

IESTI01 – TinyML

Embedded Machine Learning

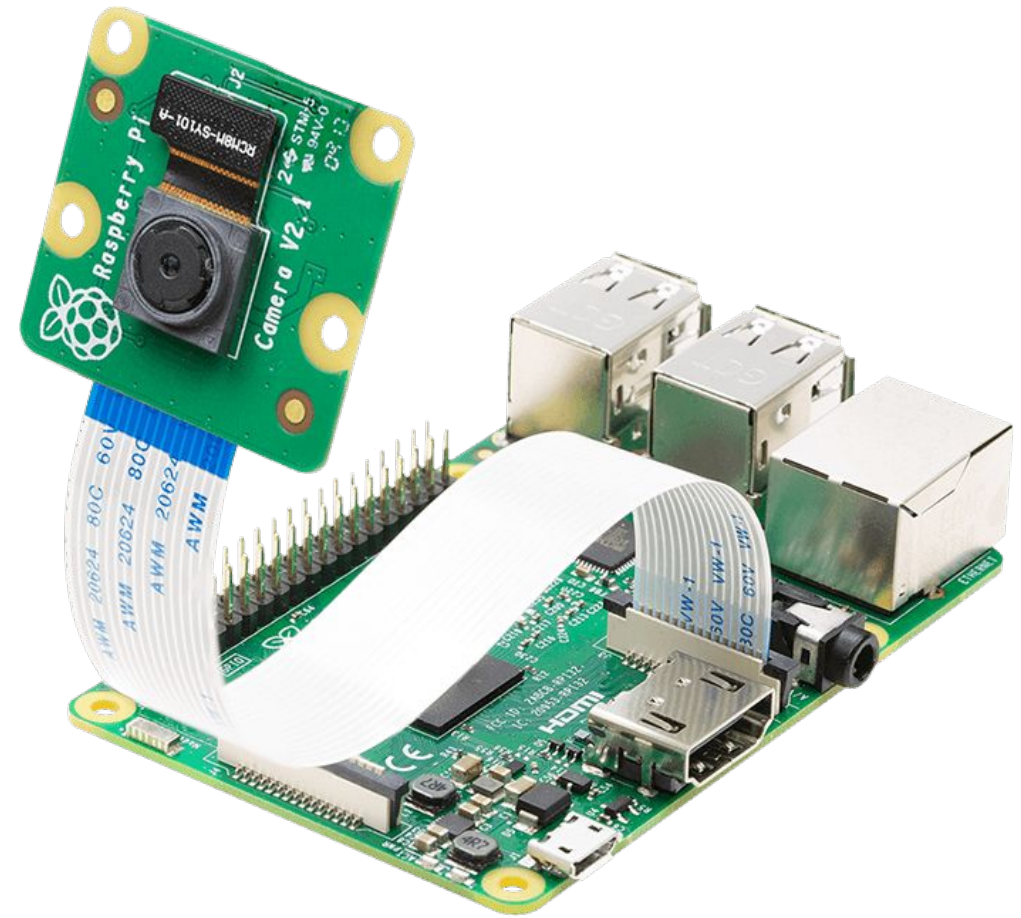
16. a EdgeML with TensorFlow Lite

Image Classification & Object Detection
Demo



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UNIFEI



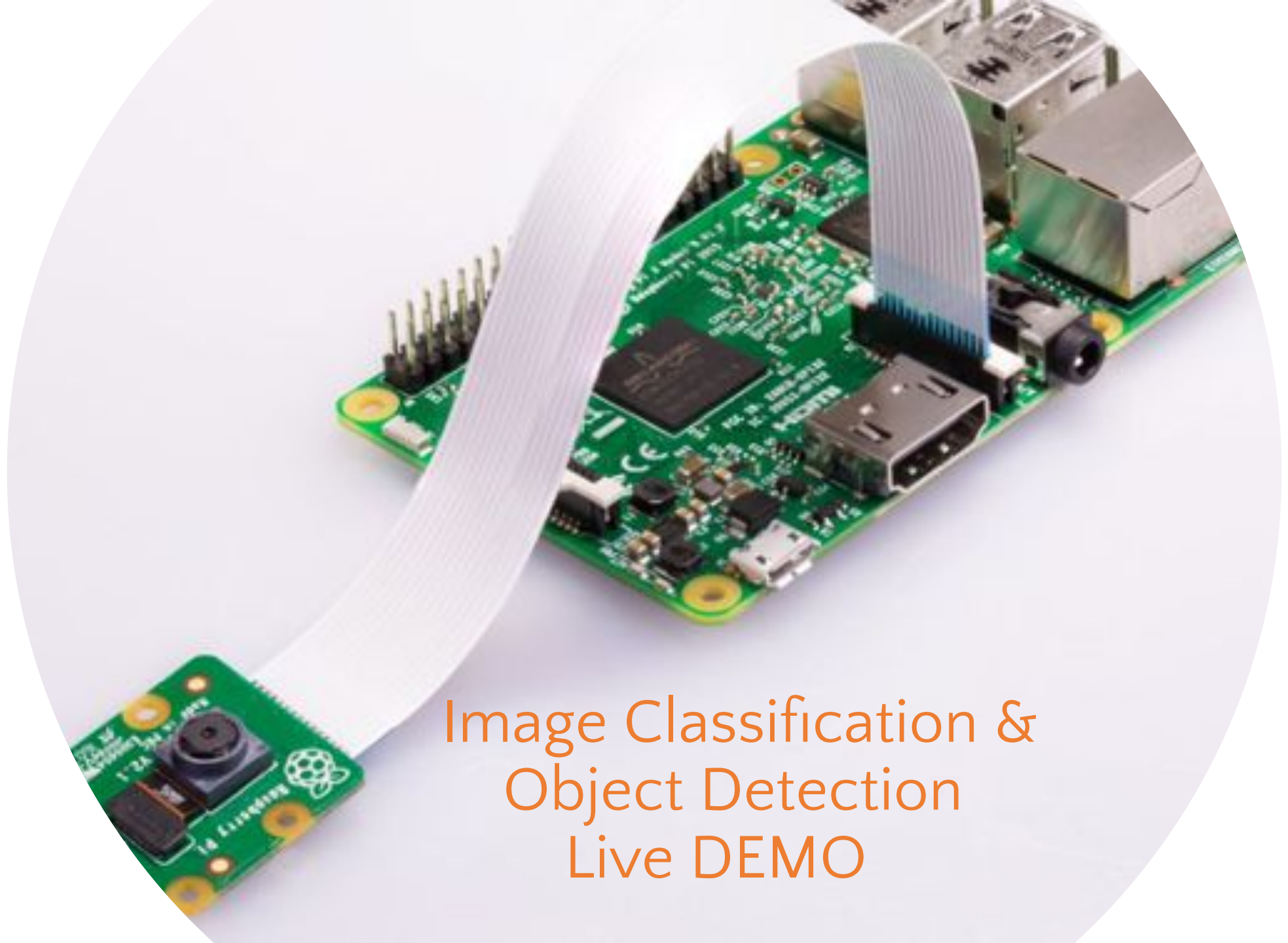
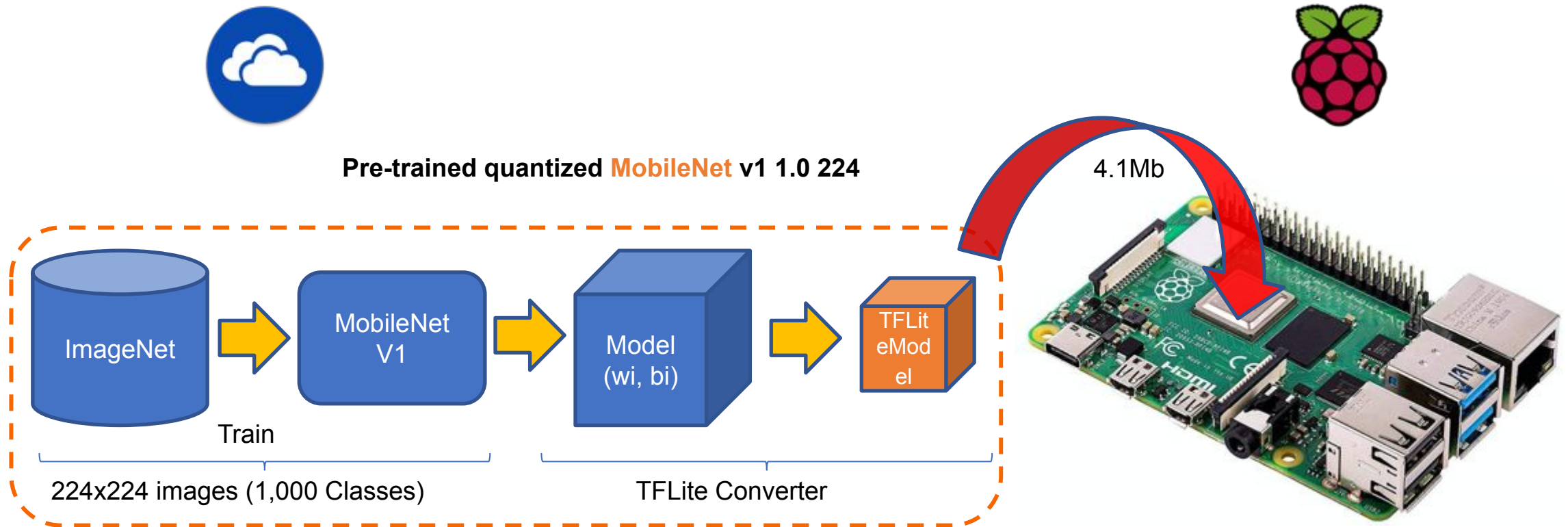


Image Classification & Object Detection Live DEMO

Demo 1 – Image Classification



TF Lite Inference – Image Classification

```
In [5]: 1 input_details
```

```
Out[5]: [{'name': 'input',  
  'index': 88,  
  'shape': array([ 1, 224, 224,  3]),  
  'dtype': numpy.uint8,  
  'quantization': (0.0078125, 128),  
  'quantization_parameters': {'scales': array([0.0078125], dtype=float32),  
  'zero_points': array([128]),  
  'quantized_dimension': 0}]]
```

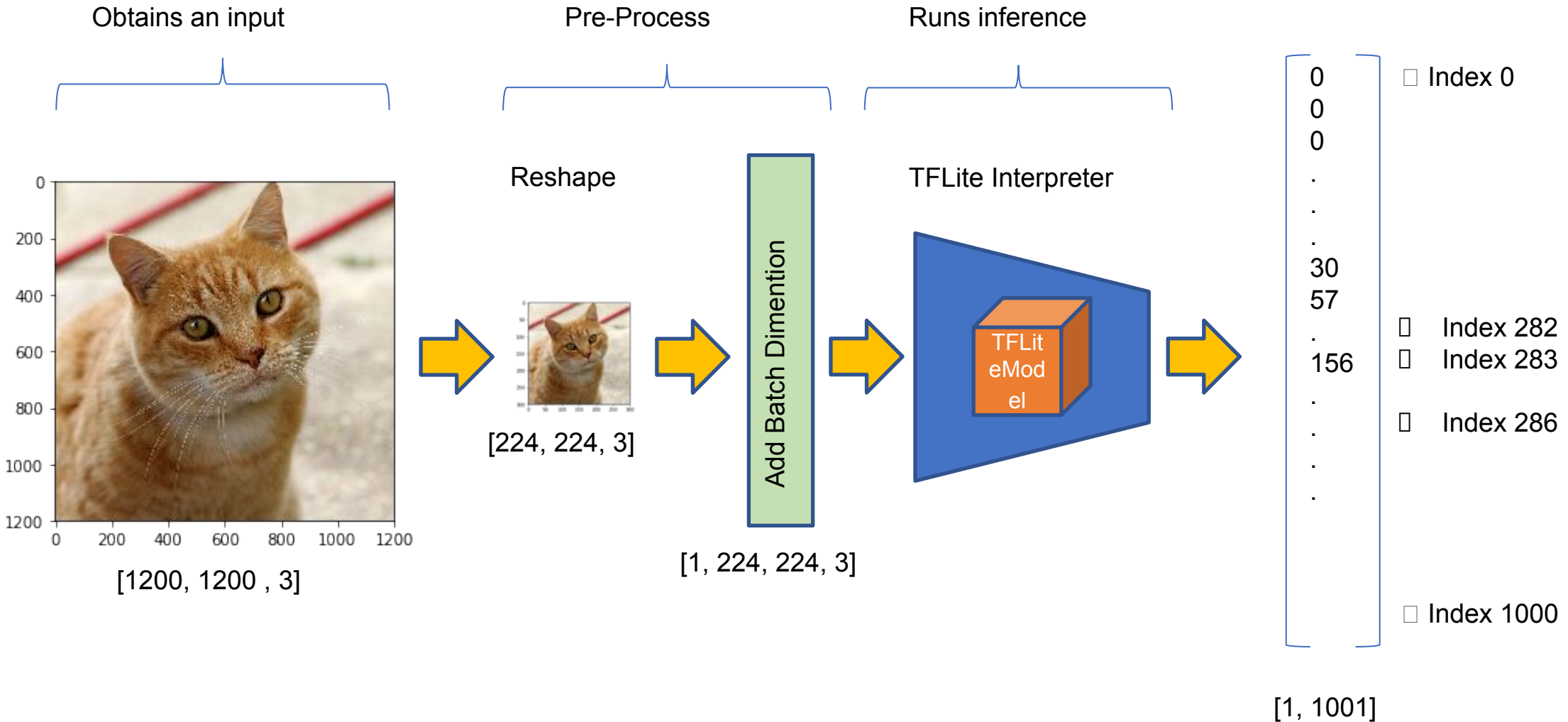
← Input Image
Shape

```
In [6]: 1 output_details
```

```
Out[6]: [{'name': 'MobilenetV1/Predictions/Reshape_1',  
  'index': 87,  
  'shape': array([ 1, 1001]),  
  'dtype': numpy.uint8,  
  'quantization': (0.00390625, 0),  
  'quantization_parameters': {'scales': array([0.00390625], dtype=float32),  
  'zero_points': array([0]),  
  'quantized_dimension': 0}]]
```

← Output
model

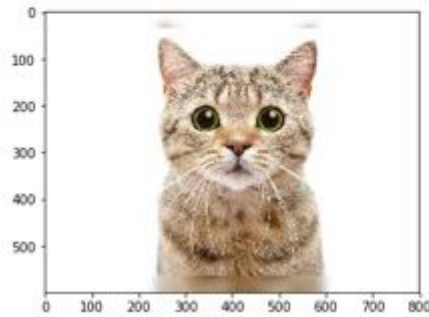
TF Lite Inference – Image Classification



TF Lite Inference – Demo – Postprocess

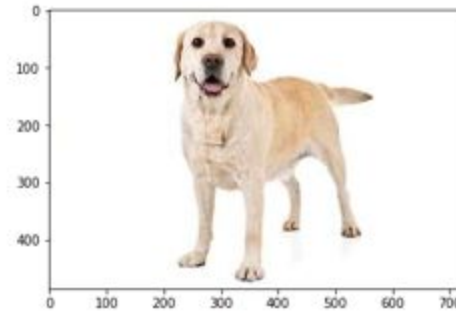
```
1 image_path = './images/cat_1.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
Egyptian cat	: 64%
tabby	: 14%
bucket	: 3%



```
1 image_path = './images/dog_1.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
Labrador retriever	: 83%
golden retriever	: 13%
bloodhound	: 0%



```
1 image_path = './images/pi_1.jpeg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
German shepherd	: 60%
dhole	: 16%
malinois	: 7%



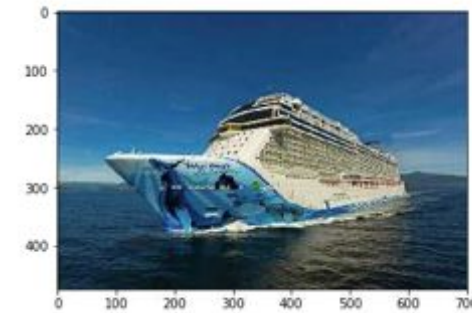
```
1 image_path = './images/car_2.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
minivan	: 76%
convertible	: 6%
jeep	: 4%



```
1 image_path = './images/ship_1.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
liner	: 85%
trimaran	: 11%
aircraft carrier	: 0%



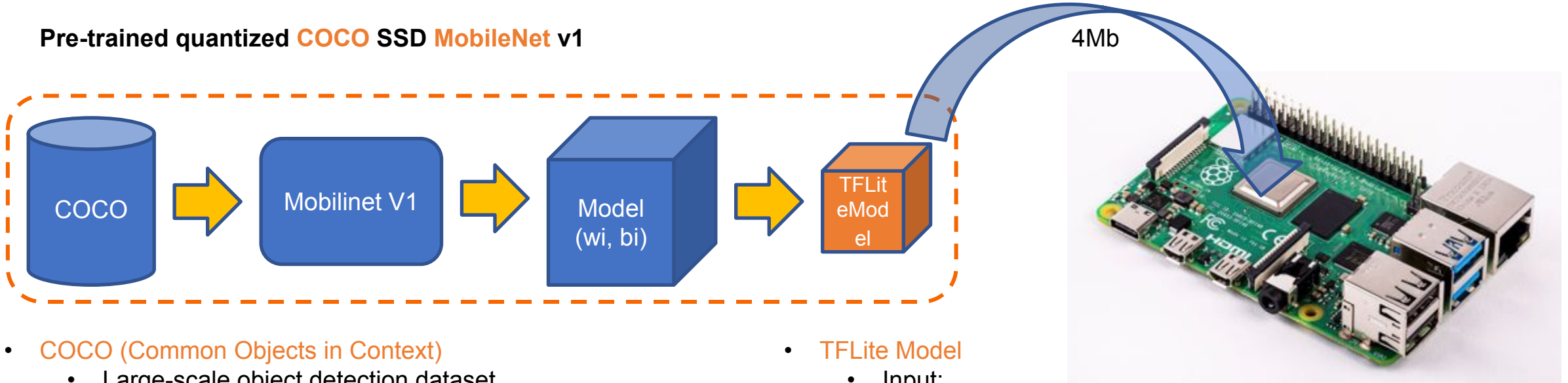
```
1 image_path = './images/table.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]	[Prob]
dining table	: 40%
table lamp	: 34%
desk	: 17%



Demo 2 – Object Detection

Pre-trained quantized **COCO SSD MobileNet v1**



- **COCO (Common Objects in Context)**

- Large-scale object detection dataset
- 200K labeled images
- 91 Stuff categories

- **MobileNet**

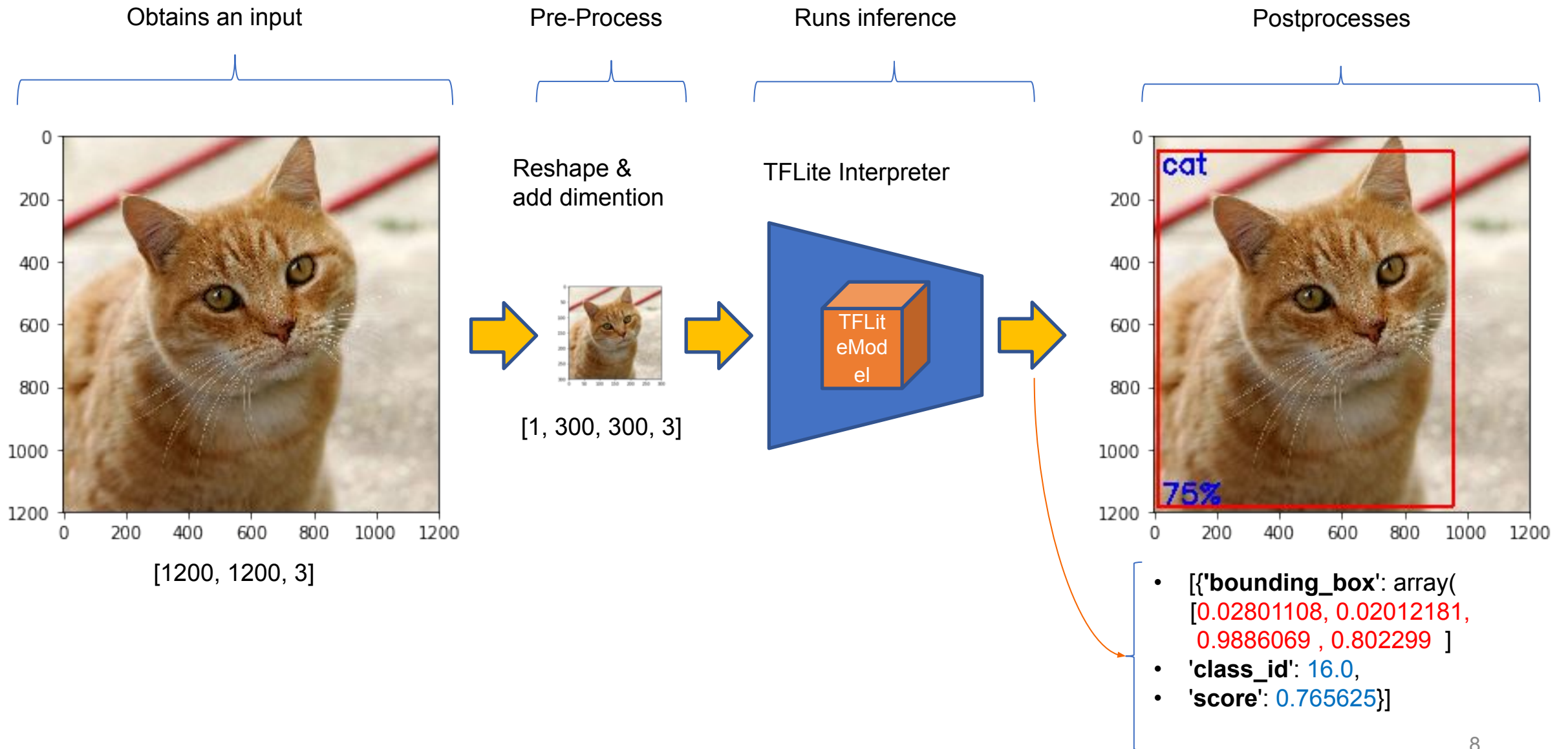
- “Depth- wise Separable” convolutions
- Introduced by Google in 2017
- Similar performance with state-of-the-art architectures (as VGG or Inception)
- Much smaller network (20% of VGG parameters (7M))

- **TFLite Model**

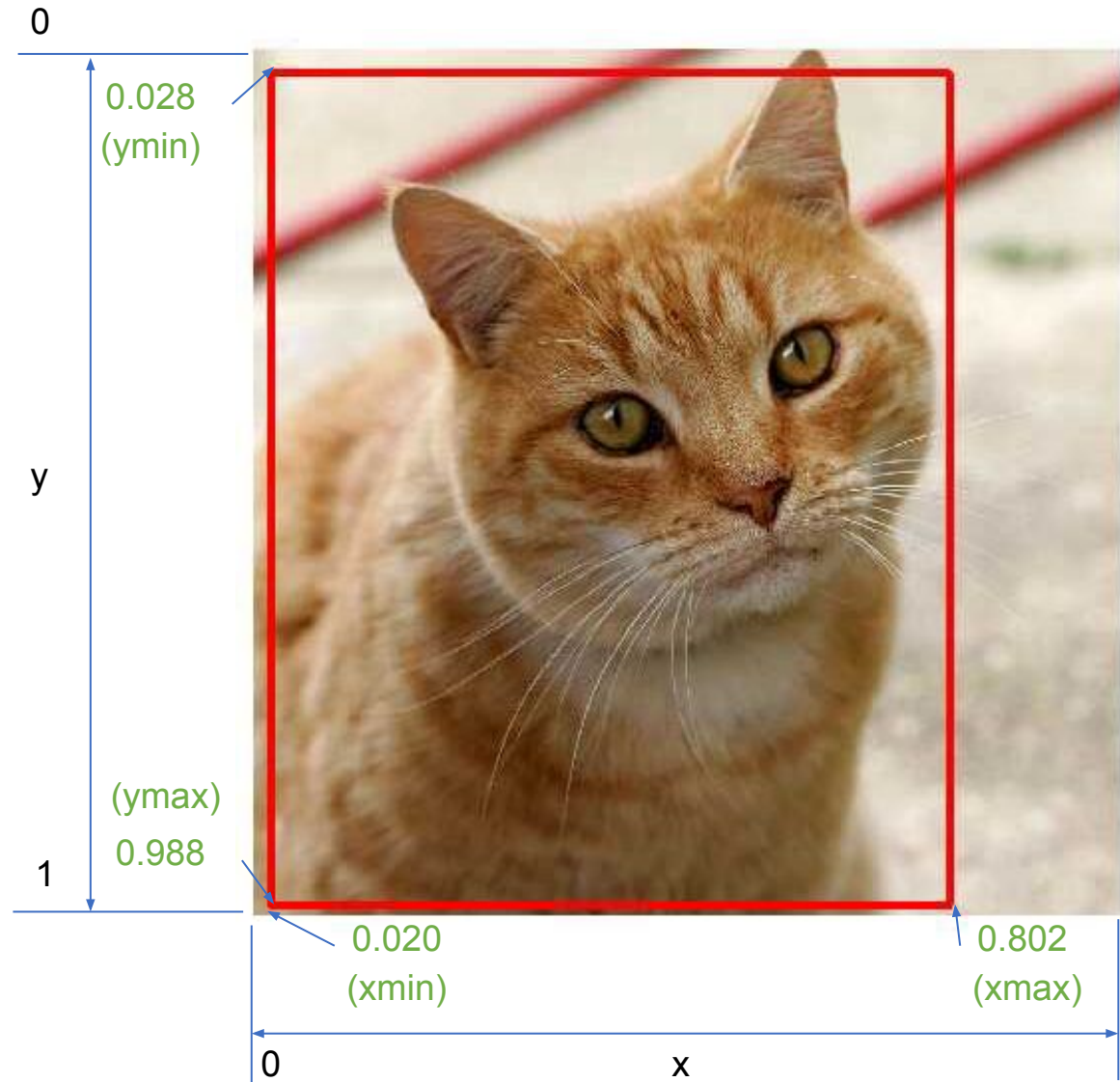
- Input:
 - Image: 300x300x3
 - Flattened Buffer: 270K bytes
 - Each byte: 0 to 255
- Output:
 - Bounding Box \square (ymin, xmin, ymax, xmax)
 - Class ID (0 to 89) ("Stuff categories")
 - Score (0 to 1)

COCO Paper:
<https://arxiv.org/pdf/1405.0312.pdf>
MobileNet paper:
<https://arxiv.org/pdf/1704.04861.pdf>

TF Lite Inference – Object Detection



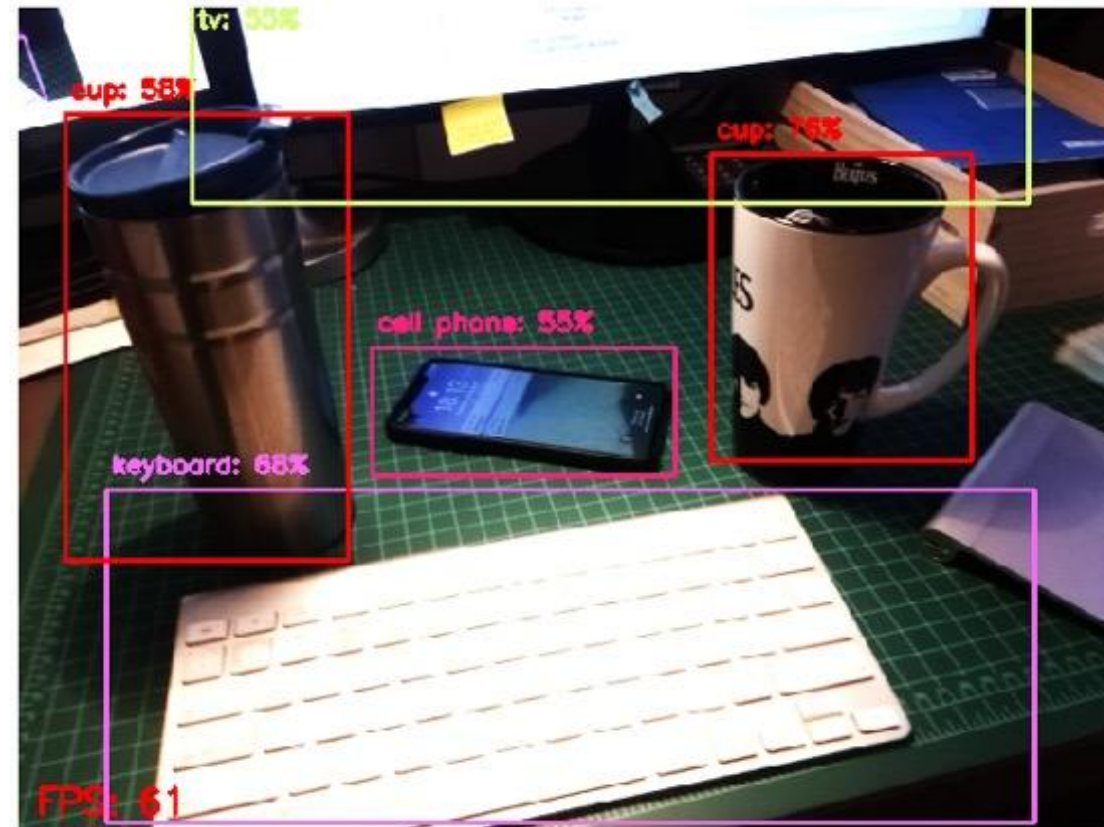
Bounding Box- Object Detection



TF Lite Inference – Demo – Postprocesses



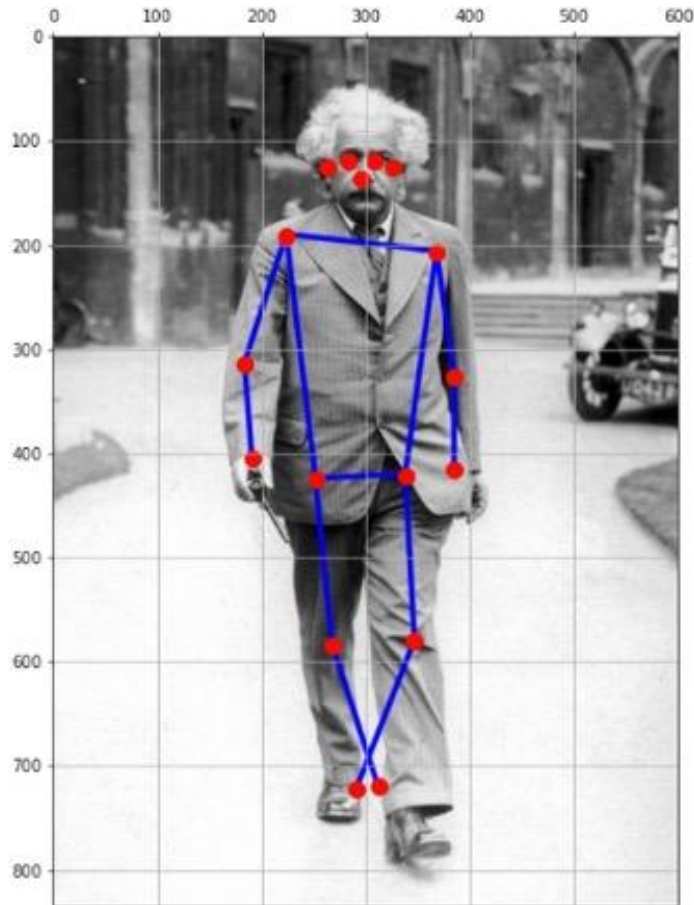
Photos



Live Video

TensorFlow Lite – Other applications

- Pose Estimation on RPi



- Gesture Recognition
- Speech Recognition
- Smart Replay
- Image Segmentation
- Text Classification
- On-device recommendation
- Style Transfer



Content Image



Style Image



Output Image

Example of style transfer

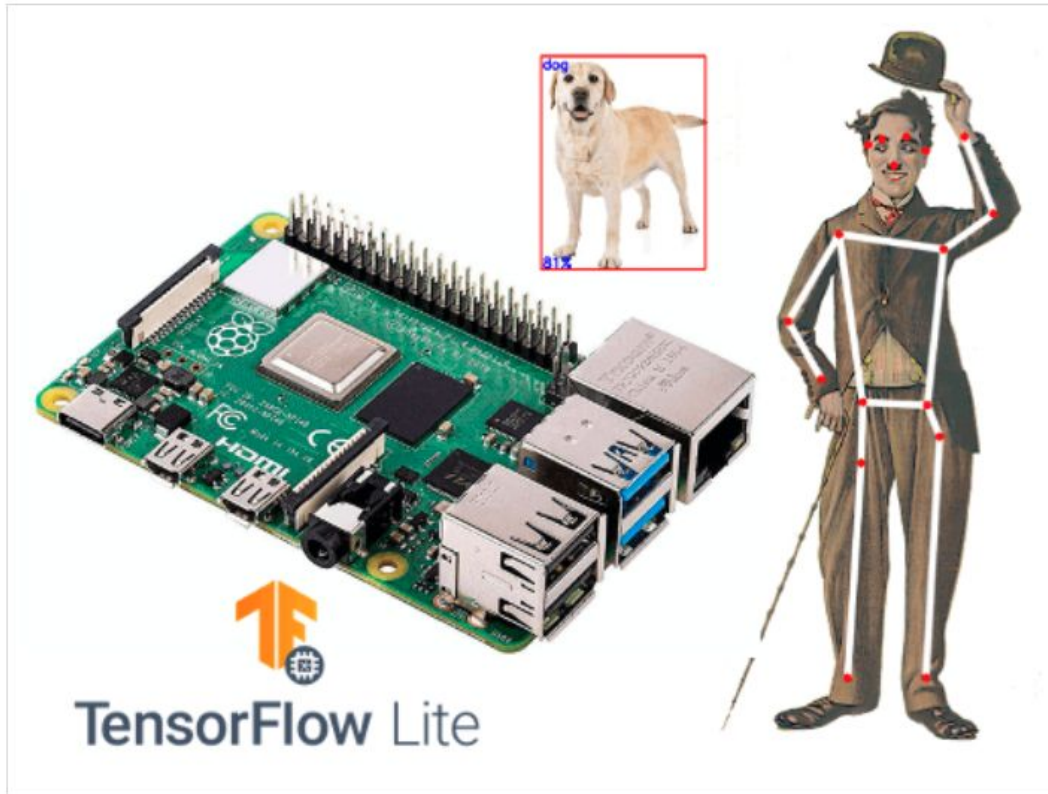


MJRoBot (Marcelo Rovai)
Published August 19, 2020 © GPL3+

Exploring IA at the Edge!

Image Recognition, Object Detection and Pose Estimation using Tensorflow Lite on a Raspberry Pi

 Intermediate  Full instructions provided  8 hours  2,231



<https://www.hackster.io/mjrobot/exploring-ia-at-the-edge-97588d>

Main references

- [Harvard School of Engineering and Applied Sciences - CS249r: Tiny Machine Learning](#)
- [Professional Certificate in Tiny Machine Learning \(TinyML\) – edX/Harvard](#)
- [Introduction to Embedded Machine Learning - Coursera/Edge Impulse](#)
- [Computer Vision with Embedded Machine Learning - Coursera/Edge Impulse](#)
- Fundamentals textbook: [“Deep Learning with Python” by François Chollet](#)
- Applications & Deploy textbook: [“TinyML” by Pete Warden, Daniel Situnayake](#)
- Deploy textbook [“TinyML Cookbook” by Gian Marco Iodice](#)

I want to thank **Shawn Hymel** and Edge Impulse, **Pete Warden** and **Laurence Moroney** from Google, Professor **Vijay Janapa Reddi** and **Brian Plancher** from Harvard, and the rest of the **TinyMLedu** team for preparing the excellent material on TinyML that is the basis of this course at UNIFEI.

The IESTI01 course is part of the **TinyML4D**, an initiative to make TinyML education available to everyone globally.

Thanks



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