

Contact Localization via Active Oscillatory Actuation

Divas Subedi¹, Elizabeth Schoemer¹, Digesh Chitrakar¹,
Yun-Hsuan Su², and Kevin Huang¹

¹Trinity College, Department of Engineering, 300 Summit St, Hartford, CT 06106

²Mount Holyoke College, Department of Computer Science, 50 College St, South Hadley, MA 01075

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Introduction

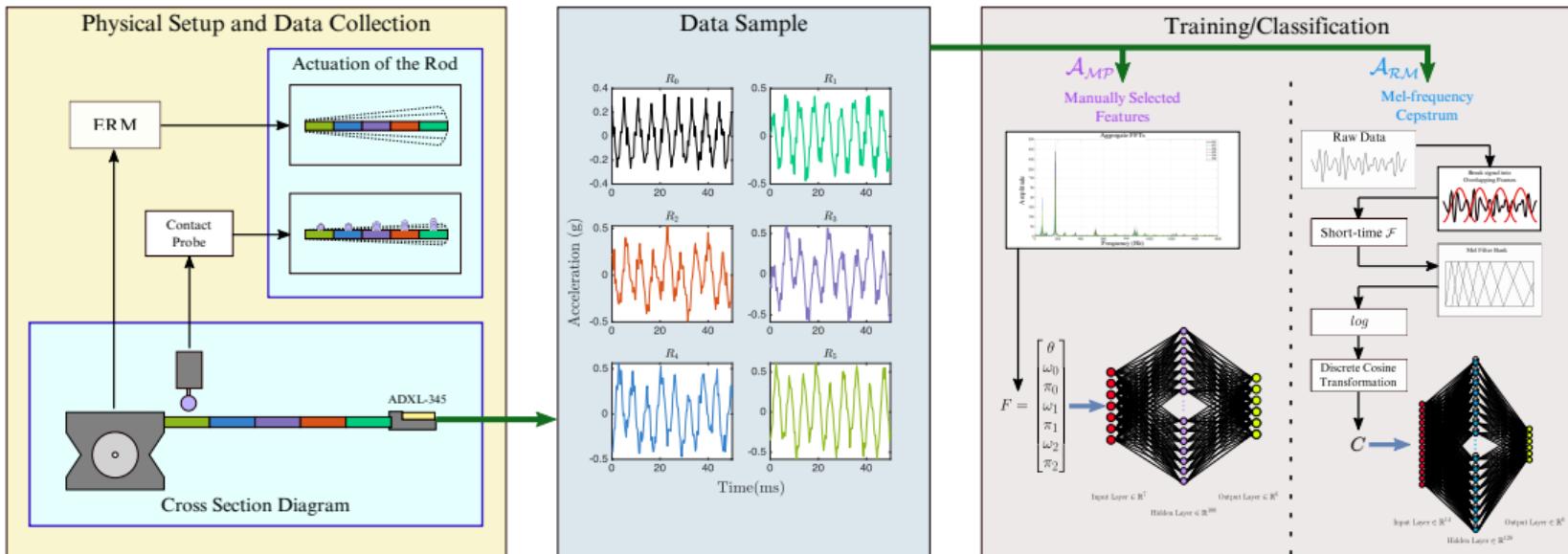
Introduction

Motivation for this study:

- ▶ Contact localization in robotics.
 - ▶ Grasping, manipulation
 - ▶ Telelocomotion
 - ▶ Medical robotics
- ▶ Current technologies are expensive.
 - ▶ Piezoelectric
 - ▶ Serial links
- ▶ Seek non-intrusive and readily implemented alternative.
 - ▶ Accelerometers are inexpensive and non-intrusive.

Procedure

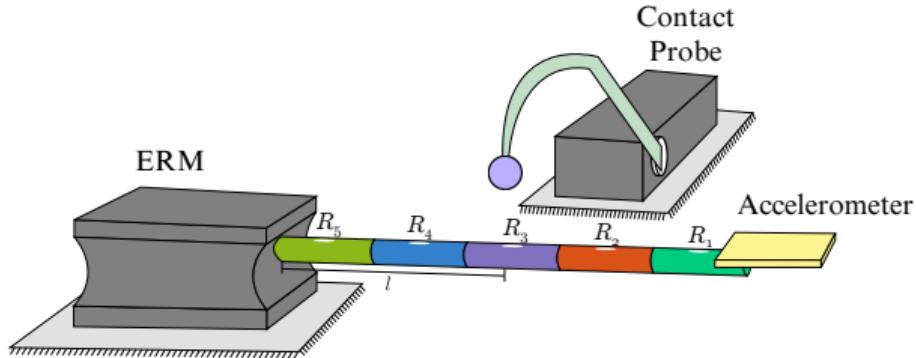
Procedure



Procedure

Hardware Setup

The hardware setup consists of a rigid link, contact probe, ERM, and an accelerometer. The ERM actively actuates the rigid link, while contacts are made at appropriate locations between $R_1 - R_5$, R_0 is untouched. Inertial data is collected via the IMU.



Contact Point	R_1	R_2	R_3	R_4	R_5
$D[\text{mm}]$	10.0	30.6	51.2	71.8	92.4

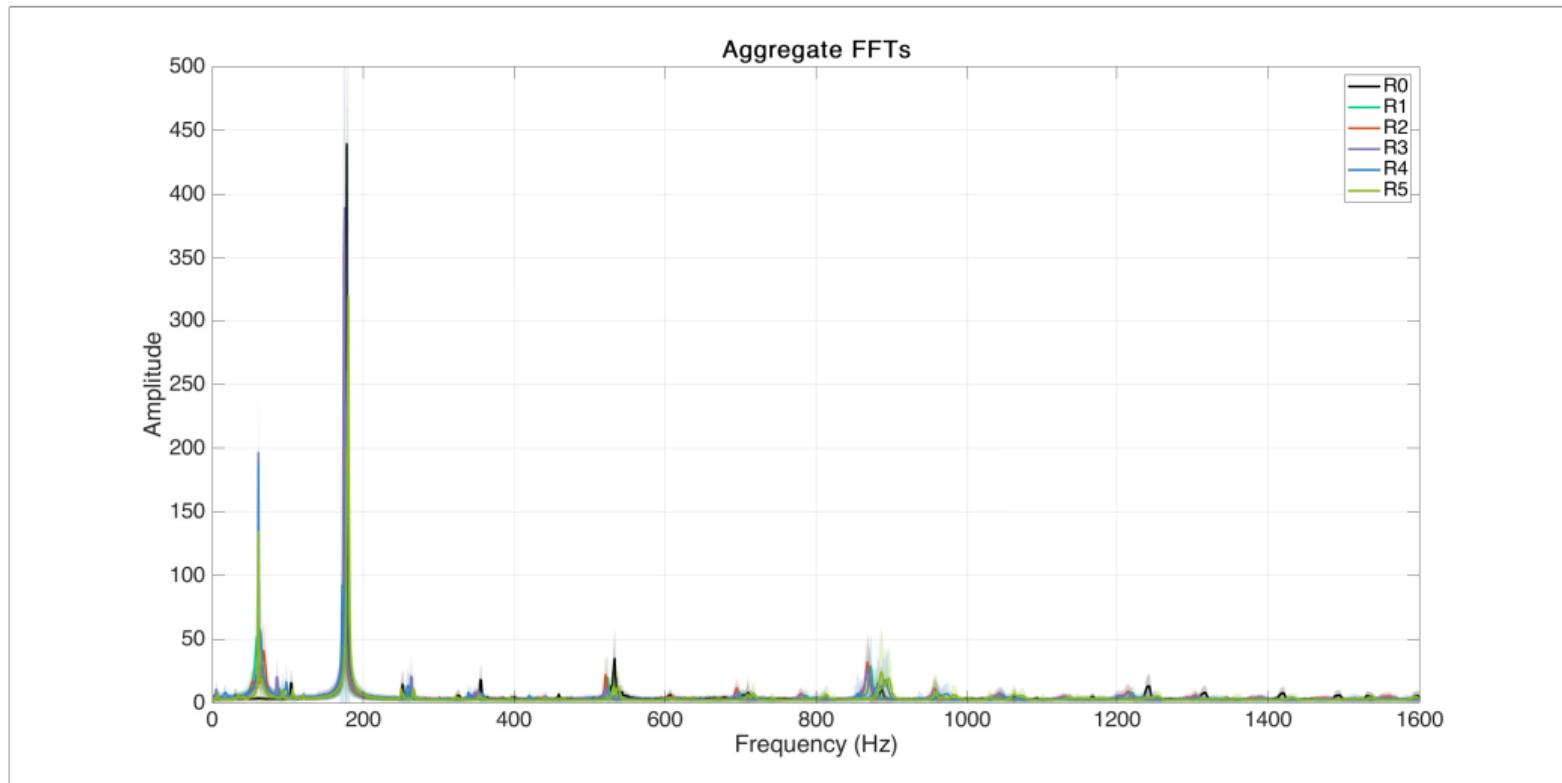
Procedure

Data Collection

- ▶ The contact probe and the ERM actuator were automated using a Raspberry Pi.
- ▶ The ERM was actuated for one second with no load and constant current.
- ▶ Accelerometer measured, at 3200 Hz, the 3-axis acceleration.
- ▶ The setup is reset and allowed to rest for 180 seconds.
- ▶ 150 samples were collected for each section to form an aggregate database of 900 one-second time series.

Classification

Manual Features



Manual Features

$$F = [\theta \quad \omega_0 \quad \pi_0 \quad \omega_1 \quad \pi_1 \quad \omega_2 \quad \pi_2]^T \quad (1)$$

Feature space constructed from seven measures

- ▶ θ : atan2 of ratio of RMS vertical and horizontal acceleration
- ▶ ω_i : peak frequency within band i
- ▶ π_i : peak prominence within band i

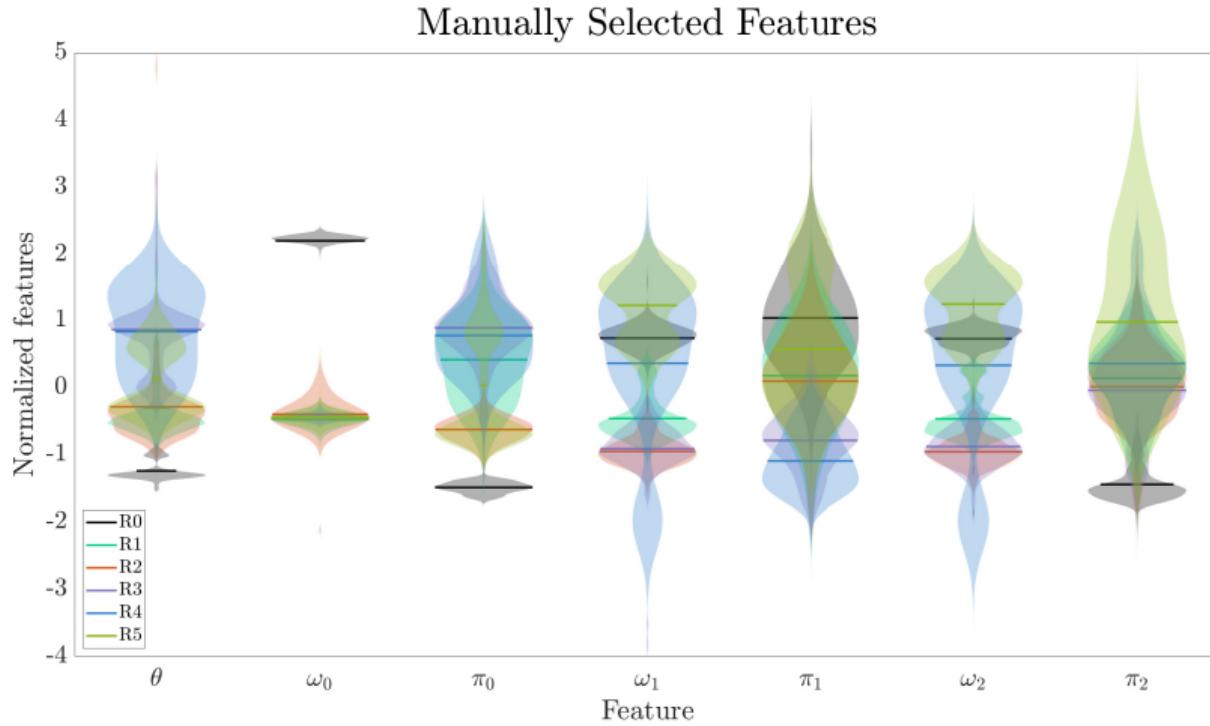
Bands were empirically defined from the Aggregate FFT.

Band 0: [0, 120] Hz

Band 1: [500-550] Hz

Band 2: [825-925] Hz

Manual Feature Space



MFCC Features

$$X(k) = \mathcal{F}\{x(n)\} \quad (2)$$

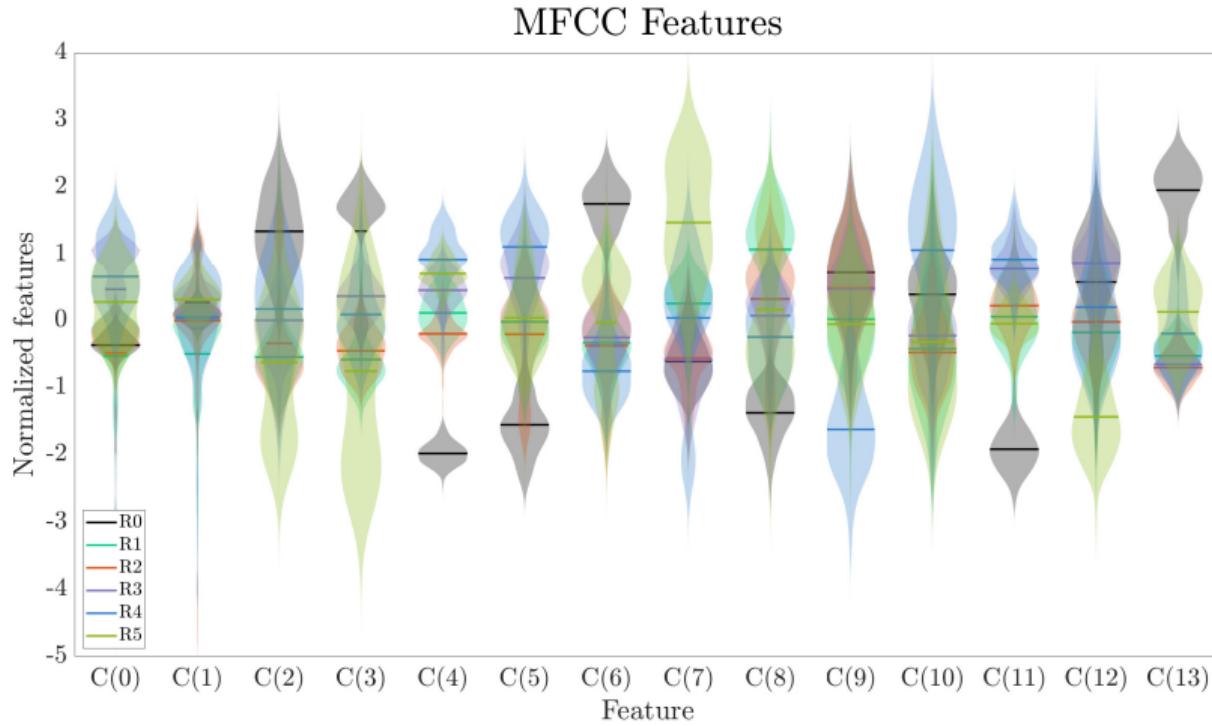
$$f_{mel} = 2595 \log_{10} \left(1 + \frac{f}{700} \right) \quad (3)$$

$$s(m) = \sum_{k=0}^{N-1} [|X(k)|^2 H_m(k)] ; \quad 0 < m < M - 1 \quad (4)$$

$$C(n) = \sum_{m=0}^{M-1} \log_{10}(s(m)) \cos \left(\frac{\pi n(m - 0.5)}{M} \right) \quad (5)$$

The first 14 coefficients constitute the \mathcal{A}_{RM} feature space, i.e. $0 \leq n \leq 13$ in (5).

MFCC Feature Space



Classification Algorithms

Two contact localization algorithms were proposed in this work:

1. \mathcal{A}_{MP} : Multiclass Perceptron with manually selected features, F

Input: Seven manually selected features in F

Structure: A single layer of 100 perceptrons

Optimization: Scaled conjugate gradient backpropagation to optimize cross entropy

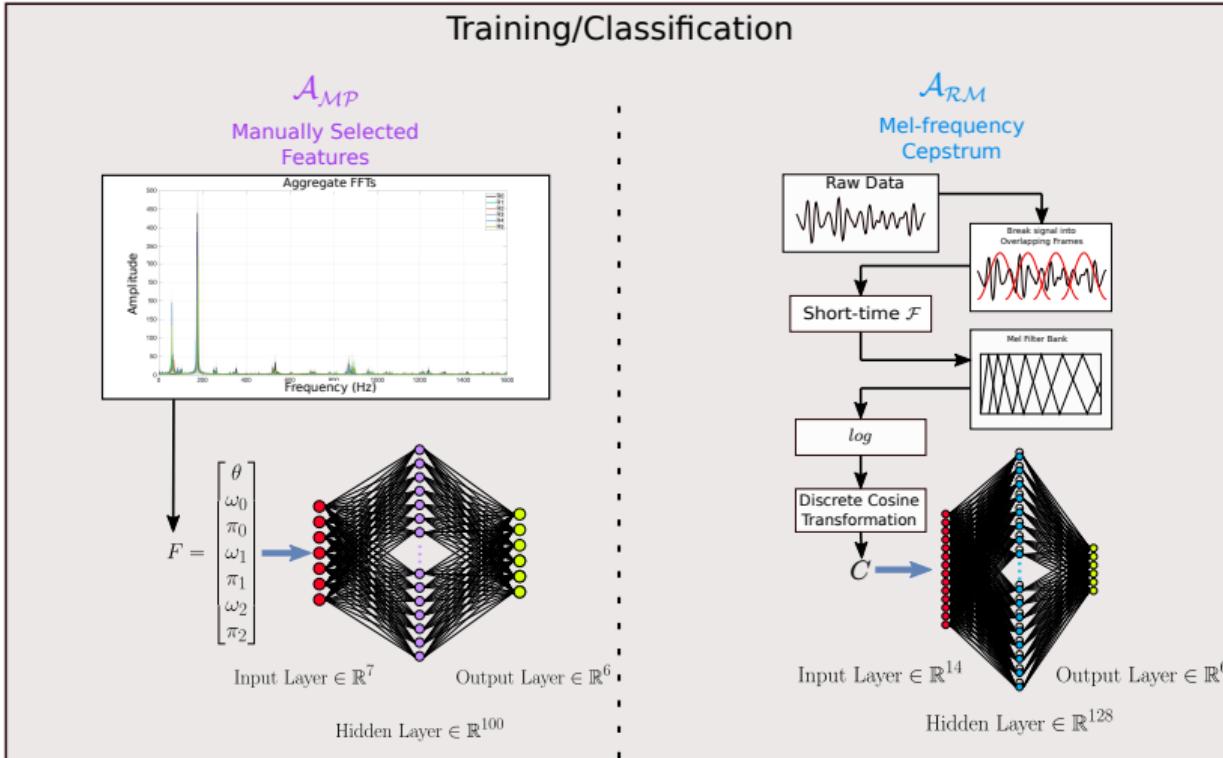
2. \mathcal{A}_{RM} : Recurrent Neural Network - with Mel-frequency cepstrum features C .

Input: Sequences of 98 C feature vectors per sample

Structure: Bidirectional LSTM recurrent neural network with 128 hidden units

Activation: Sigmoid gate activation and hyperbolic tangent state activation

Classification Algorithms



Results

Classification with $\mathcal{A}_{\mathcal{MP}}$

Confusion Matrix $\mathcal{A}_{\mathcal{MP}}$ (Feature Space F)

True Class	Predicted Class					
	R_0	R_1	R_2	R_3	R_4	R_5
R_0	150					
R_1		127	14	8	1	
R_2		3	147			
R_3		9		141		
R_4				1	148	1
R_5		1		1	1	147

Classification with \mathcal{A}_{MP}

Table 1: Multi-class Perceptron Classifier (\mathcal{A}_{MP}) Performance

Class	Precision	Recall	Accuracy	F1-score
R_0	1.0	1.0	1.0	1.0
R_1	0.9071	0.8467	0.9600	0.8759
R_2	0.9130	0.9800	0.9811	0.9453
R_3	0.9338	0.9400	0.9789	0.9369
R_4	0.9867	0.9867	0.9956	0.9867
R_5	0.9932	0.9800	0.9956	0.9866

Overall accuracy : 95.56%

Classification with $\mathcal{A}_{\mathcal{RM}}$

		Confusion Matrix $\mathcal{A}_{\mathcal{RM}}$ (Feature Space C)					
		R_0	R_1	R_2	R_3	R_4	R_5
True Class	R_0	150					
	R_1		139	4	6		1
	R_2		1	144	5		
	R_3		1		147	2	
	R_4		1		3	146	
	R_5				2		148
	Predicted Class	R_0	R_1	R_2	R_3	R_4	R_5

Classification with \mathcal{A}_{RM}

Table 2: Bi-LSTM Classifier (\mathcal{A}_{RM}) Performance

Class	Precision	Recall	Accuracy	F1-score
R_0	1.0	1.0	1.0	1.0
R_1	0.9789	0.9267	0.9844	0.9521
R_2	0.9730	0.9600	0.9889	0.9664
R_3	0.9018	0.9800	0.9789	0.9393
R_4	0.9865	0.9733	0.9933	0.9799
R_5	0.9933	0.9867	0.9967	0.9900

Overall accuracy : 97.11%

Conclusion

Summary

Designed, constructed and evaluated a repeatable active contact localization mechanism

- ▶ Actuate oscillations on physical link
- ▶ Contact link in separate regions ($R_0 - R_5$)
- ▶ Measure frequency response
- ▶ Classify response to contact region using two different models
- ▶ We observe:
 - ▶ Ease of contact sensing
 - ▶ In general, further section from the actuator are harder to identify

Future Work

- ▶ Actuation Modality
- ▶ MFCC features can be tuned to task
- ▶ Expansion of data set in multiple dimensions
- ▶ Implementation of asymmetric rigid link
- ▶ Acoustic Contact Sensing

Thank you!

Thank you for your attention. The authors would like to thank the SII 2022 organizing committee and welcome all questions via email at divas.subedi@trincoll.edu.



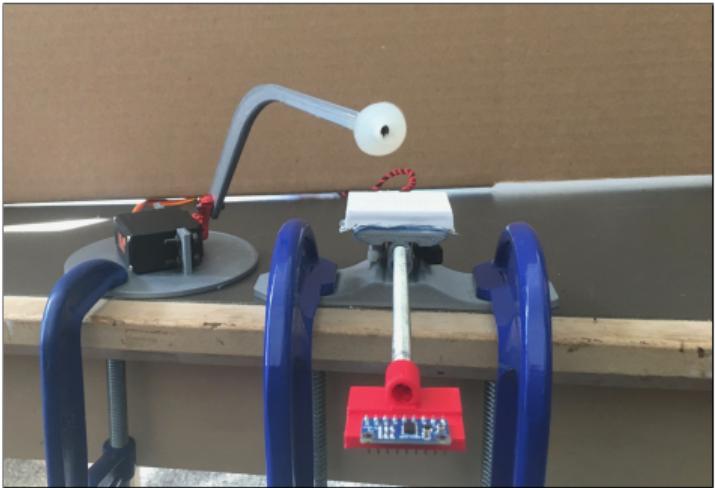
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Appendix

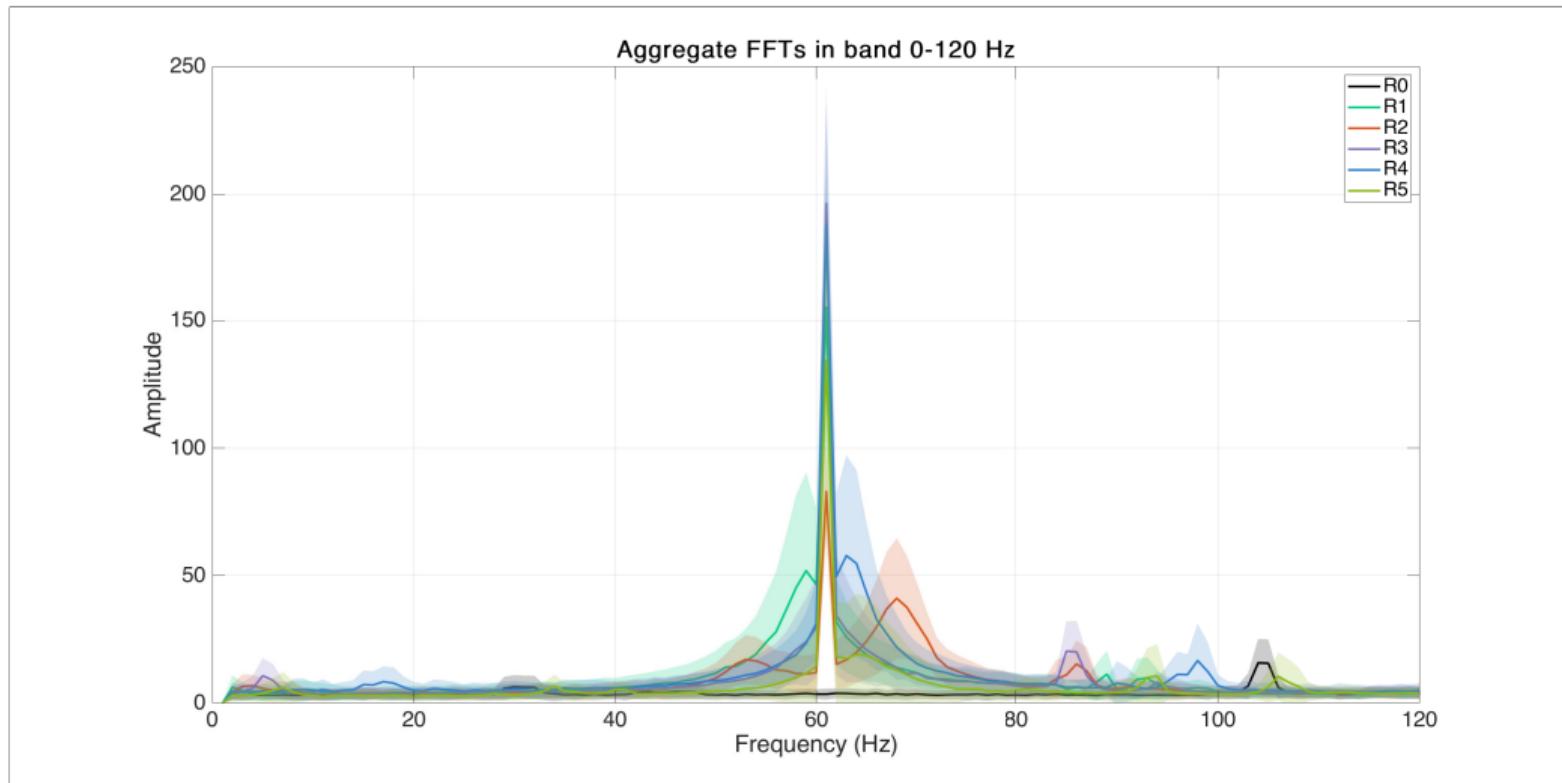
Hardware Setup Implementation

The physical setup is realized using:

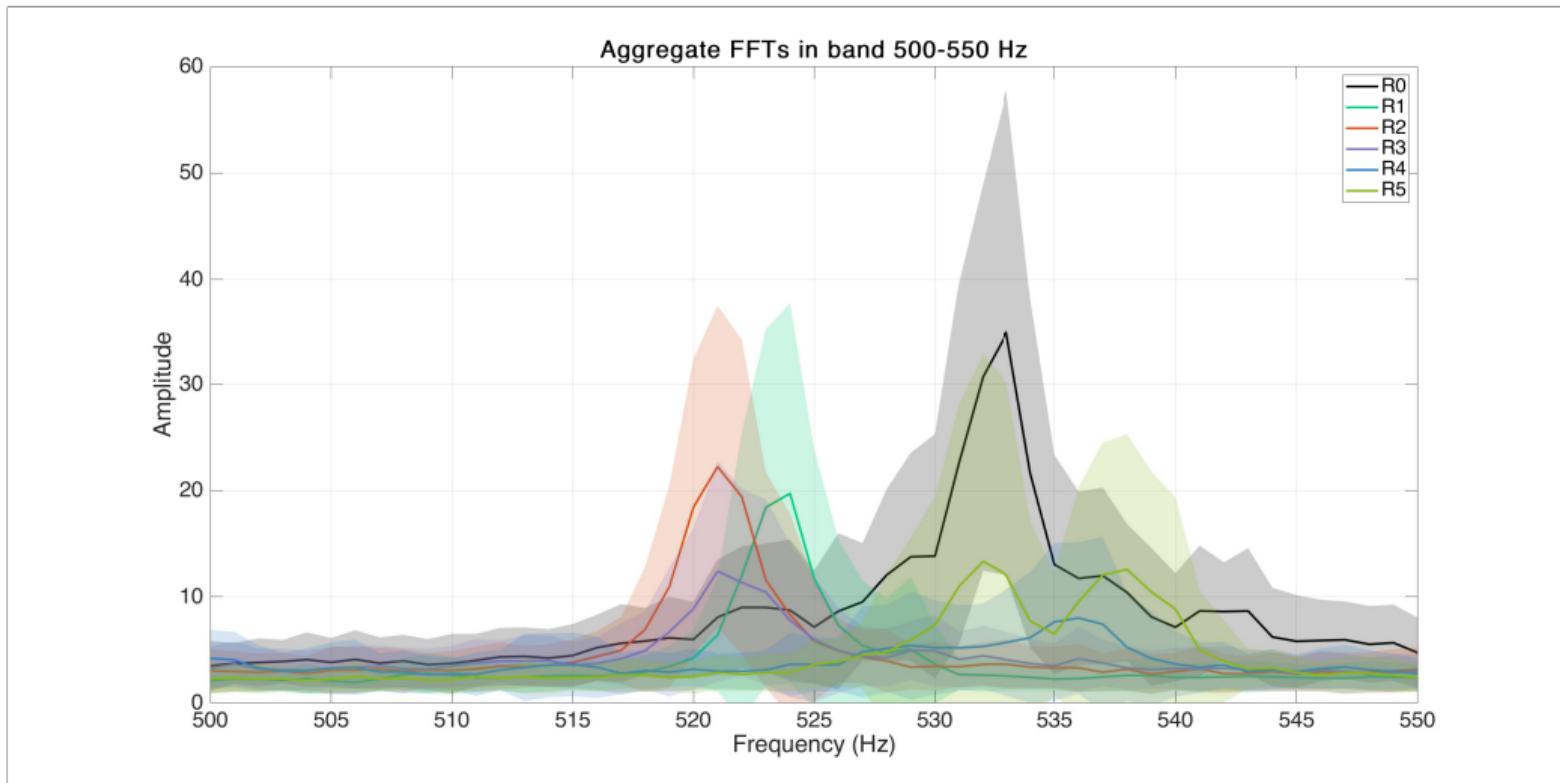
- ▶ ¼ inch Zinc Coated Steel Rod
- ▶ Secured with PLA mounts
 - ▶ ADXL 3-axis accelerometer
 - ▶ Eccentric Rotating Mass
 - ▶ MG995 metal gear servo



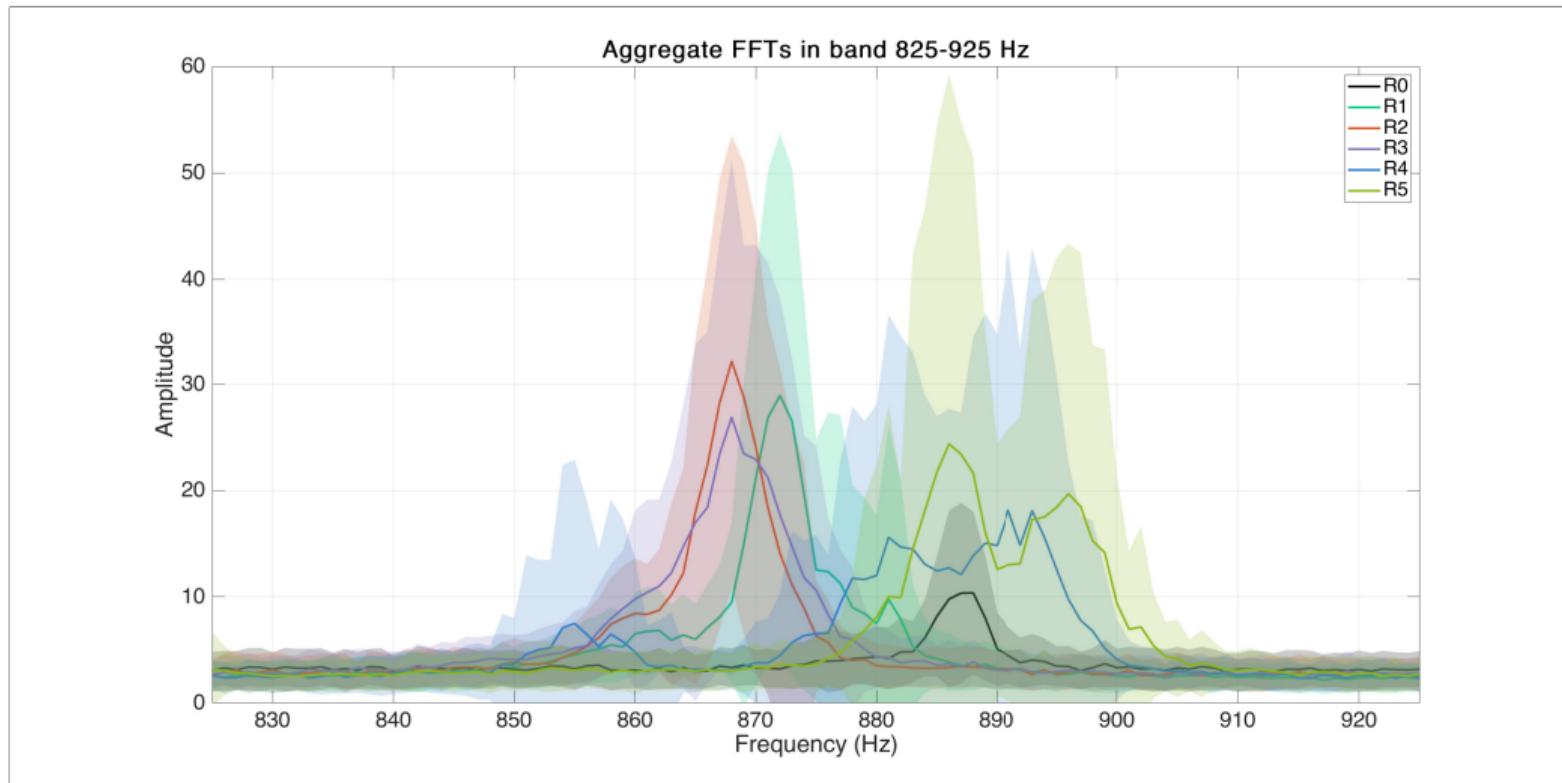
Band0



Band1



Band2



MFCC feature for a sample

