

# VARUN KUMAR KOTHAPALLI

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## PROFESSIONAL SUMMARY

- AI / Machine Learning Engineer with 5+ years of experience designing, building, and operating production-ready machine learning systems across enterprise environments.
- Strong background in applied machine learning, including supervised and unsupervised modeling, feature engineering, model training, evaluation, and performance optimization using Python and industry-standard ML frameworks.
- Proven experience delivering Generative AI and NLP solutions leveraging Large Language Models (LLMs), Retrieval-Augmented Generation (RAG), and transformer-based architectures for document intelligence.
- Hands-on expertise in deploying, monitoring, and scaling ML models using MLOps practices, cloud platforms (AWS, Azure), containerization (Docker, Kubernetes), and RESTful APIs.
- Adept at collaborating with cross-functional teams to translate business requirements into reliable, scalable AI solutions with measurable operational impact.

## TECHNICAL SKILLS

**Artificial Intelligence & Machine Learning:** Artificial Intelligence (AI), Machine Learning (ML), Deep Learning, Supervised & Unsupervised Learning, Feature Engineering, Model Training & Evaluation, Hyperparameter Optimization, Model Explainability, Bias Mitigation

**Generative AI & Large Language Models:** Generative AI, Large Language Models (LLMs), Transformer Architectures, Prompt Engineering, Fine-Tuning (LoRA, PEFT), Retrieval-Augmented Generation (RAG), Embeddings, Vector Search

**Natural Language Processing (NLP):** Text Classification, Named Entity Recognition (NER), Sentiment Analysis, Text Summarization, Text Generation, Semantic Search, Tokenization

**Programming & ML Frameworks:** Python, SQL, NumPy, Pandas, Scikit-learn, PyTorch, TensorFlow

**MLOps, Deployment & Model Serving:** MLOps, MLflow, Model Versioning, Model Deployment, Model Monitoring, Data Drift Detection, RESTful APIs, FastAPI, Flask, JSON APIs

**Cloud Platforms & Containerization:** AWS (SageMaker, EC2, S3, RDS), Azure Machine Learning, Docker, Kubernetes

**Data Engineering for Machine Learning:** ETL Pipelines, Feature Pipelines, Apache Spark, Data Warehousing

**Databases:** PostgreSQL, MySQL, Oracle, SQL Server

## PROFESSIONAL EXPERIENCE

### AI / Machine Learning Engineer

Mar 2025 - Present

#### Equifax | Saint Louis, Missouri

- Designed Python- and SQL-based machine learning pipelines with Scikit-learn that streamlined feature preparation for large consumer credit datasets, cutting recurring model build cycles by around 30%.
- Strengthened credit risk and decisioning accuracy by training supervised and unsupervised ML models, producing 18-22% gains in accuracy and recall through disciplined cross-validation and tuning.
- Delivered real-time scoring capability by serving trained models through REST APIs built with FastAPI and Flask, enabling consistent prediction access across internal Equifax platforms.
- Stabilized production performance by introducing evaluation, monitoring, and drift-detection workflows, which reduced unexpected model degradation incidents by approximately 25%.
- Converted regulatory and analytics requirements into deployable AI decision logic, allowing model outputs to directly support KPIs tied to accuracy, compliance, and operational efficiency.
- Reduced manual document review effort by applying Generative AI and transformer-based NLP models for summarization and content extraction, achieving nearly 35% efficiency improvement.
- Improved release reliability by containerizing ML services with Docker and running inference workloads on AWS EC2, lowering environment-related deployment failures by about 40%.
- Sustained continuous model improvement through iterative development and review cycles supported by Azure Machine Learning experiment tracking and monitoring, enabling faster feedback integration and steady optimization of production ML systems.

### Database Developer - Applied Machine Learning Systems

Sep 2021 - Jul 2023

#### MedPlus Health Systems | Chennai, India

- Designed SQL-based ETL pipelines to consolidate pharmacy, prescription, and sales data, increasing the availability of ML-ready datasets for analytics and modeling by approximately 35%.
- Converted high-volume transactional records into stable ML feature tables through SQL-driven feature engineering, enabling consistent and reusable inputs for forecasting and analytical models.
- Reduced training data errors by around 30% by embedding validation rules and anomaly detection logic directly into ingestion and preprocessing workflows.
- Improved model experimentation speed by optimizing database schemas, indexing, and partitioning strategies, cutting feature retrieval latency by nearly 40%.
- Shortened data refresh cycles supporting model training by tuning SQL execution plans, allowing analytics and ML teams to iterate on models more frequently.
- Provided standardized dataset access by exposing curated healthcare data through JSON-based REST APIs, enabling seamless consumption by analytics and machine learning services.

- Protected sensitive healthcare and transactional information by enforcing data integrity checks and access controls, maintaining compliance across ML data pipelines.
- Aligned database design with evolving ML lifecycle and MLOps needs through close collaboration with data scientists and ML engineers, supporting scalable feature pipelines.

## Software Engineer - Data Platforms

Mar 2019 - Aug 2021

### DXC Technologies India | India

- High-volume analytics workloads processed more efficiently once SQL Server queries, indexes, and stored procedures were optimized, increasing overall data throughput by around 30%.
- Reliability of datasets consumed by analytics and ML teams improved after introducing SQL-based preprocessing and aggregation logic, which reduced downstream data discrepancies by approximately 25%.
- Faster access to dashboards and model-preparation datasets followed the restructuring of database schemas and execution plans, cutting query response times by nearly 40%.
- Scalable compute and storage for enterprise reporting systems was enabled by supporting AWS EC2 and S3 deployments, improving availability of cloud-hosted analytics applications.
- Data platforms were better positioned for future AI and ML initiatives through active participation in architecture design discussions, ensuring warehouse structures aligned with evolving analytical needs.
- Unplanned outages became less frequent after performing root-cause analysis on data pipeline and reporting failures, reducing repeat incidents by about 20%.
- Clearer and more accurate business insights were delivered by converting client requirements into well-structured SQL data models, strengthening the quality of operational and performance reports.
- Knowledge continuity across delivery teams improved through detailed documentation of data flows, schemas, and processing logic, shortening onboarding time and reducing reliance on individual contributors.

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## PROJECTS

### GENERATIVE AI-POWERED DOCUMENT INTELLIGENCE SYSTEM

- Implemented Generative AI and Large Language Models (LLMs) to automate document understanding and summarization, enabling accurate extraction of contextual insights from unstructured enterprise documents.
- Response relevance and factual grounding improved by integrating Retrieval-Augmented Generation (RAG) with embedding models and vector search, ensuring LLM outputs referenced indexed enterprise data.
- Inference reliability at scale was achieved by serving models through FastAPI, containerizing with Docker, deploying on AWS and registering models via MLflow with SageMaker-compatible artifacts, enabling reproducible and versioned deployments.

### END-TO-END MACHINE LEARNING PIPELINE FOR PREDICTIVE ANALYTICS

- Structured datasets were prepared for modeling by orchestrating data ingestion, feature engineering, and preprocessing pipelines using Python, SQL, and Apache Spark, supporting repeatable ML experimentation.
- Predictive performance was improved through training supervised machine learning models with Scikit-learn and PyTorch, applying cross-validation and hyperparameter optimization techniques.
- Production stability and transparency were maintained by deploying models with Docker and Kubernetes, while implementing model evaluation, explainability, and monitoring aligned with MLOps practices.

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## EDUCATION

### Master of Science in Information Technology

Aug 2023 - Dec 2024

Webster University | Missouri, USA

### Bachelor of Science in Computer Science

Jun 2015 - May 2018

Loyola Academy Degree & PG College | Hyderabad, India

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## CERTIFICATIONS

- Machine Learning Specialization - [Coursera](#)
- Deep Learning Specialization - [Coursera](#)
- Generative AI with Large Language Models - [Coursera](#)
- Applied Machine Learning with Python - [Coursera](#)
- MLOps Fundamentals - [DataCamp](#)