

Natural Language Inference By Jack Pay & Dennis Kiselev



Introduction

- Two models were developed to solve Natural Language Inference (NLI), the task of assessing whether a premise semantically supports a hypothesis
- Method B focused on using RNNs whilst C finetuned pretrained Transformer architectures

Methods

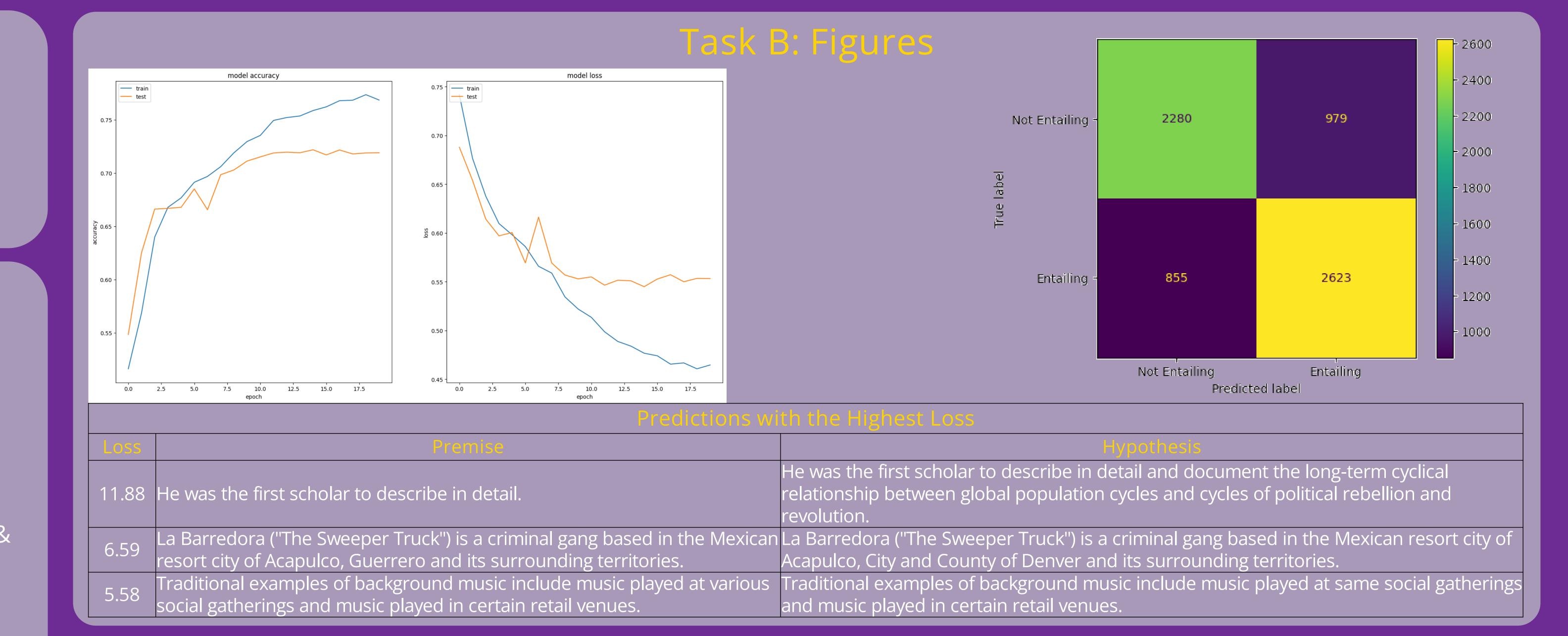
- Method B:
- Trained a **Bi-LSTM** with frozen **XLNET** embeddings
- Utilised learning rate scheduling to approach a global optimum
 - LR was reduced on metric plateau
- Employed subtractive & multiplicative sentence fusion, & attention to enhance sentence representations
- Method C:
- Finetuned the base ROBERTA Transformer model
- Employed data augmentation to bolster the training data:
- Synonym replacements & insertions
- Word deletions & swaps
- Used early stopping and LR scheduling to reduce overfitting
 - LR warmup and then decreasing was used

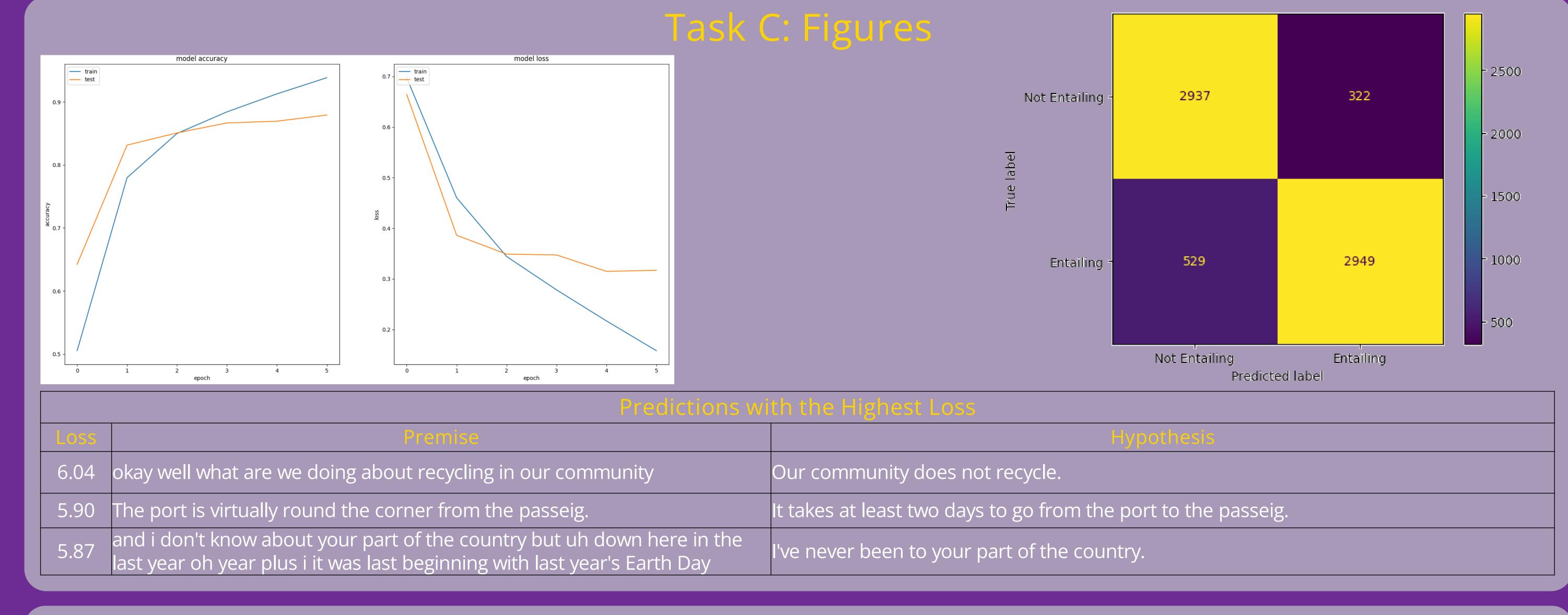
Conclusions

- Pretrained word embeddings are key to higher performance
- Transformers achieve a higher performance compared with RNN
- Basic textual augmentations do not significantly improve performance
- Models struggle when premise & hypotheses are ambiguous, contain disfluencies or OOV words

References

- S. Zhang, S. Liu and M. Liu, "Natural language inference using LSTM model with sentence fusion," 2017 36th Chinese Control Conference (CCC), Dalian, China, 2017, pp. 11081-11085, doi: 10.23919/ChiCC.2017.8029126
- Tarunesh, Ishan, Somak Aditya, and Monojit Choudhury. "Trusting roberta over bert: Insights from checklisting the natural language inference task." arXiv preprint arXiv:2107.07229 (2021).
- Sadat, Mobashir, and Cornelia Caragea. "Scinli: A corpus for natural language inference on scientific text." arXiv preprint arXiv:2203.06728 (2022).





Optimal Results

Method	Accuracy	Precision	Macro Precision	Weighted Macro Precision	Recall	Macro Recall	Weighted Macro Recall	F1-Score	Macro F1-Score	Weighted Macro F1-Score	MCC	Loss
В	0.728	0.728	0.728	0.728	0.754	0.727	0.728	0.741	0.727	0.728	0.455	0.535
C	0.874	0.902	0.874	0.875	0.848	0.875	0.874	0.874	0.874	0.874	0.749	0.333