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1. **Sriram Kumar P** and **Jac Fredo AR**, "Optimal electrodermal activity segment for enhanced emotion recognition using spectrogram-based feature extraction and machine learning", International Journal of Neural Systems, 2024.
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2. **Sriram Kumar P**, Praveen Kumar G, Nagarajan G, and **Jac Fredo AR**, "A comparative analysis of EDA decomposition methods for improved emotion recognition", Journal of Mechanics in Medicine and Biology, 2023.
   DOI: 10.1142/S0219519423400432.
   (Impact Factor: **0.89**)

3. **Sriram Kumar P**, Praveen Kumar G, Nagarajan G, and **Jac Fredo AR**, "Electrodermal activity-based analysis of emotion recognition using temporal-morphological features and machine learning algorithms", Journal of Mechanics in Medicine and Biology, 2023.
   DOI:10.1142/S0219519423400444.
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4. **Sriram Kumar P**, Praveen Kumar G, Abdul ASG, Nagarajan G, and **Jac Fredo AR**, "Deep learning-based automated emotion recognition using multimodal physi-

ological signals and time-frequency methods". IEEE Transactions on Measurement and Instrumentation, Manuscript ID: TIM-24-00978 (Under Revision).
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5. **Sriram Kumar P** and **Jac Fredo AR**, "Emotion classification through optimal segments of EDA and texture analysis of time-encoded images with artificial intelligence". IEEE Transactions on Measurement and Instrumentation, Manuscript ID: TIM-S-24-02382 (Under Revision)
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Presentations in International Conferences

1. **Sriram Kumar P**, Praveen Kumar G, Nagarajan G, and **Jac Fredo AR**, "Comparative analysis of electrodermal activity decomposition methods in emotion detection using machine learning", Medical Informatics Europe (MIE), Gothenburg, Sweden, May 22-25, 2023.
DOI: 10.3233/SHTI230067.

2. **Sriram Kumar P** and **Jac Fredo AR**, "Investigating the effects of two-class categorical emotion classification through electrodermal activity and machine learning", IEEE Fifth International Conference on Cybernetics, Cognition and Machine Learning Applications (ICCCMLA), Hamburg, Germany, October 7-8, 2023.
DOI: 10.1109/ICCCMLA58983.2023.10346868.

3. **Sriram Kumar P** and **Jac Fredo AR**, "Enhancing emotion recognition: Machine learning with phasic spectrogram texture features", IEEE Fifth International Conference on Cybernetics, Cognition and Machine Learning Applications (ICCCMLA), Hamburg, Germany, October 7-8, 2023.
DOI: 10.1109/ICCCMLA58983.2023.10346619.

4. **Sriram Kumar P** and **Jac Fredo AR**, "Identification of optimal EDA segment for improved emotion detection using time-series image encoding and machine

learning", 61st International Biomedical Sciences Instrumentation Symposium and Rocky Mountain Bioengineering Symposium, Louisiana, USA, April 11-14, 2024.

5. **Sriram Kumar P** and **Jac Fredo AR**, "A comparative study of time-frequency methods for the effective classification of categorical emotions using EDA and deep learning", 61st International Biomedical Sciences Instrumentation Symposium and Rocky Mountain Bioengineering Symposium, Louisiana, USA, April 11-14, 2024.

6. **Sriram Kumar P** and **Jac Fredo AR**, "Classification of emotional states using electrodermal activity and random forest", 22nd International Conference on Mechanics in Medicine and Biology, Bologna, Italy, September 19-21, 2022.

Curriculum Vitae

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   - **Doctor of Philosophy** (2021 - 2024)

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