



Make your IoT Smarter  
with Tensorflow Lite ...  
... to Design the Future of Vertical Farming



@alexis0duque





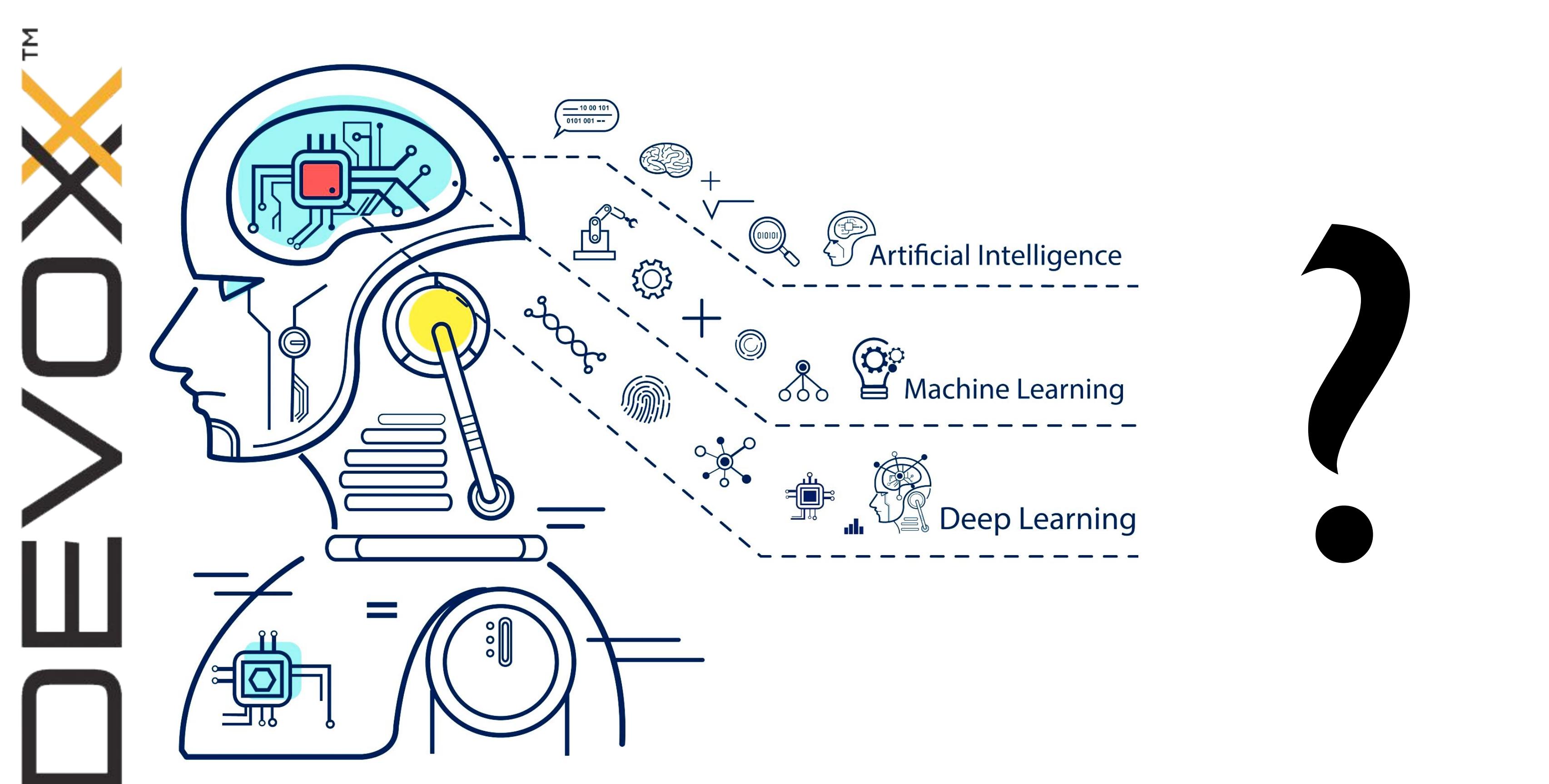
# Who am I?

**Alexis DUQUE**

Director of Research & Development

-  @alexis0duque
-  alexisduque
-  alexisd@rtone.fr
-  alexisduque.me
-  <https://goo.gl/oNUWu6>





Devoxx  
IoT  
Tensorflow  
AI



TensorFlow Lite

?

TM  
Devoxx



?

#Devoxx #IoT #Tensorflow #AI

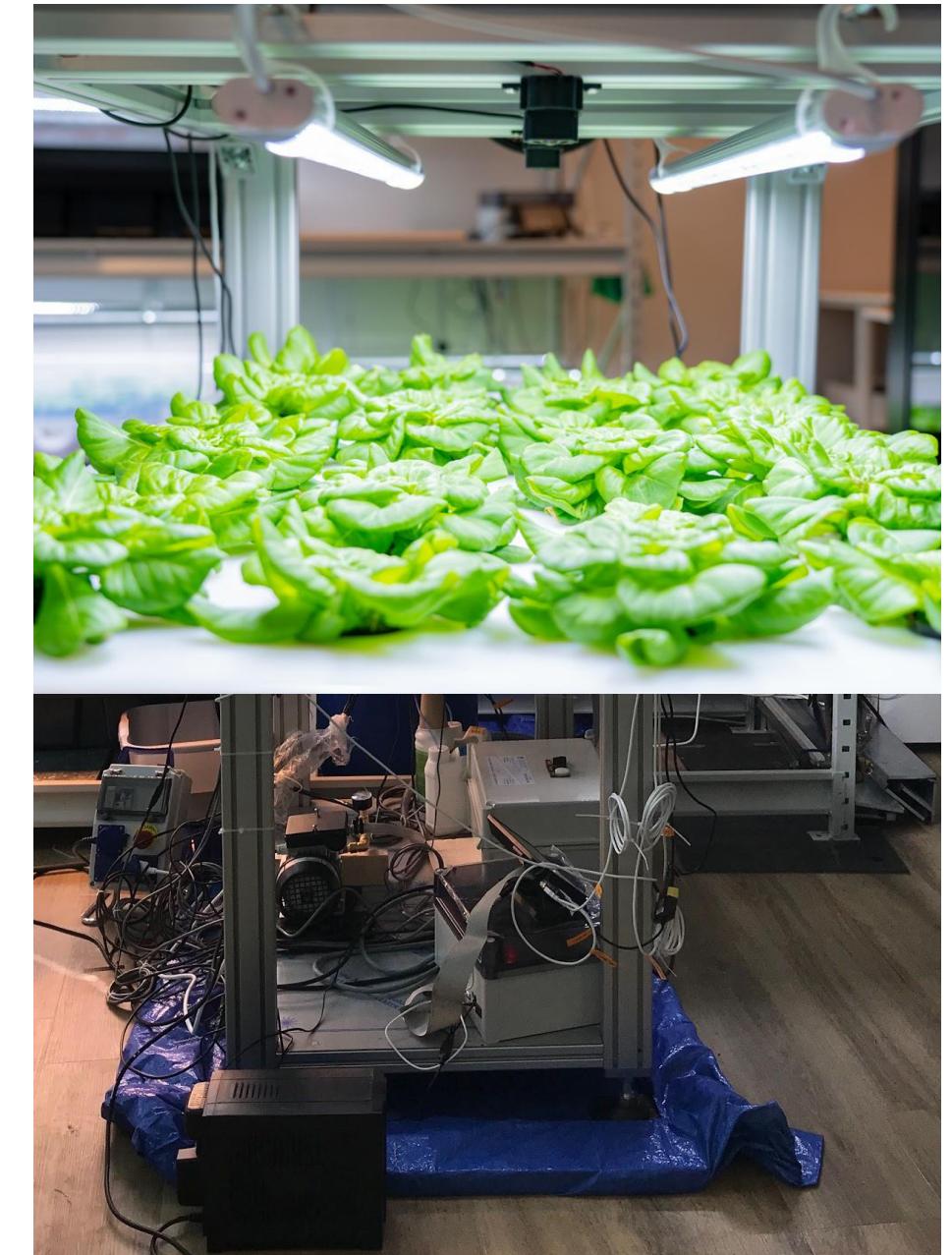
@alexis0duque

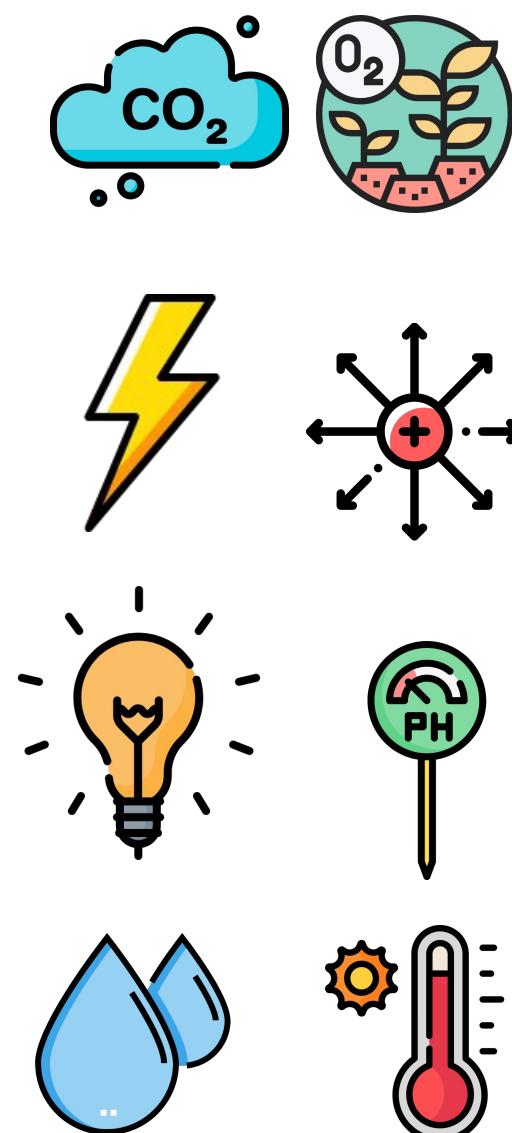
# Outline & What You Will Learn

- **Indoor Vertical Farming**
- **Why Intelligence at the Edge?**
- Introduction to **Tensorflow Lite**. How to Use It?
- **Setup** your laptop & RPI
- **Build** and **train** a small model to **predict lettuce weight**
- **Convert** and deploy it on a RPI
- Run **predictions** on IoT devices
- Benchmarks
- Further Work



# Indoor Vertical Farming





avg **co2** (double, in ppm)

avg **dissolved oxygen** (double, in ppm)

avg **electrical conductivity** (double, in  $\mu\text{S}/\text{cm}$ )

avg **RedOx potential** (double, in mV)

avg **PPFD** (Photosynthetic Photon Flux Density,  
double, in  $\mu\text{mol}/\text{m}^2/\text{s}$ )

avg **water pH** (double, in -)

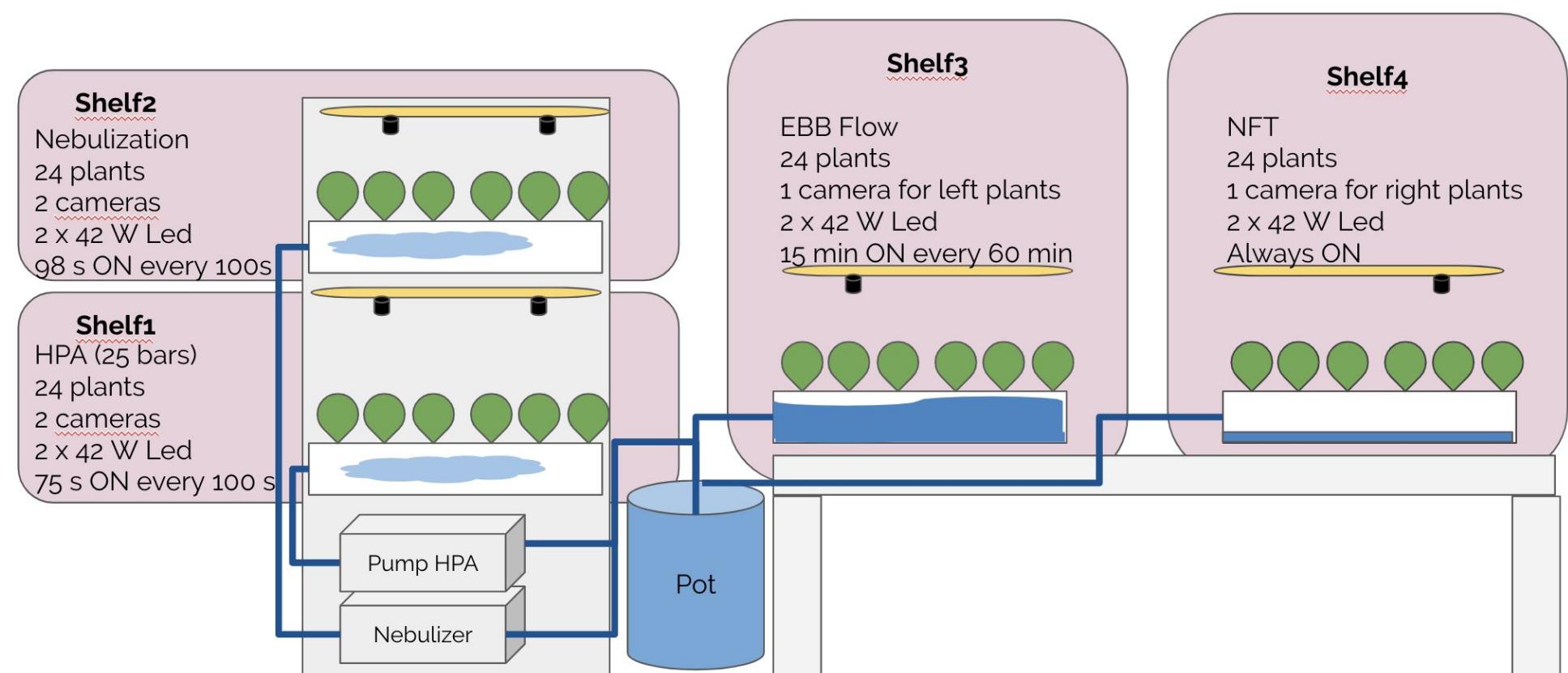
average **humidity** (double, in %)

average **temperature** (double, in  $^{\circ}\text{C}$ )



# System Architecture

## Actuators and Irrigation System



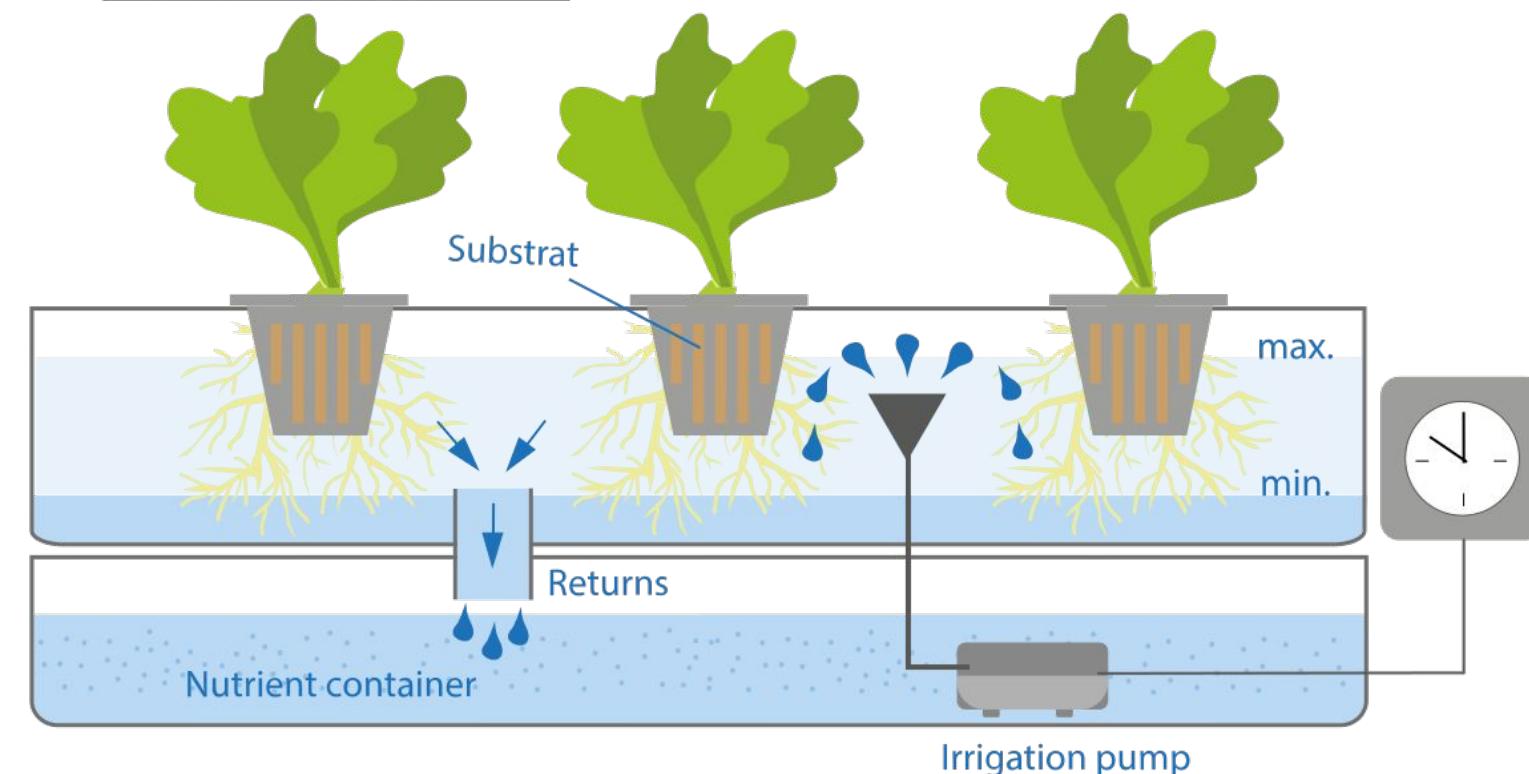
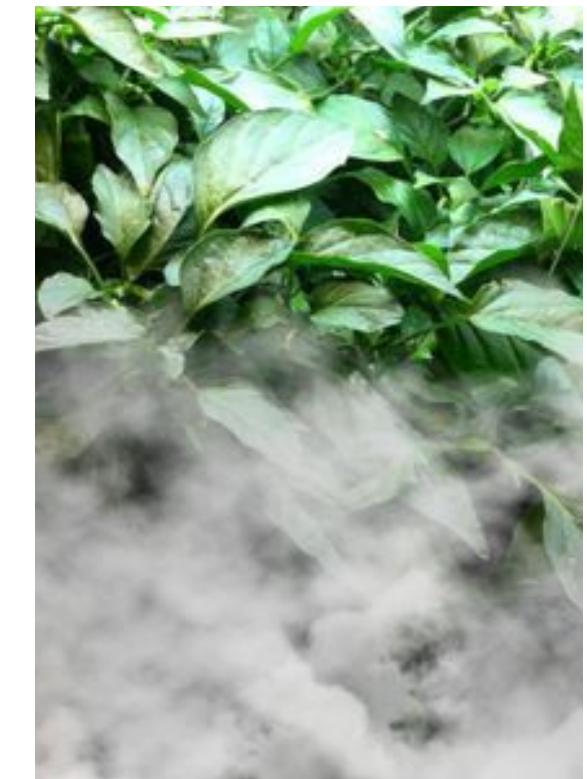
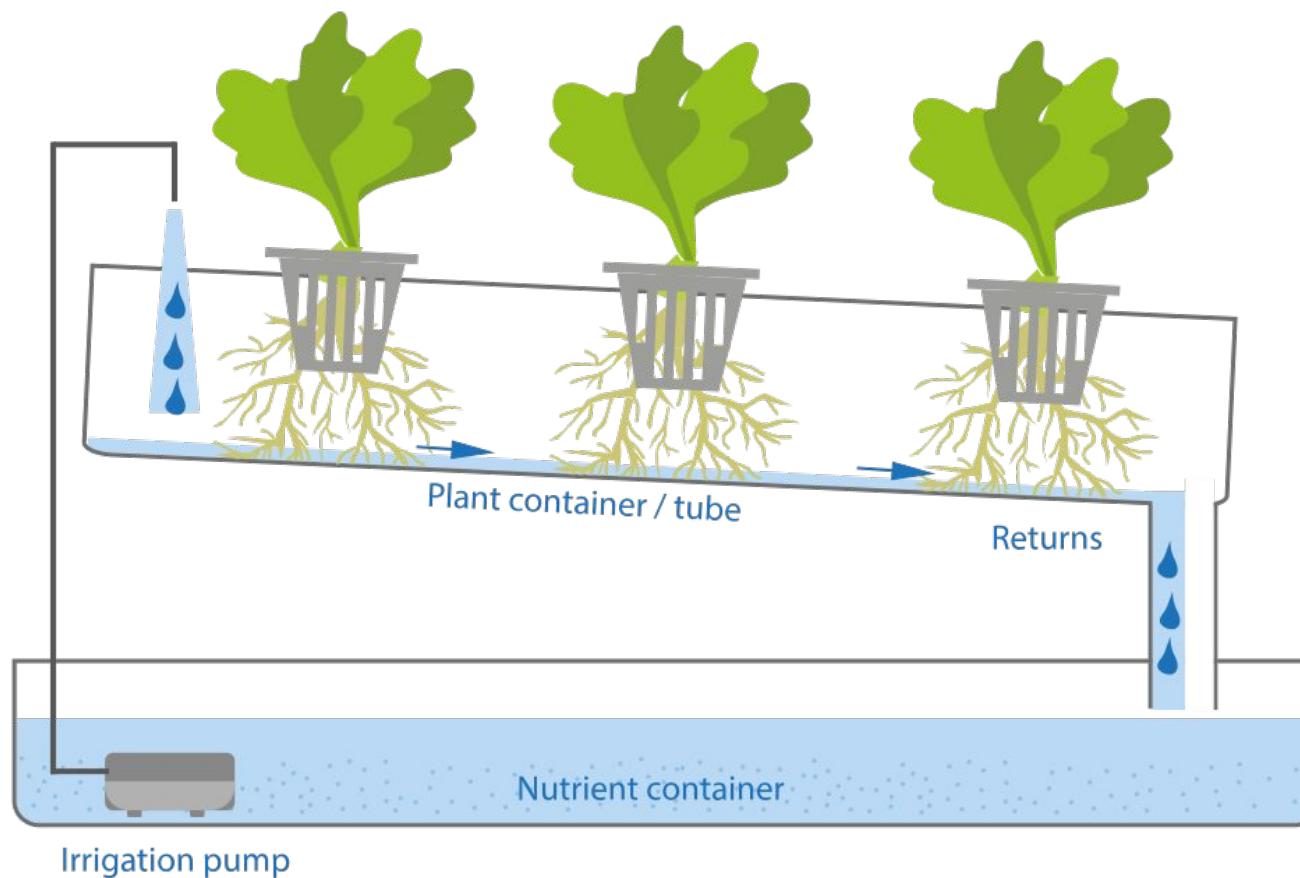
## Sensors





# Irrigation System

- HPA
- Nebulization
- NFT
- Ebb & Flow



<https://www.hydroponic-urban-gardening.com/hydroponics-guide/various-hydroponics-systems>



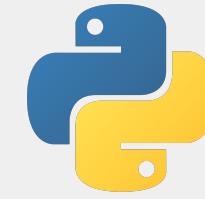
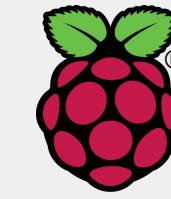
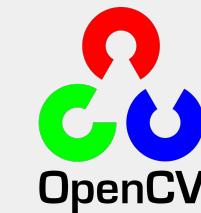
# System Architecture



Sensors



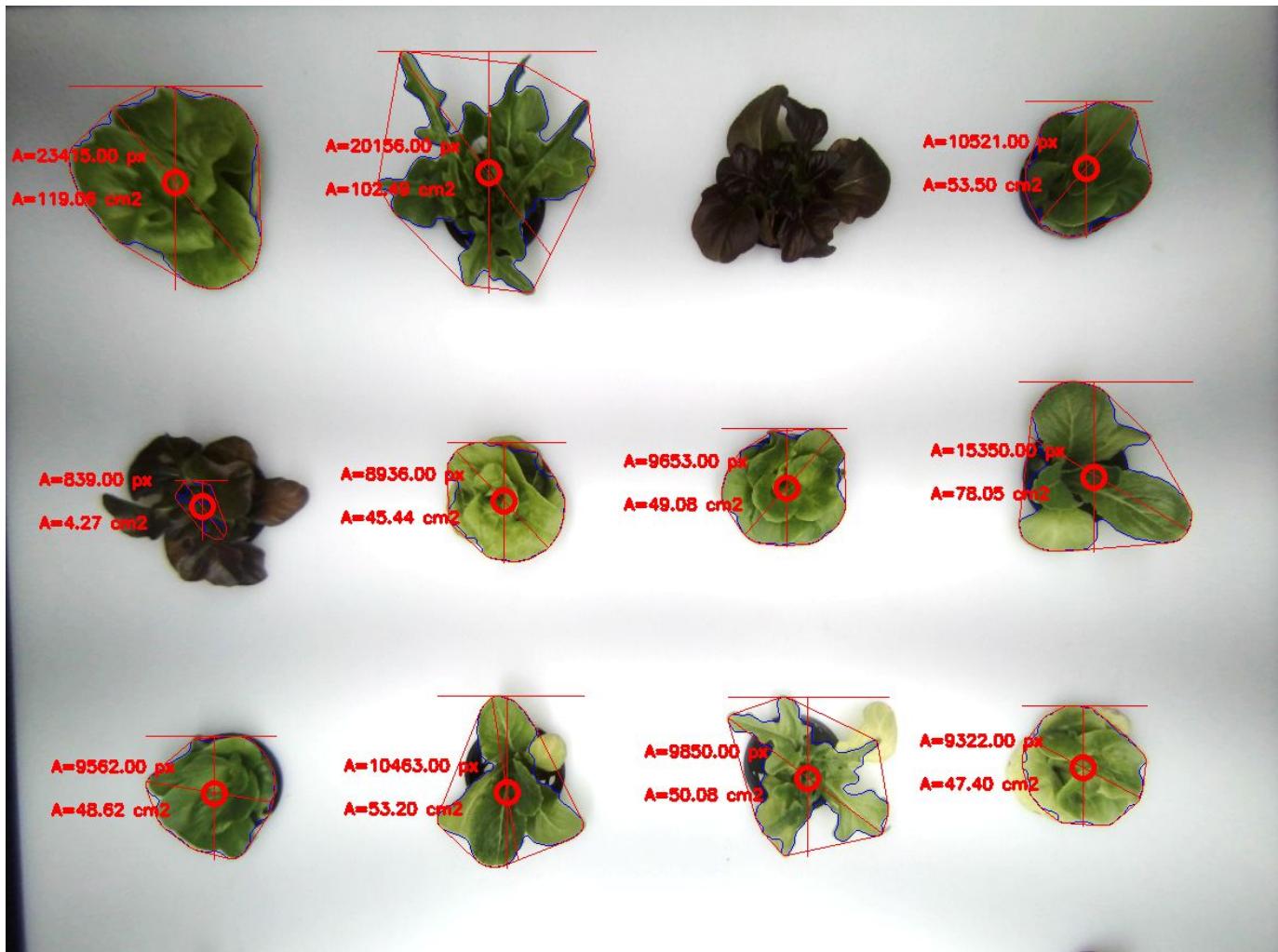
Motherboard



TensorFlowLite

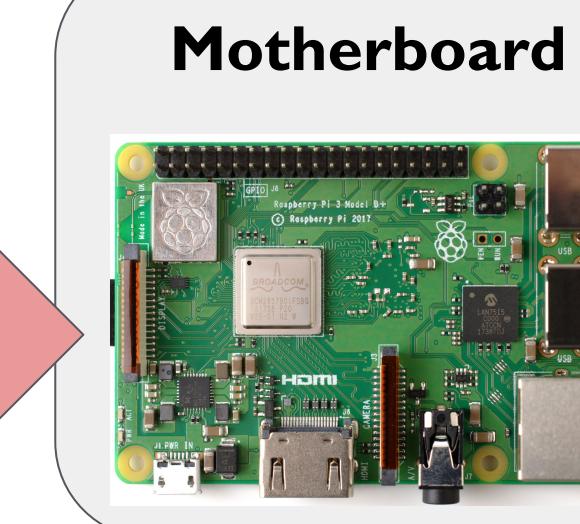


# Computer Vision

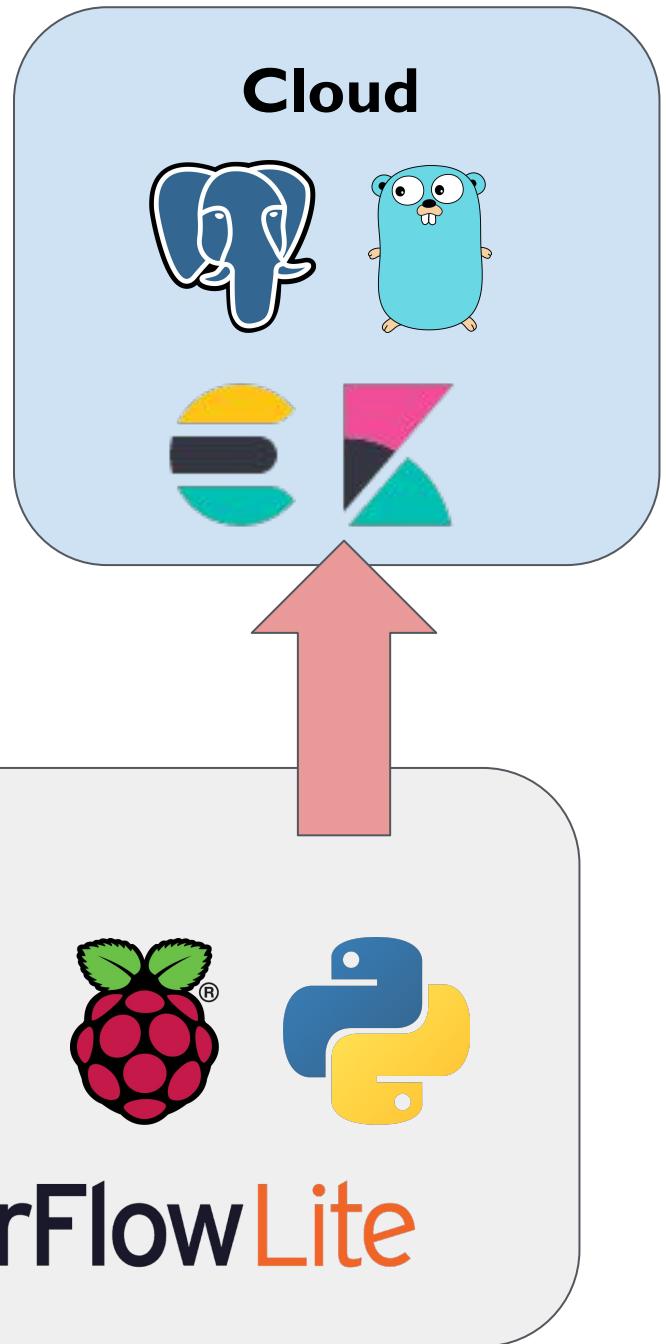
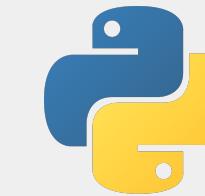
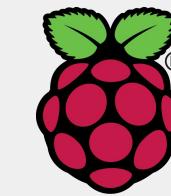




# System Architecture



Motherboard



## 5 - Unit 5 - 2 shelves Opened - EBB Flow

Activity Sensors Edition

C

### Unit sensors

- co2** co2\_atlas\_02
- Dissolved Oxygen do\_atlas\_pot\_02
- Electrical Conductivity ec\_atlas\_pot\_02
- Oxydo-Reduction Potential orp\_atlas\_pot\_02
- PPFD** par\_sq500\_01
- PPFD** par\_sq500\_02
- pH** ph\_atlas\_02
- Relative Humidity hum\_am2315\_01
- Relative Humidity hum\_am2315\_02
- Air Temp. temp\_am2315\_01
- Air Temp. temp\_am2315\_02
- Water Temp. temp\_ds18b20\_01

### Action types

- irrigation
- light
- nutrition

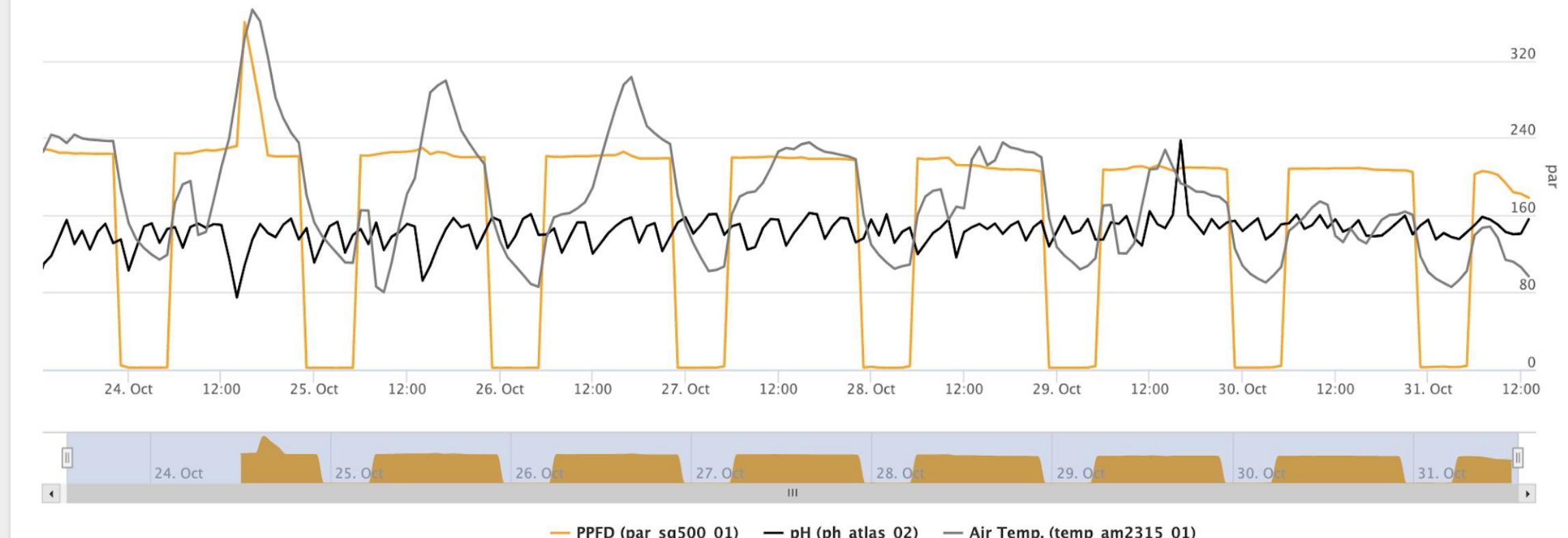
Start date  
10/24/2019, 2:56 PM

End date  
10/31/2019, 2:56 PM

Displayed: average value per hour

Zoom hour day week All

From Oct 23, 2019 To Oct 31, 2019



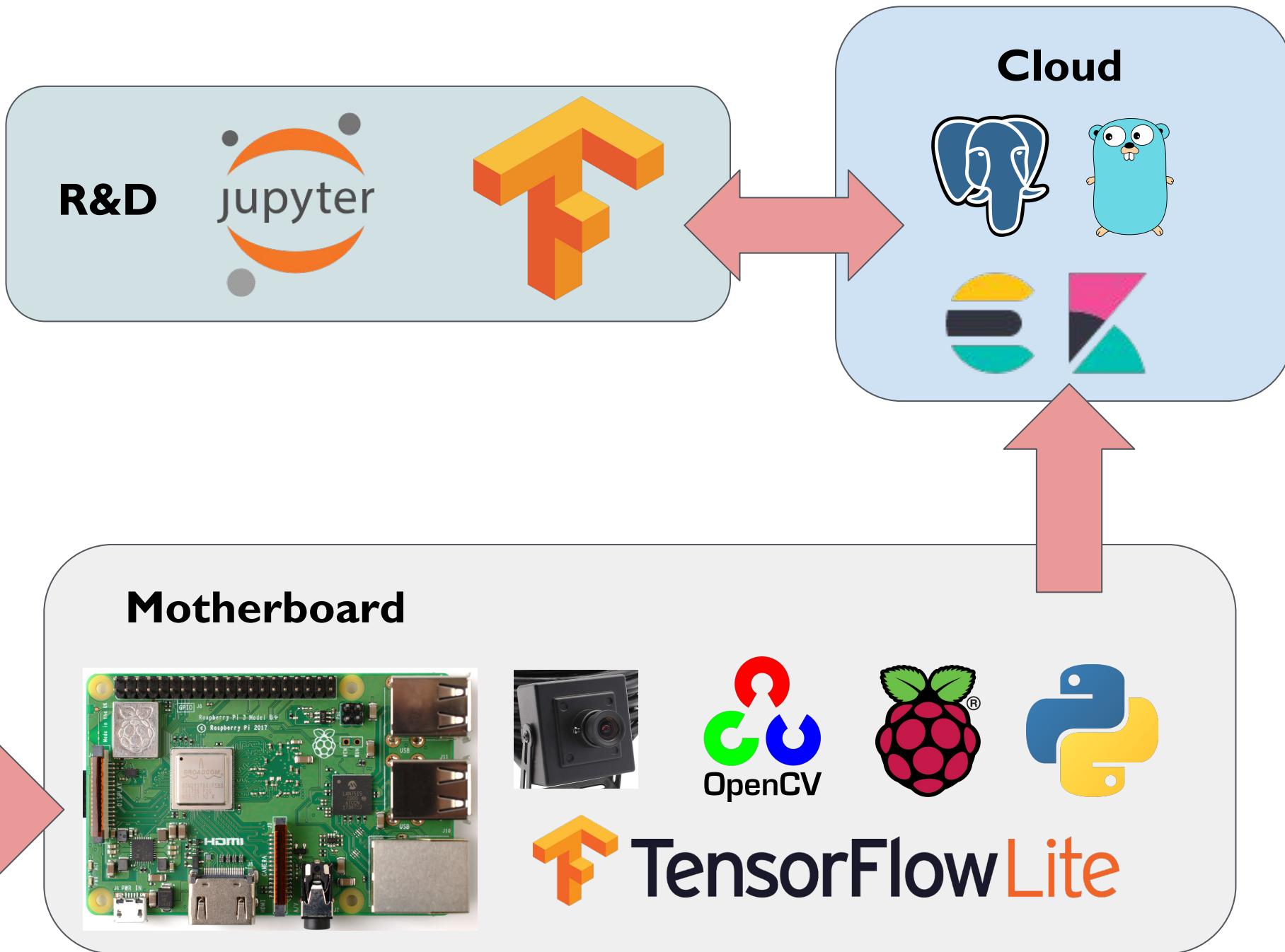
— PPFD (par\_sq500\_01) — pH (ph\_atlas\_02) — Air Temp. (temp\_am2315\_01)



# System Architecture



Sensors



TM

# Why Machine Learning?

Is the system working as expected?

When my lettuce will be ready to be eaten?

What should I do to make vegetables looks and tastes better?

What should I do to make it grow faster?

MIT researchers hacked agriculture to create what may be the tastiest basil on earth

By Martin Finucane Globe Staff, April 3, 2019, 2:18 p.m.



[\*Flavor-cyber-agriculture: Optimization of plant metabolites in an open-source control environment through surrogate modeling\*](#)

Johnson AJ, et al. (2019) *Flavor-cyber-agriculture: Optimization of plant metabolites in an open-source control environment through surrogate modeling*. PLOS ONE 14(4): e0213918.

TM



# Why Edge Computing?

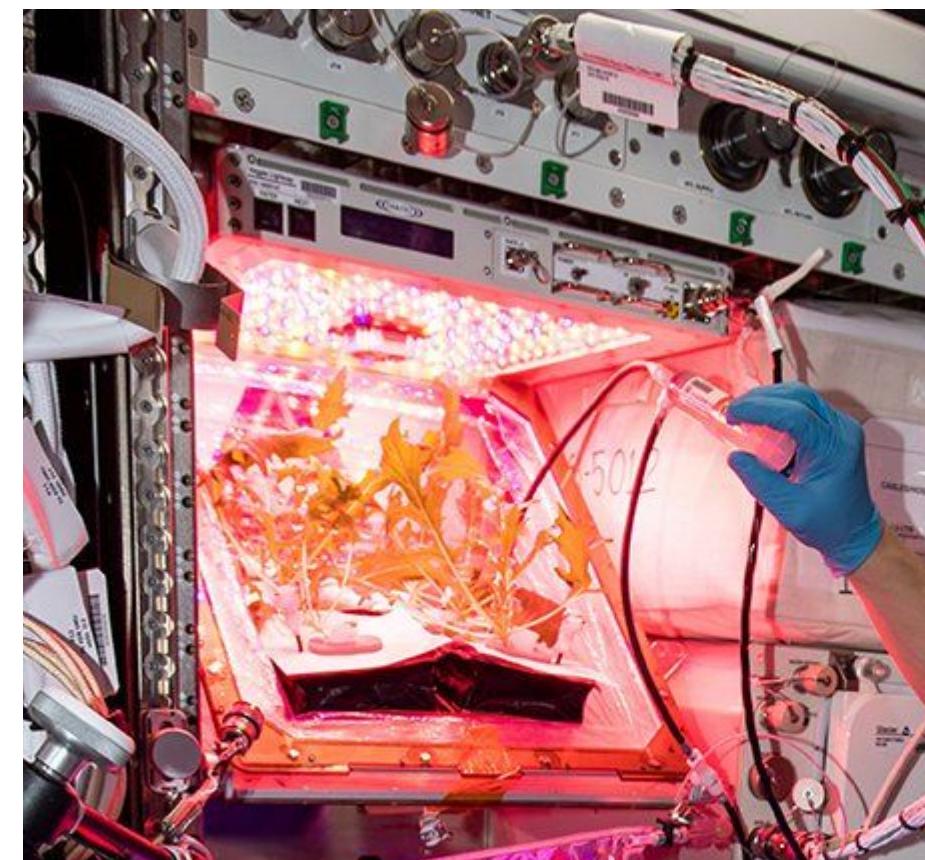
Infrastructure and cloud cost

Scalability

Must work in the field, without internet

Network Latency

Privacy





# Tensorflow



# TensorFlow

Open source library created by Google

Platform for Machine Learning

2.0 (released on October 2019)

**Create, train, debug and use various machine learning model** (neural network but not only!)

Keras, Lite, Tensorboard, Tensorflow Probability

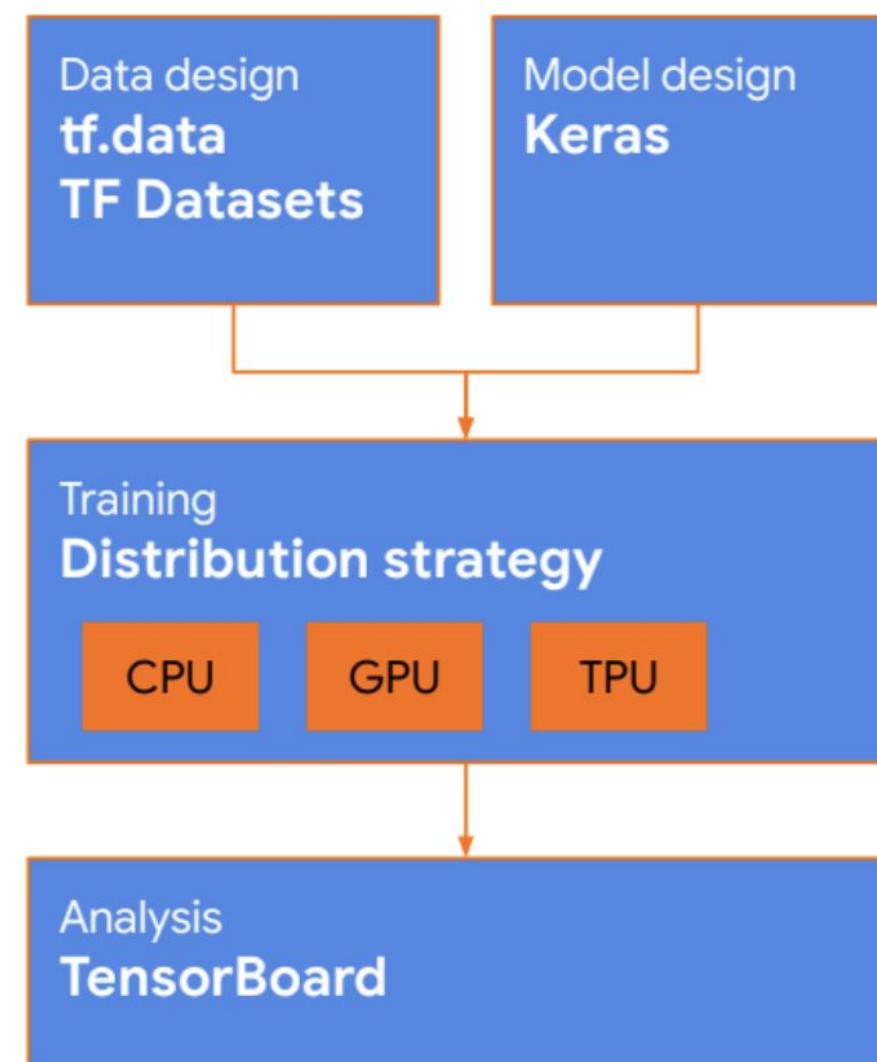


# Tensorflow



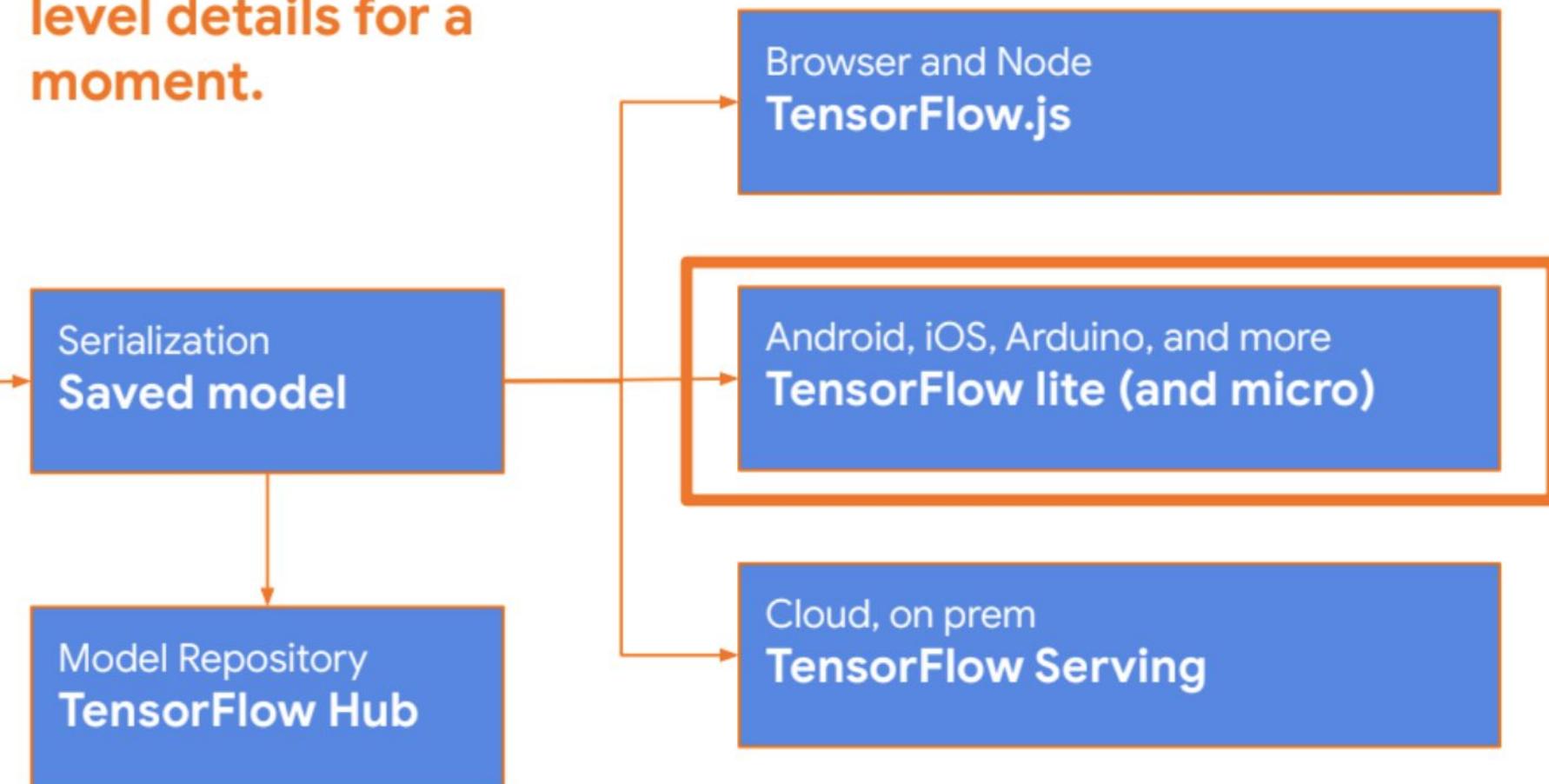
# TensorFlow

## Training



Let's look at low level details for a moment.

## Deployment





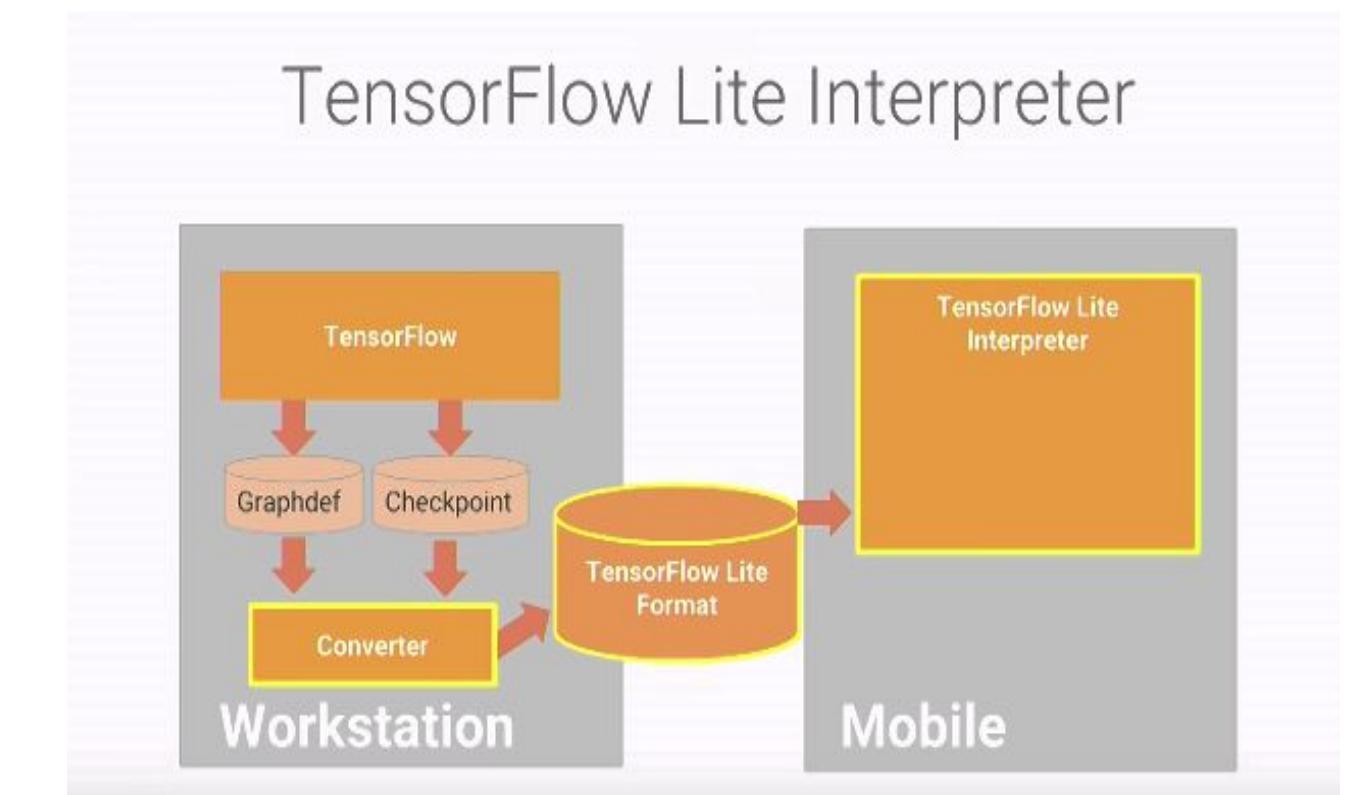
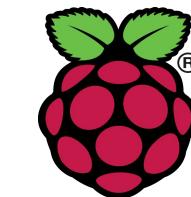
# Tensorflow Lite (TF-Lite)



**Converter + Interpreter**

**Tensorflow vs TF-Lite ?**

- smaller model size
  - faster inference
  - mobile, embedded, MCU
- but**
- no training
  - model is frozen => no re-training
  - no transfer learning





# Tensorflow Lite (TF-Lite)

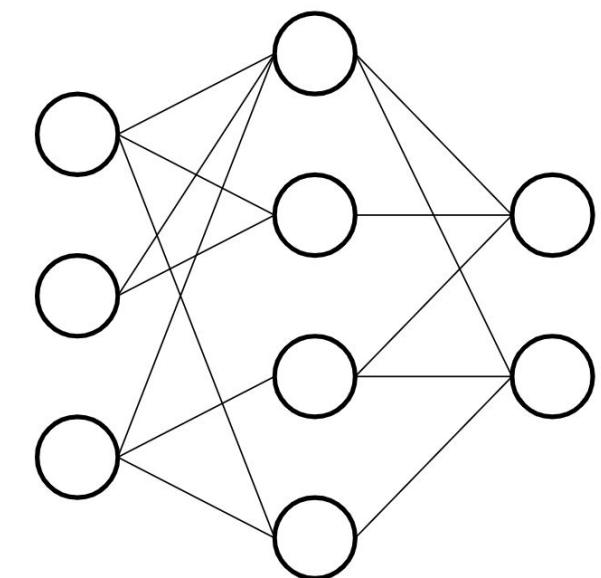
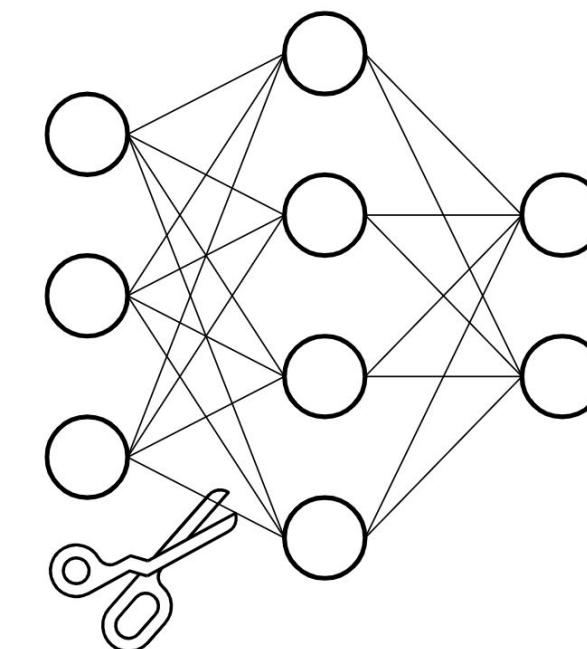


## Optimization

- Pruning
- Post Training Quantization

## Delegate to offload execution

- GPU, TPU, DSP





# ML Workflow with Tensorflow Lite



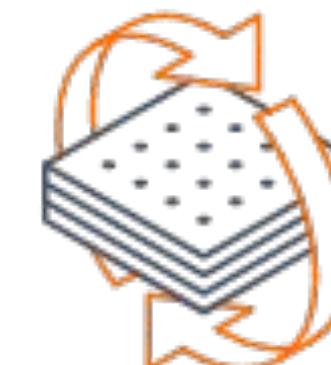
**Import your dataset**

**Work on data:** preprocessing, normalization, features selection

**Build your model with Tensorflow**

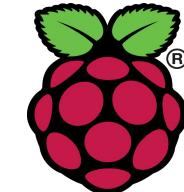
**Train your model**

**Export and convert to .tflite**





# ML Workflow with Tensorflow Lite



## TensorFlow Lite

**Load** your model (*or grab one in [github.com/tensorflow/models](https://github.com/tensorflow/models)*)

**Preprocess** input data

**Allocate Memory**

**Run** inference

**Interpret** output

# Setup on your Laptop

```
RitoneIoT@devoxx Using cached https://files.pythonhosted.org/packages/0e/3e/377007e3f36ec42f1b84ec322ee12141a9e10d808312e5738f52f80a232c/pexpect-4.7.0-py2.py3-none-any.whl Processing ./Library/Caches/pip/wheels/98/b0/dd/29e28ff615af3dda4c67cab719dd51357597eabff926976b45/backcall-0.1.0-cp37-none-any.whl Collecting pickleshare Using cached https://files.pythonhosted.org/packages/9a/41/220f49aaea88bc6fa6cba8d05ecf24676326156c23b991e80b3f2fc24c77/pickleshare-0.7.5-py2.py3-none-any.whl Collecting jedi>=0.10 Using cached https://files.pythonhosted.org/packages/55/54/da994f359e4e7da4776a200e76dbc85ba5fc319eefc22e33d55296d95a1d/jedi-0.15.1-py2.py3-none-any.whl Collecting wcwidth Using cached https://files.pythonhosted.org/packages/7e/9f/526a6947247599b084ee5232e4f9190a38f398d7300d866af3ab571a5bfe/wcwidth-0.1.7-py2.py3-none-any.whl Collecting webencodings Using cached https://files.pythonhosted.org/packages/f4/24/2a3e3df732393fed8b3ebf2ec078f05546de641fe1b667ee316ec1dcf3b7/webencodings-0.5.1-py2.py3-none-any.whl Collecting attrs>=17.4.0 Using cached https://files.pythonhosted.org/packages/a2/db/4313ab3be961f7a763066401fb77f7748373b6094076ae2bda2806988af6/attrs-19.3.0-py2.py3-none-any.whl Collecting importlib-metadata Using cached https://files.pythonhosted.org/packages/f6/d2/40b3fa882147719744e6aa50ac39cf7a22a913cbcba86a0371176c425a3b/importlib_metadata-0.23-py2.py3-none-any.whl Processing ./Library/Caches/pip/wheels/bb/46/00/6d471ef0b813e3621f0abe6cb723c20d529d39a061de3f7c51/pyrsistent-0.15.4-cp37m-macosx_10_14_x86_64.whl
```



# Setup on your Laptop

```
$ python3 --version  
$ pip3 --version  
$ virtualenv --version  
  
$ virtualenv --system-site-packages -p python3 ./venv  
$ source ./venv/bin/activate  
$ pip install --upgrade pip  
$ pip install --upgrade tensorflow=2.0  
$ pip install numpy pandas jupyter jupyterlab notebook  
matplotlib
```



# Setup on your RPI

## Tensorflow Lite Interpreter

1. Using pip and official TF release (not always up to date)
2. Cross compile Tensorflow for ARMv7 on your laptop
3. Build Bazel and Tensorflow on your RPI (> 24h)
4. **Using pip and a *community built* .whl package**



<https://github.com/PINTO0309/Tensorflowlite-bin>

TM

# Setup on your RPI

```
RtonelloT@devoxx:~$ curl -O https://raw.githubusercontent.com/tensorflow/tensorflow/master/tensorflow/lite/python/tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl  
Connecting to raw.githubusercontent.com (raw.githubusercontent.com)|151.101.120.133|:443... connected.  
HTTP request sent, awaiting response... 200 OK  
Length: 1493151 (1.4M) [application/octet-stream]  
Saving to: 'tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl'  
  
tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl          0%[  
          ] 0 --  
tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl          100%[=====>] 1.42M 8.  
=====] 65MB/s in 0.2s  
  
2019-10-26 16:57:00 (8.65 MB/s) - 'tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl' saved [1493151/1493151]  
  
(pi-devoxx-tflite) pi@raspberrypi:~ $ pip install --upgrade tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl  
Looking in indexes: https://pypi.org/simple, https://www.piwheels.org/simple  
Processing ./tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl  
Requirement already satisfied, skipping upgrade: numpy>=1.12.1 in ./pi-devoxx-tflite/lib/python3.7/site-packages (from tflite  
-runtime==2.0.0) (1.17.3)  
Installing collected packages: tflite-runtime  
Successfully installed tflite-runtime-2.0.0  
(pi-devoxx-tflite) pi@raspberrypi:~ $ python  
Python 3.7.3 (default, Apr 3 2019, 05:39:12)  
[GCC 8.2.0] on linux  
Type "help", "copyright", "credits" or "license" for more information.  
>>> import tflite_runtime
```



# Setup on your RPI

## Tensorflow Lite Interpreter

```
$ sudo apt install swig libjpeg-dev zlib1g-dev python3-dev  
python3-numpy unzip  
$ wget  
https://github.com/PINT00309/TensorflowLite-bin/raw/master/2  
.0.0/tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl  
$ pip install --upgrade  
tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl
```



# Setup on your RPI

## Full Tensorflow package

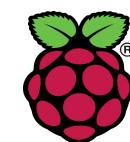
```
sudo apt-get install -y libhdf5-dev libc-ares-dev libeigen3-dev  
pip install keras_applications==1.0.8 --no-deps  
pip install keras_preprocessing==1.1.0 --no-deps  
pip install h5py==2.9.0  
sudo apt-get install -y openmpi-bin libopenmpi-dev  
sudo apt-get install -y libatlas-base-dev  
pip install six wheel mock  
wget  
https://github.com/PINT00309/Tensorflow-bin/raw/master/tensorflow-2.0.0-cp37-cp37m-linux\_armv7l.whl  
pip uninstall tensorflow  
pip install tensorflow-2.0.0-cp37-cp37m-linux_armv7l.whl
```



# Demo



**Part 1** - Build, Train and Convert a simple Neural Network model to predict lettuce weight

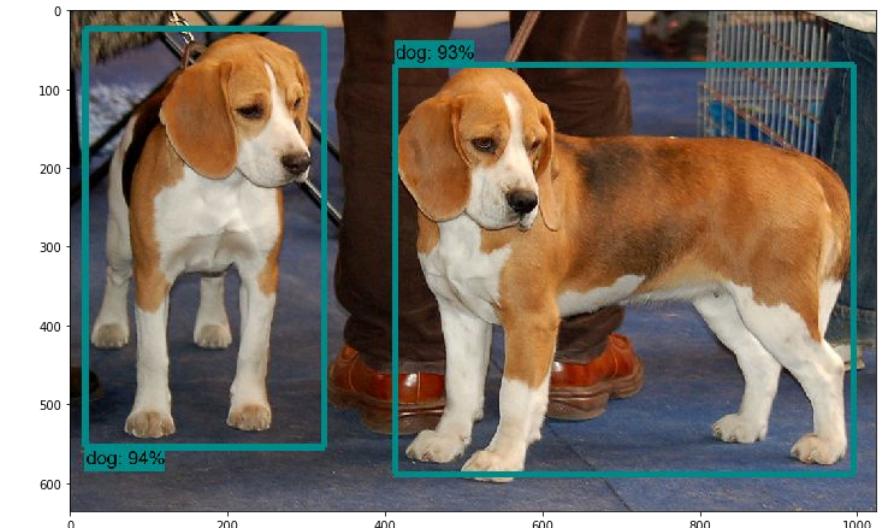
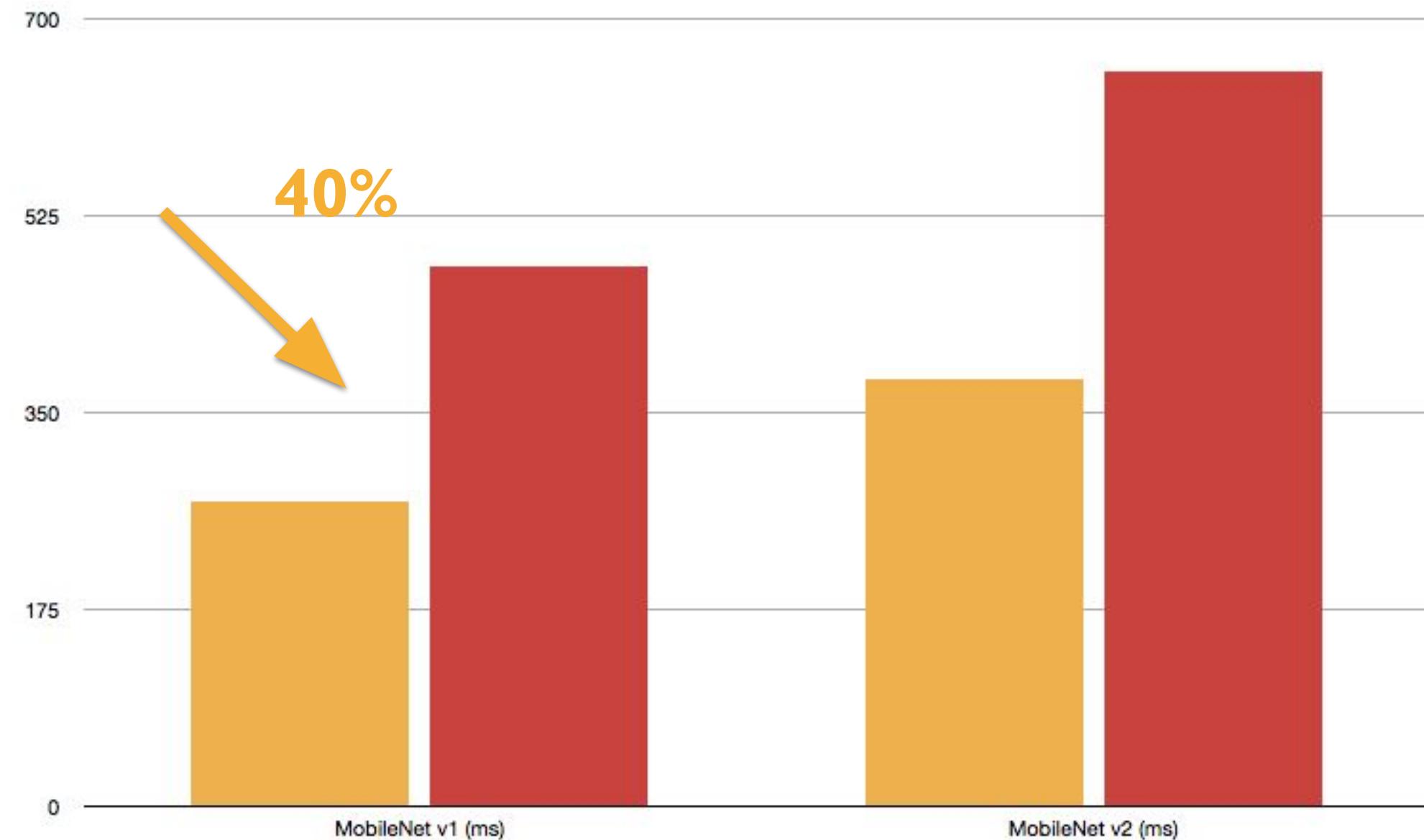


**Part 2** - Deploy your tflite model on RPI and run inference



# Tensorflow Lite Benchmark

## Inference Time

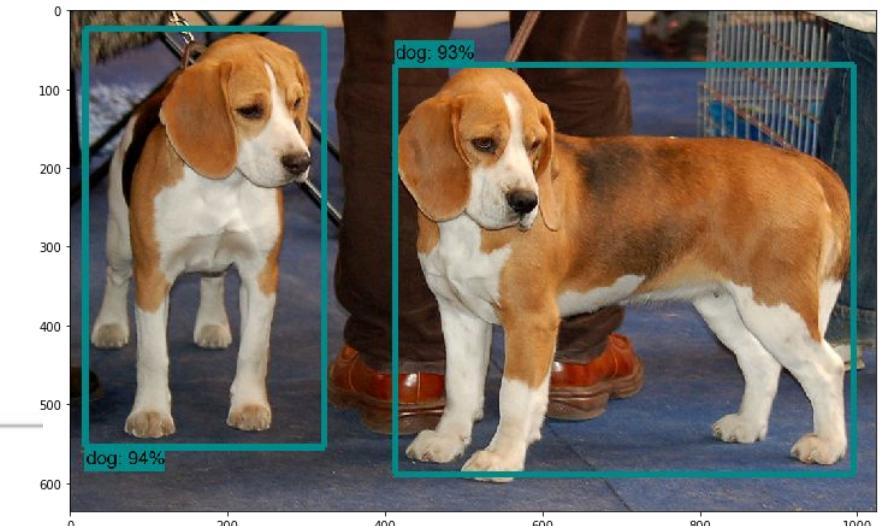
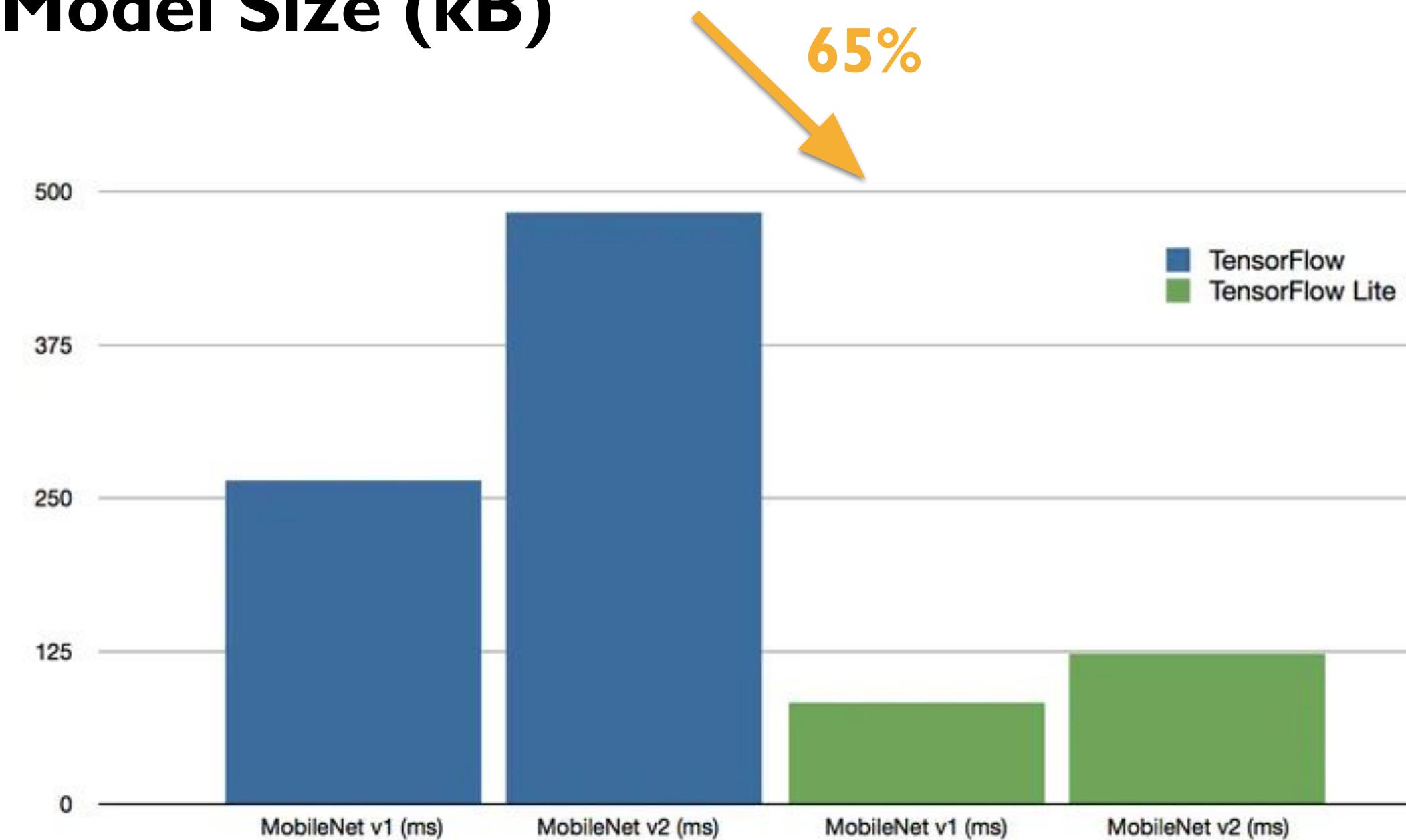


Source: Alasdair Allan



# Tensorflow Lite Benchmark

## Model Size (kB)



Source: Alasdair Allan



# Tensorflow Lite Limitations

Reinforcement Learning

Transfer Learning

Recurrent Neural Network (RNN) like LSTM

Operation Compatibility

[https://www.tensorflow.org/lite/guide/ops\\_compatibility](https://www.tensorflow.org/lite/guide/ops_compatibility)

<https://github.com/tensorflow/tensorflow/blob/master/tensorflow/lite/experimental/examples/lstm/g3doc/README.md>

TM

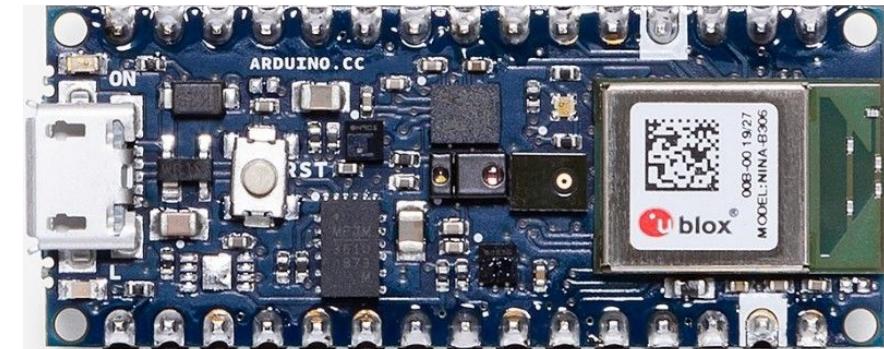


# TFLite on Microcontrollers

**Inference on Cortex-M microcontroller**

Only some operations are supported

Enough for hotword, gesture and speech  
recognition

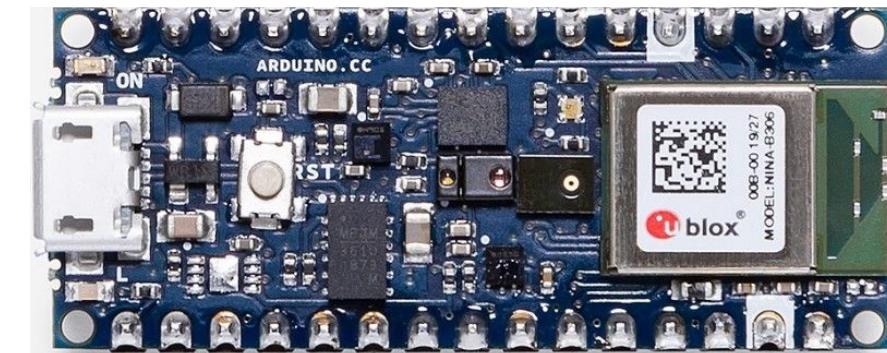


Arduino Nano 33





# TFLite on Microcontrollers



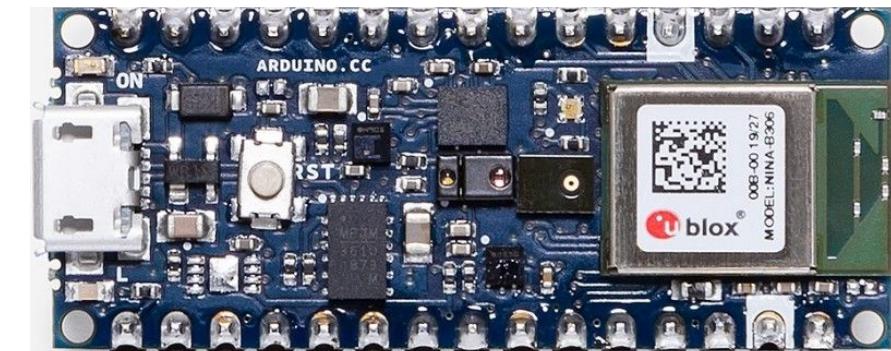
## C++ API

```
#include <TensorFlowLite.h>
// This is your tflite model
#include "lg_weight_model.h"
#include
"tensorflow/lite/experimental/micro/kernels/all_ops_resolver.h"
#include "tensorflow/lite/experimental/micro/micro_interpreter.h"
#include "tensorflow/lite/schema/schema_generated.h"

tflite::ErrorReporter *error_reporter = nullptr;
const tflite::Model *model = nullptr;
tflite::MicroInterpreter *interpreter = nullptr;
TfLiteTensor *input = nullptr;
TfLiteTensor *output = nullptr;
```



# TFLite on Microcontrollers



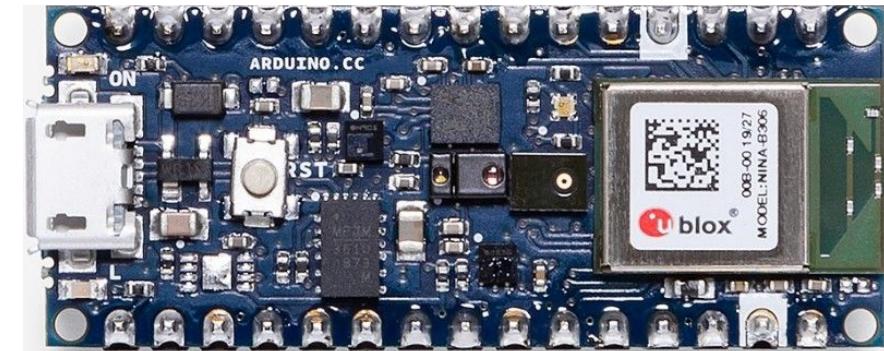
```
// Finding the min value for your model may require tests!
constexpr int kTensorArenaSize = 2 * 1024;
uint8_t tensor_arena[kTensorArenaSize];

// Load your model.
model = tflite::GetModel(g_weight_regression_model_data);
// This pulls in all the operation implementations we need.
static tflite::ops::micro::AllOpsResolver resolver;

// Build an interpreter to run the model with.
static tflite::MicroInterpreter static_interpreter(
    model, resolver, tensor_arena, kTensorArenaSize, error_reporter);
interpreter = &static_interpreter;
```



# TFLite on Microcontrollers

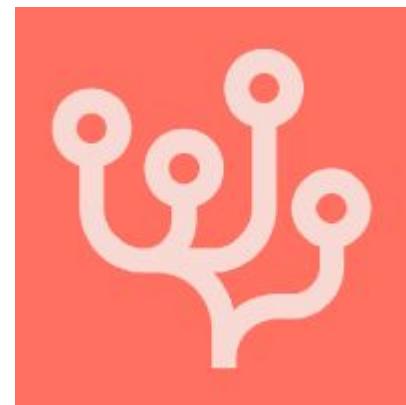
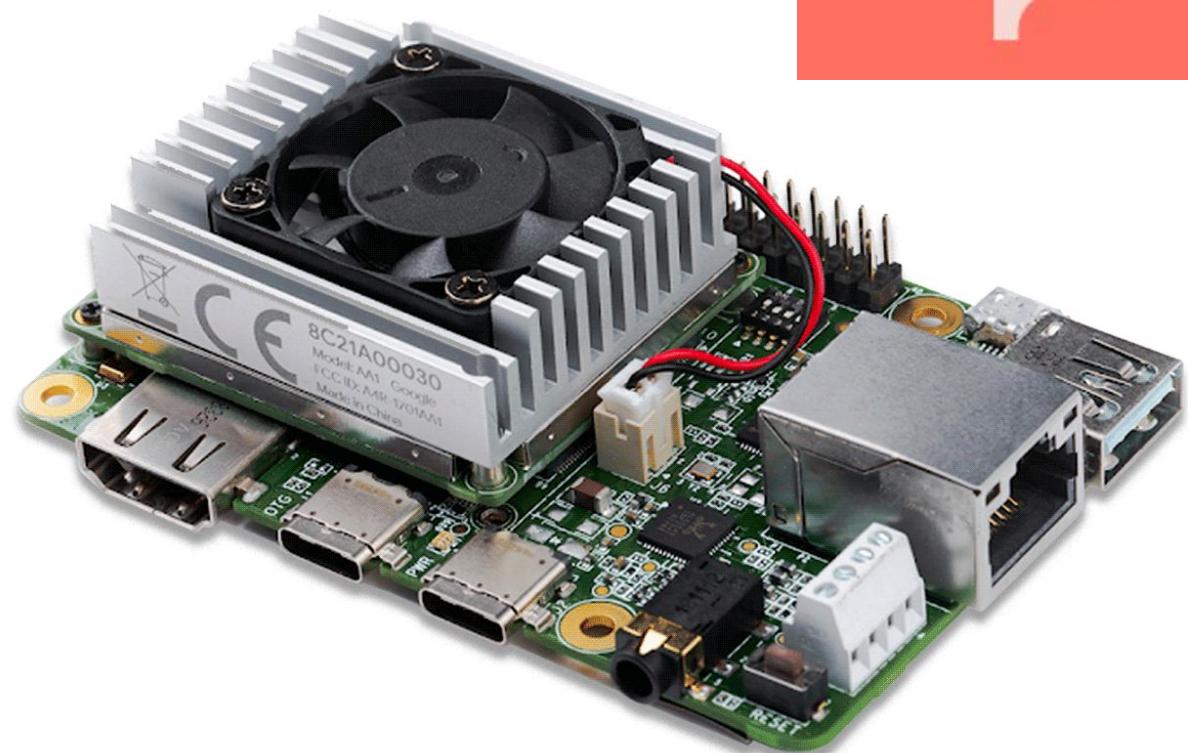


```
// Allocate memory for the model's tensors.  
TfLiteStatus allocate_status = interpreter->AllocateTensors();  
// Obtain pointers to the model's input and output tensors.  
input = interpreter->input(0);  
output = interpreter->output(0);  
// Feed the interpreter with the input value  
float x_val = random(0, 10);  
input->data.f[0] = x_val;  
  
// Run Inference  
TfLiteStatus invoke_status = interpreter->Invoke();  
// Get inference result  
float y_val = output->data.f[0];
```



# Further Work

Training at the Edge  
Transfer Learning  
Federated Learning



# Summary

Build, Train, Optimize, Convert on laptop

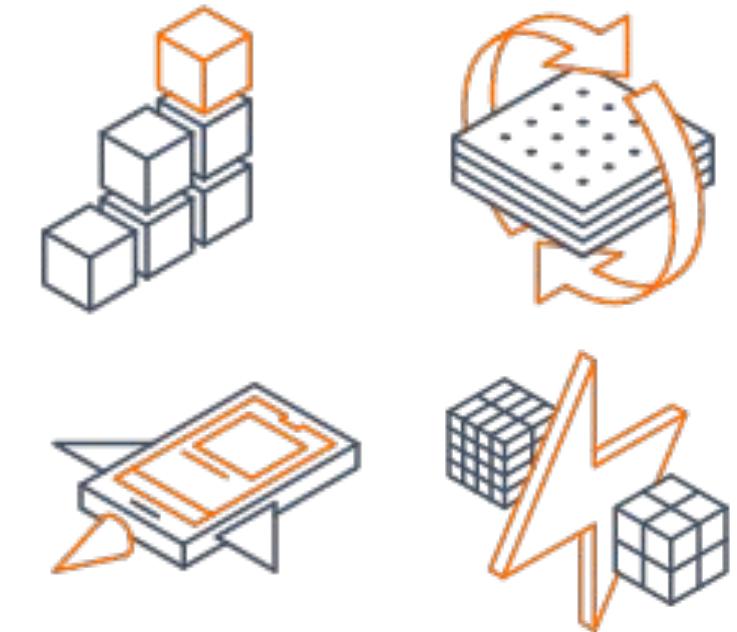
Deploy, Infer on device

Some operations are not supported

Quantization does not affect accuracy

Inference on IoT and microcontrollers is feasible

Regression, anomalies detection, objects recognition, smart reply,  
etc.

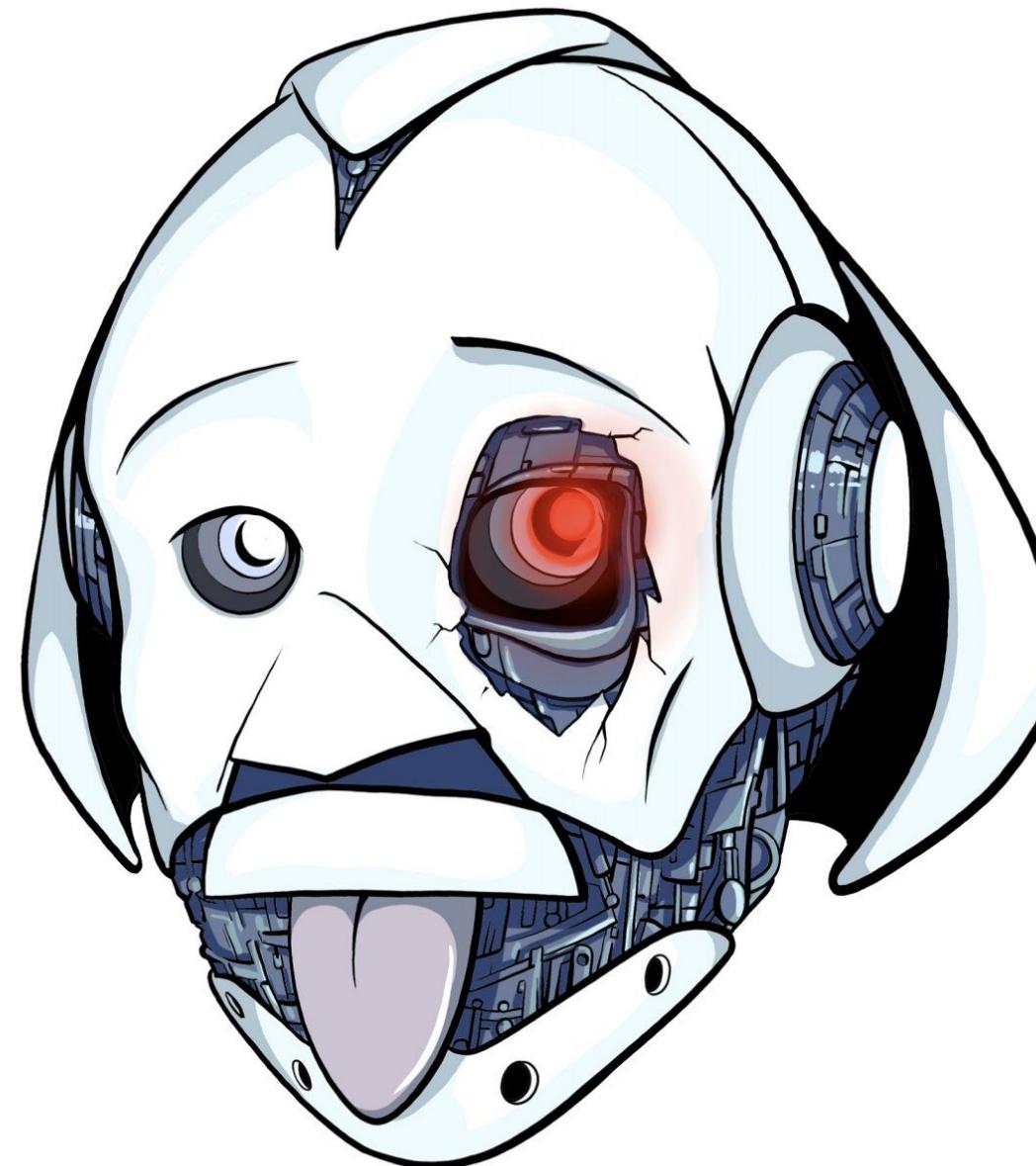




# Thanks!



[frama.link/tflite-devoxx](https://frama.link/tflite-devoxx)



[frama.link/rtone-jobs](https://frama.link/rtone-jobs)



RTONE  
IOT MAKERS



# References

- <https://medium.com/tensorflow/how-to-get-started-with-machine-learning-on-arduino-7daf95b4157>
- <https://www.tensorflow.org/lite>
- <https://www.tensorflow.org>
- <https://coral.withgoogle.com/>
- <https://arxiv.org/abs/1902.01046>
- <https://medium.com/tensorflow/tensorflow-model-optimization-toolkit-pruning-api-42cac9157a6a>