

Created by CasualCraft

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# SYSTEM ARCHITECTURE

CrediSentinel: Pre-Delinquency Risk  
Prediction Platform

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# SYSTEM ARCHITECTURE

## High level overview

Flow:

Data Sources → Data Processing → Feature Engineering → ML Model → Risk Scoring → Dashboard + Alerts



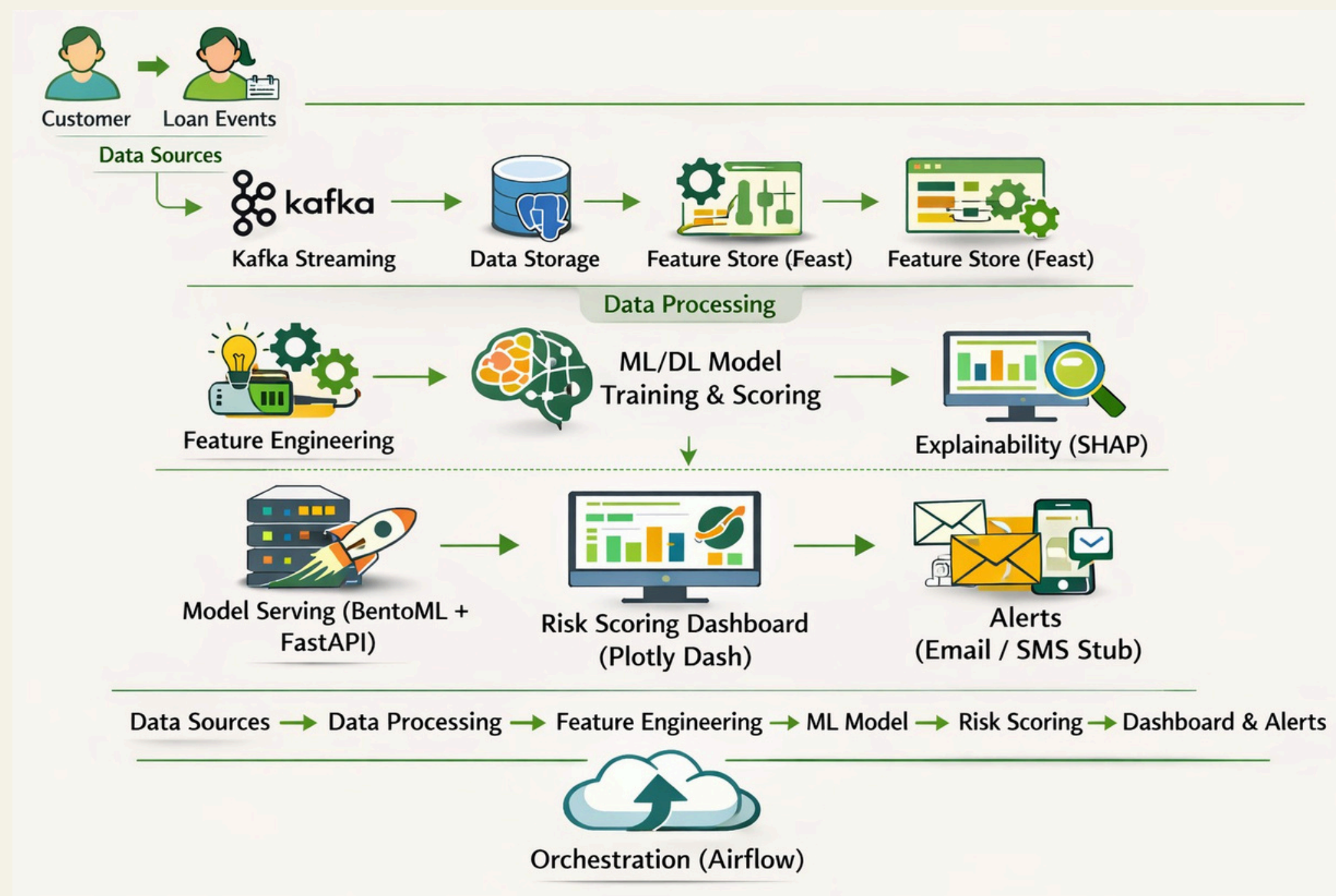


# SYSTEM ARCHITECTURE

## High level overview

### Pipeline Flow:

- Customer + Loan Events (repayments, utilization, missed EMI)
- Kafka Streaming Layer
- Data Storage (PostgreSQL)
- Feature Store (Feast)
- ML/DL Model Training & Scoring
- Explainability (SHAP)
- Model Serving (BentoML + FastAPI)
- Dashboard (Plotly Dash)
- Alerts (Email/SMS Stub)
- Orchestration (Airflow)





# STREAMING AND DATA LAYER

## Streaming

- Apache Kafka
- Captures:
  - Payment events
  - Loan updates
  - Transaction patterns

## Database

PostgreSQL

Stores:

Raw customer/loan data  
Historical repayment logs  
Risk scoring outputs

Real-Time Data Processing & Storage





# FEATURE ENGINEERING LAYER

The Feature Engineering Layer ensures consistent and reliable data preparation using a centralized feature store (Feast). Financial attributes such as income patterns, creditworthiness, spending behavior, loan exposure, and macro-risk factors are organized into structured feature groups. These engineered features are then used by both Machine Learning models (XGBoost, LightGBM) for structured data and Deep Learning models (LSTM/GRU) for sequential risk analysis. The processed data ultimately generates a probability score that classifies customers into Low, Medium, or High risk categories, supporting accurate and scalable risk prediction.

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# FEATURE STORE

- Feast
- Feature Groups Used
- Income & Employment
- Income\_monthly, employment\_type
- Income\_volatility\_6m
- Creditworthiness & Bureau
- Credit\_score
- Bureau\_inquiries\_6m
- Utilization & Spending Behaviour
- Credit\_card\_utilization
- Spend\_growth\_6m, expense\_volatility\_6m
- Balance\_slope\_6m
- Loan Exposure & EMI Burden
- personal\_loan\_outstanding,
- home\_loan\_outstanding
- total\_outstanding, total\_emi
- emi\_to\_income\_ratio
- Macro & Risk Context
- macro\_stress, delinquency\_stage



Feature Store  
**Feast**

## Feature Groups Used



### Income & Employment

- income\_monthly
- employment\_type
- income\_volatility\_6m



### Creditworthiness & Bureau

- credit\_score
- bureau\_inquiries\_6m



### Utilization & Spending Behaviour

- credit\_card\_utilization
- spend\_growth\_6m, expense\_volatility\_6m
- balance\_slope\_6m



### Utilization & Spending Behaviour

- credit\_card\_utilization
- spend\_growth\_6m, expense\_volatility\_6m
- balance\_slope\_6m



### Loan Exposure & EMI Burden

- personal\_loan\_outstanding
- home\_loan\_outstanding
- total\_outstanding\_total\_emi
- emi\_to\_income\_ratio



### Macro & Risk Context

- macro\_stress
- delinquency\_stage



# ML AND DL INTELLIGENCE LAYER

## Risk Prediction Models

ML Models

XGBoost

LightGBM

Used for:

Structured/tabular banking data

High accuracy & fast inference

## DL Models

PyTorch (LSTM/GRU)

Used for:

Sequential repayment patterns

Time-series risk behavior

## Output

Probability score (0–1)

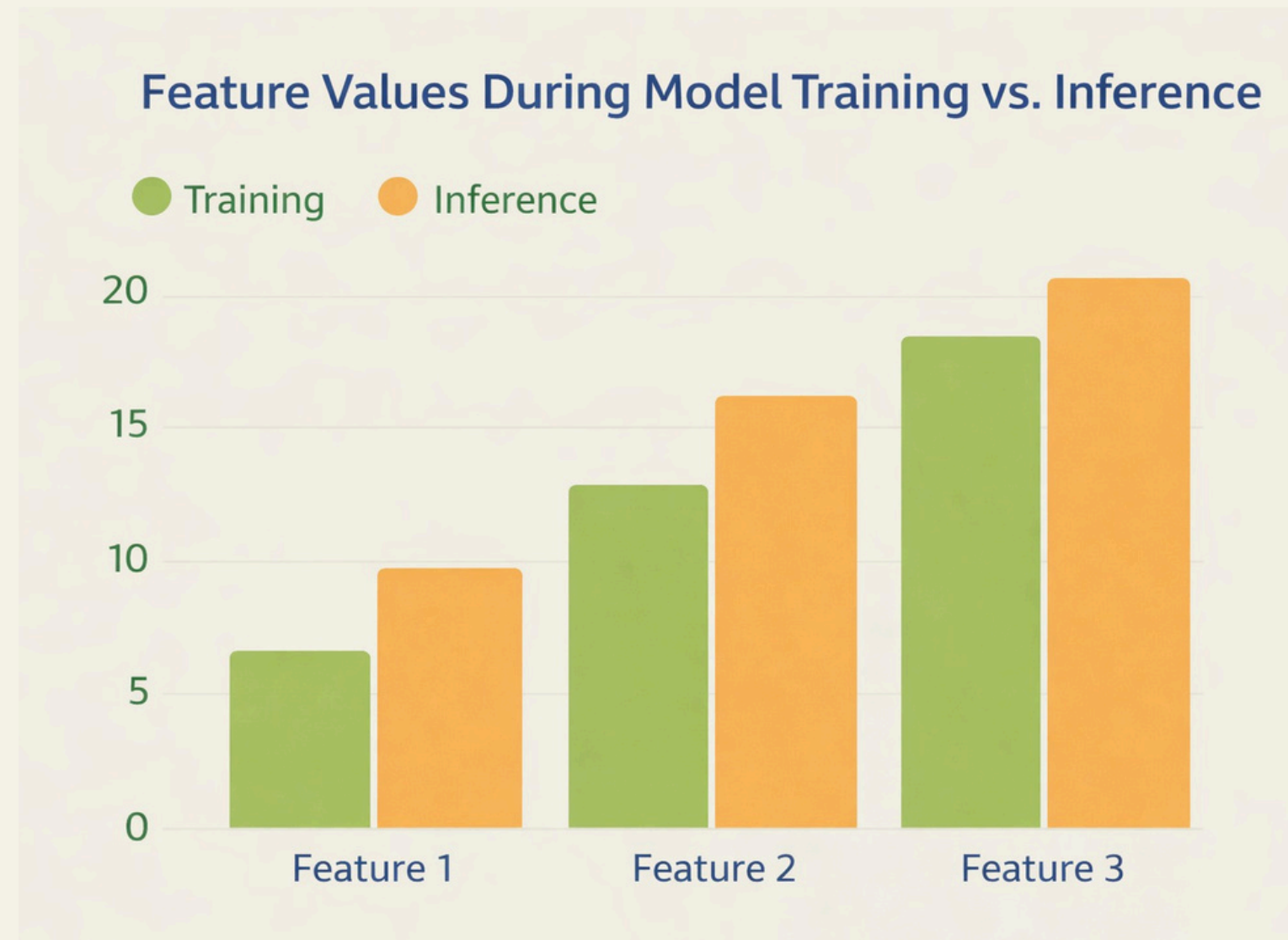
Risk bucket: Low / Medium / High





# FEATURE DISTRIBUTION

This bar chart compares key feature values used during model training and real-time inference. It helps validate that the feature distribution remains consistent across both stages, reducing the risk of data drift and ensuring the model performs reliably in production.





# EXPLAINABILITY AND PRIVACY LAYER

- In banking, accuracy alone is not enough — predictions must be transparent, auditable, and privacy-safe
- SHAP (Explainable AI) is used to explain every risk score
- SHAP highlights the top drivers behind pre-delinquency predictions, such as:
  - emi\_to\_income\_ratio
  - credit\_card\_utilization
  - balance\_slope\_6m
  - bureau\_inquiries\_6m
- This helps risk teams and collections teams trust model outputs and take informed action
- Opacus (Differential Privacy in PyTorch) is applied during training to:
  - Prevent customer data memorization
  - Reduce the risk of sensitive information leakage
- This layer makes CrediSentinel a Responsible AI system:
  - Accurate
  - Explainable
  - Privacy-compliant / deployment-ready

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## Trust, Compliance, and Responsible AI



# MODEL SERVING AND API LAYER

## Serving

BentoML + FastAPI

## API Outputs

Risk Score

Risk Band

SHAP Explanation

Suggested action label

This is enterprise-grade  
because :

- Scalable
- Fast inference
- Easy integration with bank systems



Production-Ready Model Deployment



# DASHBOARD AND ALERTS LAYER

## Dashboard

Plotly + Dash

Displays:

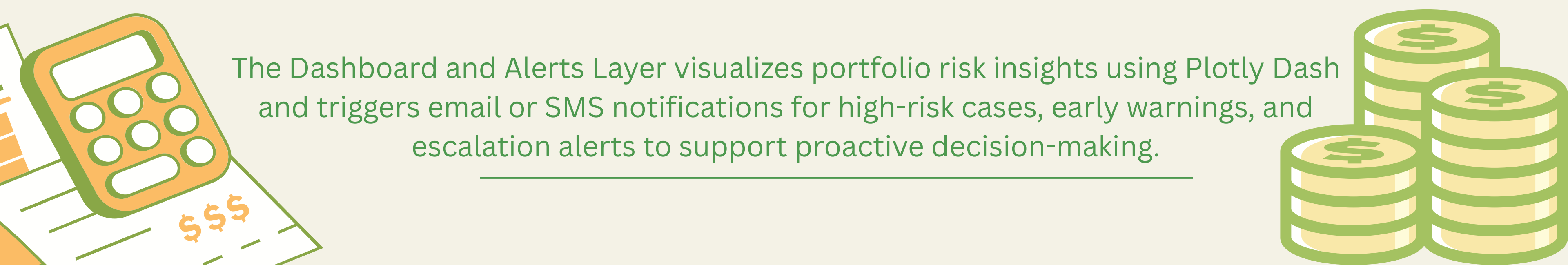
- Portfolio risk summary
- High-risk customers list
- Risk trends over time
- Customer drill-down

## Notifications

Email / SMS Stub

Triggers:

- High-risk alerts
- Early warning alerts
- Escalation alerts



The Dashboard and Alerts Layer visualizes portfolio risk insights using Plotly Dash and triggers email or SMS notifications for high-risk cases, early warnings, and escalation alerts to support proactive decision-making.



# ORCHESTRATION LAYER

Automated ML Workflow

Management

Orchestration

Apache Airflow

Scheduled Pipelines

Daily ingestion

Feature computation

Batch scoring

Weekly/monthly retraining

Monitoring & logging

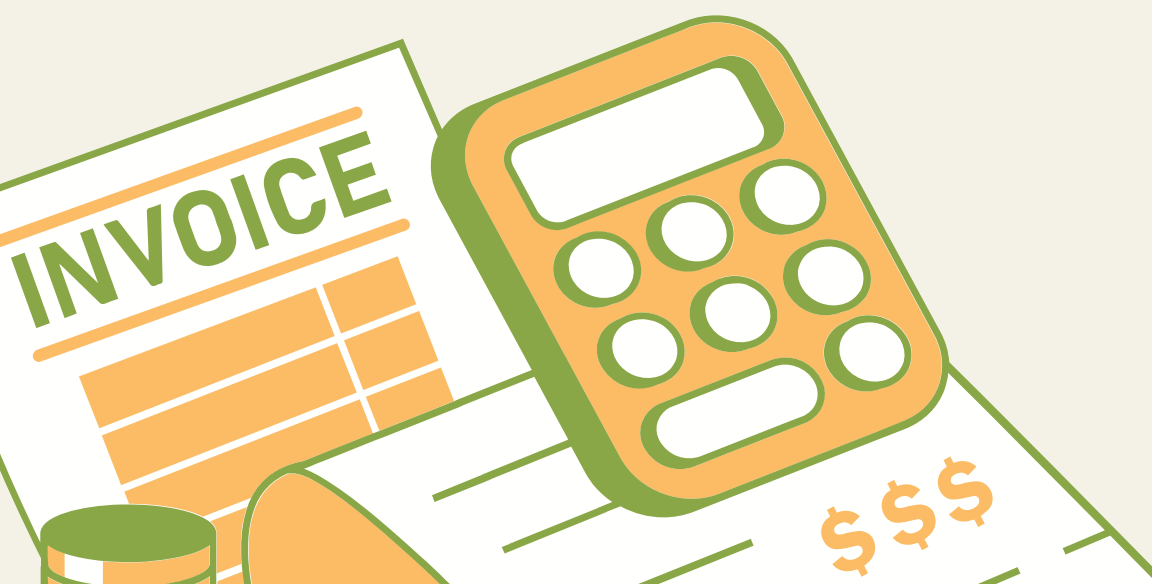


Automates ML workflows using Airflow, managing pipelines, scoring, retraining, monitoring, and scheduled data processing.



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# THANK YOU



By Team - CasualCraft

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