# Lessons learned from productising Al workloads with Open Source technology

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## Use Case - GenAl for Equities and Risk Management

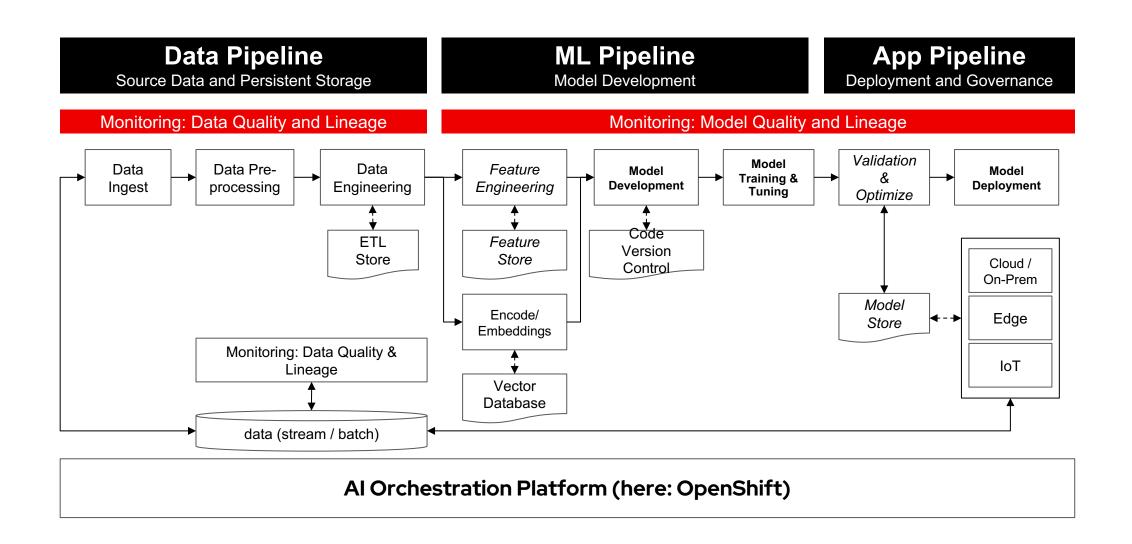
Investment Bank would like a chatbot that's trained on their own equities data so that they can avoid going to multiple sources (dashboards, applications, spreadsheets) to answer questions on portfolios, assets, and risk. They are open to RAG, Alignment Tuning, and Fine Tuning of LLMs.

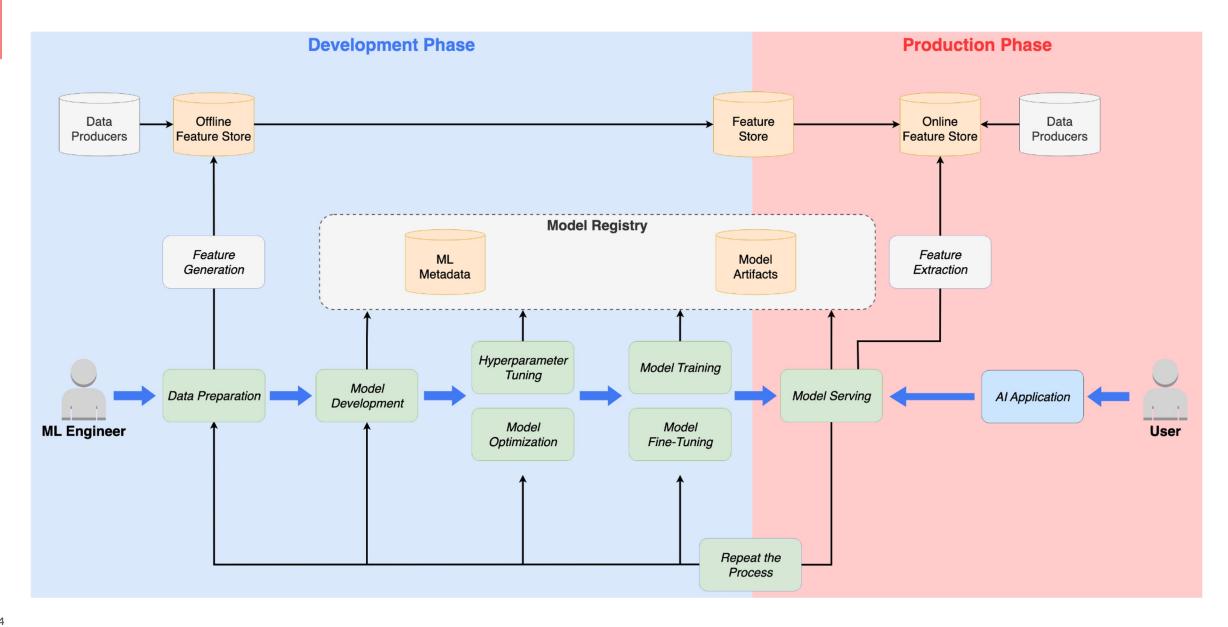


#### **Chatbot Answering on Equities and Risk**

- Portfolio Monitoring: Track portfolio performance, alerting users to significant changes or potential risks.
- Risk Assessment: Evaluate investment risk levels, suggesting adjustments based on market conditions.
- **Scenario Simulation:** Run hypothetical scenarios to predict outcomes and advise on risk mitigation strategies.
- **Regulatory Compliance:** Ensure adherence to regulations by providing guidance on compliance issues.
- Automated Reporting: Generate detailed reports on equities performance and risk metrics.
- User Queries: Answer questions about specific stocks, market conditions, and investment strategies.

#### Our initial approach to model deployments





### When going into production...

# Deployment infrastructure and scalability

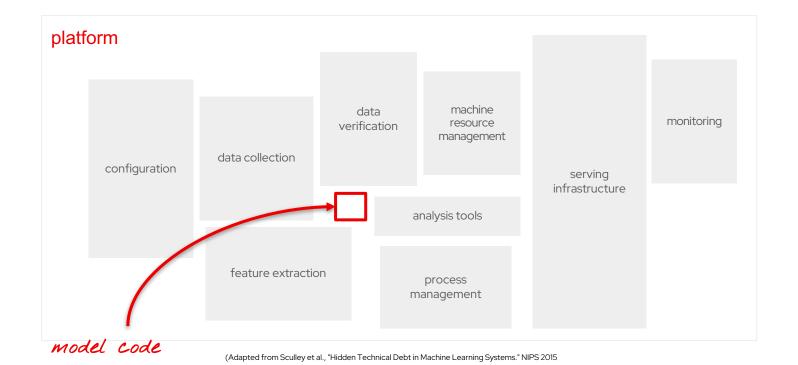
- → All Cloud: Regulatory issues
- → All on-prem: Resource constraints
- → Both!

#### **Kubeflow**

- → Keeping up with changes
- → Resource Management
- → Version Compatibility
- → Bugs, bugs, bugs

We need an operations team and engineering partner!

### Model code is one component of a larger system



Hidden Technical
Debt in Machine
Learning Systems

"Only a small fraction of real-world ML systems is composed of the ML code...The required surrounding infrastructure is vast and complex."

"Developing and deploying ML systems is relatively fast and cheap, but maintaining them over time is difficult and expensive"