

# SINJINI BANERJEE

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## TECHNICAL SKILLS

|                                 |  |
|---------------------------------|--|
| Programming Languages           | Julia   Python   C   Matlab   SQL  |
| Deep Learning Frameworks        | TensorFlow   PyTorch   JAX   Keras   |
| Software tools                  | Jupyter Notebook   Eclipse   Colab   Atom/Uber-Juno   Gurobi Optimiser   Simulink   Android Studio   Onshape |
| Hardware tools                  | Arduino   Programmable Logic Controllers (PLC)   Raspberry Pi  |
| Data Analysis and Visualization | NumPy   SciPy   Pandas   Scikit-Learn   Matplotlib   Seaborn   Plotly  |
| Statistical Analysis            | Hypothesis Testing   Regression Analysis   ANOVA   Statistical Modeling   Experimental Design                |

## EDUCATION

|   |  |             |                     |
|---|--|-------------|---------------------|
| Rutgers University, NJ                              | Ph.D. – Electrical Engineering (Specialization: Information & Signal Processing) | GPA 3.93/4  | Sept 2020 – Present |
| University at Buffalo – SUNY                        | M.S. – Electrical Engineering,<br>Thesis: Signal optimization                    | GPA 3.62/4  | June 2019           |
| Heritage Institute of Technology,<br>Kolkata, India | B.Tech – Applied Electronics & Instrumentation Engineering                       | GPA 8.35/10 | July 2016           |

## WORK EXPERIENCE

|   |                     |
|---|---------------------|
| <b>Graduate Research Assistant, Department of Electrical and Computer Engineering, Rutgers University</b> <ul style="list-style-type: none"><li><b>Measuring training variability from stochastic optimization using robust nonparametric testing</b> (TensorFlow, Pytorch, Jax)<ul style="list-style-type: none"><li>Formulated a robust hypothesis testing framework to quantify the top two critical instability factors in deep learning optimization, in collaboration with Pacific Northwest National Laboratory.</li><li>Delivered a novel model selection metric that surpasses traditional accuracy measures in capturing model similarity with 95% confidence, enabling identification of the optimal model ensemble size for a 10% improvement in reliable predictions.</li><li>Built, optimized, and fine-tuned over 1000 deep learning models (Feedforward Neural Networks, MLPs, CNNs, Vision Transformer, Large Language Models) using high-performance computing (HPC) infrastructure provided by Rutgers' Office of Advanced Research Computing, improving training time by 90%.</li></ul></li><li><b>Mitigating risks associated with prediction inconsistency of equally accurate deep net models in machine learning model markets</b> (TensorFlow, Pytorch, Jax)<ul style="list-style-type: none"><li>Designed a novel safety-critical audit tool to address the negative impacts of prediction inconsistency of deep learning models, achieving 99% accuracy across six datasets, in financial and medical risk mitigation applications.</li><li>Improved an algorithm to generate robust counterfactual explanations, providing actionable recourse to 95% of negatively impacted clients.</li></ul></li></ul> | Sept 2020 – Present |
| <b>Graduate Teaching Assistant, Department of Electrical and Computer Engineering, Rutgers University</b> <ul style="list-style-type: none"><li>Instructed over 90 first-year engineering students in developing core technical and analytical skills, including 3D CAD modeling (Onshape), MATLAB programming, and data analysis with Excel.</li><li>Developed hands-on coding assignments and real-world engineering challenges that strengthened students' understanding of MATLAB control flow, functions, and data plotting.</li><li>Guided students through 14 debugging sessions and best practices for writing efficient code.</li></ul>  | Sept 2024 – Present |

## INTERNSHIP

### **Intern, Department of Electrical Engineering, University at Buffalo (Julia)**

Aug 2019 – May 2020

- Developed a novel outlier-robust Recursive Least Squares (RLS) algorithm by integrating sparsity-aware modeling of outliers within a hierarchical optimization framework (HO-RLS), enhancing robustness in noisy environments by 20%.
- Benchmarked performance against eight state-of-the-art robust RLS variants, showing superior estimation accuracy and faster adaptation in stationary and non-stationary signal processing scenarios.
- Improved algorithm performance by 30% through parallel computing implementation on clusters available at the Center for Computation Research at the University at Buffalo.

## PAPERS

- Banerjee, S., Marrinan, T., Cannon, R., Chiang, T., & Sarwate, A. D. (2024). Measuring training variability from stochastic optimization using robust nonparametric testing. arXiv preprint arXiv:2406.08307 (To appear in IEEE Journal of Selected Topics in Signal Processing). June 2025
- Slavakis, K., & Banerjee, S. (2019). Robust hierarchical-optimization RLS against sparse outliers. IEEE Signal Processing Letters, 27, 171-175. Dec 2019

## POSTER PRESENTATIONS

- Bellairs Workshop on Machine Learning and Statistical Signal Processing for Data on Graphs. Jan 2023
- DIMACS Workshop on Modeling Randomness in Neural Network Training: Mathematical, Statistical, and Numerical Guarantees. June 2024

## ACADEMIC PROJECTS

### **Understanding tensor decomposition for spectral unmixing in hyperspectral images (Matlab)**

Sept 2021 – Sept 2022

- Applied spectral analysis techniques to process and analyze high-dimensional hyperspectral image data.
- Implemented efficient spectral unmixing through low-rank tensor decomposition to extract endmembers and generate abundance maps with 15% improvement over traditional methods, when applied to remote sensing and environmental monitoring applications.

### **Musical Instrument Recognition using harmonics**

Oct 2017 – Dec 2017

- Extracted harmonic profiles of flute and piano recordings using cepstral analysis, identifying over 10 key distinguishing features in timbre and frequency response with 95% accuracy, which enabled enhanced instrument classification.

### **Classification of cancer subgroups using microarray gene expression data**

Sept 2016 – Mar 2017

- Implemented particle swarm optimization and adaptive K-nearest neighborhood technique on lung cancer data to classify cancer subgroups.
- Extracted 14 genes that can be efficiently exploited for high-accuracy diagnostic prediction through the t-test method for dimensionality reduction.