

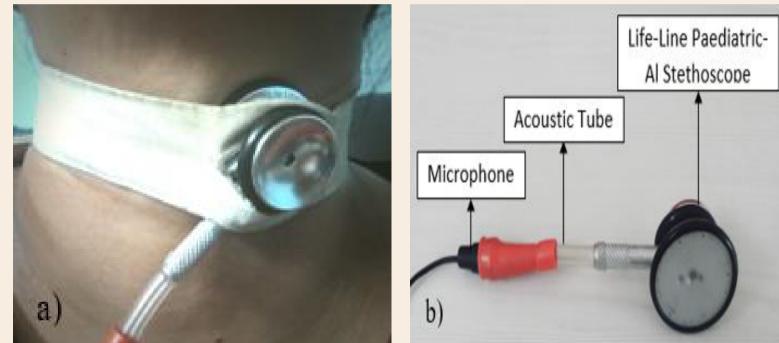


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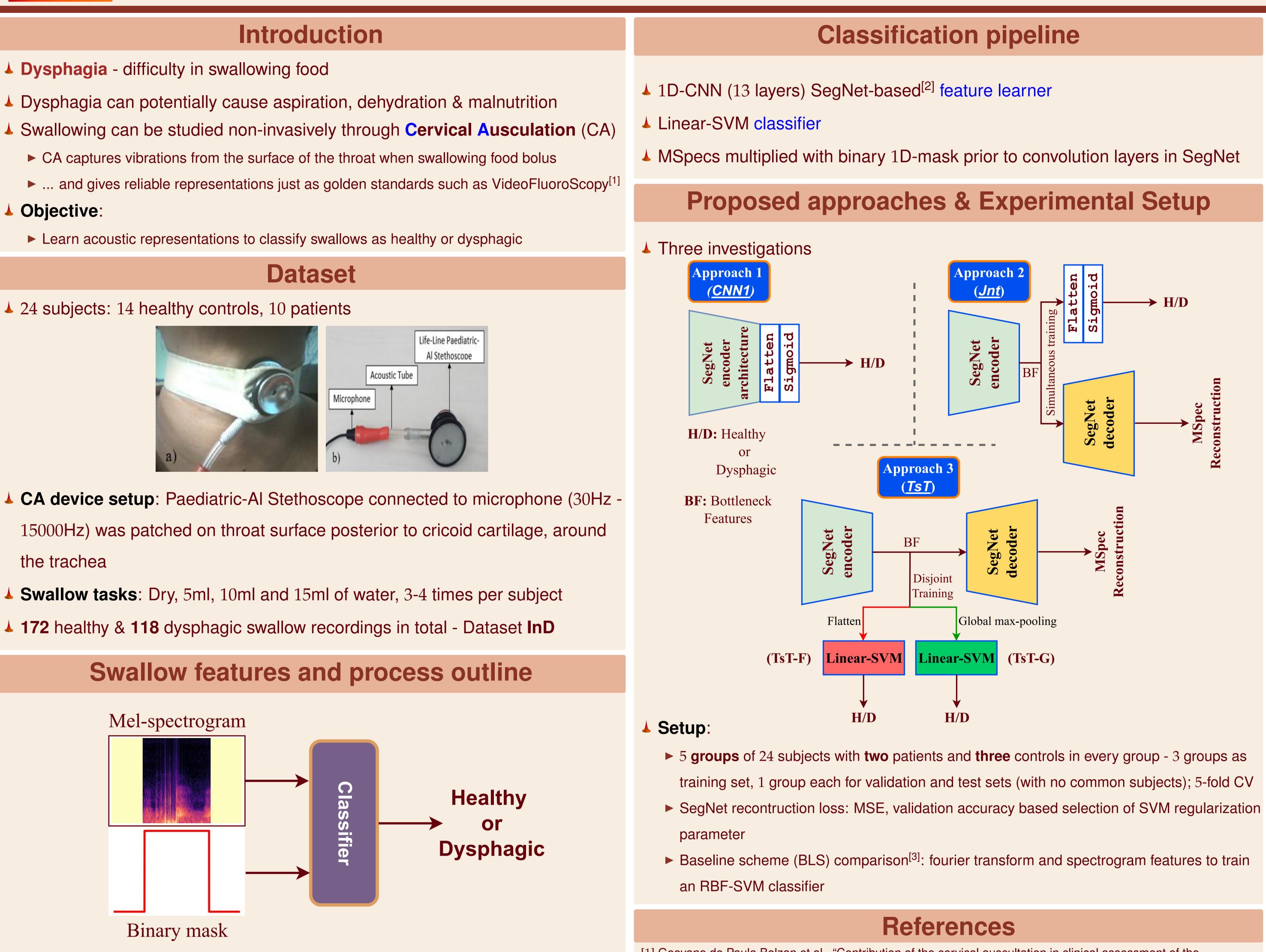
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- Dysphagia difficulty in swallowing food

- **A** Objective:



- the trachea



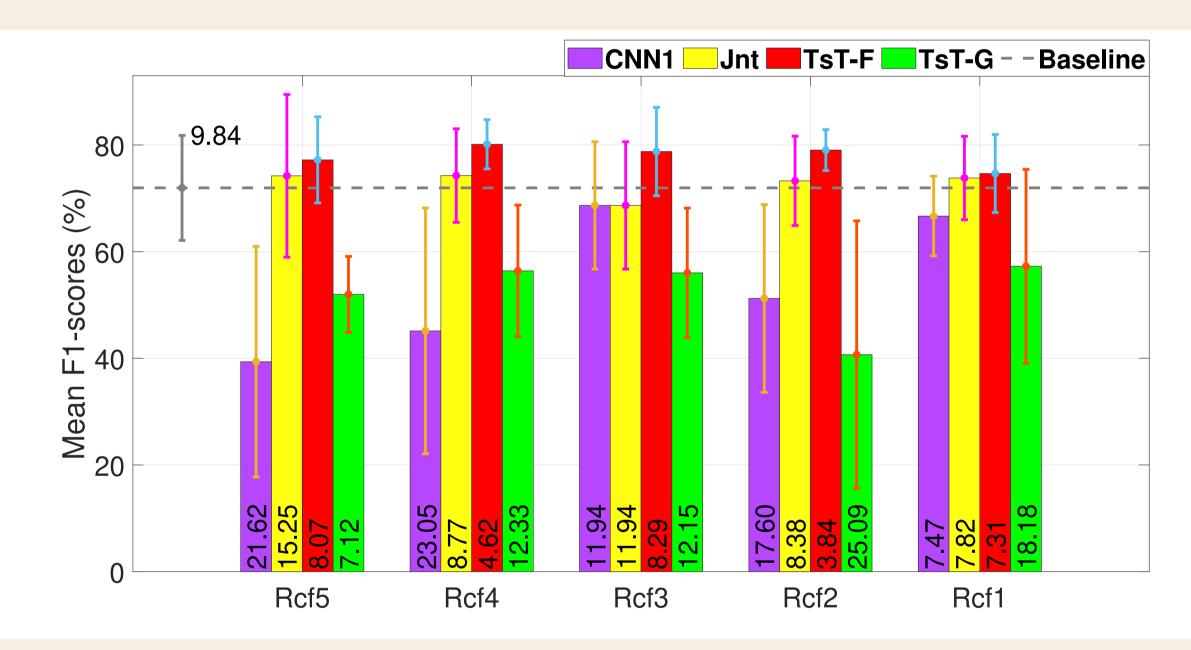
- Input feature: Mel-Spectrogram (MSpec) for every swallow signal was computed with 20ms long hamming winsow and 2ms long hop
- Pre-processing MSpecs (not signals) padded with zeros to equalize time length of all MSpecs

# **SEGNET-BASED DEEP REPRESENTATION LEARNING FOR DYSPHAGIA CLASSIFICATION**

[1] Geovana de Paula Bolzan et al., "Contribution of the cervical auscultation in clinical assessment of the oropharyngeal dysphagia," Revista CEFAC, vol. 15, no. 2, pp. 455-465, 2013. [2] Vijay Badrinarayanan et al., "Segnet: A deep convolutional encoder-decoder architecture for image segmentation," IEEE trans. on pattern analysis and machine intelligence, vol. 39, no. 12, pp. 2481–2495, 2017. [3] Shigeyuki Miyagi et al., "Classifying dysphagic swallowing sounds with support vector machines," in Healthcare Multidisciplinary Digital Publishing Institute, 2020, vol. 8, p. 103.

- progressing maxpooling layers in SegNet encoder

	Mean F1-score (%)	Mean Sensitivity (%)	Mean Specificity (%)	Mean SD (%)
CNN1	54.21	43.05	62.14	±16.33
Jnt	72.84	72.91	62.92	±10.43
TsT-F	77.96	78.29	66.98	±6.43
TsT-G	52.46	49.28	47.73	±14.97



- BLS suffered on InD (larger and imbalanced data)
- F1-score and lowest mean SD across all RCF*i*-trials

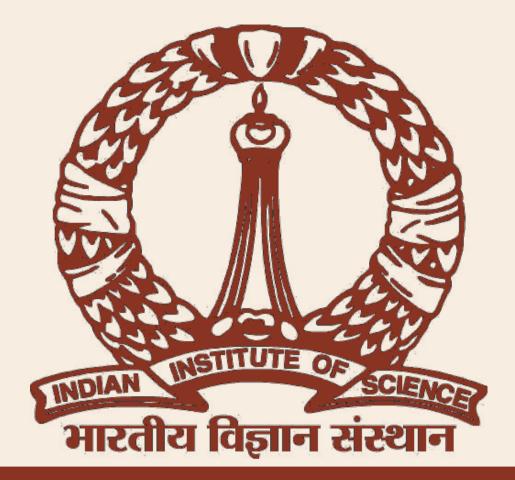
## than fully supervised training

transition patterns

### Future directions

- Adapt proposed method to identify severity of dysphagia

- of Science and Technology, Govt. of India for their support.



## **Results & Discussion**

Depth analysis: 5 "RCFi-trials" with bottleneck features from each of the 5

Learning at different depths identifies time-scale resolution that maximizes performance Each maxpooling layer has Receptive Field Size (RCF) of 3, 10, 26, 58, 122 respectively

▲ Best: TsT-F in RCF4-trial with a mean (across folds) test F1-score of 80.13%

▲ Jnt performance only competitive with BLS but still showed low F1-scores TsT-F outperformed all approaches, including TsT-G, with highest mean

### Conclusion

Unsupervised feature learning followed by supervised training performed better

Performance variation at different depths hints the influence of swallow phase

Explore time-scale analysis to outline spectral signatures pertaining to levels of severity **Acknowledgement:** The authors thank the Pratiksha Trust and the Department

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