

Deep Learning for Image Processing and Computer Vision with MATLAB

Sarah Hung
Application Engineer

2018/11/28

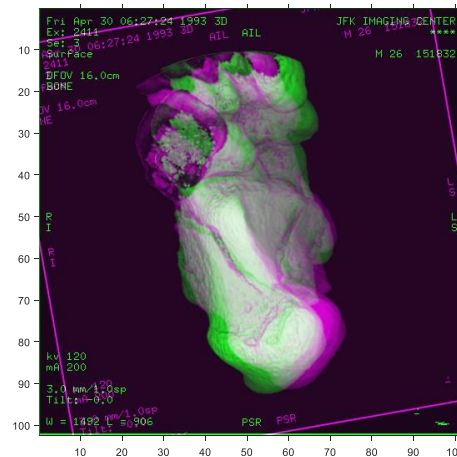
Agenda

- New Features in Image Processing
 - 3-D Image Processing
 - Deep Learning for Image Processing
- Deep Learning for Computer Vision
- Deep Learning Network Deployment

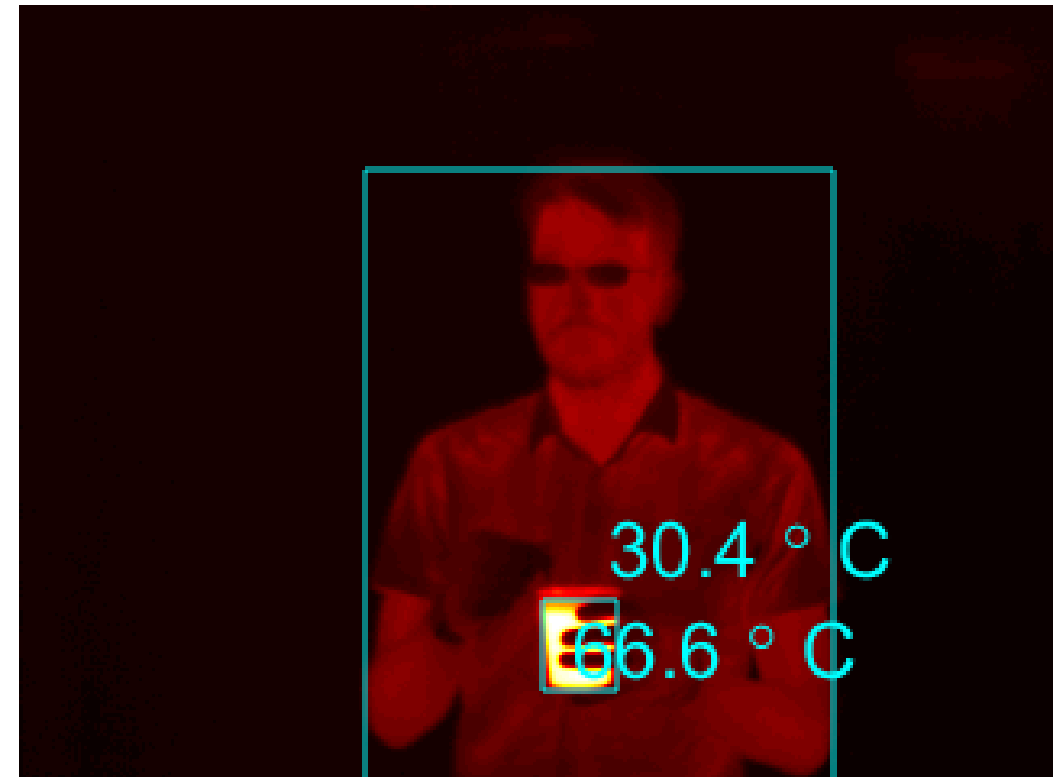
Image Processing Toolbox

Accelerate image processing and algorithm development

- Image display and exploration
- Image enhancement
- Image analysis
- Morphological operations
- Image registration



Thermal Image Segmentation



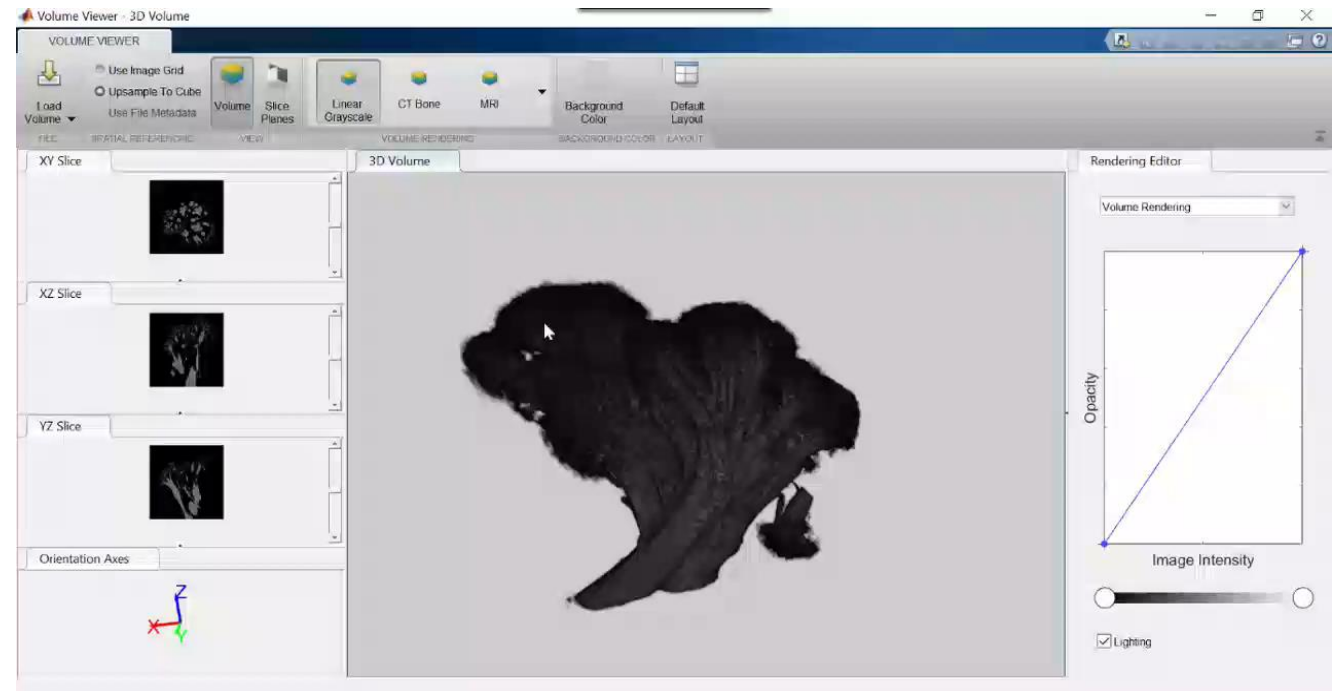
3-D Image Processing

Over 40 functions support 3-D volumetric image processing

Capabilities Includes:

- Image arithmetic
- Morphology
- Segmentation
- Geometric transforms
- Enhancement

Volume Viewer App for exploration



Segment Lungs from 3-D Chest Scan

- Create seed mask
- Active contour technique

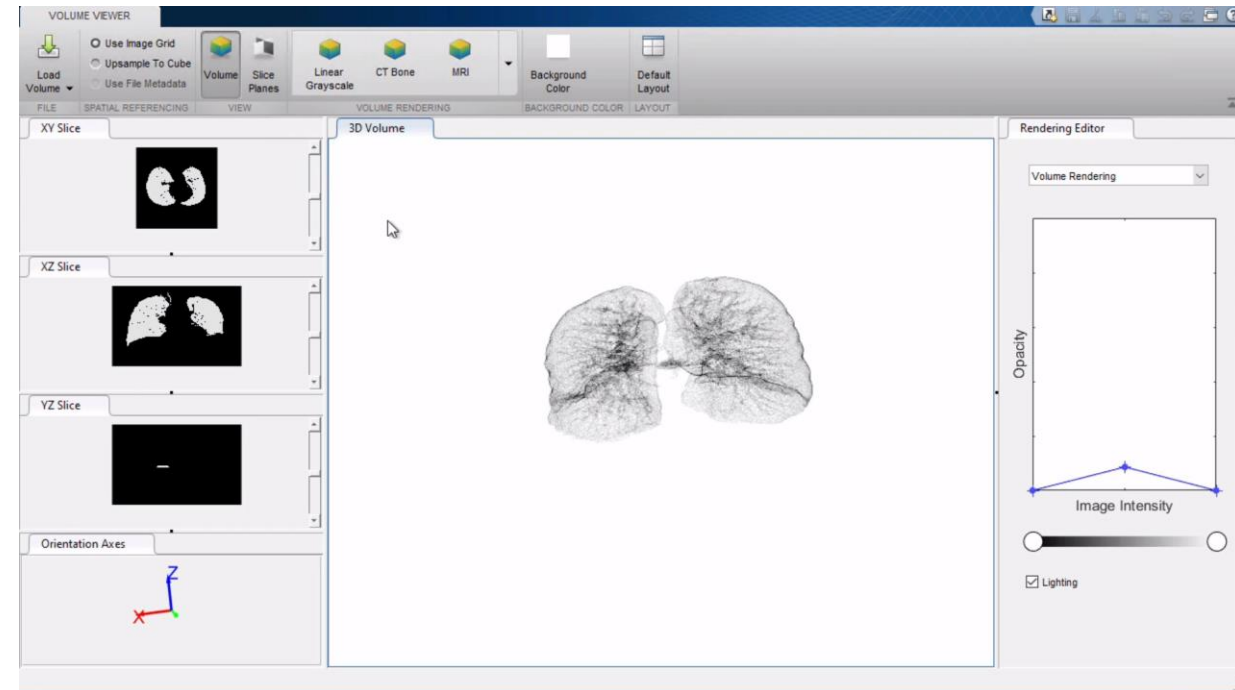
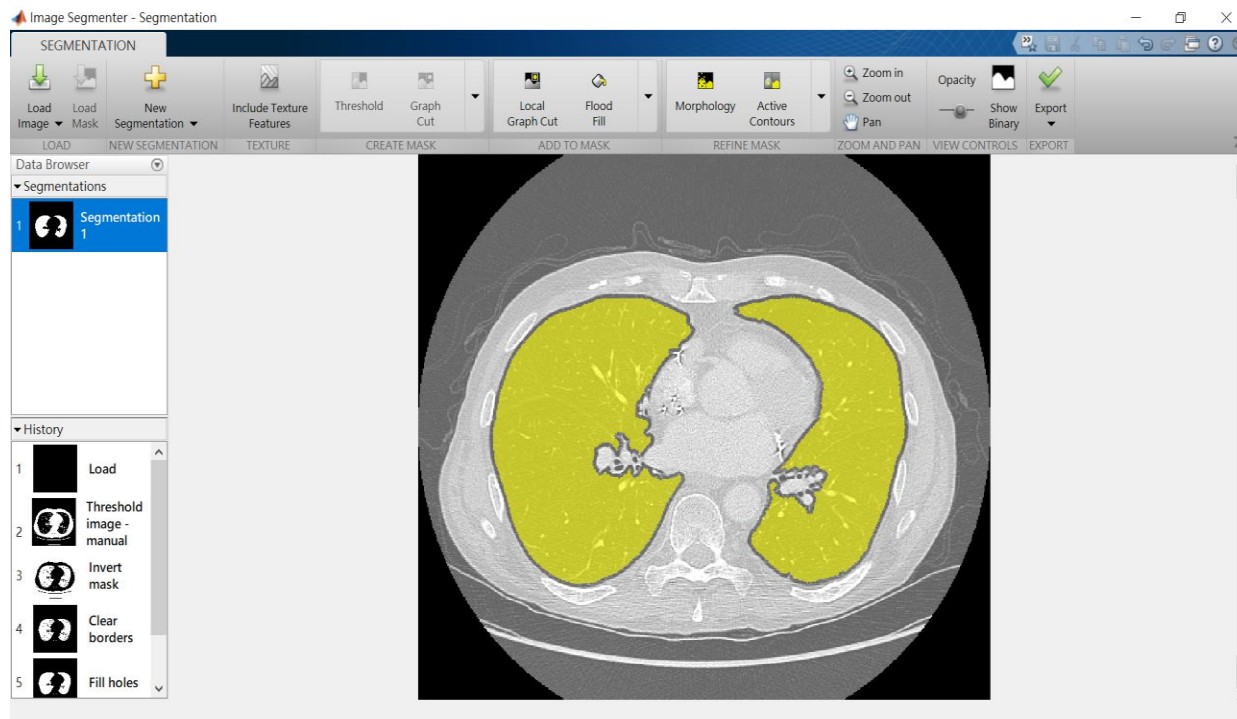
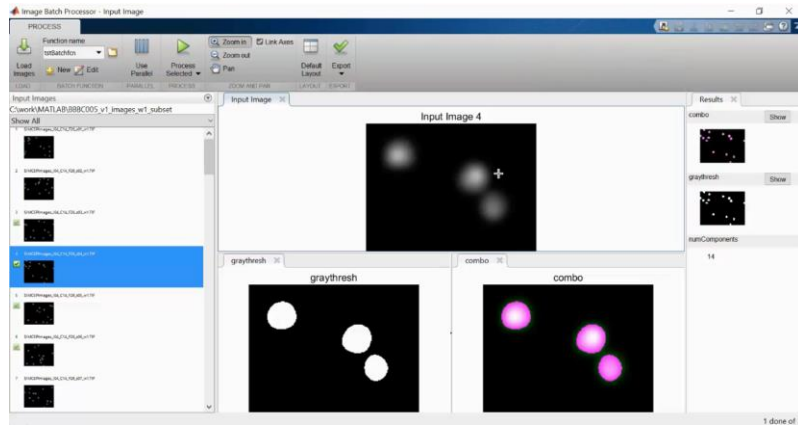
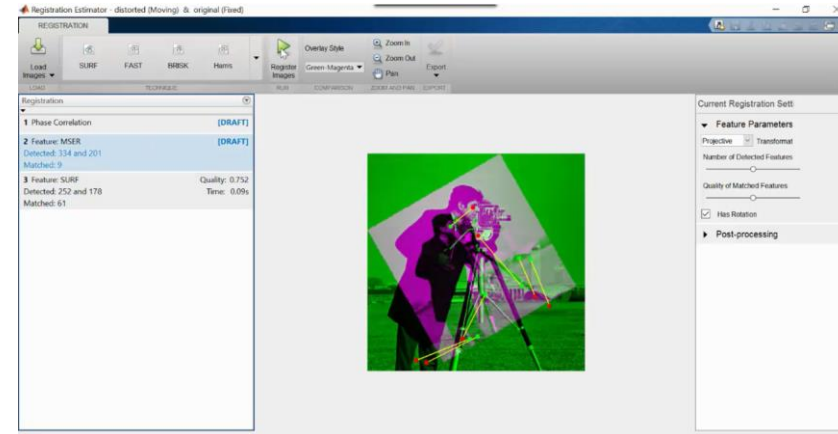


Image Segmenter & Volume Viewer App

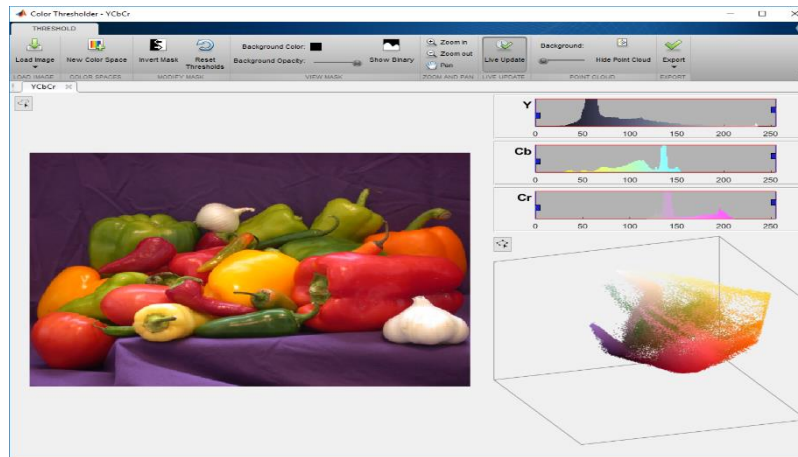
Image Processing Apps



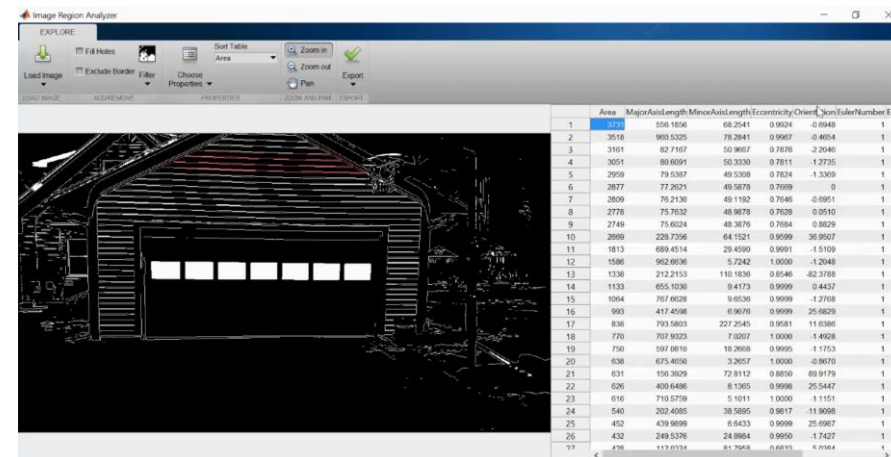
Batch Processing



Registration Estimation

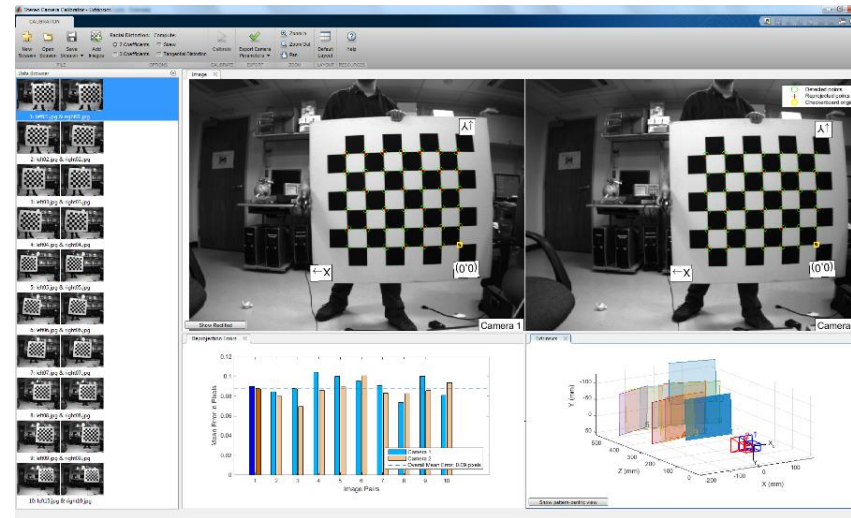
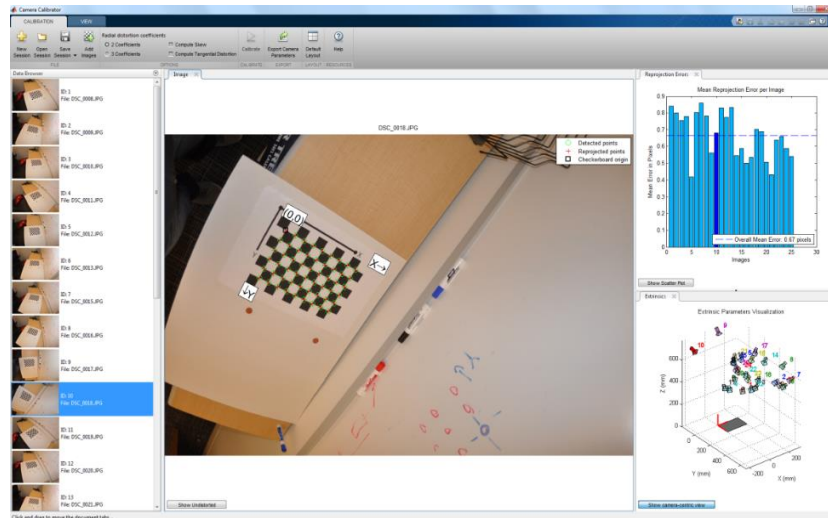


Color Thresholder



Region Analysis

Computer Vision Apps



Single and Stereo Camera Calibration

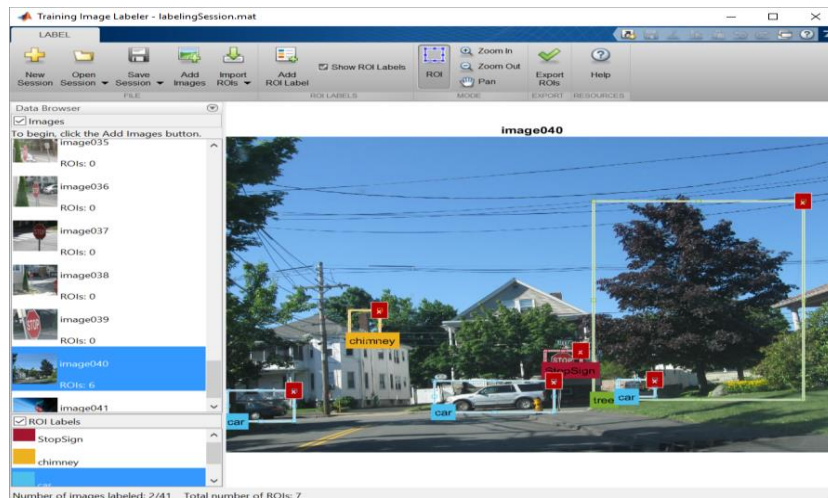
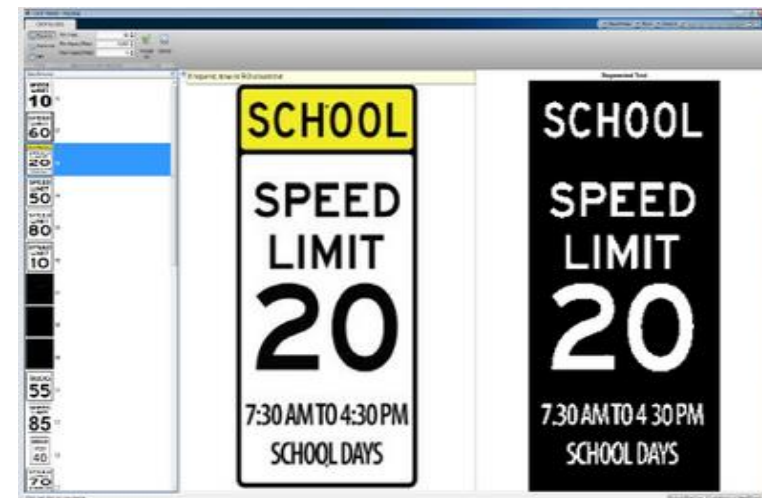


Image and Video Labeler




OCR Training

Deep Learning for Image Processing

Perform image processing using convolutional neural networks

[Documentation](#) [All](#) [Examples](#) [Functions](#) [Apps](#)

 CONTENTS




Image Processing Toolbox

Perform image processing, analysis, and algorithm development

Image Processing Toolbox™ provides a comprehensive set of reference-standard algorithms and workflow apps for image processing, analysis, visualization, and algorithm development. You can perform image segmentation, image enhancement, noise reduction, geometric transformations, image registration, and 3D image processing.

Image Processing Toolbox apps let you automate common image processing workflows. You can interactively segment image data, compare image registration techniques, and batch-process large datasets. Visualization functions and apps let you explore images, 3D volumes, and videos; adjust contrast; create histograms; and manipulate regions of interest (ROIs).

You can accelerate your algorithms by running them on multicore processors and GPUs. Many toolbox functions support C/C++ code generation for desktop prototyping and embedded vision system deployment.

Getting Started

Learn the basics of Image Processing Toolbox

Import, Export, and Conversion

Image data import and export, conversion of image types and classes

Display and Exploration

Interactive tools for image display and exploration

Geometric Transformation and Image Registration

Scale, rotate, perform other N-D transformations, and align images using intensity correlation, feature matching, or control point mapping

Image Filtering and Enhancement

Contrast adjustment, morphological filtering, deblurring, ROI-based processing

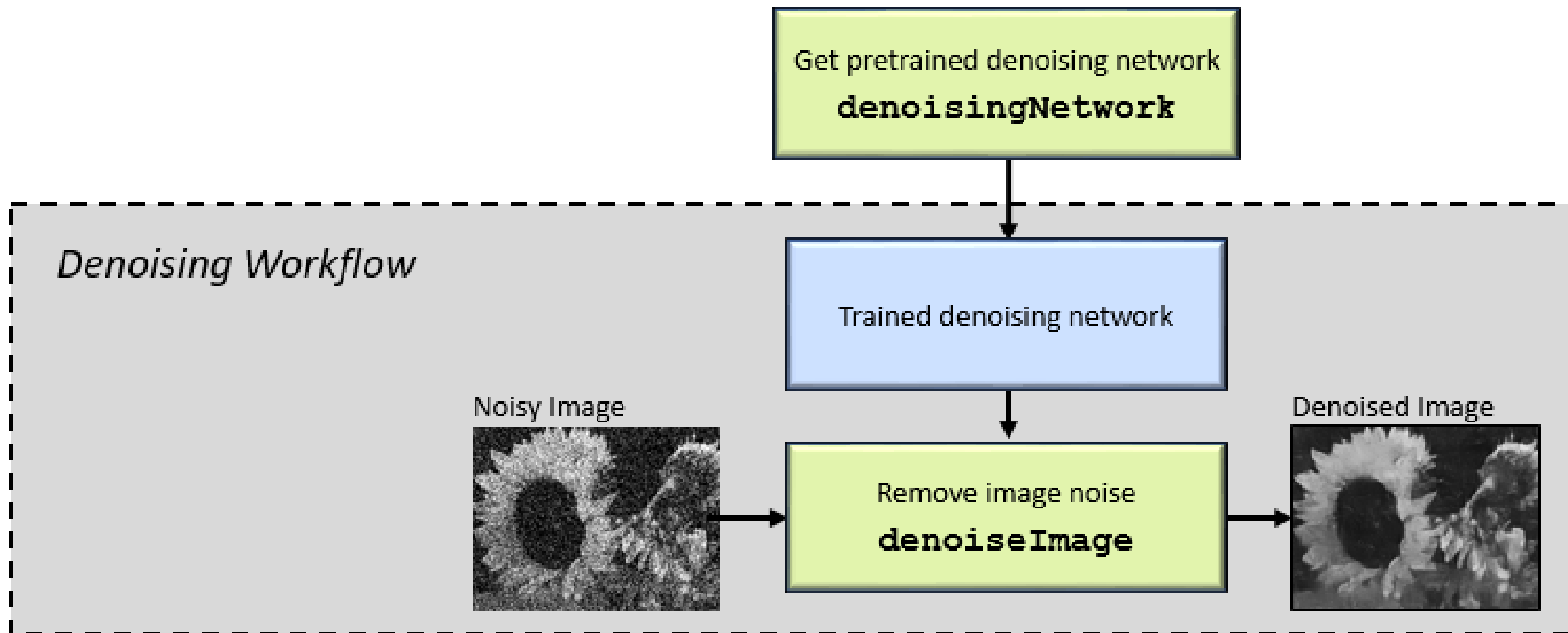
Image Segmentation and Analysis

Region analysis, texture analysis, pixel and image statistics

Deep Learning for Image Processing

Apply Denoising Neural Networks

- Built-in pretrained **DnCNN network** to remove Gaussian noise.



Train Denoising Neural Networks

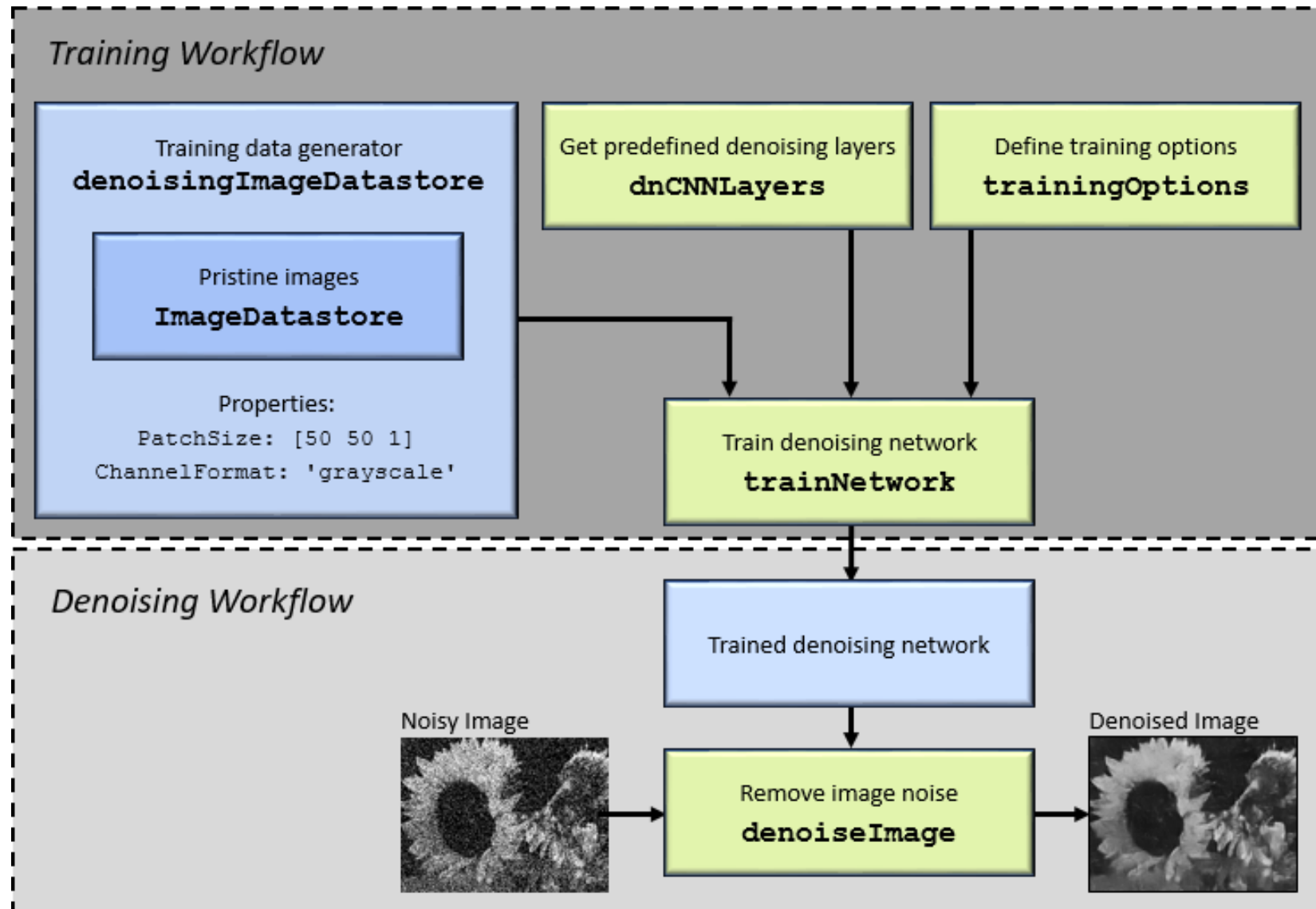


Image Super-Resolution – (1)

- Very-Deep Super-Resolution (VDSR) neural network.
- Estimate a high-resolution image from a single low-resolution image.

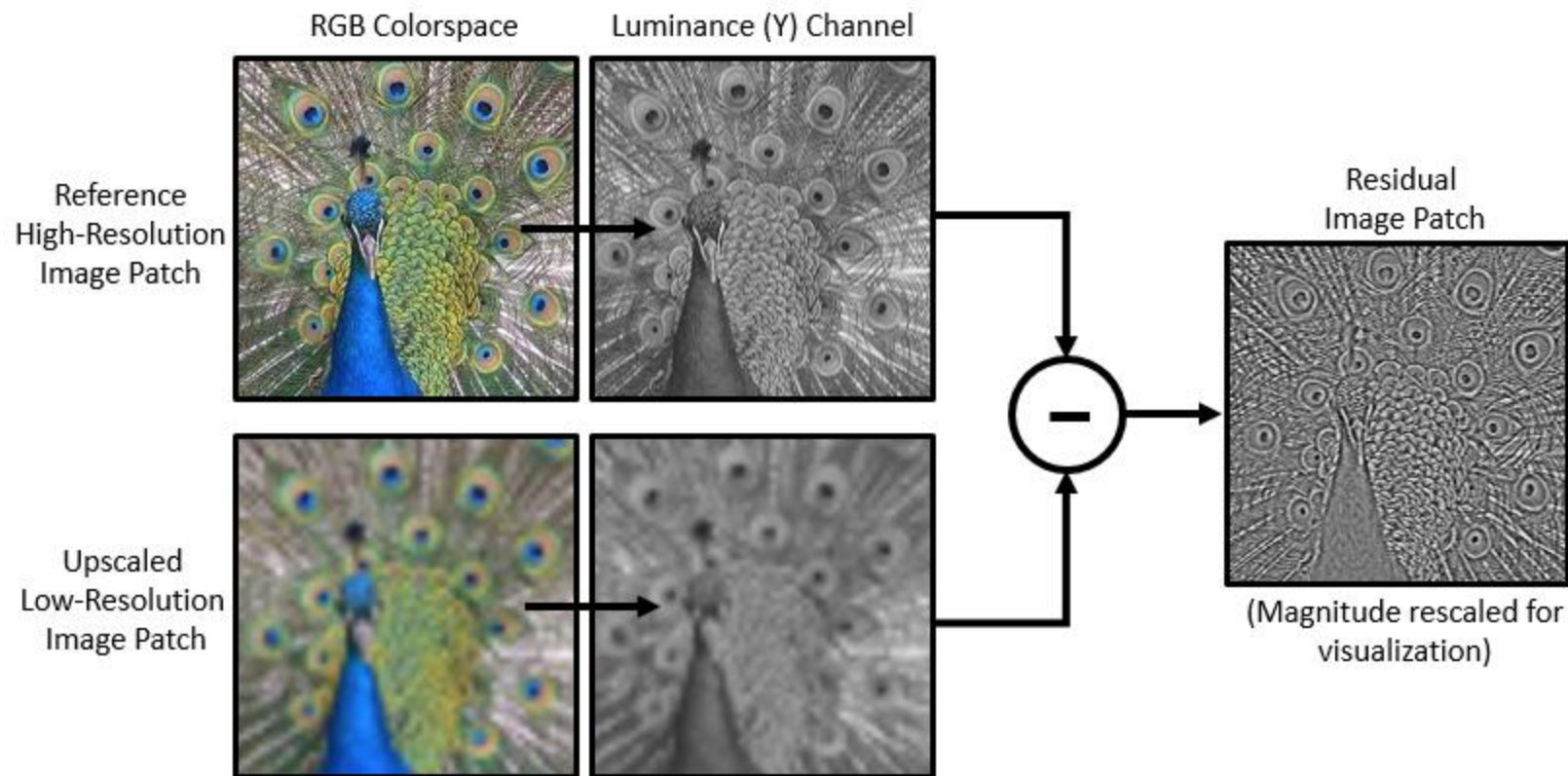


Image Super-Resolution – (2)

High-Resolution Results Using Bicubic Interpolation (Left) vs. VDSR (Right)



Technical Articles and Newsletters

Search Technical Articles

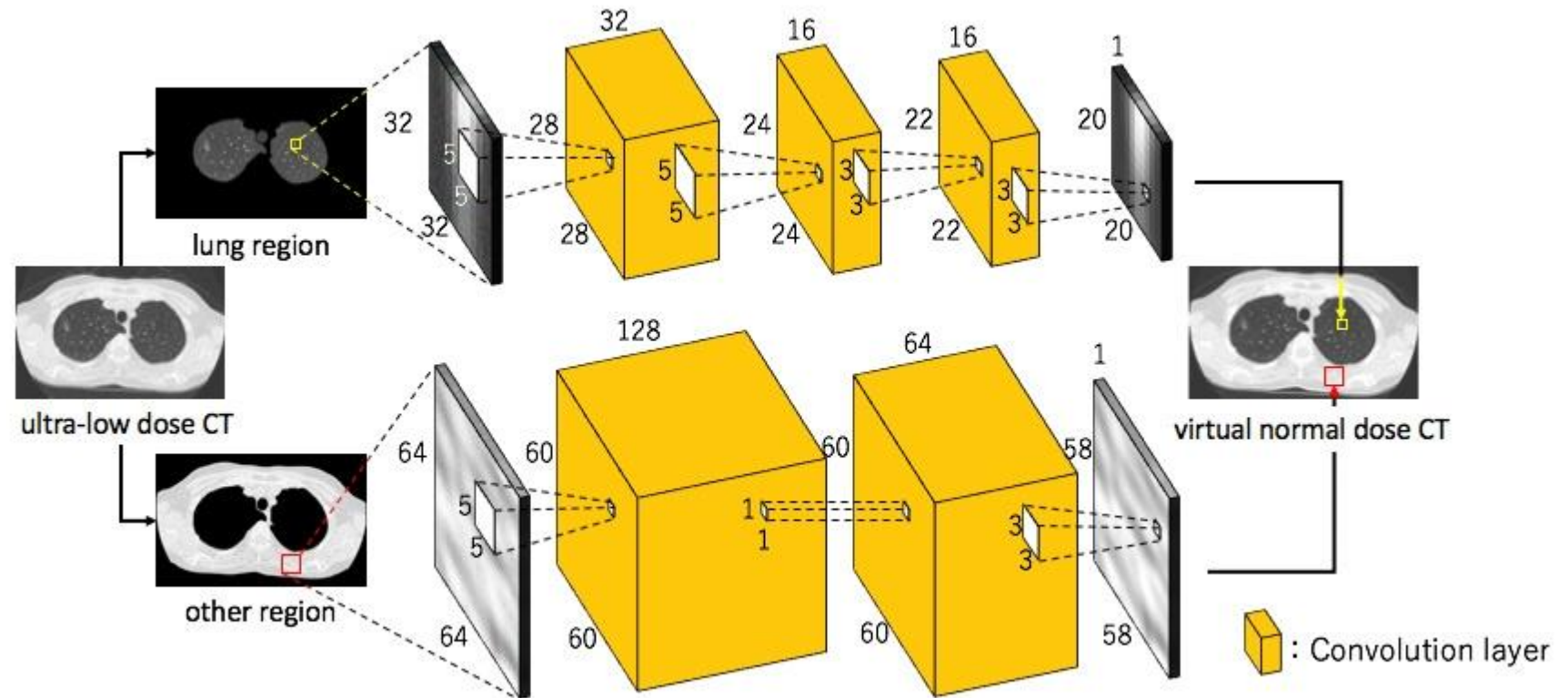
Overview | Search Technical Articles | Newsletters ▼ | Cleve's Corner Collection | Sign Up

Using Deep Learning to Reduce Radiation Exposure Risk in CT Imaging

By Dr. Ryohei Nakayama, Ritsumeikan

Because they produce 3D images of organs, CT scans have a significantly greater diagnostic value than X-rays. However, exposure to potentially harmful radiation together by computer software. As a result, a typical CT scan delivers 7 milliseverts (mSv), 350 times higher risk; guidelines limit the radiation dose

Medical researchers want to limit radiation exposure. One approach is the use of ultra-low-dose CT. A principal drawback of ultra-low-dose CT is that it is difficult for physicians to see organs, fat, and



Agenda

- New Features in Image Processing
 - 3-D Image Processing
 - Deep Learning for Image Processing
- Deep Learning for Computer Vision
- Deep Learning Network Deployment

Deep Learning is a Neural Network

Inspired by the structure and function of the brain

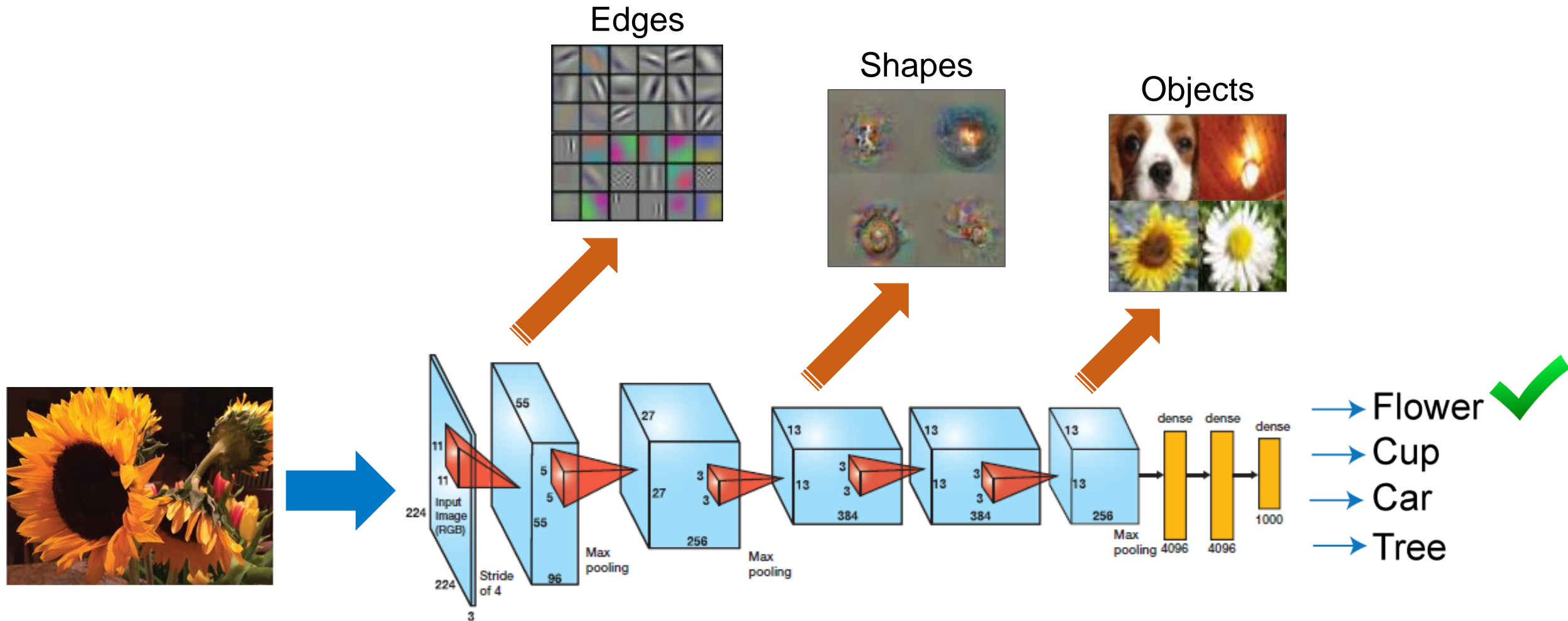
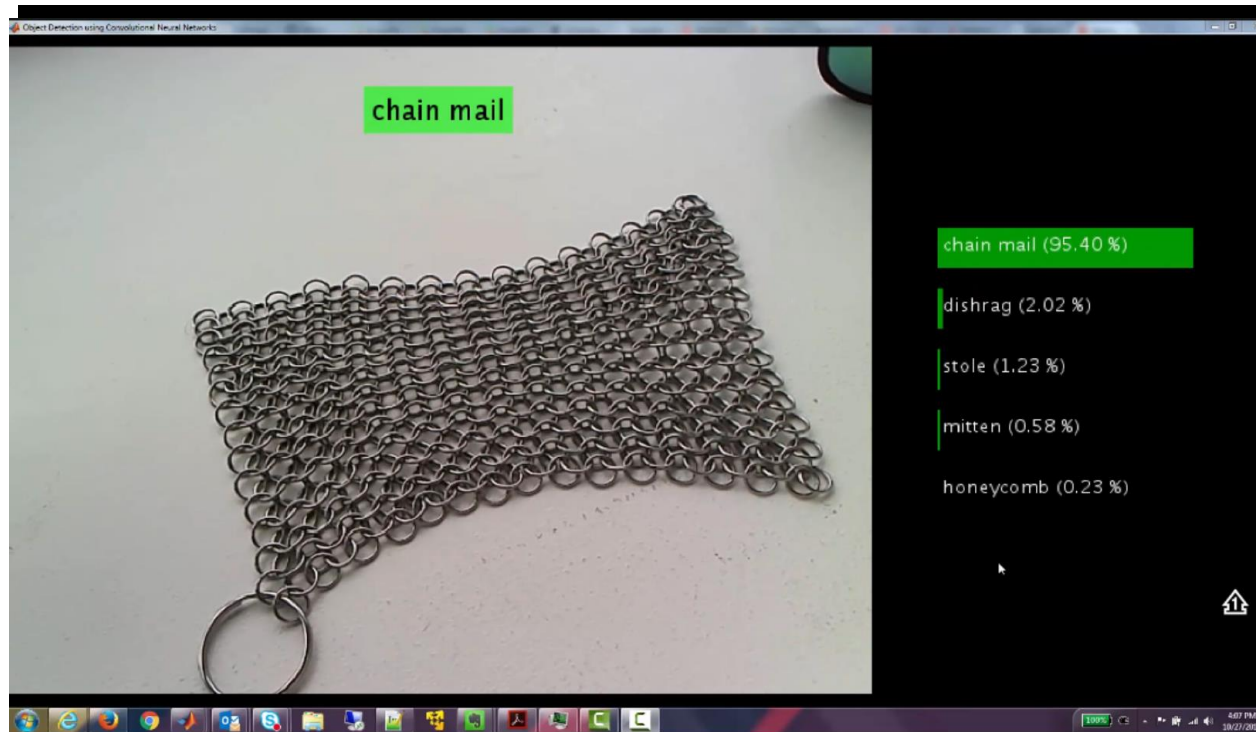


Image Classification using Deep Learning



```
camera = webcam; % Connect to the camera
nnet = alexnet; % Load the neural net

while true
    picture = camera.snapshot; % Take a picture
    picture = imresize(picture,[227,227]); % Resize the picture

    label = classify(nnet, picture); % Classify the picture

    image(picture); % Show the picture
    title(char(label)); % Show the label
    drawnow;

end
```

**Training
(GPU)**

Millions of images from 1000 different categories (AlexNet)

Prediction

Real-time object recognition using a webcam connected to a laptop

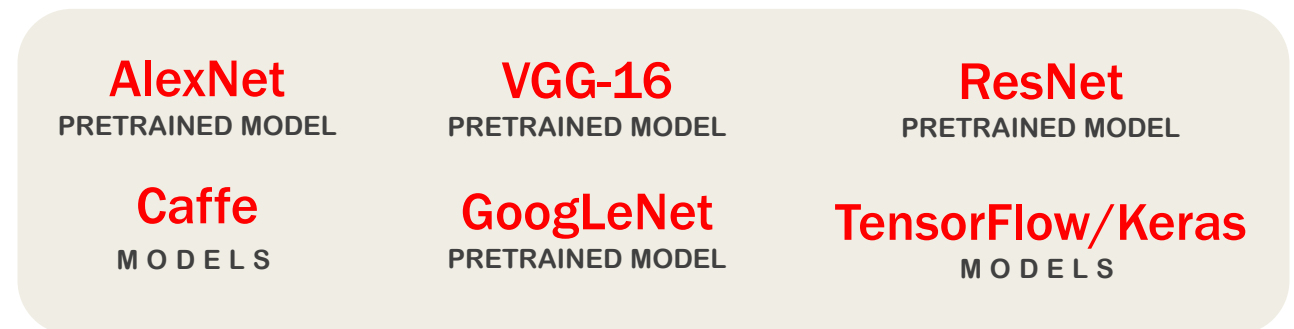
MATLAB Integrates with Open Source Frameworks

- Pretrained Models*
 - AlexNet
 - VGG-16
 - VGG-19
 - GoogLeNet
 - InceptionV3
 - ResNet18
 - ResNet50
 - ResNet101
 - Inception-ResNet-v2
 - SqueezeNet
 - DenseNet-201

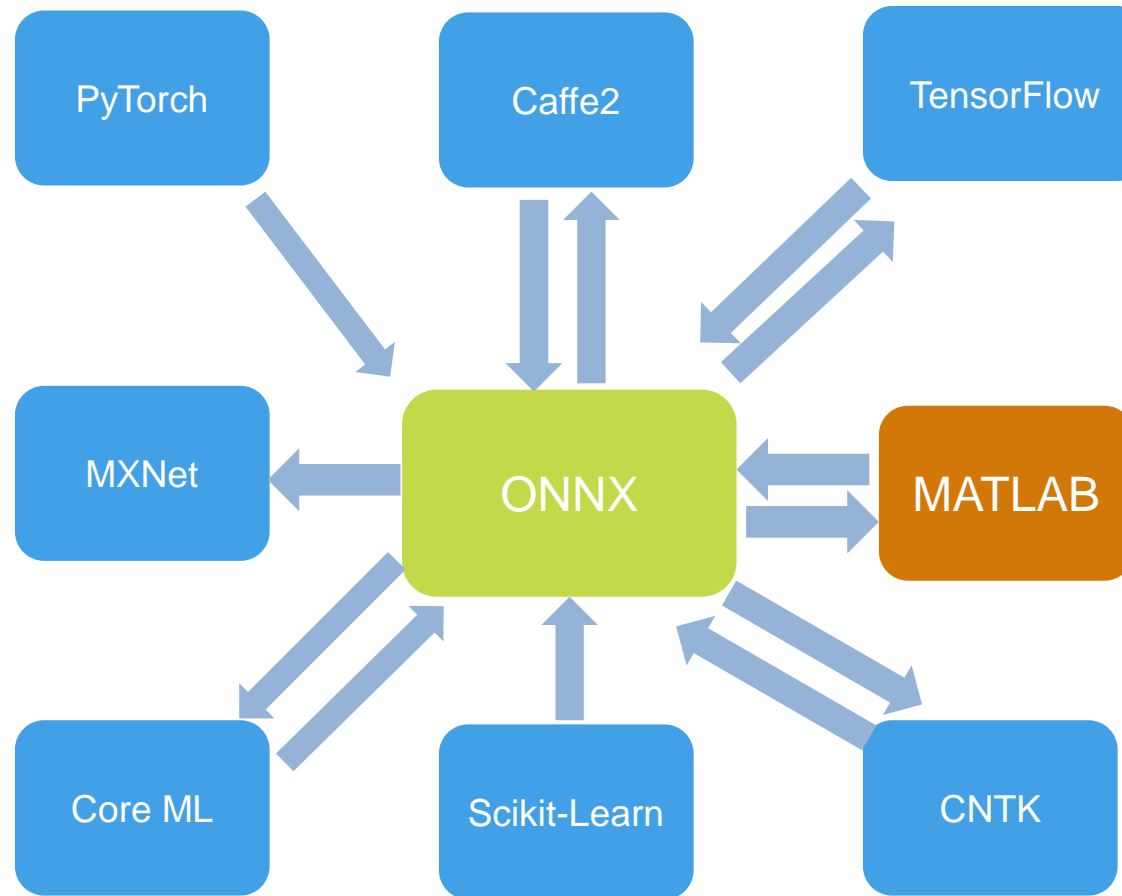
...

* single line of code to access model

- Import Models from Frameworks
 - Caffe Model Importer
 - TensorFlow-Keras Model Importer
- Converter for ONNX Model Format



ONNX is an open format to represent deep learning models

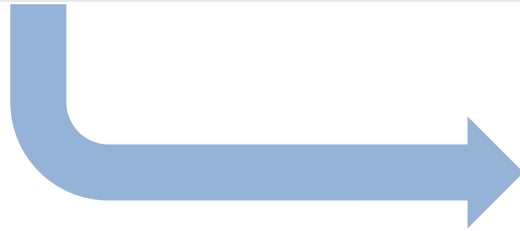


ONNX = Open Neural Network Exchange Format

ONNX Converter

Import from ONNX format

```
modelfile = 'cifarResNet.onnx';
classes = ["airplane" "automobile" "bird"];
net = importONNXNetwork(modelfile,'Output');
analyzeNetwork(net)
```



Export to ONNX format

```
filename = 'fishdetector.onnx';
exportONNXNetwork(net,filename)
```

net
Analysis date: 04-Jul-2018 08:16:54

77 layers 0 warnings 0 errors

ANALYSIS RESULT

	NAME	TYPE	ACTIVATIONS	LEARNABLES
1	Input_input 32x32x3 images	Image Input	32x32x3	-
2	input_Sub Layer adds bias to the input	Bias	32x32x3	-
3	conv1np 16 3x3x3 convolutions with stride [1 1] and padding [1 1 1]	Convolution	32x32x16	Weights 3x3x3x16 Bias 1x1x16
4	BN1np Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16
5	relu1np ReLU	ReLU	32x32x16	-
6	S1U1_conv1 16 3x3x16 convolutions with stride [1 1] and padding [1 1 1]	Convolution	32x32x16	Weights 3x3x16x16 Bias 1x1x16
7	S1U1_BN1 Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16
8	S1U1_relu1 ReLU	ReLU	32x32x16	-
9	S1U1_conv2 16 3x3x16 convolutions with stride [1 1] and padding [1 1 1]	Convolution	32x32x16	Weights 3x3x16x16 Bias 1x1x16
10	S1U1_BN2 Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16
11	add11 Element-wise addition of 2 inputs	Addition	32x32x16	-
12	relu11 ReLU	ReLU	32x32x16	-
13	S1U2_conv1 16 3x3x16 convolutions with stride [1 1] and padding [1 1 1]	Convolution	32x32x16	Weights 3x3x16x16 Bias 1x1x16
14	S1U2_BN1 Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16
15	S1U2_relu1 ReLU	ReLU	32x32x16	-
16	S1U2_conv2 16 3x3x16 convolutions with stride [1 1] and padding [1 1 1]	Convolution	32x32x16	Weights 3x3x16x16 Bias 1x1x16
17	S1U2_BN2 Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16

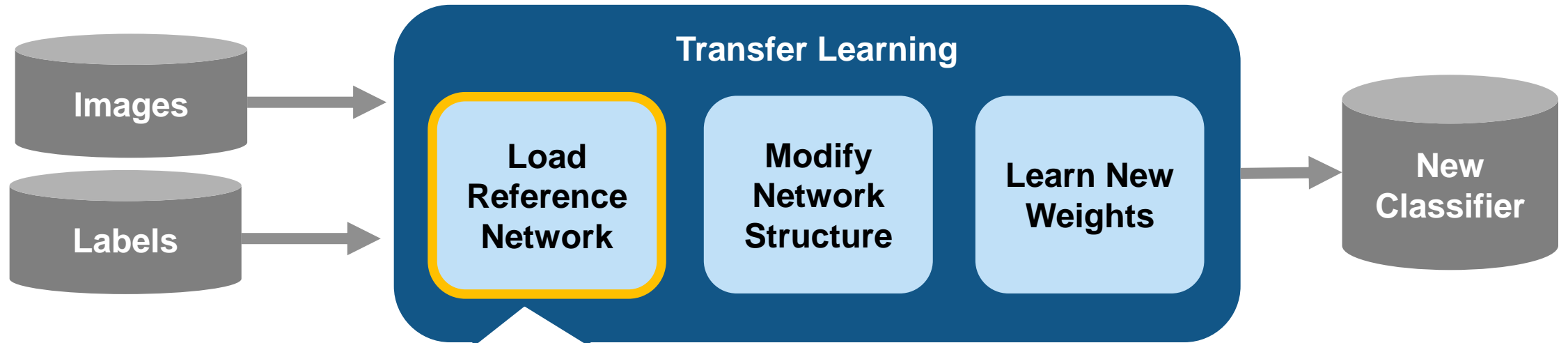
The background of the slide is a close-up photograph of a wooden surface, likely a table or countertop, showing a natural wood grain pattern with various shades of brown and tan. Two rectangular text boxes are overlaid on the center of the image. The top box is dark gray with white text, and the bottom box is white with teal text.

Deep Learning Demo

Image Classification

Transfer Learning

Access Reference Models in MATLAB



Easily Load Reference Networks

Access Models with 1-line of MATLAB Code

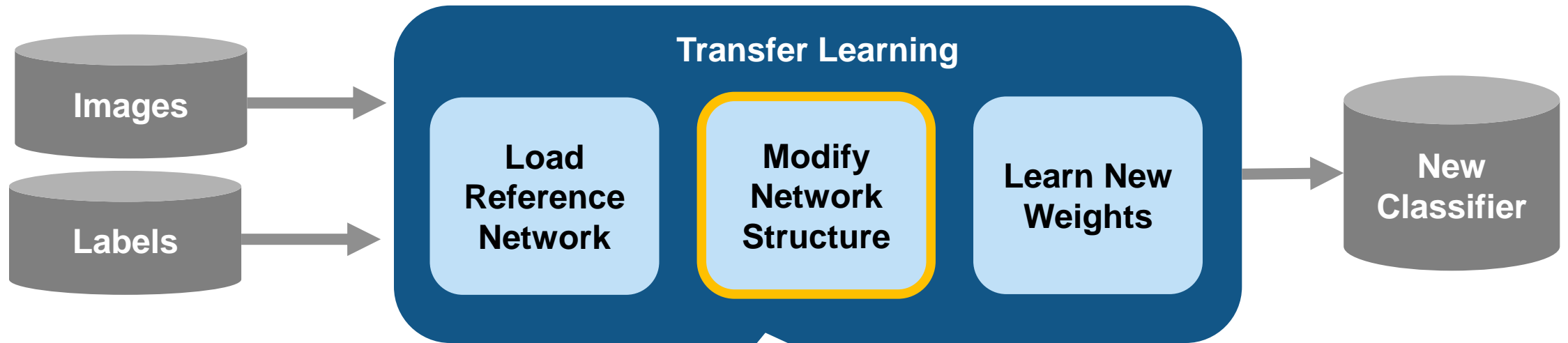
```
Net1 = alexnet
```

```
Net2 = vgg16
```

```
Net3 = vgg19
```

Transfer Learning

Modify Network Structure



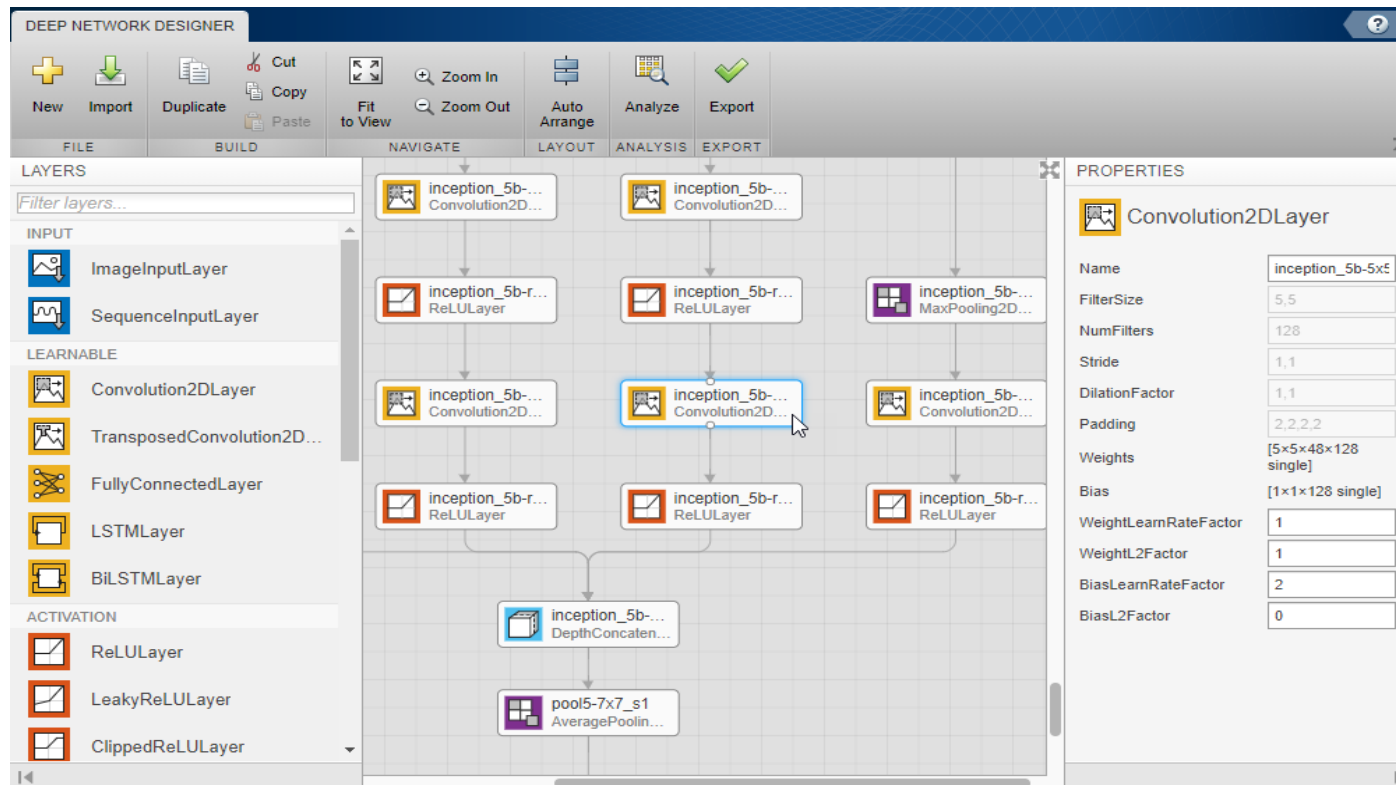
Simple MATLAB API to modify layers:

```
layers(23) = fullyConnectedLayer(5, 'Name', 'fc8');  
layers(25) = classificationLayer('Name', 'VehicleClassifier')
```

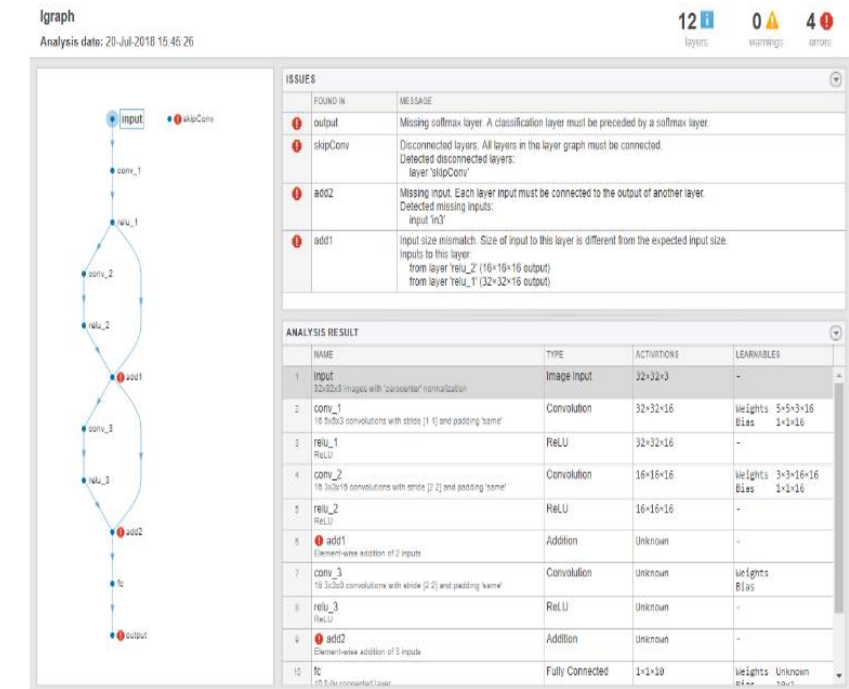
Deep Network Designer App

- Graphically design and modify networks
- Check for errors in network architectures

>> deepNetworkDesigner

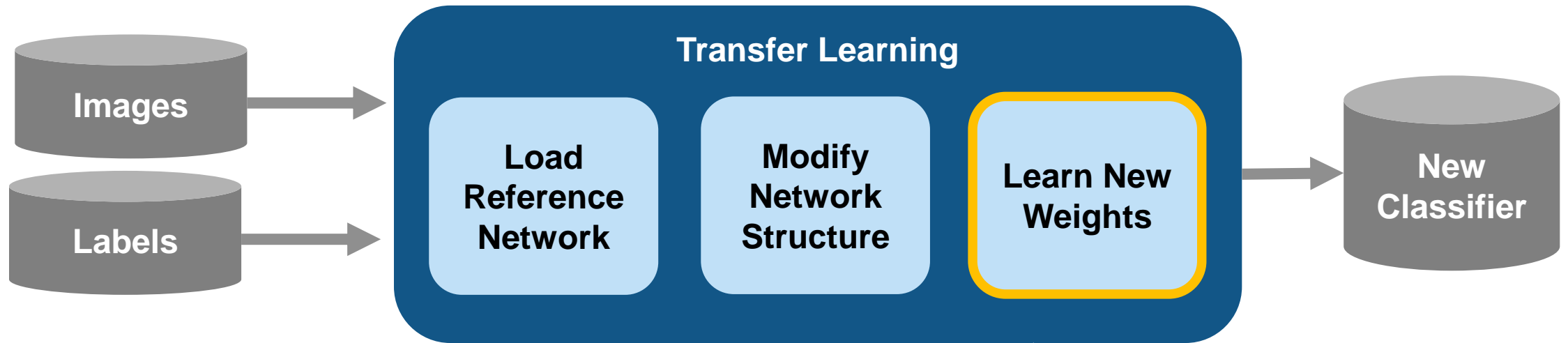


Design deep networks



Check for errors

Transfer Learning Training Network



Train Any Network

```
trainNetwork(datastore, layers, options)
```

Deep Learning on CPU, GPU, Multi-GPU and Clusters

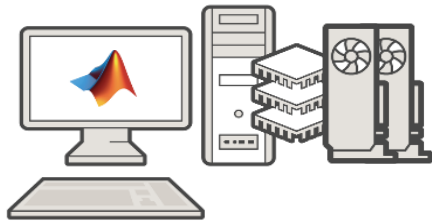
HOW TO TARGET?



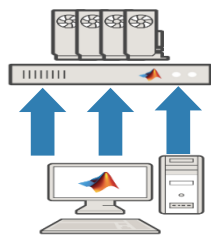
Single
CPU



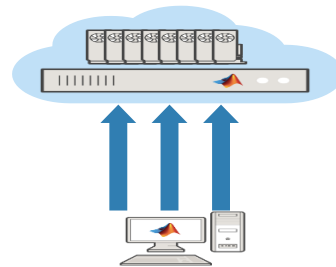
Single CPU
Single GPU



Single CPU, Multiple GPUs



On-prem server with
GPUs



Cloud GPUs
(AWS)

```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...
    'ExecutionEnvironment', 'auto' );
```

```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...
    'ExecutionEnvironment', 'multi-gpu' );
```

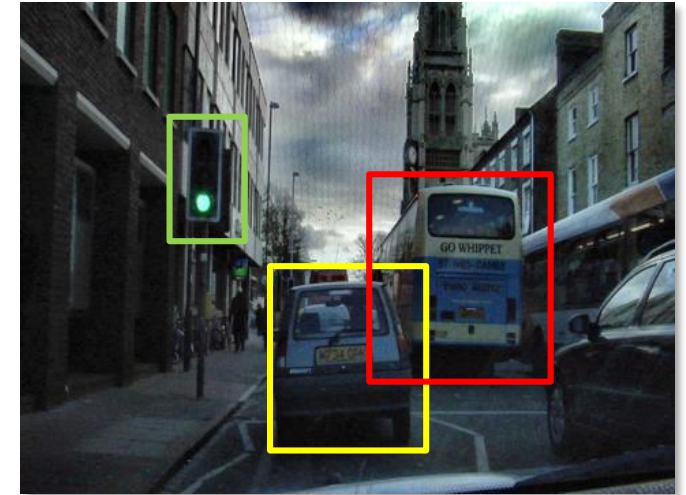
```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...
    'ExecutionEnvironment', 'parallel' );
```

Object Detection

Original Image



ROI detection



Pixel classification



Semantic Segmentation



CamVid Dataset

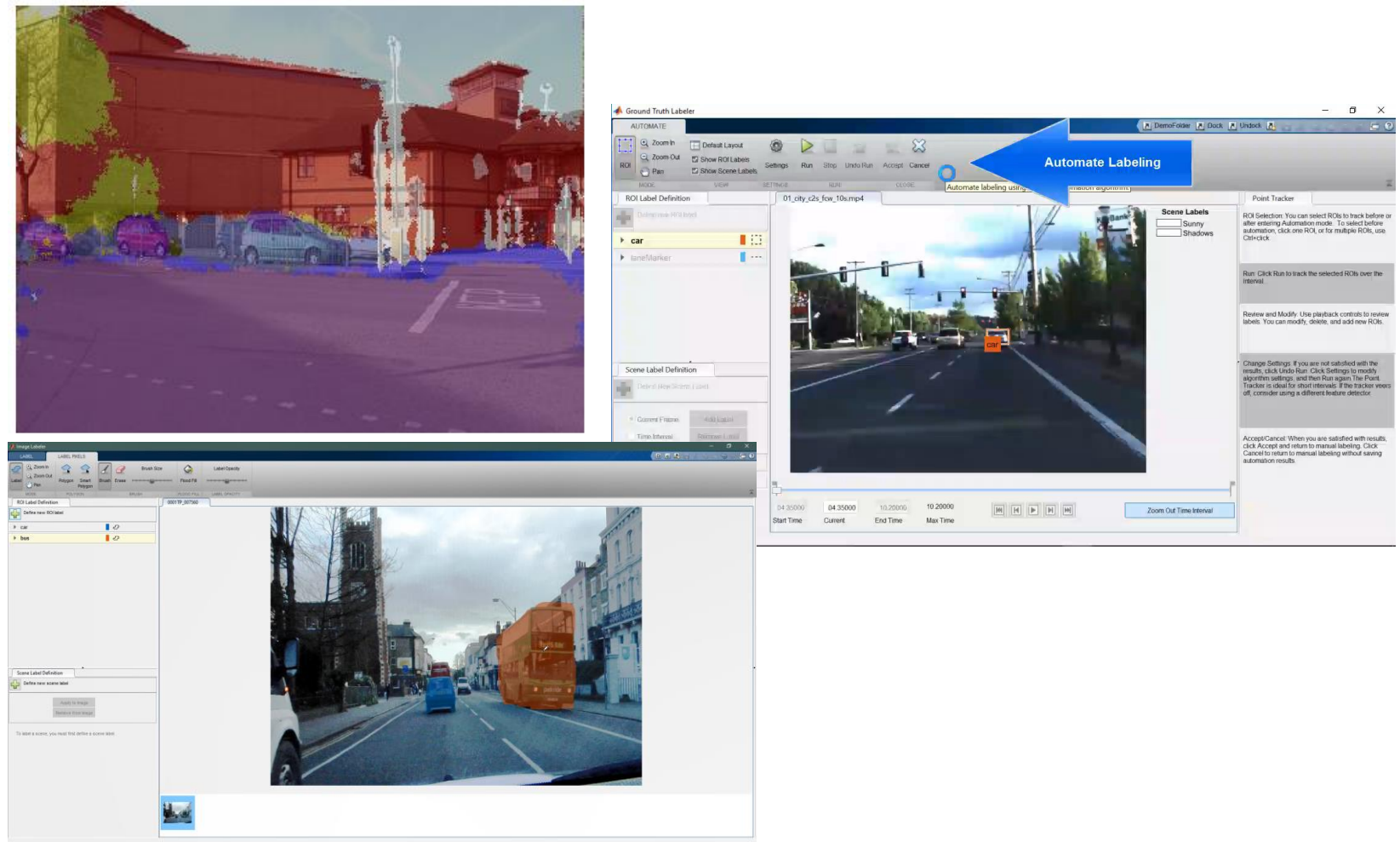
1. Segmentation and Recognition Using Structure from Motion Point Clouds, ECCV 2008
2. Semantic Object Classes in Video: A High-Definition Ground Truth Database ,Pattern Recognition Letters

Ground truth Labeling

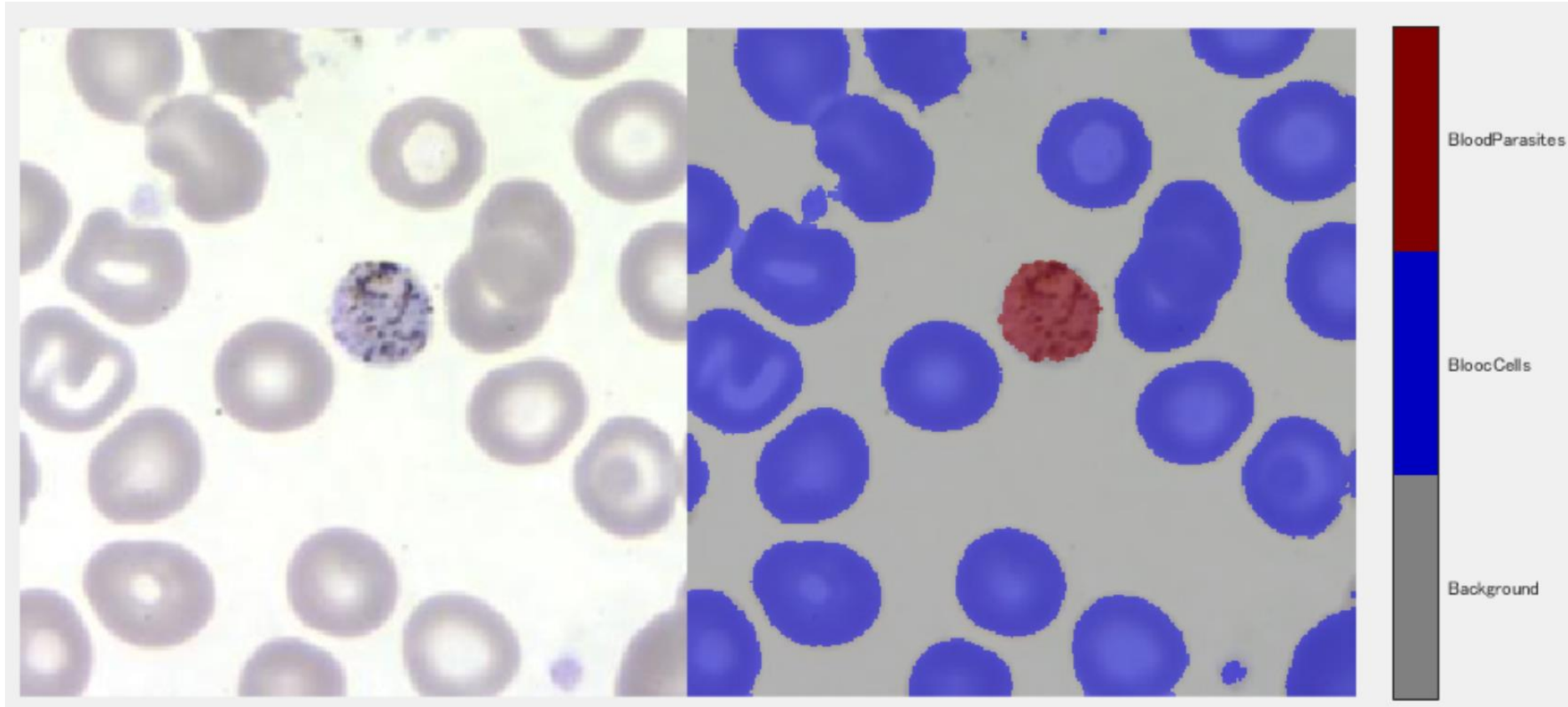
“How do I *label* my data?”

**New App for
Ground Truth
Labeling**

Label pixels
and regions for
semantic
segmentation



Application for Semantic Segmentation



[Medical Image Segmentation Using SegNet](#)

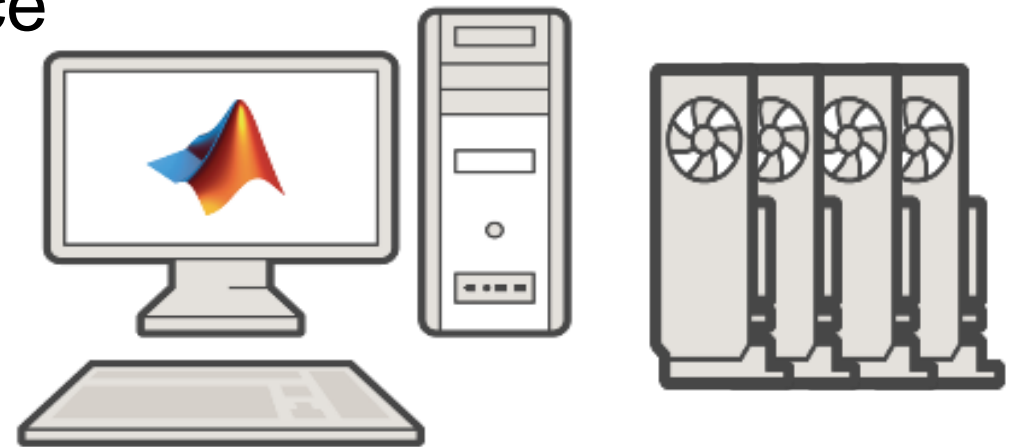
Agenda

- New Features in Image Processing
 - 3-D Image Processing
 - Deep Learning for Image Processing
- Deep Learning for Computer Vision
- Deep Learning Network Deployment

MATLAB is Fast for Deployment

- Target a GPU for optimal performance
- NVIDIA GPUs use CUDA code
- We only have MATLAB code.

Can we translate this?



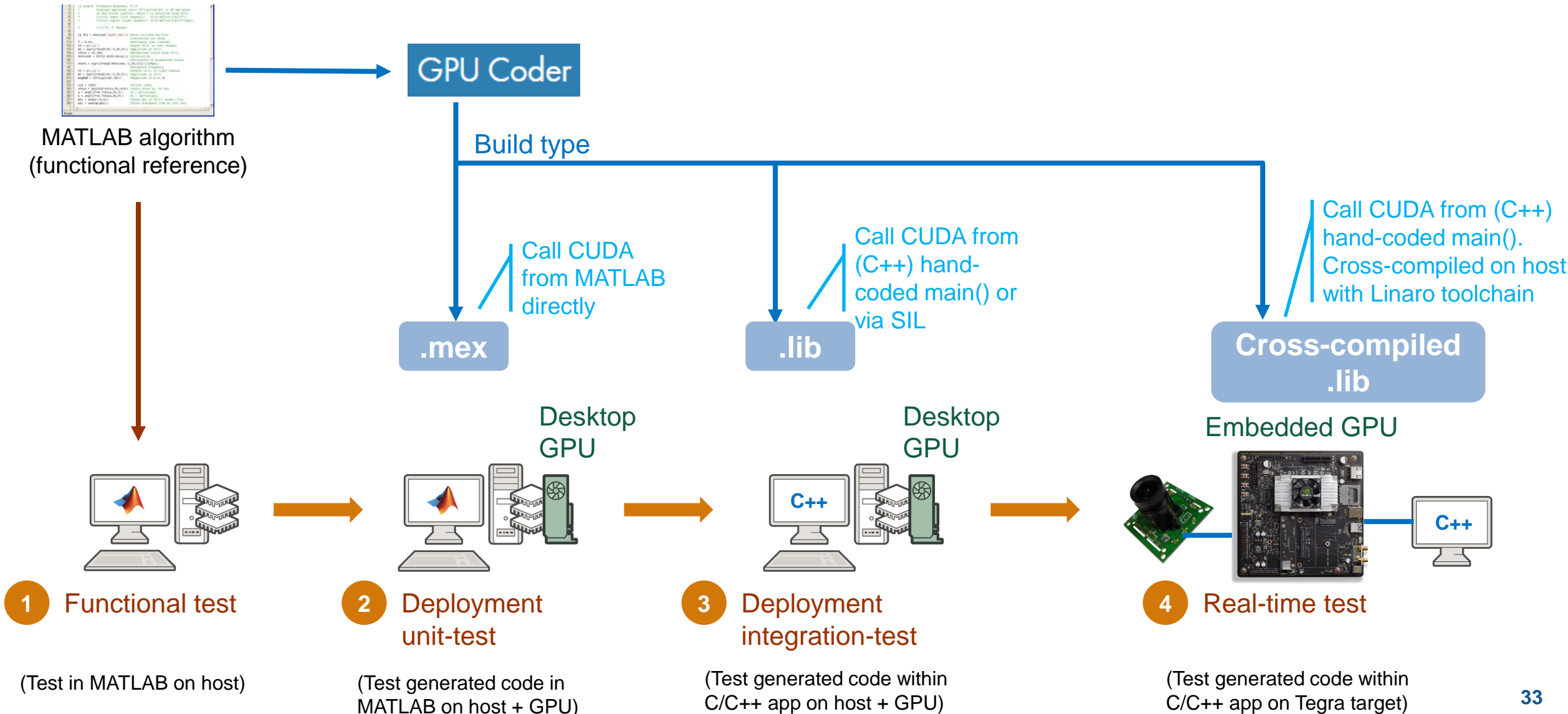
GPU Coder

- Automatically generates **CUDA** Code from MATLAB Code
 - can be used on NVIDIA GPUs

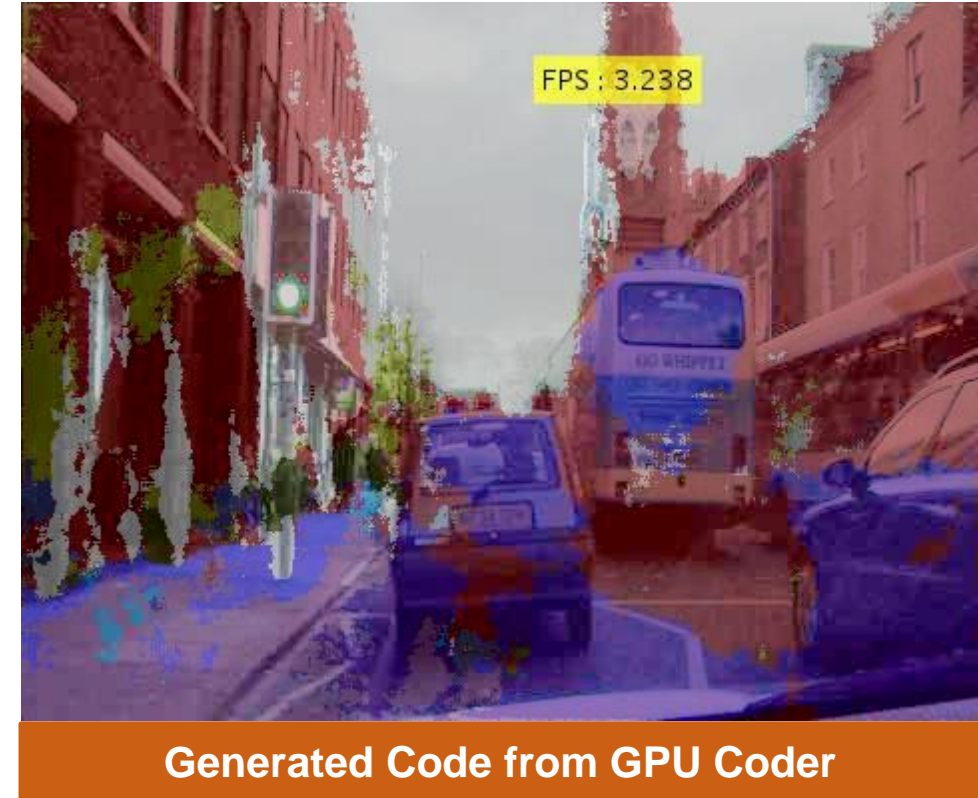


- CUDA extends C/C++ code with constructs for parallel computing

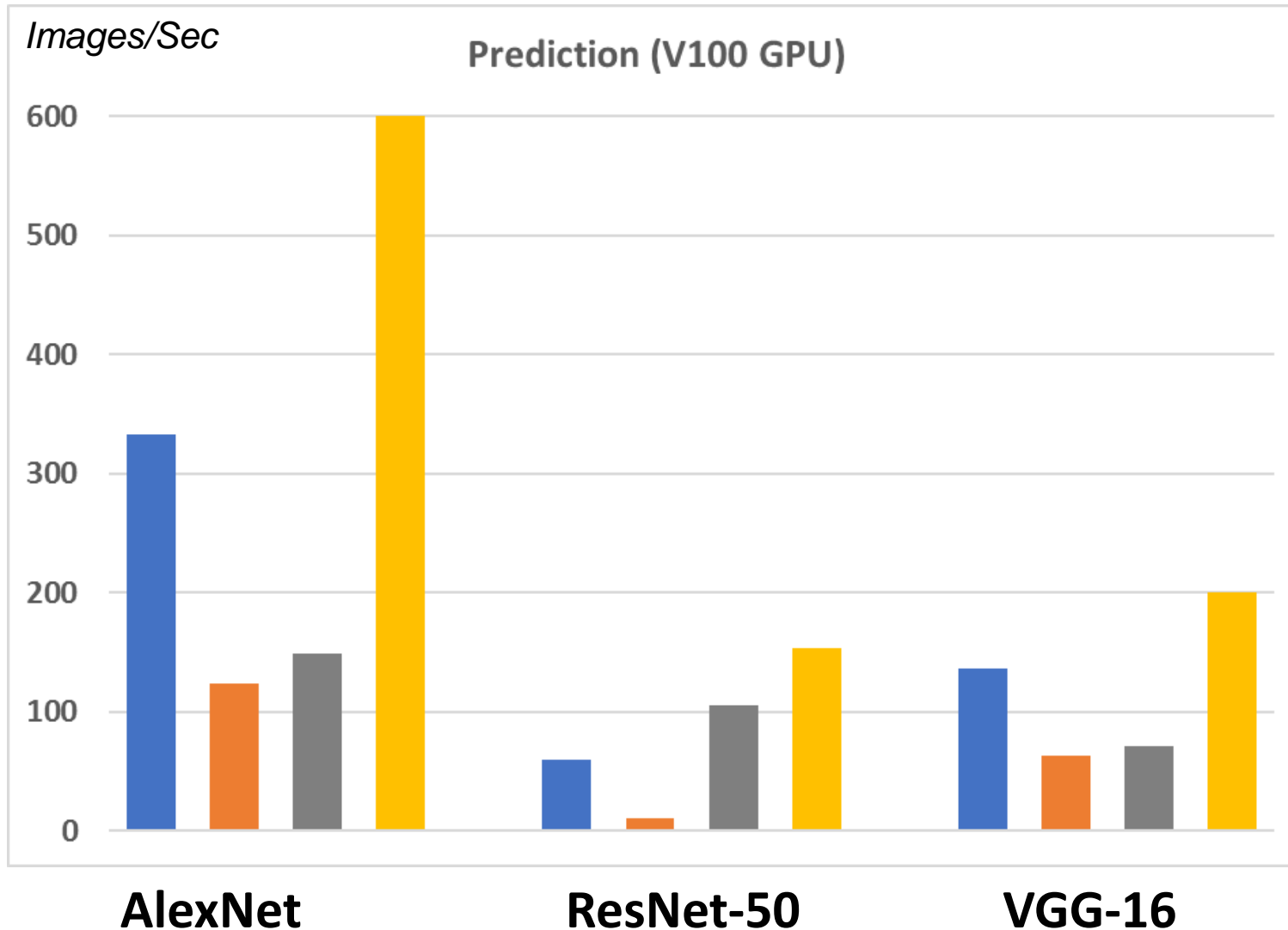
How to Use GPU Coder? Workflow to Embedded Tegra GPU



GPU Coder Performance



Prediction Performance: Fast with GPU Coder



Why is GPU Coder so fast?

- Analyzes and optimizes network architecture
- Invested 15 years in code generation

TensorFlow

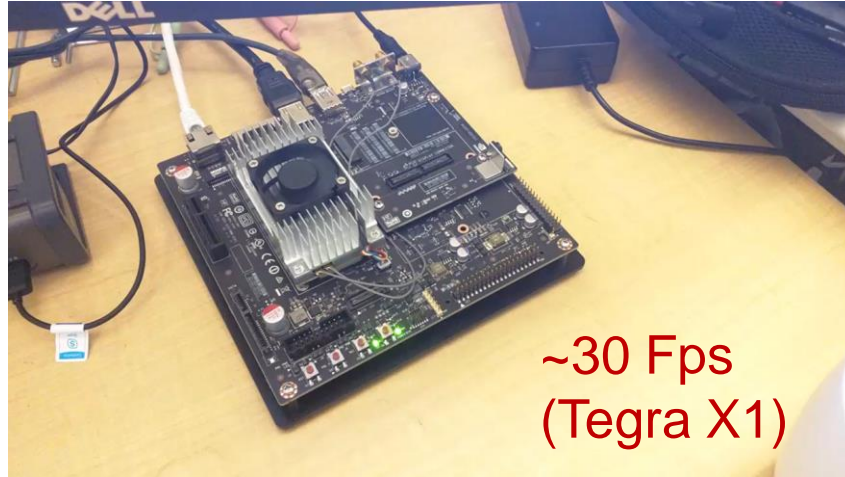
MATLAB

MXNet

GPU Coder

Networks deployed with GPU Coder

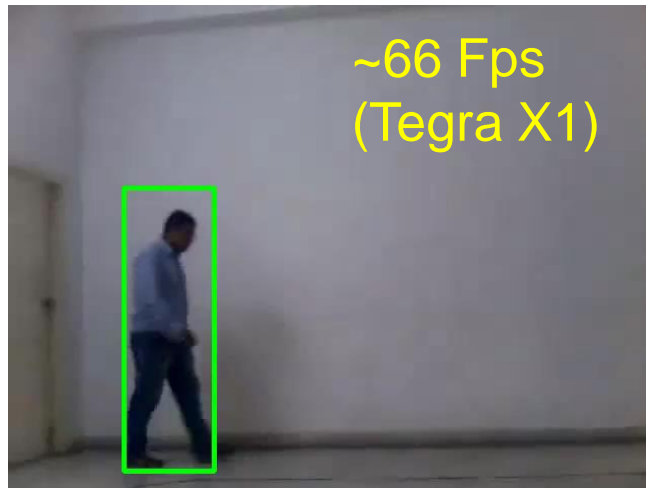
Alexnet



Vehicle
Detection



People detection



Lane detection



Thank you!