VINEET SRIVASTAVA

Data Analysis & Cleaning ► Machine Learning ► Big Data Analytics ► IOT & Wireless Communication

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Professional Experience

• Qualcomm Bengaluru, India- Senior Engineer (11/2021- 07/2022)

Technologies/Tech-Stack: Bluetooth Low Energy(BLE) and LE Audio (Basic Audio Profile-BAP, Common Audio Profile- CAP, GATT, Enhanced GATT Caching, BLE Connections), Python, Embedded C, Wireless Data Packets Analysis, LE Audio Profiles Automation Framework Development, Machine Learning & Predictive modelling

• Capgemini Pune, India- Consultant (Client: Dexcom San Diego, USA) (02/2021-10/2021)

Technologies/Tech-Stack: Continuous Glucose Monitoring System (CGMS) using BLE, IOT Healthcare Analytics, Python, EDA & Feature Engineering, BLE packets debugging & Automation.

• Mirafra Hyderabad, India - Senior Software Engineer (10/2019 - 02/2021)

Technologies/Tech-Stack: Smart Building Automation, Zigbee, Data Analysis and Machine Learning, Python, Data pipelining and Cloud Deployments, Object Tracking using Computer Vision

• <u>L&T Technology Services, India</u>- Engineer (09/2016- 02/2019)

Technologies/Tech-Stack: Indoor Localization & People Tracking using BLE, Predictive Machine learning models- such as KNN, Linear Regression, BLE RSSI Filtering- Kalman Filter, Moving Average, Published Patent on designing a fine-tuned Trilateration Algorithm for indoor localization, Python, Embedded C

Industrial Projects

a. Title: Deriving Critical Metrices & Create Predictive Model to identify Bluetooth Connection Failures

Qualcomm India, Senior Engineer, Project-Tenure: (02/2022- 06/2022)

Problem Statement:

- The production line team was facing regular **BLE connection failures** between wearable devices like **Fitbit and BLE receiver(Phone)** thereby leading to loss of real time sensor data (BLE packets). The team wanted to identify the important metrices and understand the pattern behind connection loss.
- The task was to create a pipeline to analyze the metrices from sniffed BLE packets and perform quick analytics. A predictive
 classification-based machine learning model was to be designed to predict the connection failures based on identified critical
 parameters during packet exchange between BLE Transmitter (Wearable device) and Receiver (Phone).

Tech-Stack/Tools: Bluetooth Low Energy (BLE), BLE packets debugging, Python, PySpark, AWS-S3, Glue, Glue Studio, Athena SQL Queries, QuickSight, ML Models-Logistic Regression, Decision Tree, Random Forest, AdaBoost, XgBoost.

Roles & Contribution:

- Data Pipelining & Analytics
- ✓ Ingested sniffed CSV files into S3 Raw data bucket as part of the data lake architecture. Data was catalogued using AWS Glue.
- ✓ Created an AWS Glue job to read the raw CSV files from S3-raw bucket, convert to parquet format and ingest in S3-Clean Bucket.
- ✓ Created another glue job to **read the parquet files from s3-Clean Bucket and perform exploratory data analysis and feature engineering using Python** to derive **critical metrices behind connection failures** and store the same in data lake **(S3-Analytics bucket)**and create a database for the same for **Athena queries**.
- ✓ Used AWS Athena Services to query the database of BLE metrices created as part of AWS Glue jobs to be used by AWS QuickSight dashboard for analytics and visualization.
- Machine Learning (ML) Pipeline
- ✓ Developed a **classification-based ML** algorithm by reading the data from AWS-S3 Analytics bucket created as part of AWS glue jobs to predict the connection failure/signal loss based on critical parameters identified such as BLE time to connect, Time in connection, Time to disconnect, Encryption validation, ACK of sensor-data.
- ✓ The **Adaboost performed better** when compared to other models with **accuracy of 66% and recall of 72%.** The trained model along with other artifacts was stored in **AWS-S3 artifacts model**.

Results: The entire pipeline, analytics on Dashboard along with trained model, helped the production line team quickly to identify the faulty wearable devices or the ones performing poor due to frequent connection failures

b. Title: <u>Predicting Lifecycle of IOT Sensor by designing an end-to-end ML Pipeline (Building Automation)</u>

Mirafra Technologies India, Client: Honeywell, Senior Software Engineer, Project-Tenure: (04/2020- 01/2021)

Problem Statement:

- As part of the smart building automation, Zigbee-based end-devices such as THL (temperature, humidity, and light) sensors are
 used to automatically adjust room temperature or light based on their readings. These sensors were, however, subjected to battery
 drain, and hence their lifecycle was compromised.
- The team was endowed with the responsibility of **creating an end-to-end ML pipeline** with cloud deployment to predict the sensor's lifecycle (**whether endangered or working normally**), thereby **reducing the cost of frequent setup and installation**.

Tech-Stack/Tools: ZigBee, Python, Confluent-Kafka, NoSQL-MongoDB, Docker, EDA & Feature Engineering, Machine Learning-Classification based Algorithms, MQTT, GitHub Actions, AWS- EC2, ECR, S3, CI/CD Pipeline, Deployments, Kanban, Jira, VS-code

Roles & Contribution:

- Data Pipelining & Engineering
- ✓ Developed a **python-based Gateway Application** running on Raspberry-Pi to receive the readings from THL (Temperature, Humidity & Lux) Sensor along with Target Feature (Sensor-Life Endangered:1, Working Normally: 0), serialize & preprocess the data and send to **Confluent-Kafka Topic** and **MQTT server for actions.**
- ✓ Developed a Consumer application on a Linux Machine to consume the available data from **Kafka-Topic** and send the same to **MongoDB NoSQL** database cluster to be used for analytics & predictive modelling.
- Machine Learning (ML) Pipeline
- ✓ Developed and actively supported by team to create ML components by consuming data from MongoDB- Data Ingestion, Data Validation, Data Transformation, Model Creation, Model Evaluation and Model Pusher.
- ✓ Developed Classification based ML algorithms like **Decision Tree**, **Random Forest**, **AdaBoost & Logistic Regression** with EDA, and Feature Engineering. **Handled data imbalance** using techniques such as **SMOTE and Oversampling**.
- ✓ Discussed with **Technical Architects and Product owners** to counter issues of **Data Leakage and Data Drift** during data validation phase.
- ✓ Evaluated the model using different metrices such **Precision, ROC-Curve, Recall & F-Score** and selected the best model.
- ✓ **The Random Forest model** gave better accuracy and precision, **close to 80**% and helped in proper identification of faulty sensors. The hardware team was quick to fix those sensors in case, anything was wrong with their controllers.
- CI/CD Pipelining for DOCKER image deployment on AWS EC2(self-hosted runner) using GitHub Actions

Results: The entire pipeline developed helped to reduce the burden of frequent setup. The installation costs went down by almost 30%.

c. Title: <u>Construction and Design of Synthetic Syndromic Surveillance Cloud Architecture-Azure</u> *UIC School of Public Health*, Graduate Research Assistant, Project-Tenure: (10/2022- Present)

Tech-Stack/Tools: Python, PySpark, Azure Synapse Analytics, ETL pipelining, Microsoft SQL, Data Cleaning and Feature Engineering, Machine Learning

- Analyzed the Illinois in/out-patient data acquired by COMPdata informatics and statistically matched the patient records with household population records from the RTI US Synthetic Household population database to generate synthetic population household data.
- Analyzed the cumulative data (in Tbs) in Azure Synapse Analytics using Python (Azure PySpark). The data generated was stored into
 Azure Data Lake and after filtering with ICD-10 codes was ingested into an Azure SQL Server.

Skills & Technologies

- Programming: Python, R, C#, Embedded C, OOPs using Python, Linux System Programming, Multithreading
- Data Structures: Algorithm Complexity, Arrays/Lists, Strings, Linked-Lists, Trees, Graphs, Stacks & Queues
- Exploratory Data Analysis: Pandas, NumPy, Seaborn, Matplotlib, Data Cleaning, Feature Engineering, familiar working & performing analytics on different file formats- CSV, Parquet, Excel, JSON, YAML, Config.ini, etc.
- Statistics & Machine Learning: Hypothesis & Chi-Square Testing for feature selection, Central Limit Theorem, ANOVA, T-Test, Linear & Logistic Regression, Gradient Descent, Regularization- Ridge and Lasso, Cross Validation, Hyperparameter Tuning, Decision Tree, Ensemble Techniques- Bagging & Boosting (Random Forest, AdaBoost, XgBoost), KNN, K-Means, Hierarchical Clustering, DBSCAN, Naive Bayes, Data Drift and Data Leakage concepts.
- Natural Language Processing & Deep Learning: NLTK, Spacy, Tensor-Flow, Gensim, Text Processing- Tokenization, Stopwords,
 Stemming/Lemmatization, Bag-of-words, TF-IDF, Word Embedding- word2vec, fasttext, RNN, LSTM, NER, Artificial Neural network (ANN), CNN-Image Classification, Computer Vision- Objection detection & Tracking, OpenCV, NLP on Clinical Trial Datasets to develop keyword search.
- Big Data Analytics & Cloud: SQL with Azure, Oracle/MySQL, PySpark, AWS EC2, IAM, S3, Lambda, AWS Glue, Athena, Azure Synapse Analytics, SQL Pool, Kafka, Airflow, NoSQL- MongoDB, ETL pipelining, CI/CD pipelining using GitHub Actions
- Visualization Dashboards- Power-BI, Azure ArcGIS for geospatial mapping, AWS QuickSight, Plotly, Dash
- Tools & Process: Appium for Mobile App Automation, Docker, GitHub, Web-Scrapping (REST-APIs, Beautiful-Soup), Android (ADB) Debugging, Agile
 Process, Kanban, Testing Frameworks- PyTest and Unittest, HIPAA compliant.

Research/Patient

- Working on the healthcare-research as part of the <u>synthetic syndromic surveillance project</u>, presented our research work poster at 2022 <u>Institute for Public Health and Medicine Population Health Forum</u> at Northwestern University, nominated as a <u>finalist</u> for the <u>Rowland "Bing" Chang excellence in Research Award.</u>
- US Patent: <u>Detecting Elevator Mechanics inside Elevator Systems</u>.- RSI data analysis & creation of fine-tuned Trilateration Algorithm.

Education

- Master's in Business Analytics, University of Illinois at Chiago- USA, CGPA: 3.78/4.0
- Mary 2012 Mary 2016
- **B. Tech in Electronics & Communication**, VIT Chennai- India, CGPA: 8.16/10

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