Multimodal Conversational Emotion Recognition



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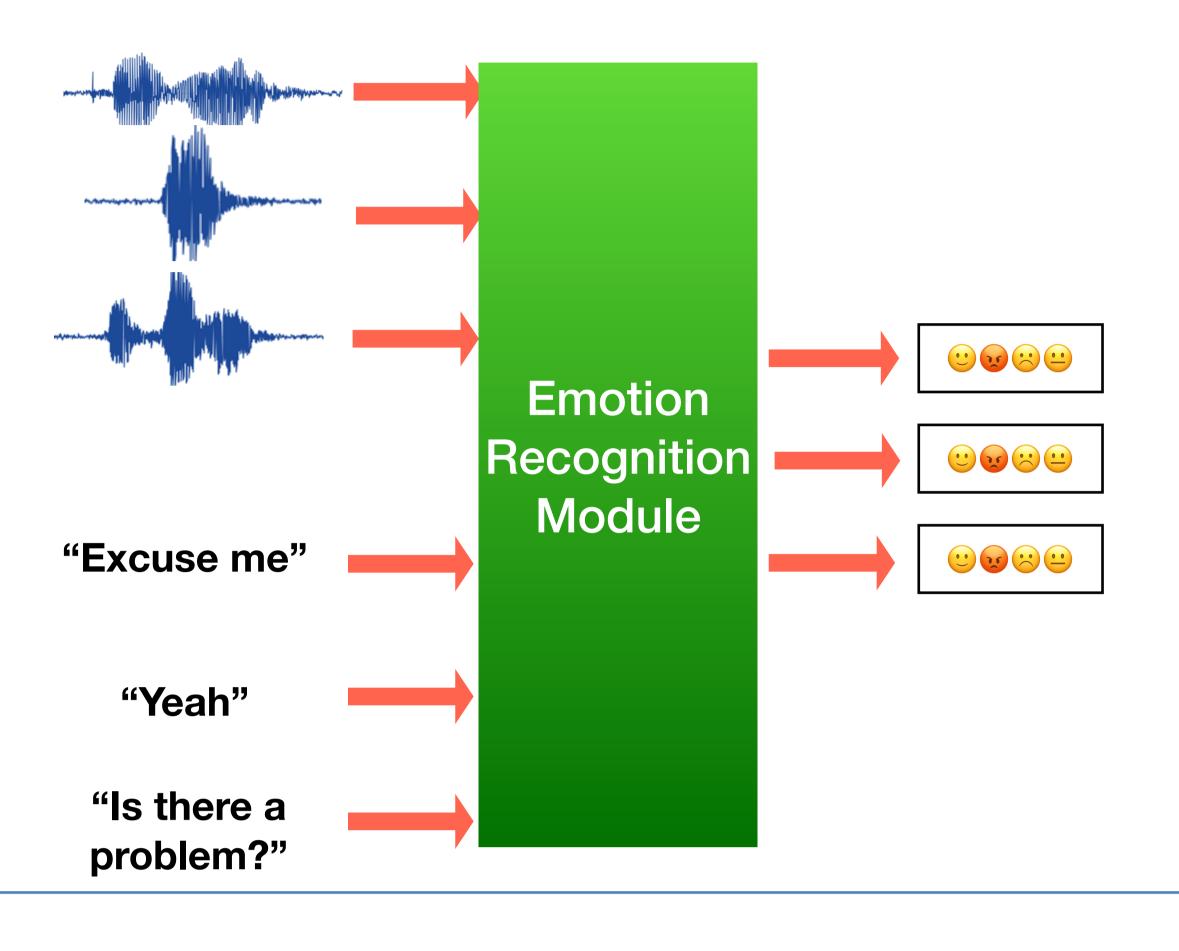
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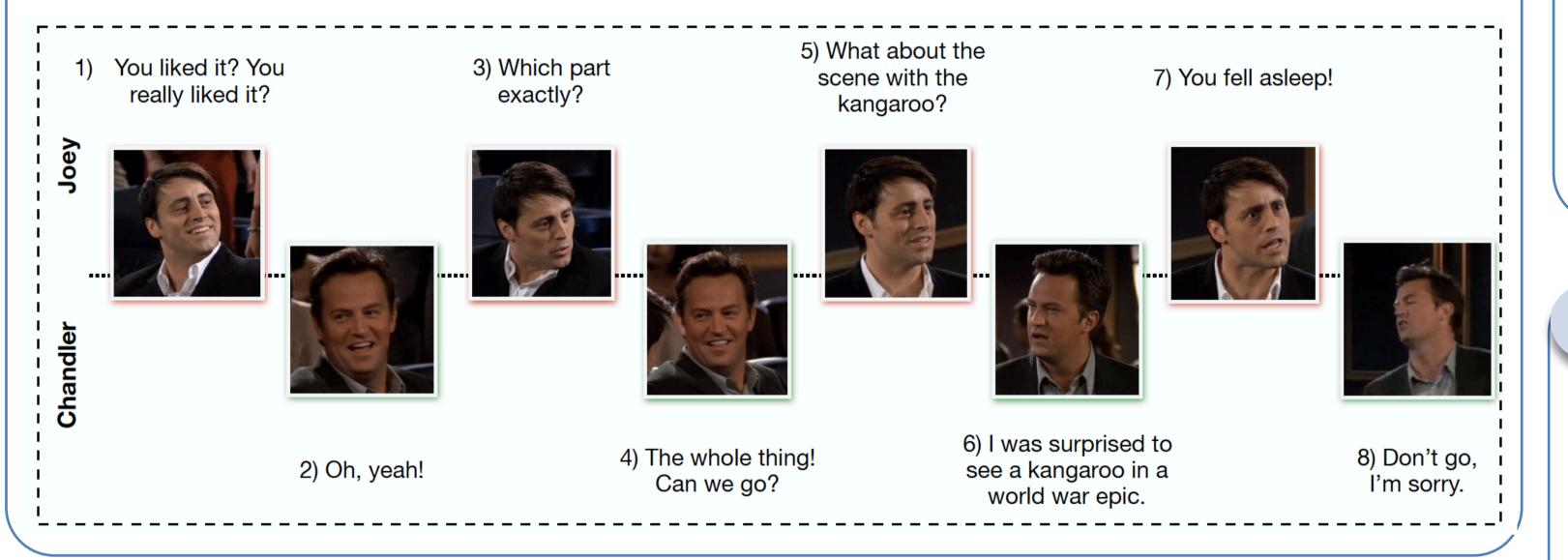
Introduction

Recognise emotions of speakers engaging in a conversation, taking cues from multiple modalities such as speech and provided text transcriptions



Major Challenges

- Multiple speakers contextual decisions are key
- Information across multiple modalities fusion is key

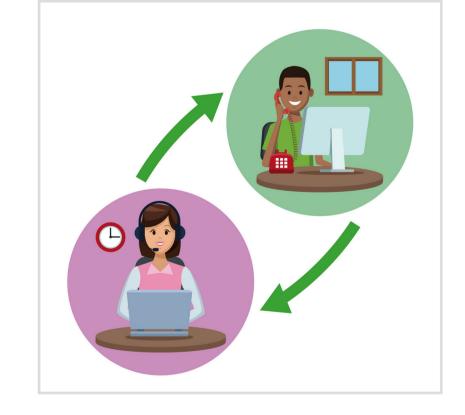


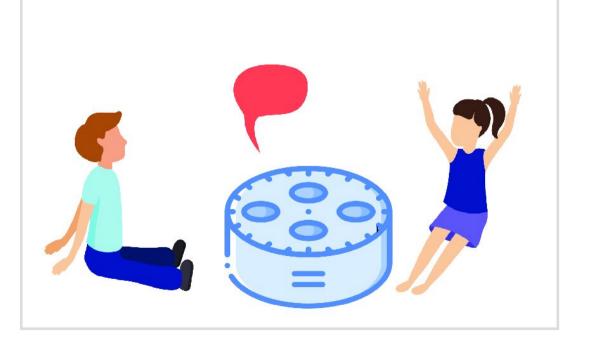
Applications



Continuous monitoring of emotions - better recognition of mental health issues

Analysis of customer care call centre conversations





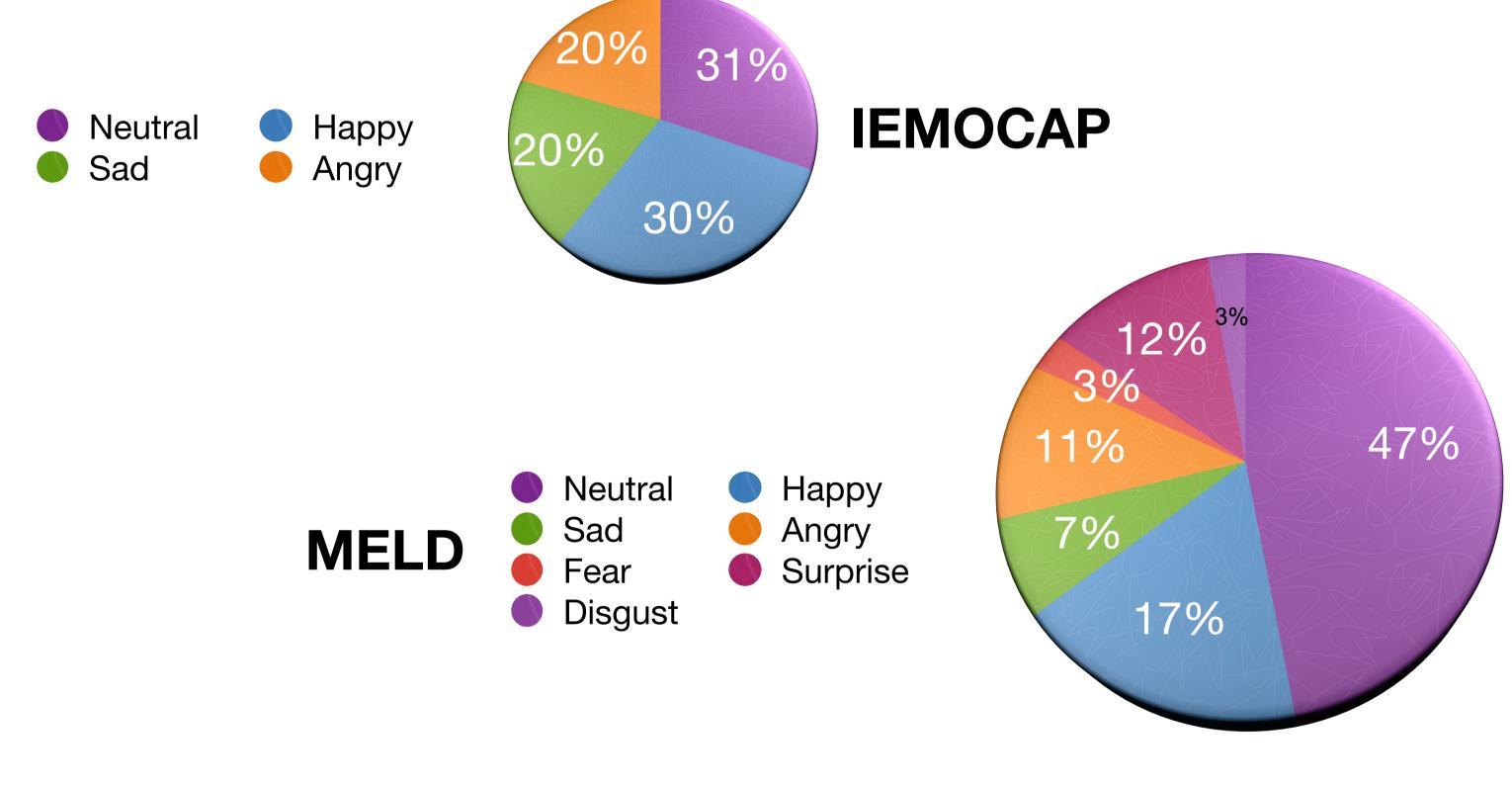
Improving human machine interaction

Approach Audio Classifier Audio Classifier Context addition network Fusion Network Fusion Network Label Label "Yeah" "Is there a problem?"

- This is a hierarchical model which first trains the audio and text classifier to classify individual speech and text utterances
- Context from other utterances in conversations is added by neural networks suited for temporal data - RNNs, LSTMs, GRUs
- Fusion of the two modalities is done by aligning the audio and text representations by means of similarity transformer

Datasets and Results

- Major and most popular Datasets IEMOCAP and MELD
- Unbalanced datasets Metric is Weighted F1 score



- IEMOCAP recognition score ~85%; MELD recognition score ~ 66%
- Further research required for better recognition of sparsely represented classes

References

- Busso et al. "IEMOCAP: Interactive emotional dyadic motion capture database." LREC 42.4 (2008): 335-359.
- Poria et al. "Meld: A multimodal multi-party dataset for emotion recognition in conversations." arXiv preprint arXiv:1810.02508.
- Dutta et al. "Multimodal Transformer with Learnable Frontend and Self Attention for Emotion Recognition." ICASSP 2022.