Raghu Ram Sattanapalle

J 347-873-2177 | **☑** raghurams95@gmail.com | **ऻ** raghuramsattanapalle | **۞** RaghuRamSatt | **⊕** raghurams.com

EDUCATION

Northeastern University

Boston, MA

Master of Science in Data Science (GPA: 3.867 / 4.00)

Jan. 2023 - May 2025

New York University

New York, NY

Master of Science in Mechanical Engineering

Sep. 2016 - May 2018

EXPERIENCE

ML Engineer

June 2025 - Present

Grant X Remote

- Implemented reasoning-based document segmentation using PageIndex hierarchical tree structures to extract 50+ structured fields from funding documents without traditional chunking or vector databases.
- Architected hybrid search engine combining semantic search (ELSER), vector embeddings, and BM25 retrieval with LLM-based query generation using Gemini models, achieving 90.3% MRR and 75.6% precision@10.
- Built data ingestion pipeline processing IRS Form 990-PF fillings from 756,000+ private foundations, parsing XML tax returns to extract organizational metadata, grant histories, and financials with cursor-based pagination and retry logic.
- Deployed production system on GCP/Kubernetes serving 25K+ federal and private opportunities with 91ms p50 latency, 99.9% uptime; optimized enrichment pipeline by 4-6x through batch database fetching.

Engineering Data Scientist (Co-op)

Jan. 2024 – June 2024

Veeco Instruments

San Jose, CA

- Unified 10+ years of manufacturing data from multiple storage drives using Python and SQL pipelines to build a robust dataset for Machine Learning driven wafer analysis.
- Developed convolutional neural networks (CNNs) using TensorFlow and PyTorch to predict boron wafer resistance, achieving a 6% average error rate by employing data augmentation, group normalization, and various CNN architectures.
- Built a standalone wafer visualization tool (Python) to identify trends and anomalies, improving manufacturing efficiency by 30% through data-driven process control.
- Collaborated with senior engineers on **code reviews** (GitHub) and **refactoring**, ensuring robust and **maintainable ML pipelines** aligned with software engineering best practices.

PhD Candidate / Researcher

Sept. 2018 - Aug. 2022

NYU Dynamical Systems Laboratory

Brooklyn, NY

- Analyzed the MIMIC dataset (300M+ clinical observations) using SQL for data extraction and supervised ML models for analysis, achieving 90% accuracy in predicting ICU mortality in collaboration with NYU Langone clinicians.
- Led a causal inference study on mass shootings, media coverage, and firearm acquisition using time series analysis (ARIMA, Tramo/Seats) and transfer entropy, resulting in a *Nature Human Behaviour* publication.
- Developed a mathematical model using **stochastic differential equations** in **MATLAB** and **Mathematica** to examine collective behavior, contributing to a publication in *Flow* (Cambridge Core).
- Secured a **\$2.1M NSF grant** as co-author to investigate the U.S. firearm ecosystem, utilizing **Tableau** for visually compelling preliminary results included in the proposal.

TECHNICAL SKILLS

Programming Languages: Python (Pandas, NumPy), SQL, C++, Java, JavaScript, Scala, R, MATLAB

Machine Learning: TensorFlow, PyTorch, Deep Learning, NLP, RAG, LangChain, Text Embeddings, Ranking/Retrieval Models, XGBoost, Anomaly Detection, Fraud Detection Models, Model Deployment, Model inference optimization, A/B Testing

Data Engineering & Cloud: AWS (EMR, S3, EC2, SageMaker), GCP, Azure, Hadoop, Spark, BigQuery, Hive

Databases: MySQL, PostgreSQL, MongoDB (NoSQL), Snowflake, Elasticsearch, Supabase, Vector DBs (Pinecone, FAISS, Qdrant)

Visualization: Tableau, Power BI, Matplotlib, Seaborn, Plotly, ggplot2, D3.js

Tools: Git, Docker, Kubernetes, Linux, CI/CD, Bash, Postman

Projects

FraudFusion: Synthetic Fraud Data Generation | Python, PyTorch, XGBoost, Diffusion Models Jan. 2025 - Apr. 2025

- Developed diffusion models to generate synthetic credit card fraud data, addressing extreme data imbalance (0.5% fraud rate) to improve anomaly detection and fraud identification capability.
- Implemented specialized feature engineering techniques and custom loss functions to capture complex fraud patterns, iteratively improving through multiple model versions to achieve high-quality synthetic data generation.
- Improved fraud detection performance of **XGBoost classifiers** by increasing detection rate from 82% to ≈90%, with minimal impact on false positive rates.

Scalable Music Similarity Analysis with Spark | Spark, Scala, AWS, Docker

Oct. 2024 - Dec. 2024

- Developed a music similarity system using **Spark** on the **Million Song Dataset** (**10K songs, 110K users**), integrating **K-Means** for audio feature analysis with **collaborative filtering** based on user listening history for song recommendations.
- Engineered two K-Means parallelization strategies in Spark to improve efficiency, achieving a 4.58x speedup on AWS EMR, and implemented H-V partitioning with a sparse matrix for efficient similarity computation.
- Evaluated K-Means using Silhouette Score and Davies-Bouldin Index and applied user-based rating normalization to enhance collaborative filtering, focusing on model accuracy and performance.