

### The Second DiCOVA Challenge:

Dataset And Performance Analysis for Diagnosis of COVID-19 using Acoustics

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# Sound based Diagnosis



#### COVID-19 diagnosis methods:

- RT-PCR testing
- RAT testing
- Point-of-care testing (POCT)
- Advantages of sound based diagnosis of COVID-19





### **Motivation**



#### DiCOVA Challenge series

- Diagnosis of COVID-19 using acoustics
- COSWARA dataset [1]
- First DiCOVA Challenge [2]: 12 Feb 23rd March, 2021
  - Focused on cough audio recordings
  - Special Session in Interspeech 2021
  - 29 teams from around the world
  - 19 outperformed baseline

<sup>2.</sup> Sharma, N. K., Muguli, A., Krishnan, P., Kumar, R., Chetupalli, S. R., & Ganapathy, S. (2022). Towards sound based testing of COVID-19—Summary of the first Diagnostics of COVID-19 using Acoustics (DiCOVA) Challenge. Computer Speech & Language, 73, 101320.



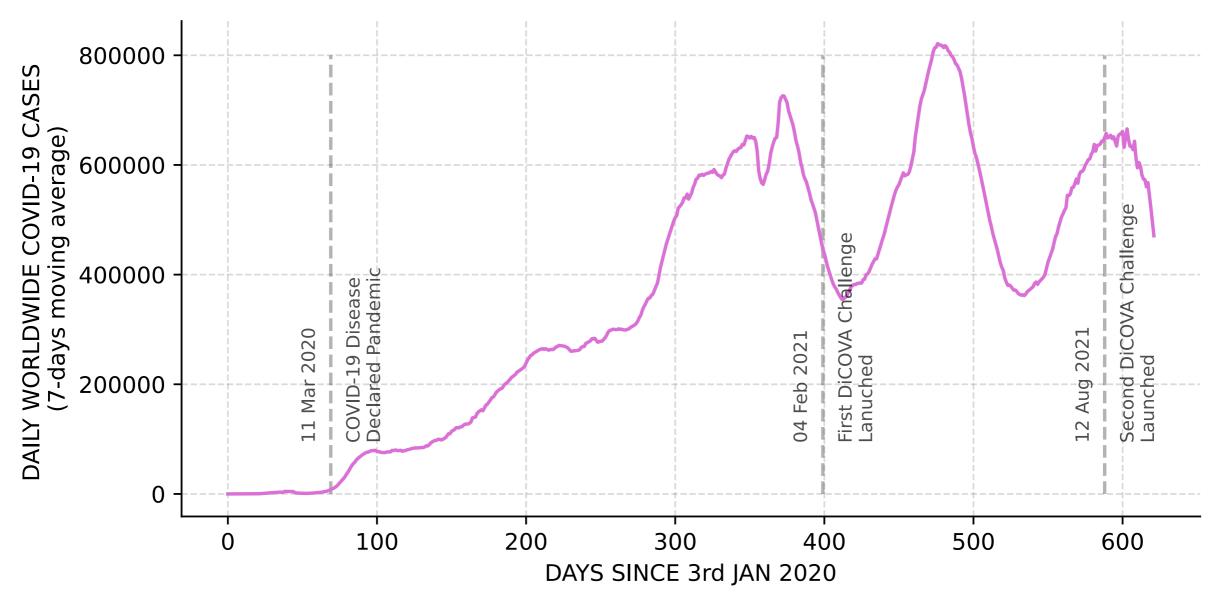


<sup>1.</sup> Neeraj Sharma, Prashant Krishnan, Rohit Kumar, Shreyas Ramoji, Srikanth Raj Chetupalli, R Nirmala, Prasanta Kumar Ghosh, and Sriram Ganapathy, "Coswara – a database of breathing, cough, and voice sounds for COVID-19 diagnosis," in Proc. Interspeech, 2020, pp. 4811–4815

### **Motivation**



#### **Second DiCOVA Challenge**:







# Second DiCOVA Challenge!





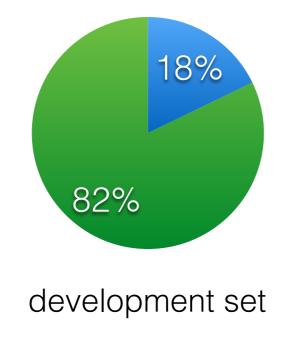
http://dicovachallenge.github.io/

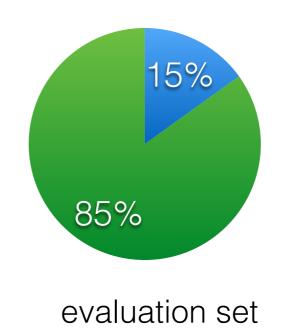






- Development set: 965 individuals, 172 are COVID-19 positive
  - released as 5 train-val folds
- Evaluation set: 471 individuals, 71 are COVID-19 positive



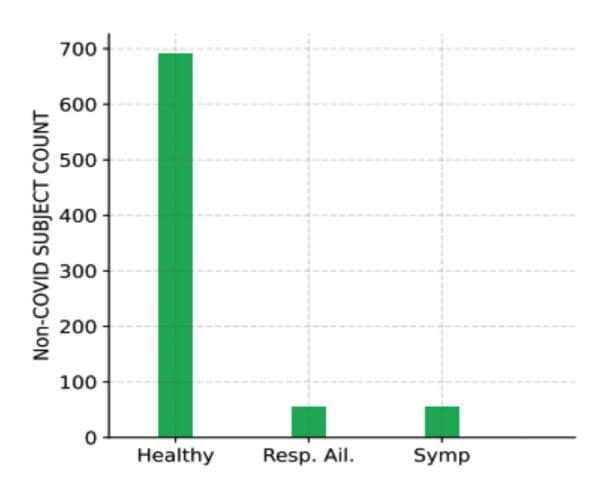








#### Non-COVID subject distribution



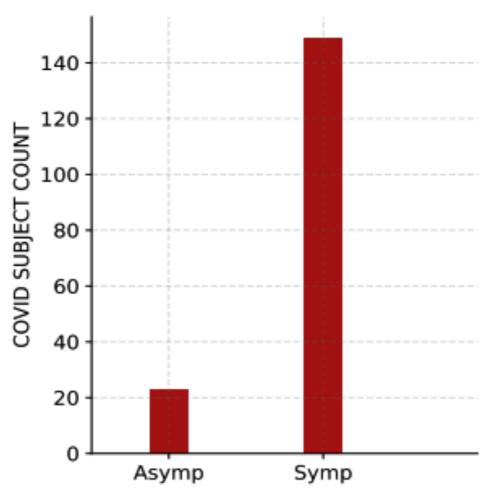
- healthy (no symptoms)
- resp. ail. (asthma, chronic lung disease, pneumonia)
- symptoms (cold, cough, fever, loss of taste or smell)







#### COVID subject distribution



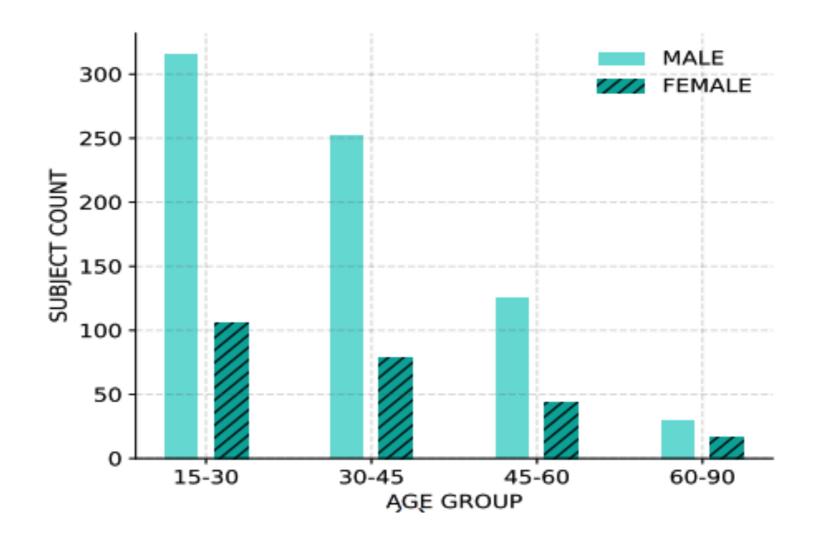
- Asymp (COVID-19 positive without symptoms)
- Symp(COVID-19 positive with symptoms)







Gender and age distribution

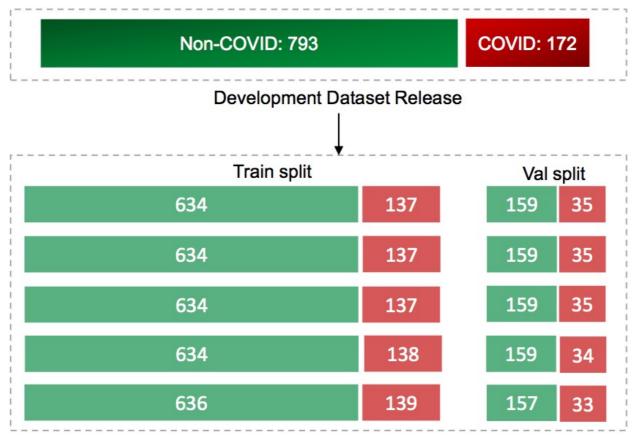








- 5 fold cross-validation
- Tuned based on average validation performance
- Evaluation



Divided into five-fold Train-Val (~80-20%) split





### **Track Details**



#### 4 tracks:

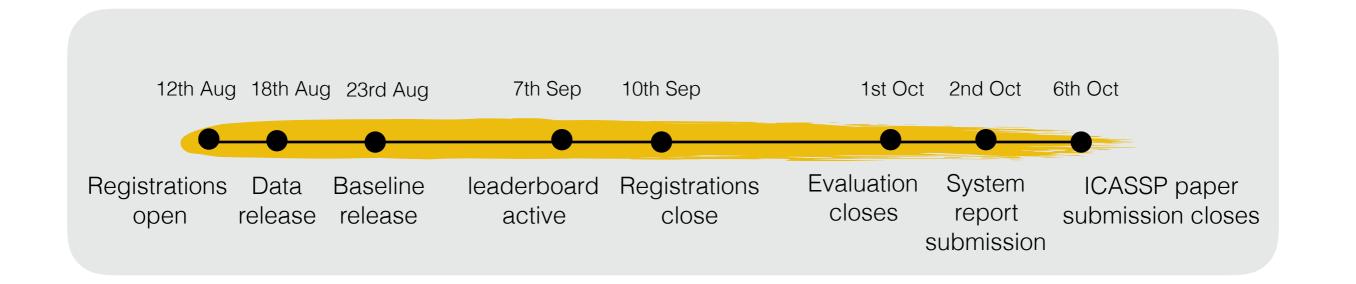
- Track 1, 2 and 3 on breathing, cough and speech sounds
- Track 4 fusion of the first 3 modalities





### **Timeline**









# **Baseline System**



- Bi-LSTM based baseline classifier
- log mel-spectrogram features
- · Baseline codes are made open source to encourage further work on it.

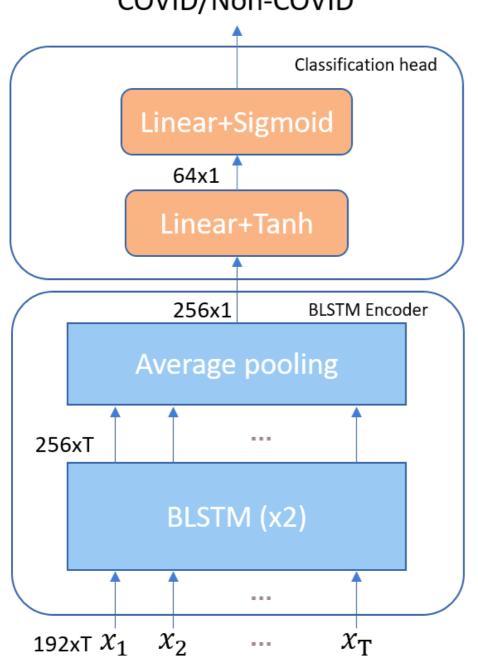




# **Baseline System**



#### COVID/Non-COVID



- Two bi-directional long-short term memory (BiLSTM) layers and a fully connected layer
- Trained on segments of utterances
- Inference based on average probability scores over segments

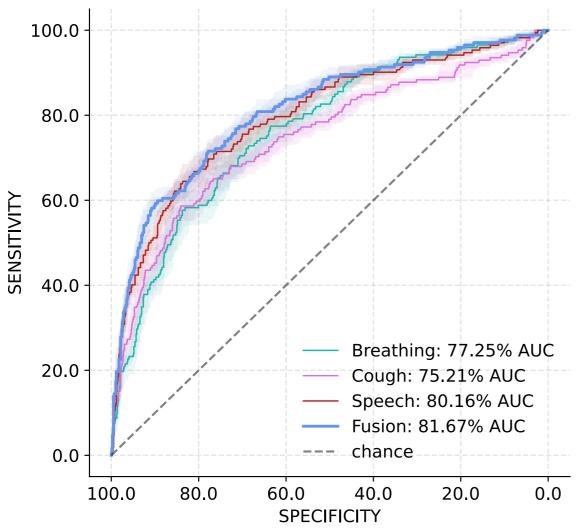




### **Baseline Results**



- Receiver operating characteristics (ROC) curve
- AUC-ROC



Validation	AUC-ROC Performance (in %)			
	Breathing	Cough	Speech	Fusion
fold-0	74.8	71.8	75.4	77.3
fold-1	73.9	78.2	87.2	82.4
fold-2	74.3	77.2	80.6	81.8
fold-3	80.0	74.0	78.2	80.3
fold-4	83.2	74.9	79.5	86.6
Avg. Validation	77.3	75.2	80.2	81.7
Test	84.5	74.9	84.3	84.7

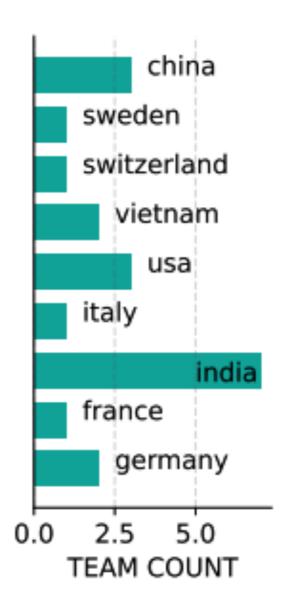
**Table 1**. Baseline system performance on the validation folds in the development dataset, and the blind test dataset.







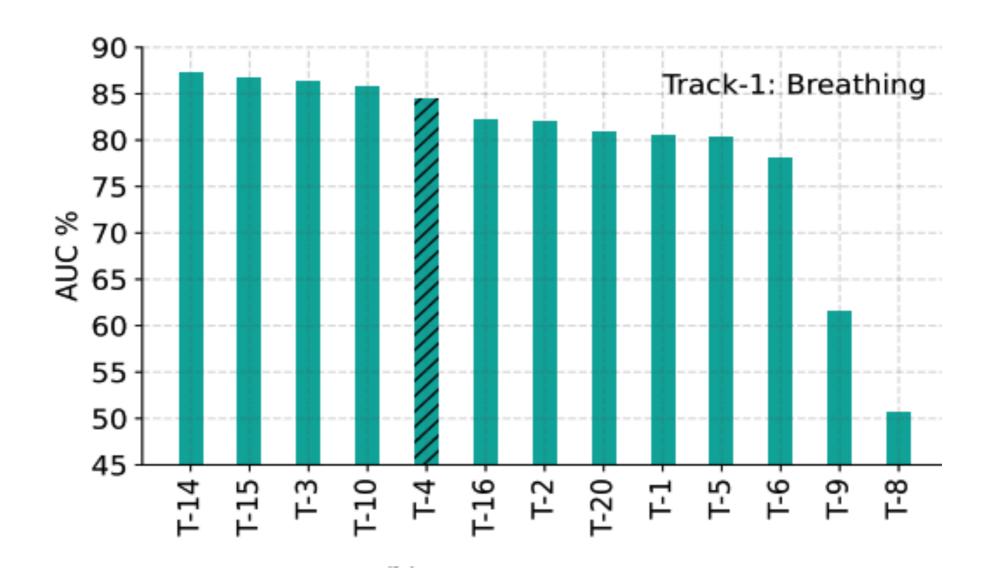
Distribution based on country of origin







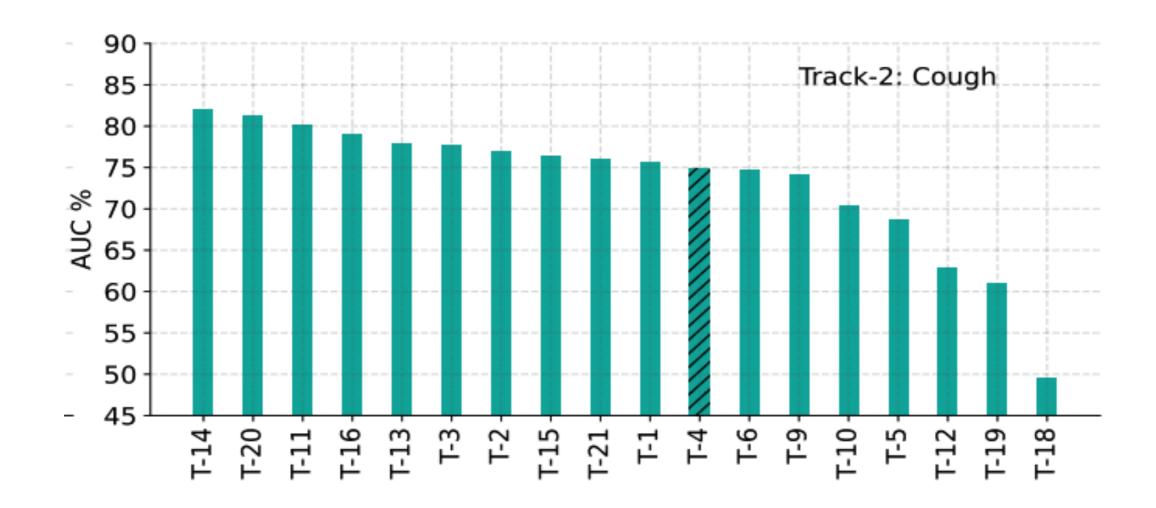








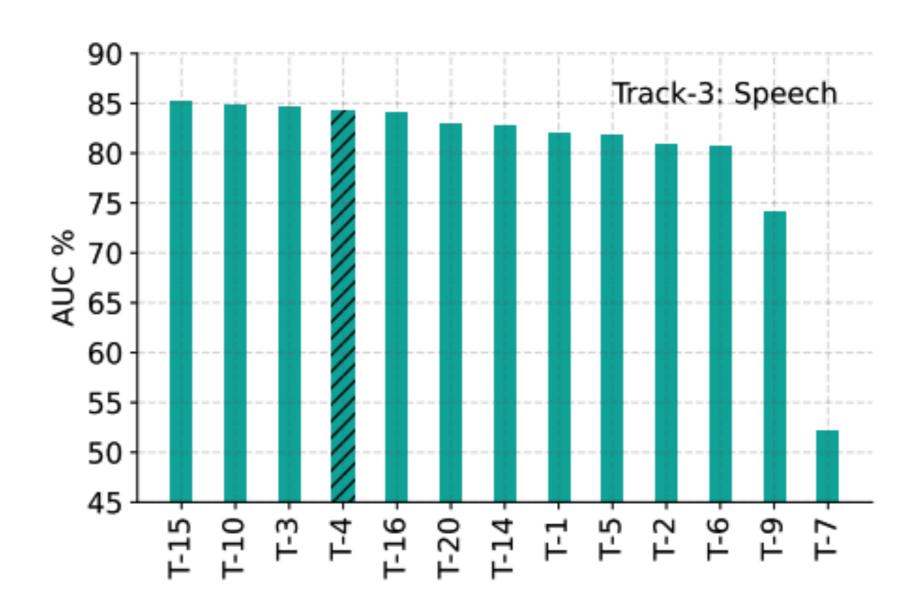








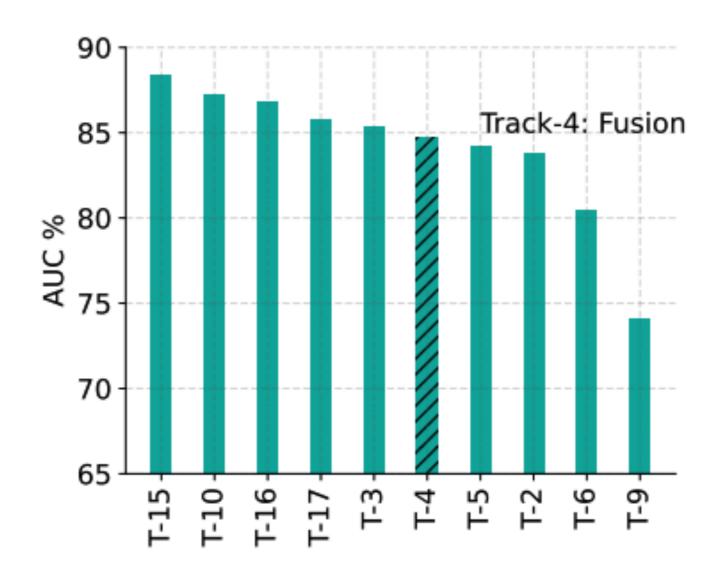










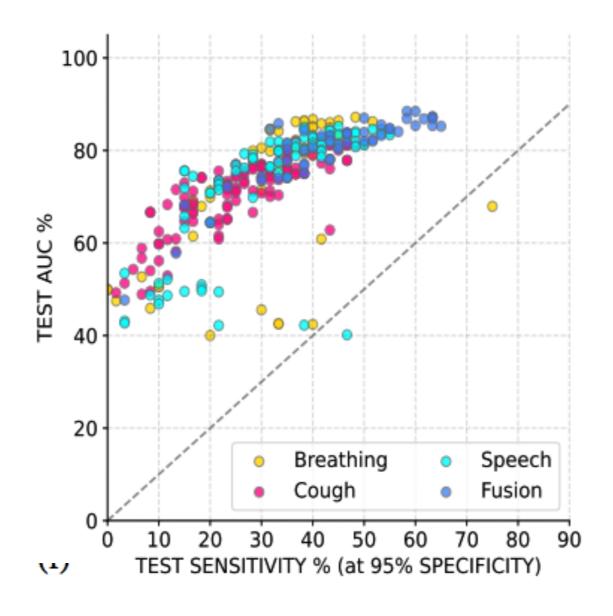








Blind Test AUC versus Sensitivity







### **Conclusions**



- Few teams surpassed the baseline
- Breathing and speech signals showed effectiveness
- Effectiveness of fusion strategy







# Q&A



