

MATLAB EXPO 2019

Deep Learning and Reinforcement Learning Workflows in A.I.

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Principal Product Manager

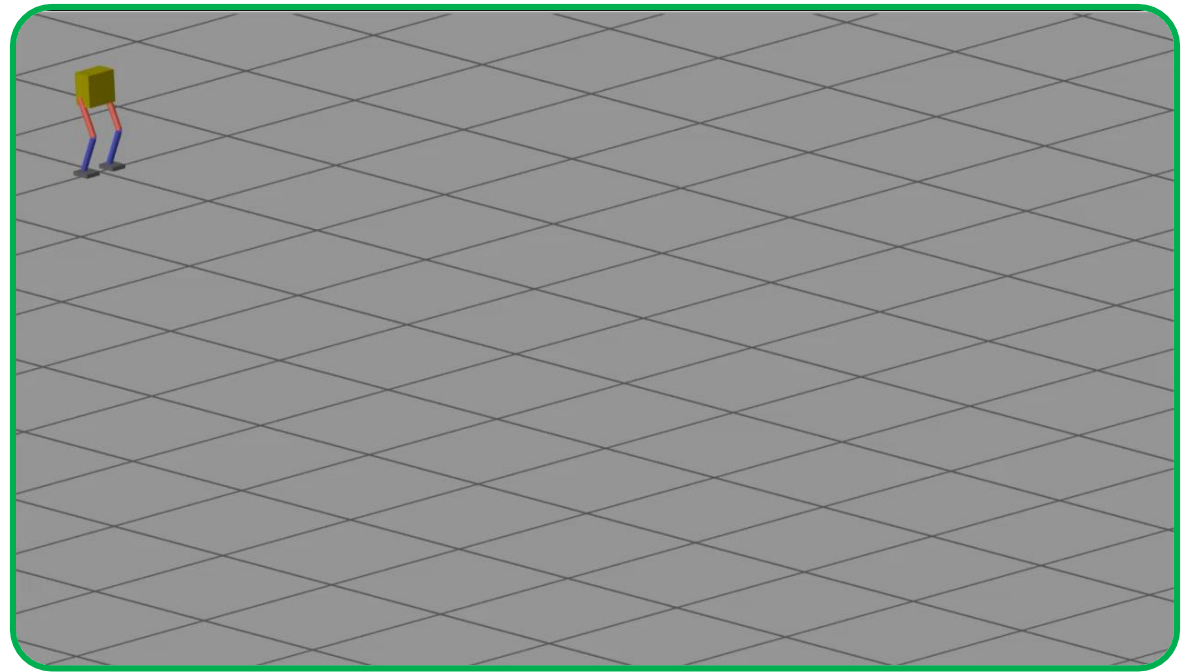
Deep Learning, Computer Vision, Automated Driving



Why MATLAB for Artificial Intelligence?

Artificial Intelligence

Development of computer systems to perform tasks that normally require human intelligence



A.I. Applications



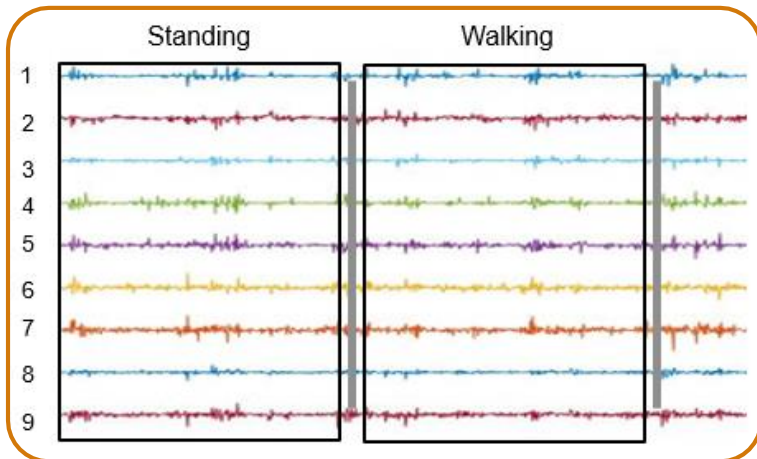
Object Classification



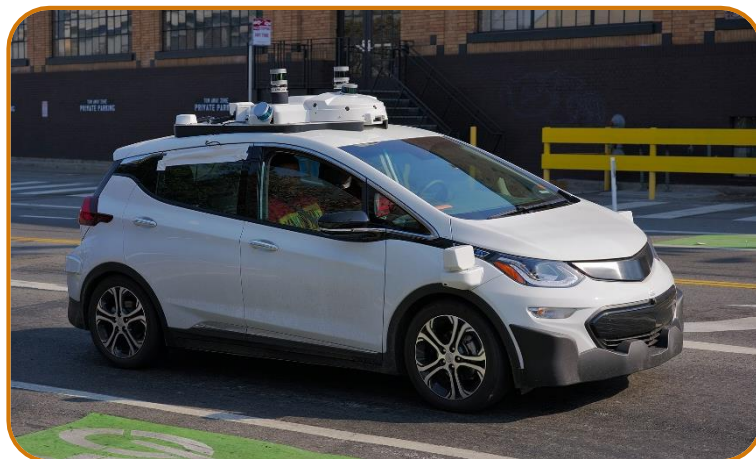
Speech Recognition



Predictive Maintenance



Signal Classification

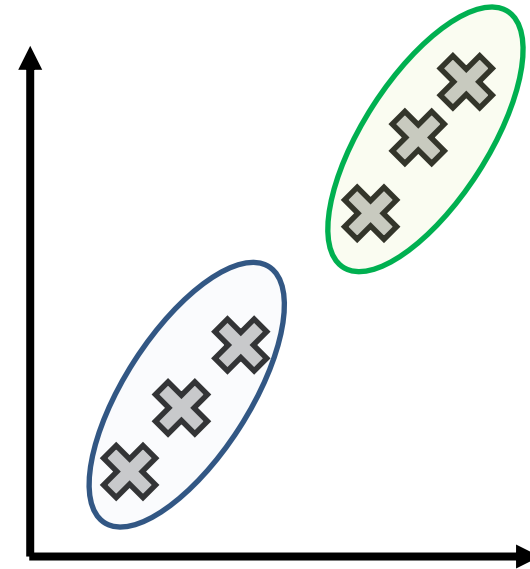
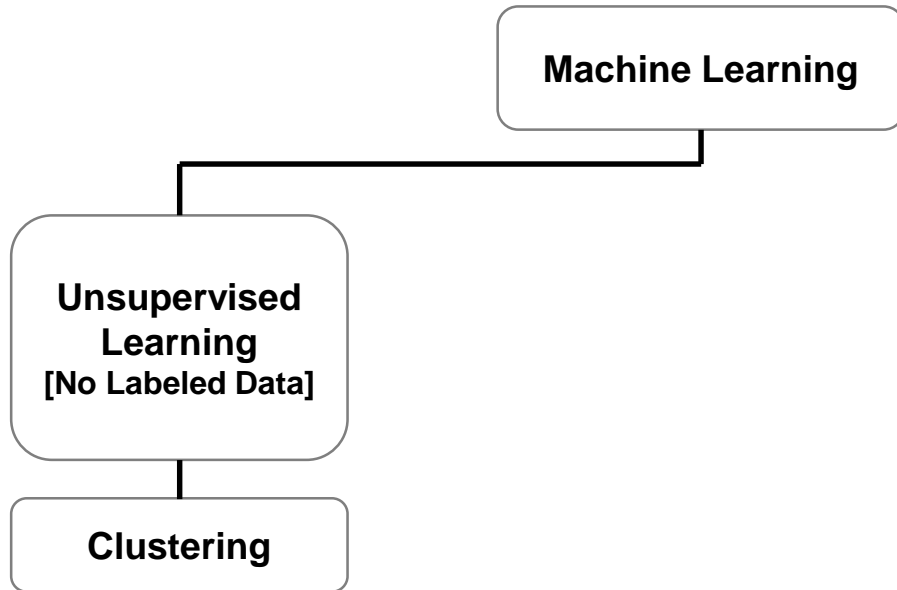


Automated Driving

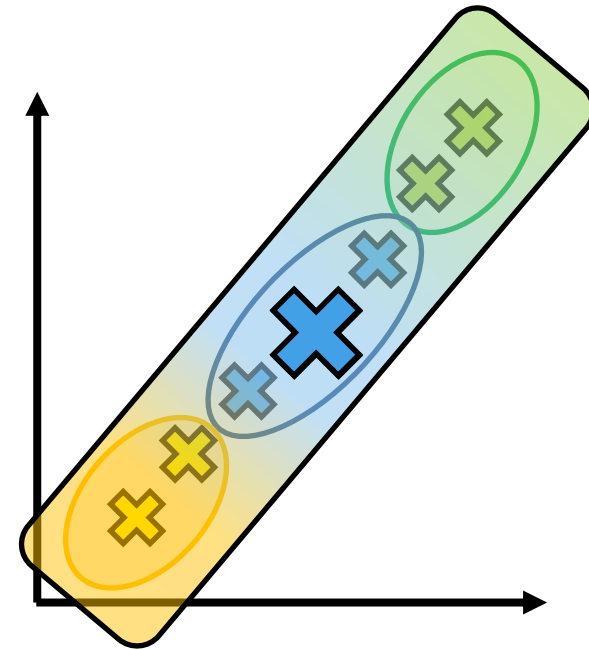
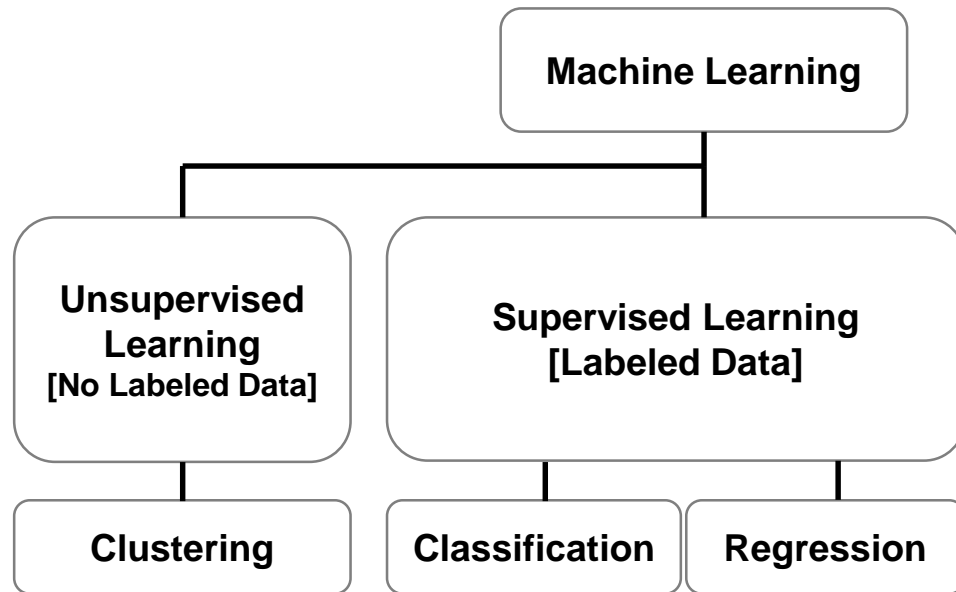


Stock Market Prediction

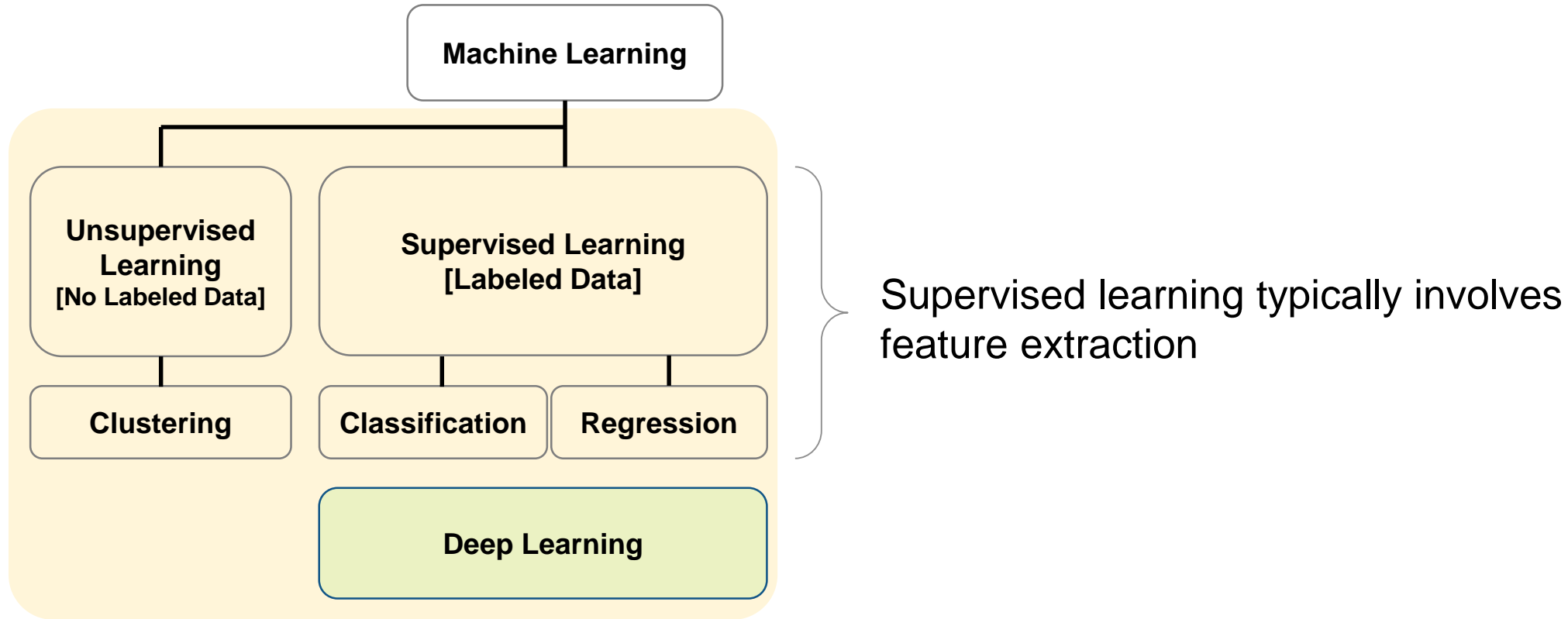
Machine Learning and Deep Learning



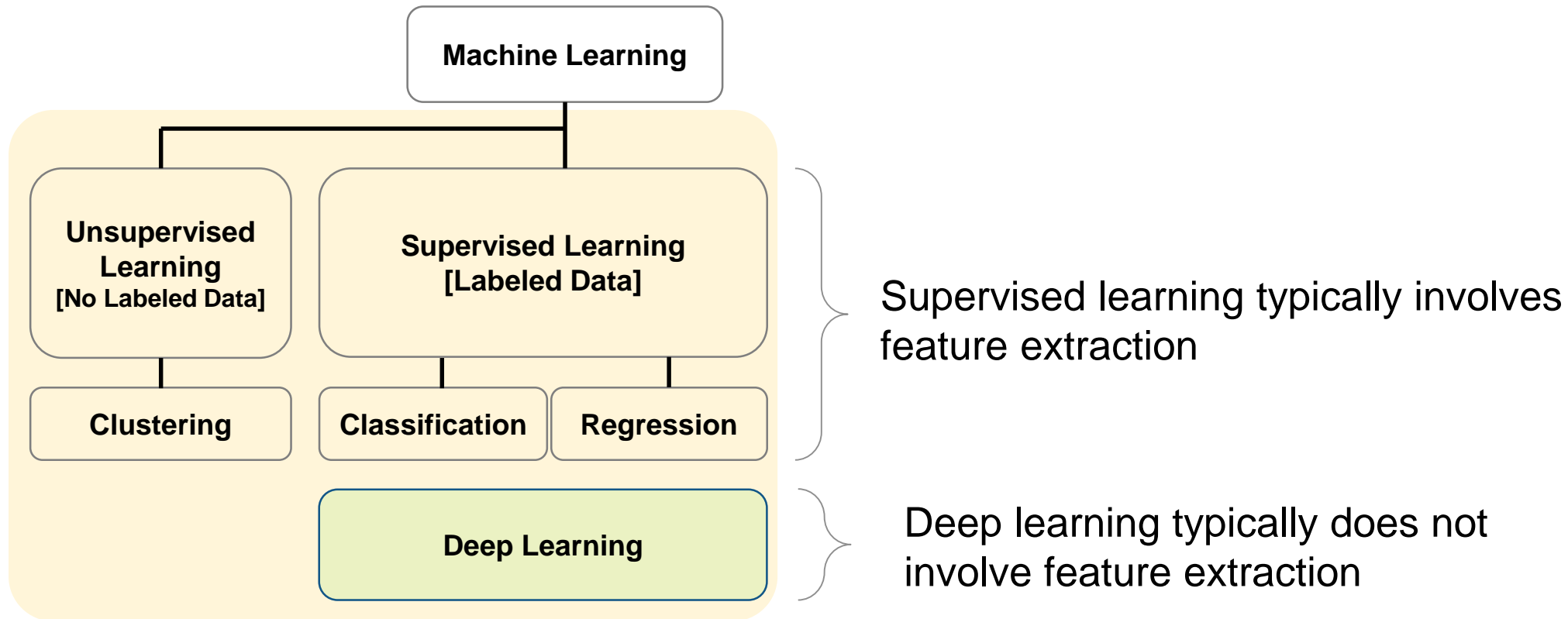
Machine Learning and Deep Learning



Machine Learning and Deep Learning

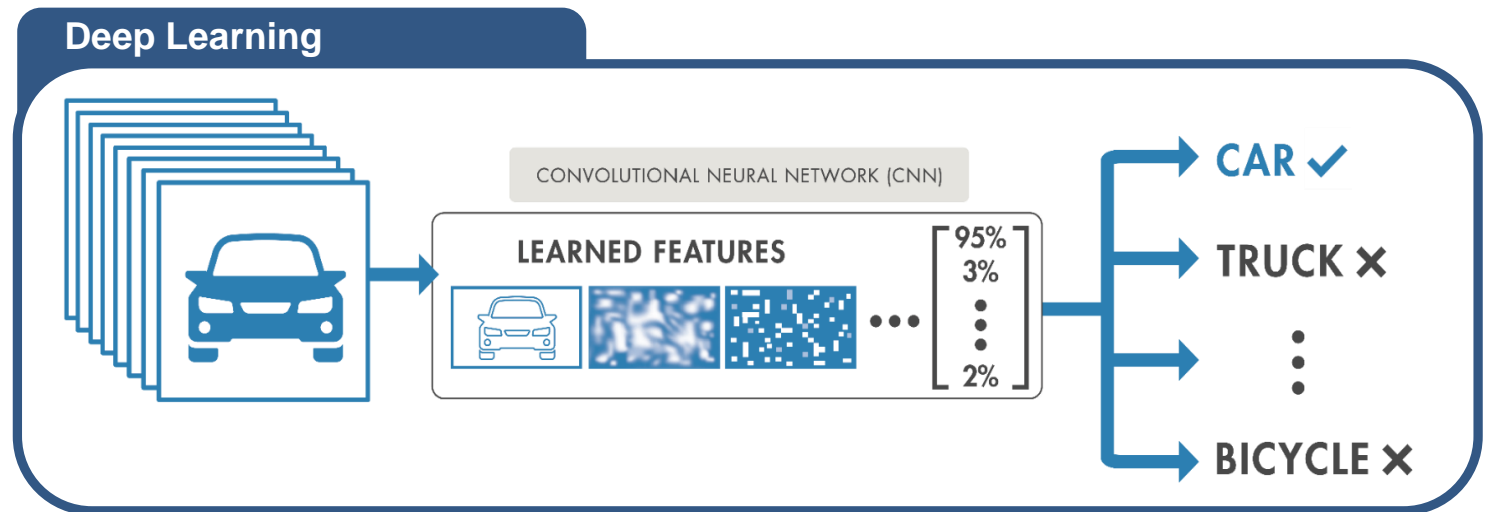
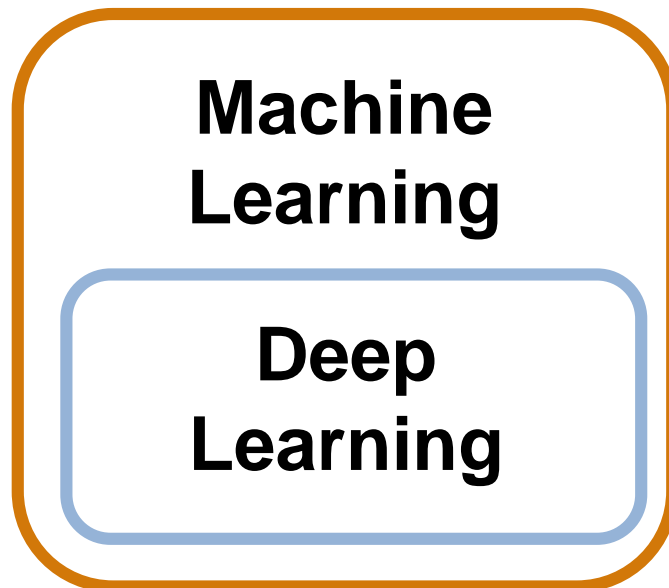


Machine Learning and Deep Learning

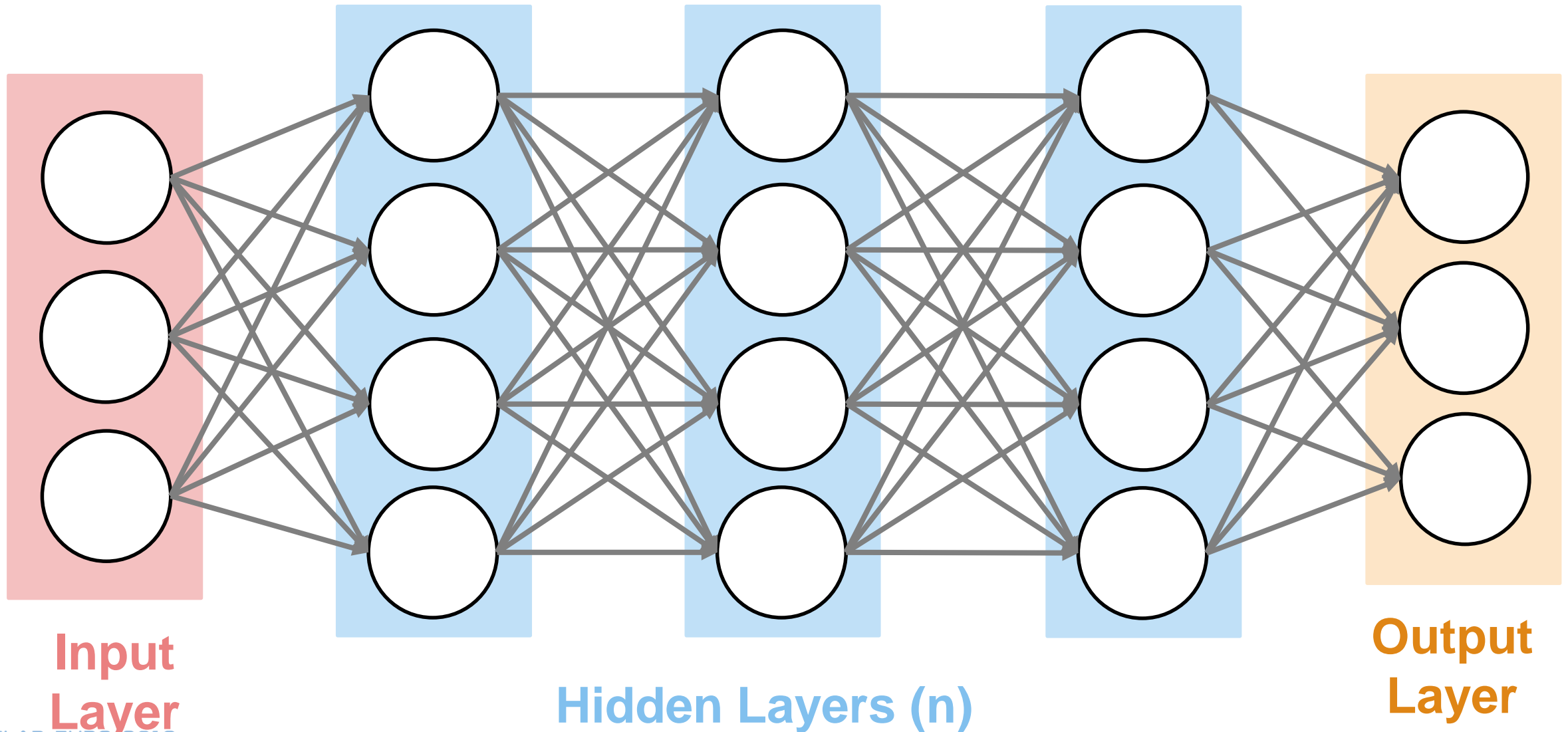


Deep Learning

- Subset of machine learning with **automatic feature extraction**
 - Learns features and tasks directly from data
 - More Data = better model



Deep Learning Uses a Neural Network Architecture



Deep Learning Datatypes

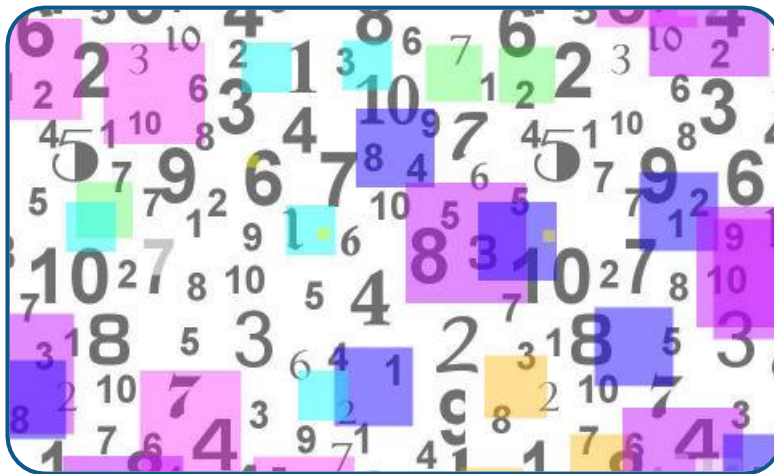
Image



Signal



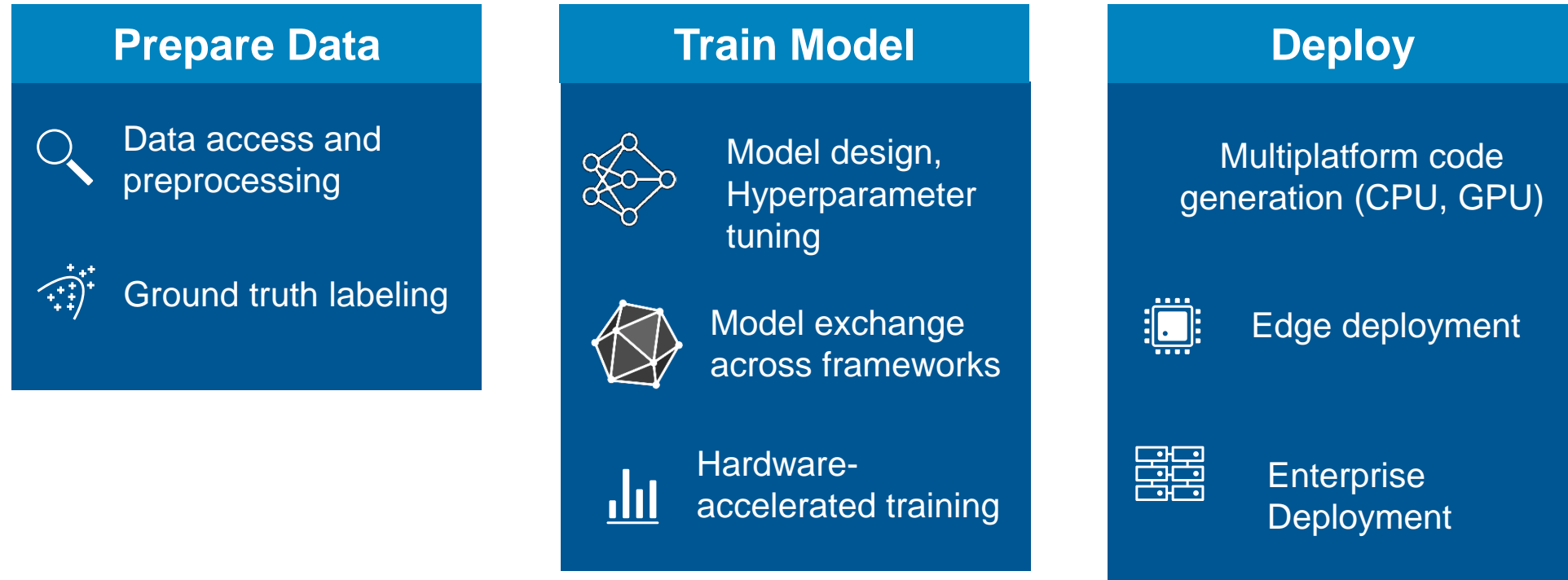
Numeric



Text



Deep Learning Workflow



Why MATLAB for A.I. Tasks?

Increased productivity with interactive tools

Generate simulation data for complex models and systems

Ease of deployment and scaling to various platforms

Full A.I. workflows that cannot be easily replicated by other toolchains

Why MATLAB for A.I. Tasks?

Increased productivity with interactive tools

Labeling

Training

**Model
Exchange**

**Full A.I. workflows that cannot be easily
replicated by other toolchains**

**Labeling for deep learning is repetitive,
tedious, and time-consuming...**

but necessary

ROI Label Definition

Define new ROI label

Vehicle

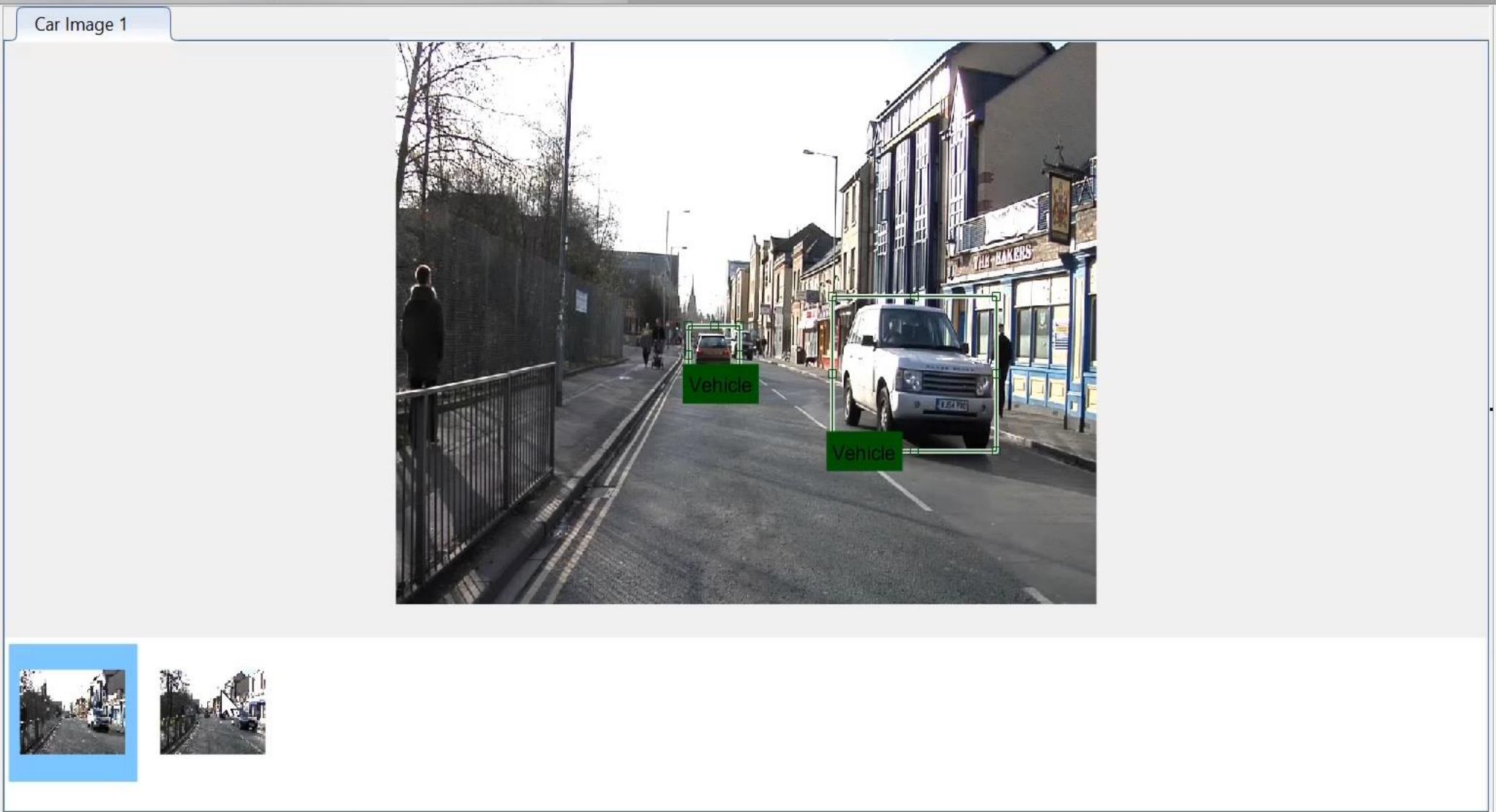
Scene Label Definition

Define new scene label

Apply to Image

Remove from Image

To label a scene, you must first define a scene label.



LABEL

FILE: Load, Save, Import Labels

MODE: Label, Zoom In, Zoom Out, Pan

VIEW: Layout, Show ROI Labels, Show Scene Labels

AUTOMATE LABELING: Algorithm: Select Algorithm, Automate, Configure Automation

SUMMARY: View Label Summary

EXPORT: Export Labels

ROI Label Definition

Label, Sublabel, Attribute

Lane

Scene Label Definition

Define new scene label

Current Frame: Add Label

Time Interval: Remove Label

To label a scene, you must first define a scene label.

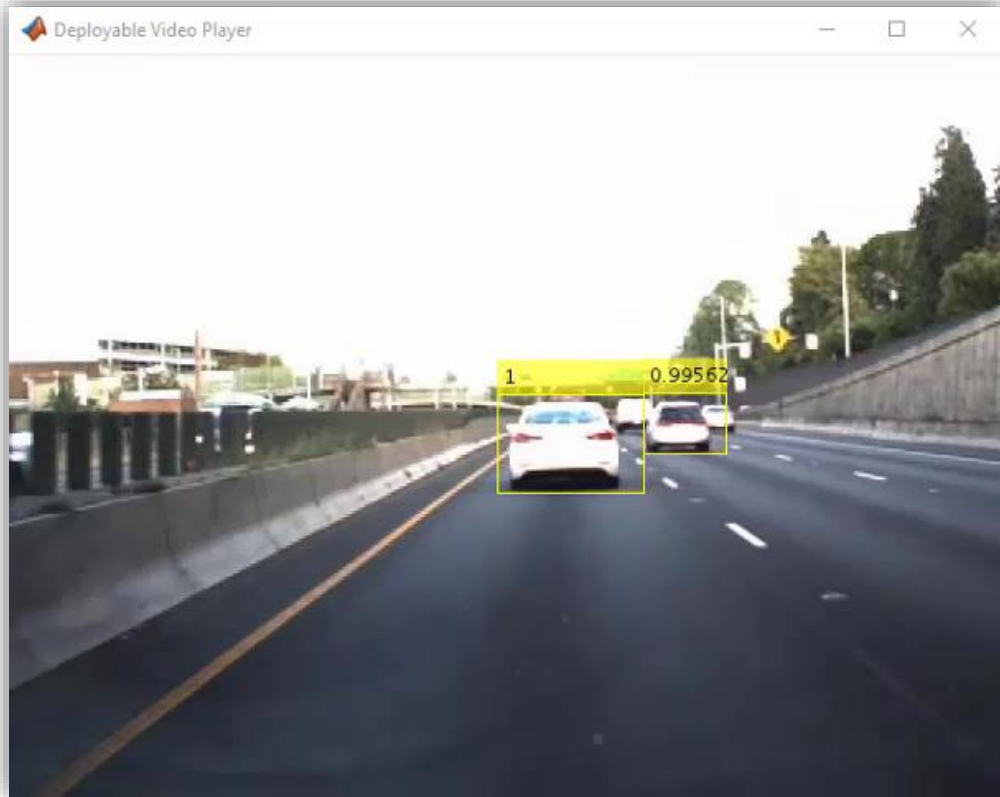
05_highway_lanechange_25s.mp4



Timeline: 00.00000 (Start Time), 05.80000 (Current), 25.00000 (End Time), 25.00000 (Max Time)

Navigation: Previous, Play, Next, Stop

Zoom In Time Interval



Signal Labeling – annotate signals with labels/sublabels, export to workspace for training

Define Labels

View properties of labels

The interface shows the 'Label' tab with a 'Point' definition for 'TrillPeaks'. The 'Label Definition' section shows a tree view with 'WhaleType', 'MoanRegions', 'TrillRegions', and 'TrillPeaks'. The 'Labeled Signal Set' table is as follows:

Name	Plot	Value	Location (Min)	Location (Max)
whale1	<input checked="" type="checkbox"/>	blue		
WhaleType		blue		
MoanRegions				
	<input checked="" type="checkbox"/>	true	6.13604115...	7.763
	<input type="checkbox"/>	true	16.37525	18.153984...
	<input type="checkbox"/>	true	11.4020000...	13.120148...
TrillRegions				
	<input type="checkbox"/>	true	1.4357724...	3.275
TrillPeaks				
	<input type="checkbox"/>	1	1.77425	
	<input type="checkbox"/>	2	2.44375	
	<input checked="" type="checkbox"/>	3	2.74225	
whale2	<input checked="" type="checkbox"/>	blue		
WhaleType		blue		
MoanRegions				
	<input checked="" type="checkbox"/>	true	2.44511966...	3.5605
	<input type="checkbox"/>	true	5.7136928...	8.113
	<input type="checkbox"/>	true	15.3215	16.712880...
TrillRegions				
	<input type="checkbox"/>	true	10.91475	13.152470...
TrillPeaks				
	<input type="checkbox"/>	1	11.50975	
	<input type="checkbox"/>	2	11.88	
	<input checked="" type="checkbox"/>	3	12.32975	

The top plot shows two signals, 'whale1' (blue) and 'whale2' (orange), over time. The bottom plot shows the 'MoanRegions', 'TrillRegions', and 'TrillPeaks' for both signals, with interactive annotations like 'true' and '1 2 3'.

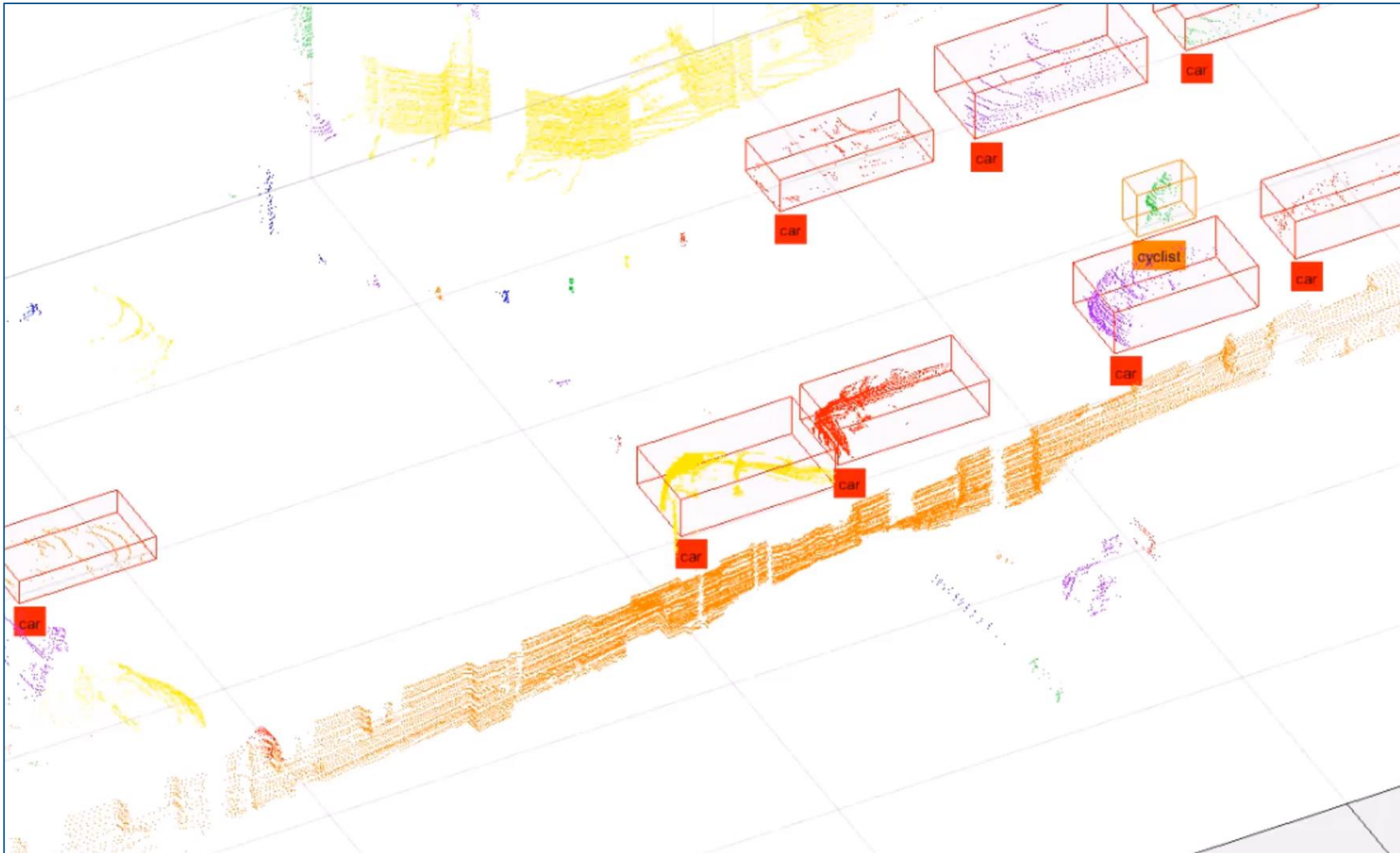
Interactively Label Signals

User Story – Veoneer (Autoliv)

- Automotive
 - Software and hardware for active safety, autonomous driving, occupant protection, and brake control
- Building radar sensor – check accuracy using LiDAR-based verification
- Human analyzes hours of recorded data
- Used MATLAB to semi-automate labeling and tracking of 3D LiDAR point clouds.



Manual Labeling for 25 events took over 20 minutes.
After full automation with MATLAB's tools, it took 5 minutes



Why MATLAB for A.I. Tasks?

Increased productivity with interactive tools

Labeling

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**Model
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LAYERS

INPUT

-  ImageInputLayer
-  SequenceInputLayer

LEARNABLE

-  Convolution2DLayer
-  TransposedConvolution2DLayer
-  FullyConnectedLayer
-  LSTMLayer
-  BiLSTMLayer

ACTIVATION

-  ReLUActivation
-  LeakyReLUActivation
-  ClippedReLUActivation

NORMALIZATION AND DROPOUT



PROPERTIES

Number of layers	0
Number of connections	0
Input type	None
Output type	None

Import pre-trained model as starting point

DESIGNER

FILE BUILD NAVIGATE LAYOUT ANALYSIS EXPORT

LAYER LIBRARY

Filter layers...

- clippedReluLayer
- tanhLayer
- eluLayer

NORMALIZATION AND UTILITY

- dropoutLayer
- batchNormalizationLayer
- crossChannelNormalizationLayer
- crop2dLayer

POOLING

- averagePooling2dLayer
- averagePooling3dLayer
- maxPooling2dLayer
- maxPooling2dLayer (for unpooling)
- maxUnpooling2dLayer
- maxPooling3dLayer
- roiMaxPooling2dLayer

COMBINATION

- additionLayer
- depthConcatenationLayer
- concatenationLayer

OBJECT DETECTION

- regionProposalLayer
- yolov2ReorgLayer
- yolov2TransformLayer

OUTPUT

- softmaxLayer
- classificationLayer
- regressionLayer
- rpnSoftmaxLayer
- rcnnBoxRegressionLayer
- rpnClassificationLayer
- pixelClassificationLayer
- yolov2OutputLayer

YOLOv2 Object Detection Layers
R2019a

PROPERTIES

Number of layers	0
Number of connections	0
Input type	None
Output type	None

Import Pre-trained Models for Transfer Learning

Inception-v3

ResNet-101

VGG-16

Inception-ResNet-v2

ResNet-18

GoogLeNet

DenseNet-201

VGG-19

SqueezeNet

AlexNet

ResNet-50

Places365-GoogLeNet

MobileNet-v2

Xception

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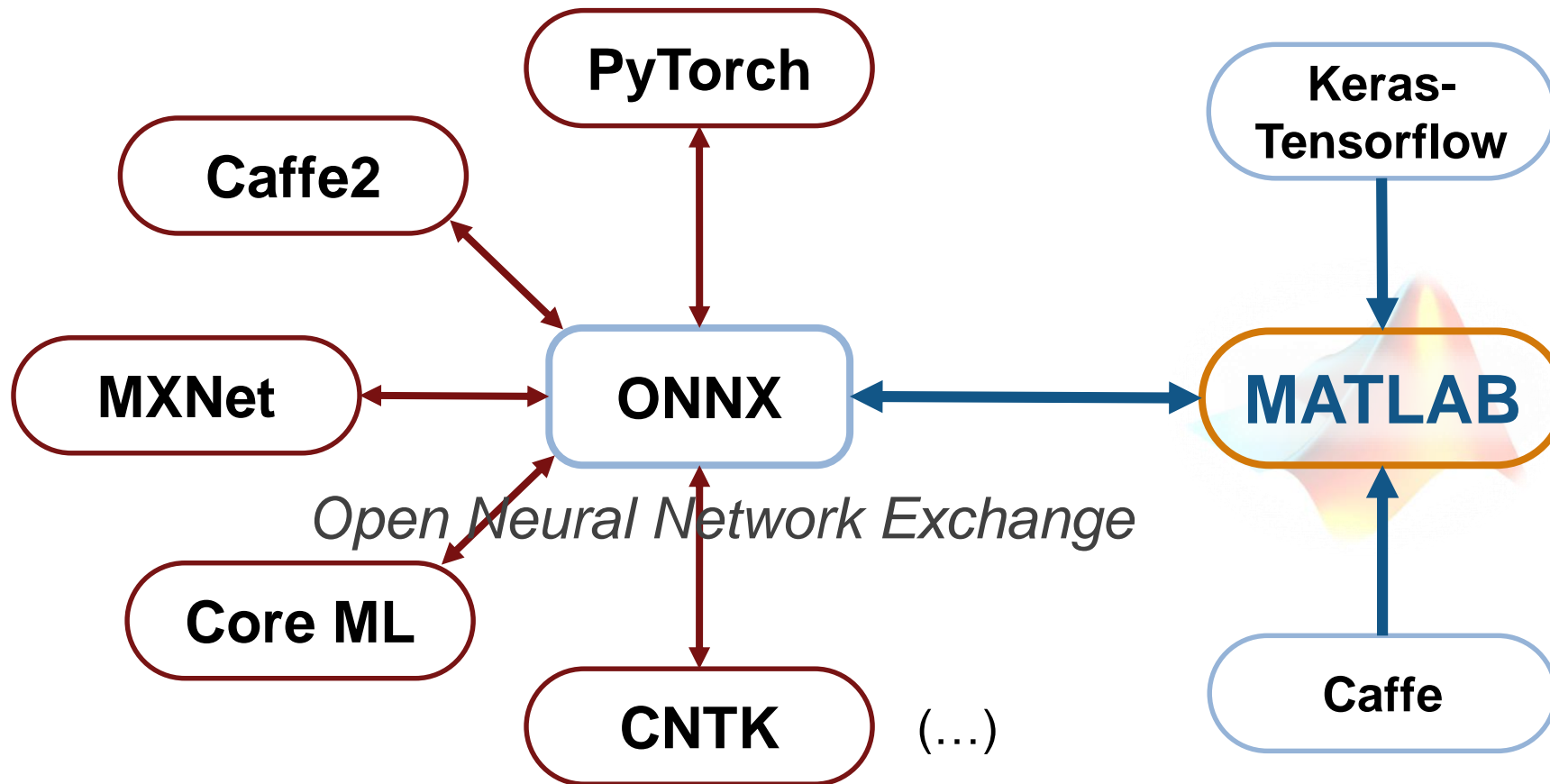
Labeling

Training

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Model Exchange with MATLAB



Why MATLAB for A.I. Tasks?

Increased productivity with interactive tools

Generate simulation data for complex models and systems

Ease of deployment and scaling to various platforms

Full A.I. workflows that cannot be easily replicated by other toolchains

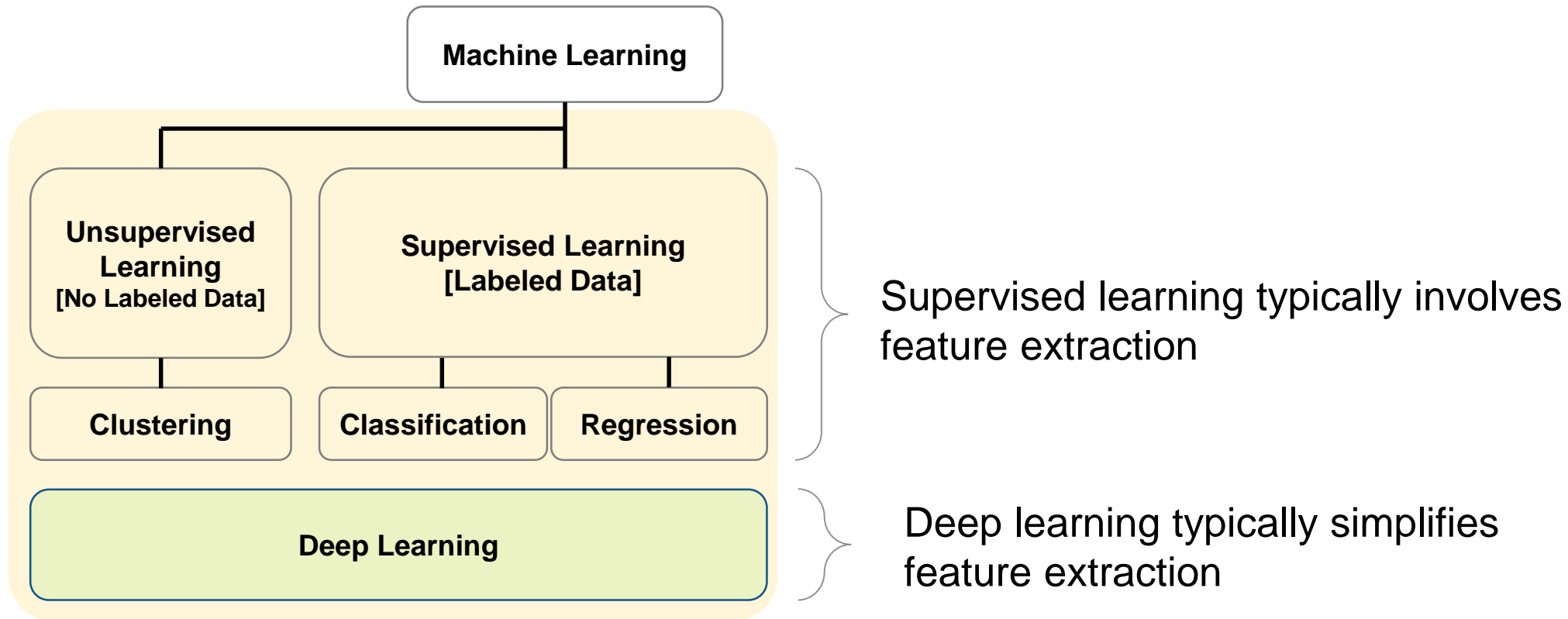
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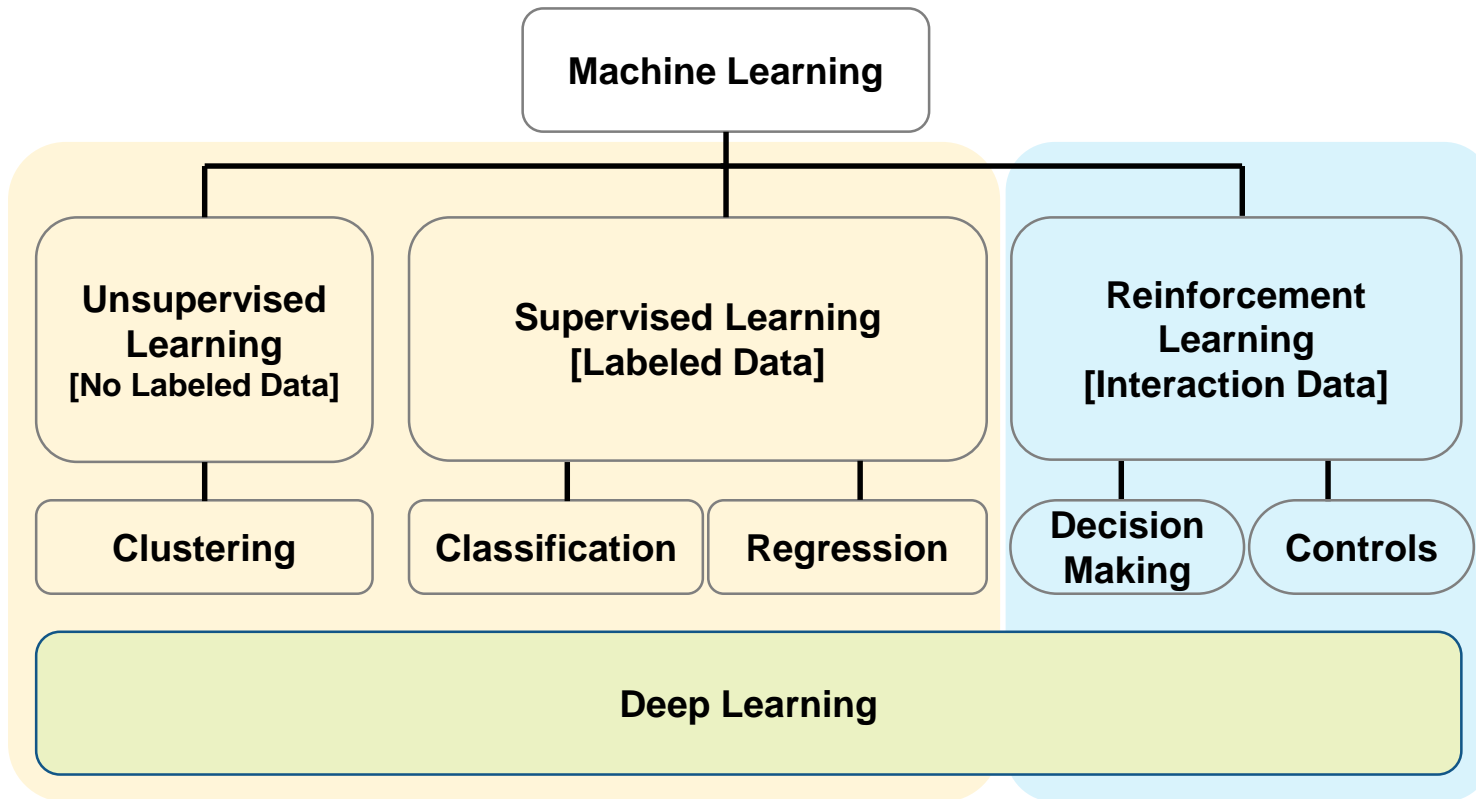
**Reinforcement
Learning**

**Full A.I. workflows that cannot be easily
replicated by other toolchains**

Reinforcement Learning vs Machine Learning vs Deep Learning



Reinforcement Learning vs Machine Learning vs Deep Learning

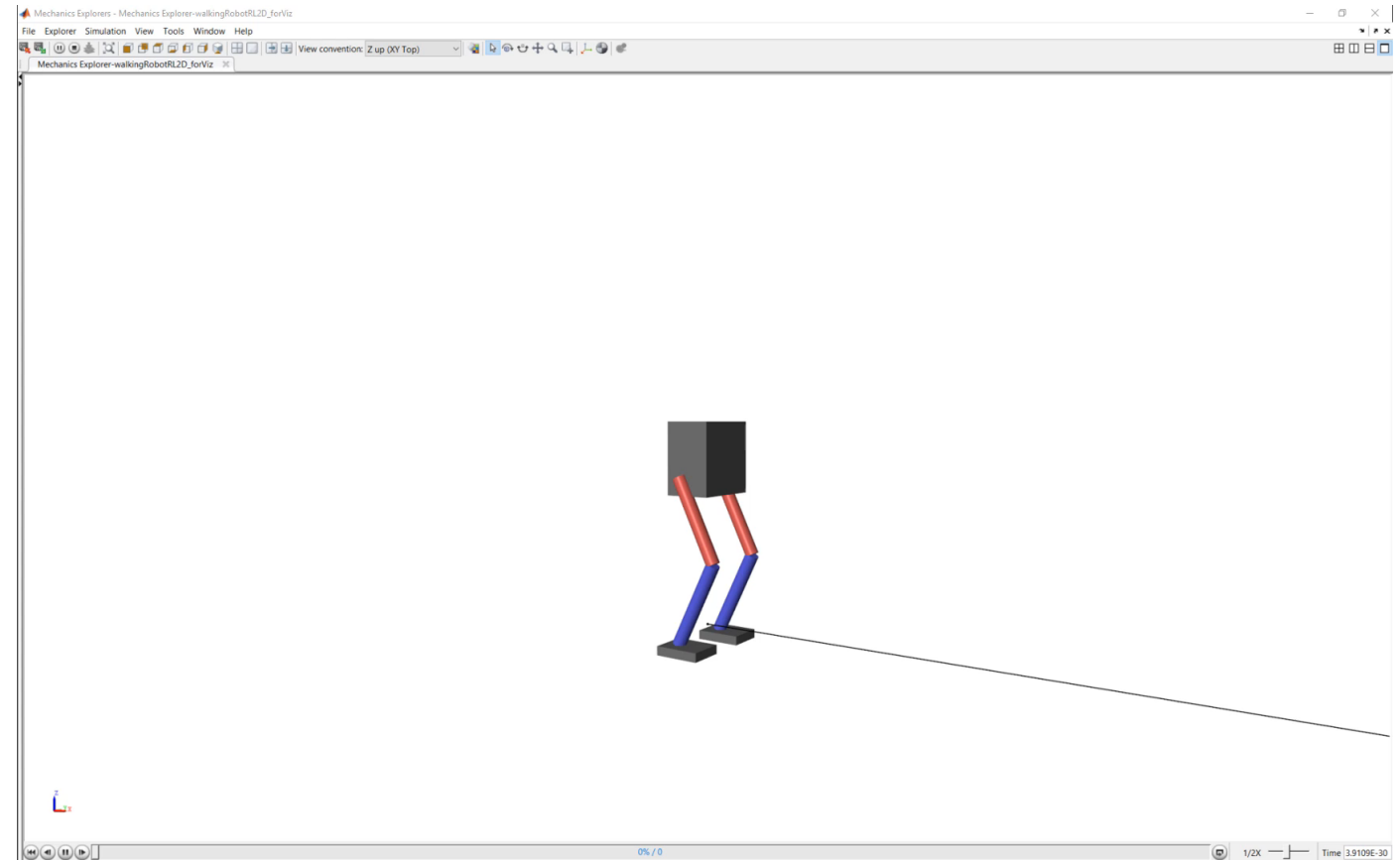


Reinforcement learning:

- Learning through trial & error [*interaction*]
- It's about learning a **behavior** or accomplishing a **task**

What is Reinforcement Learning?

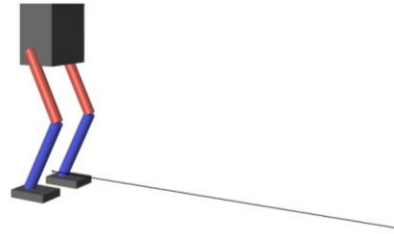
- What is Reinforcement Learning?
 - Type of machine learning that trains an **'agent'** through repeated interactions with an environment
- How does it work?
 - Through a trial & error process that uses a reward system to maximize success



Reinforcement Learning enables the use of Deep Learning for Controls and Decision Making Applications



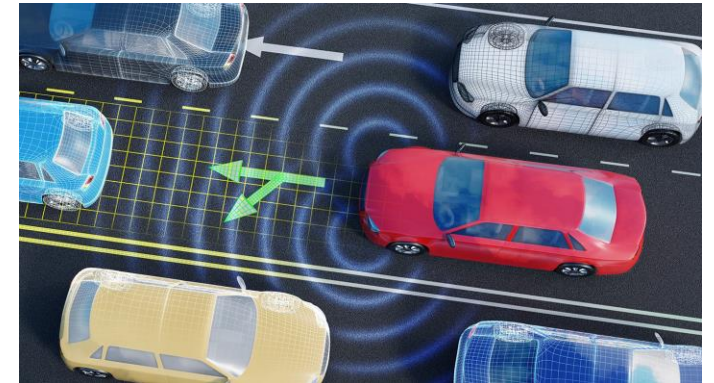
Controls



Robotics

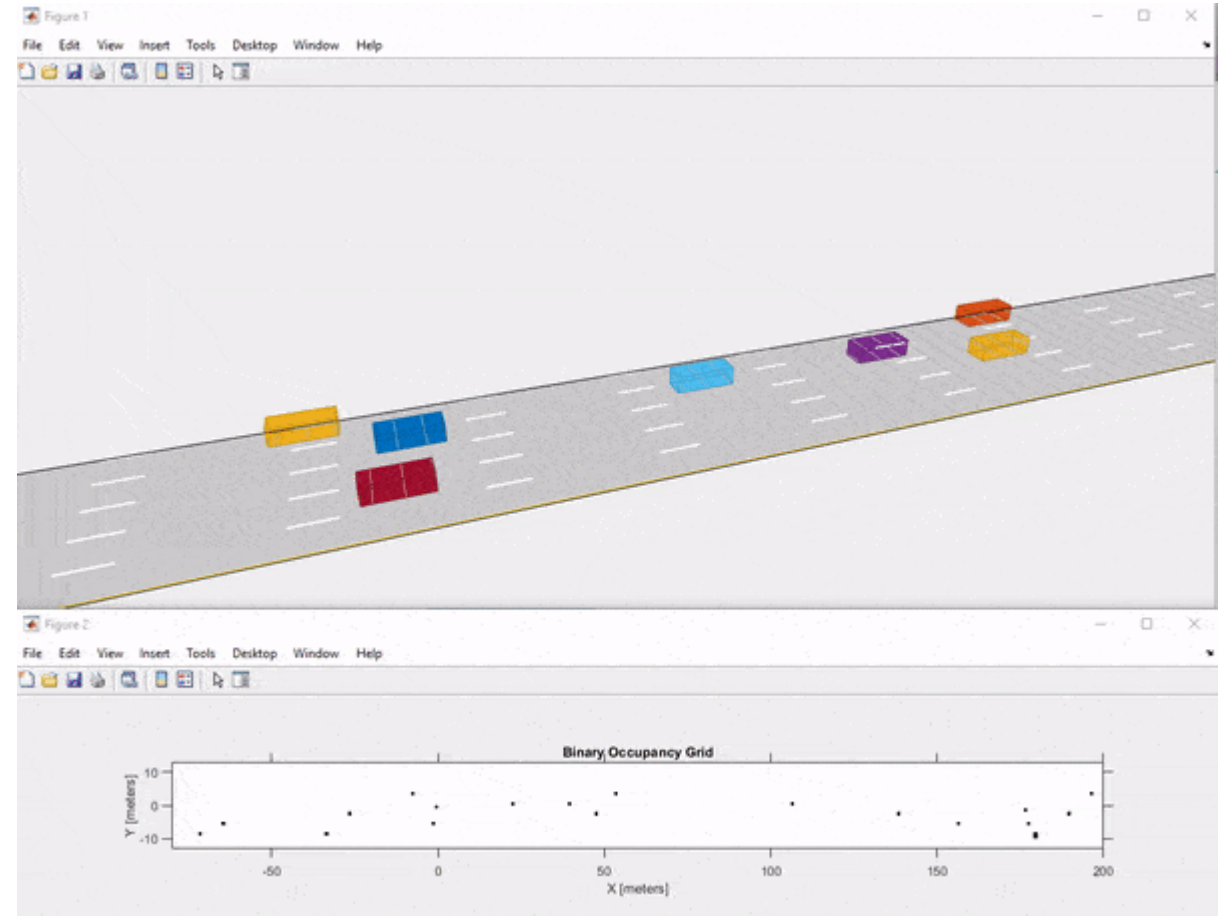
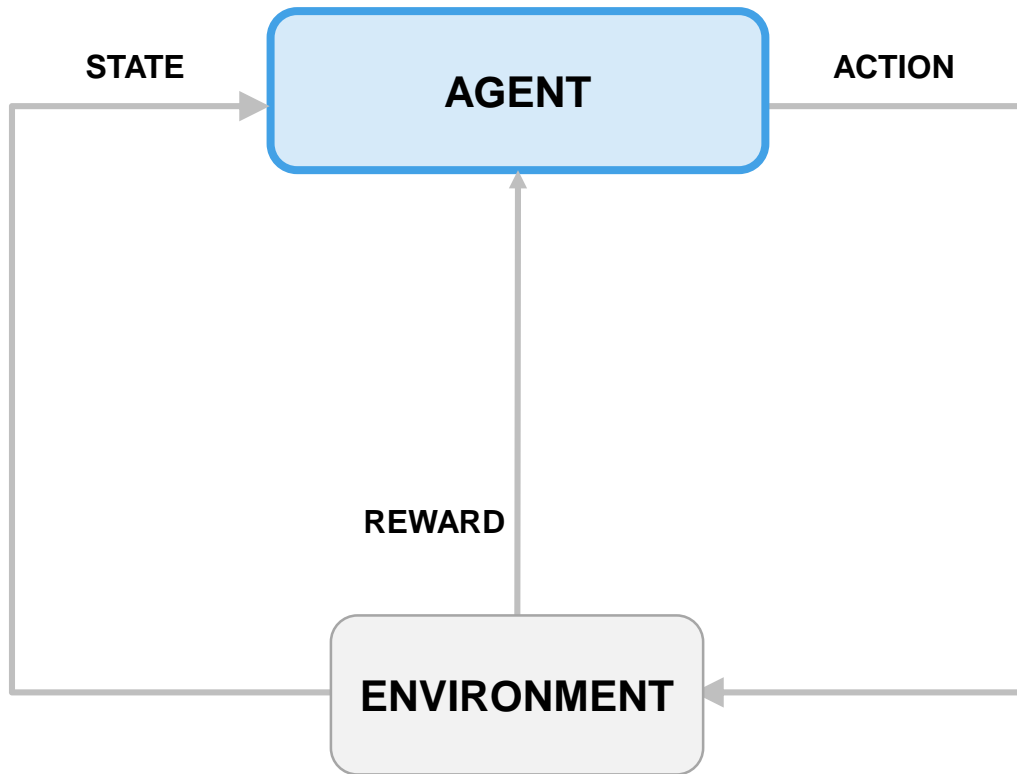


A.I. Gameplay



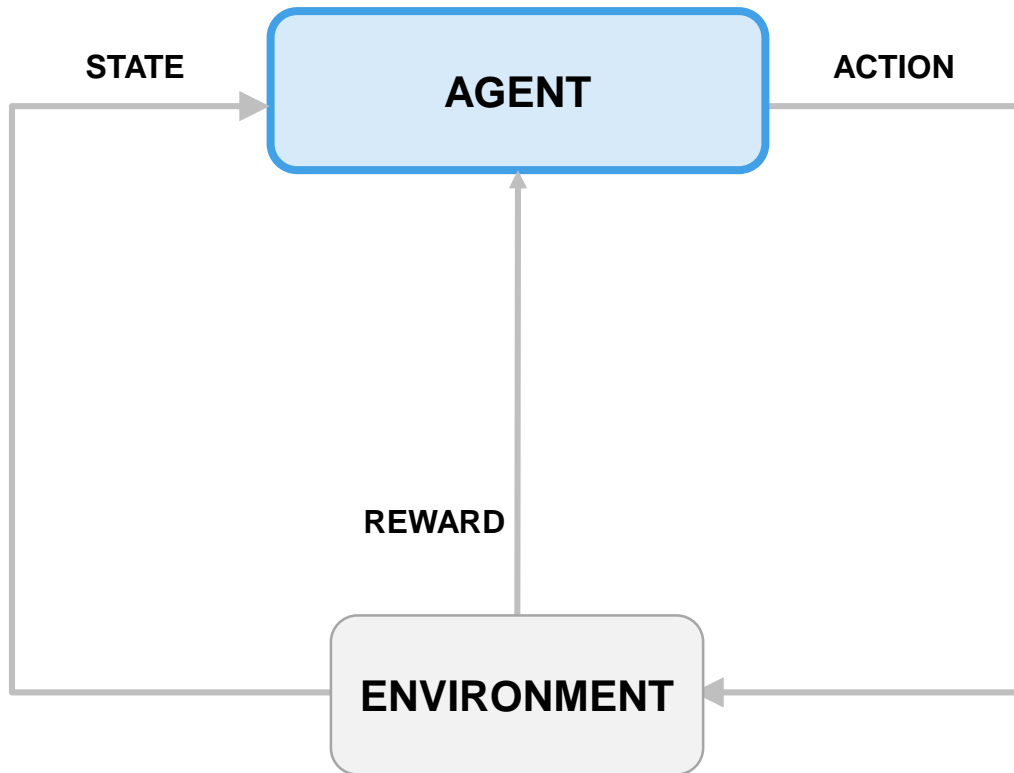
Autonomous driving

How Does Reinforcement Learning Work?



A Practical Example of Reinforcement Learning

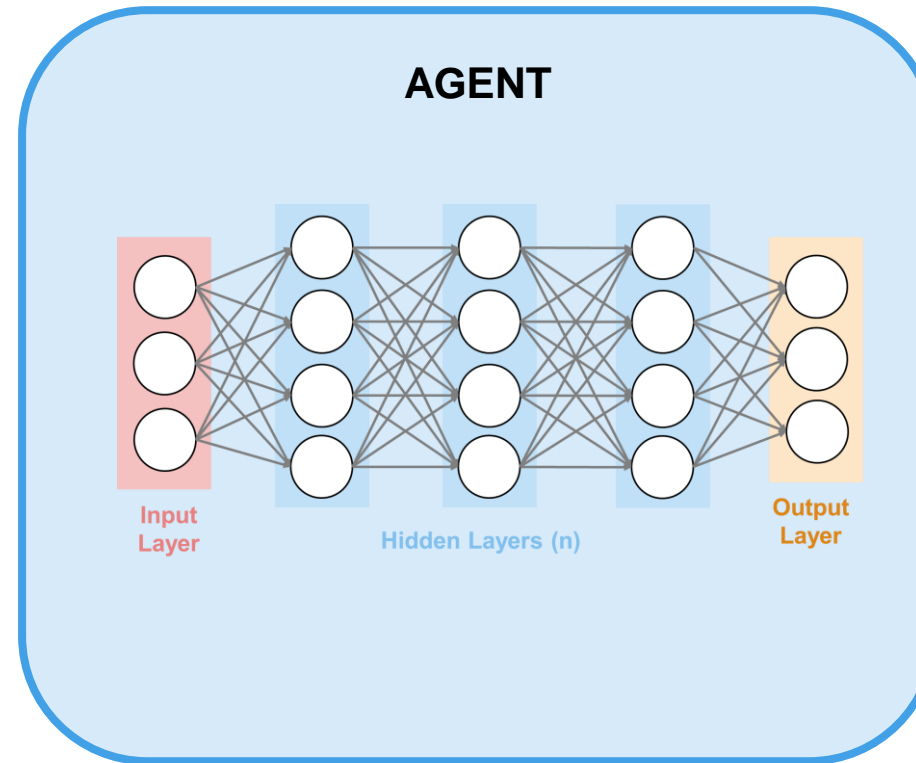
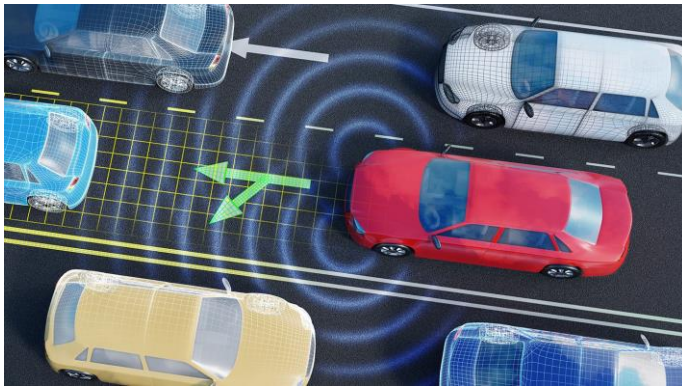
Training a Self-Driving Car



- Vehicle's computer learns how to drive...
(**agent**)
- using sensor readings from LIDAR, cameras,...
(**state**)
- that represent road conditions, vehicle position,...
(**environment**)
- by generating steering, braking, throttle commands,...
(**action**)
- to avoid collisions and lane deviation...
(**reward**).

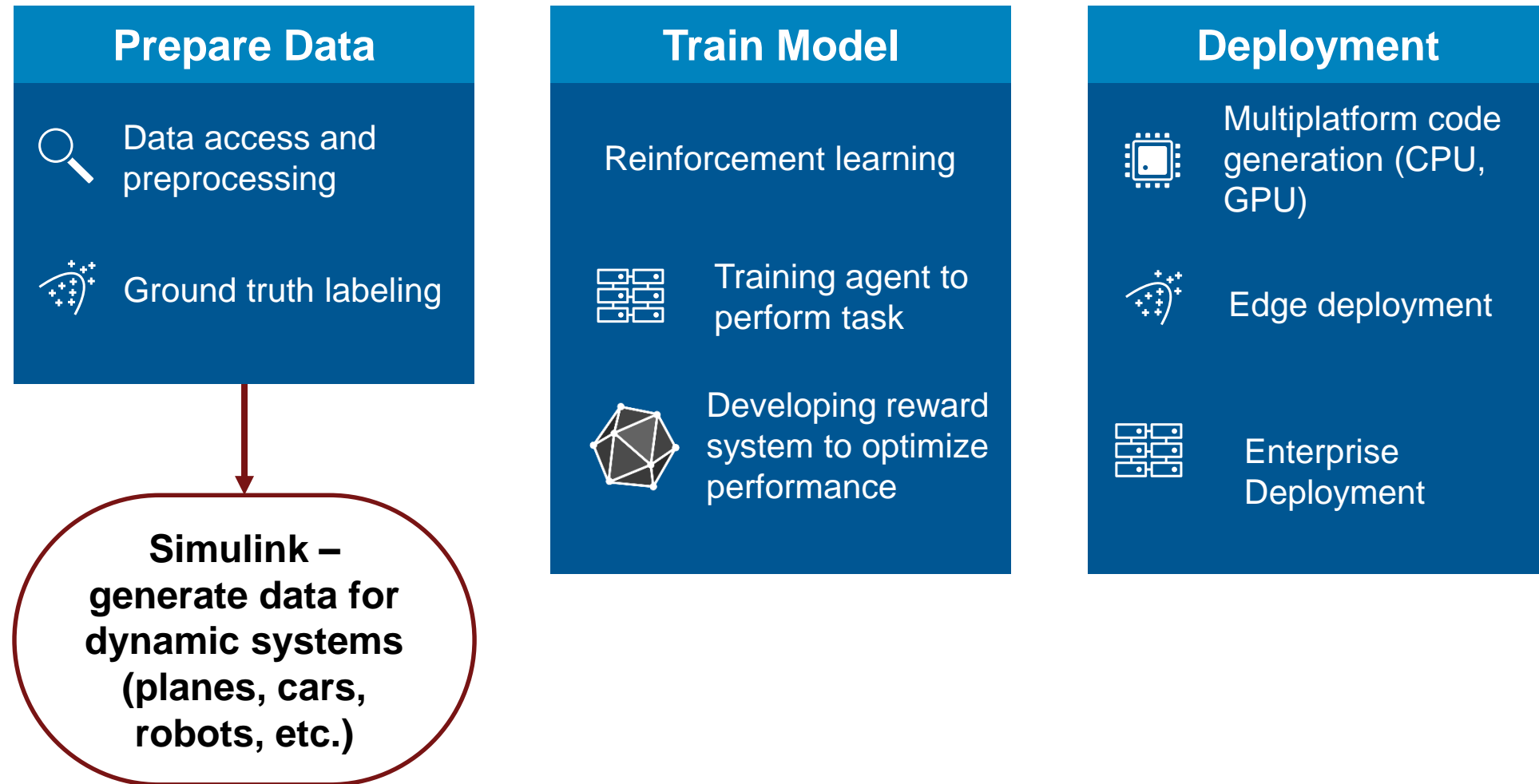
The goal of Reinforcement learning is for the agent to find an optimal algorithm for performing a task

Deep Networks are commonly found in the agent, because they can model complex problems.



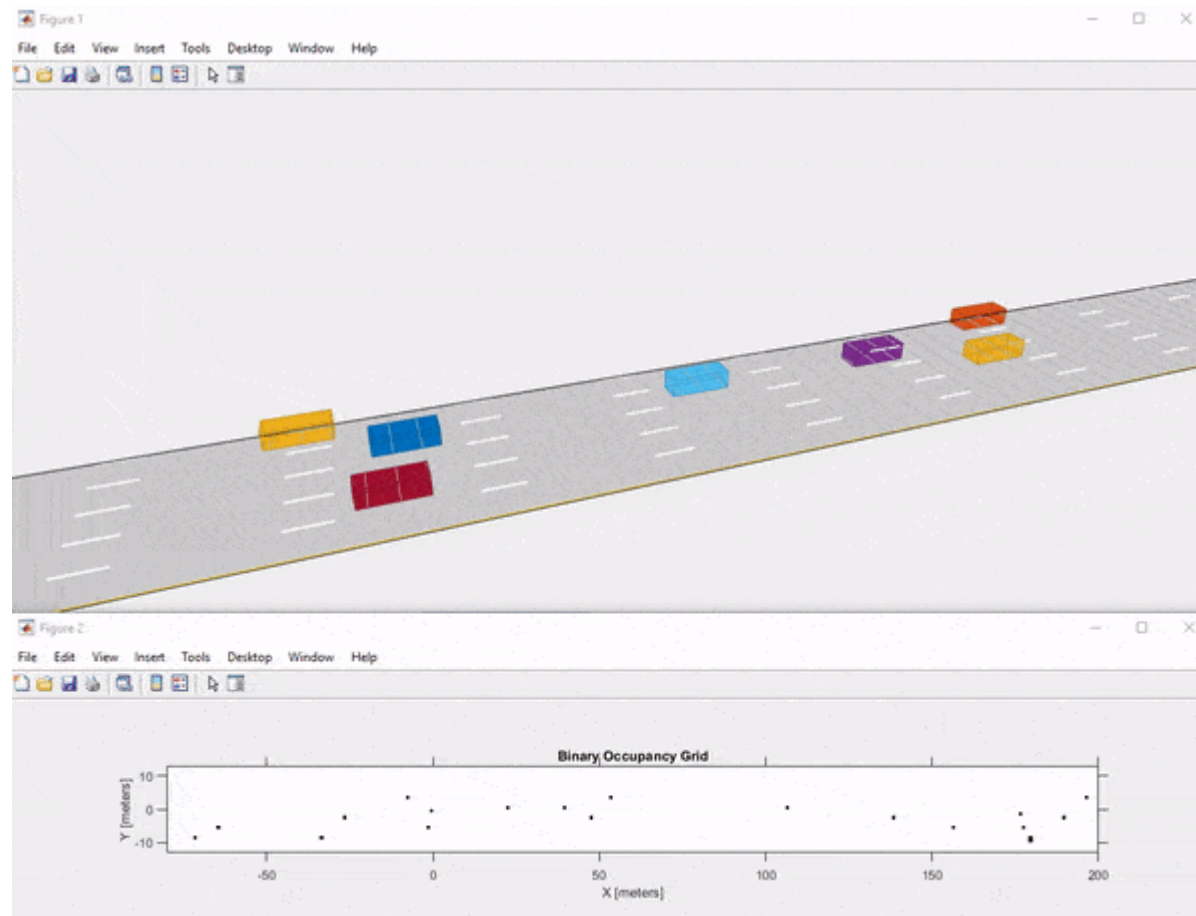
- **Turn left**
- **Turn right**
- **Brake**
- **Accelerate**

Reinforcement Learning Workflow



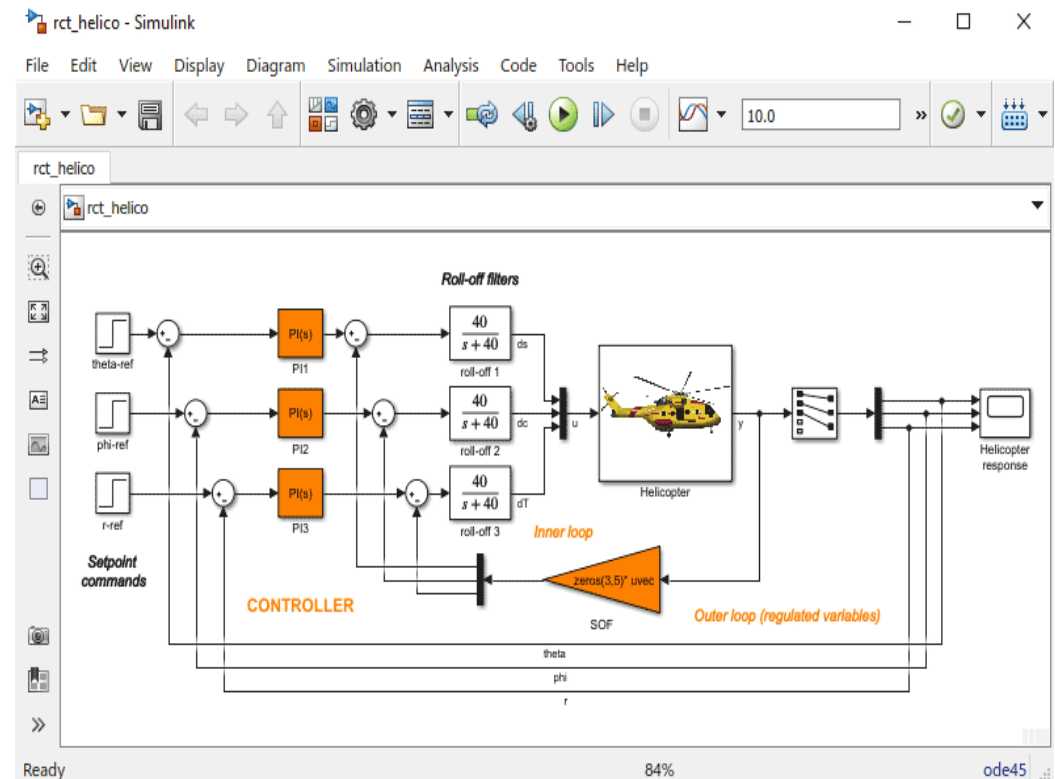
Why MATLAB and Simulink for Reinforcement Learning?

Virtual models allow you to simulate conditions hard to emulate in the real world.



Using MATLAB and Simulink for Reinforcement Learning

- Reinforcement learning is a dynamic process
- Decision making problems
 - Financial trading, calibration, etc.
- Controls-based problems
 - Lane-keep assist, adaptive cruise control, robotics, etc.



Why MATLAB for A.I. Tasks?

Increased productivity with interactive tools

Generate simulation data for complex models and systems

Ease of deployment and scaling to various platforms

Full A.I. workflows that cannot be easily replicated by other toolchains

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Code

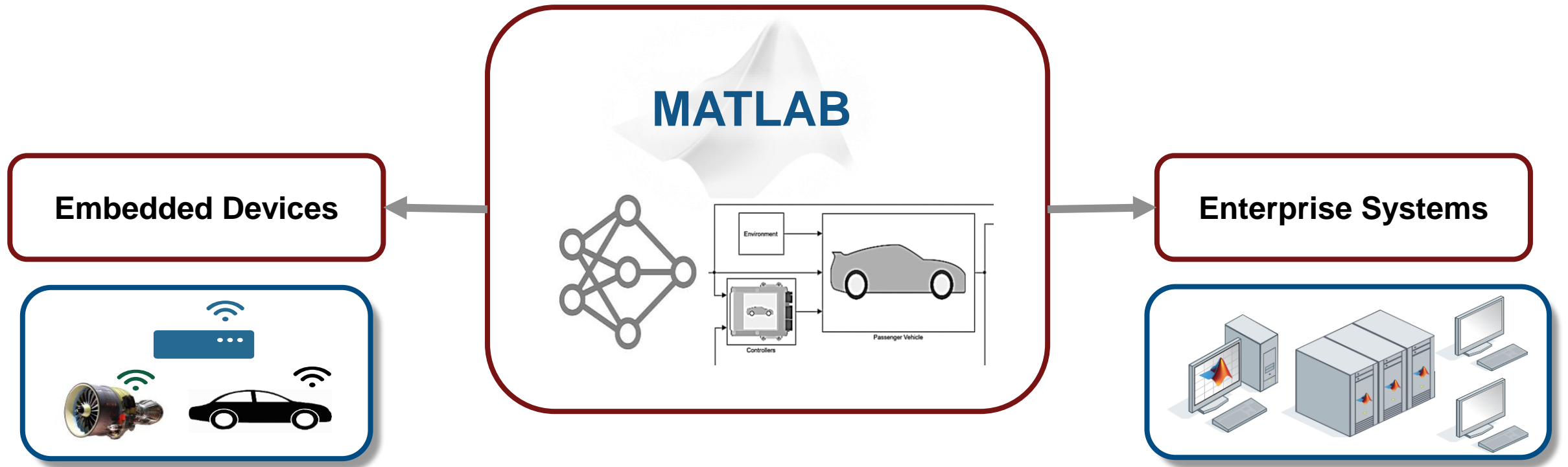
Embedded

Enterprise

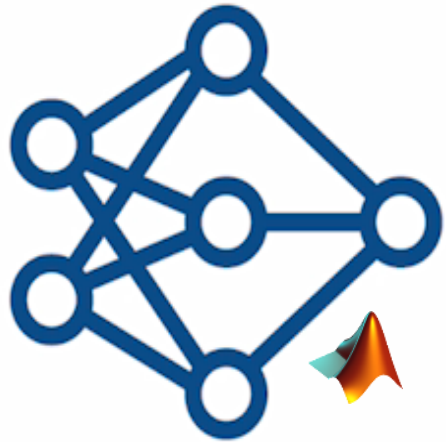
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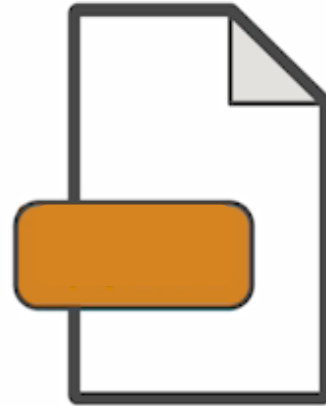
Deployment and Scaling for A.I.



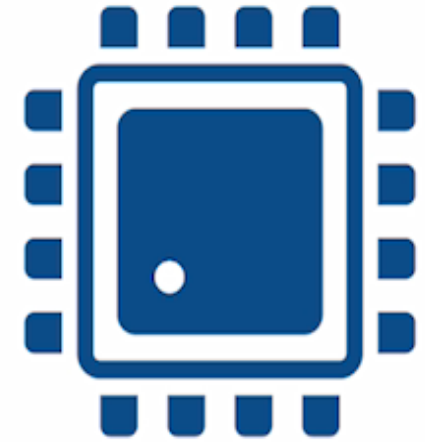
Embedded Devices – Automatic Code Generation



MATLAB Code

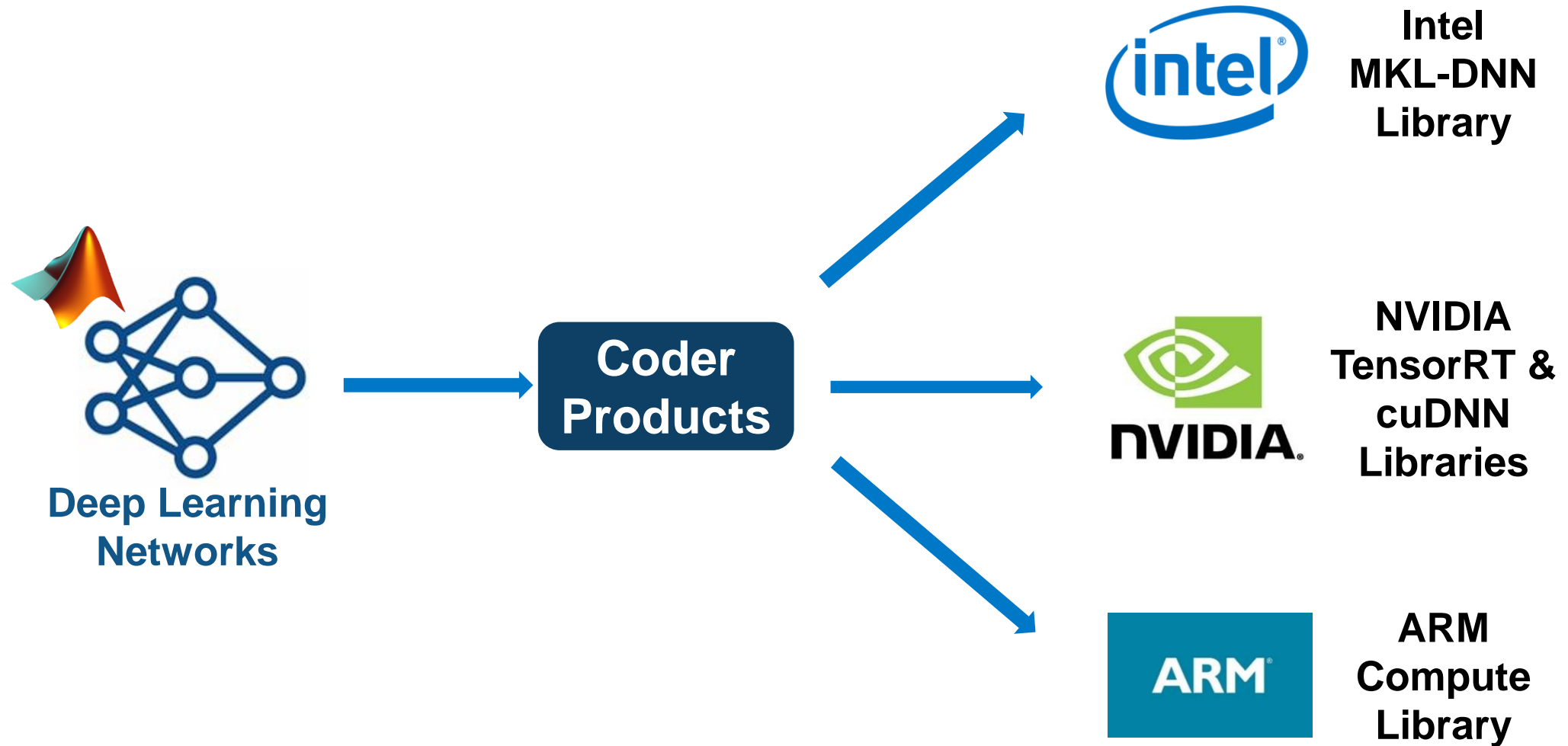


Auto-generated
Code
(C/C++/CUDA)

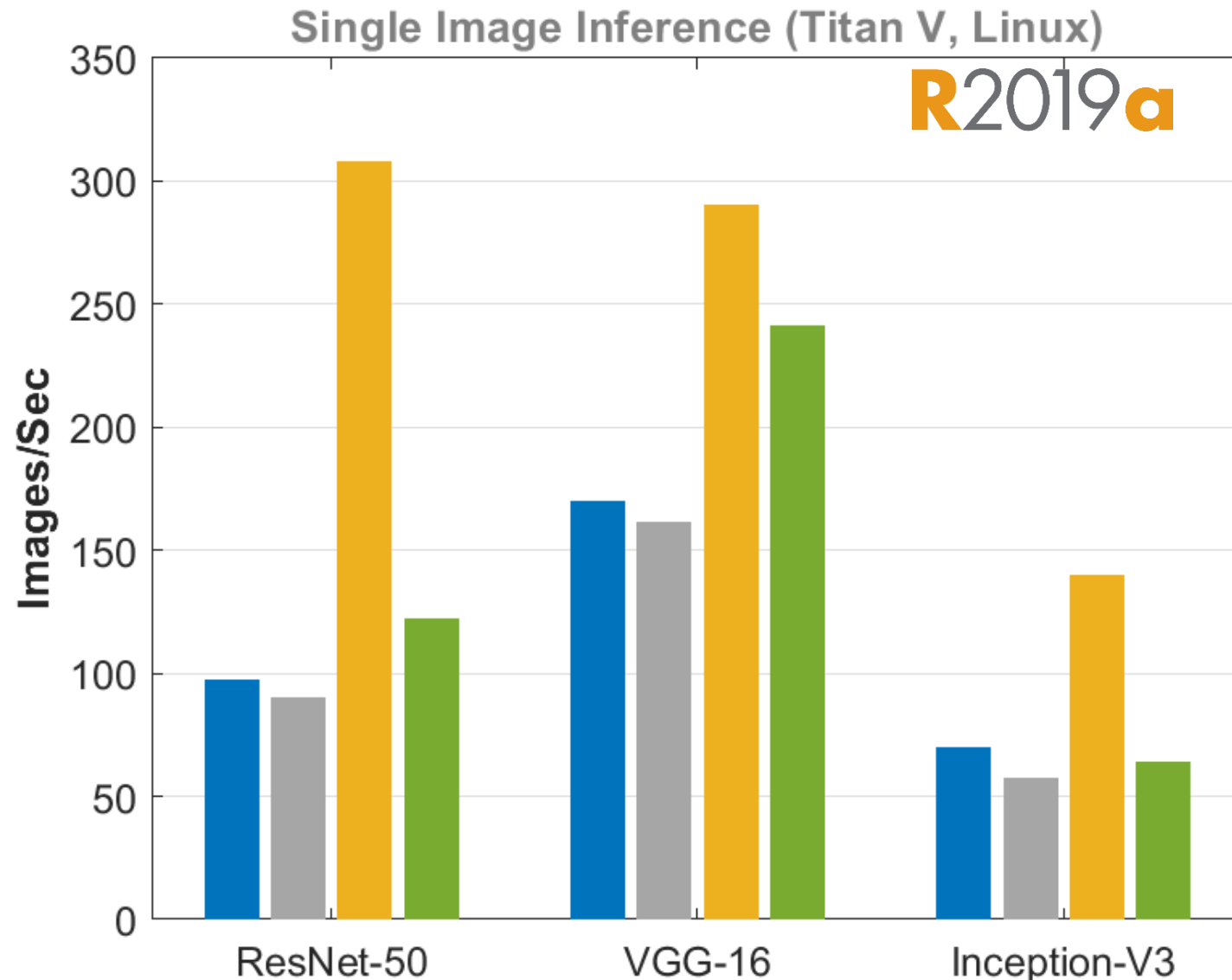


Deployment
Target

Deploying Deep Learning Models for Inference



With GPU Coder, MATLAB is fast

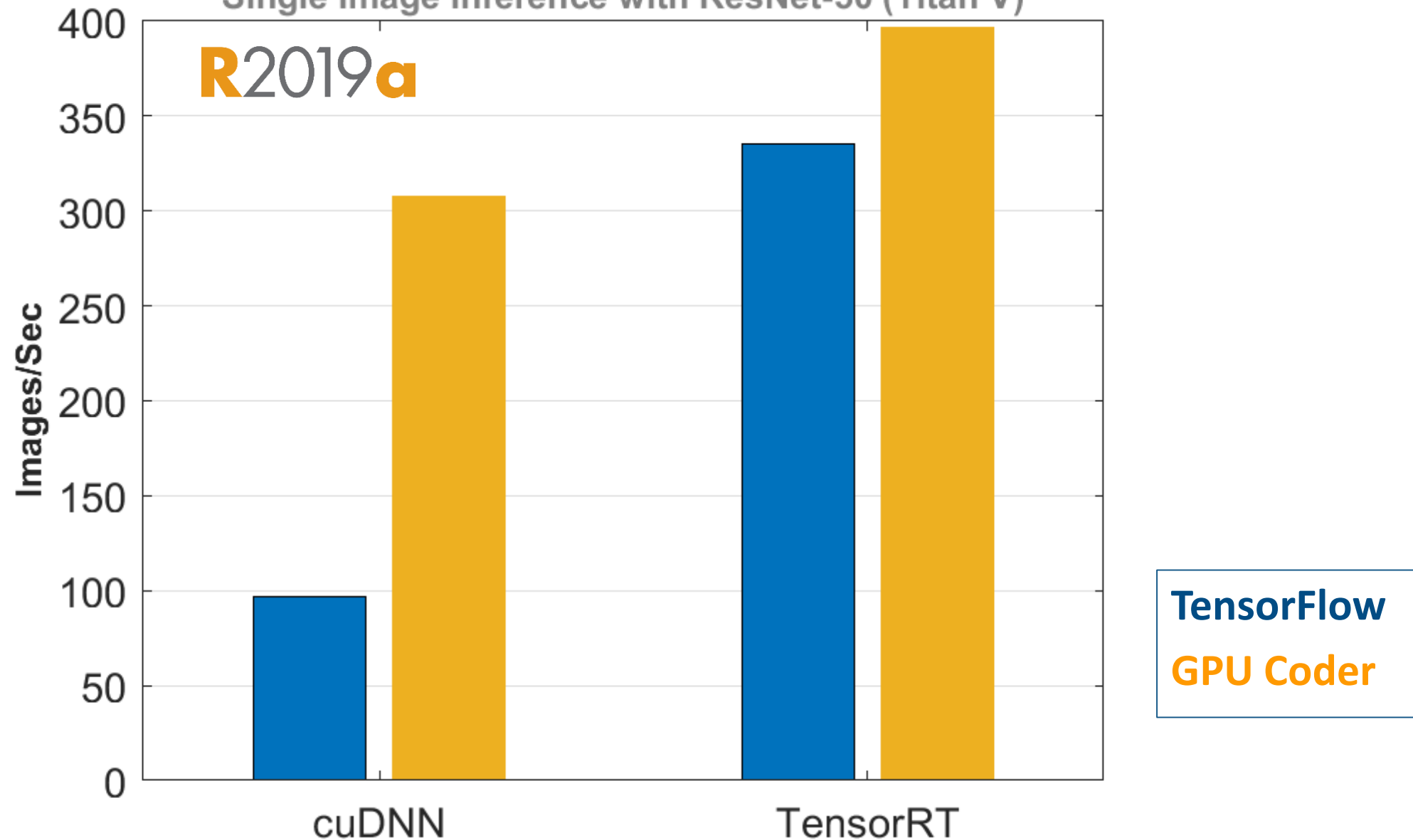


GPU Coder is faster than TensorFlow, MXNet and Pytorch

- TensorFlow
- MXNet
- GPU Coder
- PyTorch

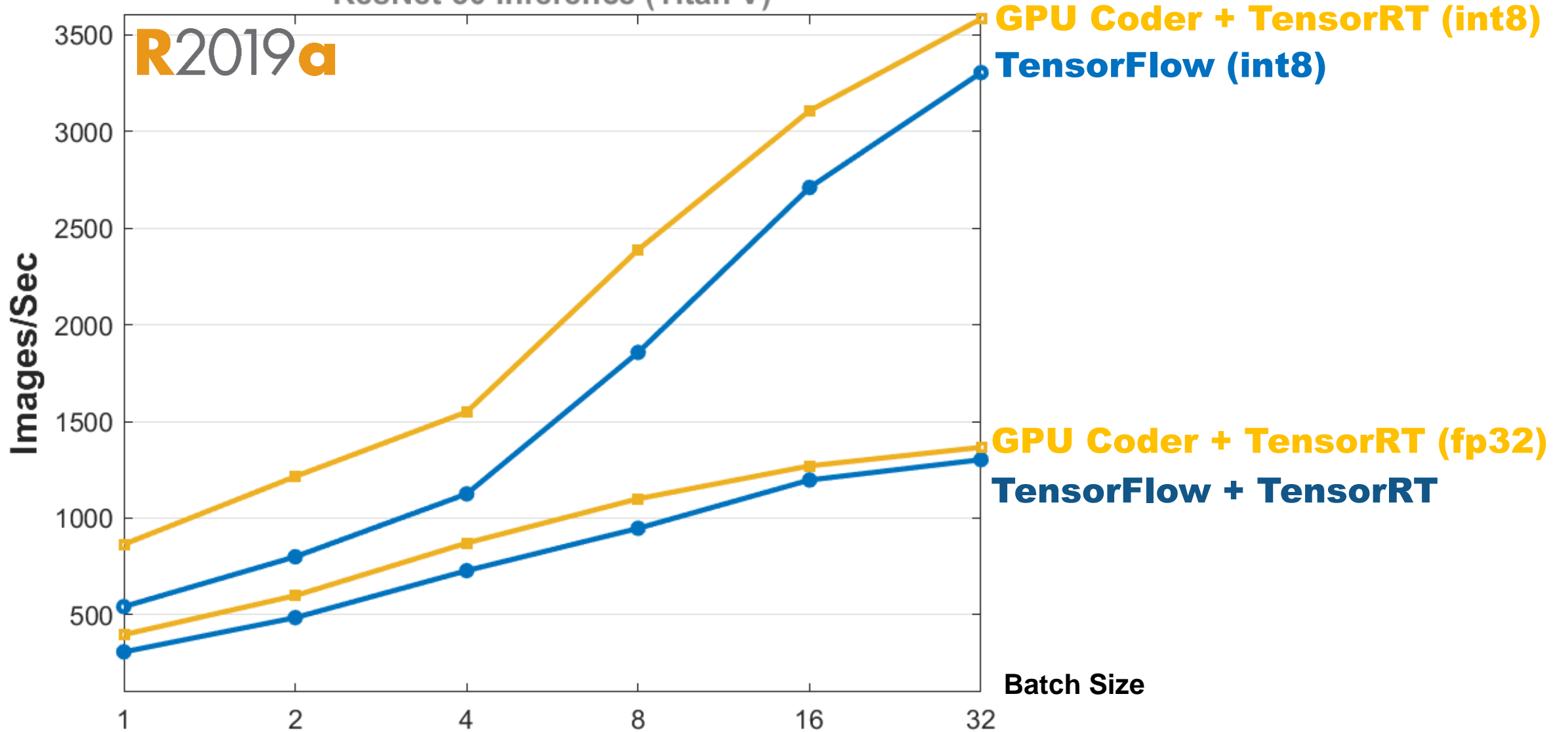
TensorRT speeds up inference for TensorFlow and GPU Coder

Single Image Inference with ResNet-50 (Titan V)

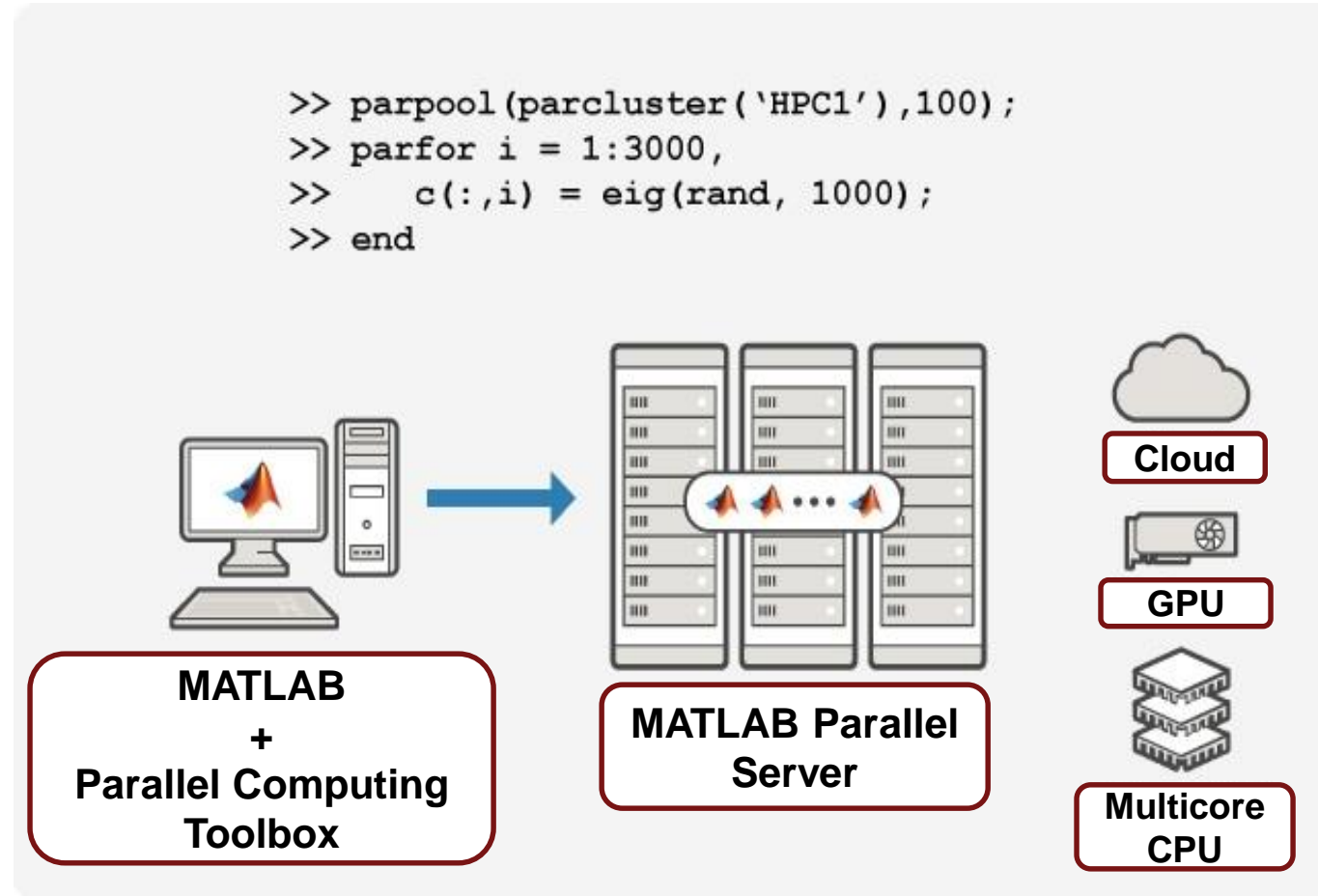


Even higher Speeds with Integer Arithmetic (int8)

ResNet-50 Inference (Titan V)

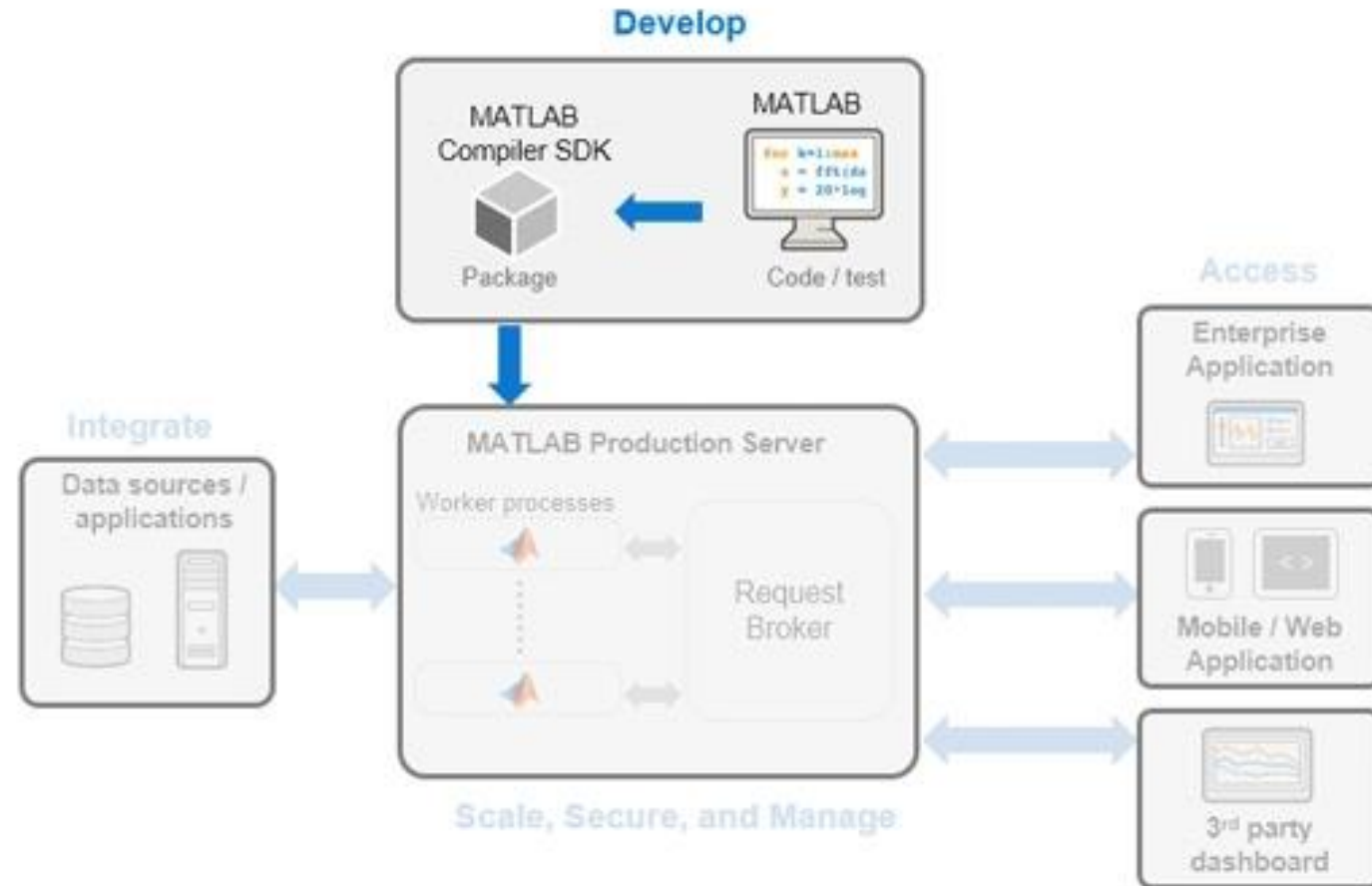


Enterprise Deployment



Run thousands of simulations in parallel with MATLAB Parallel Server to save hours of training time.

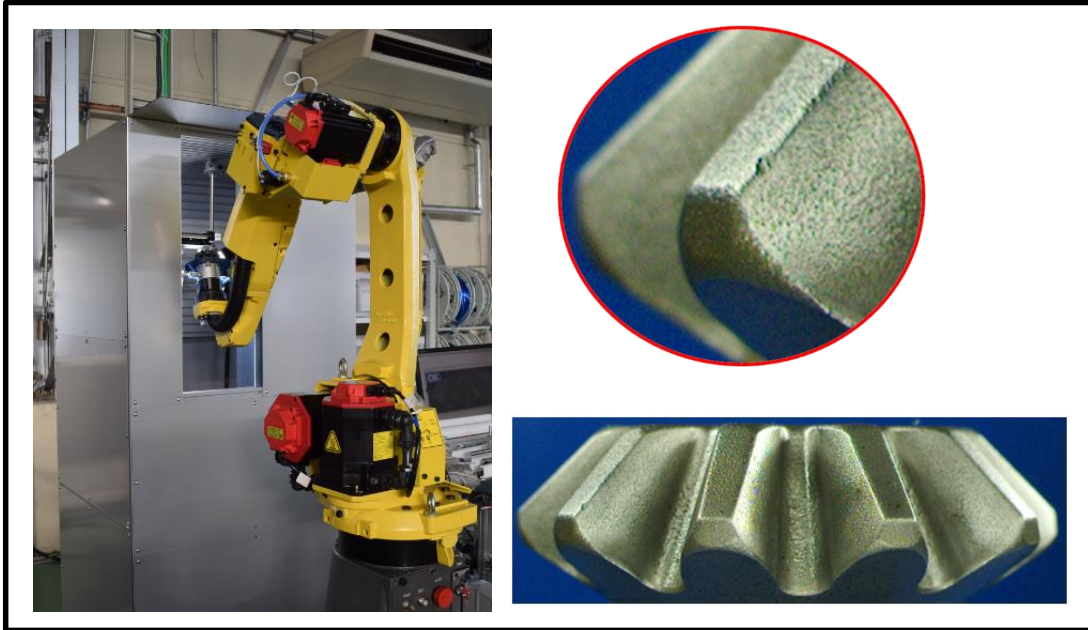
Enterprise Deployment



Deployment to the cloud with MATLAB Compiler and MATLAB Production Server

Musashi Seimitsu Industry Co.,Ltd.

Detect Abnormalities in Automotive Parts



Automated visual inspection of 1.3 million
bevel gear per month

MATLAB use in project:

- Preprocessing of captured images
- Image annotation for training
- Deep learning based analysis
 - Various transfer learning methods
(Combinations of CNN models, Classifiers)
 - Estimation of defect area using Class Activation Map (CAM)
 - Abnormality/defect classification
- Deployment to NVIDIA Jetson using GPU Coder



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