From Paperwork to Pipeline: Building MLOps for Autonomous Mortgage Al

Transforming mortgage processes with autonomous AI for compliance, efficiency, and scalability





About the Speaker

Interests / Achievements

- Delivered \$8M+ annual savings via enterprise automation & Al integration
- Architected **generative AI chatbot** for loan servicing & compliance
- Designed AI analytics platform for multi-channel customer insights
- Built intelligent document processing pipeline for financial workflows
- Led **cloud migration** & governance for enterprise automation ecosystem

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Mortgage Industry Challenges and MLOps Opportunities



Highlight complexity of traditional mortgage processing

Mortgage loan processing involves over 47 manual steps and takes 30-60 days on average per application, creating inefficiencies and delays.

2

Emphasize high manual effort in document review

Between 60-80% of document reviews require human expertise, increasing labor costs and potential for human error in mortgage processing.

3 ➪2

Address challenges from legacy systems and integration limits

Multiple legacy mortgage systems lack robust API integration, complicating data flow and automation efforts across platforms.

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Stress importance of regulatory compliance in mortgage industry

Mortgage processing must comply with strict regulations such as GDPR, CCPA, and Fair Lending Acts, requiring careful data handling and auditability.

5 (6)

Promote MLOps as a transformational solution

Implementing autonomous AI pipelines via MLOps can reduce mortgage processing times by 90% while ensuring regulatory compliance and reducing manual workload.

Scale Requirements



Process 100K+ loan applications daily with sub-second responses and 24/7 zero downtime availability

Regulatory Complexity



Ensure explainable AI for audits, fair lending compliance, and adapt to multi-jurisdictional rules

Legacy Integration



Integrate 1980s mainframe systems and batch workflows with real-time AI processes



Unique MLOps Challenges in **Autonomous** Mortgage Al

Key technical and regulatory hurdles for scalable, compliant mortgage AI systems

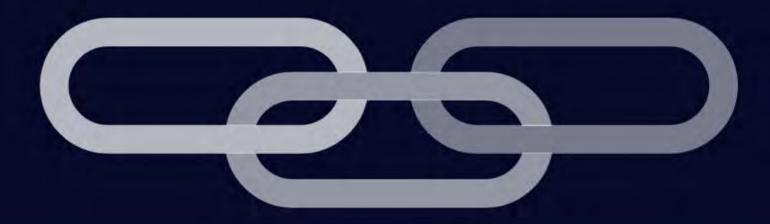
Core MLOps Foundations Adapted for Mortgage Al

Data Governance for Compliance

Includes audit trails, PII encryption, anonymization, and dataset version control with lineage tracking.

Deployment Strategy

Employs blue-green deployments, shadow mode testing with human validation, and gradual rollout with rollback.



Model Development Practices

Uses containerized training, automated bias detection, fairness validation, and compliant documentation generation.

Data Pipeline Architecture for Mortgage Al

Key components and challenges in building AI-driven mortgage data pipelines

1

Data Ingestion Challenges

Handling unstructured documents, real-time feeds, and integrating with legacy systems to ensure seamless mortgage data ingestion.

2

Document Processing

Using **OCR**, **NER**, and structured data extraction techniques to convert complex mortgage documents into usable, machine-readable formats.

3

Feature Engineering

Deriving key features like **income**, **debt-toincome ratios**, and **risk scores** from raw mortgage data for AI modeling.

4

Data Validation

Ensuring data completeness, detecting anomalies, and enforcing strict quality gates to maintain data integrity.

5

Privacy Compliance

Implementing automated **PII masking** and enforcing **data retention policies** to meet regulatory requirements effectively.

6

Storage Strategy

Leveraging a **feature store** for low-latency retrieval, **data lake** for historical storage, and **real-time streaming** for fast decision making.

Model Development & Training Pipeline for Mortgage Al

Comprehensive process integrating data management, training, and validation for mortgage AI models

Training Data Management

Manage historical loan data, generate synthetic data for edge cases and bias mitigation, and collect ongoing production outcomes.



Validation Gates

Implement business KPI simulation, fairness testing, automated regulatory reporting, and stress testing against market volatility before deployment.

Model Training Architecture

Utilize multi-model ensemble including credit risk, fraud detection, and document validation with automated hyperparameter tuning and cross-validation for stability.

Canary Release (30-60 days)

- Limited traffic deployment •
- Route 5-10% of traffic to new models with real-time KPI and error rate monitoring, enabling automated rollbacks to maintain performance standards during early adoption.

Shadow Mode (0-30 days)

- · Parallel model validation
- AI models run alongside existing systems
 without influencing decisions, allowing
 human agents to validate recommendations
 and gather performance metrics for detailed
 analysis.

Full Deployment (60+ days)

- Gradual traffic shift and optimization
- Transition to full traffic with continuous monitoring, employing A/B testing to optimize models and benchmarking outcomes against business objectives to ensure sustained value.

Progressive Deployment Strategies for Mortgage Al Models Phased approach to ensure reliable, optimized AI integration in residential lending

Continuous Integration & Delivery (CI/CD) for Mortgage AI

Key components and automation processes for reliable mortgage AI deployment

Version Control

Manages model code and **configuration** changes securely and collaboratively with Gitbased systems.

Model Registry

Tracks model **metadata** and ensures integrity through artifact signing for secure deployments.

Dependency Mgmt

Automatically handles dependencies and performs **security scanning** to ensure safe builds.

Unit Testing

Validates components like data transformation and feature engineering for accuracy.

Integration Test

Verifies interoperability of model deployment infrastructure components for stable serving.

Workflow Tests

Simulates full loan processing scenarios to ensure end-to-end system functionality.













Orchestration & Workflow Automation in Mortgage MLOps

Verification tasks depend on each other, requiring careful sequencing and coordination to maintain workflow integrity and avoid delays.

Service Int.

Maintaining state allows the system to be resilient during failures, enabling recovery without data loss or process interruption.

Loan Time

Mortgage loans take 30-60 days with multiple verification steps involving credit checks, appraisals, and compliance reviews to ensure accuracy and approval.

Dependencies

Integration with credit bureaus, appraisers, and other third-party services is essential for real-time data validation and loan processing.

State Mgmt

Orchestration & Workflow Automation in Mortgage MLOps

Event-Driven

Using events and message queues creates a scalable, loosely coupled orchestration framework that enhances responsiveness and fault tolerance.

Tools like Airflow and Prefect orchestrate DAGs for complex task scheduling, ensuring reliable and timely execution of loan processes.

Workflow Eng.

Retry Logic

Implementing retry mechanisms and circuit breakers protects the system by handling transient failures gracefully and preventing cascading errors.

Event sourcing provides detailed audit trails and supports accurate recovery of system state changes after failures or restarts.

Event Source

Orchestration & Workflow Automation in Mortgage MLOps

Retraining

relevance.

Automated model **retraining** triggers when performance metrics drop below predefined thresholds, ensuring continued accuracy and

Automated data quality checks halt pipelines immediately when anomalies or failures are detected, preventing downstream errors.

Resource Alloc

Data Quality

Dynamic resource allocation scales computing and storage based on loan volume, optimizing operational efficiency and cost.



Compliance & Security Considerations in Mortgage Al MLOps



Ensure regulatory compliance through explainability and bias detection

Implement model
explainability for audits
and adverse action
notices, detect bias to
maintain fair lending
compliance, track data
lineage with audit trails,
and monitor model
performance across
demographic groups.



Implement robust security with encryption and access control

Use end-to-end
encryption for data in
transit and at rest,
enforce role-based
access control following
the least privilege
principle, maintain audit
logs of all model
interactions, and
automate compliance
reporting for regulators.



Apply MLOps security patterns for secure model operation

Secure model serving with authentication and authorization, anonymize PII in training pipelines and feature stores, conduct vulnerability scanning of dependencies and containers, and establish incident response procedures for security violations.

Optimize & Scale AI for Real-Time Mortgage Decisions

Implement efficient strategies to meet performance demands and ensure seamless scaling in residential lending AI systems

Achieve sub-second latency

Ensure **real-time loan decisions**with inference times under one
second for fast customer responses.

Apply model quantization & pruning

Reduce model size and latency using **quantization** and **pruning** while keeping prediction accuracy.

Optimize batch inference

Process loan applications in optimized batches to increase throughput and handle high volumes.

Cache frequent features & predictions

Use **caching** to reduce redundant computations, speeding up responses and cutting resource use.

Enable workload-based auto-scaling

Adjust compute resources dynamically by tracking queue depth and processing times for efficiency.

Monitor latency and resource use

Use dashboards to track **latency percentiles** and resource use, preventing bottlenecks.

Balance cost with predictive scaling

Leverage **predictive analytics** to optimize resources, lowering costs while maintaining performance.

Automate SLA breach alerts

Implement alerts to notify teams instantly on **SLA breaches**, ensuring rapid incident response.

Incident Management & Anomaly Detection in Mortgage Al

Key failure modes, detection systems, and response procedures in residential lending AI



Common failure modes in Mortgage Al

Includes model degradation from data drift, infrastructure failures, data quality issues, and external service outages.



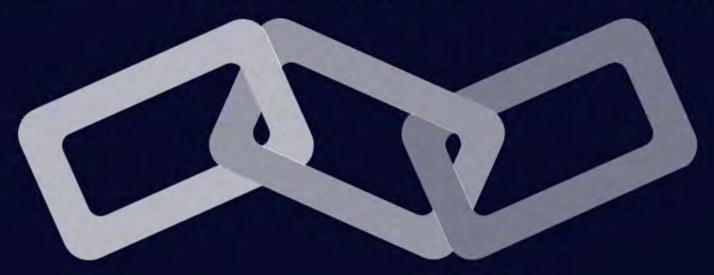
Detection and response systems

Utilizes statistical drift detection, performance anomaly detection, infrastructure health checks, and business impact monitoring.



Incident response procedures

Features automated rollback, expert escalation, root cause analysis, and communication protocols for incident management.



Future Directions & Key Takeaways

Explore advanced technologies and best practices driving automotive AI success and operational excellence

1 LLMs

Use **LLMs** to automate docs.

2 Federated

Train models with **privacy**.

3 Quantum

Solve tasks with quantum.

4 Edge

Run AI on **edge** devices.

5 Regulatory

Build **compliance** pipelines.

6 Deployment

Use **canary** and shadow modes.

7 Monitoring

Track **KPIs** and metrics.

8 Legacy

Integrate with **old** systems.

9 Feedback

Enable **continuous** updates.

10 Speedup

Cut **processing** time 90%.

11 Uptime

Keep **99.9%** availability.

12 Compliance

Ensure **zero** violations.



Let's align, adapt, and accelerate for a breakthrough transformation.