

MACHINE LEARNING PLATFORM FOR PRE-PROCESS AND POST PROCESS INSPECTION

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Quality Control at Boeing

- Manufacturing processes require inspection for quality control
- One example is the appearance of air pockets in an additive manufacturing process
- With the ability to collect image data, Boeing is interested in automating quality control with modern computer vision techniques.

Platform Design

- The purpose of the ML Platform is to make machine-learning approaches accessible to domain scientists who are interested in apply machine learning, but do not have the resources to learn about the technologies in detail.
- With accessibility in mind, we designed the platform in a low-code, autoML manner.
- The user simply provides training data and a model type. The platform will automatically train and save all weights, metrics, and other model information in a reusable output directory.

Platform Implementation

- The platform is designed with a standard Model-View-Controller (MVC) UI architecture.
- Application is written in PyTorch Lightning and PyQt.
- The model logic contains scripts for training and performing inference with a suite of machine learning models.
- The view logic contains logic for visually rendering the UI to the screen. • The controller logic coordinates all interactions between the model logic and
- the view logic, . The controller has responsibilities such as responding to user events and writing to the database.
- The primary advantage of the MVC architecture is that each core component of the application is decoupled from the others. This modularity is helpful organization for developers who may seek to extend the application in the future.

Platform Requirements

To use application, we have provided Boeing with a private GitHub. Included in the GitHub are instructions to setup the application environment and details about directory formatting or raw data.

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Model Training Panel:

- Selection among suite of model types.
- Hyperparameter Configuration
- GPU Support
- Tensorboard Logging

Model Prediction Panel: Selection among trained models.

- Rename and Delete
- Functionality
- Image Explorer for Visualization of Model Results

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	Model Training				
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	1) Upload Training Data				
\bigcirc	Uploaded Directory: NEU Metal Surface Defects Data				
$\mathcal{I}_{\mathcal{S}}$		_			
	2) Select Model Type				
	Image_Classification				
	Default Model Hyperparameters				
	length 224width 224gpus 0max_epochs 10				
	batch_size 32num_workers 2 lr 1e-3				
	Ontionals Madifield means a	-			
	Optional: Modify Hyperparameters				
	To override the default hyperparameters, please provide a sequence of flags.				
	Model Hyperparameters:				
	Trainer Hyperparameters:gpus 1				
		-			
	3) Submit Training Run				
	Experiment Name: Experiment 1				
	Train Open Dashboard				
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	Madel Duadiation				
ŝ	Model Prediction				
É					
•	1) Upload Image Directory				
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\mathbb{O}	Uploaded Directory: Test_Images				
-0		_			
	2) Select Trained Model				
	Image_Classification				
	Experiment 1				
	Experiment 2				
	Experiment 3				
	Current Selections Experiment 2				
	Current Selection: Experiment 3 Refresh				
		_			
	3) View Model Results				
	Open Image Explorer				



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Platform Features

Image Explorer Panel:

- Resizable window for framing any size input • Image Navigation with arrow keyboard
- shortcuts Report Generation containing information of trained model metrics and model predictions.



Future Work and Extensions

 Client-Server Application Architecture packages such as Ray.

• Database Infrastructure.

here as well including Airflow and Prefect.

• Application Scale.

Each component of the application may be containerized and managed through Kubernetes for greater modularity, maintainability, and parallelization.



[2]

Platform Results

- Web UI streamlines deployment and accessibility to end user.
- Enables application to scale compute with distributed computing achieved through

At scale, the application should have a dedicated data warehouse, as well as an automated means to access and transform the raw data into training data. Large ecosystem of tools exist