

Microsoft Machine Learning Projects

Project 1: Customer Churn Prediction & Automated Retraining System

Project Overview

Develop a production-grade customer churn prediction system that:

- Ingests streaming customer behavior data
- Builds and serves a churn prediction model
- Implements automated retraining based on data drift
- Provides dashboards and alerts for business stakeholders

Milestone 1: Data Collection, Preprocessing, and Exploration

Objectives

- Gather and prepare churn data from various sources (internal CRM, logs, billing)
- Engineer features and understand data quality and imbalance

Tasks

1. Data Collection

- a. Collect historical and streaming customer transactional and usage data from SQL databases, logs, or Azure Blob Storage
- b. Ensure comprehensive labels for churn vs non-churn outcomes

2. Data Preprocessing

- a. Clean and normalize features (scaling, encoding, null handling)
- b. Address class imbalance (SMOTE or weighting)

3. Exploratory Data Analysis

- a. Visualize churn correlates (tenure, usage patterns)
- b. Investigate imbalance and bias issues

Deliverables

- Cleaned and Processed Dataset
- Preprocessing Pipeline Documentation
- EDA Report with insights and issues

Milestone 2: Model Development & Evaluation

Objectives

- Build and evaluate churn prediction models with production focus

Tasks

1. Model Building

- a. Implement models such as Gradient Boosting, XGBoost, Random Forest
- b. Perform feature selection

2. Model Evaluation

- a. Use confusion matrix, ROC-AUC, precision-recall curves
- b. Validate with cross-validation

3. Optimization

- a. Hyperparameter tuning (Grid/Random/Bayesian)
- b. Consider threshold tuning for business impact

Deliverables

- Trained and Optimized Churn Prediction Model
- Model Evaluation Report with business-relevant metrics

Milestone 3: Advanced Techniques & Cloud Deployment

Objectives

- Deploy the churn model as a scalable prediction service on Azure

Tasks

1. Transfer Learning/Ensembling

- a. Implement ensemble techniques for improved performance

2. Azure Deployment

- a. Deploy via **Azure Machine Learning** as a REST endpoint
- b. Use Docker containers for standardized serving

3. API Integration

- a. Build REST API to receive customer data and return churn risk scores

Deliverables

- Deployed Azure ML Churn Predictor
- Integrated API for real-time scoring

Milestone 4: MLOps, Monitoring & Web Dashboard

Objectives

- Implement MLOps best practices for lifecycle management and business integration

Tasks

1. Experiment Tracking

- a. Use MLflow or Azure ML tracking for models and metrics

2. Web Dashboard

- a. Create a dashboard (Power BI or Streamlit) showing churn predictions & business KPIs

3. Model Monitoring

- a. Set drift detection triggers, performance alerts

4. Automated Retraining Strategy

- a. Schedule pipelines to retrain the model on fresh data based on drift metrics

Deliverables

- Monitoring & Retraining Pipeline
- MLOps Documentation
- Interactive Dashboard

Milestone 5: Final Documentation & Presentation

Objectives

- Summarize the project and demonstrate full functionality

Tasks

1. **Final Report**
 - a. Explain architecture, challenges, business impact
2. **Presentation**
 - a. Demo real-time scoring and dashboard insights
3. **Future Roadmap**
 - a. Suggest improvements like A/B testing or cross-region deployment

Deliverables

- Final Written Report
- Polished Final Presentation
- Future Enhancements List

Project 2: Real-Time Anomaly Detection for IoT Sensor Streams

Project Overview

Build a production anomaly detection system for streaming IoT sensor data (e.g., manufacturing or smart infrastructure). The system should:

- Detect anomalies in real time
- Scale in production
- Trigger alerts and dashboard updates

Milestone 1: Data Collection, Preprocessing, and Exploration

Objectives

- Collect multi-source time-series streams, clean and prepare

Tasks

1. Data Collection

- a. Ingest IoT streams via Azure Event Hubs or Kafka
- b. Store raw data in Azure Blob / Data Lake

2. Preprocessing

- a. Windowing, smoothing, feature extraction
- b. Handle missing streams and noise

3. EDA

- a. Statistical patterns and correlation analysis
- b. Identify anomaly patterns

Deliverables

- Processed Time-Series Dataset
- Preprocessing Pipeline Doc
- Anomaly Patterns EDA Report

Milestone 2: Model Development & Evaluation

Objectives

- Train models that detect anomalies

Tasks

1. Modeling

- a. Evaluate Autoencoders, Isolation Forests, LSTM-based approaches

2. Evaluation

- a. Precision, recall on anomaly detection
- b. Real-time detection simulation

3. Optimization

- a. Hyperparameter search

Deliverables

- Trained Anomaly Detection Models
- Evaluation Report

Milestone 3: Cloud Deployment & Integration

Objectives

- Deploy pipeline to production

Tasks

1. Azure Deployment

- a. Use Azure Stream Analytics + Functions for inference
- b. Containerize model

2. Scalable API

- a. Expose REST service for incoming sensor data

3. Integration

- a. Alerts via Azure Monitor, Logic Apps

Deliverables

- Azure ML Real-Time Anomaly Service
- API & Alert System

Milestone 4: MLOps & Monitoring Dashboard

Objectives

- Enable ongoing monitoring and automated reactions

Tasks

1. Tracking & Versioning

- a. Track experiments and retrains with MLflow
- 2. Live Dashboard**
 - a. Power BI or Grafana real-time
- 3. Retraining Automation**
 - a. Trigger retrain based on drift metrics

Deliverables

- Monitoring Dashboard
- MLOps Pipeline Doc

Milestone 5: Final Documentation & Presentation

Objectives

- Complete documentation and demonstration

Deliverables

- Full Final Report
- End-to-End Demo Presentation
- Future Enhancements Proposal

Project 3: Intelligent Support Ticket Classification with RAG

Project Overview

Build a **production-ready support system** that:

- Ingests support tickets
- Uses NLP + vector search + transformer models
- Generates accurate, context-aware automatic responses

This reflects real roles where ML Engineers integrate NLP with production systems and **deliver measurable impact on operations.**

Milestone 1: Data Collection & Preprocessing

Data Collection: Support tickets from databases

Preprocessing: Tokenization, cleaning, embeddings

EDA: Text length, class frequencies

Deliverables: Cleaned corpus, preprocessing doc, EDA report

Milestone 2: Model Development

- Train vector search + transformer models
- Evaluate using metrics like accuracy, recall, BLEU
- Compare traditional vs RAG

Deliverables: Models & Evaluation Report

Milestone 3: Azure Deployment

- Deploy via Azure ML as API
- Integrate vector DB (Azure Cognitive Search)
- Backend service

Deliverables: Deployed API & Integration Doc

Milestone 4: MLOps & Monitoring

- Track experiments (MLflow)
- Dashboard for KPIs
- Retraining triggers

Deliverables: Monitoring Setup, Pipelines Doc

Milestone 5: Final Documentation & Presentation

Deliverables: Final Report, Demo, Future Enhancements

Project 4: Real-Time Anomaly Detection & Alerting System (IoT / Time-Series)

Project Overview

Develop a cloud-based anomaly detection system for streaming IoT sensor data. The system must detect anomalies in real time, trigger alerts, and provide analytics dashboards.

Milestone 1: Data Collection, Preprocessing, and Exploration

Objectives

- Collect real or simulated time-series IoT data
- Preprocess and explore seasonal patterns and noise

Tasks

1. Data Collection:

- a. Ingest data from multiple sensors (temperature, pressure, etc.) via simulated streams or public IoT datasets.
- b. Store raw data in Azure Data Lake or Blob Storage.

2. Data Preprocessing:

- a. Handle missing values, smoothing, and window segmentation.
- b. Create engineered features (lag features, rolling means).

3. Exploratory Data Analysis (EDA):

- a. Analyze trends, seasonal effects, and noise patterns.
- b. Identify baseline normal behavior and potential anomalies.

Deliverables

- Cleaned and Structured Time-Series Dataset
- Data Preprocessing Pipeline Documentation
- EDA Report with Visualizations

Milestone 2: Model Development & Evaluation

Objectives

- Train reliable anomaly detection models suitable for streaming data

Tasks

1. Model Development:

- a. Implement models such as LSTM Autoencoders, Isolation Forests, or Prophet-based threshold detection.

2. Evaluation:

- a. Evaluate using precision, recall, F1-score for anomaly classes.
- b. Test on held-out test sets and synthetic anomalies.

3. Optimization:

- a. Adjust thresholds and retrain pipelines for robustness.

Deliverables

- Trained Anomaly Detection Models
- Model Evaluation Report

Milestone 3: Advanced Techniques & Cloud Integration

Objectives

- Deploy the model in real time on Azure

Tasks

1. Cloud Deployment:

- a. Use **Azure Stream Analytics** and Azure Functions for real-time scoring.
- b. Containerize model inference with Docker deployed through Azure Container Instances.

2. REST API:

- a. Build a REST API for real-time anomaly scoring.

3. Integration:

- a. Integrate with Azure Event Hubs for stream ingestion.

Deliverables

- Deployed Real-Time Anomaly Detection Service
- Real-Time API Endpoints

Milestone 4: MLOps, Monitoring, & Dashboard

Objectives

- Implement production-grade monitoring, alerts, and retraining automation

Tasks

1. MLOps Tracking:

- a. Track model versions and experiments with MLflow or Azure ML.

2. Monitoring:

- a. Use Azure Monitor and custom metrics for drift and performance monitoring.

3. Dashboard:

- a. Build a dashboard (Power BI or Grafana) showing anomaly rates in real time.

4. Retraining Plan:

- a. Set up scheduled retraining pipelines triggered by drift detection.

Deliverables

- Monitoring & Alerting Setup
- Dashboard & KPI Visualization
- MLOps Pipeline Documentation

Milestone 5: Final Documentation & Presentation

Objectives

- Summarize and demonstrate end-to-end solution

Tasks

1. Final Report:

- a. Document pipeline, deployment, challenges, and business impact.

2. Presentation:

- a. Live demo of streaming anomaly detection with alerts and dashboards.

3. Future Enhancements:

- a. Propose edge deployment optimization or predictive maintenance features.

Deliverables

- Final Project Report
- Final Presentation Deck
- Future Improvements Document

Project 5: Customer Support RAG-Powered Intelligent Chatbot

Project Overview

Build a production-ready support automation system using **Retrieval-Augmented Generation (RAG)** and vector search. It should serve real queries and integrate with business backend systems.

Milestone 1: Data Collection & Preprocessing

Objectives

- Aggregate support ticket logs, FAQs, documentation, and knowledge base

Tasks

1. Data Ingestion:

- a. Collect historical support tickets, product manuals, and FAQ entries.

2. Preprocessing:

- a. Clean and unify text data, remove noise, and tokenize.

3. EDA:

- a. Analyze common queries, topic distributions, and response effectiveness.

Deliverables

- Processed Text Corpus
- Preprocessing Pipeline Doc
- Support Data EDA Report

Milestone 2: Model Development & Evaluation

Objectives

- Train and evaluate RAG retrieval + generation pipeline

Tasks

1. Model Building:

- a. Create vector store (e.g., Azure Cognitive Search or embeddings + vector DB).
- b. Configure RAG model using Hugging Face or similar.

2. Evaluation:

- a. Evaluate with metrics like BLEU, ROUGE, and relevance scoring.

3. Optimization:

- a. Fine-tune RAG components based on relevance and response quality.

Deliverables

- Trained RAG Pipeline
- Model Evaluation Report

Milestone 3: Advanced Techniques & Azure Deployment

Objectives

- Deploy intelligent chatbot as part of production support

Tasks

- 1. Azure Integration:**
 - a. Deploy the RAG service via Azure Machine Learning or Azure App Service.
- 2. API & Workflow Integration:**
 - a. Create REST API and integrate with support portal.
- 3. Security & Access:**
 - a. Secure endpoints with Azure AD or API keys.

Deliverables

- Deployed Chatbot Service
- Integration & Security Doc

Milestone 4: MLOps & Monitoring

Objectives

- Track performance and automate retraining

Tasks

- 1. Experiment Tracking:**
 - a. Use MLflow tracking for RAG variations.
- 2. Monitoring:**

- a. Monitor accuracy, latency, user satisfaction scores.

3. Retraining Mechanism:

- a. Schedule refresh of embeddings and model weights.

Deliverables

- Monitoring Dashboard
- Retraining Pipeline

Milestone 5: Final Documentation & Presentation

Deliverables

- Final Report
- Demo Presentation
- Business KPI Impact Analysis

Project 6: Demand Forecasting & Inventory Optimization Engine

Project Overview

Design a scalable system that predicts product demand using time-series and ML models, and builds inventory recommendations integrated with a dashboard.

Milestone 1: Data Collection & EDA

Deliverables

- Forecast Dataset
- Preprocessing & EDA Reports

Milestone 2: Forecast Model Development

Deliverables

- Trained Forecast Model
- Evaluation Report

Milestone 3: Azure Deployment + API

Deliverables

- Deployed Service
- API Endpoints

Milestone 4: MLOps + Dashboard

Deliverables

- Monitoring & Retraining Automation
- Dashboard

Milestone 5: Final Report & Presentation

Deliverables

- Documentation
- Demo

Project 7: Personalized Recommendation System with Feedback Loop

Project Overview

Build a scalable **personalized recommendation engine** for e-commerce or content platforms that:

- Generates tailored suggestions
- Adapts to user feedback in real time
- Runs on Azure with monitoring and retraining

Real production recommendation systems like those used by Booking.com or similar platforms are central to customer engagement and have measurable business impact.

Milestone 1: Data Collection, Preprocessing & Exploration

Objectives

- Gather user interaction and item metadata
- Prepare unified datasets for training and validation

Tasks

1. Data Collection

- a. Aggregate user ratings, purchase history, browsing logs
- b. Include item metadata (categories, tags, content)

2. Data Preprocessing

- a. Clean, normalize, and encode features
- b. Create user-item interaction matrices

3. EDA

- a. Analyze user behavior patterns
- b. Visualize sparsity and coverage of interactions

Deliverables

- Cleaned Recommendation Dataset

- Preprocessing Pipeline Doc
- EDA Report with insights

Milestone 2: Model Development & Evaluation

Objectives

- Develop recommendation models

Tasks

1. Model Development

- a. Content-based filtering, collaborative filtering, matrix factorization
- b. Optionally integrate neural models (e.g., embeddings)

2. Evaluation

- a. Use metrics: Precision@K, Recall@K, MAP, NDCG

3. Optimization

- a. Hyperparameter tuning and feature selection

Deliverables

- Trained Recommendation Models
- Model Evaluation Report

Milestone 3: Cloud Deployment & Integration

Objectives

- Deploy system as a scalable service

Tasks

1. Azure Deployment

- a. Use Azure Machine Learning to host the model
- b. Containerize model with Docker

2. REST API Service

- a. Build FastAPI/Flask API for live predictions

3. Integration

- a. Connect API with front-end or consumer app

Deliverables

- Deployed Recommendation API
- Integration Documentation

Milestone 4: MLOps & Monitoring

Objectives

- Enable continuous monitoring and retraining

Tasks

1. Experiment Tracking

- a. Track models and metrics with MLflow

2. Monitoring

- a. Monitor recommendation quality using real-time logs

3. Retraining Strategy

- a. Schedule pipeline to retrain based on usage trends

Deliverables

- MLOps Documentation
- Monitoring Setup
- Retraining Pipeline

Milestone 5: Final Documentation & Presentation

Objectives

- Comprehensive report and demo

Tasks

1. Final Report

- a. Document methods, results, business impact

2. Presentation

- a. Live demo of recommendations and feedback loop

3. Future Improvements

- a. Ideas like A/B testing or cross-platform integration

Deliverables

- Final Report
- Presentation Deck
- Future Roadmap

Project 8: Enterprise Feature Store & Model Registry Pipeline

Project Overview

Build an **enterprise-grade feature store + model registry pipeline** to manage features and models consistently across training and serving. Feature stores are a core part of modern production ML workflows.

Milestone 1: Feature Engineering Pipeline

Objectives

- Define reusable features for multiple models

Tasks

1. Data Collection

- a. Gather raw data streams and batch sources

2. Feature Pipelines

- a. Transform raw data to features using scheduled ETL

3. EDA

- a. Validate feature distributions and quality

Deliverables

- Feature Dataset
- Feature Pipeline Docs
- EDA Report

Milestone 2: Model Training Integrating Feature Store

Objectives

- Train models using centrally stored features

Tasks

1. Model Training

- a. Use features to train classification/regression models

2. Evaluation

- a. Evaluate with appropriate metrics (ROC, MAE)

3. Optimization

- a. Hyperparameter tuning

Deliverables

- Trained Models
- Evaluation Report

Milestone 3: Cloud Deployment with Registry

Objectives

- Deploy models with version control

Tasks

1. Model Registry

- a. Register models in MLflow or Azure registry

2. Deployment

- a. Deploy chosen model version as Azure endpoint

3. API Integration

- a. Expose feature + model scoring API

Deliverables

- Registered & Deployed Model
- API Service

Milestone 4: MLOps & Lifecycle Monitoring

Objectives

- Monitor feature quality and model performance

Tasks

1. Tracking & Logging

- a. Monitor drift in feature distributions

2. Alerts

- a. Set alerts on data and model deviation

3. Automated Retraining

- a. Retrain when drift thresholds are met

Deliverables

- Feature & Model Monitoring
- Retraining Pipelines

Milestone 5: Final Documentation & Presentation

Deliverables

- Full Report
- Demo
- Roadmap

Project 9: Automated Explainable AI (XAI) System

Project Overview

Build an ML workflow that includes **explainability dashboards** so business stakeholders can understand model decisions. Explainability is increasingly required in regulated domains.

Milestone 1: Data and Preprocessing

Tasks

- Gather a tabular dataset (e.g., credit scoring)
- Clean and preprocess
- EDA showing feature influence

Deliverables

- Processed Data
- EDA + Feature Analysis Doc

Milestone 2: Model Training & Explainability Integration

Tasks

- Train models (e.g., Gradient Boosting)
- Integrate SHAP or LIME explanations

Deliverables

- Trained Model + XAI Artifacts
- Evaluation & XAI Report

Milestone 3: Cloud Deployment

Tasks

- Deploy model + explanation API via Azure ML
- Build dashboard interface for visual explanations

Deliverables

- Deployed XAI Service
- API & Dashboard

Milestone 4: Monitoring & MLOps

Tasks

- Track model performance and explanation stability
- Monitor for drift

Deliverables

- MLOps Documentation
- Monitoring Setup

Milestone 5: Final Documentation & Presentation

Deliverables

- Report
- Demo
- Future Enhancements

Project 10: Secure & Compliant ML Security Pipeline

Project Overview

Implement a **secure, compliant ML pipeline** for **regulated domains** (e.g., finance/healthcare), including encryption, audit logs, and secure deployment.

Milestone 1: Data Security & Preprocessing

Tasks

- Collect data respecting privacy compliance
- Encrypt sensitive fields
- EDA on encrypted data

Deliverables

- Secure Dataset
- Security Preprocessing Docs

Milestone 2: Model Development & Risk Analysis

Tasks

- Train models (e.g., fraud detection)
- Risk evaluation with privacy metrics

Deliverables

- Trained Model
- Risk + Compliance Report

Milestone 3: Secure Deployment

Tasks

- Deploy via Azure with Key Vault & secure access
- Harden endpoints

Deliverables

- Securely Deployed Service
- Security Architecture Docs

Milestone 4: MLOps & Audit Monitoring

Tasks

- Track model changes securely
- Create audit logs

Deliverables

- Monitoring + Audit Logs
- Compliance Pipeline

Milestone 5: Final Documentation & Presentation

Deliverables

- Secure ML Report
- Demo
- Enhancement Ideas