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**Multimodal deep-learning system for anxiety and depression detection**

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# Overview



- ❖ Motivation
- ❖ Dataset
- ❖ Model
- ❖ Results
- ❖ Conclusion

## Motivation



- Mental health care is an integral part of providing holistic care for patients
- Mental health screening remains a barrier to many who wish to access mental health care
- Improvements in the diagnosis or monitoring process can lead to huge benefits to those who suffer from depression and anxiety



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# WHY SPEECH?

## WHY SPEECH?



Speech is a rich source of information about  
**disease severity & progression.**

Alzheimer's  
Disease

Depression

Schizo-  
phrenia

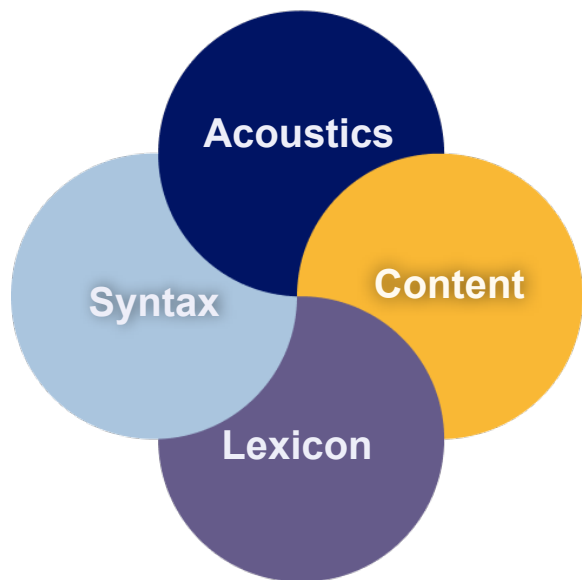
Anxiety

Fronto-  
temporal  
dementia

Disease states produce measurable changes in rate of speech, number of pauses, amount of detail provided, and types of words used.

## SPEECH ASSESSMENTS

Automated speech analysis simultaneously assesses different **domains of speech and language**.



### Benefits of using speech:

- ❖ Ecologically valid
- ❖ Low patient burden
- ❖ Functionally relevant
- ❖ Can be assessed remotely and at high frequency
- ❖ Objective



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# DATA COLLECTION

# Dataset



- Data sourced from the DEPAC corpus, collected over Amazon mTurk
- Individuals were prompted to record themselves completing self-administered speech tasks
- These were then automatically transcribed to form an associated transcript for each audio file
- Individuals also complete a questionnaire including demographic details and questions following the GAD-7 and PHQ-8 scale



- Journaling
  - Subjects are asked to describe their day in as much detail as they would like
- Prompted narrative
  - Subjects are asked to describe hobbies or travel experiences with as much details as they would like
- Positive fluency
  - Subjects are asked to list as many positive events that they expect to experience in the near future

# Dataset Features



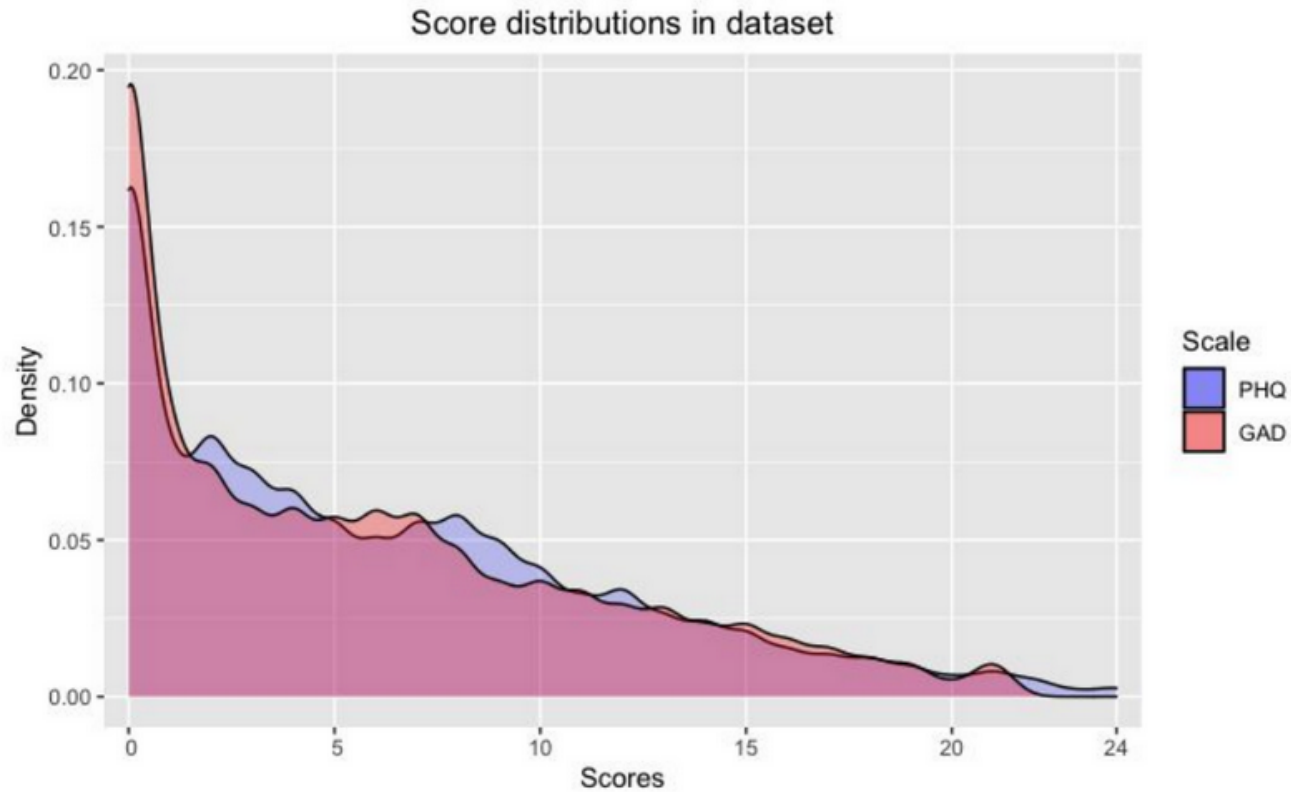
| Feature Group                    | Motivations   |
|----------------------------------|---|
| Intensity (auditory model based) | Perceived loudness in $dB$ relative to normative human auditory threshold.  |
| MFCC 0-12                        | MFCC 0-12 and energy, their first and second order derivatives are calculated on every 16 ms window and step size of 8 ms, and then, averaged over the entire sample. |
| Zero-crossing rate (ZCR)         | Zero crossing rate across all the voiced frames showing how intensely the voice was uttered.  |
| $F_0$                            | Fundamental frequency in Hz.  |
| Harmonics-to-noise-ratio (HNR)   | Degree of acoustic periodicity.   |
| Jitter and shimmer               | Jitter is the period perturbation quotient and shimmer is the amplitude perturbation quotient representing the variations in the fundamental frequency.               |
| Durational features              | Total audio and speech duration in the sample.  |
| Pauses and fillers               | Number and duration of short ( $< 1s$ ), medium ( $1 - 2s$ ) and long ( $> 2s$ ) pauses, mean pause duration, and pause-to-speech ratio.                              |
| Phonation rate                   | Number of voiced time windows over the total number of time windows in a sample.  |

# Dataset Features



| <b>Feature Group</b>            | <b>Motivations</b>  |
|---------------------------------|---|
| Discourse mapping               | Techniques to formally quantify utterance similarity and disordered speech via distance metrics or graph-based representations. |
| Local coherence                 | Coherence and cohesion in speech is associated with the ability to sustain attention and executive functions.                   |
| Lexical complexity and richness | Language pattern changes in particular related to the irregular usage patterns of words of certain grammatical categories.      |
| Syntactic complexity            | Measures of syntactic complexity of utterances.   |
| Utterance cohesion              | Measures of tense and concordance within utterances.  |
| Sentiment                       | Features such as valence, arousal, and dominance.   |
| Word finding difficulty         | Metrics related to disfluency and filled pauses in speech.  |

# Dataset

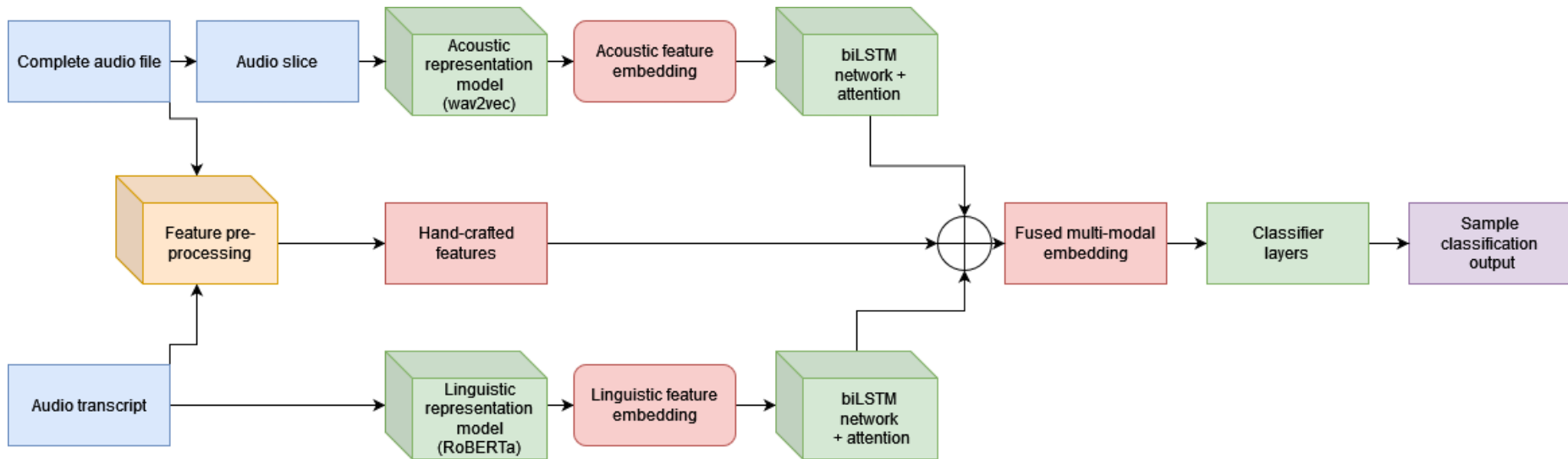




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# MODEL ARCHITECTURE

# Model Architecture



extends work presented in AudiBERT - Toto, Ermal, M. L. Tlachac, and Elke A. Rundensteiner. "AudiBERT: A deep transfer learning multimodal classification framework for depression screening." *Proceedings of the 30th ACM International Conference on Information & Knowledge Management*. 2021.



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# RESULTS

# Results

|                           | Anxiety                    |        |      |                                      |        |             | Depression                 |        |      |                                      |        |             |
|---------------------------|----------------------------|--------|------|--------------------------------------|--------|-------------|----------------------------|--------|------|--------------------------------------|--------|-------------|
|                           | Hand-crafted features only |        |      | Deep-learned + hand-crafted features |        |             | Hand-crafted features only |        |      | Deep-learned + hand-crafted features |        |             |
|                           | Precision                  | Recall | F1   | Precision                            | Recall | F1          | Precision                  | Recall | F1   | Precision                            | Recall | F1          |
| No diagnosis (score < 10) | 0.81                       | 0.65   | 0.72 | 0.76                                 | 0.72   | <b>0.73</b> | 0.73                       | 0.78   | 0.75 | 0.77                                 | 0.83   | <b>0.80</b> |
| Diagnosis (score ≥ 10)    | 0.28                       | 0.41   | 0.33 | 0.37                                 | 0.42   | <b>0.40</b> | 0.31                       | 0.42   | 0.35 | 0.48                                 | 0.39   | <b>0.43</b> |
| Overall                   |                            |        | 0.54 |                                      |        | <b>0.57</b> |                            |        | 0.58 |                                      |        | <b>0.63</b> |

- The addition of hand-crafted features improves our performance for anxiety and depression classification over the baseline
- This is reflective of existing work that shows that the addition of language models like BERT can improve depression classification



## Discussion



- There is an effect caused by data imbalance between the diagnosis and no diagnosis classes
  - Most data falls under the diagnosis cutoff (only 12.8% and 25.3% of the anxiety and depression samples respectively have scores above the cutoff)
- Depression classification overall has higher performance than anxiety



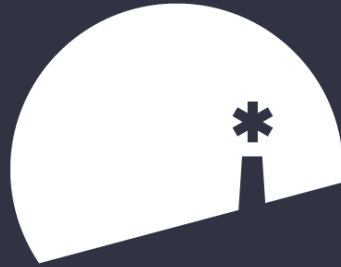
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# CONCLUSION

## Conclusion



- Speech is an effective modality for the diagnosis of depression and anxiety
- There is value in combining deep-learned and hand-crafted features for depression and anxiety detection
- Machine learning and crowdsourcing pose a new and exciting opportunity to potentially improve mental health care, and make it more accessible to all



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