

## MOTIVATION

COVID-19 Rapid Testing is needed in order to control the pandemic.

### Shortcomings of conventional testing:

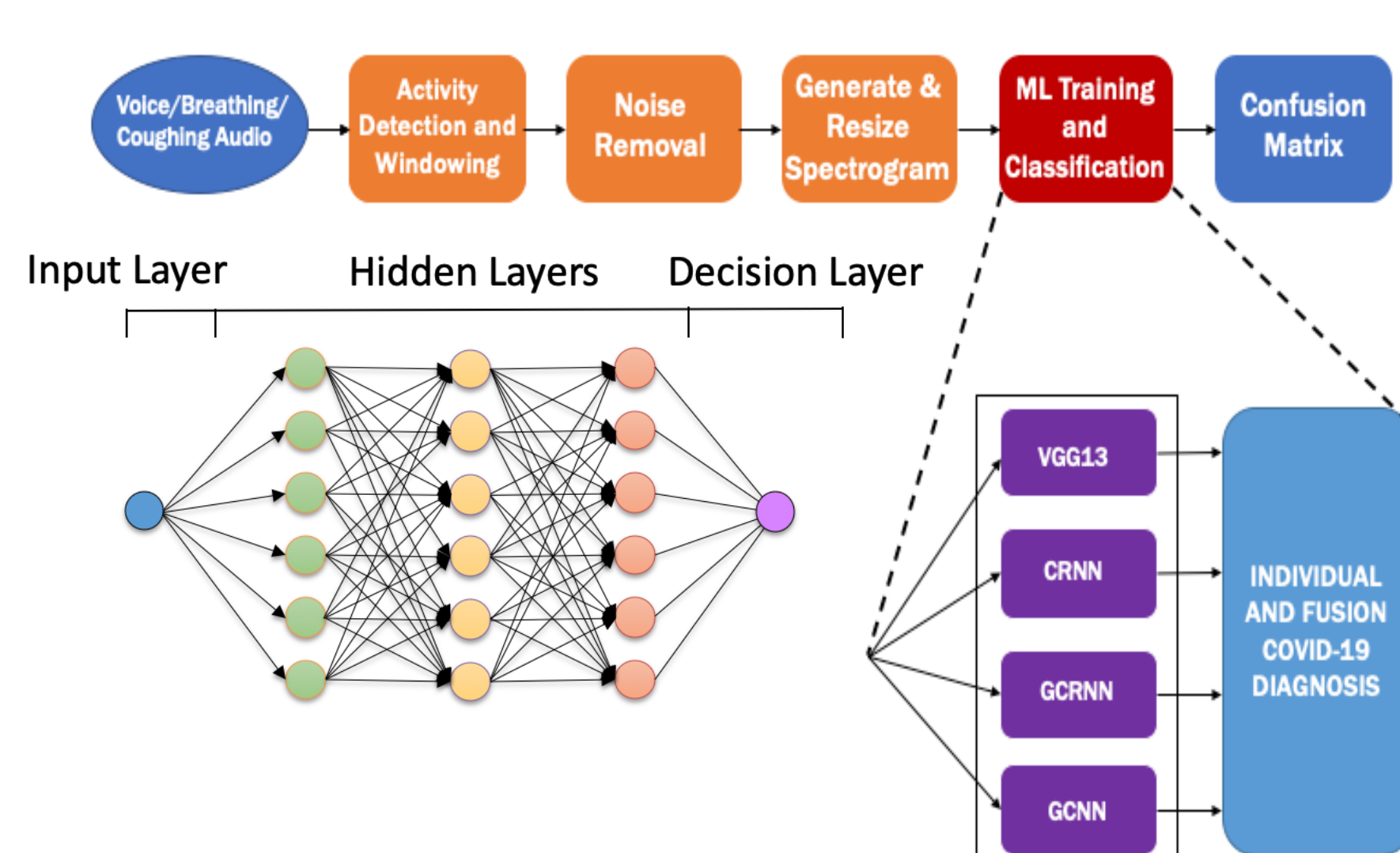
- Limited availability
- Results can take long to obtain

Machine learning can be used to detect COVID-19 using audio samples

## PROJECT AIM

- Create a robust algorithm for detecting COVID-19 using only audio samples.
- Simulate algorithm and obtain results
- Deploy the algorithm in a smartphone app for widespread accessibility

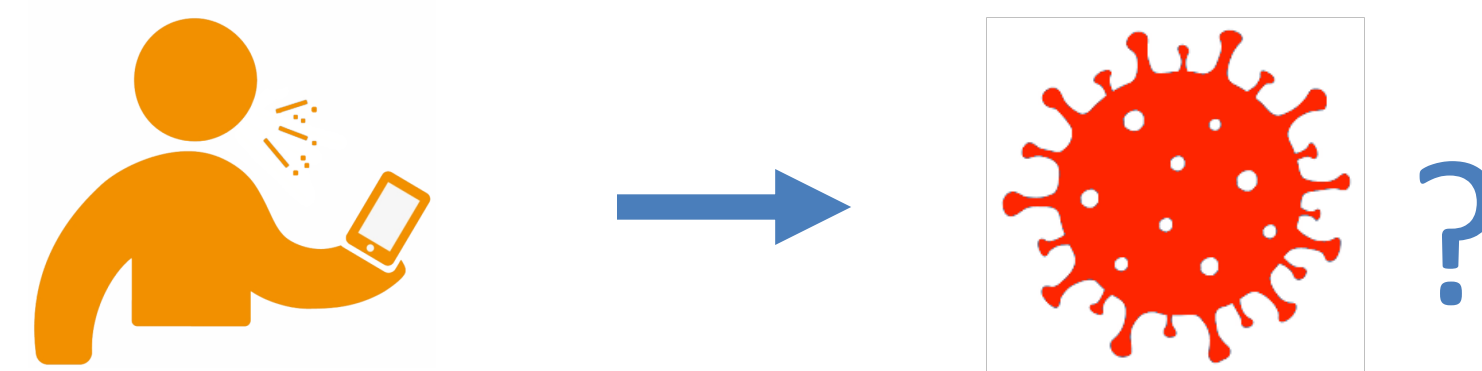
## COVID-19 DETECTION ALGORITHM



## DATA COLLECTION AND ALGORITHMS

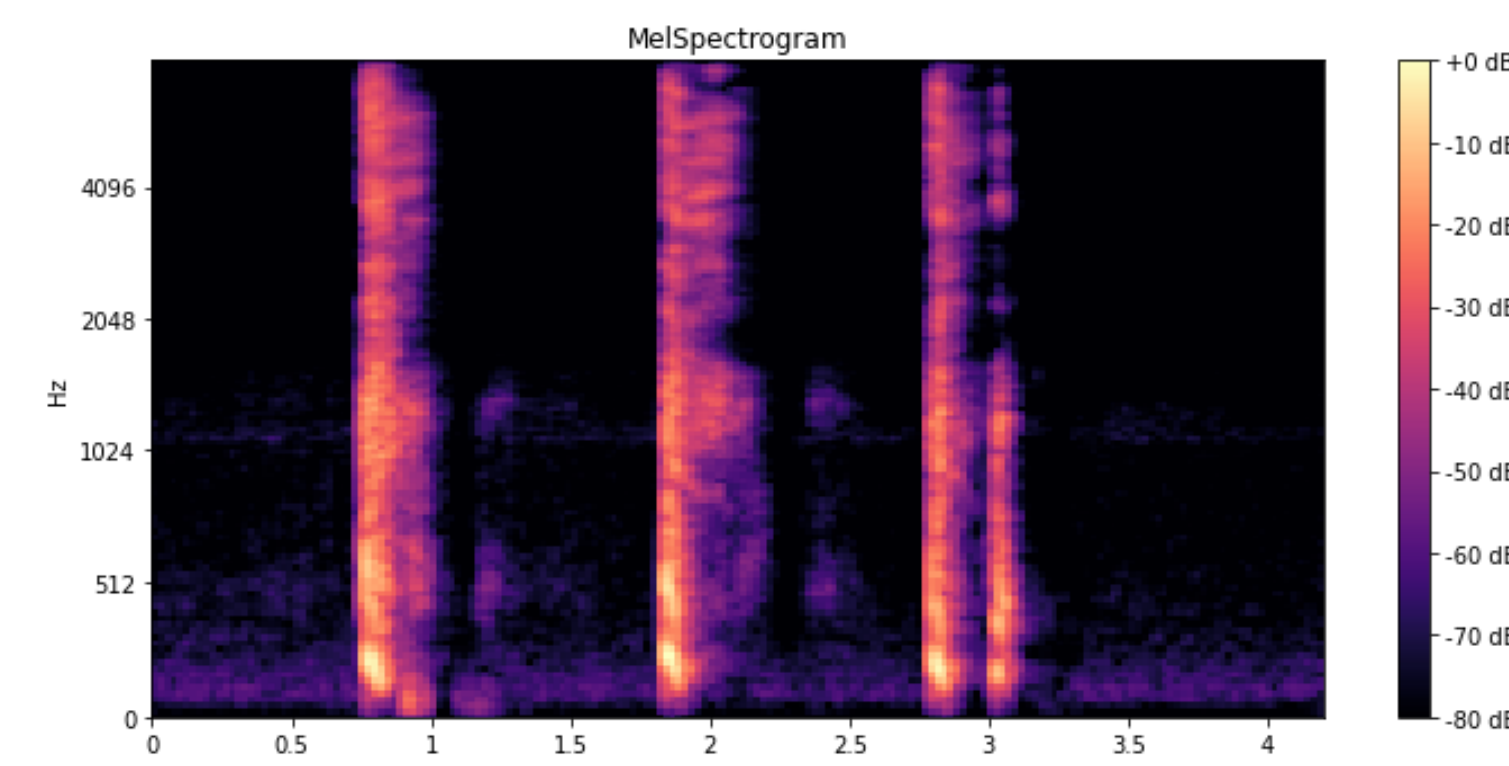
- Data Collection: Use of Coswara [1] and COUGHVID [2], public datasets containing internationally crowdsourced audio data
- Algorithms: Visual Geometry Group 13 (VGG13) [3], Convolutional Recurrent Neural Network (CRNN) [4], Gated Convolutional Neural Network (GCNN) [5], Gated Convolutional Recurrent Neural Network (GCRNN) [6], and fusion method [7].

## MOBILE PHONE SOLUTION



## SPECTROGRAM OF COUGHING PATTERNS

- Spectrogram features capturing cough patterns
- Images Converted and passed to Neural Net
- Other Spectral Features and ML Algorithms considered [8, 9]
- NN Trained using Mel spectrograms generated from Coswara [1] and COUGHVID [2] data.



## ACKNOWLEDGEMENTS

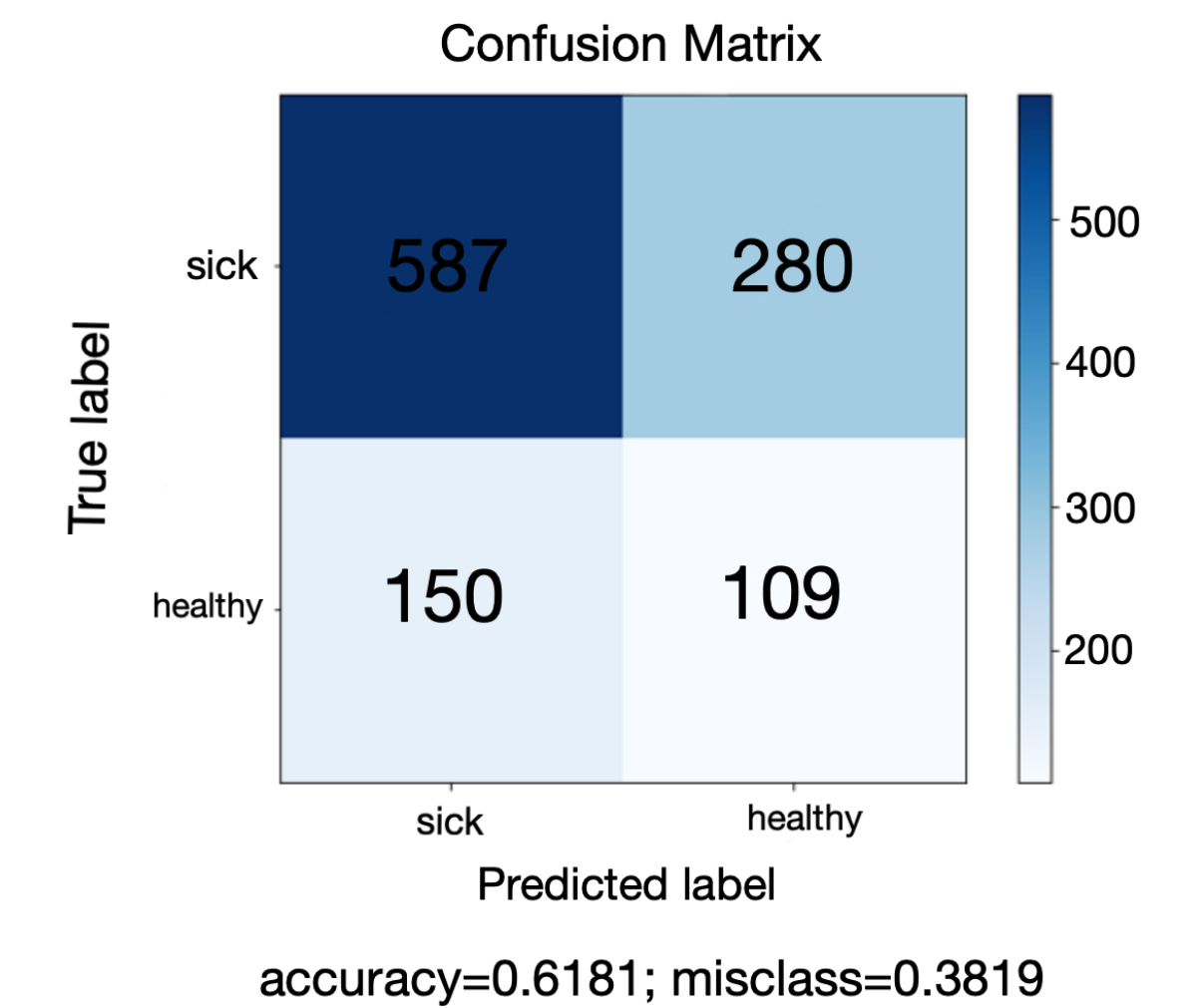
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## PRELIMINARY RESULTS

### Developed Python Implementation

```
import features
# Use a logmel representation for feature extraction
extractor = features.LogmelExtractor(cfg.sample_rate,
                                    cfg.n_window,
                                    cfg.hop_length,
                                    cfg.n_mels,
                                    )
```

Class	MAP@1	F-score	Precision	Recall
COVID-19	0.320	0.292	0.268	0.320
healthy	0.738	0.761	0.784	0.738
Macro Average	0.529	0.526	0.526	0.529
Micro Average	0.642	0.642	0.642	0.642



## CONCLUSION

- Feasibility of method demonstrated
- Initial results are encouraging:
  - 61.81% accuracy with REU algorithm (more recent results in SenSIP at 80%)
- 9M-22M parameters per model
- Training requires high speed computing

## ONGOING & PLANNED WORK

- Detect COVID-19 with 80% or more sensitivity with high specificity.
- Develop software for use on smartphones
- Explore the use of additional features for classification
- Paper invited to IEEE Asilomar Conference
- Obtain additional classification results using DiCOVA dataset and revised algorithms [10]
- Method to be applied to detect other breathing abnormalities

## REFERENCES

[1] N. Sharma, P. Krishnan, R. Kumar, S. Ramoji, S. Chetupalli, Nirmala R., P. Ghosh, and S. Ganapathy, "Coswara - A Database of Breathing, Cough, and Voice Sounds for COVID-19 Diagnosis," Interspeech 2020.

[2] L. Orlandic, T. Teijeiro, and D. Atienza, "The COUGHVID crowdsourcing dataset: A corpus for the study of large-scale cough analysis algorithms," ArXiv abs/2009.11644

[3] HK. Simonyan and A. Zisserman, "Very deep convolutional networks for large-scale image recognition," in Proc. 3rd ICLR, San Diego, 2015.

[4] M. Schuster and K. K. Paliwal, "Bidirectional recurrent neural networks," IEEE Trans. Signal Process., vol. 45, no. 11, pp. 2673-2681, 1997.

[5] Y. Xu, Q. Kong, W. Wang and M. Plumbley, "Large-Scale Weakly Supervised Audio Classification Using Gated Convolutional Neural Network," 2018 IEEE ICASSP 2018, Calgary, 2018.

[6] J. Chung, C. Gulcehre, K. Cho, and Y. Bengio, "Empirical evaluation of gated recurrent neural networks on sequence modeling," arXiv preprint arXiv:1412.3555, 2014.

[7] D. H. Wolpert, "Stacked generalization," Neural Networks, vol. 5, pp. 241-259, 1992.

[8] A. Spanias, T. Painter, V. Atti, Audio Signal Processing and Coding, Wiley, 2007.

[9] U. Shanthamallu, A. Spanias, C. Tepedelenioglu, M. Stanley, "A Brief Survey of Machine Learning Methods and their Sensor and IoT Applications," Proc. IEEE IISA 2017, Larnaca, Aug. 2017.

[10] Diagnosing COVID-19 using Acoustics (DiCOVA), A Special Session at Interspeech 2021, <https://dicova2021.github.io>