# Saikiranmansa Sunnam

## Machine Learning Engineer

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#### **≅ SUMMARY**

Machine Learning Engineer with expertise in NLP, computer vision, and anomaly detection, leveraging PyTorch, TensorFlow, and Scikit-learn to build scalable ML pipelines and deploy APIs via FastAPI/Docker. Proven success in fraud detection (Isolation Forests), Transformer fine-tuning, and RAG-based QA systems, with strong MLOps and cloud (AWS) experience. Currently pursuing an M.Sc. in Computer Science at the University at Buffalo, blending academic rigor with hands-on ML deployment.

## **➡** PROFESSIONAL EXPERIENCE

#### Machine Learning Intern, Rivach LLP

05/2022 - 03/2023 | India

- Designed an end-to-end machine learning pipelines across multiple domains such as computer vision, regression, and anomaly detection, utilizing TensorFlow, PyTorch, and Scikit-learn to deploy models with 98% accuracy and R<sup>2</sup> scores > 0.95 for regression.
- Constructed robust data preprocessing systems for 20,000+ labeled images and 100,000+ structured records, integrating SMOTE, label encoding, and feature scaling to reduce model training errors by 30%+ and eliminate data skew.
- Engineered scalable CNN-based deep learning architectures for image classification, improving predictive accuracy by 22%, while reducing model inference time by 18% using architectural pruning techniques.
- Implemented ensemble learning models, including Random Forest and Gradient Boosting, resulting in a 15% increase in F1-score and 12% gain in recall, significantly outperforming traditional classifiers.
- Tuned model hyperparameters through cross-validation and grid search strategies (e.g., n\_estimators, max\_depth, learning rate), reducing false positive rates by 25%+ and improving model variance by 20%.
- Built and defined unsupervised anomaly detection pipelines using Isolation Forest, One-Class SVM, and Local Outlier Factor, enhancing fraud detection precision by 19% on highly imbalanced datasets.
- Conducted deep regression modeling on public health datasets using PyTorch, optimizing for MSE = 0.034 and achieving an explained variance of 0.96, resulting in accurate life expectancy predictions.

## TECHNICAL PROJECTS

#### Domain-Specific QA System Using DeepSeek and RAG

- Built a production-grade question-answering system leveraging DeepSeek and Retrieval-Augmented Generation (RAG), combining FAISS-based dense vector retrieval and a language model to deliver 30%+ improvement in response accuracy.
- Processed and indexed over 1 million domain-specific documents using FAISS and Elasticsearch, enabling efficient, low-latency (<250ms) query retrieval at scale.</li>
- Enhanced semantic search quality by integrating Sentence-BERT embeddings, boosting BLEU and ROUGE scores by 15–20%, and improving relevance in context-sensitive queries.
- Benchmarked QA performance using BLEU, ROUGE, and Exact Match, achieving 30%+ lift over baseline retrieval and generation models, validating end-to-end model effectiveness.
- Launched a scalable QA pipeline on AWS using FastAPI and Docker, reducing document retrieval time by 50%.

#### Advanced Anomaly Detection and Text Classification Using Deep Learning

- Engineered an anomaly detection framework using three autoencoder variants to detect patterns in a time-series dataset of 5,315 records, achieving a maximum R<sup>2</sup> of 0.9916 and detecting 25–74 anomalies depending on the model.
- Developed a Transformer-based text classifier using PyTorch and fine-tuned it on the AG News dataset, increasing classification accuracy from 90.08% to 90.53% via L2 regularization and dropout.
- Preprocessed and visualized 10,000+ tokens, leveraging tokenization, normalization, and data visualization with Matplotlib to identify trends and feature importance for classification.
- Optimized training using advanced regularization techniques, fine-tuning dropout rates and learning rates, which resulted in a 0.45% improvement in accuracy and 12% reduction in overfitting.
- Conducted comprehensive model evaluation using R<sup>2</sup>, precision, recall, F1-score, confusion matrix, and improved ROC AUC from 0.82 to 0.91, validating model performance and generalization.

#### **ℰ** EDUCATION

State University of New York at Buffalo, Master of Science in Computer Science

12/2024 | Buffalo, NY

**Courses:** Machine learning, Deep learning, Computer Vision & Image Processing, Operating Systems, Algorithms Analysis and Design, Data Intensive Computing, Computer Security, Data Mining and Query Language, Software Engineering

## **P** SKILLS

Core ML & AI: Generative AI, LLMs (LLaMA, GPT, BERT), RAG, NLP, Computer Vision, Anomaly Detection

Frameworks & Libraries: PyTorch, TensorFlow, Scikit-learn, Hugging Face, Keras

Programming: Python, SQL, R, Java, C

Data & Visualization: Pandas, NumPy, Matplotlib, Seaborn, Plotly

DevOps & Cloud: Docker, FastAPI, REST APIs, AWS (EC2, S3, Lambda, SageMaker, DynamoDB, CloudWatch), CI/CD

Big Data & Databases: Hadoop, Spark, MySQL, Oracle Database

Tools & Collaboration: Git, Jupyter Notebooks, Data Warehousing, Testing, Code Reviews, Documentation

#### PUBLICATIONS

Published in F1000Research: Link ☑