

Vocal Fold Dynamics for Automatic Detection of Amyotrophic Lateral Sclerosis from Voice

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Problem Overview

This project explores the use of estimated vocal fold dynamics for automatic detection of Amyotrophic Lateral Sclerosis (ALS) from voice recordings. The project objectives are:

- ✓ Testing the viability of proposed method of analysis.
- ✓ Devising a set of effective and reliable features for ALS detection.

Background Information

- Human voice** carries enormous information about the physical and physiological states of the speaker and is thus suitable for the extraction of diagnostic features.
- ALS** is a neuro-degenerative disease that affects the motor neurons. The complexity of current diagnostic methods calls for expedient, objective, and reliable diagnostic aids.
- Prior studies** on the subject showed promising results but have not indicated the existence of conclusive biomarkers for ALS diagnosis.

Data Preparation

Dataset: voice recordings of prolonged phonation of "ah" of 42 individuals (32 ALS, 10 non-ALS)

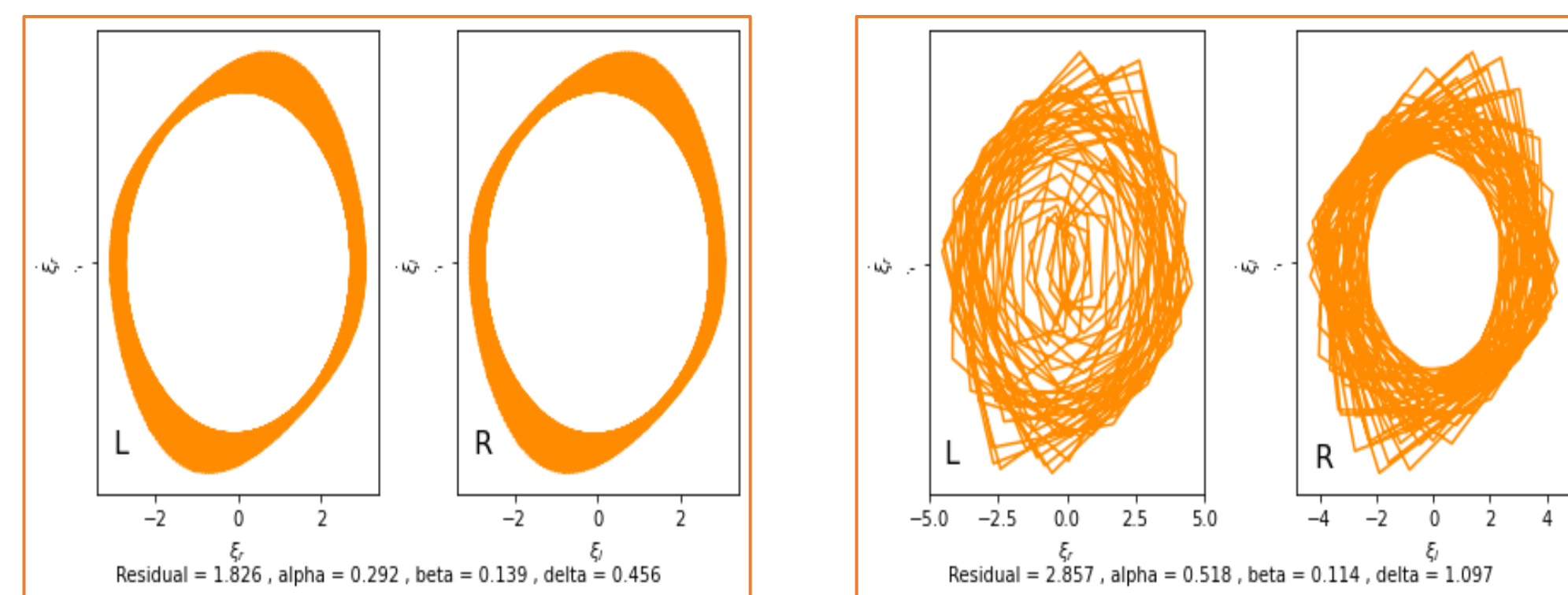


Fig. 1 Phase portraits showing estimated vocal fold velocity vs. displacements for 2 analysis windows of a single speaker.

Adjoint Least-Squares (ADLES) algorithm [1] based on a model of asymmetric vocal fold oscillation is used to estimate the vocal fold dynamics for each speaker.

The experiment setup: 10-way cross-validation conducted, with 70:30 train-test split. The whole dataset was used in each run; there were no overlap between training and test data.

Feature Set 1 (FS1): Simple Statistical Measurements

20 features investigated:

- ✓ **avg, max, min, var** of vocal fold displacement (*disp*);
- ✓ **avg, max, min, var, center, skew** of max range of displacement (*MROD*)

7 features show significant separation between case and control after FDR correction (10 before correction).

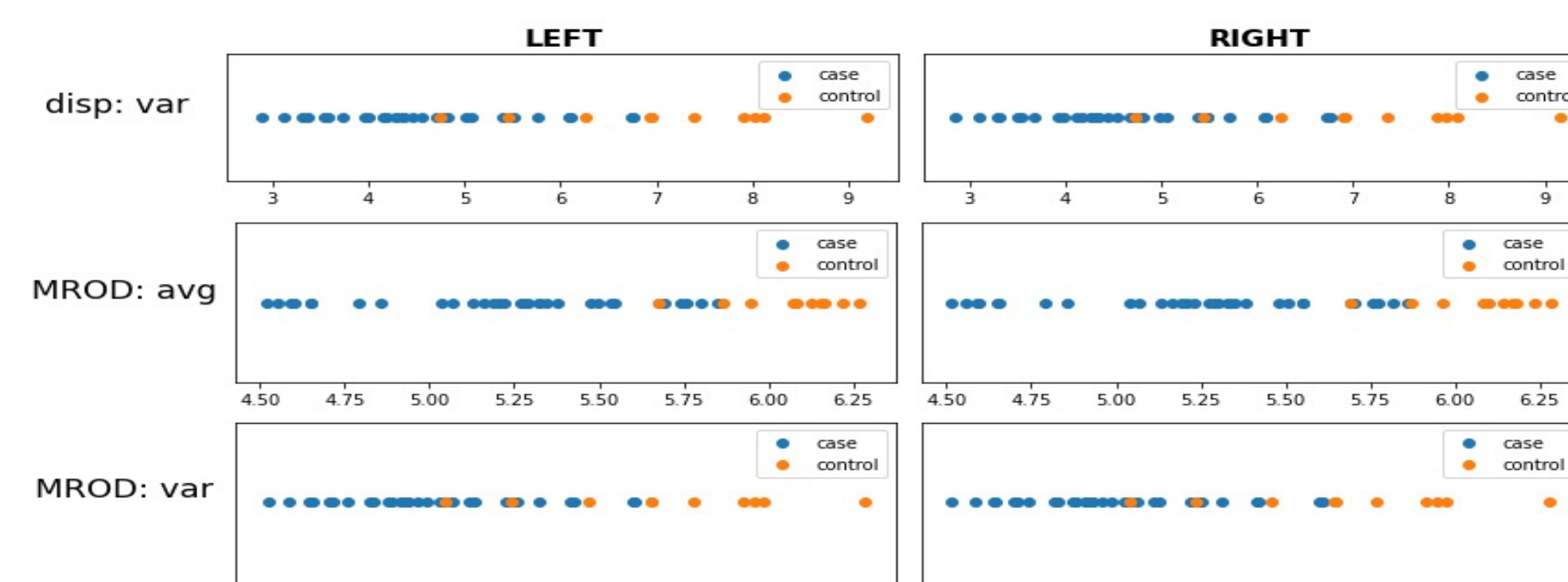


Fig. 2: Distributions of some FS1 features (significant separation after correction)

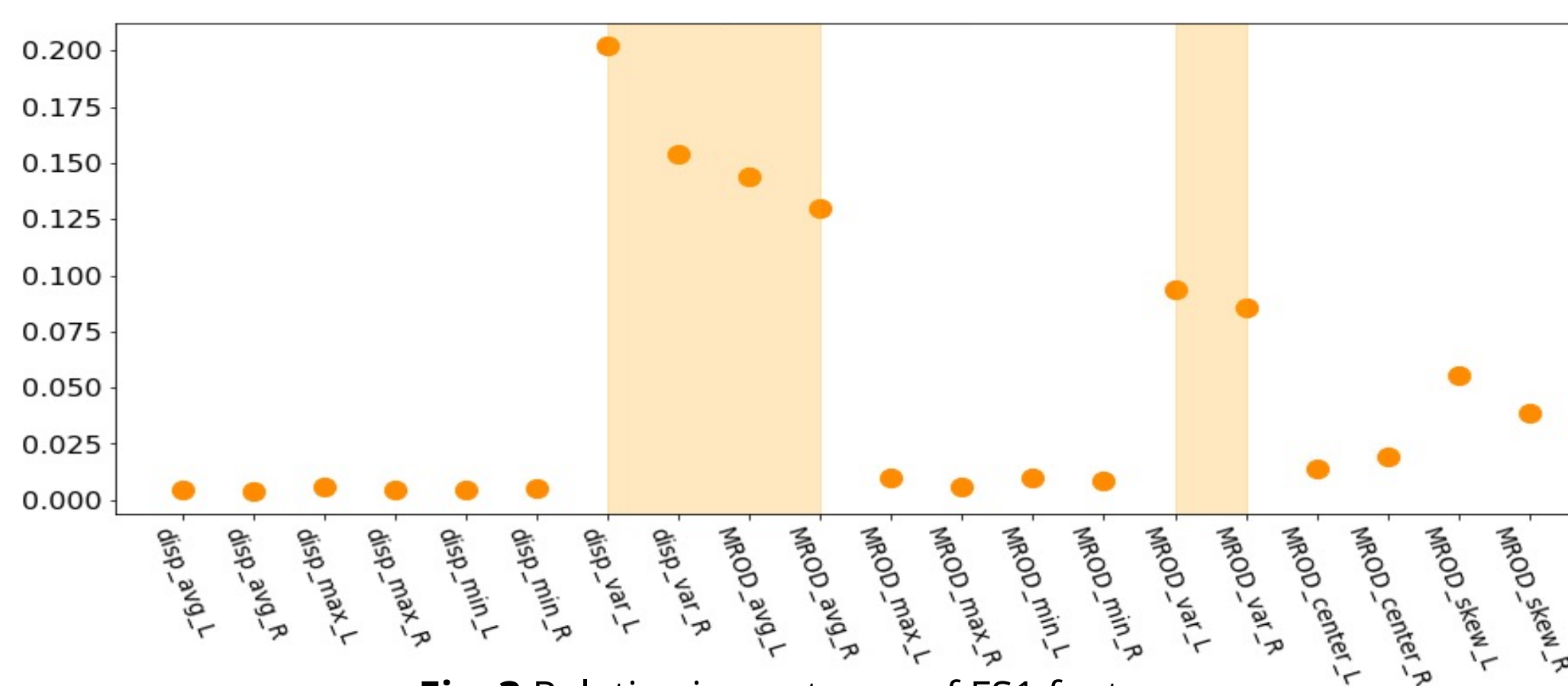


Fig. 3 Relative importance of FS1 features. *disp: var, MROD: {var, avg}* account for 80.7% of feature weight.

Classification Results:

- Random Forest Classifiers (max-depth=2) with FS1 achieved **average AUC-ROC of 99.667%** (var=0.0001).

Conclusions

- ALS decreases range of motion and capability of motion variation of the vocal folds.
- Vocal fold dynamics is useful for separating ALS from non-ALS individuals.
- It's **surprisingly easy** to detect the presence of ALS with high accuracy with proposed features!

Reference

[1] Zhao, W. and Singh, R., 2020, May. Speech-based parameter estimation of an asymmetric vocal fold oscillation model and its application in discriminating vocal fold pathologies. In *ICASSP 2020-2020 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP)* (pp. 7344-7348). IEEE.

Feature Set 2 (FS2): Phase-Space Characterizations

Features based on Lyapunov and Hurst exponents are extracted to measure the stability and auto-correlation of inferred dynamics.

17 features investigated:

- ✓ **max 1-5** among all *Lyapunov* exponents
- ✓ **avg, max, var** of *Hurst* exponents for velocity and displacement data of both sides

1 feature shows significant separation between case and control after FDR correction (12 before correction).

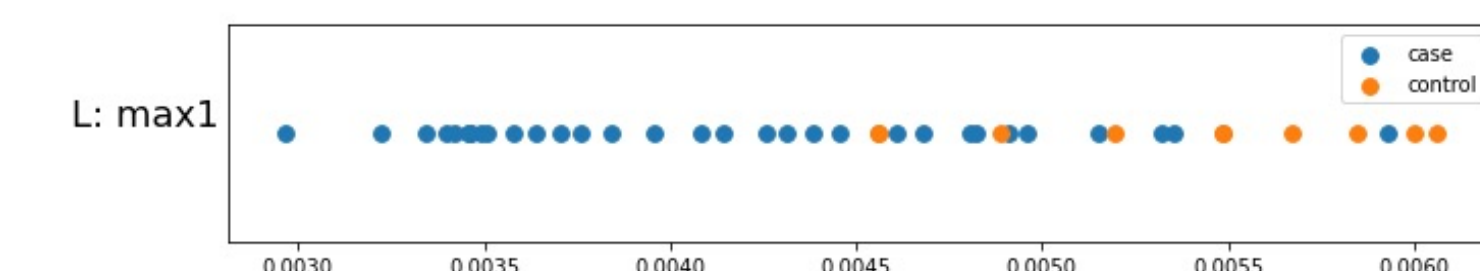


Fig. 4 Distribution of *Lyapunov: max1* (significant separation after correction)

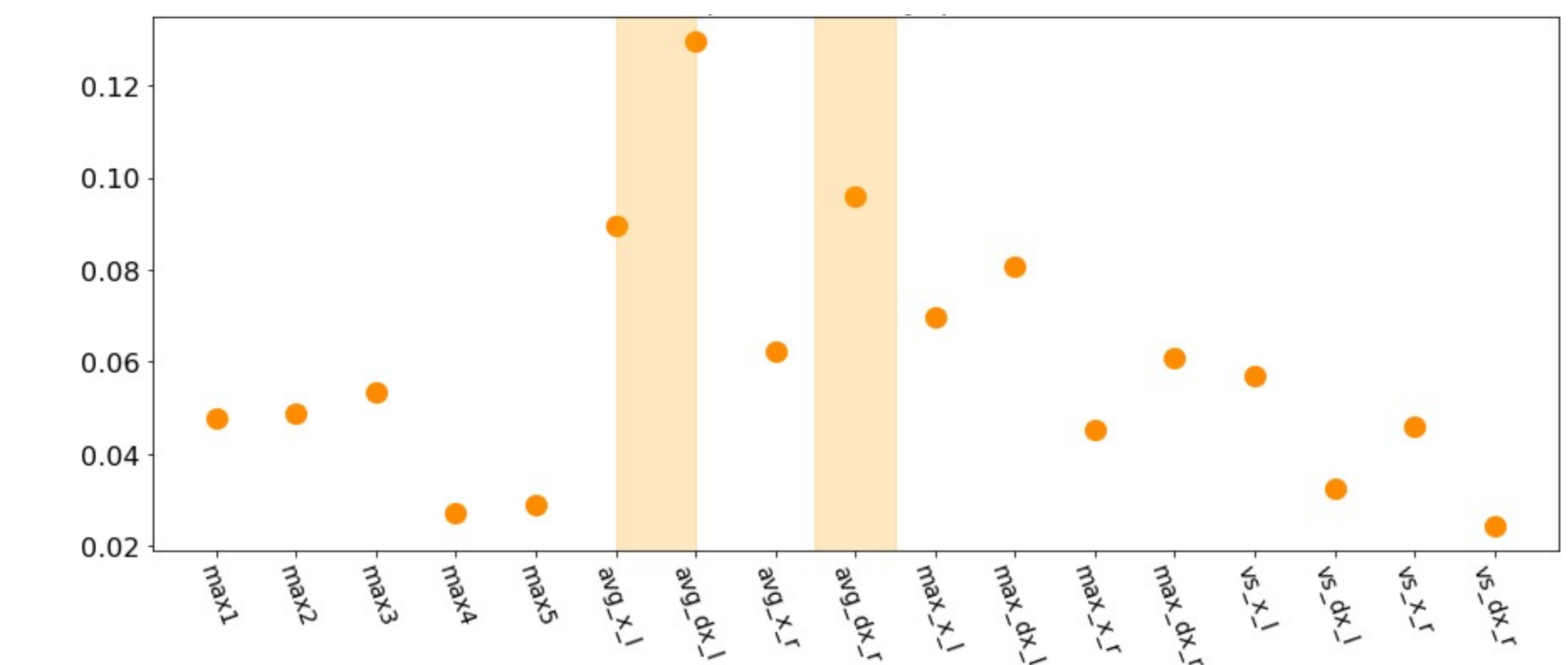


Fig. 5 Relative importance of FS2 features. *Hurst: {avg_dx_L, avg_dx_R, avg_x_L}* account for 31.5% feature weight.

Classification Results:

- Random Forest Classifiers (max-depth=2) with FS2 achieved **average AUC-ROC of 82.333%** (var=0.01312).

Future Directions

- Increase cohort size; improve sample case-control, gender balance; increase sample diversity.
- Increase the complexity of the physical model of phonation.
- Relate findings (feature distributions, relative weights) with physiological and disease mechanisms.