MATLAB EXPO

Automated Optical Inspection and Defect Detection with Deep Learning

Harshita Bhurat Product Manager – Image Processing and Computer Vision



What is Automated Optical Inspection?

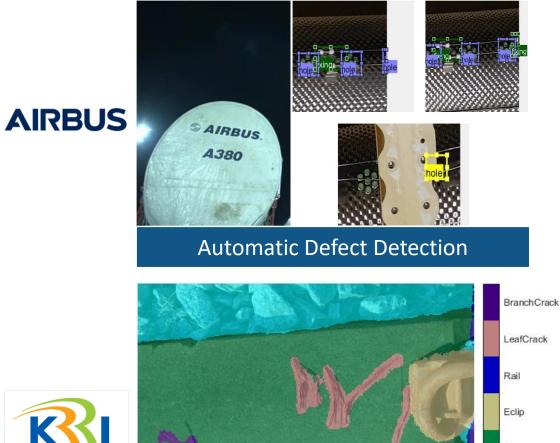
"Automated optical inspection is the **image-based** or **visual inspection** of manufacturing parts where a camera scans the device under test for both **failures** and **quality defects**"

Automated Defect Detection Machine Vision Visual Inspection Automated Inspection





Customer References



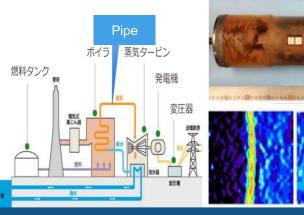


LeafCrack Eclip Sleeper Ballast

Defect Detection in Railway Components



Visual Inspection of Automotive Parts

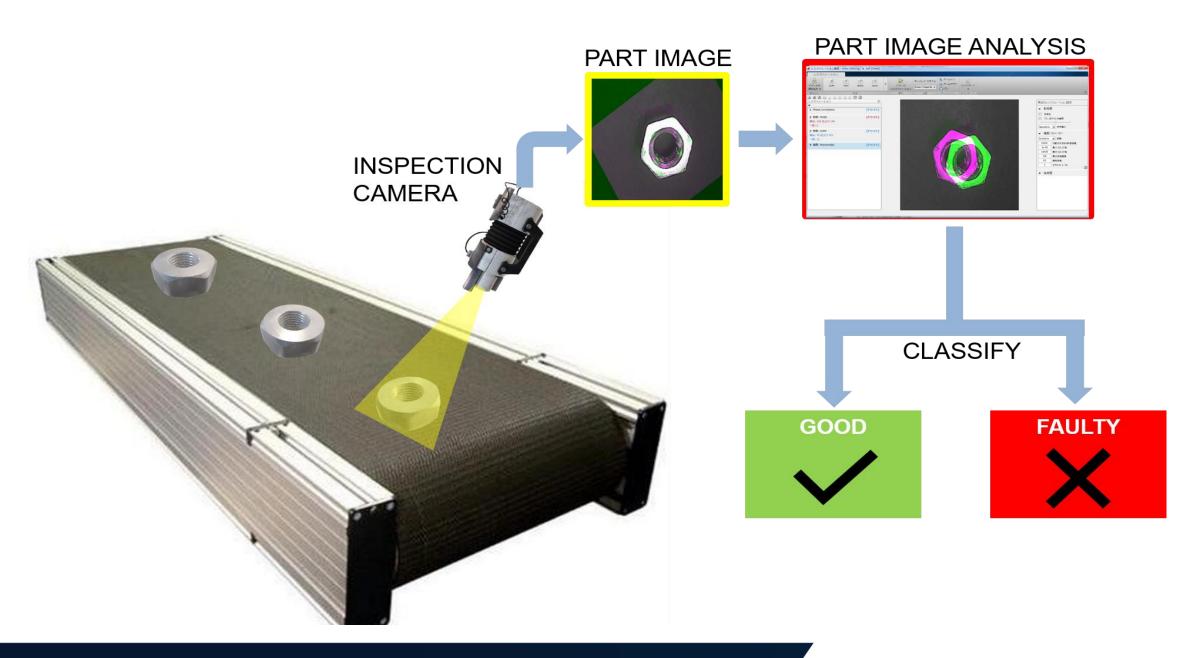




Assess Pipe Weld Damage at Power Plants

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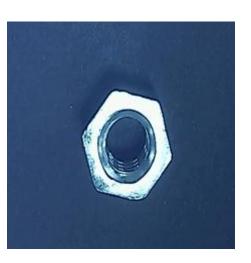




Can you find the defective hex nut?



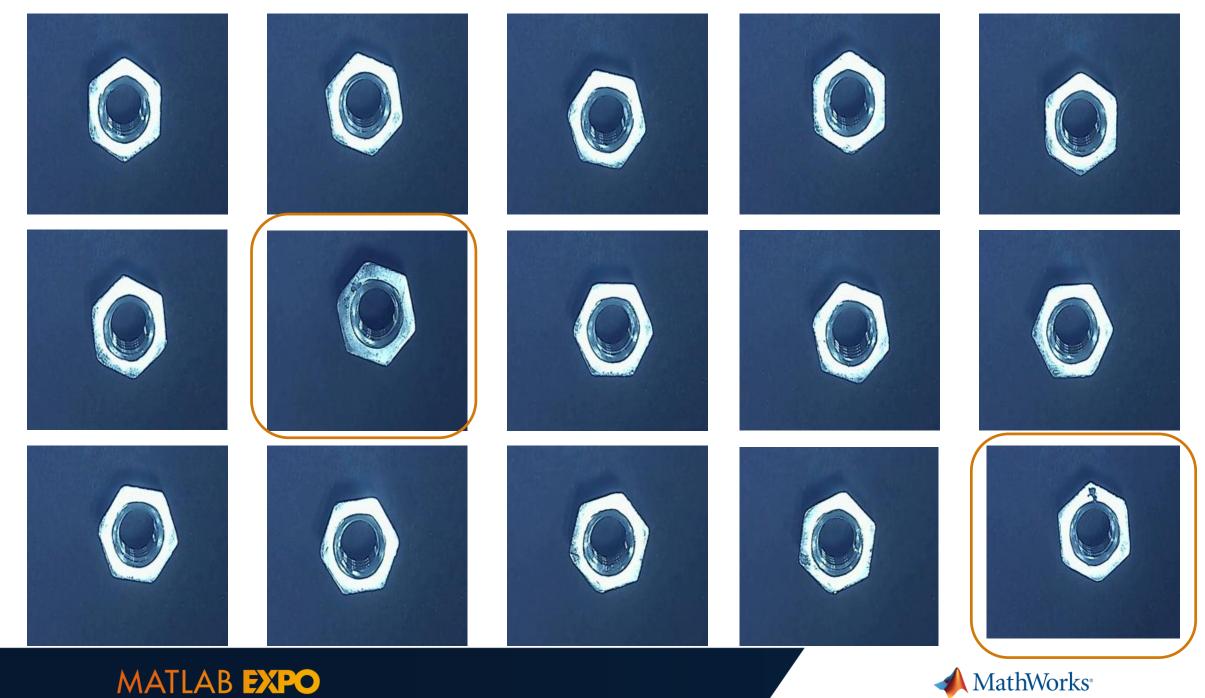






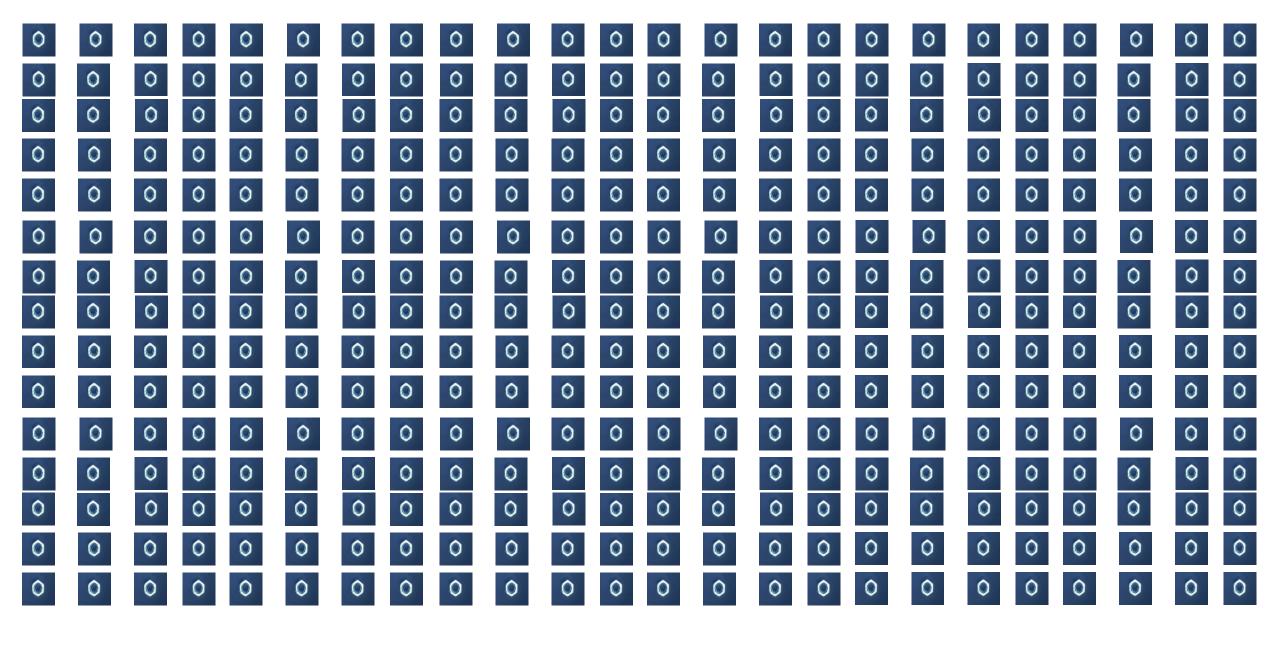




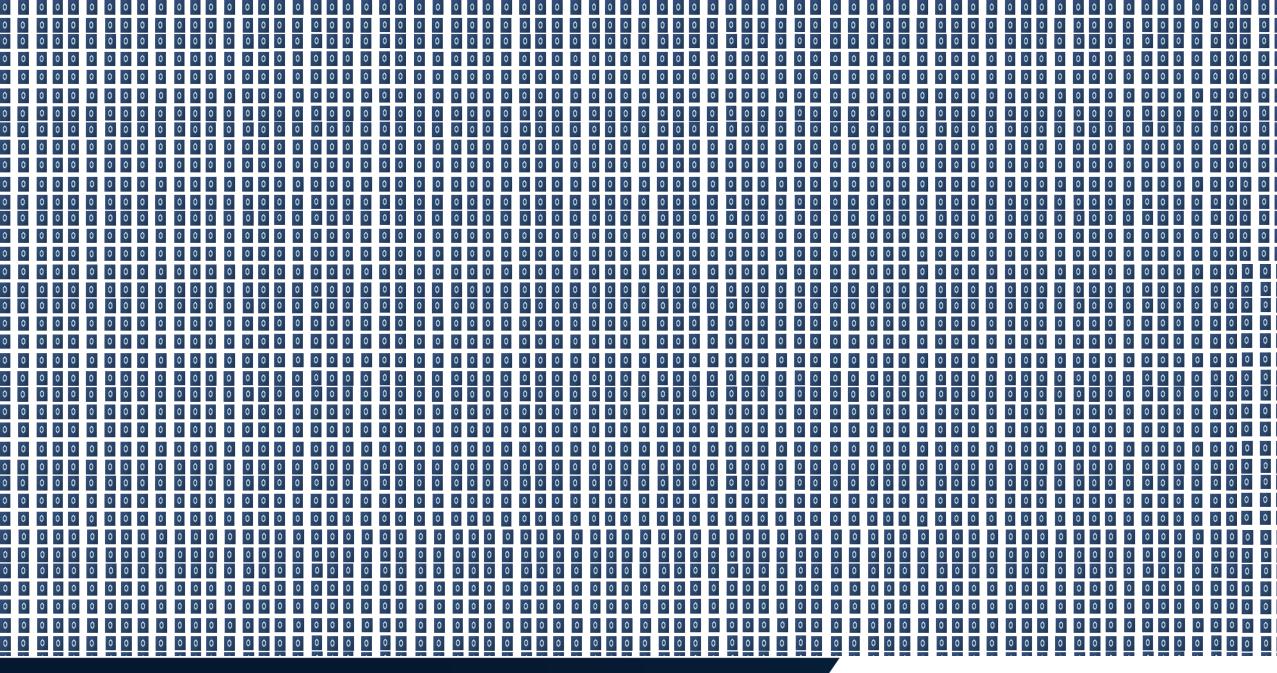


MATLAB **EXPO**











Finding Defective Hex Nuts

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Defective



1.bmp



3.bmp



2.bmp

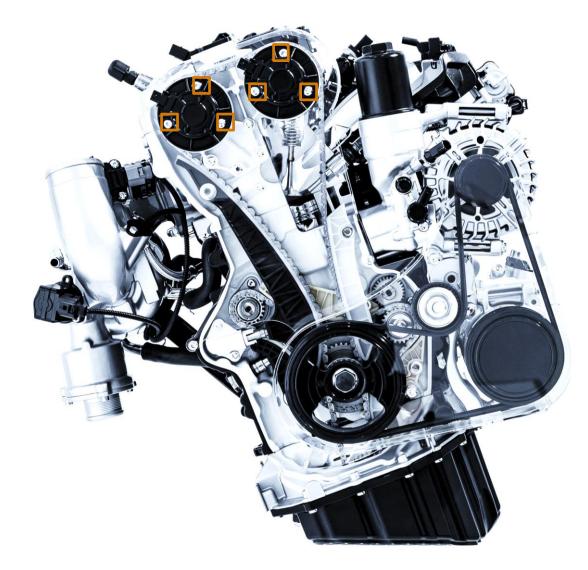


4.bmp





Detecting Parts

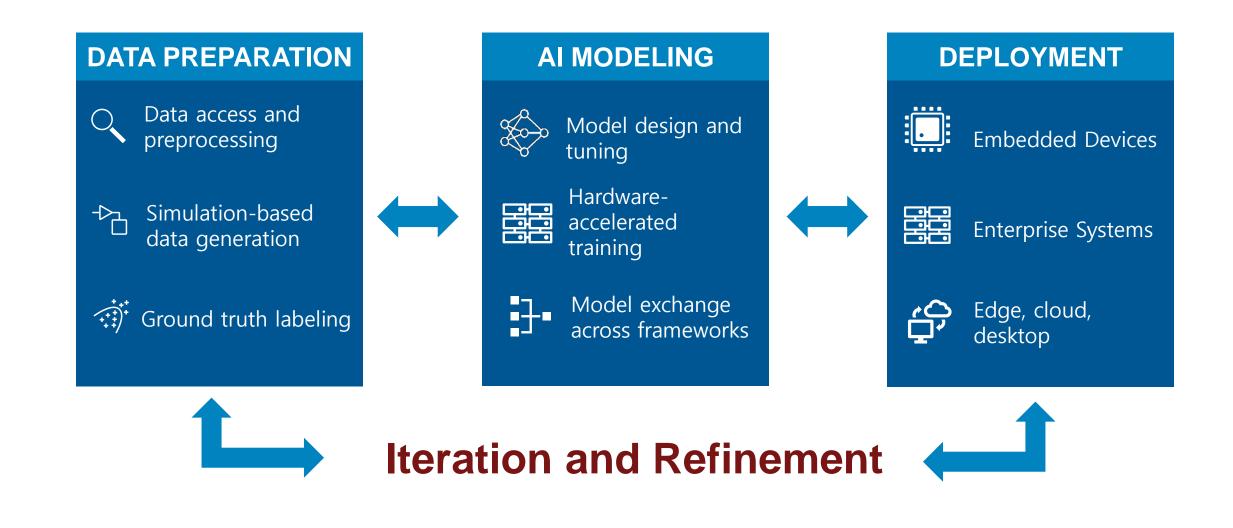






Defect Detection Workflow

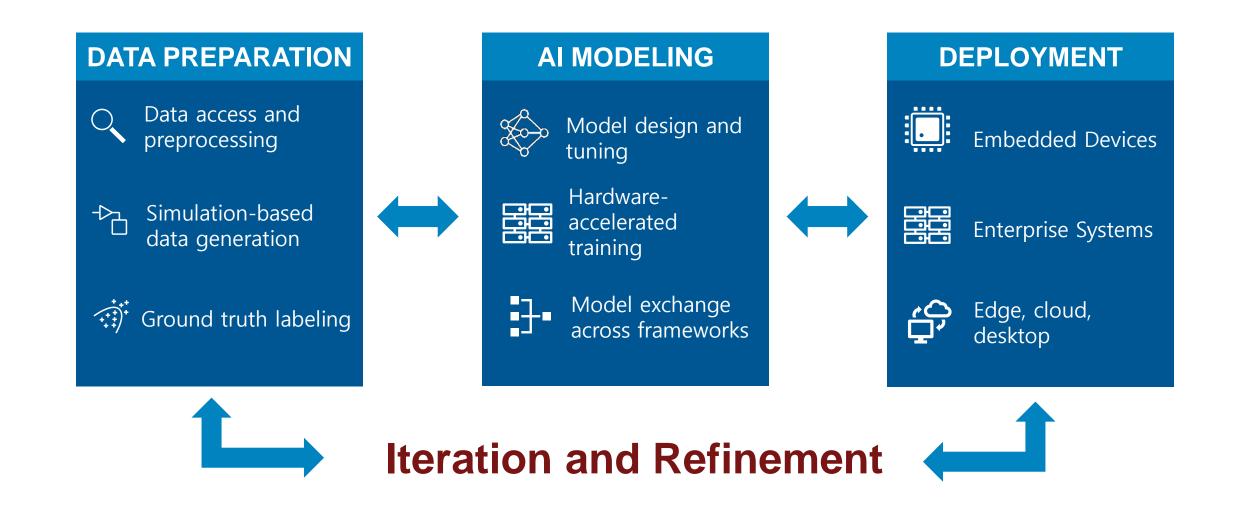
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Defect Detection Workflow

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Data Access and Preprocessing – Common Challenges

How do I access large data that might not fit in memory?

How do I preprocess data and get the right features?

How do I label my data faster?

What if I have an imbalanced dataset or don't have enough data?





Data Access and Preprocessing – Common Challenges

How do I access large data that might not fit in memory?





How do I load and access large amounts of data?

Datastores	Tall Arrays		<section-header><section-header><text></text></section-header></section-header>		
Loads image/signal data into memory as and when needed >> imageDatastore >> audioDatastore >> fileDatastore	Work with out-of- memory numeric data – Train deep neural networks for numeric arrays				
Name A Folder Dataset bed bid bid cat Custom Datastores also On available on available seven sheila six	Mile Tights by North and Yare, 1977. 2011 Image: A start of the start	Image: bit of the set of	Image: constraint of the experiment of the experim		Each red box is a 1024-by-1024 tile in the file. L1's dimensions = 29,600 x 46,000 L2's dimensions = 14,800 x 23,000 L3's dimensions = 7,500 x 12,000 Rows = 29600 Columns = 46000 TileSizeIntrinsic = [1024 1024] ResolutionLevelSizes = [29600 46000 14800 23000 7500 12000] CoarsestLevel = 3 FinestLevel = 1 PixelSpacings = [1 1; 2 2; 3.947





Data Access and Preprocessing – Common Challenges

How do I preprocess data and get the right features?





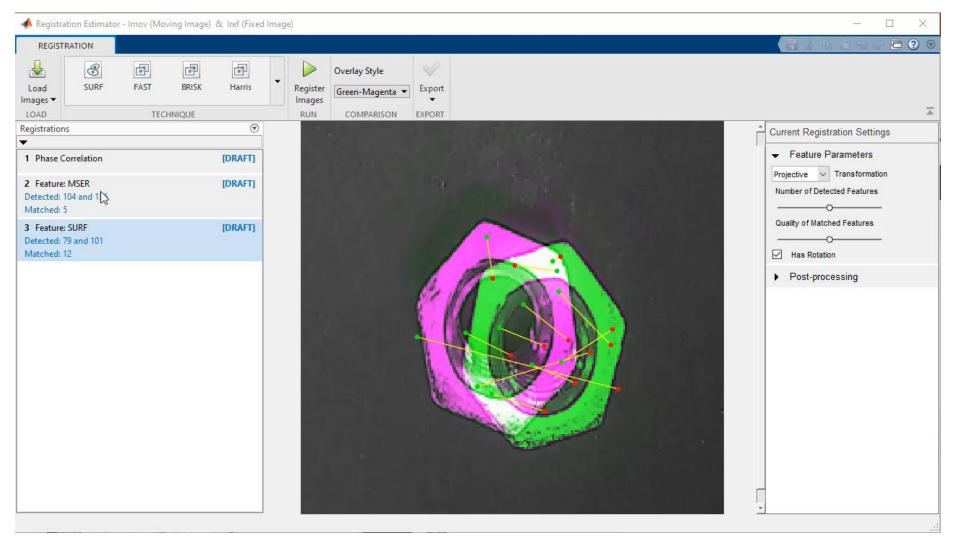
Data Access and Preprocessing – Common Challenges

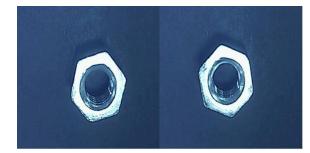
How do I preprocess data and get the right features?

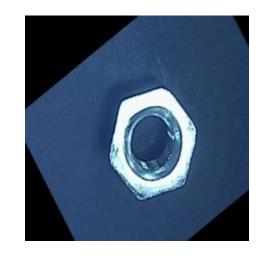




Pre-processing Data – Registration Estimator App



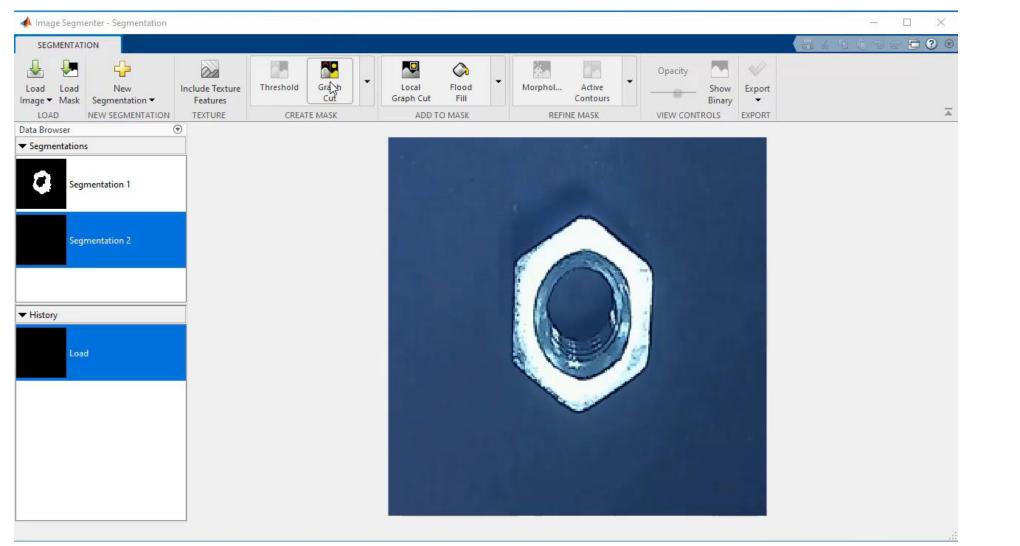




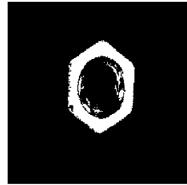
MATLAB EXPO



Pre-processing Data – Image Segmenter App



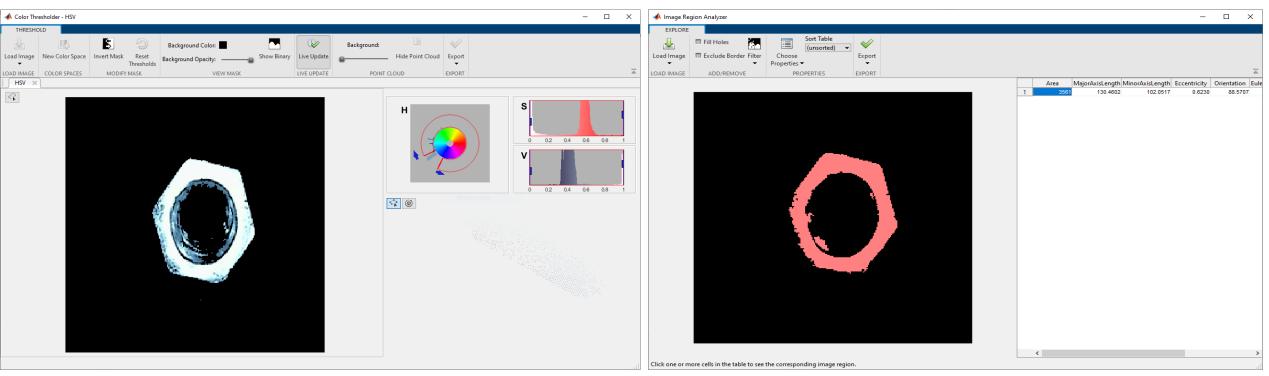








Preprocessing Data - Apps



Color Thresholder

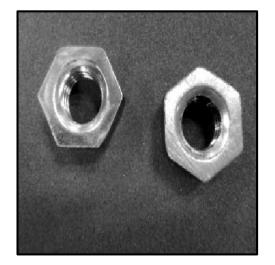
Image Region Analyzer



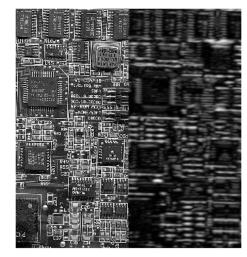


Pre-processing Data – Built-in Algorithms

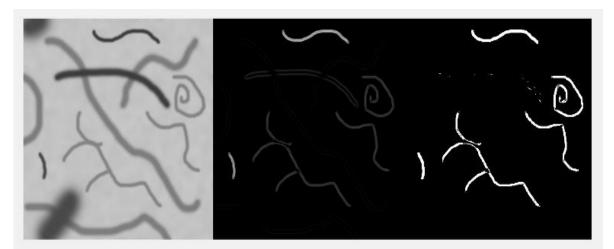
imadjust



imgaborfilt



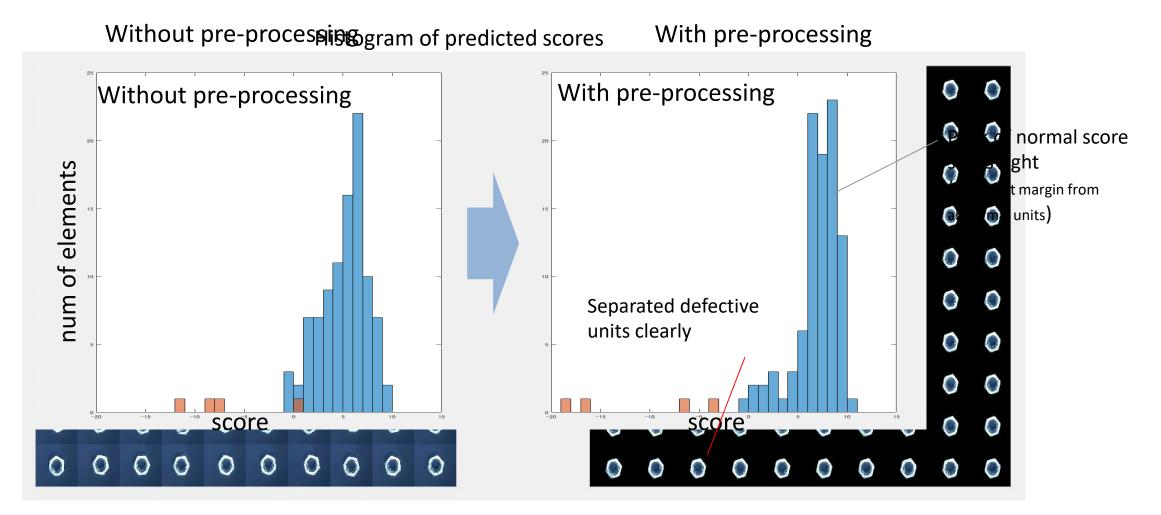
fibermetric







Defect detection using AlexNet: Results with preprocessing







Data Access and Preprocessing – Common Challenges

How do I label my data faster?





Data Preprocessing - Labeling

📣 MATLAB R2020a

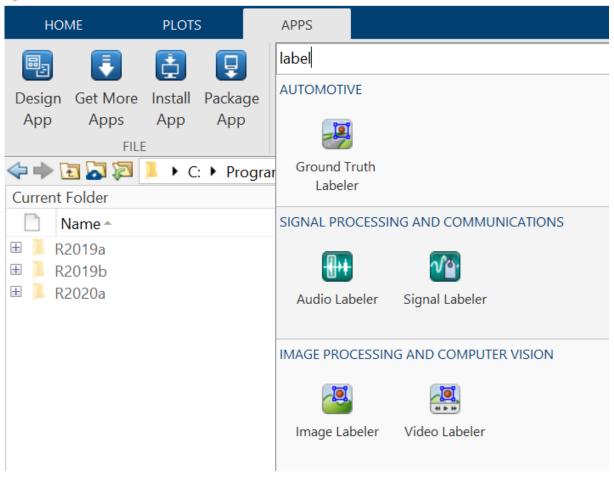


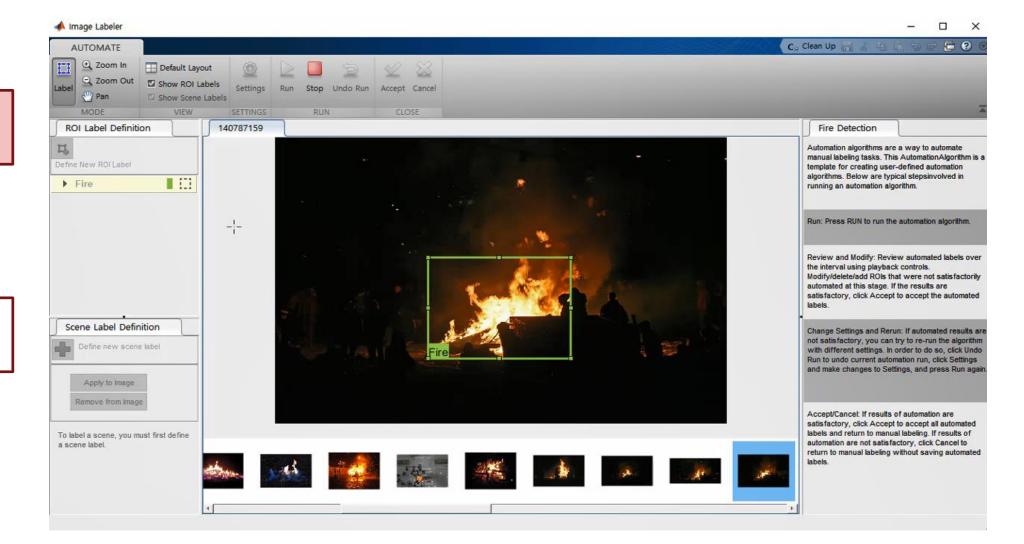




Image & Video Labeler

Image Labeler + Video labeler

Big-Image Labeler







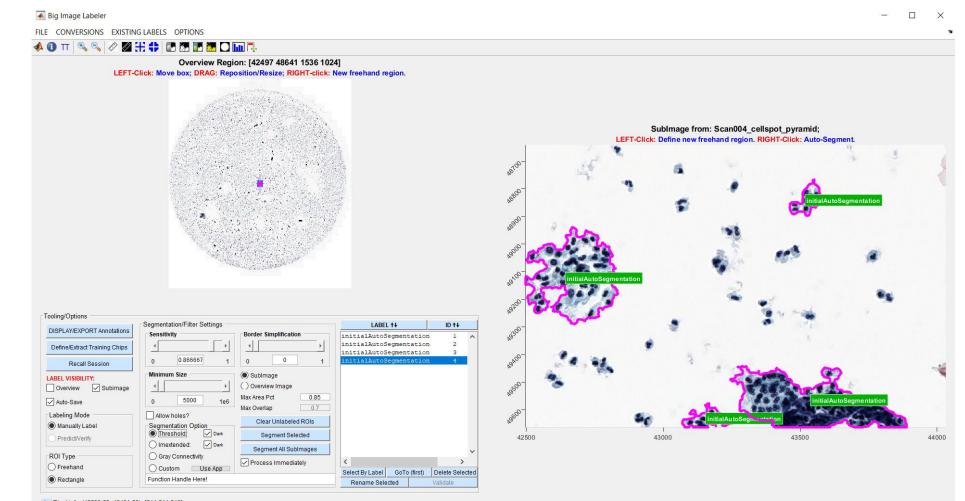
Big Image Labeler

Image Labeler

+ Video labeler

Big-Image

Labeler



14 Pixel info: (43690.68, 49491.69) [244 244 245]



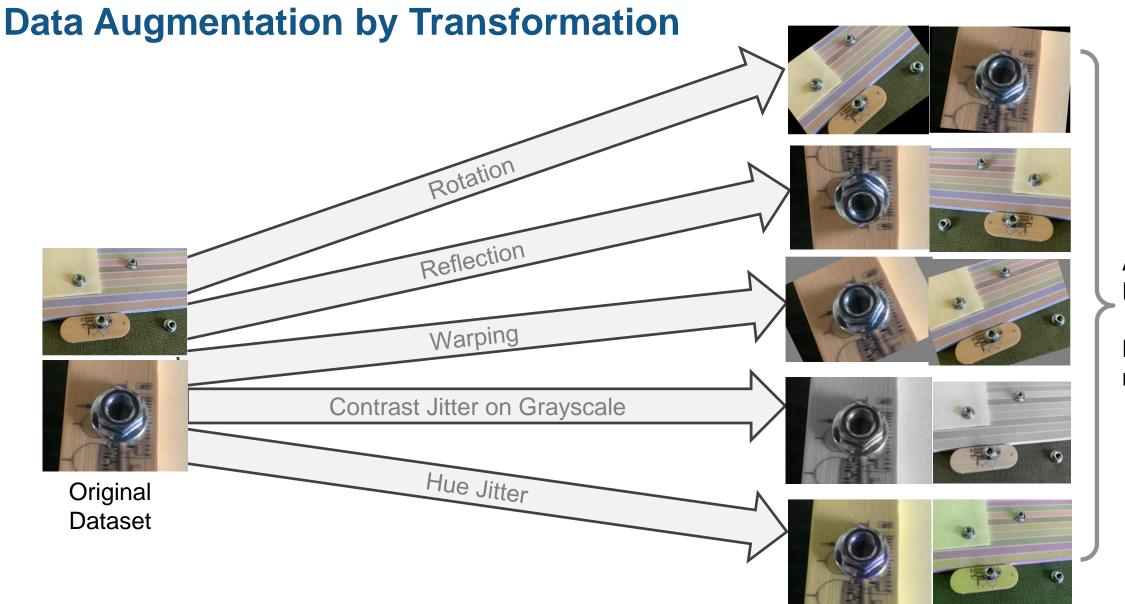


Data Access and Preprocessing – Common Challenges

What if I have an imbalanced dataset or don't have enough data?







Augmented Dataset

N times as much data

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Data Augmentation : Generative Adversarial Networks (GANs)

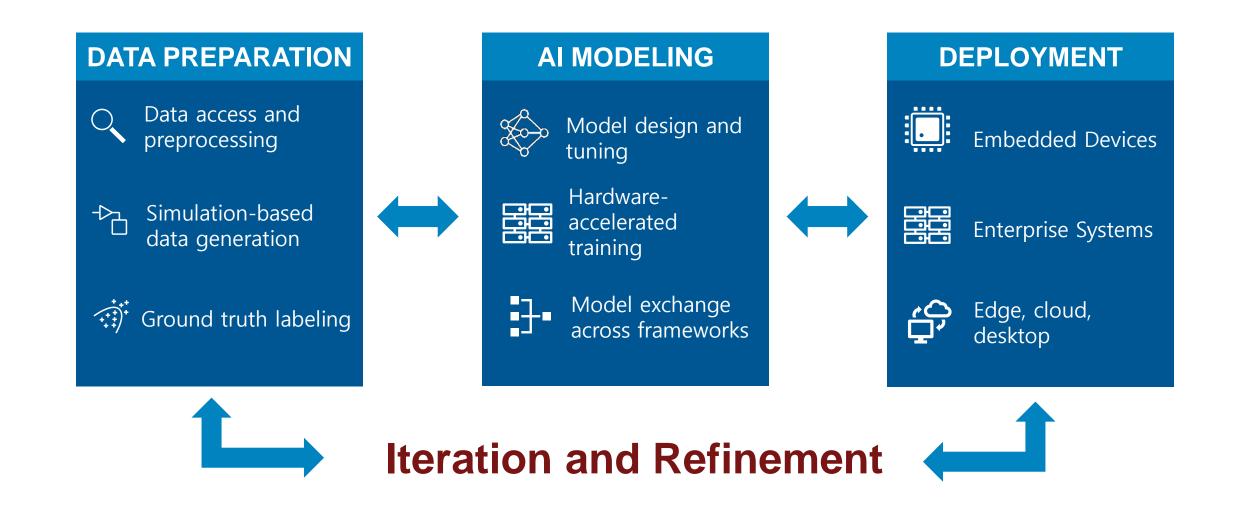






Defect Detection Workflow

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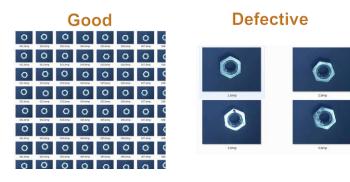




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Deep Learning for Defect Detection

Deep learning for Classification



Deep Learning for Object Detection







Deep Learning for Defect Detection – Multiple techniques

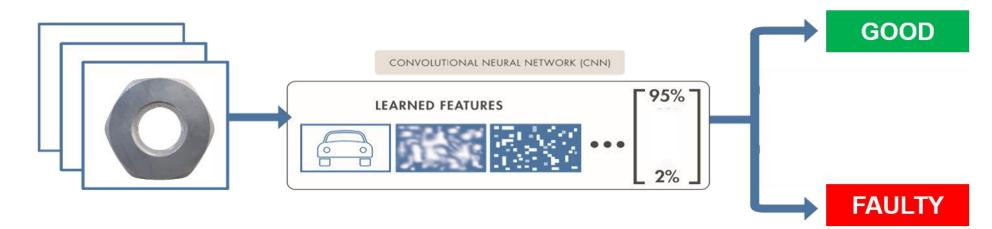
Deep learning for Classification



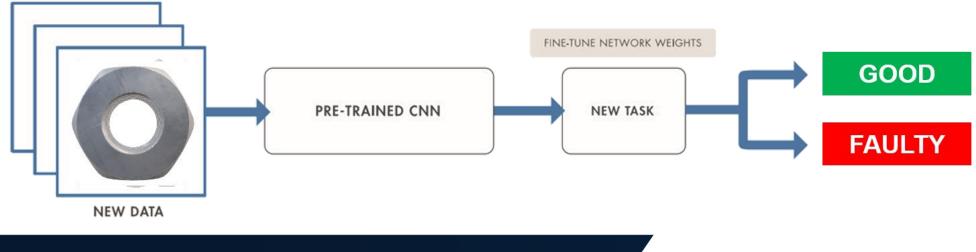


Two Approaches for Deep Learning

1. Train a deep neural network from scratch



2. Fine-tune a pre-trained model (transfer learning)







Train a Deep Neural Network from Scratch

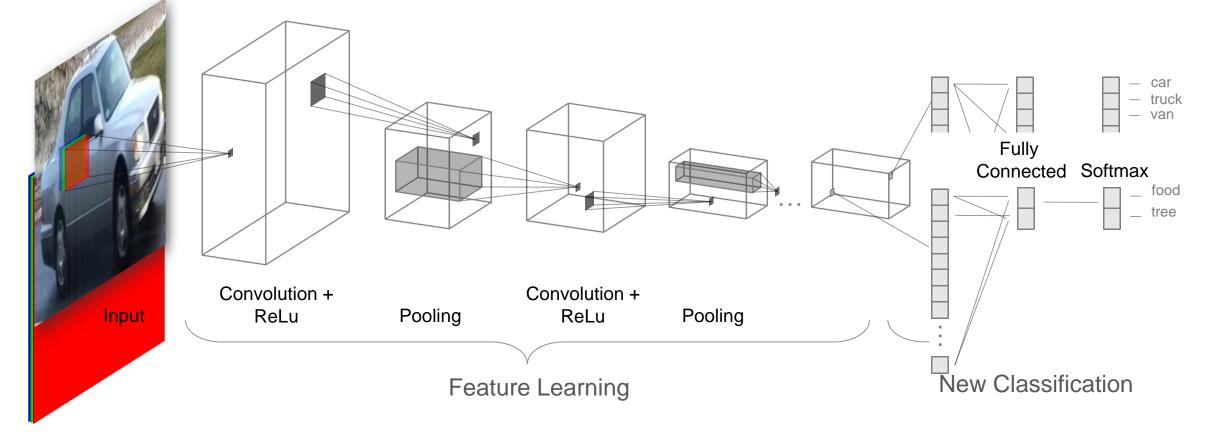
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DESIGNER	?
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LAYER LIBRARY	
OBJECT DETECTION	fullyConnectedLayer
OBJECT DETECTION	Name fc
yolov2ReorgLayer	InputSize auto OutputSize 10
yolov2TransformLayer	Weights []
OUTPUT	Bias []
III. softmaxLayer	WeightLearnRateFactor 1 WeightL2Factor 1
classificationLayer	BiasLearnRateFactor
regressionLayer	
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rcnnBoxRegressionLayer	
rpnClassificationLayer	OVERVIEW
pixelClassificationLayer	
dicePixelClassificationLayer	
yolov2OutputLayer	a
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Two approaches for Deep learning

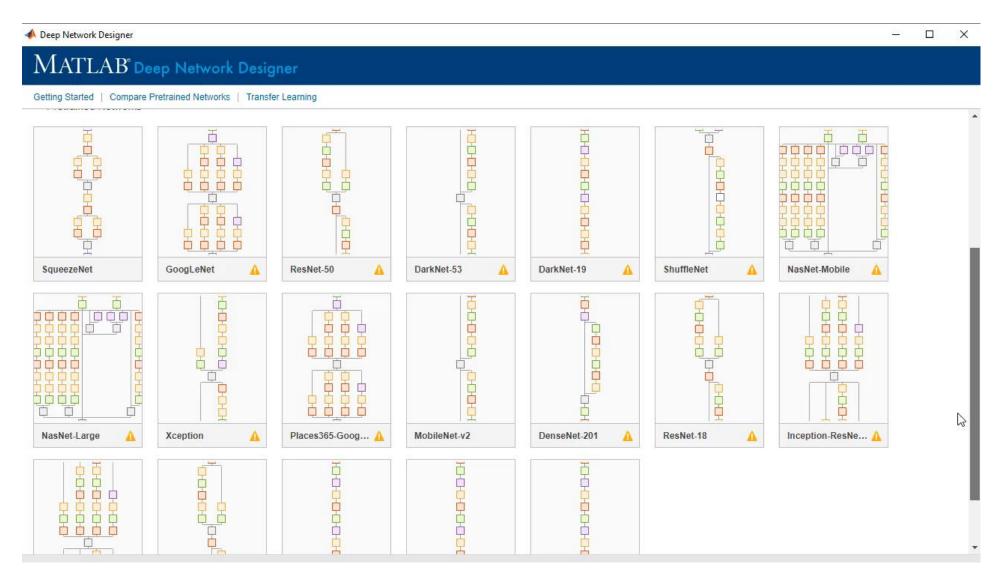
Approach 2. Fine-tune a pre-trained model (Transfer learning)







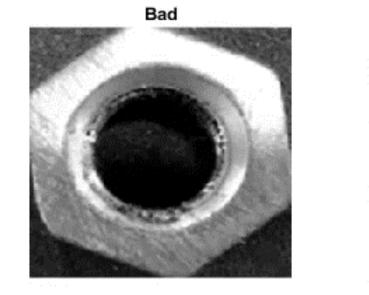
Fine-tune a Pre-trained Model (Transfer Learning)







Classification with Trained MobileNetV2

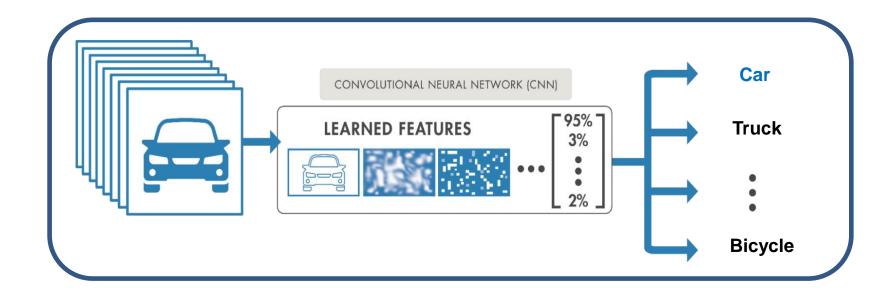


Good





Challenges with Deep Learning Models



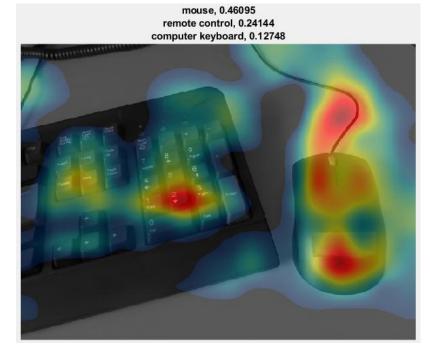
Explainable Al is required

- <u>Class Activation Mapping(CAM)</u>
- Grad-CAM

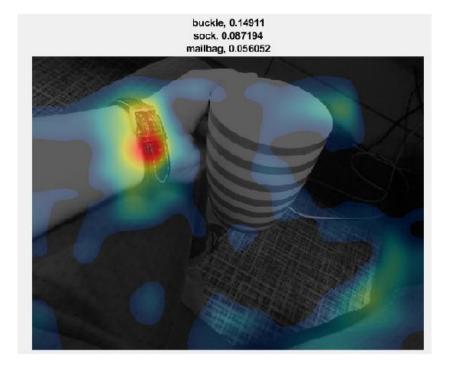




Class Activation Mapping to Investigate Network Predictions



Classified as "keyboard" due to the presence of the mouse

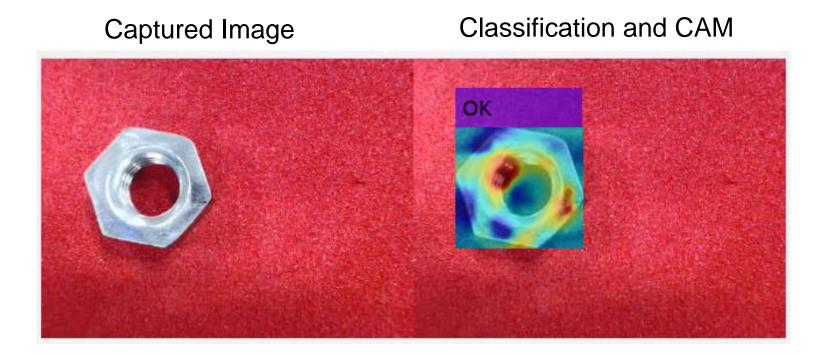


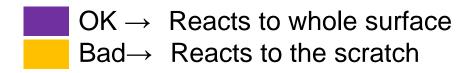
Incorrectly classified "coffee mug" as "buckle" due to the watch

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Visualization of Features with CAM









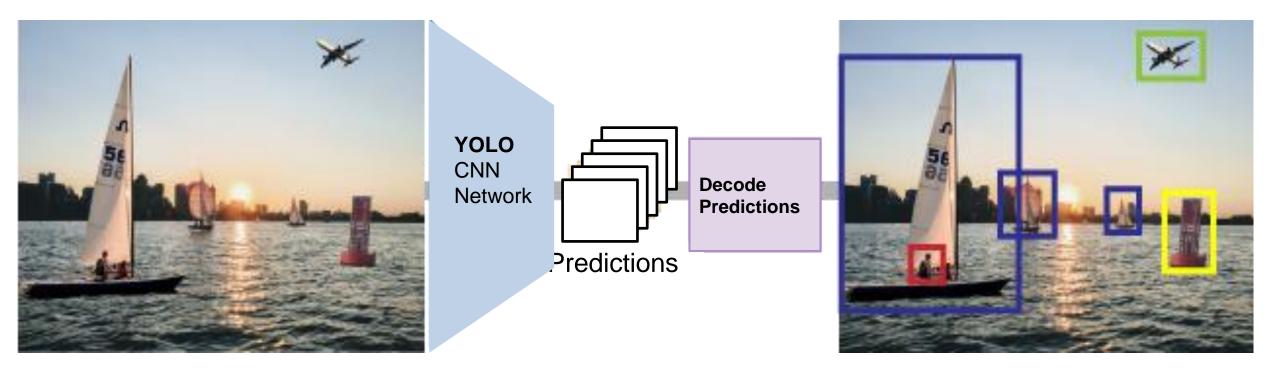
Deep Learning for Defect Detection

Deep Learning for Object Detection





Detecting Objects with You Only Look Once (YOLO) v2

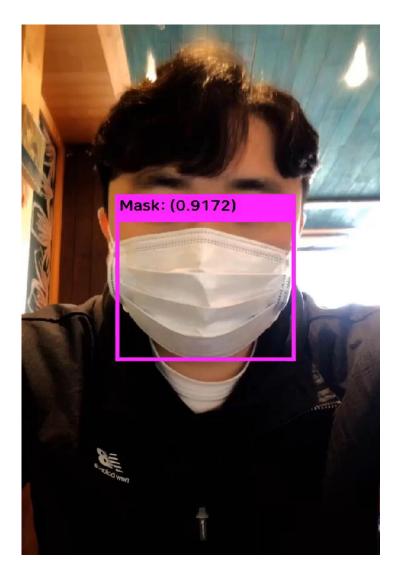


Build, test, and deploy a deep learning solution that can detect objects in images and video

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Mask Detection with YOLO v2







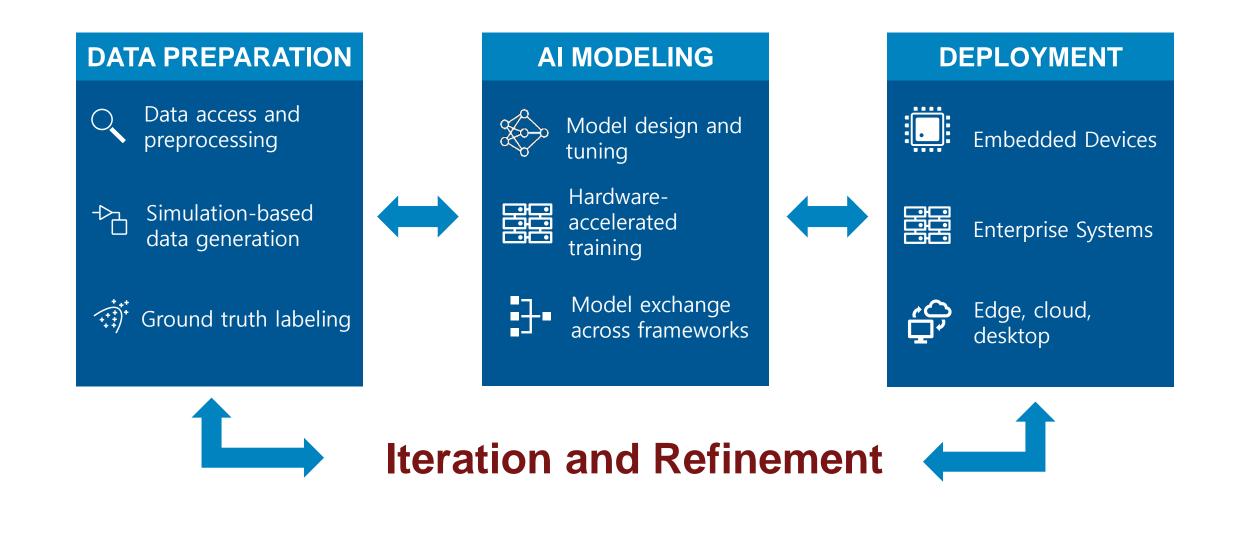
Experiment Manager

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14											





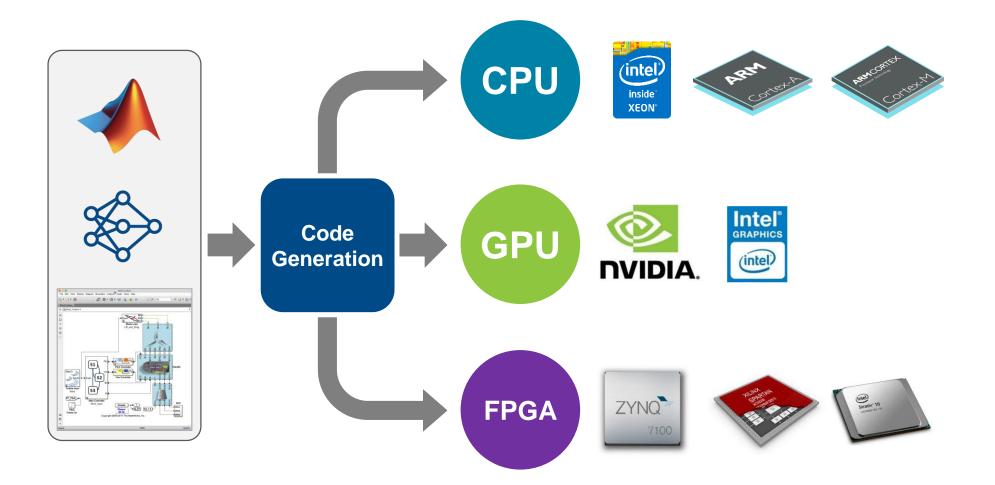
Defect Detection Workflow







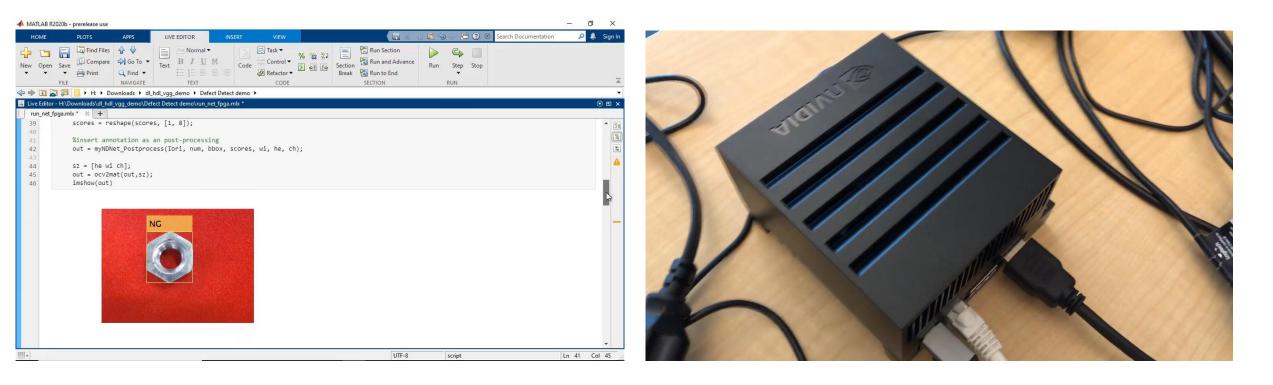
Deploy to Any Processor with Best-in-class Performance







Deploy to Hardware



Deploy defect detection algorithms from MATLAB to ZCU102 board from Xilinx

Deploy defect detection algorithms from MATLAB to Jetson AGX Xavier





Deploy to Hardware

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	root	rt	0	0	Θ	0		0.0	0.0	0:00.14 migration/0	
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	root	20	0	0	0	0		0.0	0.0	0:00.28 ksoftirqd/1	
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19	root	rt	0	0	Θ	0		0.0	0.0	0:00.14 migration/2 0:00.22 ksoftirqd/2	
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Defect detection deployed on

ARM Cortex-A microprocessor

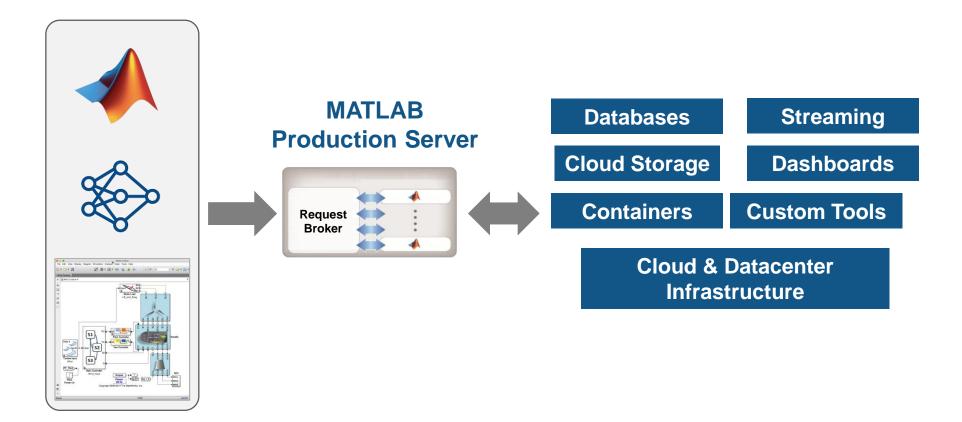
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Resources:

- **Deploying Deep Neural Networks to GPUs and CPUs Using MATLAB Coder and GPU Coder**
- Using GPU Coder to Prototype and Deploy on **NVIDIA Drive, Jetson**
- **Real-Time Object Detection with YOLO v2 Using GPU Coder**
- Image Classification on ARM CPU: SqueezeNet on Raspberry Pi
- **Deep Learning on an Intel Processor with MKL-**DNN



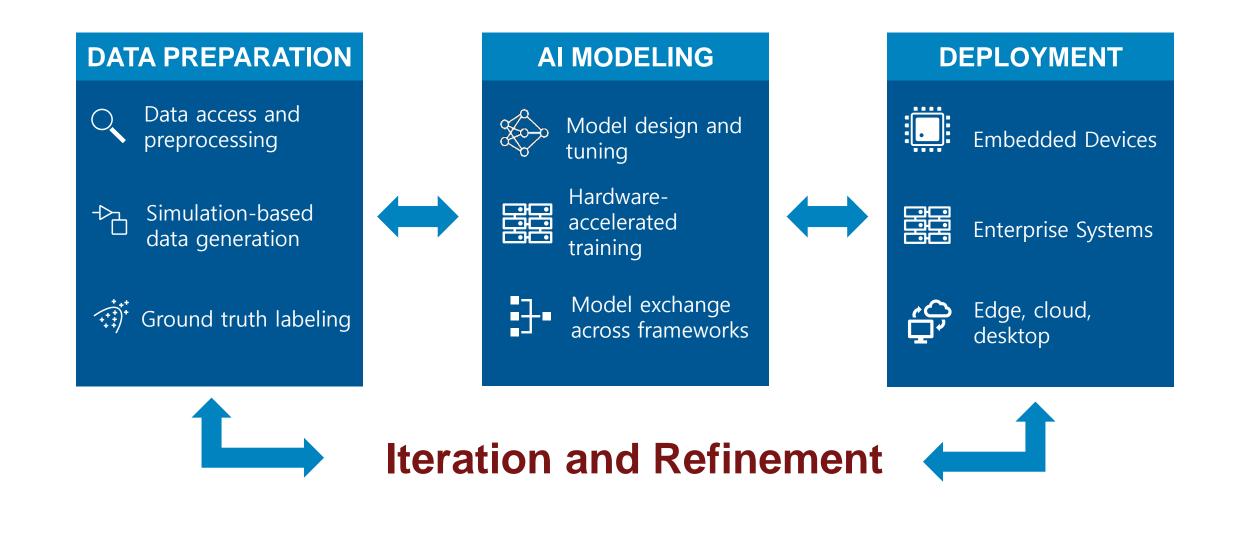
Deploy to Enterprise IT Infrastructure







Defect Detection Workflow







Key Takeaways

- Interactive and easy to use apps help explore, iterate and automate workflows
- Flexibility and options to choose networks and optimizations based on data and requirements
- MATLAB provides an easy and extensible framework for defect detection from data access to deployment







THANK YOU!



