## MATLAB EXPO 2018

## Demystifying Deep Learning

"Let the computers do the hard work"

Nilgün Sever, FİGES









## Why MATLAB for Deep Learning?

- MATLAB is Productive
- MATLAB is Fast
- MATLAB Integrates with Open Source

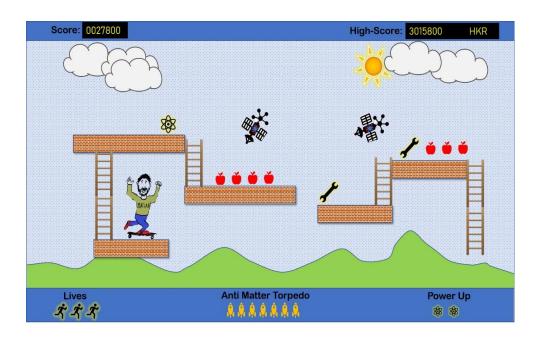


## **Deep Learning Applications**

Voice assistants (speech to text)

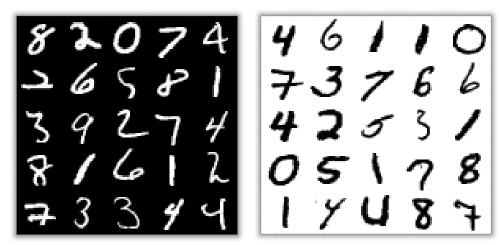
Teaching character to beat video game

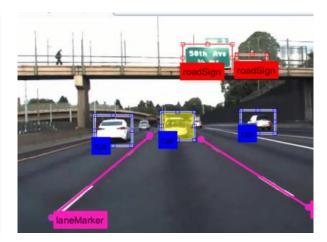
Automatically coloring black-and-white images

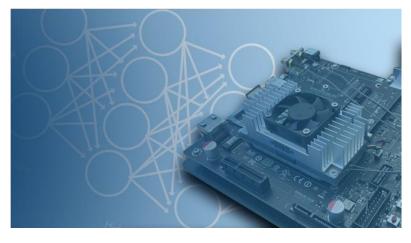












## What is Deep Learning?





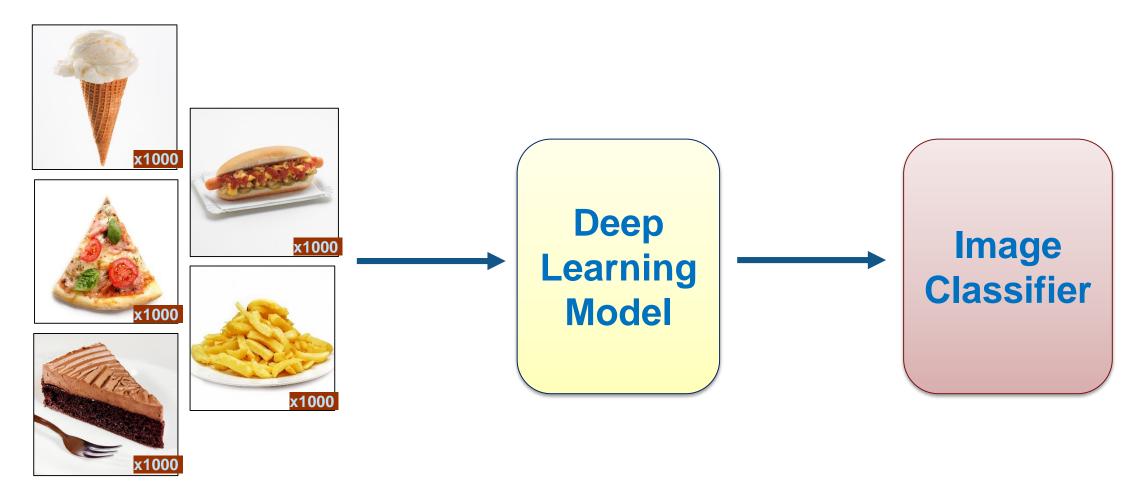


<b>12</b>	<b>0</b>	100%
40.0%	0.0%	0.0%
<b>0</b>	<b>18</b>	100%
0.0%	60.0%	0.0%
100%	100%	100%
0.0%	0.0%	0.0%



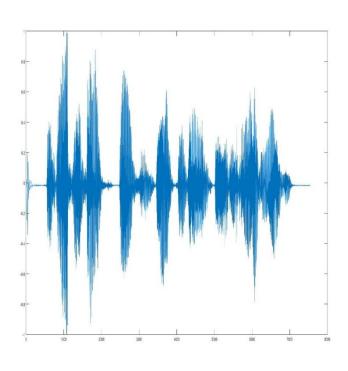
## **Deep Learning**

Model learns to perform classification tasks directly from data.

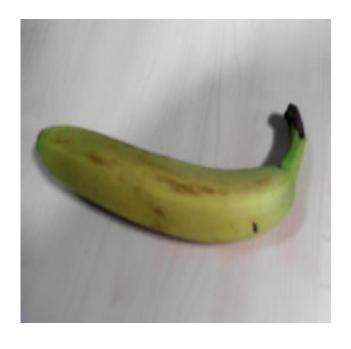




#### **Data Types for Deep Learning**







Signal

**Text** 

Image



#### **Deep Learning is Versatile**



Detection of cars and road in autonomous driving systems



Rain Detection and Removal<sup>1</sup>









*Iris Recognition – 99.4% accuracy*<sup>2</sup>

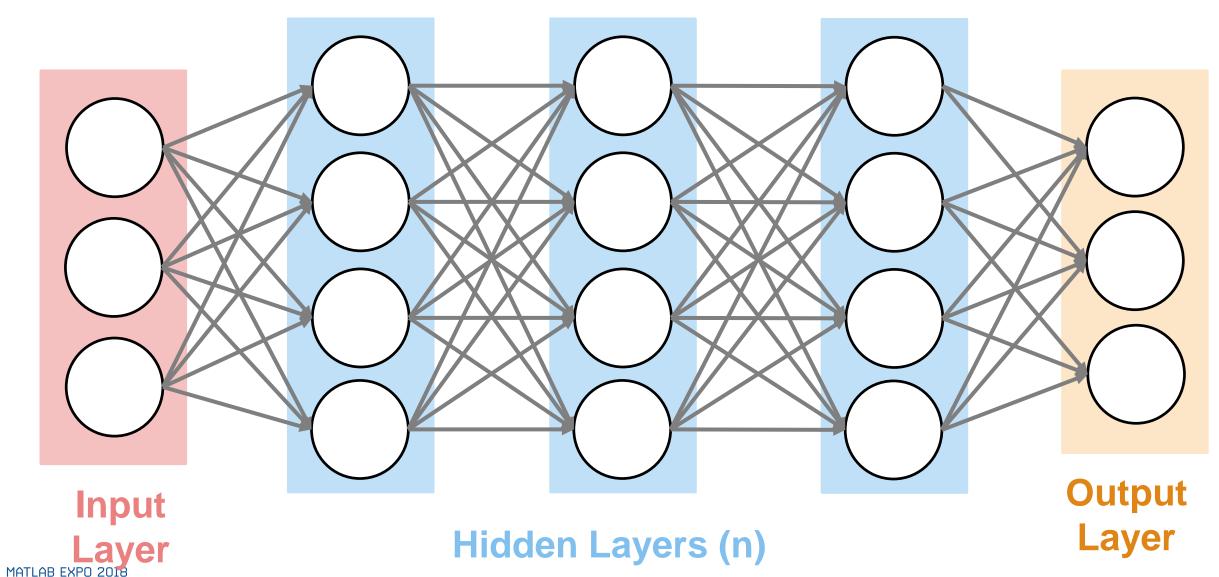
- 1. Deep Joint Rain Detection and Removal from a Single Image" Wenhan Yang, Robby T. Tan, Jiashi Feng, Jiaying Liu, Zongming Guo, and Shuicheng Yan
- 2. Source: An experimental study of deep convolutional features for iris recognition Signal Processing in Medicine and Biology Symposium (SPMB), 2016 IEEE Shervin Minaee; Amirali Abdolrashidiy; Yao Wang; An experimental study of deep convolutional features for iris recognition



## How is deep learning performing so well?



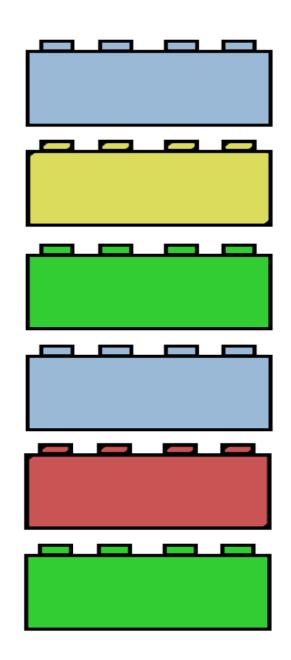
#### Deep Learning Uses a Neural Network Architecture





## **Thinking about Layers**

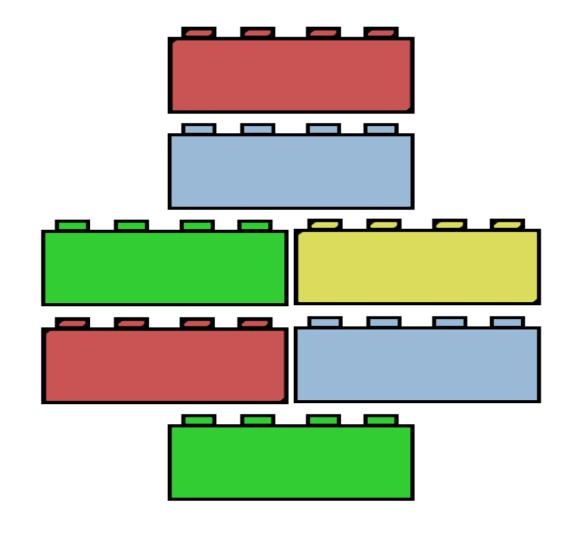
- Layers are like blocks
  - Stack on top of each other
  - Replace one block with a different one
- Each hidden layer processes the information from the previous layer





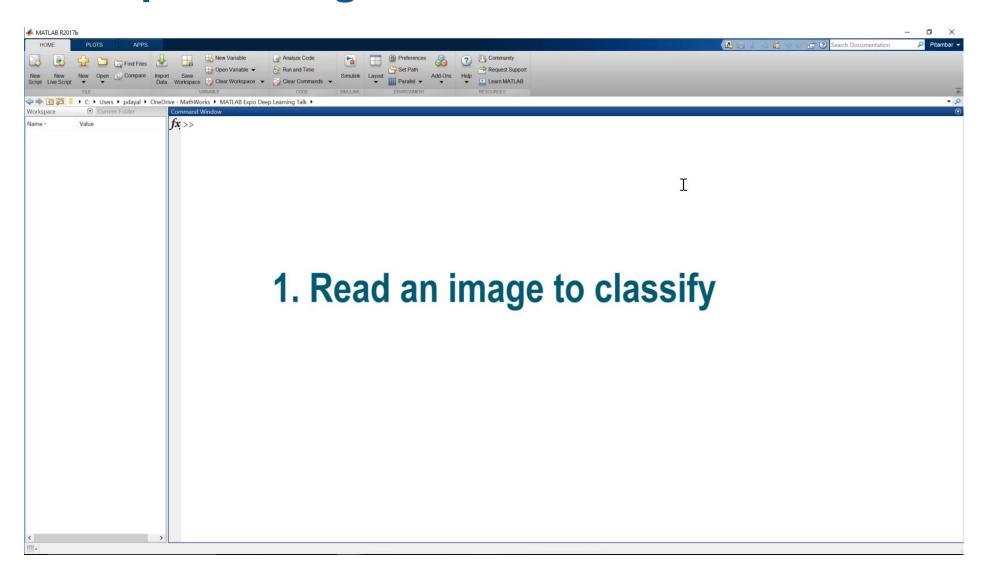
## **Thinking about Layers**

- Layers are like blocks
  - Stack them on top of each other
  - Replace one block with a different one
- Each hidden layer processes the information from the previous layer
- Layers can be ordered in different ways





#### **Deep Learning in 6 Lines of MATLAB Code**





## Why MATLAB for Deep Learning?

- MATLAB is Productive
- MATLAB is Fast
- MATLAB integrates with Open Source



# "I love to label and preprocess my data"

~ Said no engineer, ever.



#### **Caterpillar Case Study**



- World's leading manufacturer of construction and mining equipment.
- Similarity between these projects?
  - Autonomous haul trucks
  - Pedestrian detection
  - Equipment classification
  - Terrain mapping



## Computer Must Learn from Lots of Data

ALL data must first be labeled to create these autonomous systems.





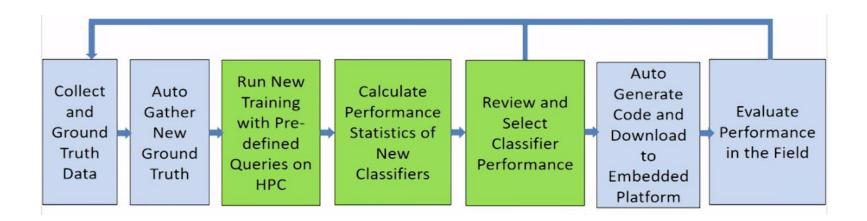
"We were spending way too much time ground-truthing [the data]"

--Larry Mianzo, Caterpillar



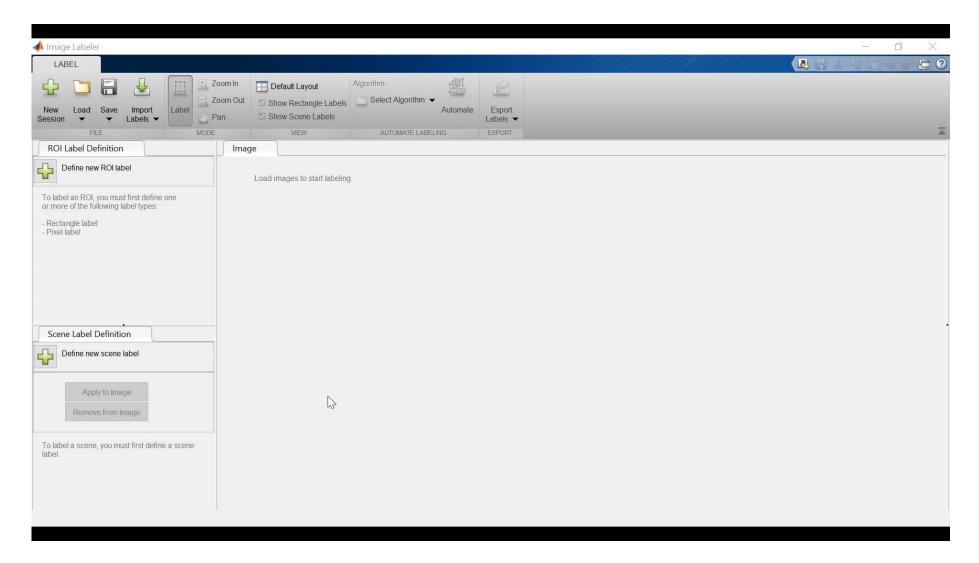
## **How Did Caterpillar Do with Our Tools?**

- Semi-automated labeling process
  - "We go from having to label 100 percent of our data to only having to label about 80 to 90 percent"
- Used MATLAB for entire development workflow.
  - "Because everything is in MATLAB, development time is short"

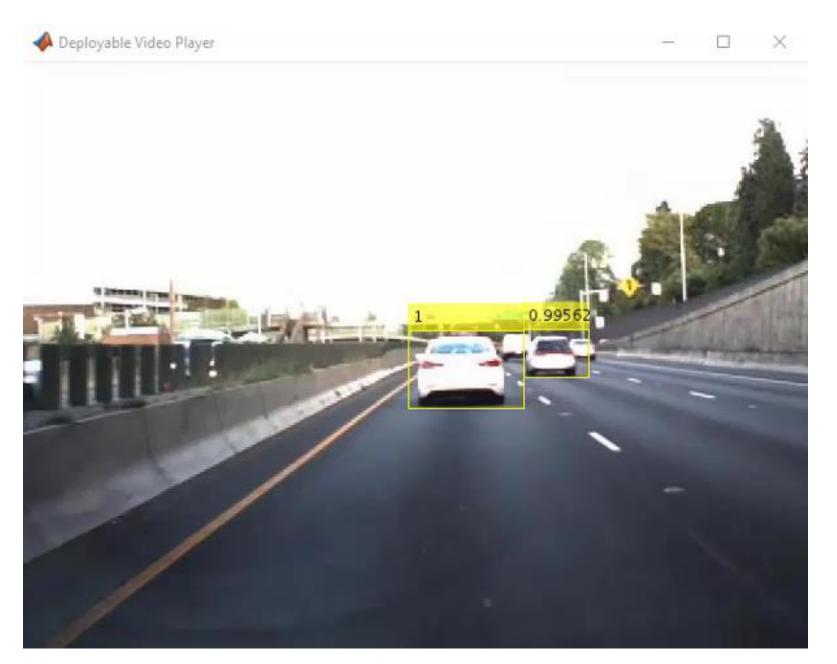




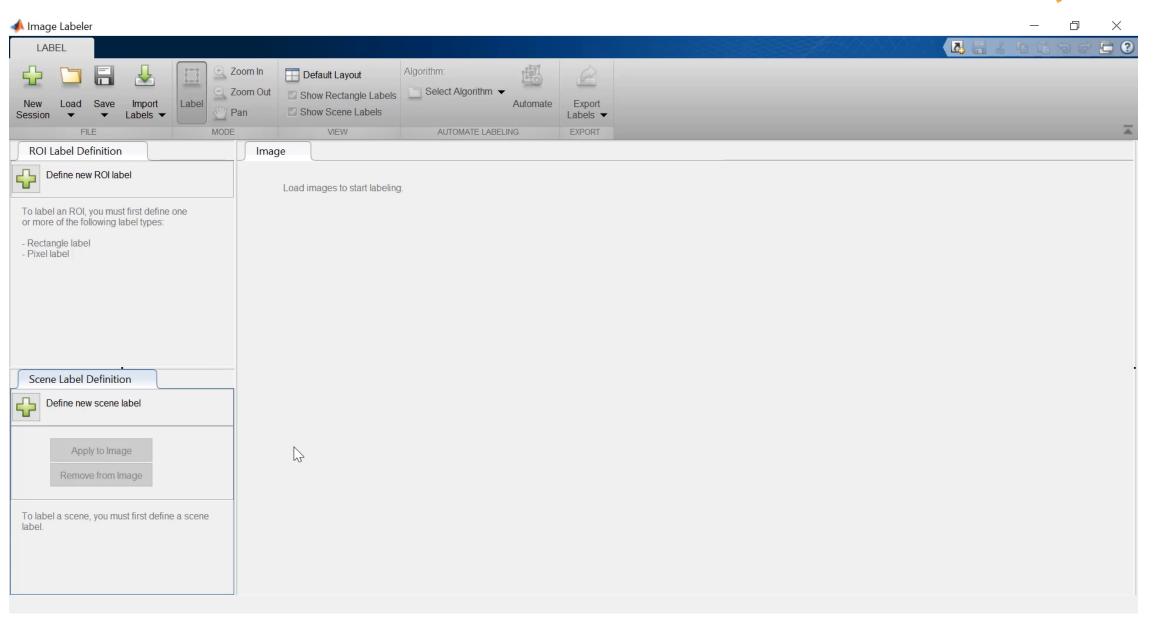
## **How Does MATLAB Come into Play?**

















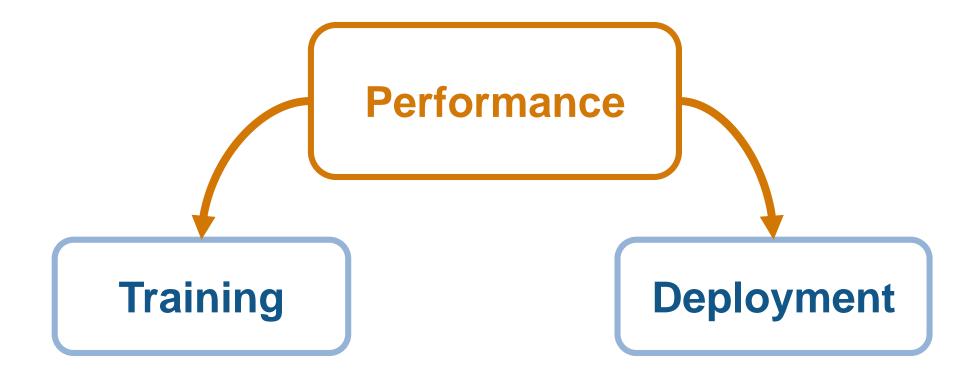
#### **MATLAB** is Productive

- Image Labeler App semi-automates labeling workflow
- Bootstrapping
  - Improve automatic labeling by updating algorithm as you label more images correctly.

Easy to load metadata even when labeling manually



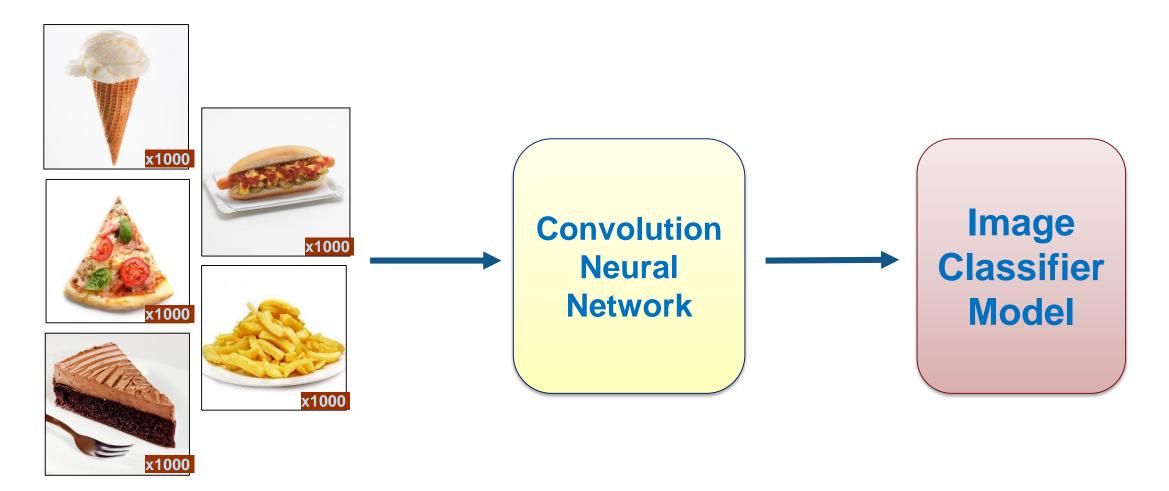
## **MATLAB** is Fast





## What is Training?

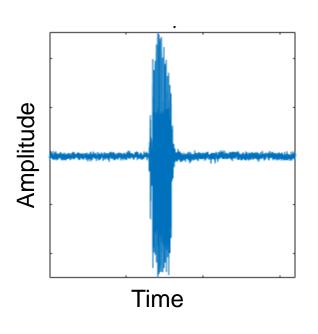
Feed labeled data into neural network to create working model

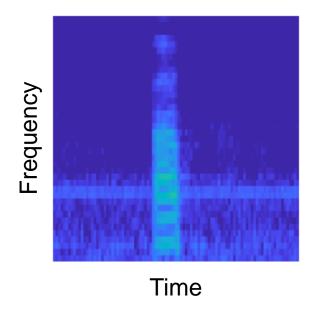


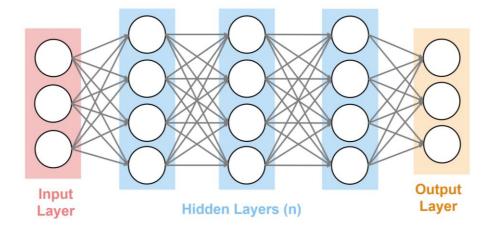


## **Speech Recognition Example**

Audio signal → Spectrogram → Image Classification algorithm



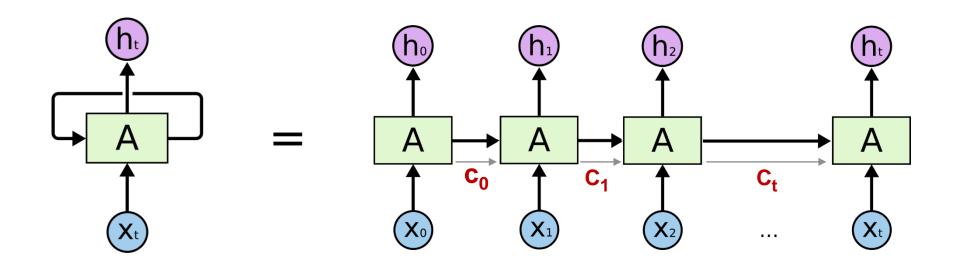






## **Another Network for Signals - LSTM**

- LSTM = Long Short Term Memory (Networks)
  - Signal, text, time-series data
  - Use previous data to predict new information
- I live in France. I speak \_\_\_\_\_\_.

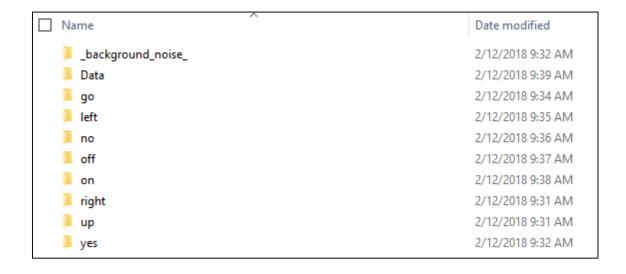




#### 1. Create Datastore

 Datastore creates reference for data

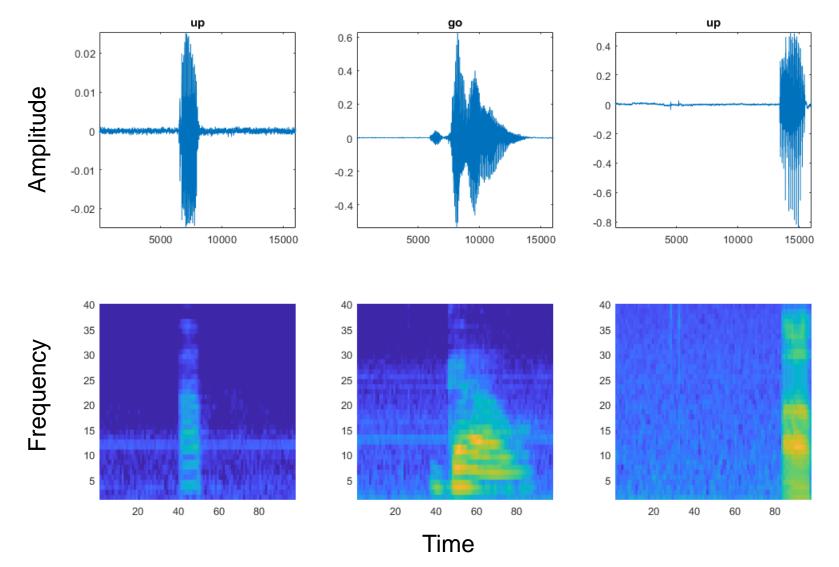
 Do not have to load in all objects into memory



```
datafolder = fullfile(tempdir,'speech_commands_v0.01');
addpath(fullfile(matlabroot,'toolbox','audio','audiodemos'))
ads = audioexample.Datastore(datafolder, ...
   'IncludeSubfolders',true, ...
   'FileExtensions','.wav', ...
   'LabelSource','foldernames', ...
   'ReadMethod','File')
```



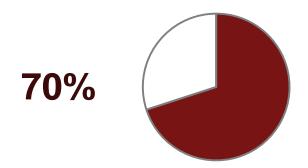
## 2. Compute Speech Spectrograms





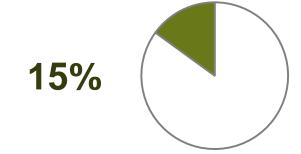
## 3. Split datastores

#### **Training**



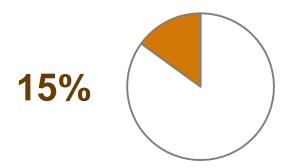
- Trains the model
- Computer "learns" from this data

#### **Validation**



 Checks accuracy of model during training

#### **Test**



- Tests model accuracy
- Not used until validation accuracy is good



#### 4. Define Architecture and Parameters

```
layers = [
    imageInputLayer(imageSize)

convolution2dLayer(3,16,'Padding','same')
    batchNormalizationLayer
    reluLayer

maxPooling2dLayer(2,'Stride',2)

convolution2dLayer(3,32,'Padding','same')
    batchNormalizationLayer
    reluLayer

maxPooling2dLayer(2,'Stride',2,'Padding',[0,1])

dropoutLayer(dropoutProb)
    convolution2dLayer(3,64,'Padding','same')
    batchNormalizationLayer
    reluLayer

dropoutLayer(dropoutProb)
```

```
convolution2dLayer(3,64, 'Padding', 'same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2, 'Stride',2, 'Padding',[0,1])
dropoutLayer(dropoutProb)
convolution2dLayer(3,64,'Padding','same')
batchNormalizationLayer
reluLayer
dropoutLayer(dropoutProb)
convolution2dLayer(3,64, 'Padding', 'same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer([1 13])
fullyConnectedLayer(numClasses)
softmaxLayer
weightedCrossEntropyLayer(classNames,classWeights)];
```

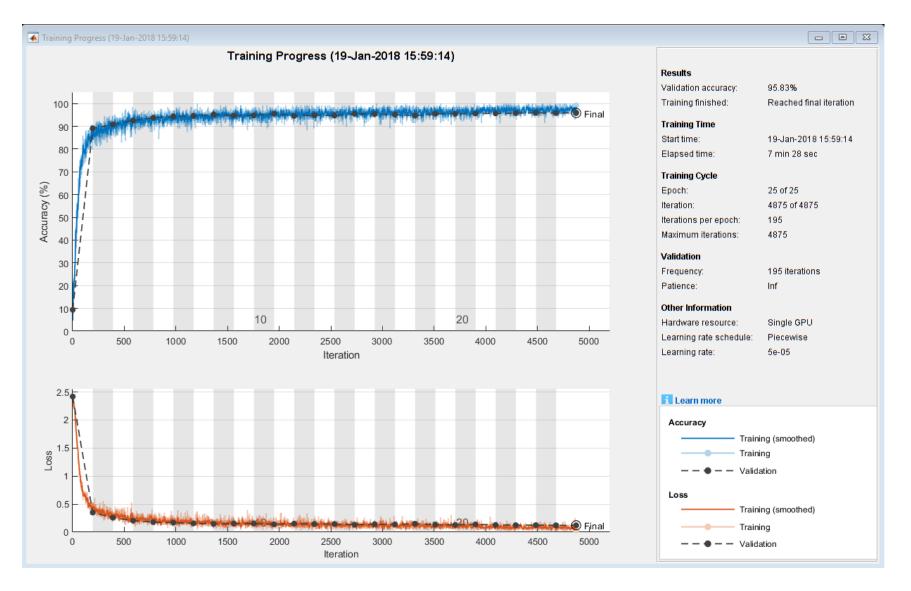
```
miniBatchSize = 128;
validationFrequency = floor(numel(YTrain)/miniBatchSize);
options = trainingOptions('adam', ...
    'InitialLearnRate',5e-4, ...
    'MaxEpochs',25, ...
    'MiniBatchSize',miniBatchSize, ...
    'Shuffle','every-epoch', ...
    'Plots','training-progress', ...
    'Verbose',false, ...
    'ValidationData',{XValidation,YValidation}, ...
    'ValidationFrequency',validationFrequency, ...
    'ValidationPatience',Inf, ...
    'LearnRateSchedule','piecewise', ...
    'LearnRateDropFactor',0.1, ...
    'LearnRateDropFeriod',20);
```

**Model Parameters** 

**Neural Network Architecture** 



#### 5. Train Network





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



Single CPU



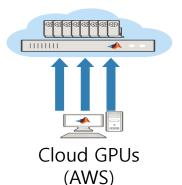
Single CPU Single GPU



Single CPU, Multiple GPUs



On-prem server with GPUs



#### HOW TO TARGET?

```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...

'ExecutionEnvironment', 'auto' );
```

```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...

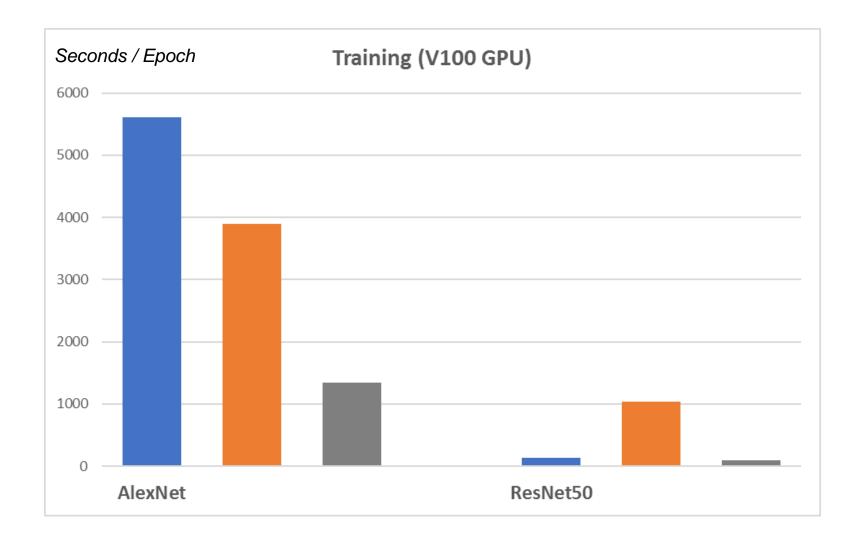
'ExecutionEnvironment', 'multi-gpu' );
```

```
opts = trainingOptions('sgdm', ...
    'MaxEpochs', 100, ...
    'MiniBatchSize', 250, ...
    'InitialLearnRate', 0.00005, ...

'ExecutionEnvironment', 'parallel' );
```



#### **Training Performance**



TensorFlow
MATLAB
MXNet
Batch size 32



#### **Training is an Iterative Process**

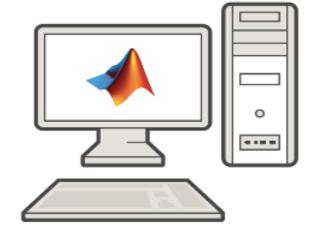
```
miniBatchSize = 128;
validationFrequency = floor(numel(YTrain)/miniBatchSize);
options = trainingOptions('adam', ...
    'InitialLearnRate',5e-4, ...
    'MaxEpochs',25, ...
    'MiniBatchSize', miniBatchSize, ...
    'Shuffle', 'every-epoch', ...
    'Plots', 'training-progress', ...
    'Verbose', false, ...
    'ValidationData', {XValidation, YValidation}, ...
    'ValidationFrequency', validationFrequency, ...
    'ValidationPatience', Inf, ...
    'LearnRateSchedule', 'piecewise', ...
    'LearnRateDropFactor',0.1, ...
    'LearnRateDropPeriod',20);
```

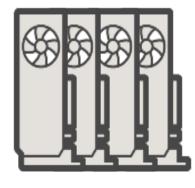
Parameters adjusted according to performance



## **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

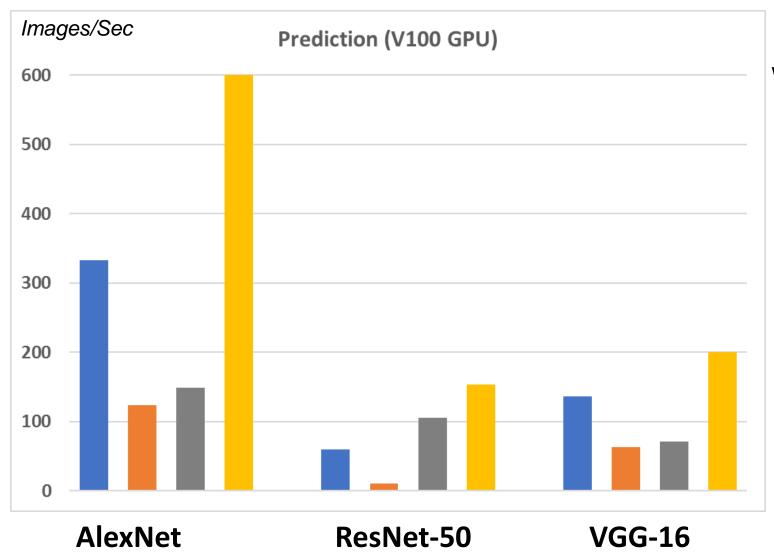


#### **GPU Coder Performance**





#### **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** 



## Why MATLAB?

- MATLAB is Productive
- MATLAB is Fast
- MATLAB Integrates with Open Source



#### **Used MATLAB and Open Source Together**



1. Deep Joint Rain Detection and Removal from a Single Image" Wenhan Yang, Robby T. Tan, Jiashi Feng, Jiaying Liu, Zongming Guo, and Shuicheng Yan

- Used Caffe and MATLAB together
- Achieved significantly better results than an engineered rain model.
- Use our tools where it makes your workflow easier!



## **MATLAB Integrates with Open Source Frameworks**

- Access to many pretrained models through add-ons
- Users wanted to import latest models

- Import models directly from Tensorflow or Caffe
  - Allows for improved collaboration

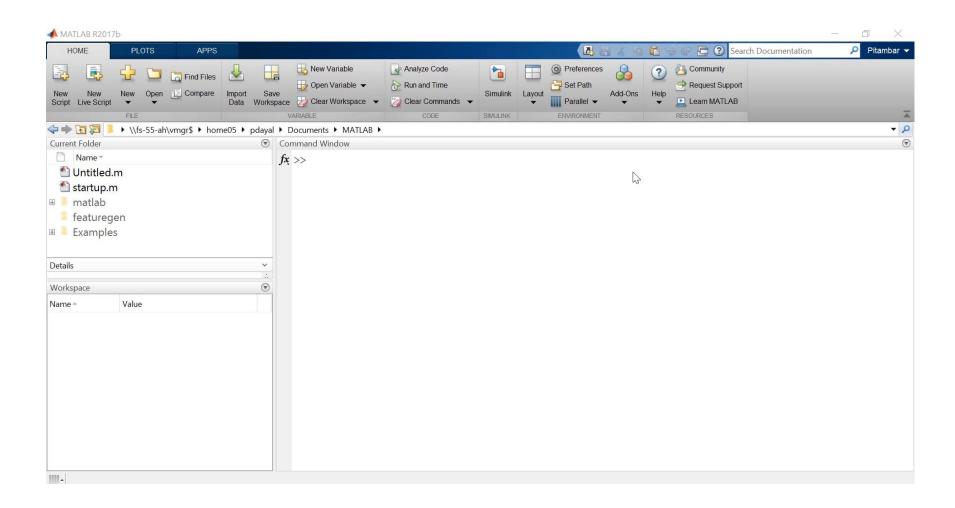
#### KERAS IMPORTER

Importer for TensorFlow-Keras Models





#### **Keras-Tensorflow Importer**





# MATLAB Integrates with Open Source Frameworks

- MATLAB supports entire deep learning workflow
  - Use when it is convenient for your workflow
- Access to latest models
- Improved collaboration with other users



## Why MATLAB for Deep Learning?

- MATLAB is Productive
- MATLAB is Fast (Performance)
- MATLAB Integrates with Open Source (Frameworks)