

# ModelOps

Current Trends and Applications

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**Demonstration Only**

See Slide Masters



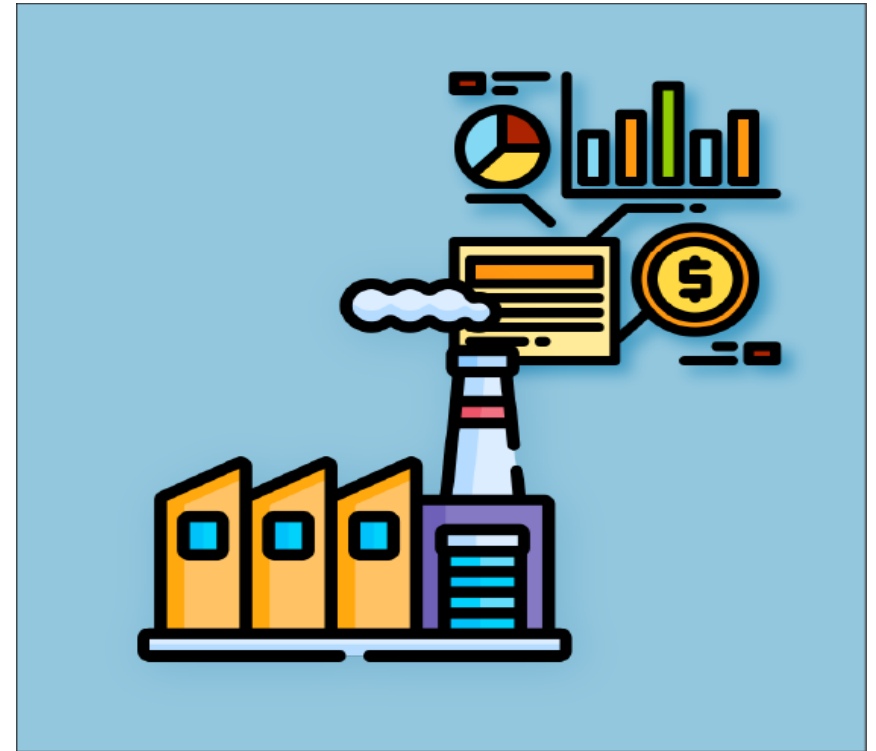
THE UNIVERSITY OF TEXAS  
AT ARLINGTON

# Key Benefits Of ModelOps In AI Projects

- Automation by ModelOps allows the deployment, monitoring, and updating of models, therefore freeing the teams from their burdensome tasks to enhance operational efficiency.
- It amplifies collaboration among data scientists, machine learning engineers, and IT operations teams by standardizing processes and tooling for model development and deployment.
- Real-time monitoring provides immediate feedback when models start to perform poorly. Models may be easily retrained or fine-tuned to keep the results at an optimum.
- ModelOps natively brings in governance and audit trails for doing audits and compliance checks, which becomes quite critical in regulated industries such as finance and health. What this means is that it provides the ability to track the basis of decisions made by the model so regulatory requirements are followed.

# Use cases In Manufacturing

- AI models predict equipment failure before they happen and bring down machine downtime.
- ModelOps ensures continuous updates of the model with real-time machine data for retraining and continuously improving the accuracy.
- Vision-based AI models identifies product defects during the production. ModelOps ensures updating these models with new product specifications and updated data, elevating quality assurance processes.





# Use cases In Insurance

- AI models immediately evaluate insurance claims for their speedy processing and approval.
- ModelOps ensures that the models continue being accurate by retraining them with new claims data and changes in regulation.
- Models identify fraudulent claims through the pattern of claims data. ModelOps ensures that these models are always monitored and updated to capture emerging fraud tactics and risks.



# Tools and Technologies



## Model Deployment & Serving Platforms

Provide lightning-fast, scalable deployment of models for real-time and batch predictions.

Example: Kubeflow, TensorFlow, MLflow.

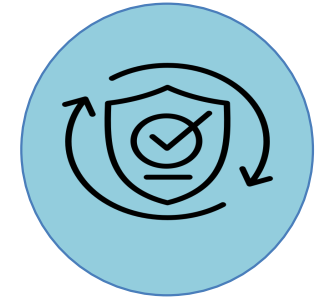


## Monitoring & Performance Management Systems

Record model accuracy, latency, and drift over time.

Utilize alerts and dashboards to enable issue identification much, much faster.

Example: Neptune.ai, Datadog.



## Model Governance & Compliance Frameworks

Make sure regulatory and ethical standards have been met.

Allow model transparency, audit trails, and bias detection.

Example: IBM Watson, DataRobot.

# Tools and Technologies



## Experimentation & Version Control Mechanisms

Enable reproducibility and tracking of model versions and experiments.

Facilitate collaboration and rollback to previous model states.

Example: Data Version Control(DVC), Git.



## Feature Stores

Centralize and manage features for consistent training and production.

Enable real time and batch access to reliable feature data.

Example: Feast, Tacton.



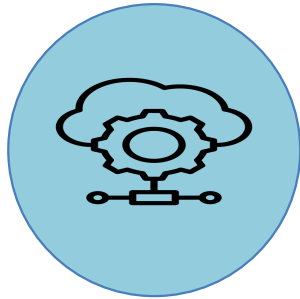
## Containerization & Virtualization Technologies

Containerization would ensure that a particular model is reproduced consistently across various environments.

Simplify the scaling and deployment of models in production.

Example: Docker, OpenShift.

# Tools and Technologies

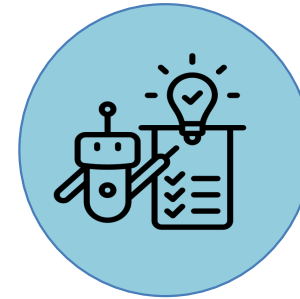


## **Workflow Orchestration & Automation Systems**

Automate the workflows of data pipelines, training, and deployment.

Manage dependencies and automate tasks in this process to ensure a smooth operation.

Example: Apache Airflow, Prefect.



## **Model Explainability & Fairness Tools**

Provide insight into model decision-making, hence transparency.

To detect and mitigate biases, thus enabling ethical model use.

Example: Fairlearn, LIME.



# ModelOps Market Ecosystem

