# Evaluating catastrophic forgetting of state-of-the-art NLP models in predicting moral values.

CSE3000 Research Project by Florentin Arsene f.arsene@student.tudelft.nl

# Background

Why would we want to predict moral values?

- 1. Personal moral values drive people's day-to-day actions.
- 2. Improve collaboration between AI and humans.

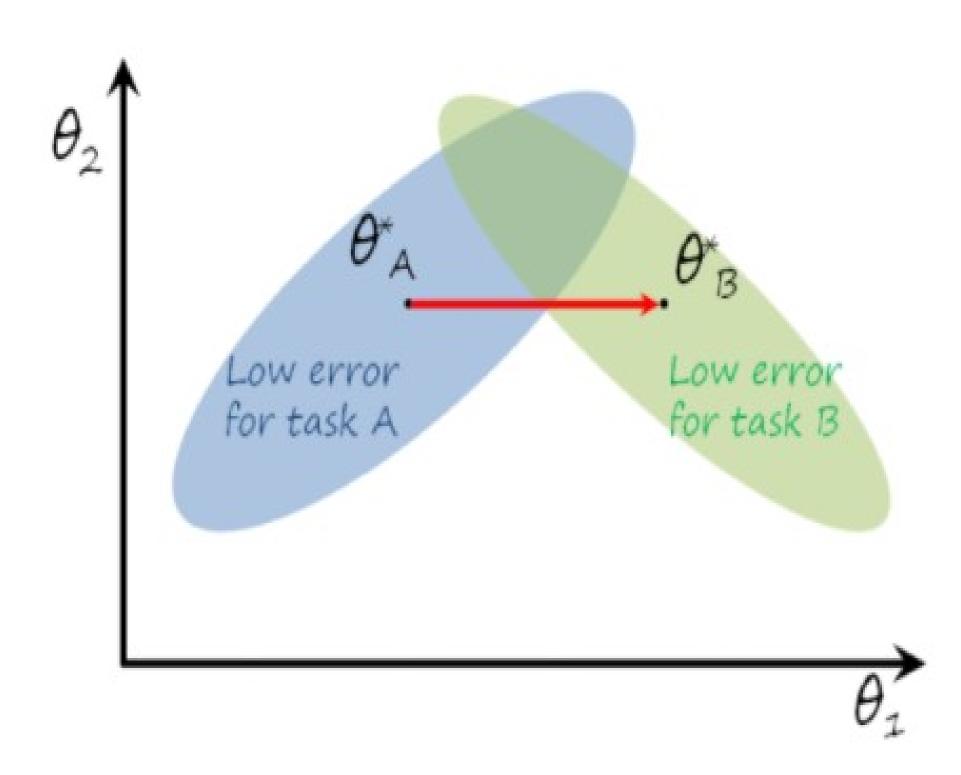
The moral values are analyzed according to the Moral Foundation Theory



# Research Question

Why evaluate catastrophic forgetting?

- 1. Big weakness of Deep Neural Networks.
- 2. Useful in real world applications aimed at sequentially learning new tasks, while not forgetting to perform old tasks.



# Methodology

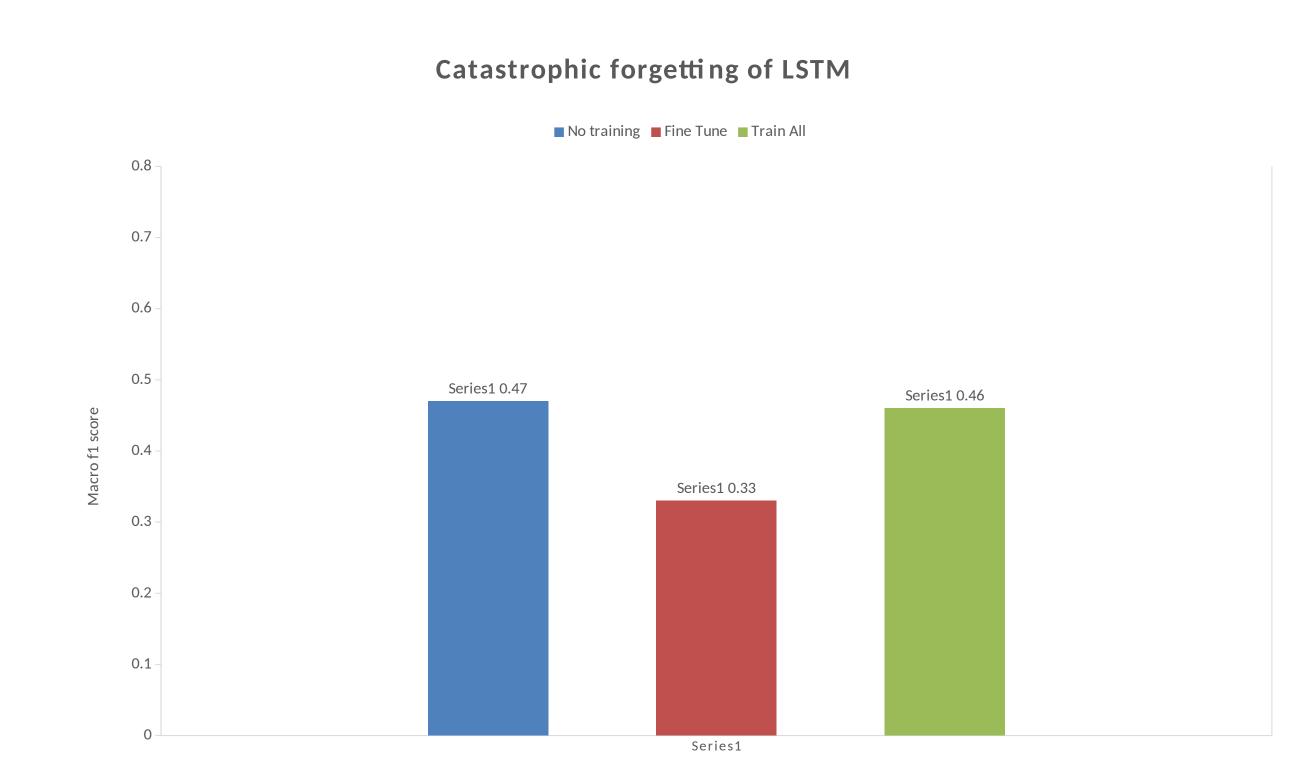
- Preprocess 7 datasets of tweets, corresponding to 7 socially relevant domains: ALM, BLM, Baltimore Protests, hate speech(Davidson), 2016 Presidential election, #MeToo movement and Hurricane Sandy.
- Implement 3 models:
  - LSTM
  - fastText
  - BERT
- Pre-train the models on 6 datasets, then sequentially train on the 7<sup>th</sup> dataset. Repeat this 7 times, each of the 7 datasets being, in turn, the new dataset.
- Evaluate catastrophic forgetting of the models.

# Experiments

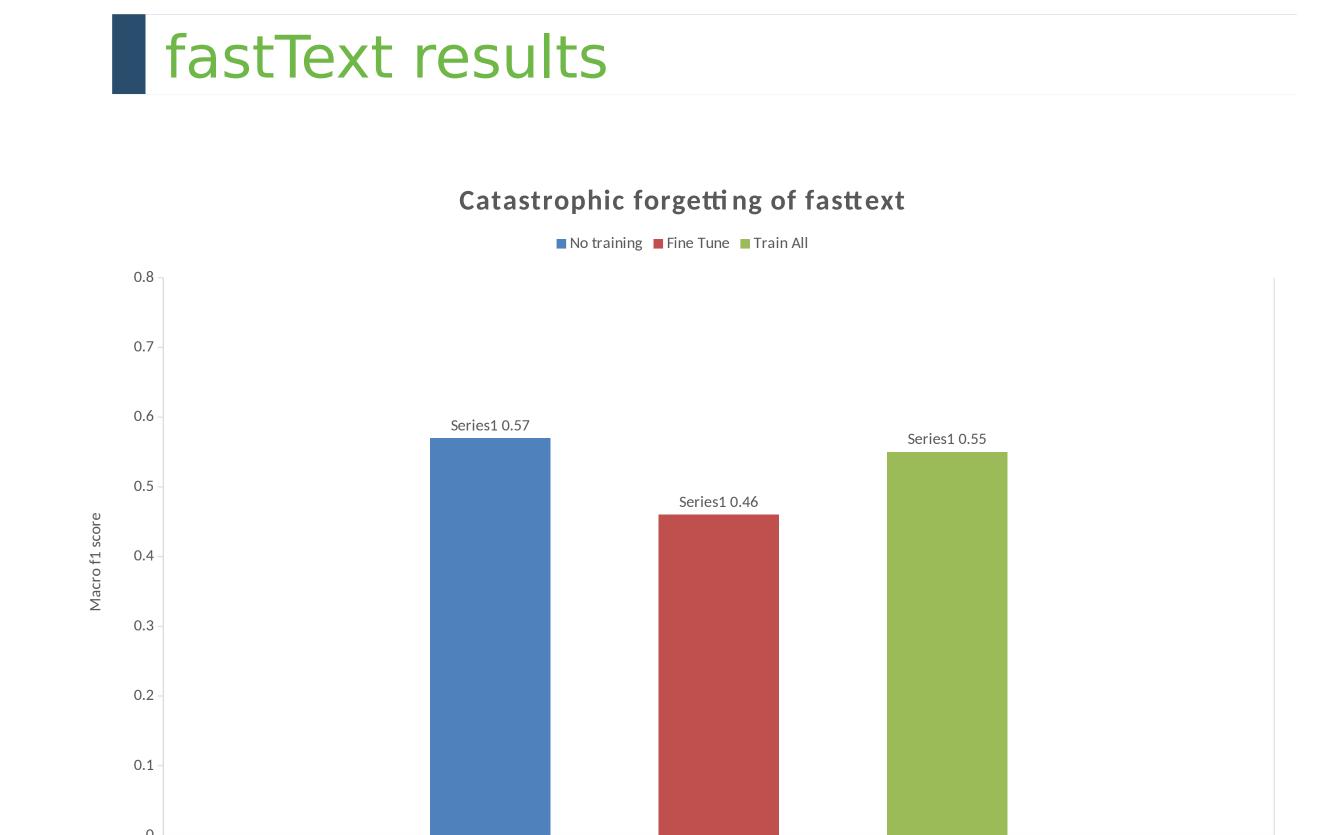
Experiments assume a model is pre-trained on 6 datasets and received a new dataset that it wants to train on. We have 3 types of experiments:

- No training: The model does not train on the new dataset.
- Fine Tune: The model trains on the new dataset.
- Train All: The model trains on a combination of the new dataset and an amount of old pre-training data equal to the new dataset size.

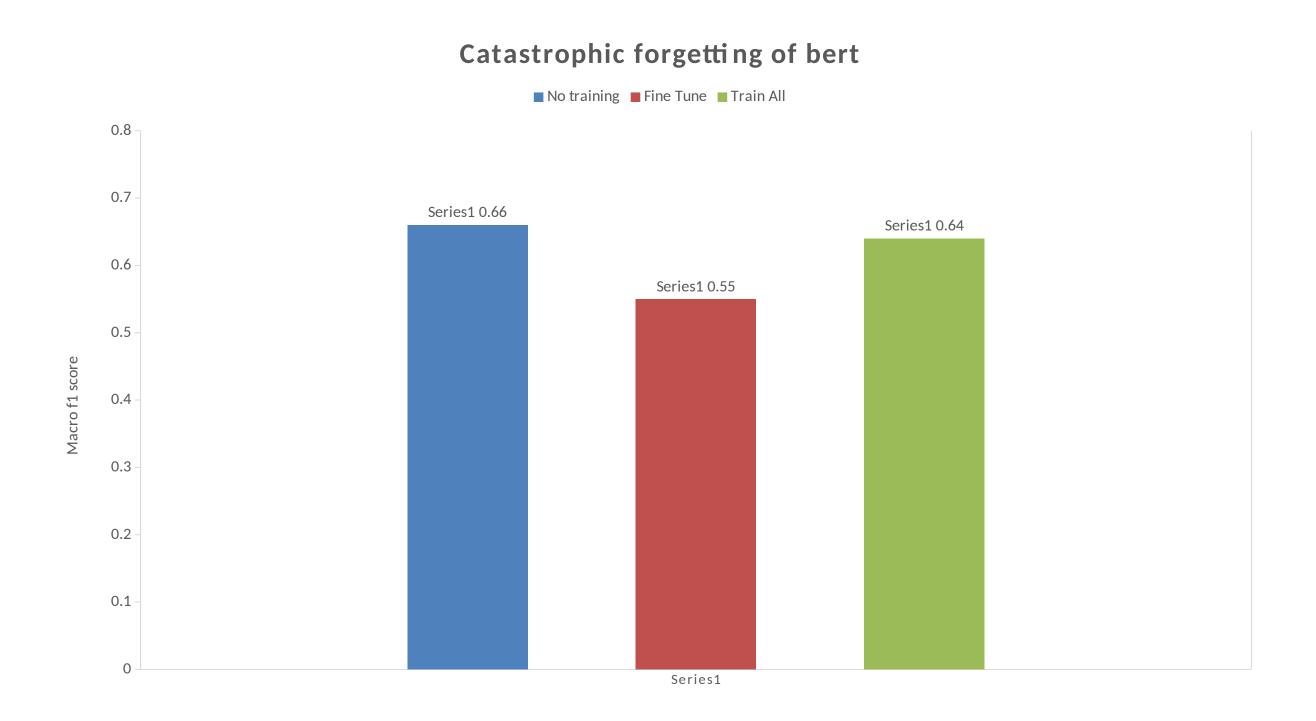
#### LSTM results



#### Contact details:



### BERT results



### Conclusion

- BERT outperforms all the models in terms of performance, taking a lot of time to train.
- fastText is the fastest in terms of training time, however it is hard to configure the model.
- Catastrophic forgetting occurs irrespective of the model.
- Catastrophic forgetting can be mitigated by training on a combination of old data with new data.

