

Phonet: a Tool Based on Gated Recurrent Neural Networks to Extract Phonological Posteriors from Speech

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Introduction

- High-dimensional feature sets like MFCCs or embeddings from neural networks are rarely used for medical applications due to their lack of interpretability.
- Phonological features can be more comprehensible for clinicians than the traditional acoustic features used in speech processing.
- Phonological features are represented by a vector with explainable information about the mode and manner of articulation of the speaker.
- These features are commonly understood by clinicians since they are related with movements of the articulators in the vocal tract.

Aim:

1. To map high-dimensional feature vectors into explainable feature vectors called **phonological posteriors** that can be comprehensible for the medical community.
2. To create an open source toolkit to estimate phonological posteriors based on bidirectional RNNs with gated recurrent units (GRUs).

Phonological posteriors

- The phonetic units of a language can be grouped into phonological classes based on the mode and manner of articulation of the sounds.
- 18 Phonological classes considered in this study.

Phonological classes		
1. Vocalic	7. Nasal	13. Lateral
2. Consonantal	8. Stop	14. Strident
3. Back	9. Continuant	15. Labial
4. Anterior	10. Flap	16. Dental
5. Open	11. Trill	17. Velar
6. Close	12. Voiced	18. Pause

Table 1: List of phonological classes.

- The phonological posteriors will be the posterior probability of a speech frame to belong to one or more phonological classes.
- The phonological posteriors are estimated with a bank of parallel RNNs.
- An additional model for phoneme recognition is also included.

Data

- CIEMPIESS corpus [1].
- 17 hours of FM podcasts in Mexican Spanish.
- The corpus was forced-aligned at phoneme level [2].
- The aligned phonemes were used as labels to train models for phoneme recognition and for the estimation of the phonological posteriors.

References

- [1] Hernández-Mena, C. D., et. al. (2014). CIEMPIESS: A new open-sourced Mexican Spanish radio corpus. In *LREC* (pp. 371-375).
- [2] Schiel, F. (1999). Automatic phonetic transcription of non-prompted speech. In *ICPhS* (pp. 371-375).

<https://github.com/jcvasquezc/phonet>

Contact



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Phonet

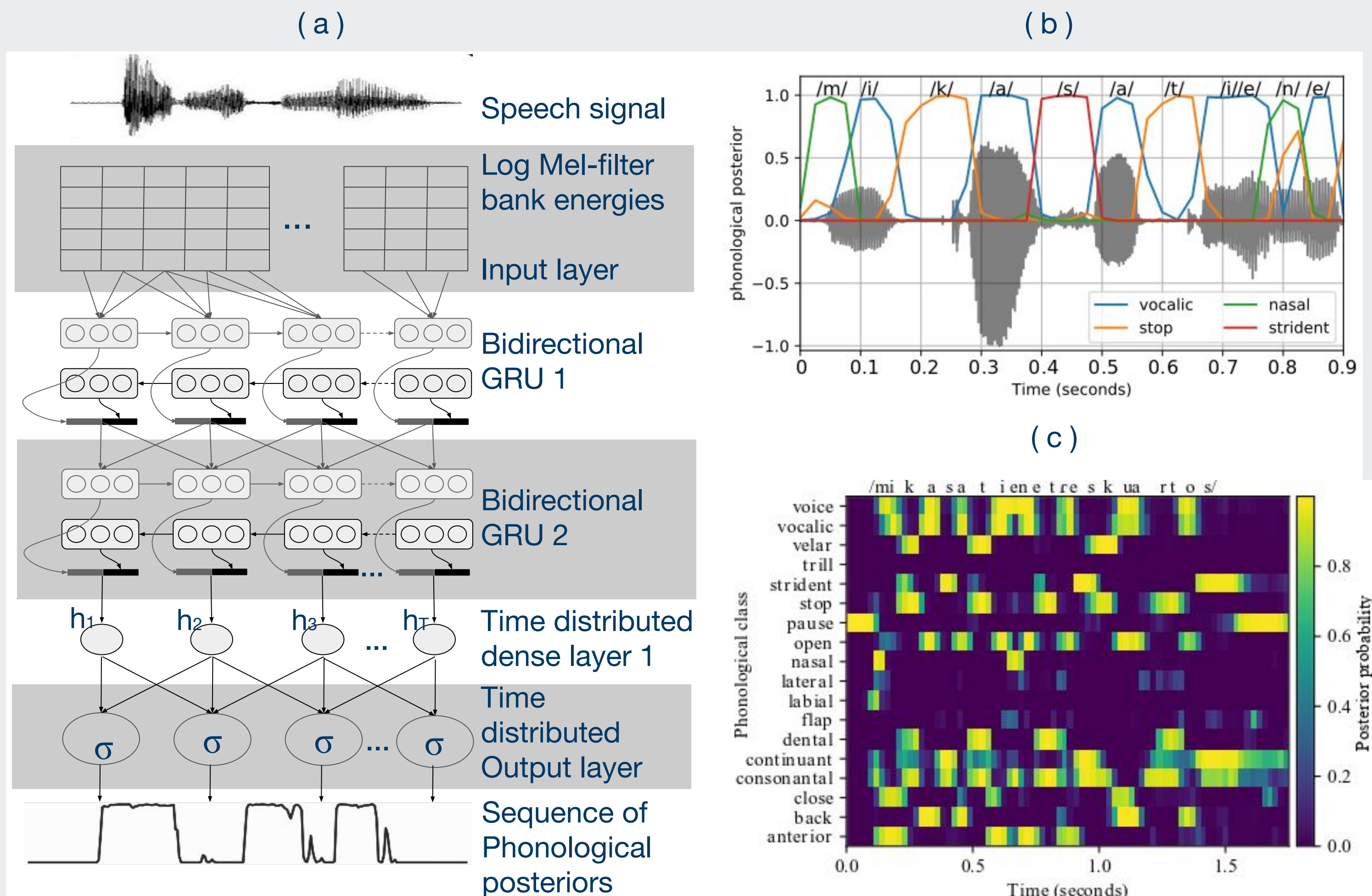


Figure 1: (a) Architecture of Phonet, (b) Example of the vocalic, stop, nasal, and strident phonological posteriors estimated for the sentence "mi casa tiene" in Spanish language. (c) Posteriogram obtained for the Spanish sentence "mi casa tiene tres cuartos".

Results

• Estimation of Phonological posteriors

Class	F-score	Class	F-score	Class	F-score
1. Vocalic	0.841	7. Nasal	0.907	13. Lateral	0.915
2. Cons.	0.831	8. Stop	0.855	14. Strident	0.932
3. Back	0.882	9. Continuant	0.861	15. Labial	0.885
4. Anterior	0.889	10. Flap	0.895	16. Dental	0.898
5. Open	0.841	11. Trill	0.986	17. Velar	0.930
6. Close	0.901	12. Voiced	0.885	18. Pause	0.958

Table 2: Recognition of the phonological classes using Phonet

• Phoneme recognition

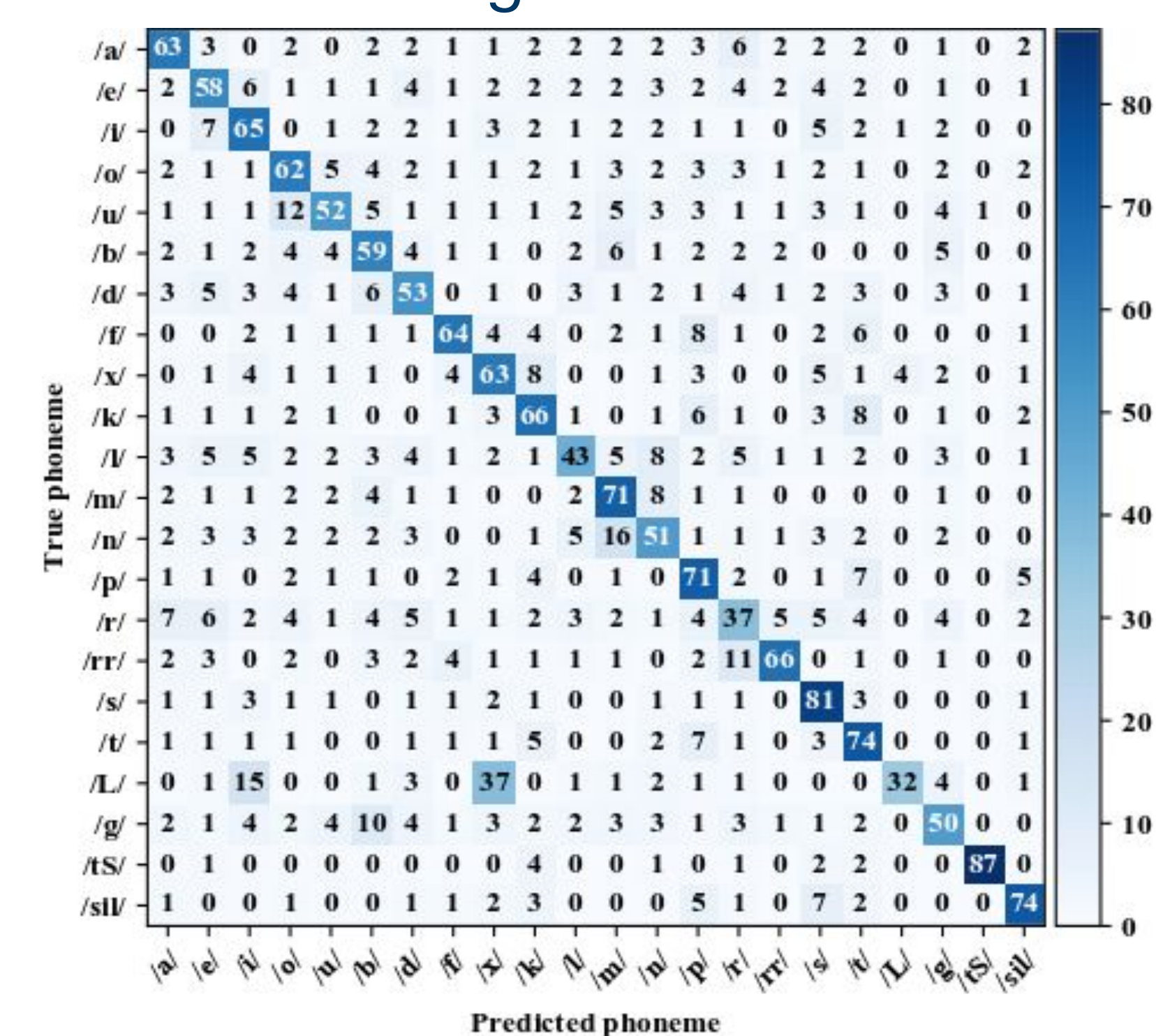


Figure 2: Normalized confusion matrix (in %) for the phoneme recognition task using the proposed approach. The blue bar indicates the percentage of samples classified for each phoneme.

Conclusion

- The accuracy of the models ranges from 80.4% to 93.3%, depending on the phonological class.
- Phonet was trained to recognize posteriors in Spanish. However, the training process can be adapted to other languages.
- Future models will include the estimation of phonological posteriors for the English and German languages.
- The trained models are available as an open-source toolkit.

Acknowledgements



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