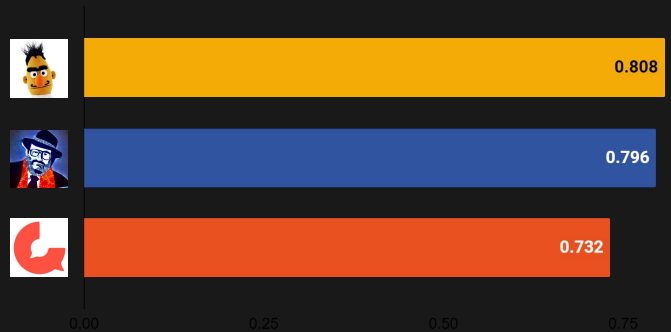
Ternary classification
Muti et al., AMI 2020

Experimental results

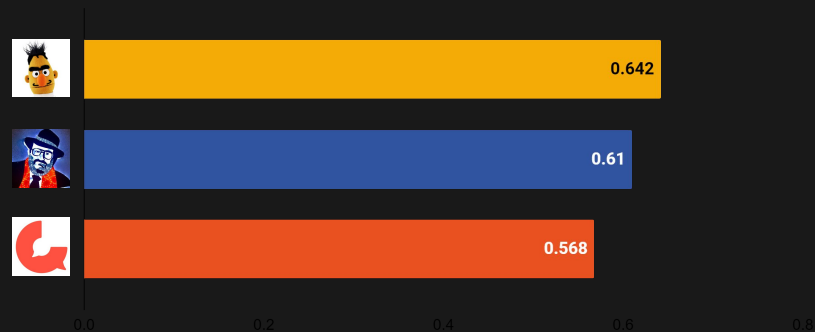
Binary Classification (Accuracy)



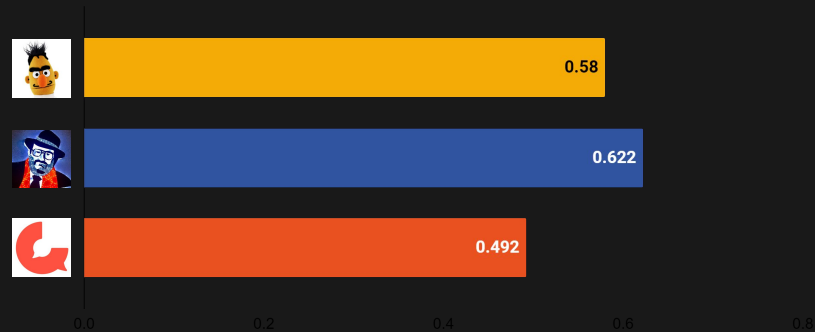
With Data Augmentation



Ternary Classification (weighed F1-score)



With Data Augmentation



Final considerations



Superior performance of ALBERTo and UmBERTo in both binary and ternary classification tasks



Domain- and language-specific model performed better: importance of dedicated models for specific languages and tasks



Negative impact of employed data augmentation techniques:

- may have introduced artificial patterns that dampened the model's ability to generalize to new data
- Solution: alternative DA techniques or tools like weighted loss function or undersampling



Thank you for your attention!

Please feel free to ask any questions



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