

# Machine Learning Platform Life-Cycle Management

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# OVERVIEW

- What is a machine learning platform?
- What is the ML platform lifecycle?
- Why ML platform lifecycle management?
- Artifacts and their associations
- Use cases at Intuit

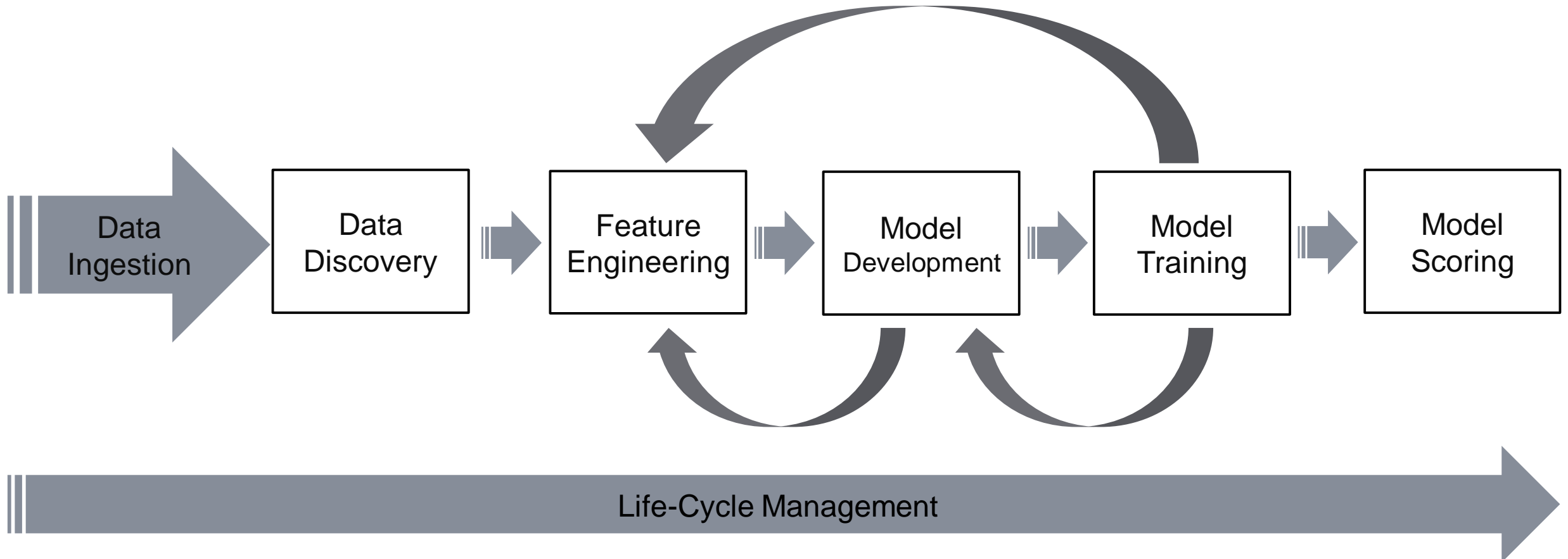


# What is a Machine Learning Platform?

- Manages the entire lifecycle of an ML model
- Includes automating and accelerating the delivery of ML applications

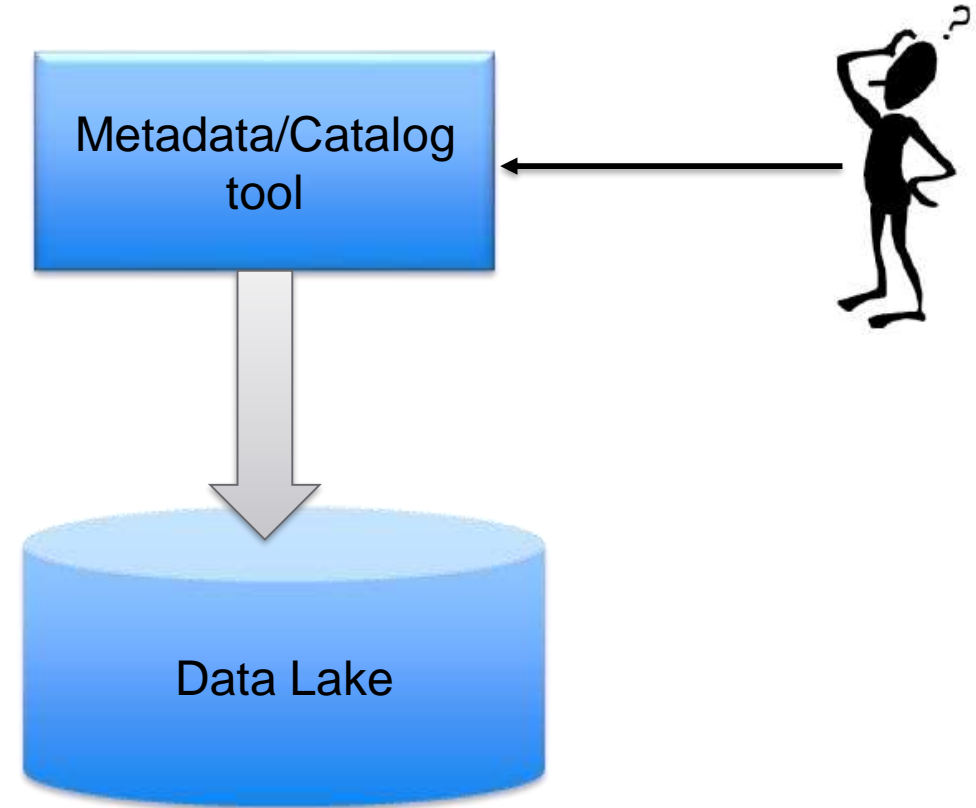


# ML Platform Life-Cycle



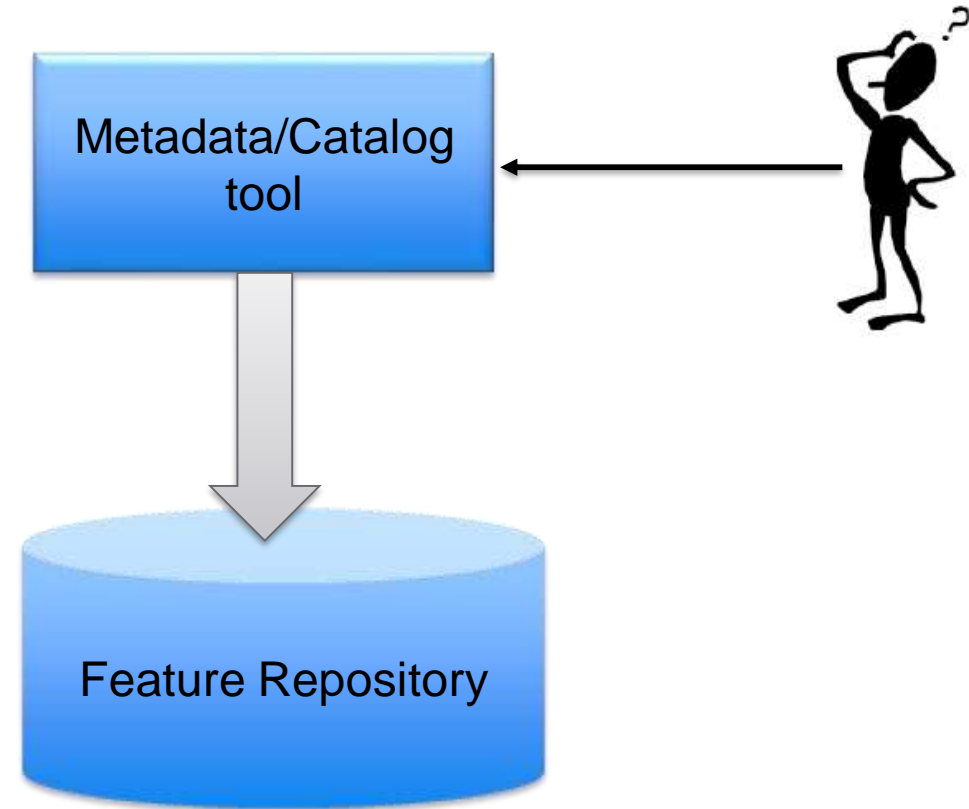
# Data Discovery

- Metadata/catalog tool
- Accessible data source  
(Raw attributes & Data lineage)



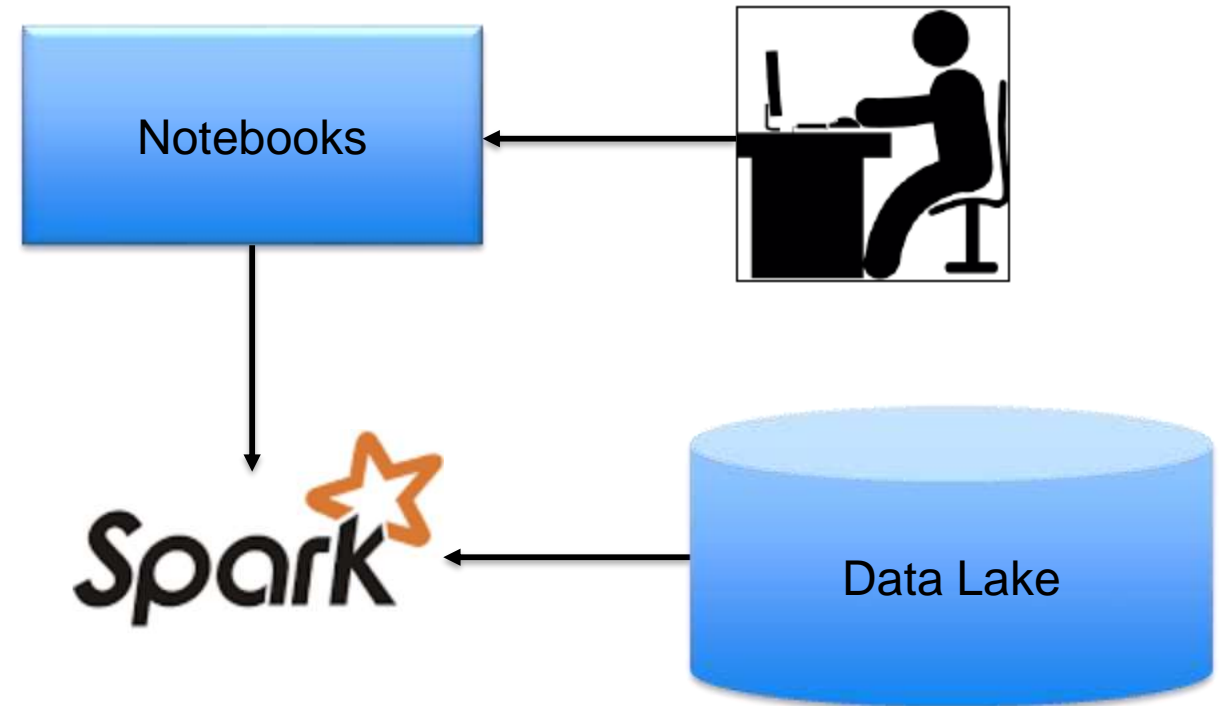
# Feature Engineering

- Output : features
- Reproducible
- Reusable



# Model Development

- Collaborative environment
- Access data lake



# Model Training

Support ability for:

- Being triggered either manually/via automation
- Creation and management of training sets
- Re-training
- Optimizing *hyper parameter tuning* through parallelization of model training execution





# Model Scoring

- Support online/offline(depend on use cases)
- Ability to be triggered either manually/via automation



# Big Mess!

- No central artifact management solution
- Hard to reuse existing features/data/algorithms/toolings
- Inability to scale for large datasets
- Lack of automation/orchestration across the ML life-cycle
- Lack of rigor/discipline in the ML development life-cycle
- **Slow down delivery of Machine Learning applications**



## Ideal Status

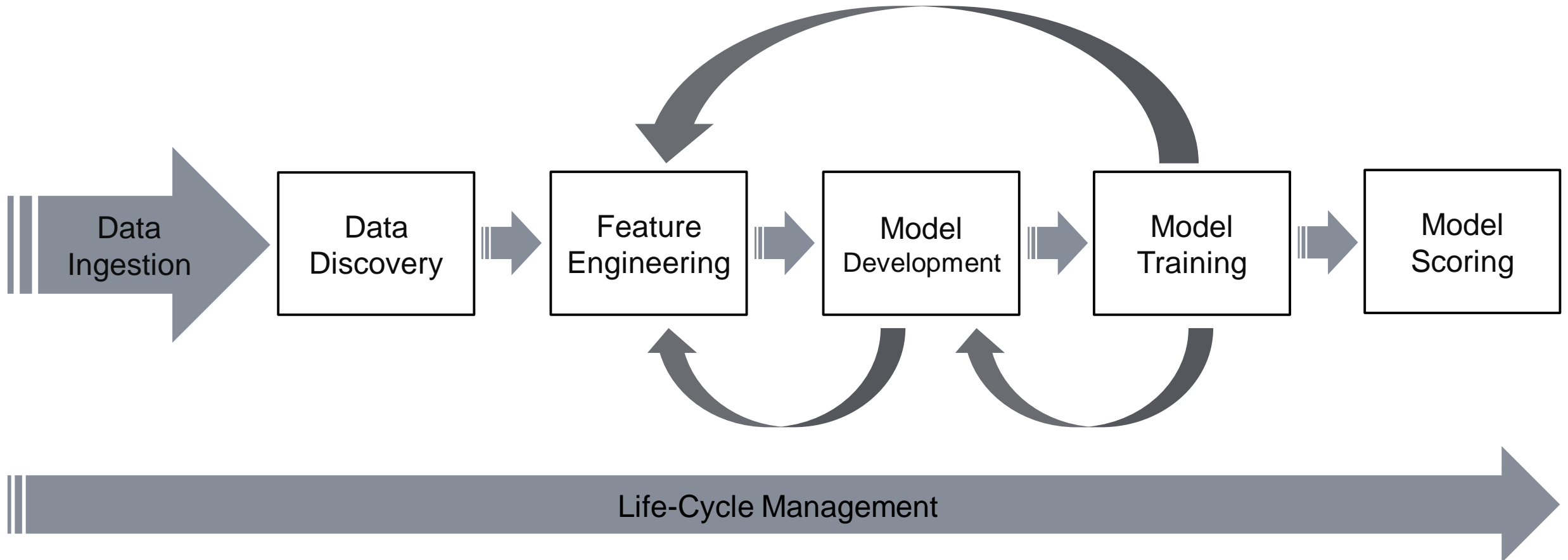
- Optimizing data scientists' engineering process
- Tie ML components together into a cohesive platform, support the life-cycle of ML artifacts end-to-end
- Increase efficiency of delivering ML predictions at scale

## Artifact Management



- **Data Artifacts**
  - Features
  - Training sets
- **Model Artifacts**
  - Model code
  - Trained models
  - Performance metrics
  - Hyper parameter values
- **Environment Artifacts**
  - Languages & language versions
  - Packages & Package versions





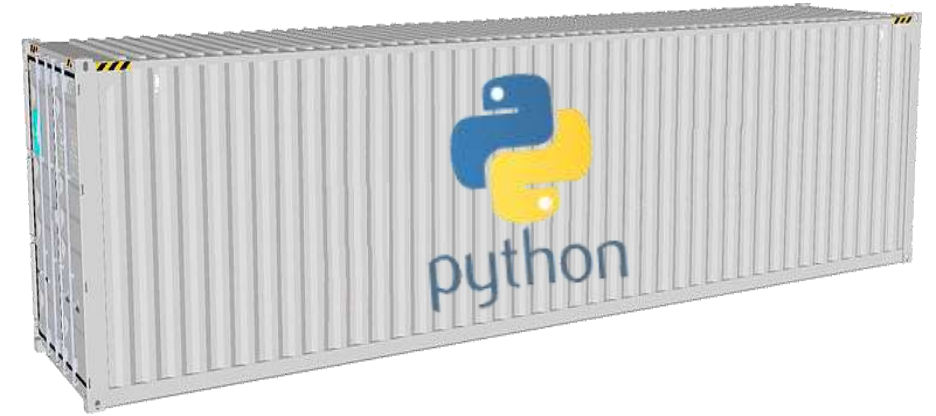
# CONTAINERIZATION!



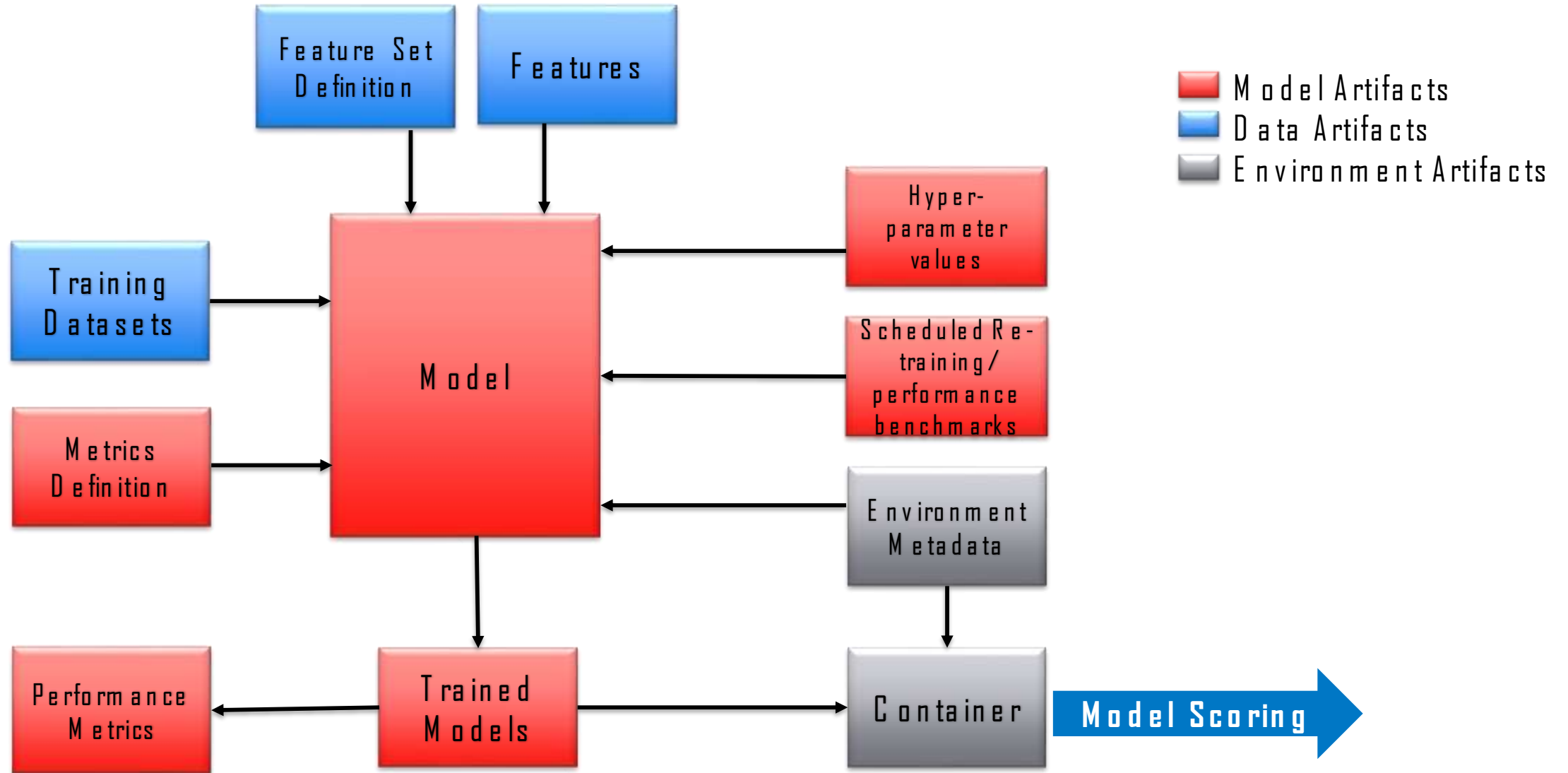
**Environment in Container**



# Benefit Of Containerization



- **Flexibility** : Model has specific environment
- **Consistency** : Model has same behavior throughout the life-cycle





# Model Code

- Developed in notebooks
- Multiple versions
- Each version associate with an externalized environment artifact



# Environment Artifacts

- Environment must be consistent for development, training, scoring
- Externalized as metadata
- Model/Execution environments constructed from metadata and deployed into containers (Docker, Yarn, Conda, etc.)



## Examples of containers/virtual environments

- Tailored to the environment (built based on externalized environment metadata)
- Used for model development, training, execution

Container/virtual environment	Usage
Docker container	<ul style="list-style-type: none"> <li>• Model development</li> <li>• Model training</li> <li>• Online scoring</li> </ul>
Yarn container	On Spark cluster <ul style="list-style-type: none"> <li>• Distributed training</li> <li>• Batch offline scoring</li> </ul>
Conda environment	Model development



# Features

- Used as data input of the model
- Stored in discoverable feature repository
- Metadata defines the model specific feature-sets



# Trained Models

- Serialized, weighted model files
- Associate with a version of model code and training set



# Training Sets

- Datasets used to train, validate and test the model
- Associated to a trained model



# Feature Set Definitions

- Define what feature sets this model requires



# Hyper Parameter Values

- Set up values before learning process
- Model specific





## Metric Definition

- Defines the metrics to collect and thresholds to evaluate models against.



# Performance Metrics

- Metrics to evaluate model effectiveness
- Model metrics including: ROC curve, confusion matrix, F1 score, precision, recall, etc.

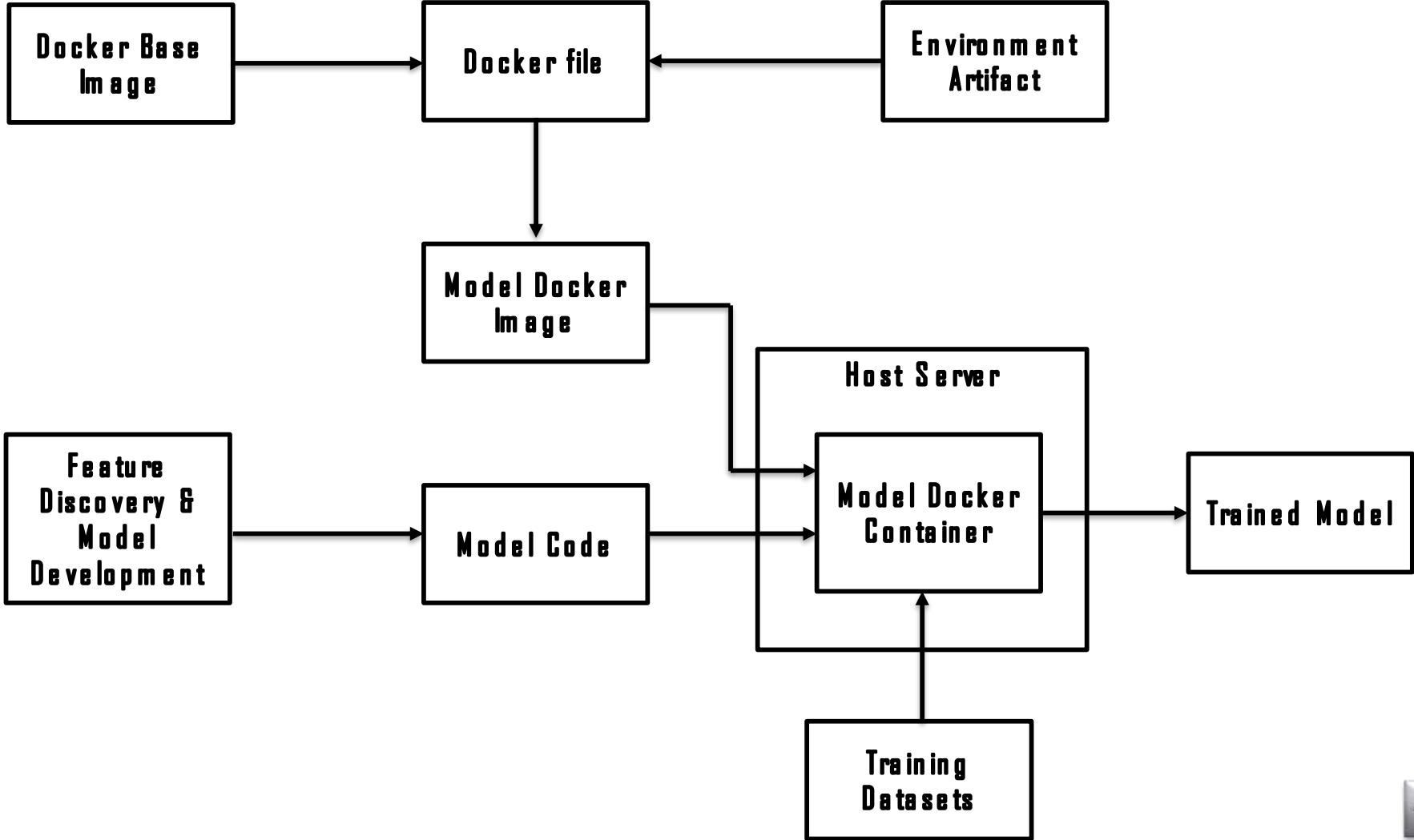


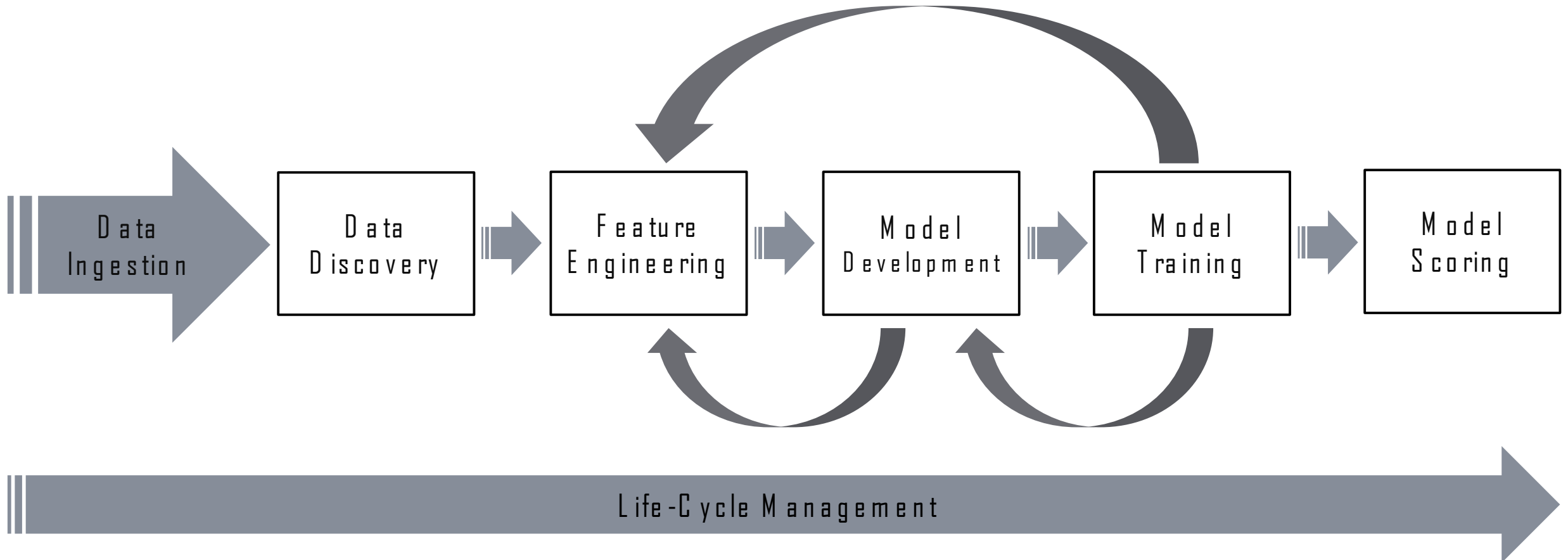
## Scheduled Re-training & performance benchmarks

- To automate the re-training and deployment of updated models
- Model specific

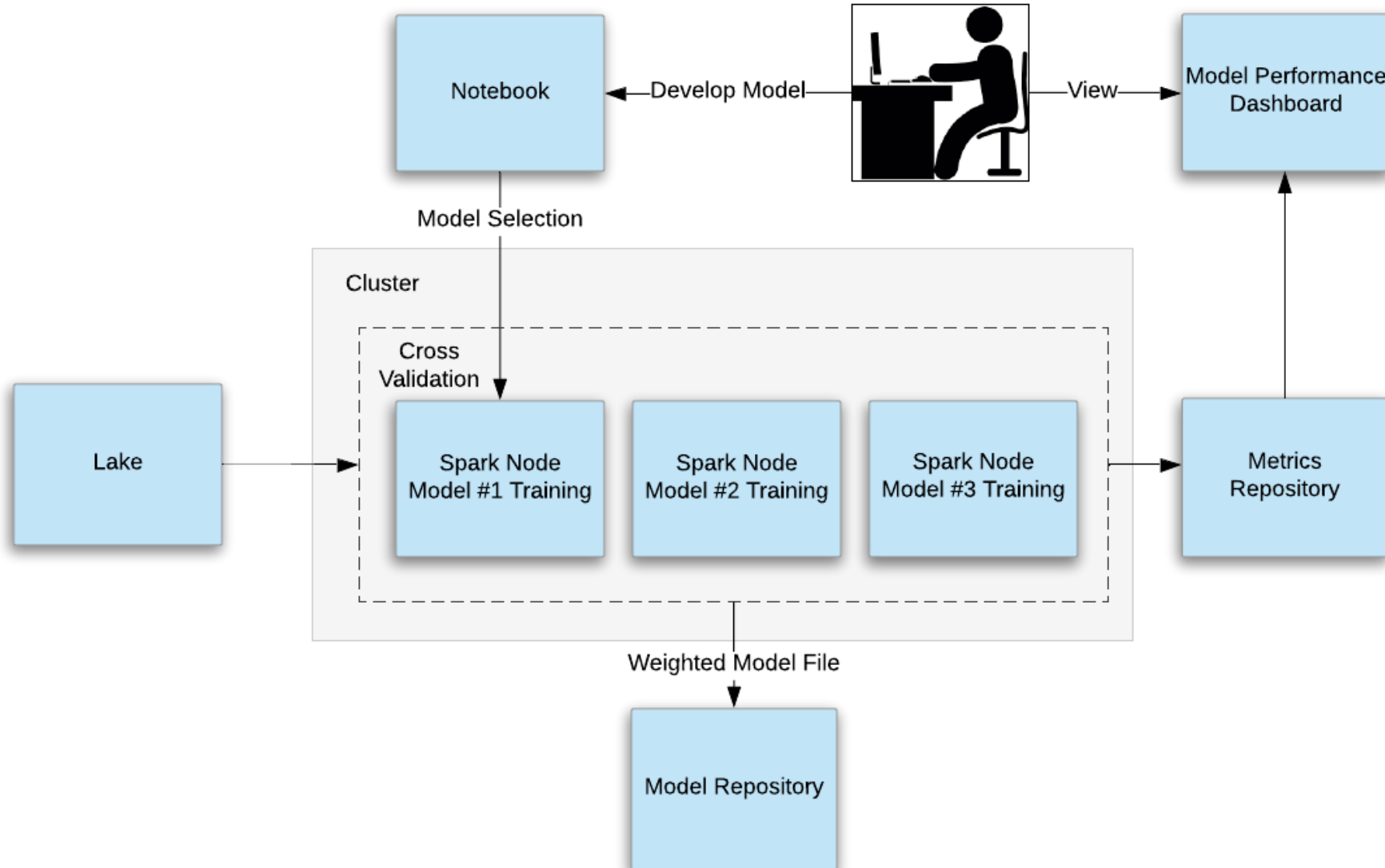


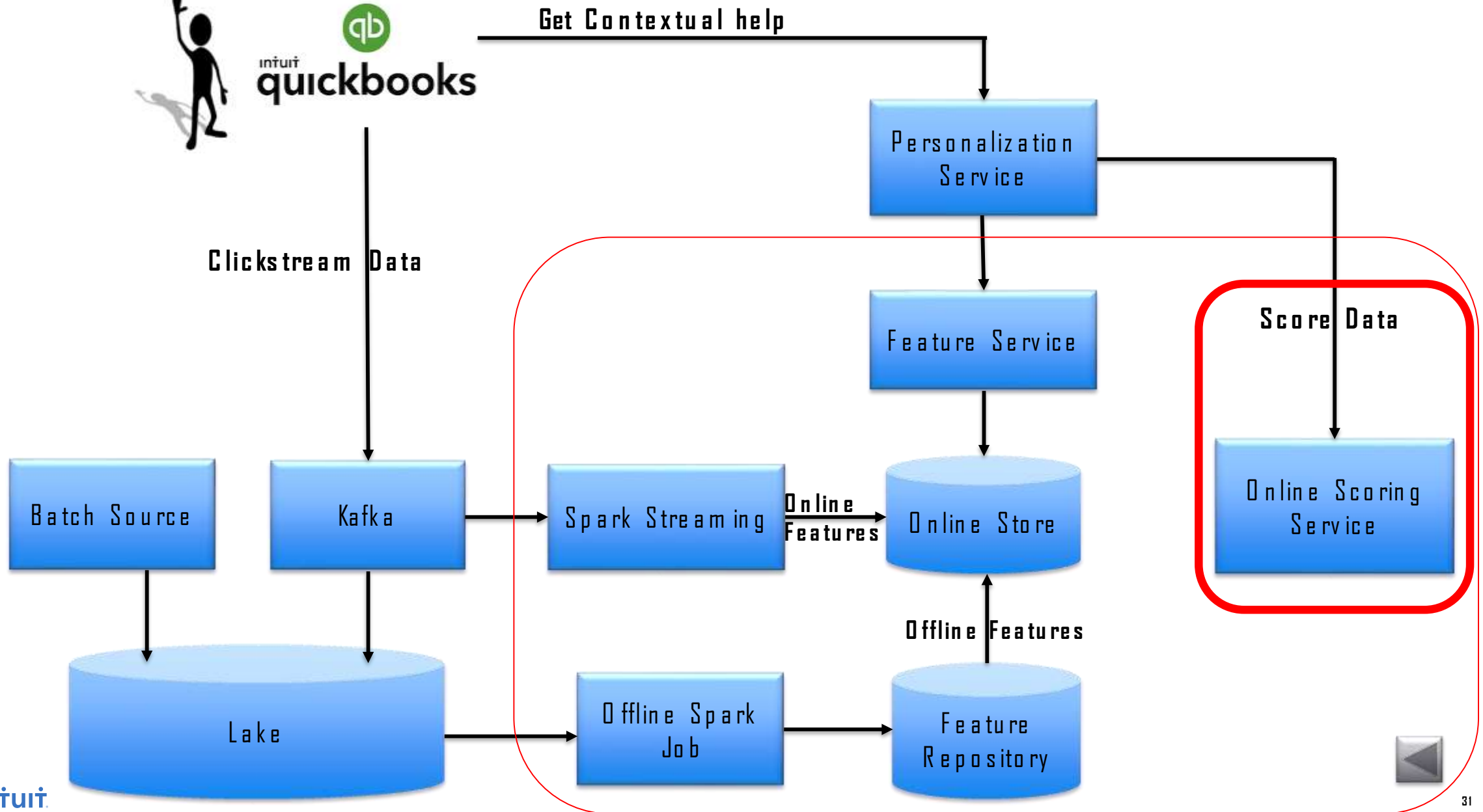
# Model Training in Docker Container



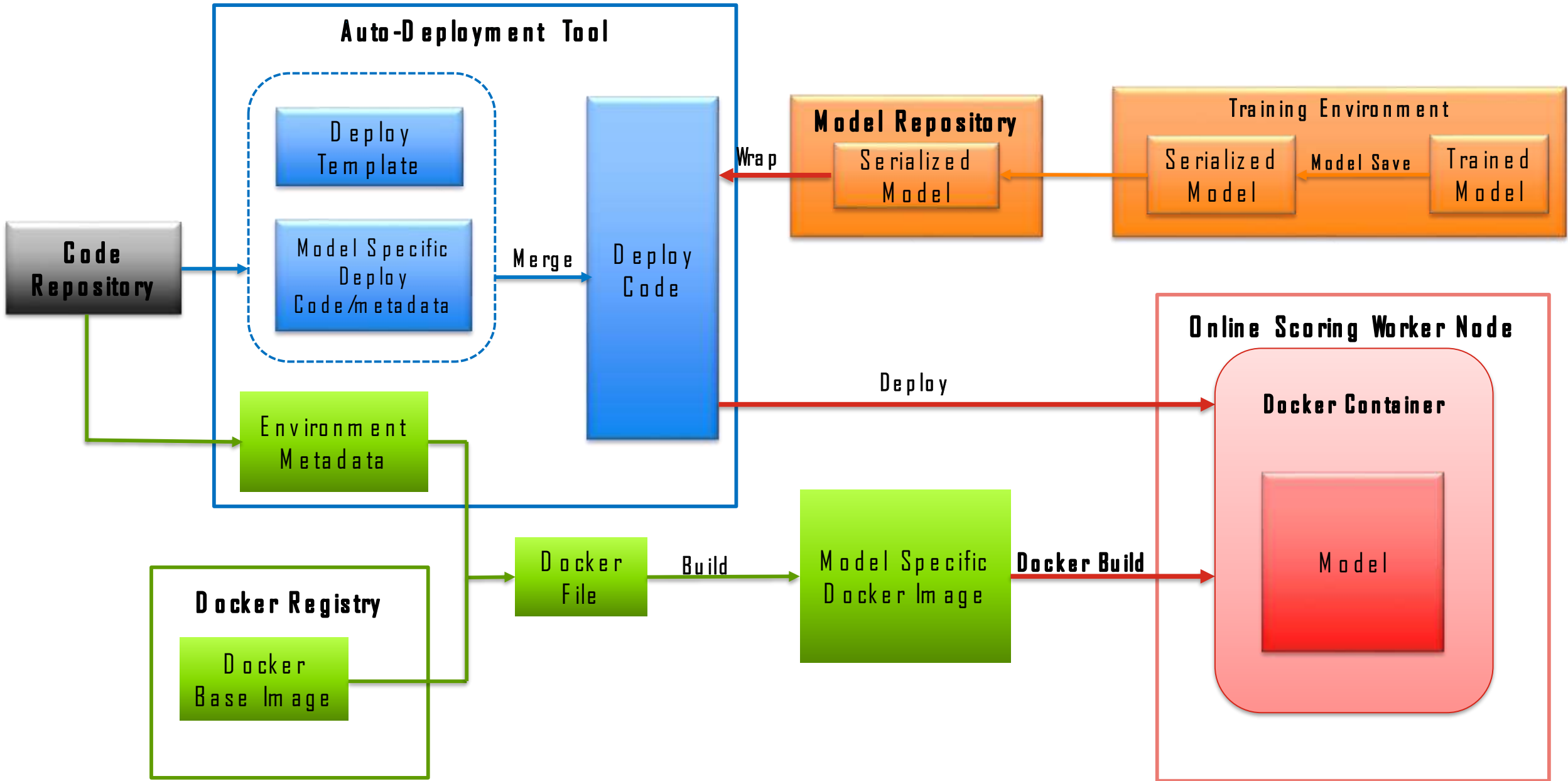


## Model Development & Training & Tuning

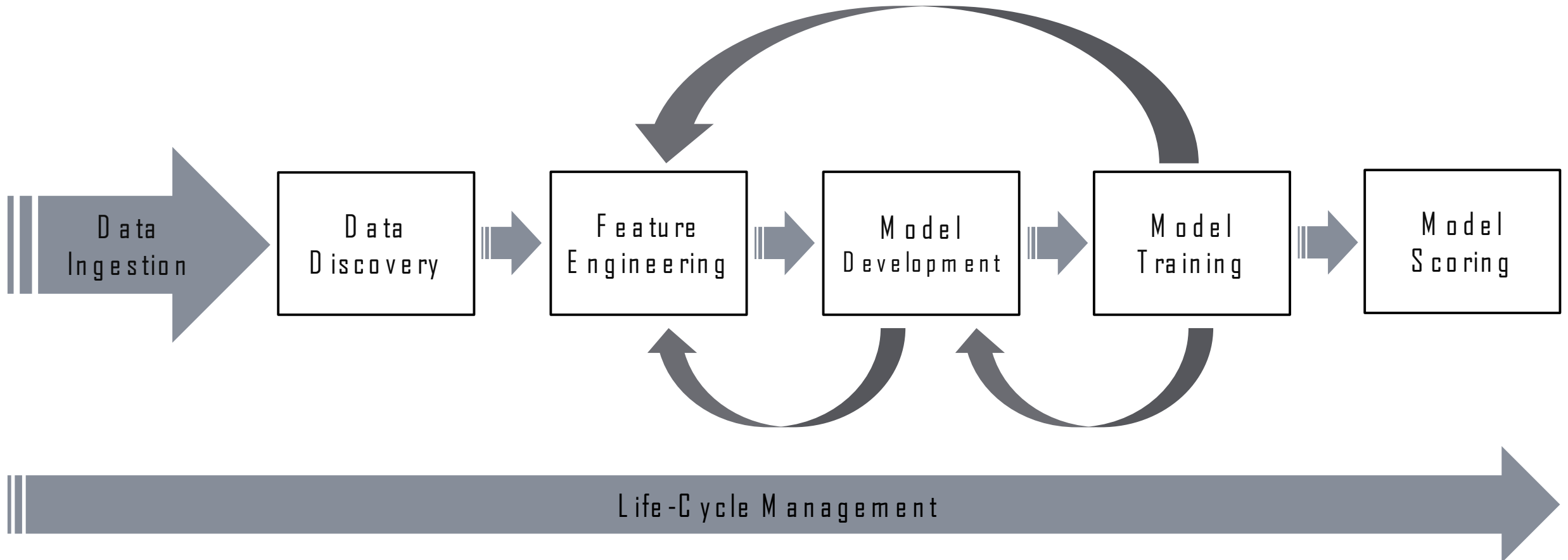




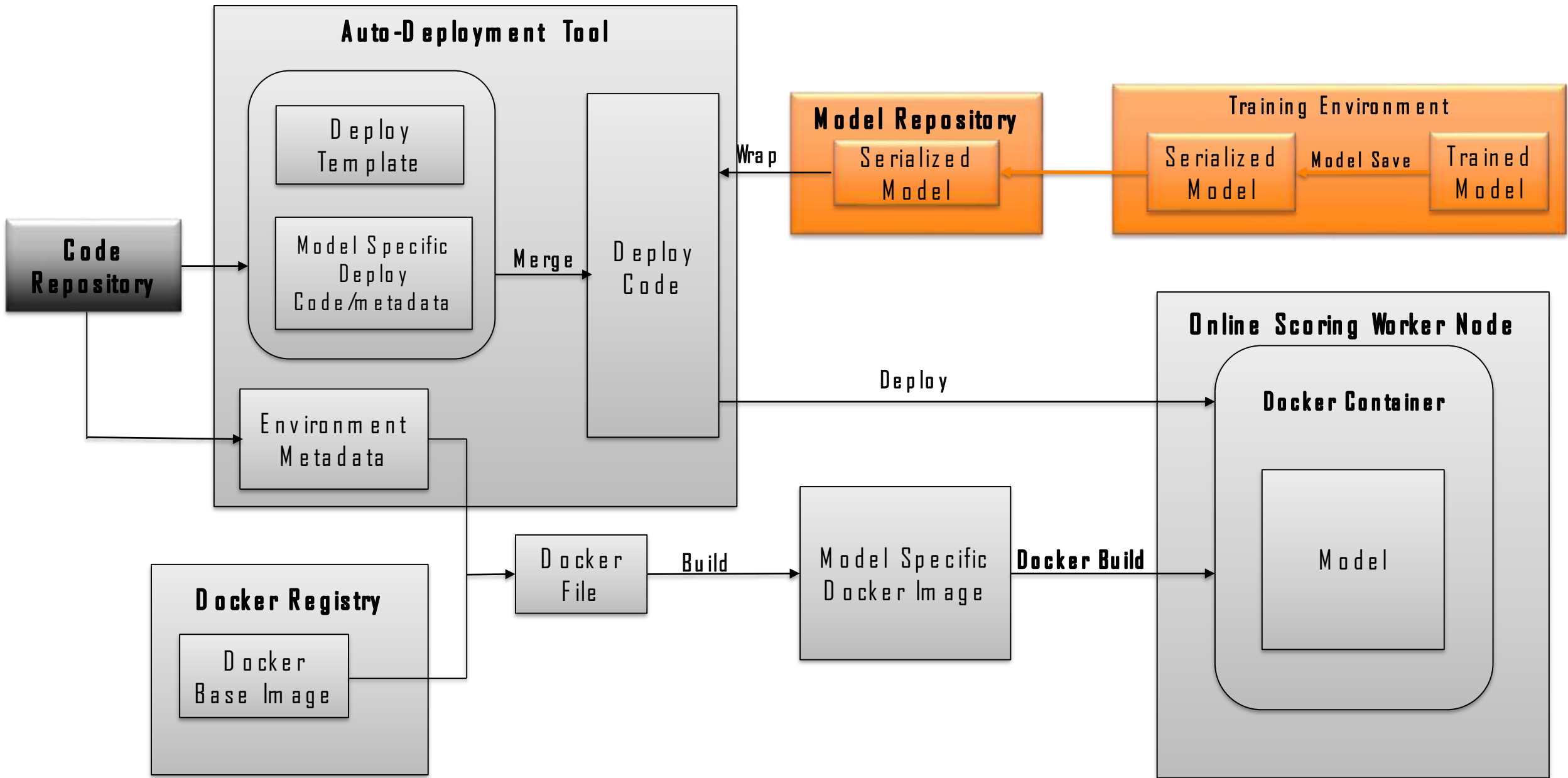
# Example: Online Scoring Service Deployment Diagram



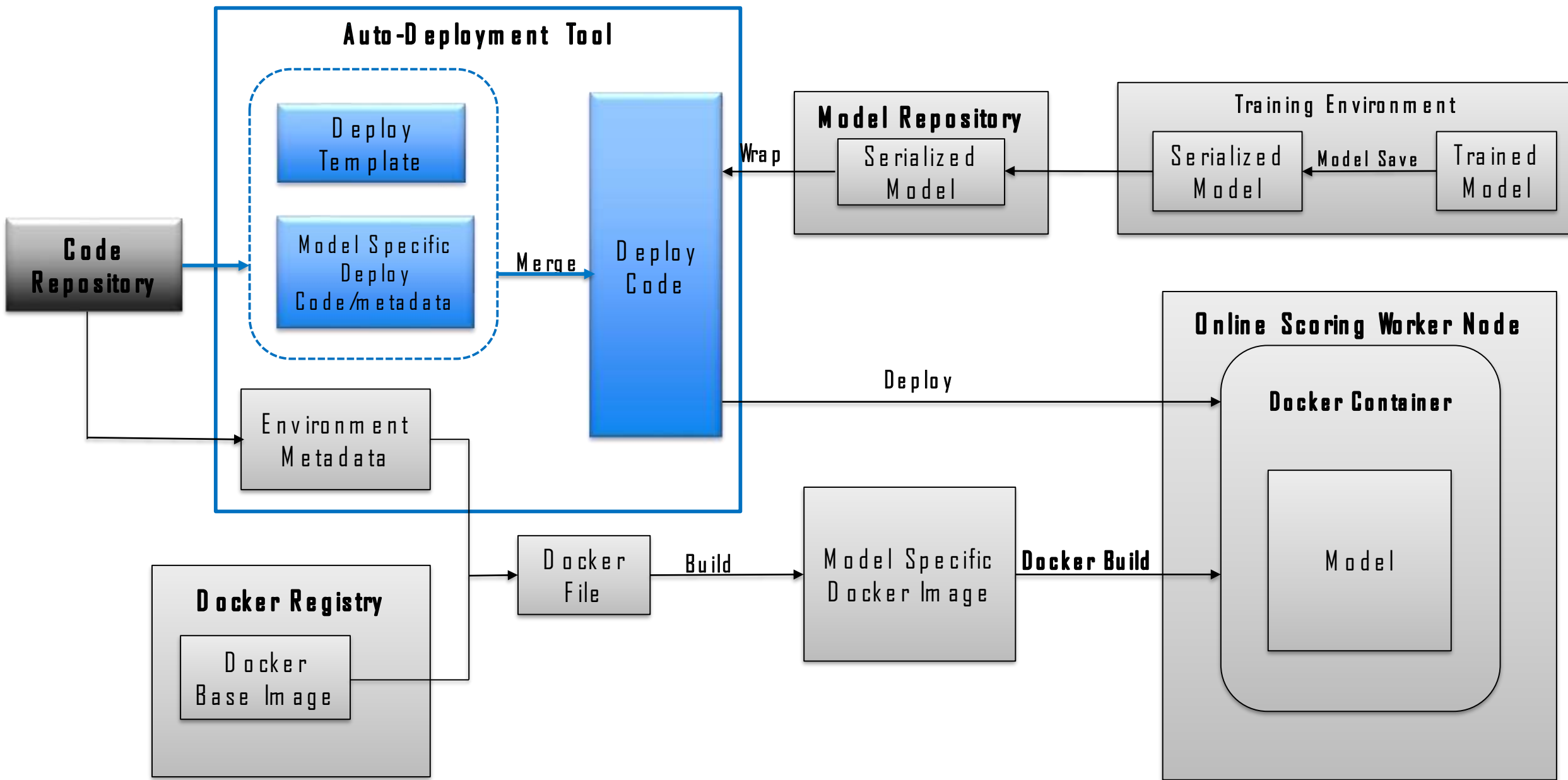




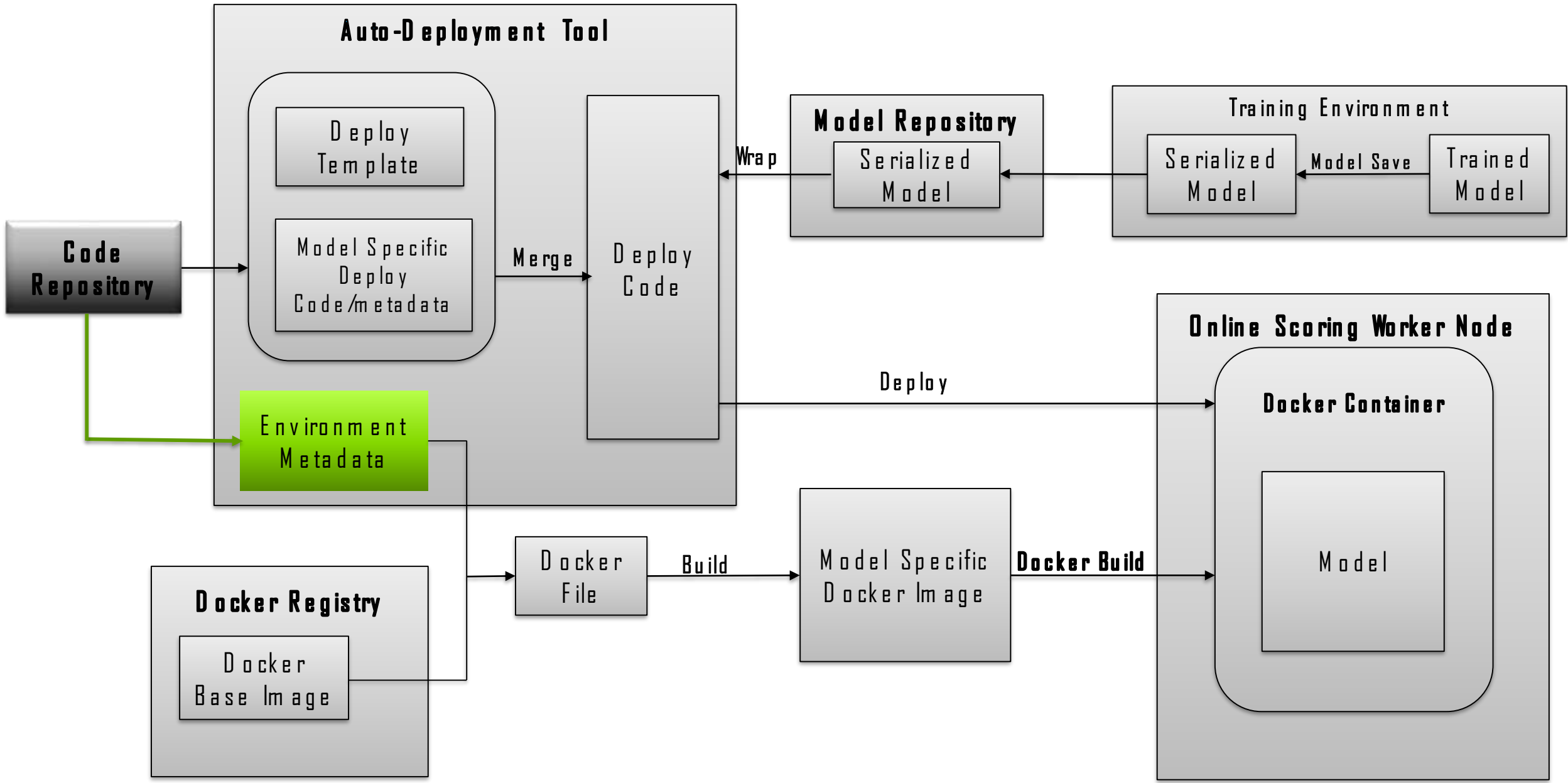
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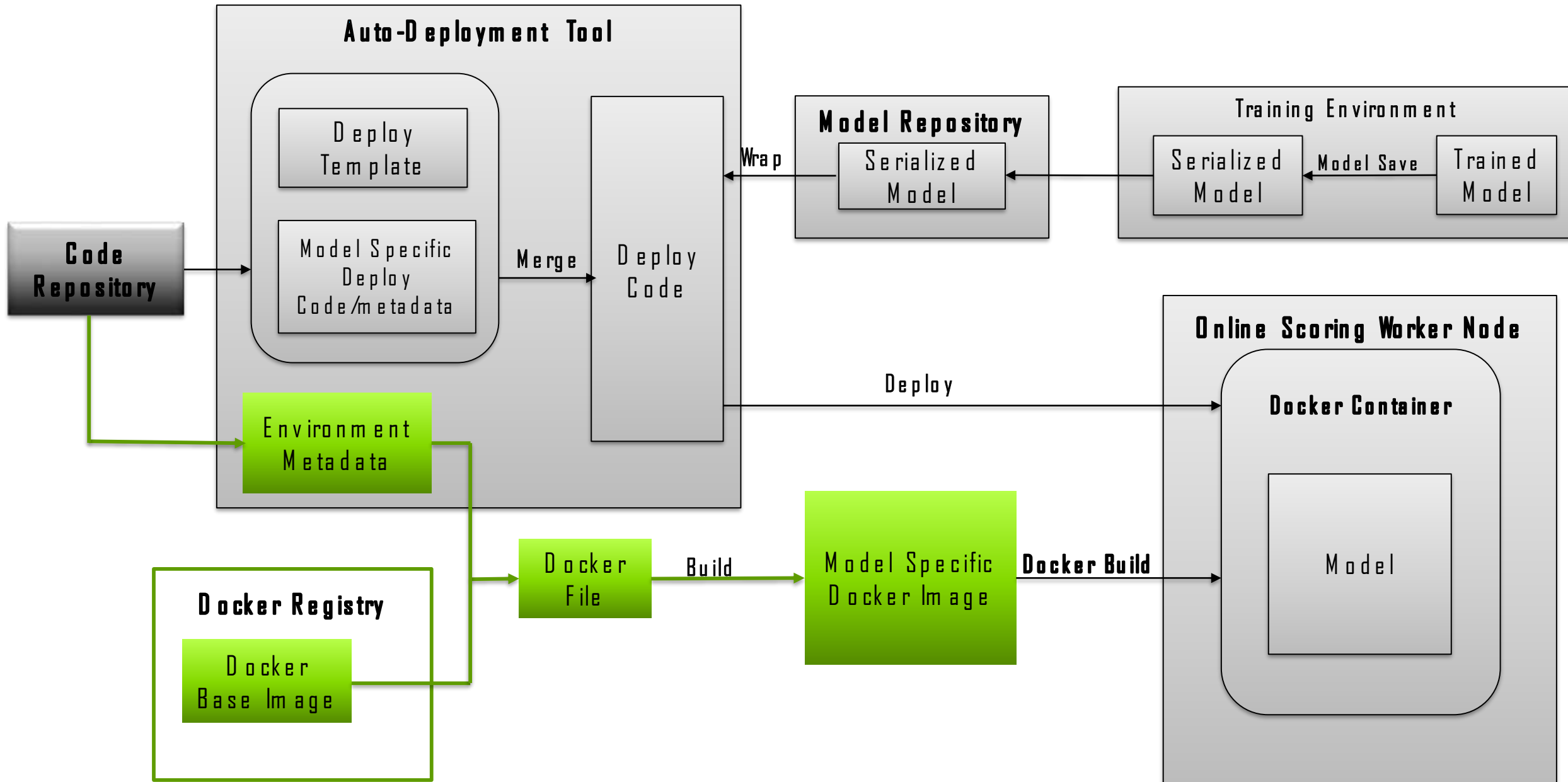
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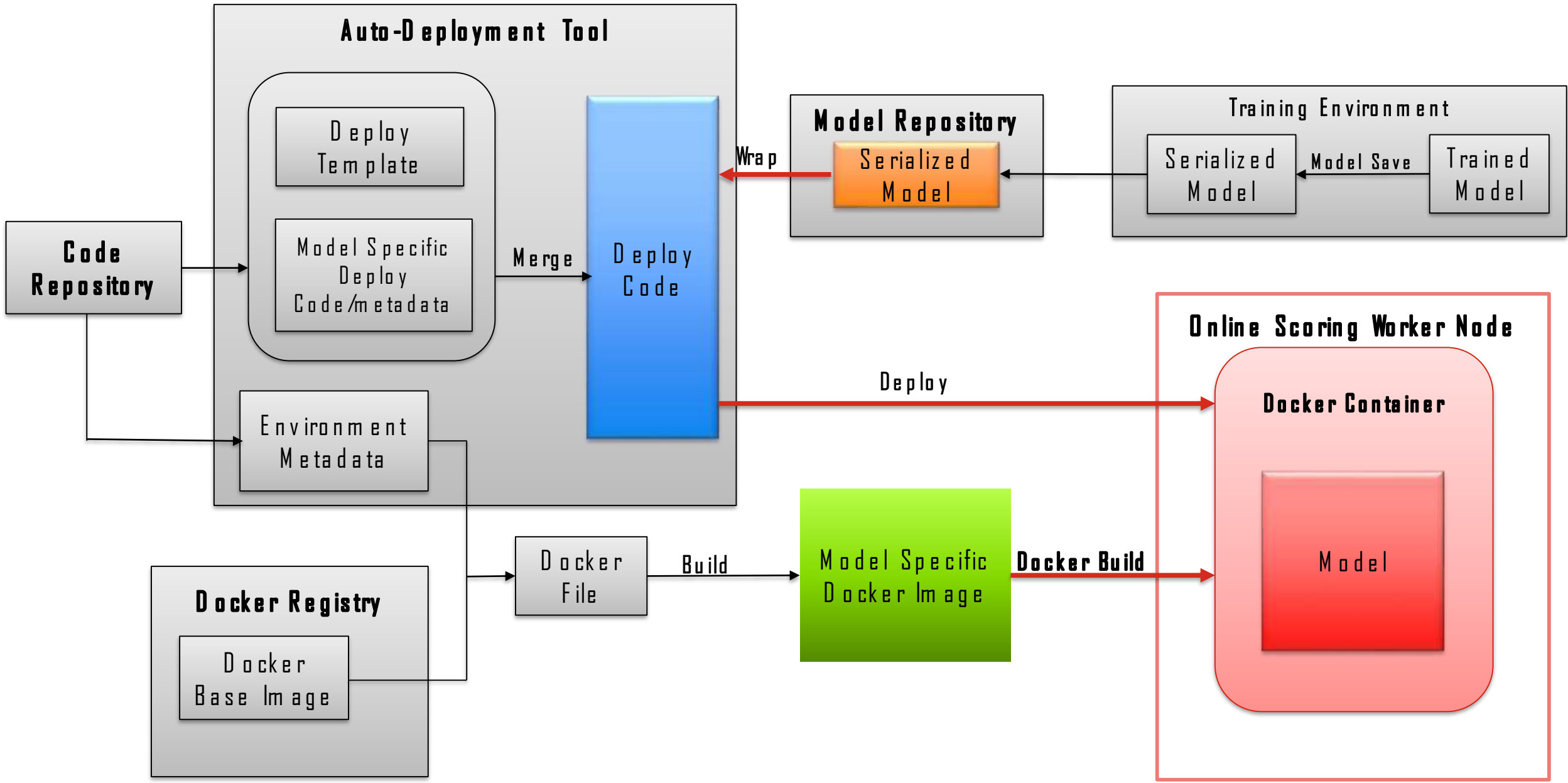
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# CONTAINERIZATION!



Environment in Container



# Thank you



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