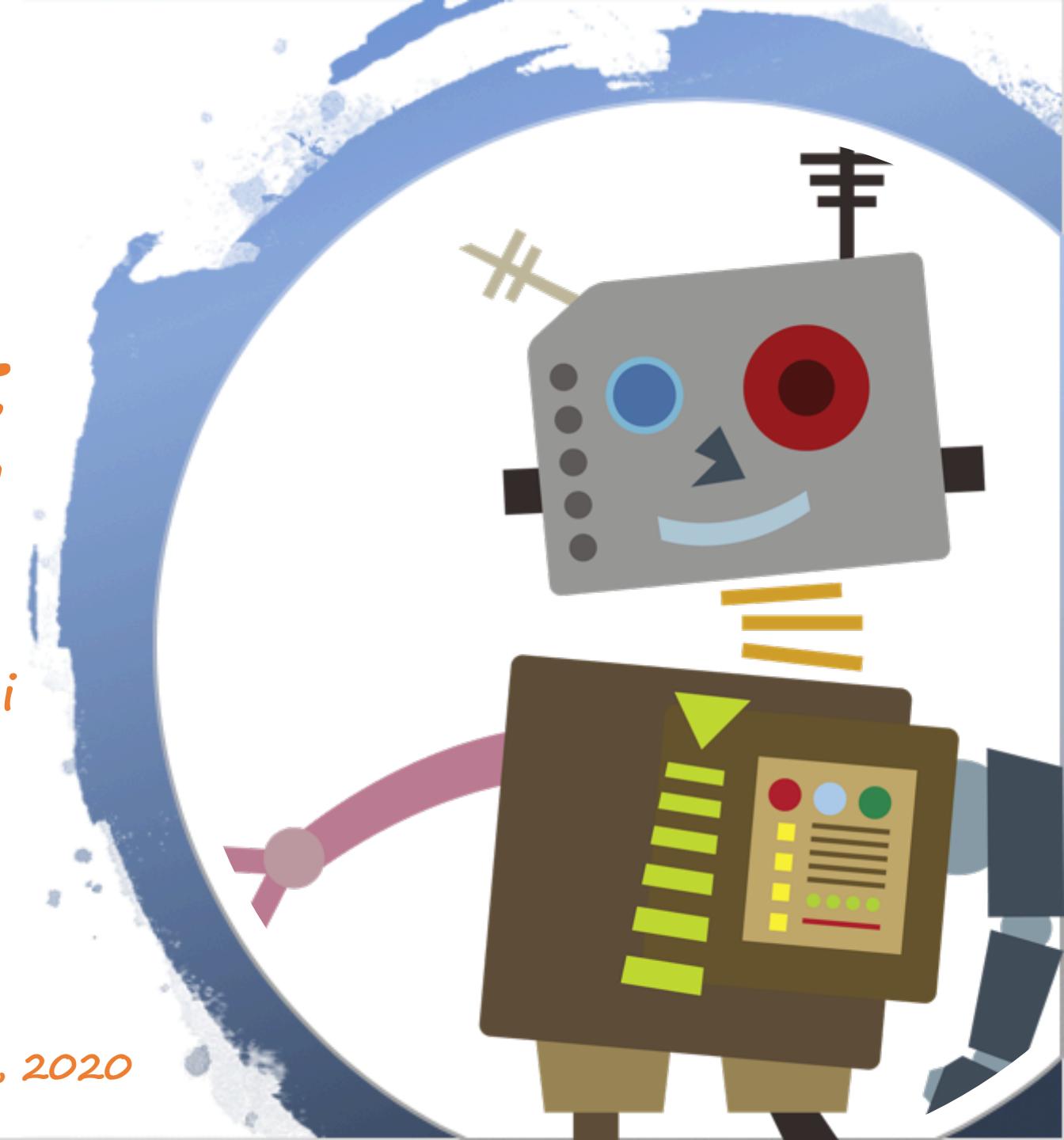


Exploring AI at the Edge!

By Marcelo Rovai

August 27th, 2020





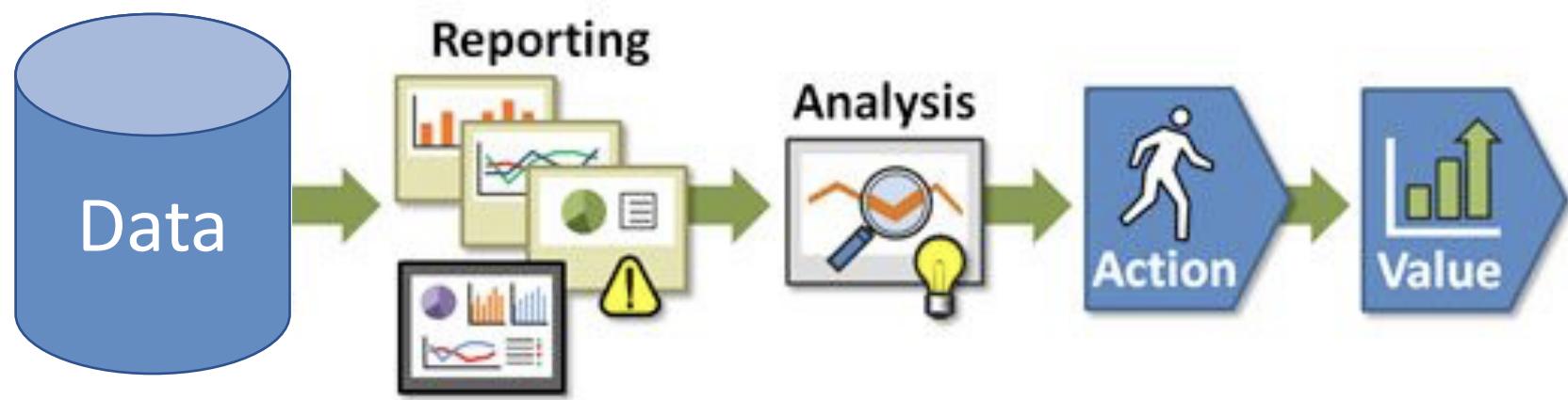
Marcelo Rovai

Brazilian from São Paulo, Master in Data Science by UDD, Chile, and MBA by IBMEC (INSPER), Brazil. Marcelo graduated in 1982 as an Engineer from UNIFEI with pos graduation by Poli/USP, both in Brazil. He worked as a teacher, engineer, and executive in several companies in the technology area such as AVIBRAS Aeroespacial, SID Informática, ATT-GIS, NCR, DELL, COMPAQ (HP) and more recently at IGT where he continues as Senior Advisor.

In 2016, Marcelo began writing about electronics, publishing his works in sites of the area as MJRoBot.org (Editor/Writer), Hackster.io (#1 Contributor), Instructables.com, and Medium.com (TDS – Towards Data Science). Besides winning several Instructables competitions in the areas of electronics, robotics, and IoT.

Marcelo lives with his wife Ilza in Santiago, Chile, where he divides his time between his consultant work and sharing ideas in the field of Data Science, Electronics, IoT, Physical Computing and Robotics.

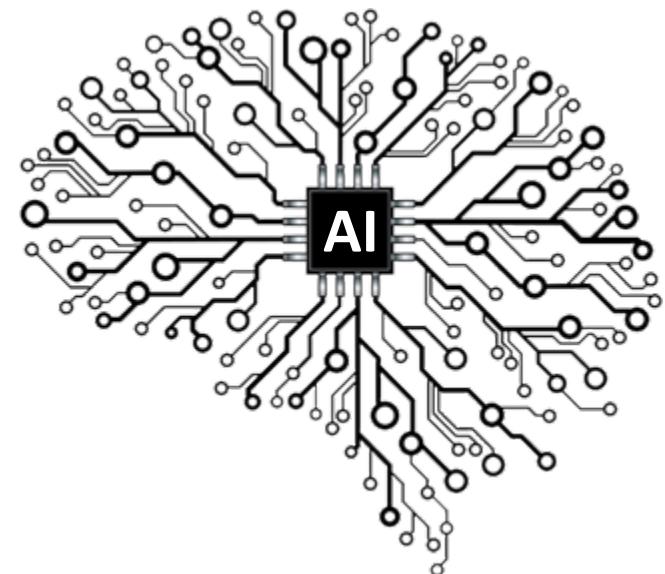
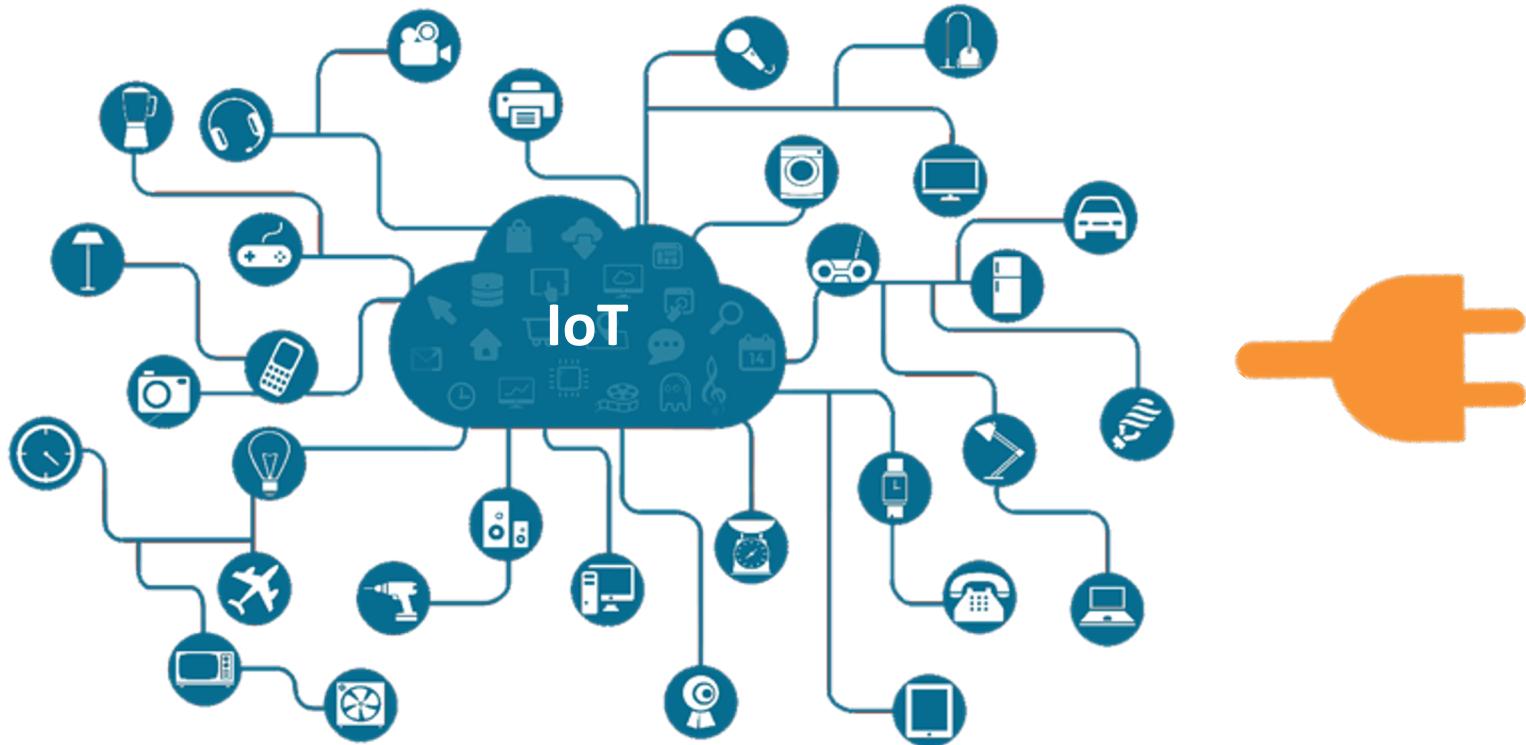
Data → ... → Value



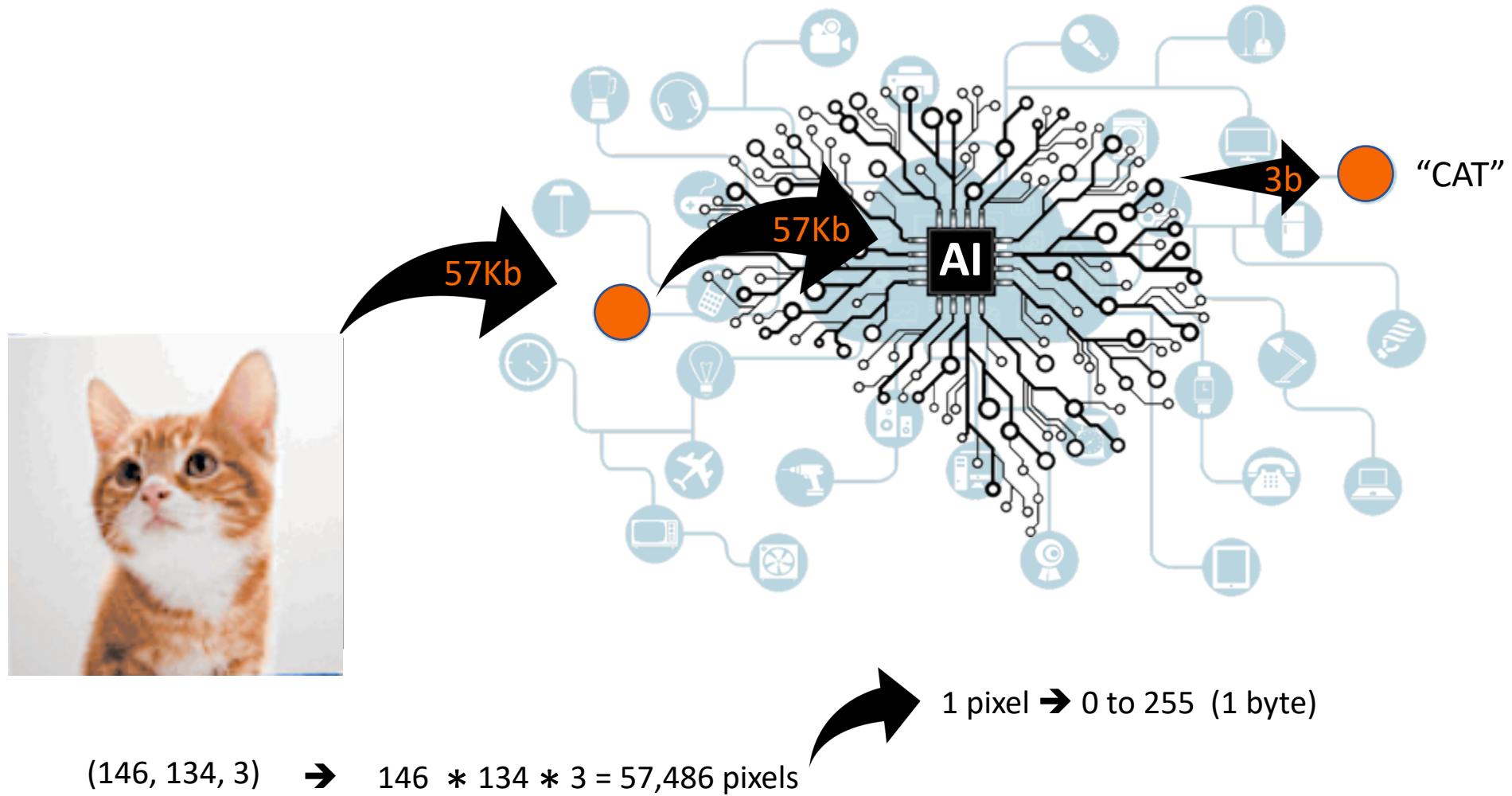
Data → ...→ Value



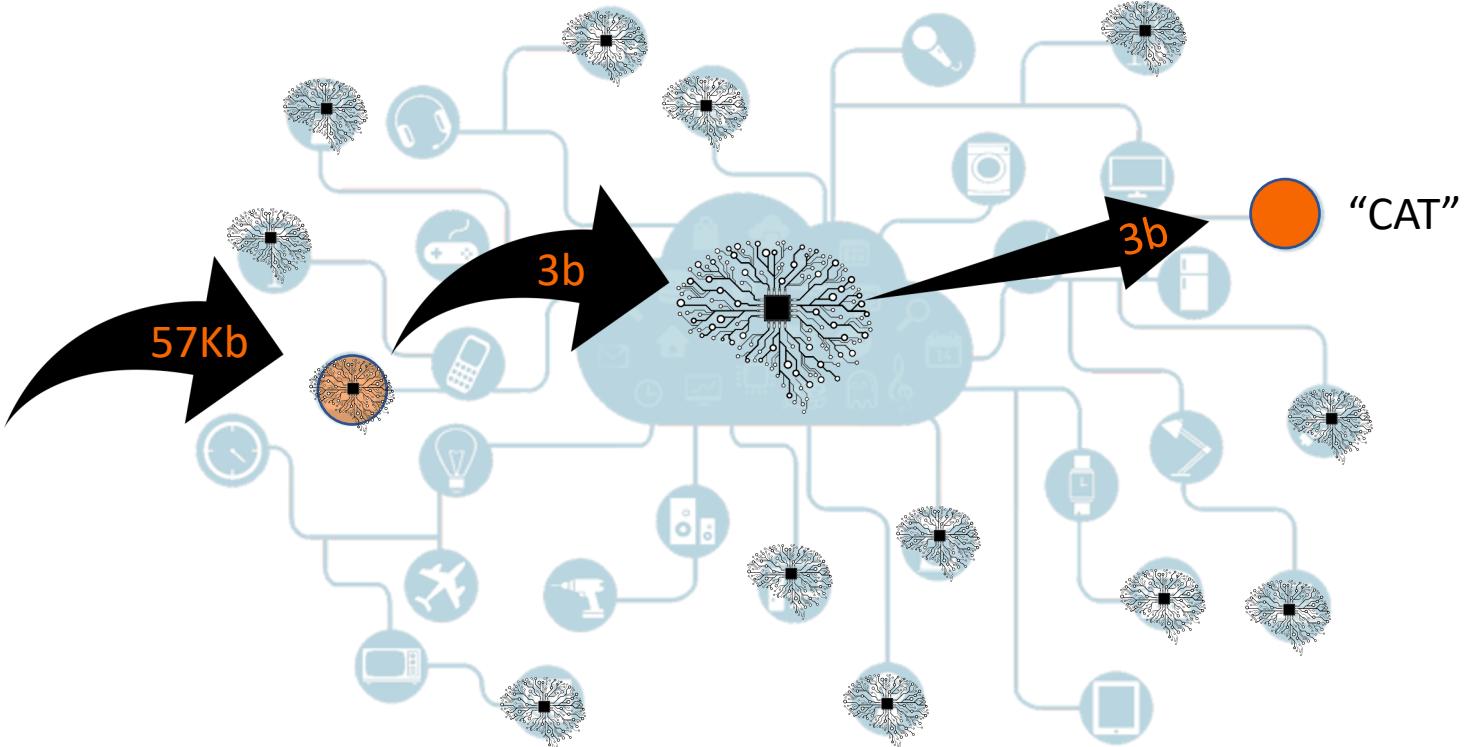
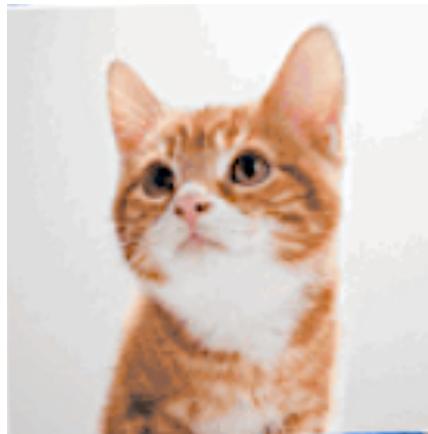
Data (IoT) → Value (AI)



Data (IoT) → Value (AI)



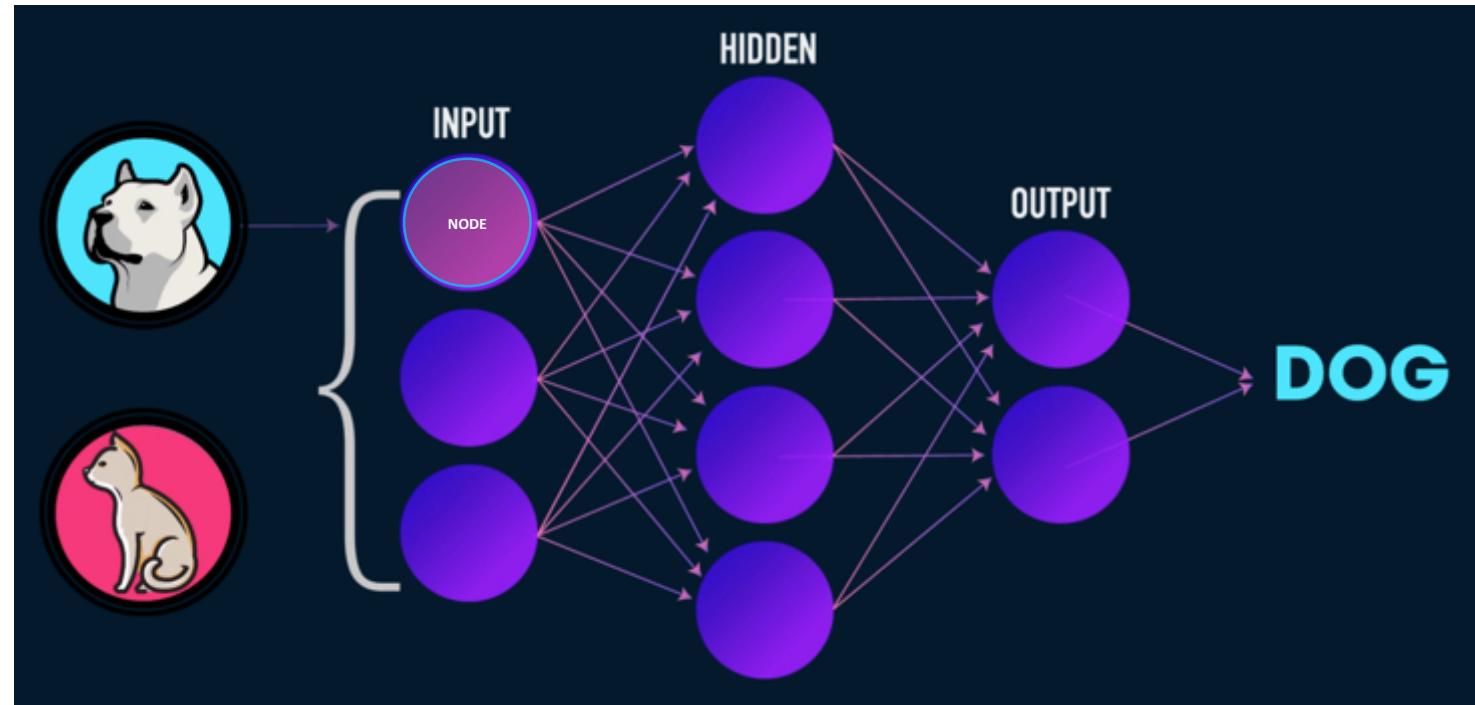
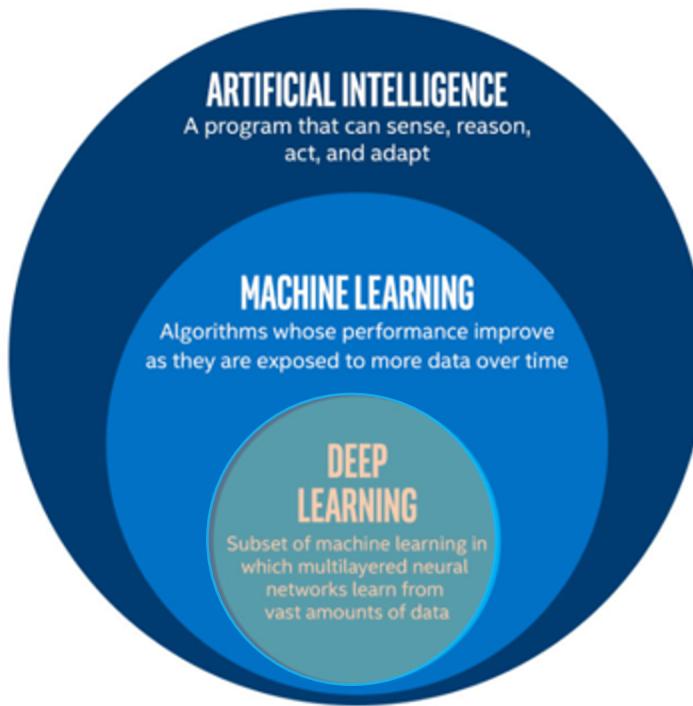
AI at the Edge → Edge AI



"The term "Edge AI" might be the new buzzword of 2019/2020, much like "Internet of Things" was in 2016/2017. "

AI → Deep Learning (DL)

Deep Learning: Subset of Machine Learning in which multilayered neural networks learn from vast amounts of data

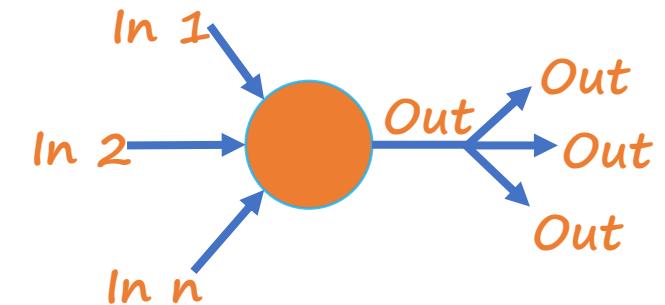
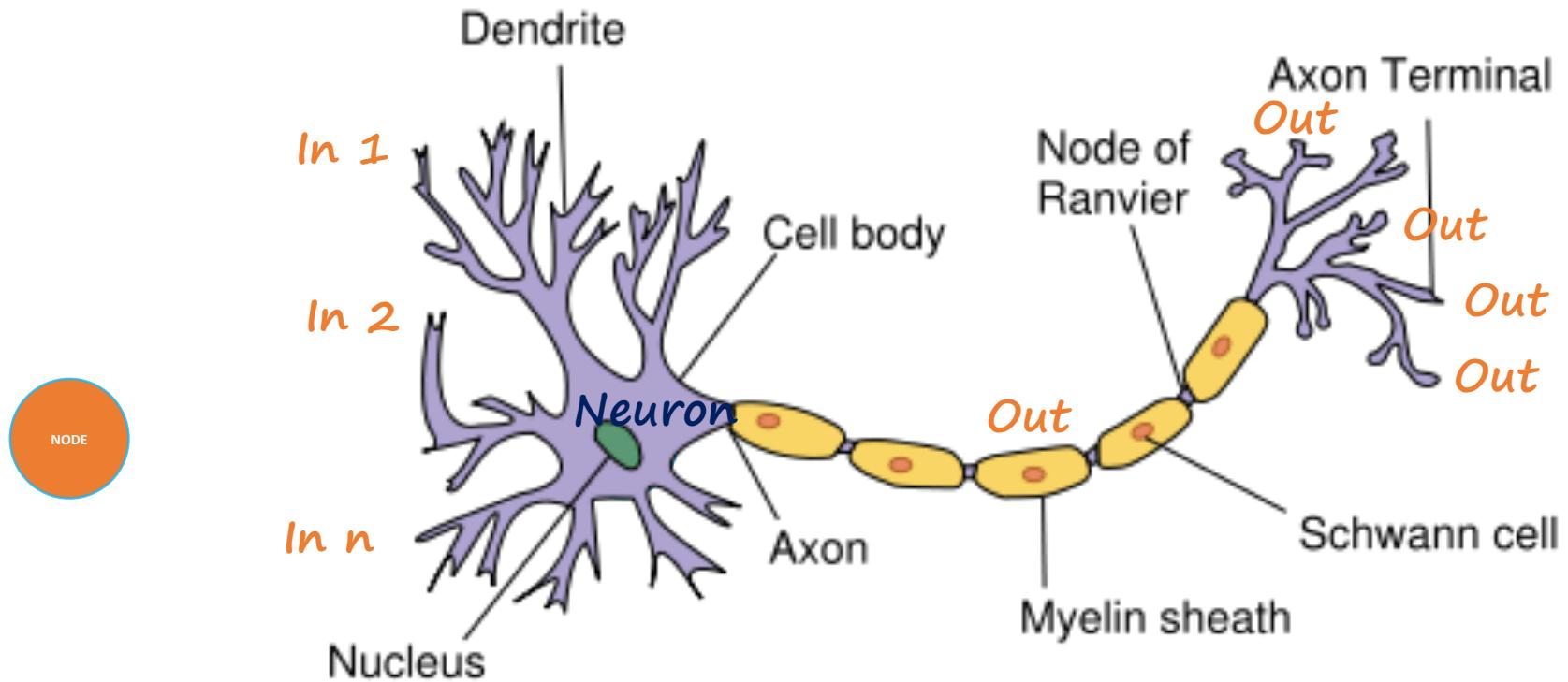


Source Image: <https://towardsdatascience.com/an-introduction-to-deep-learning-af63448c122c>

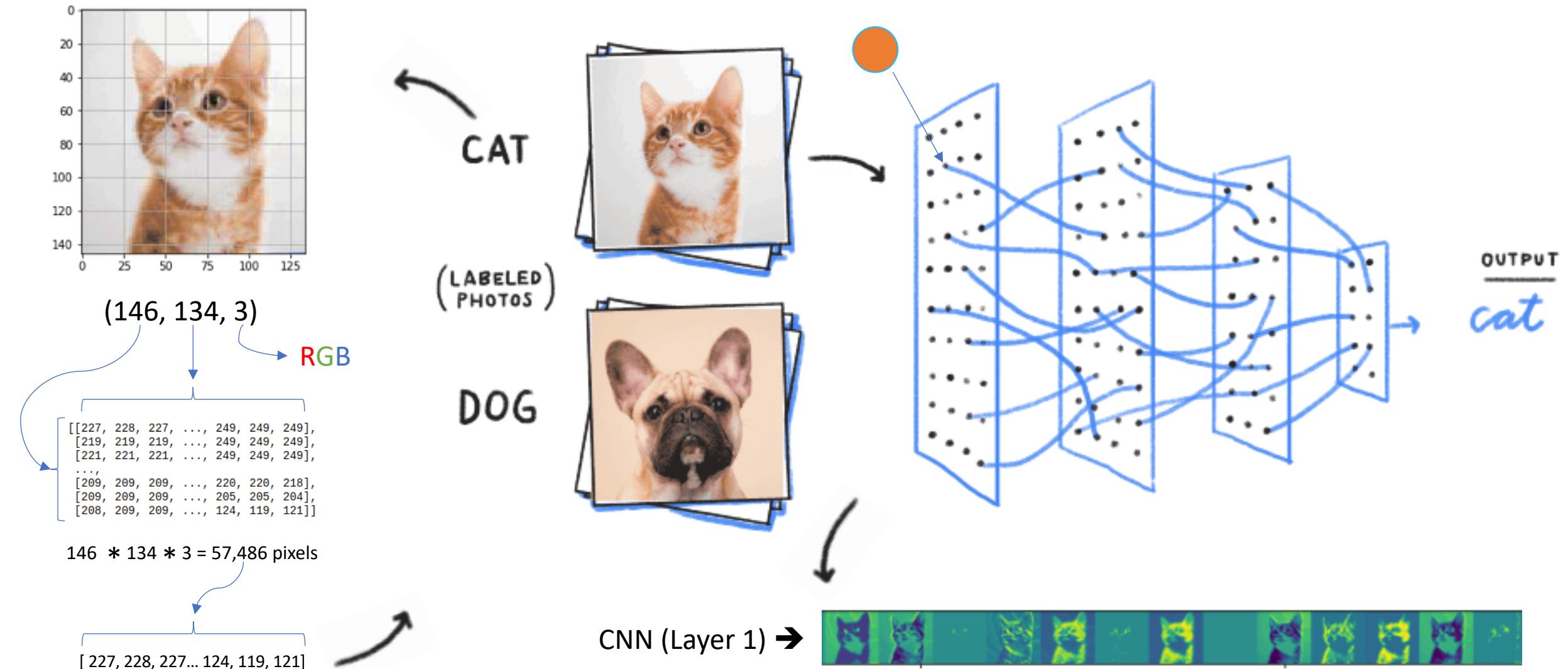
Neuron

meetup

IoT hacking & Data
Science Innovators



Artificial Neural Network



The Neural Network Zoo

meetup

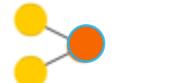
IoT hacking & Data
Science Innovators

- Input Cell
- Backfed Input Cell
- △ Noisy Input Cell
- Hidden Cell
- Probabilistic Hidden Cell
- △ Spiking Hidden Cell
- Capsule Cell
- Output Cell
- Match Input Output Cell
- Recurrent Cell
- Memory Cell
- △ Gated Memory Cell
- Kernel
- Convolution or Pool

A mostly complete chart of Neural Networks

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Perceptron (P)



Feed Forward (FF)



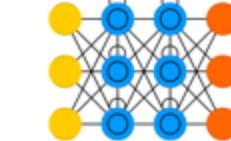
Radial Basis Network (RBF)



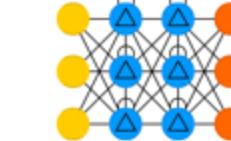
Recurrent Neural Network (RNN)



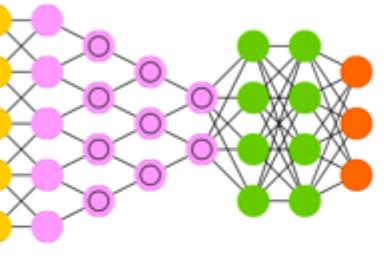
Long / Short Term Memory (LSTM)



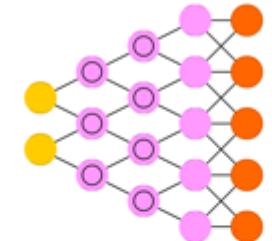
Gated Recurrent Unit (GRU)



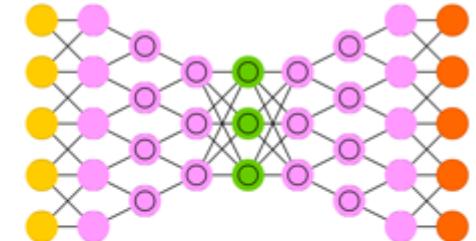
Deep Convolutional Network (DCN)



Deconvolutional Network (DN)



Deep Convolutional Inverse Graphics Network (DCIGN)



Auto Encoder (AE)



Variational AE (VAE)



Denoising AE (DAE)



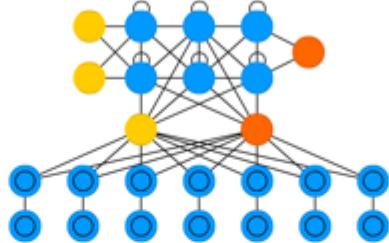
Sparse AE (SAE)



Deep Residual Network (DRN)



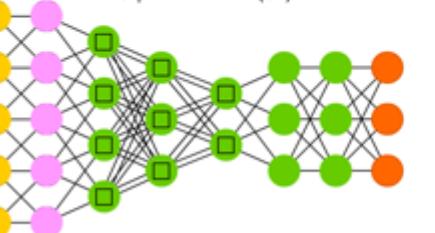
Differentiable Neural Computer (DNC)



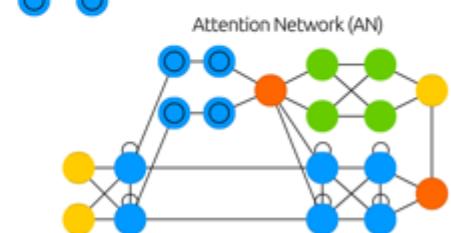
Neural Turing Machine (NTM)



Capsule Network (CN)



Kohonen Network (KN)



The Neural Network Zoo

meetup

IoT hacking & Data
Science Innovators

- Input Cell
- Backfed Input Cell
- △ Noisy Input Cell
- Hidden Cell
- Probabilistic Hidden Cell
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- Output Cell
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- Recurrent Cell
- Memory Cell
- △ Gated Memory Cell
- Kernel
- Convolution or Pool

A mostly complete chart of Neural Networks

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Perceptron (P)

Feed Forward (FF)

Radial Basis Network (RBF)

Recurrent Neural Network (RNN)

Long / Short Term Memory (LSTM)

Gated Recurrent Unit (GRU)

RNN
(NLP and Time Series)

Auto Encoder (AE)

Variational AE (VAE)

Denoising AE (DAE)

Sparse AE (SAE)

AE
(Filtering)

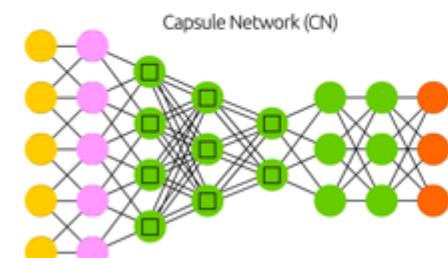
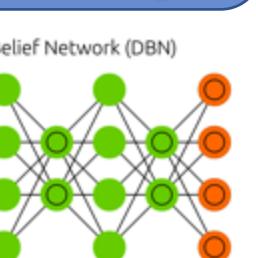
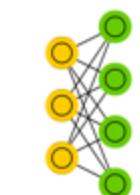
Markov Chain (MC)

Hopfield Network (HN)

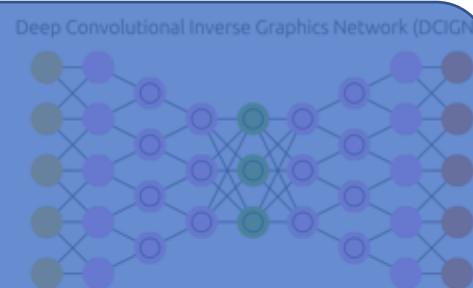
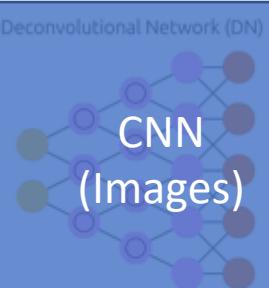
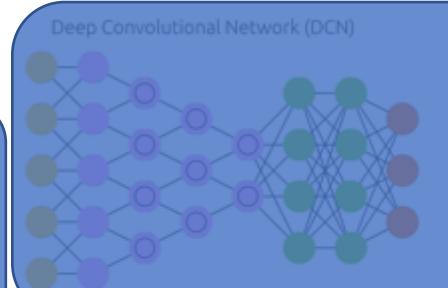
Boltzmann Machine (BM)

Restricted BM (RBG)

Deep Belief Network (DBN)



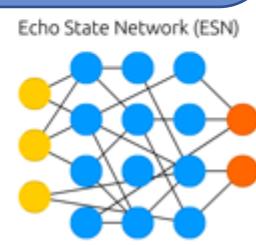
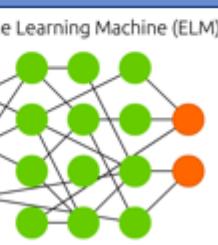
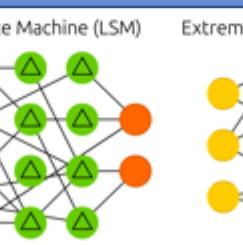
Deep Feed Forward (DFF)
(ML)



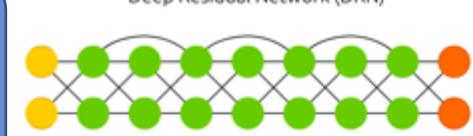
GAN
(Deep Fake)



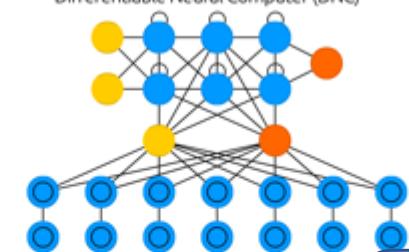
Liquid State Machine (LSM)



Deep Residual Network (DRN)



Differentiable Neural Computer (DNC)



Neural Turing Machine (NTM)

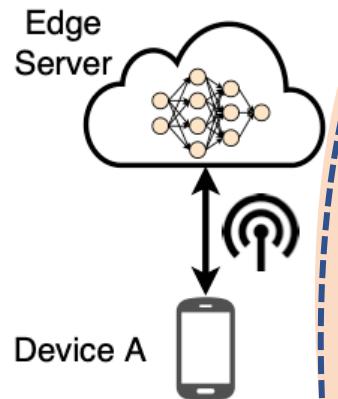
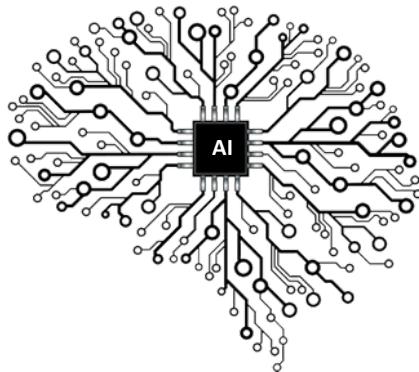


Kohonen Network (KN)

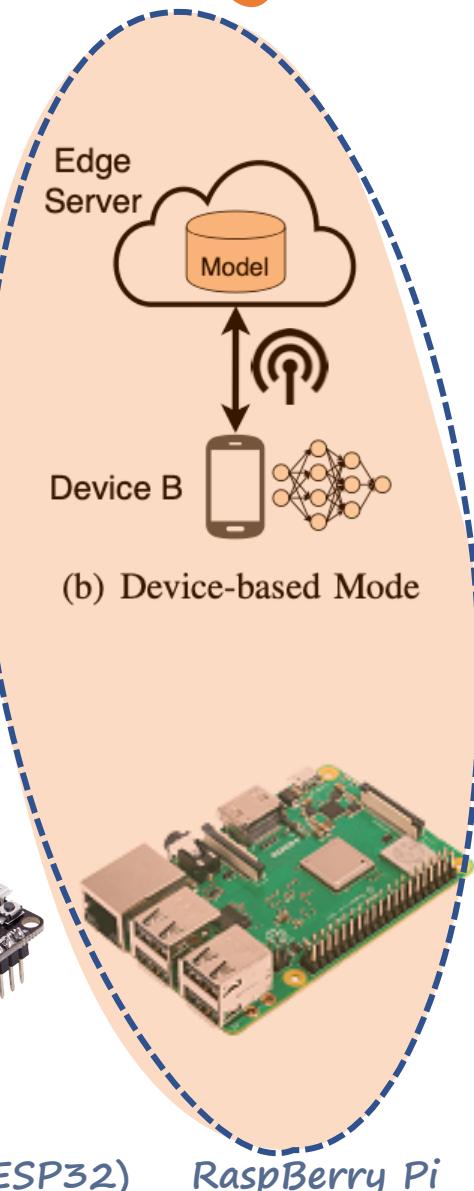


AE
(Transformer)
(GPT)

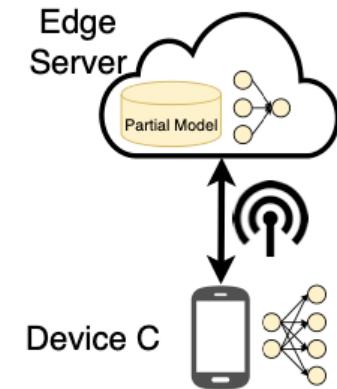
AI → Edge Computing



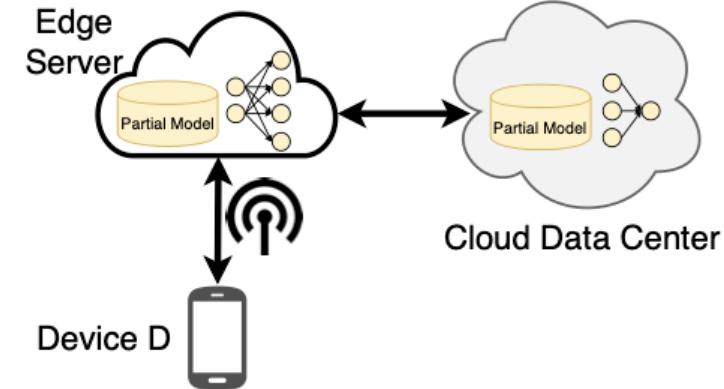
(a) Edge-based Mode



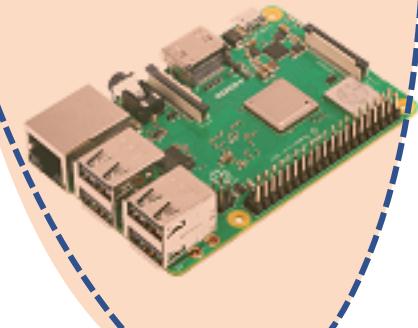
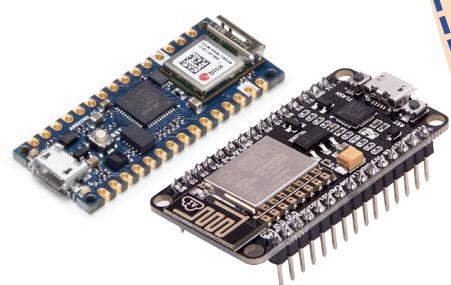
(b) Device-based Mode



(c) Edge-Device Mode

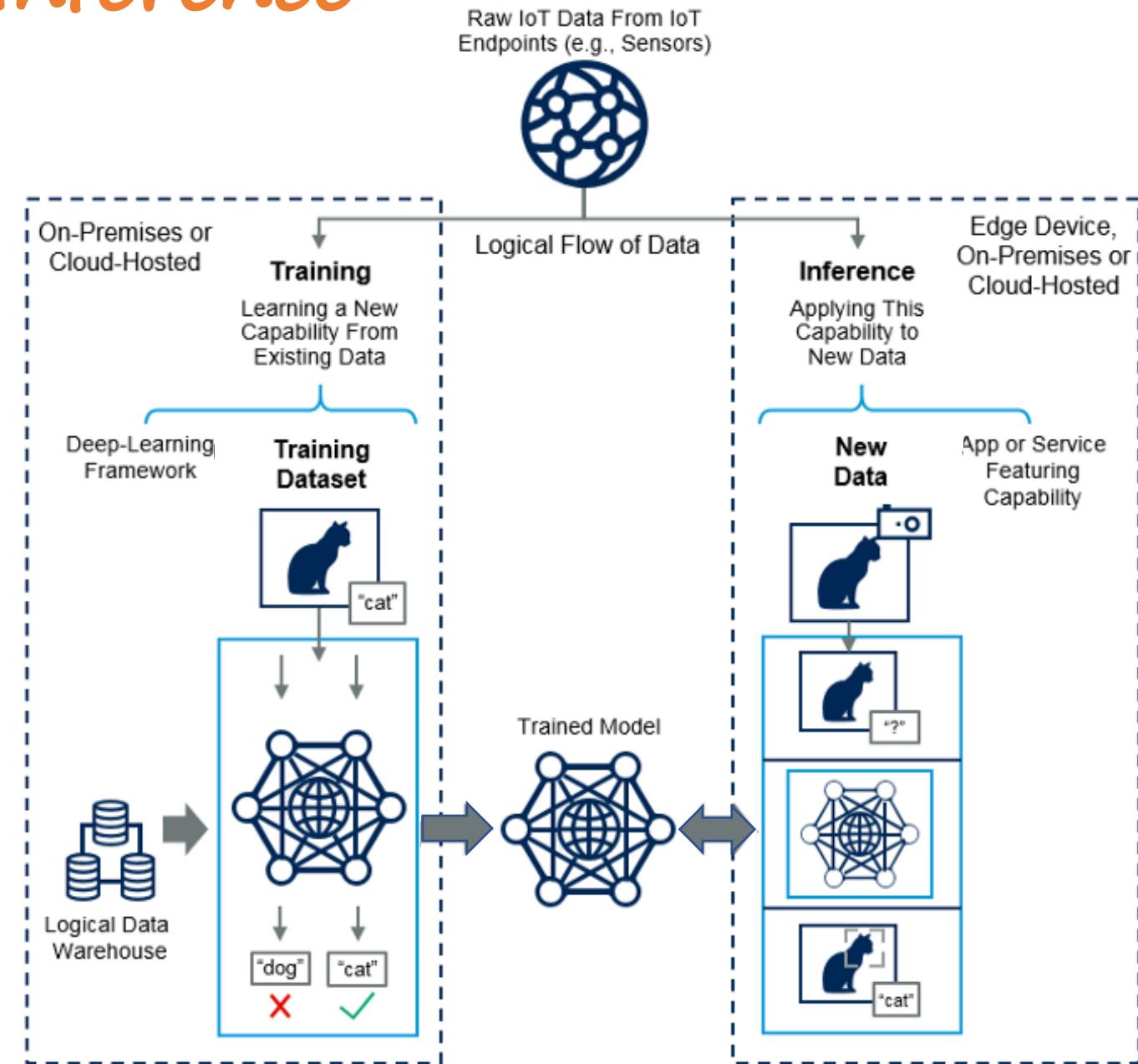


(d) Edge-Cloud Mode



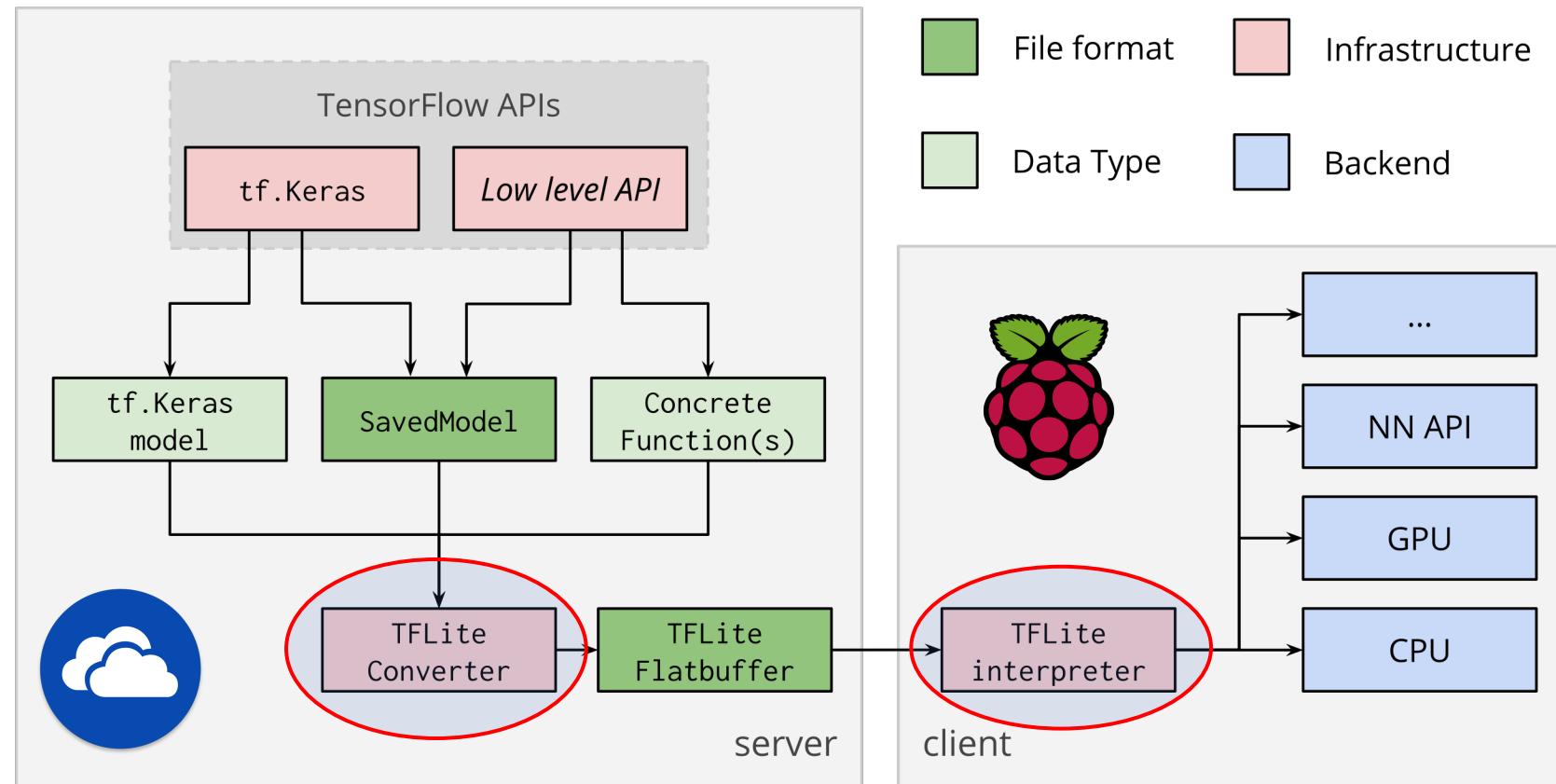
<https://arxiv.org/pdf/1905.10083.pdf>

Training vs. Inference





TensorFlow Lite



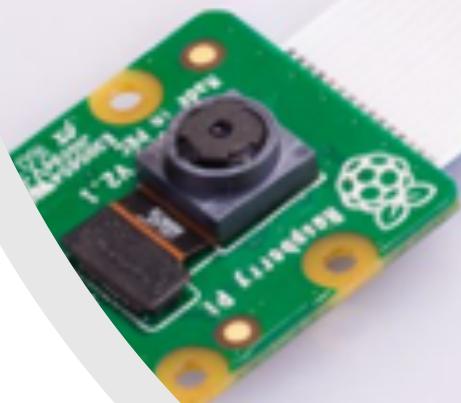
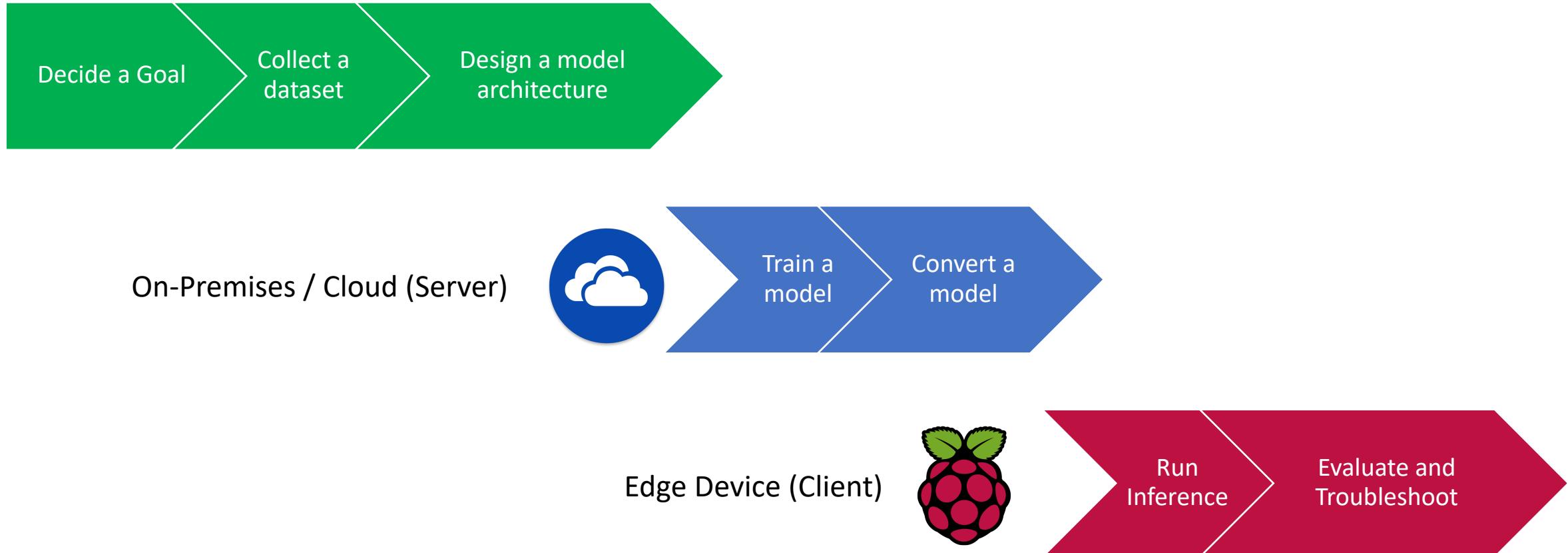
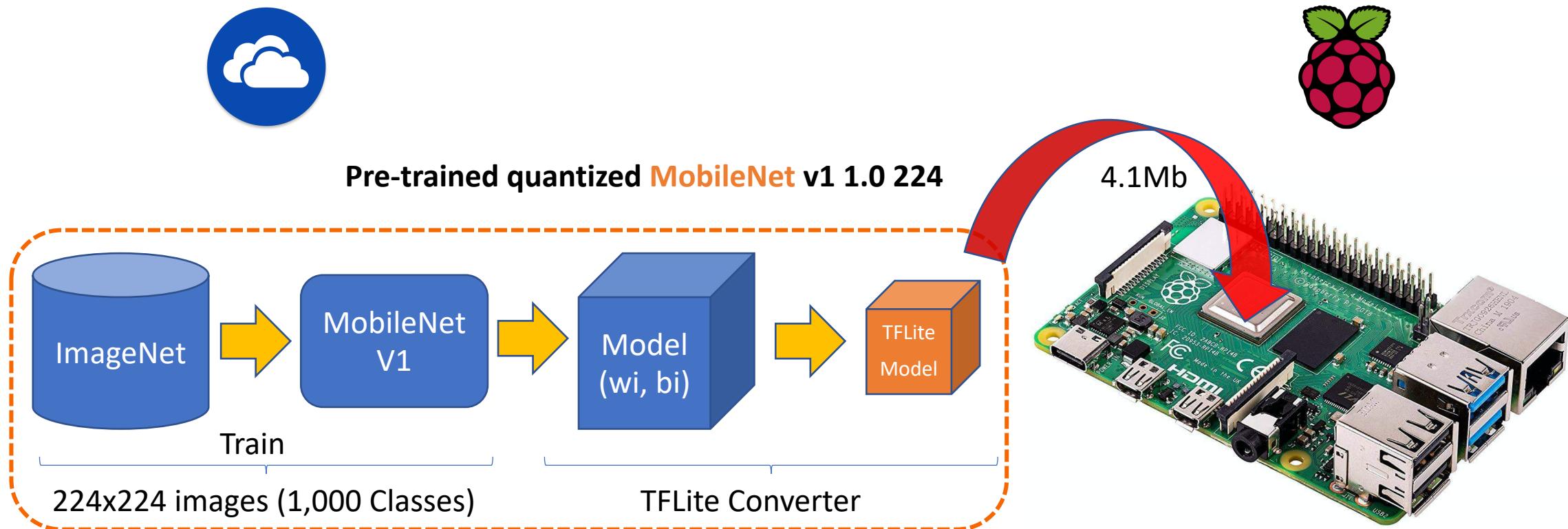


Image Classification &
Object Detection
Live DEMO

Deep Learning Workflow



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}]]
```

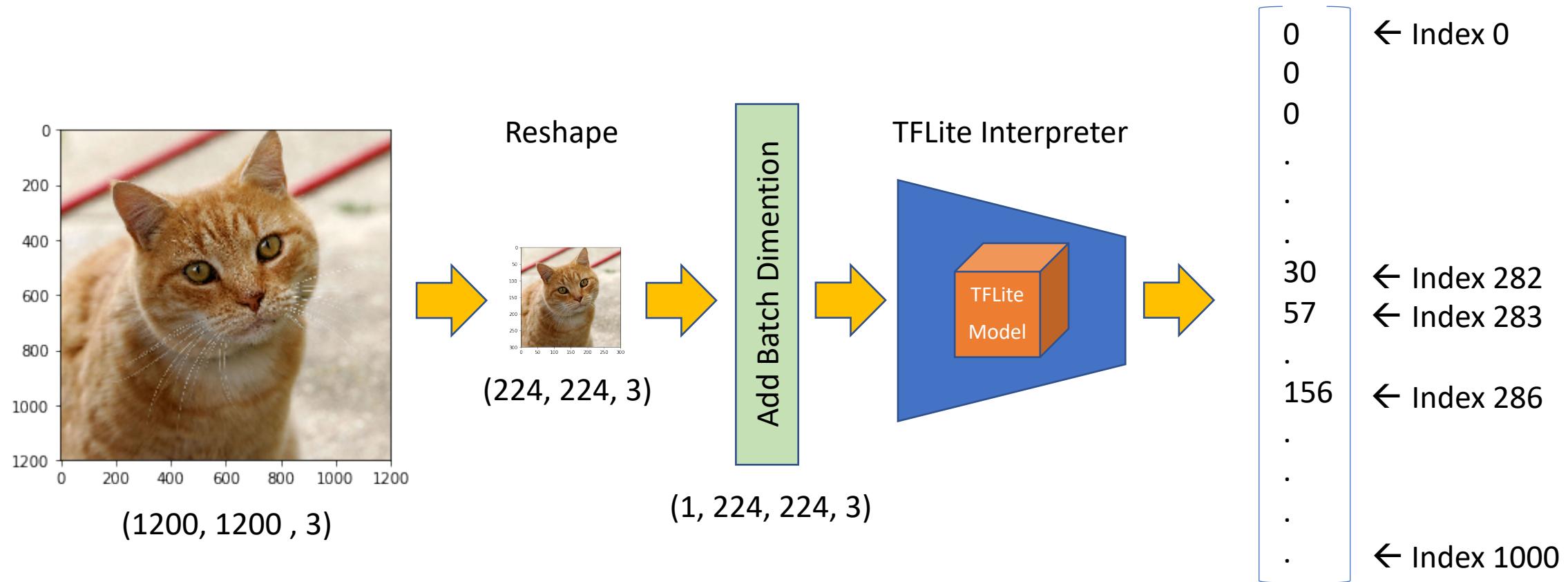
Input Image Shape

```
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 – Results

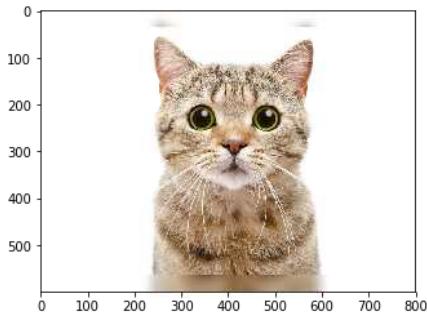
meetup

IoT hacking & Data
Science Innovators

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

[PREDICTION]

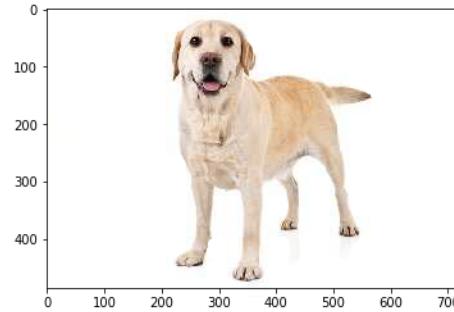
Egyptian cat : 64%
tabby : 14%
bucket : 3%



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

[PREDICTION]

Labrador retriever : 83%
golden retriever : 13%
bloodhound : 0%



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

[PREDICTION]

German shepherd : 60%
dhole : 16%
malinois : 7%



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

[PREDICTION]

minivan : 76%
convertible : 6%
jeep : 4%



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

[PREDICTION]

liner : 85%
trimaran : 11%
aircraft carrier : 0%



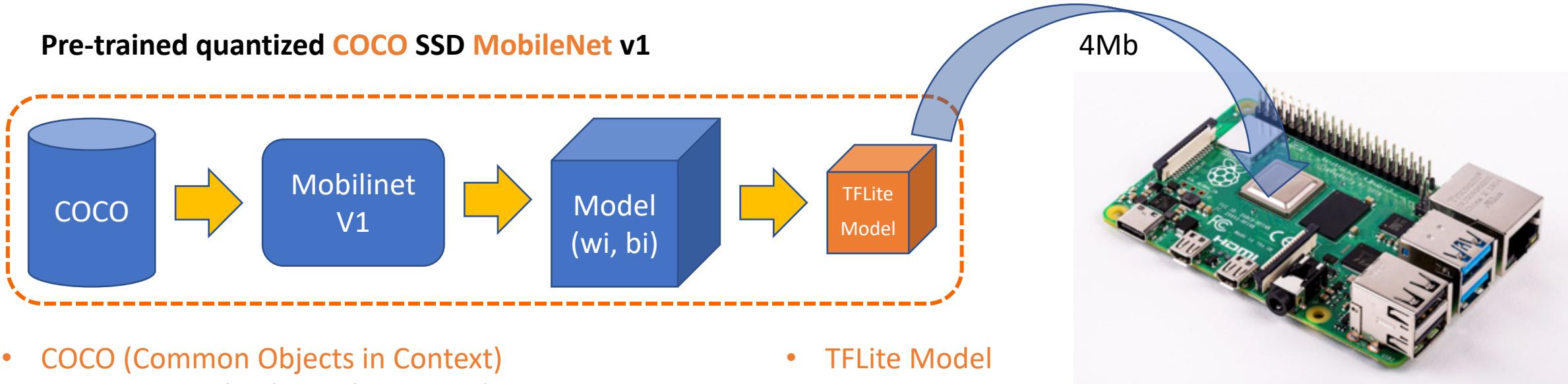
```
1 image_path = './images/table.jpg'  
2 image_classification(image_path, labels)
```

[PREDICTION]

dining table : 40%
table lamp : 34%
desk : 17%



Demo 2 - Object Detection

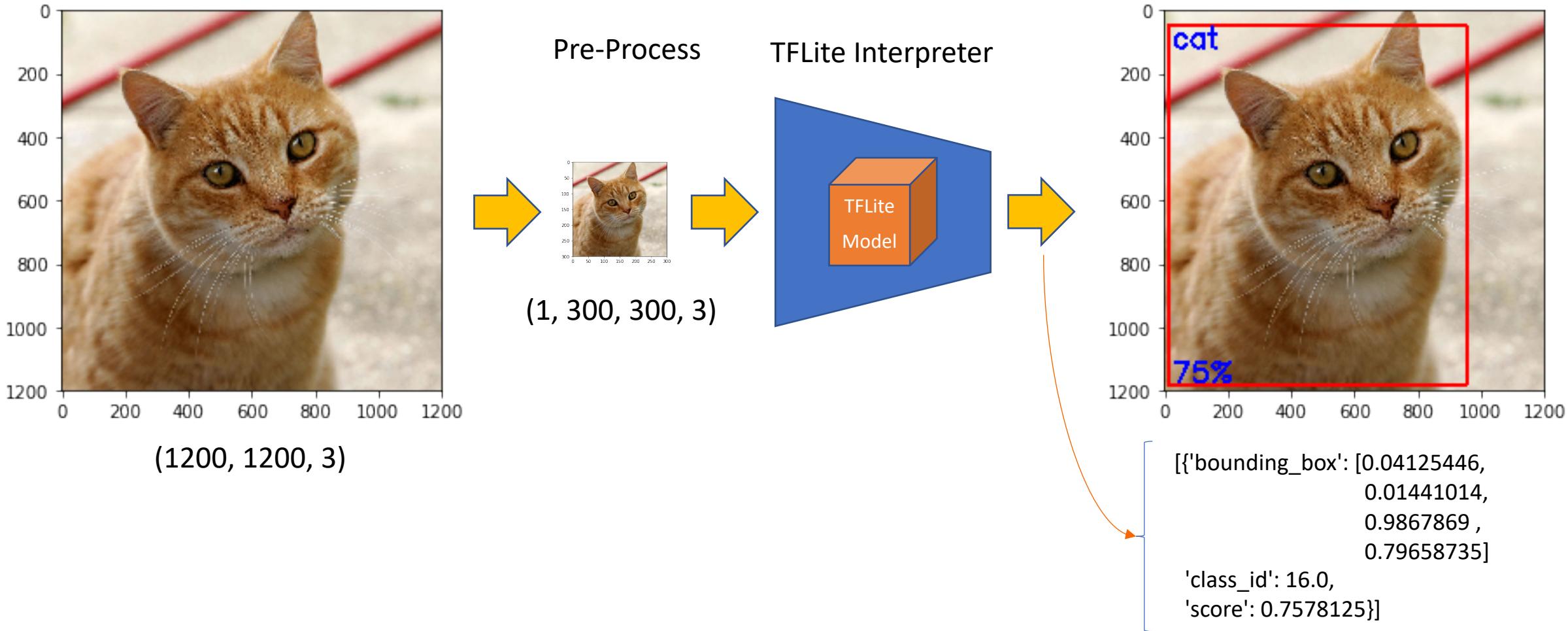


- **COCO (Common Objects in Context)**
 - Large-scale object detection dataset
 - 200K labeled images
 - 91 Stuff categories
- **MobileNet**
 - “Depth- wise Separable” convolutions
 - Introduced by Google in 2017
 - Similar performance with state-of-the-art architectures (as VGG or Inception)
 - Much smaller network (20% of VGG parameters (7M)
- **TFLite Model**
 - Input:
 - Image: 300x300x3
 - Flattened Buffer: 270K bytes
 - Each byte: 0 to 255
 - Output:
 - Bounding Box → (ymin, xmin, ymax, xmax)
 - Class ID (0 to 89) ("Stuff categories")
 - Score (0 to 1)

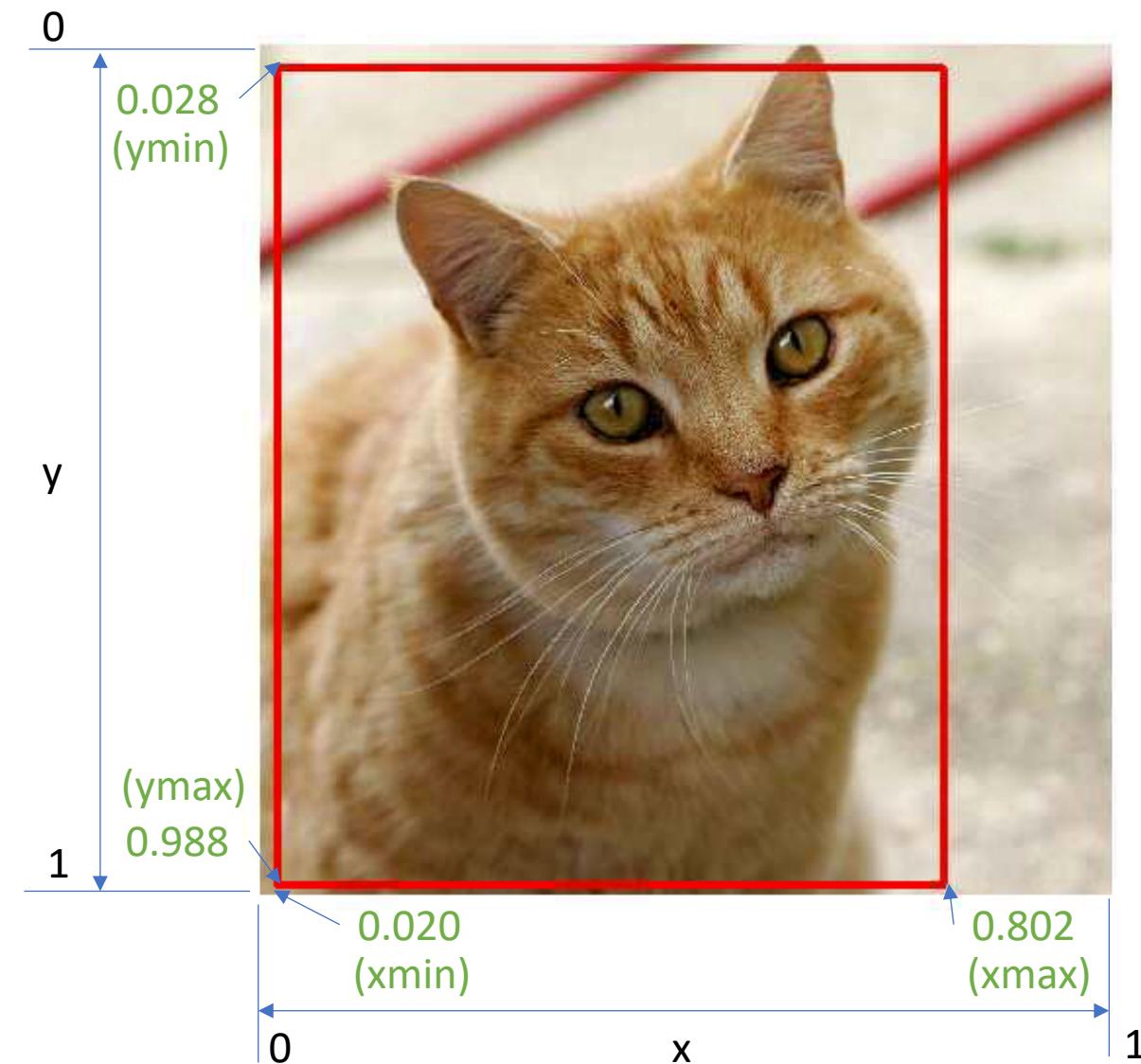
COCO Paper: <https://arxiv.org/pdf/1405.0312.pdf>

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

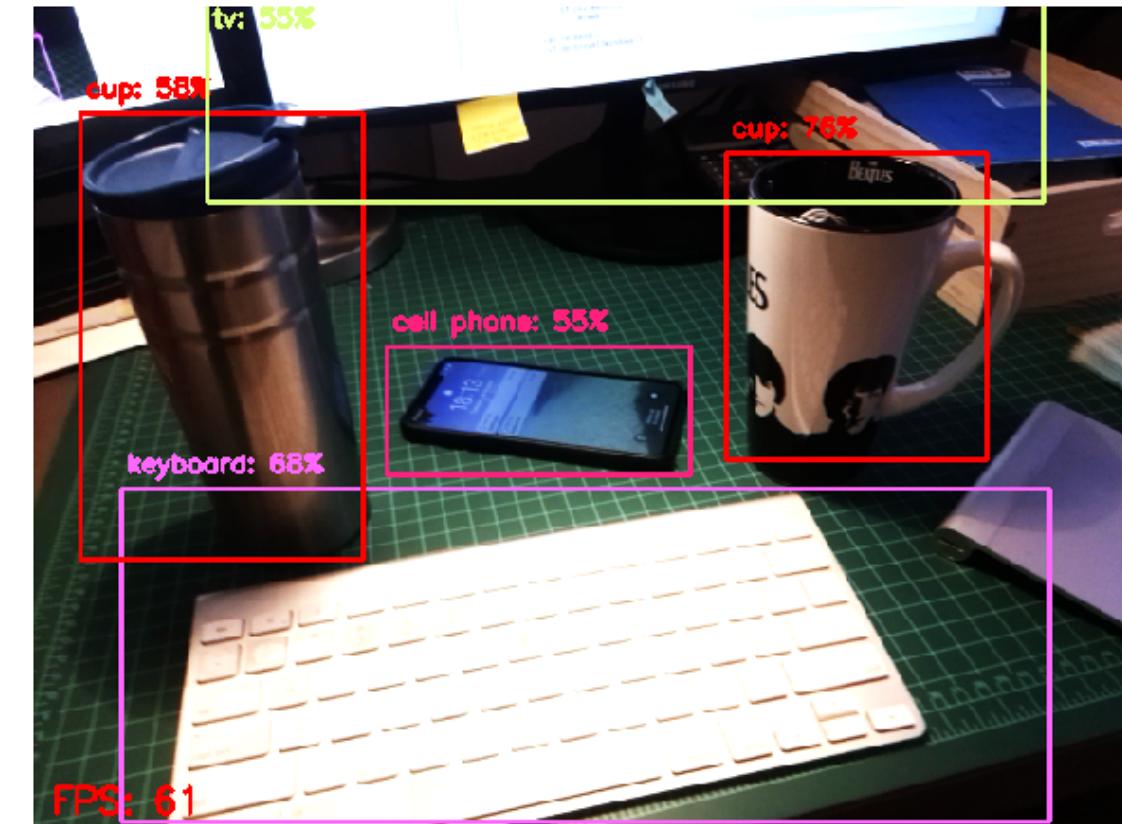
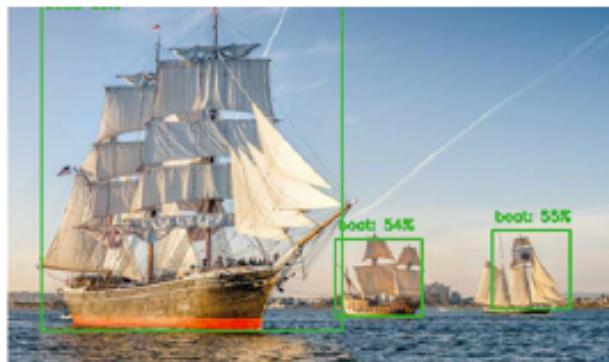
TF Lite Inference - Object Detection



Bounding Box- Object Detection



TF Lite Inference – Demo – Results

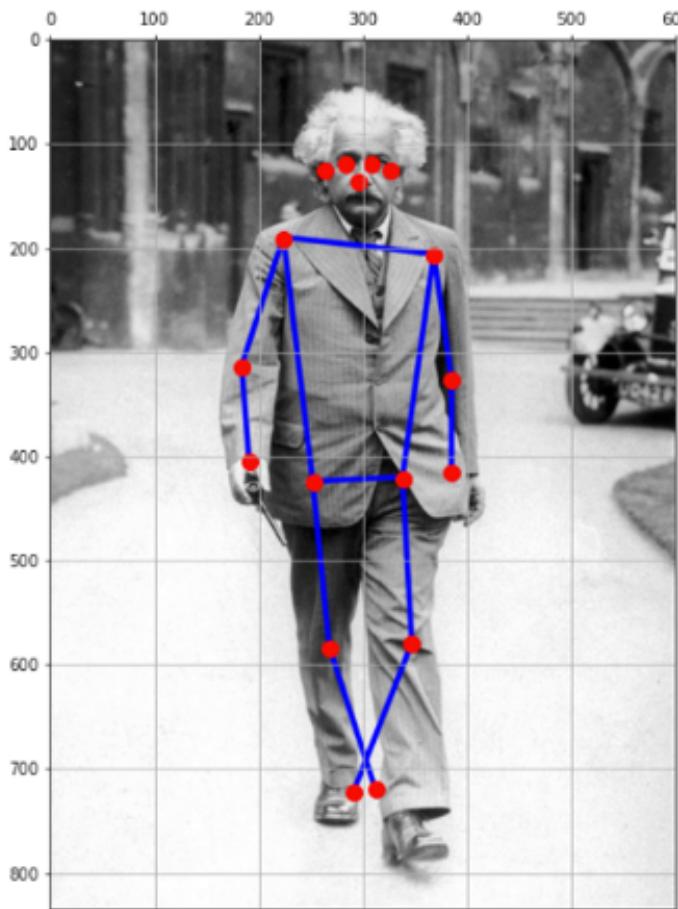


Photos

Live Video

TF Lite Inference – Other applications

Pose Estimation on RPi



TinyML

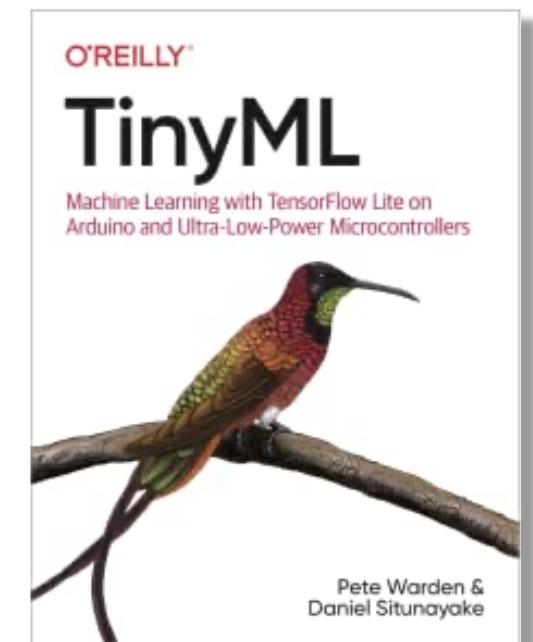
– Machine Learning on Ultra-Low-Power Microcontrollers

The world of Embedded devices:

- Run a neural network model at an energy cost of below 1 mW
- Devices that can run on coin battery by a year
- 32 bit Microcontrollers that cost less than \$1
- ARM's Cortex-M series of chips

Applications as:

- Speech Recognition (Wake-Word Detection)
- Detecting Gestures (iWatch)
- Detection of Objects on camera



Learning more about Edge AI

- Deploy machine learning models on mobile and IoT devices:
 - <https://www.tensorflow.org/lite>
- Teachable Machine: Train a computer to recognize your own images, sounds, & poses:
 - <https://teachablemachine.withgoogle.com>
- Machine Learning on Mobile and Edge Devices with TensorFlow Lite: Daniel Situnayake at QCon SF
 - <https://www.infoq.com/news/2019/11/tensorflow-lite-edge-qcon/>
- What is Edge AI? Machine Learning + IoT
 - <https://www.digikey.com/en/maker/projects/what-is-edge-ai-machine-learning-iot/4f655838138941138aad62c170827af>
- Exploring AI at the Edge!
 - <https://towardsdatascience.com/exploring-ia-at-the-edge-b30a550456db>

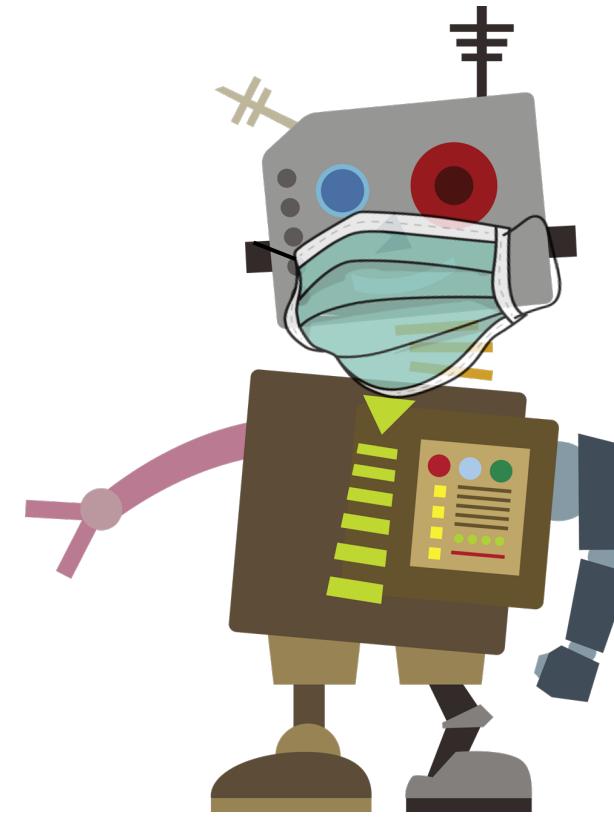
meetup

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“To infinity and beyond!”





MJRoBot.org

github.com/Mjrovai

hackster.io/mjrobot

medium.com/@rovai

[instructables.com/member/mjrovai](https://www.instructables.com/member/mjrovai)



Thanks
And keep safe!