



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

@alexis0duque



DEVOXX™
MOROCCO





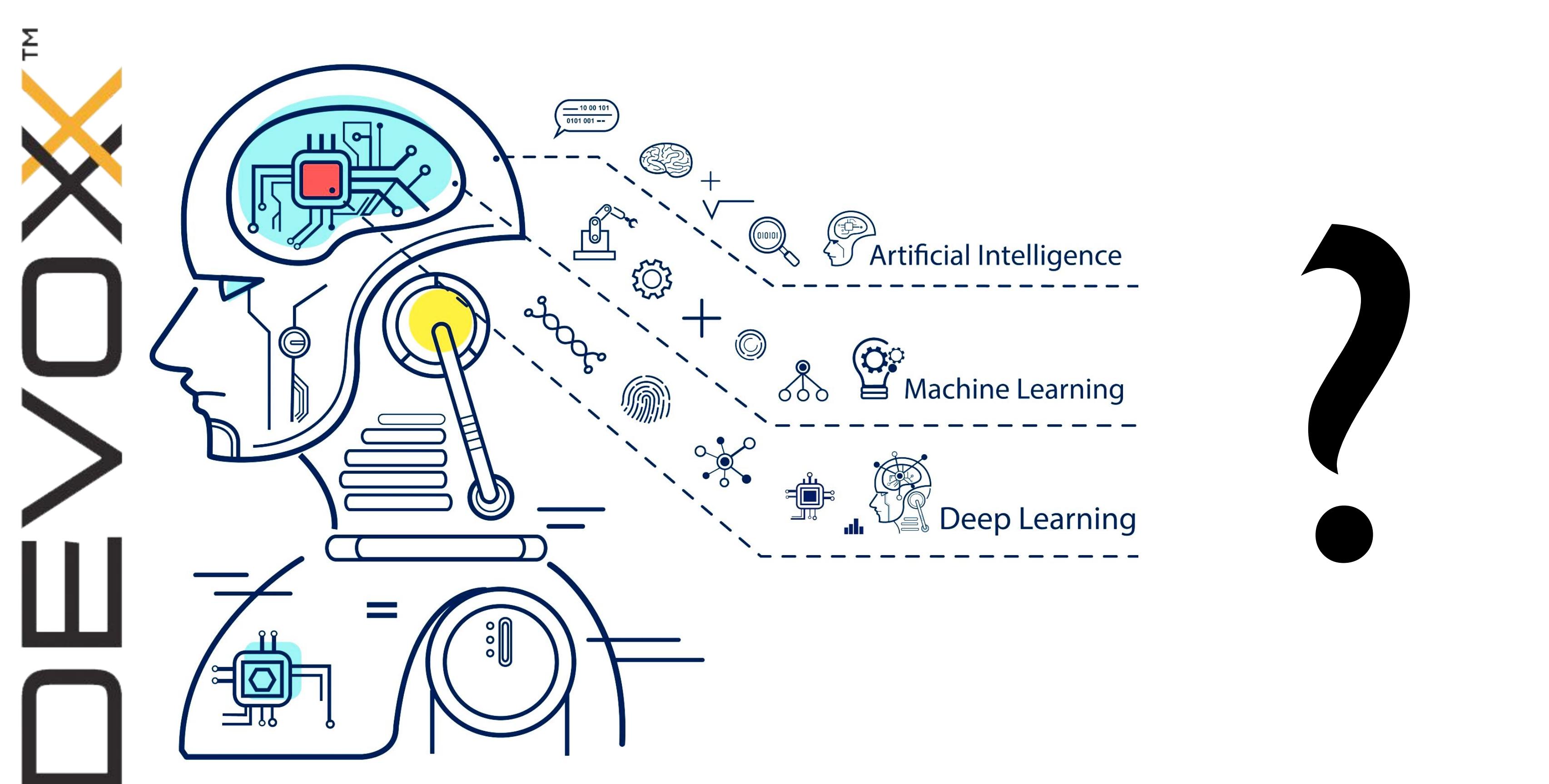
Who am I?

Alexis DUQUE

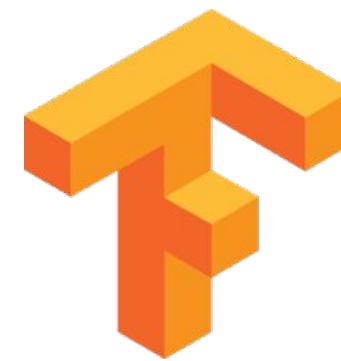
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-  alexisduque.me
-  <https://goo.gl/oNUWu6>





TM
Devoxx
MA



TensorFlow Lite

?

TM
Devoxx
DE



?

#DevoxxMA #IoT #Tensorflow #AI

@alexis0duque



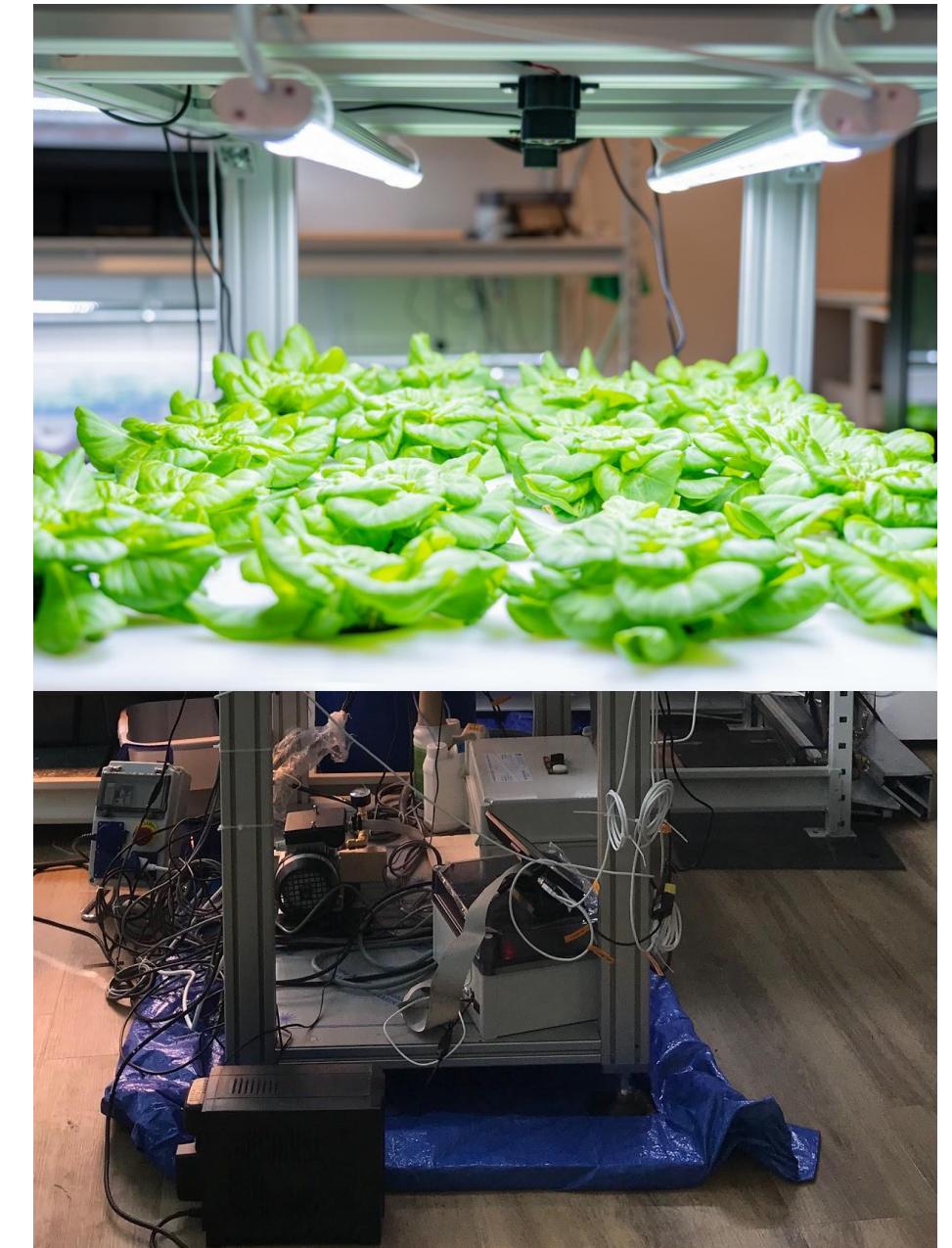
Outline & What You Will Learn

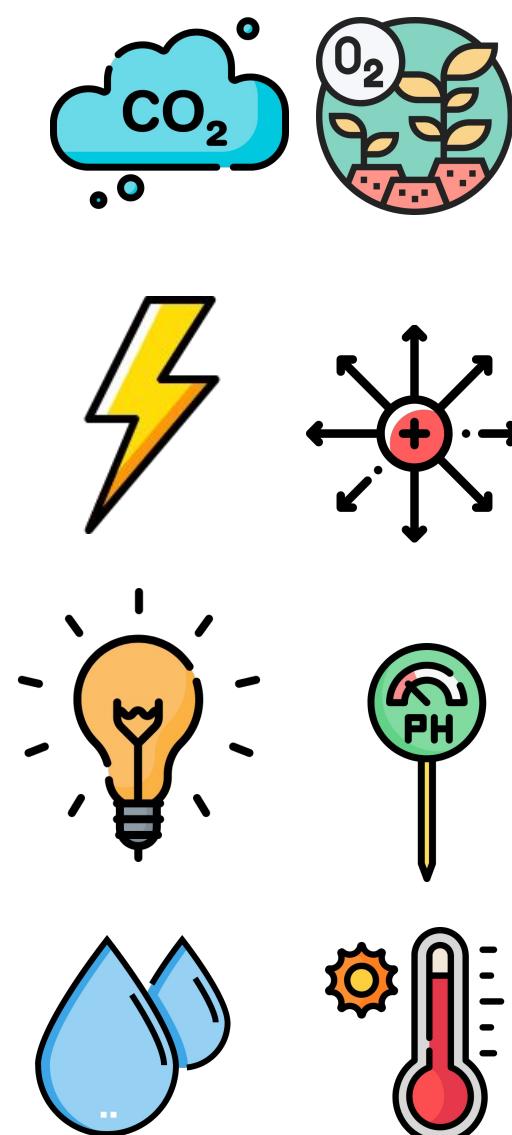
- **Indoor Vertical Farming**
- **Why Intelligence at the Edge?**
- Introduction to **Tensorflow Lite**. How to Use It?
- **Setup** your laptop & RPI
- **Build** and **train** a small model to **predict lettuce weight**
- **Convert** and deploy it on a RPI
- Run **predictions** on IoT devices
- Benchmarks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

TM
X
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D
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Indoor Vertical Farming

LA
GRANGETTE





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 -)

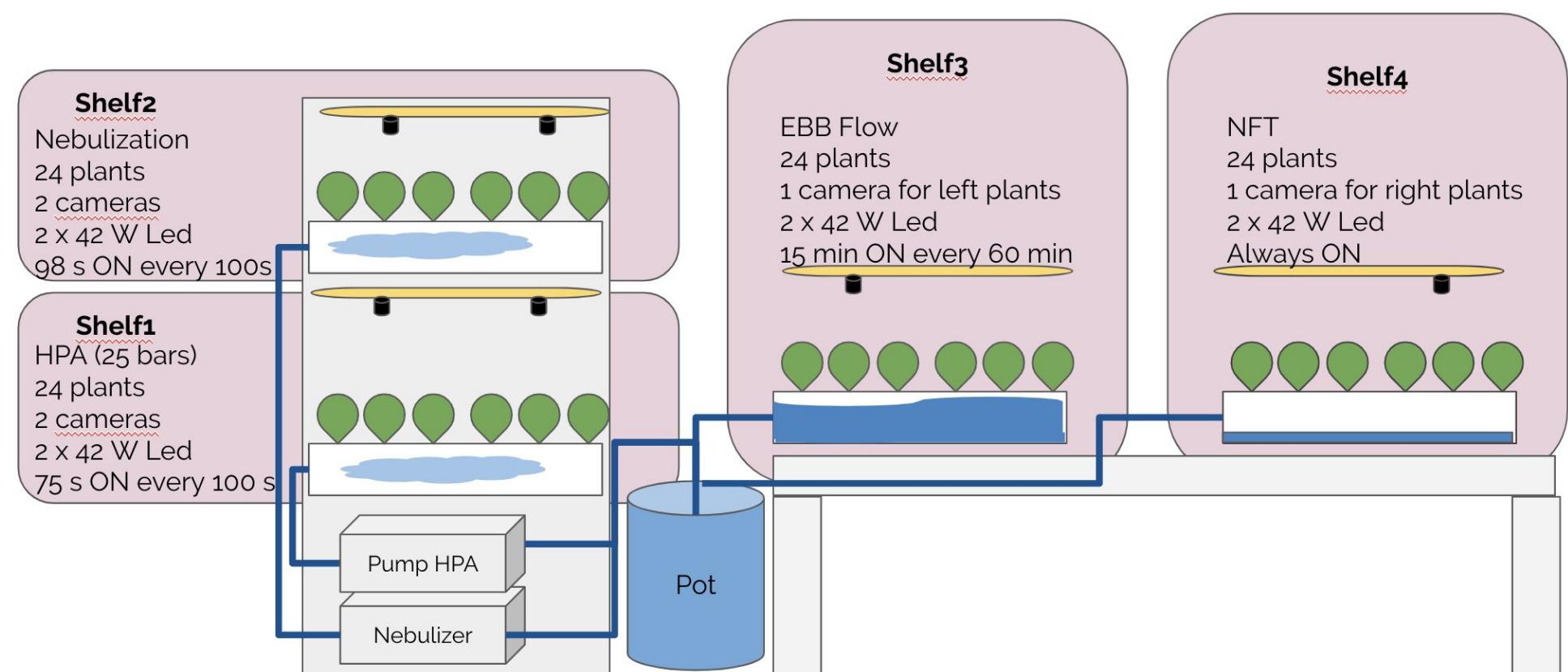
average **humidity** (double, in %)

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



System Architecture

Actuators and Irrigation System



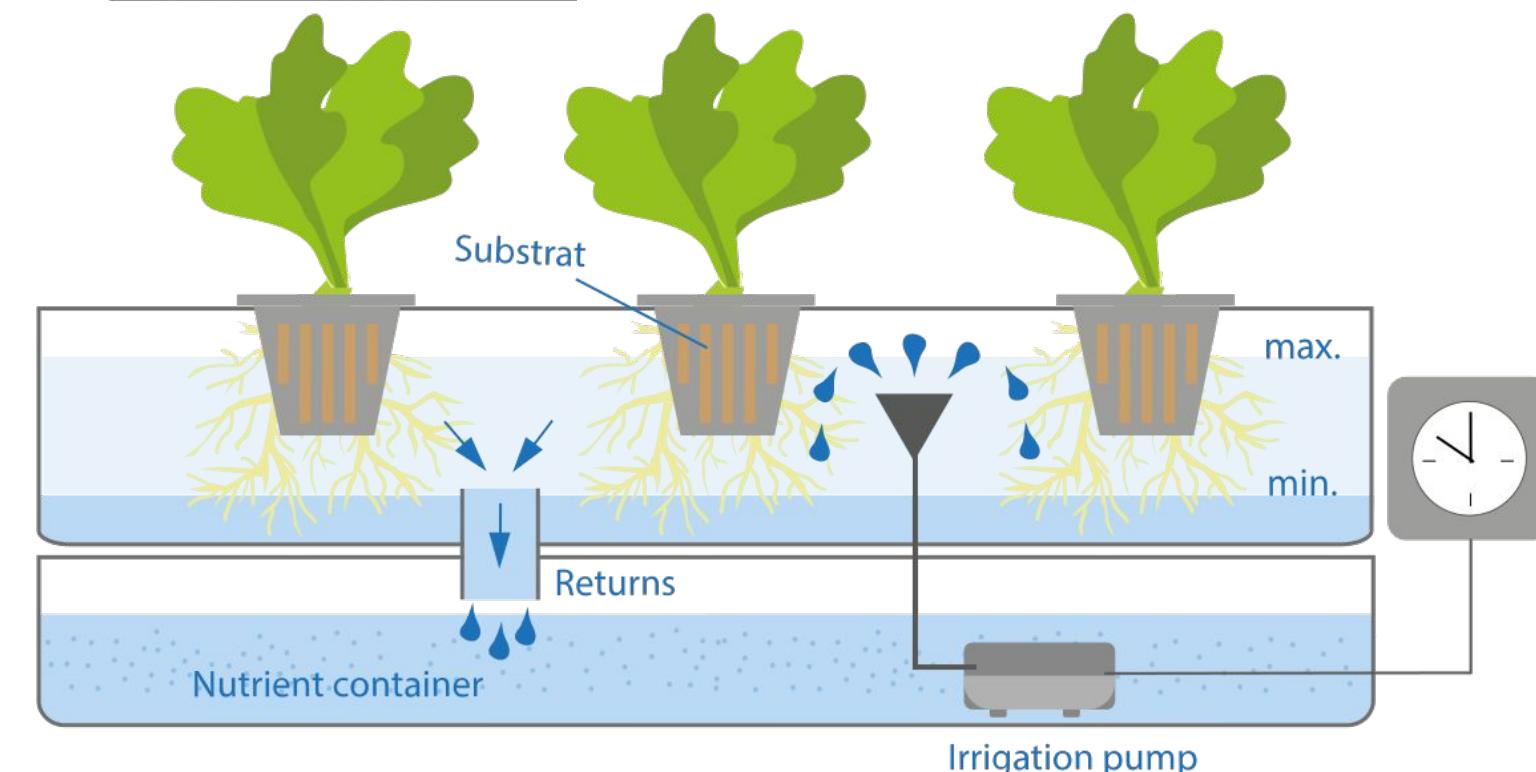
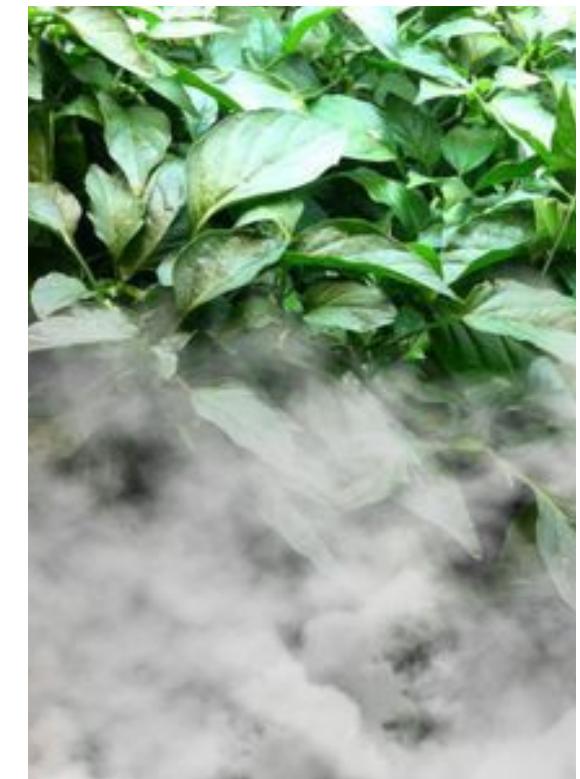
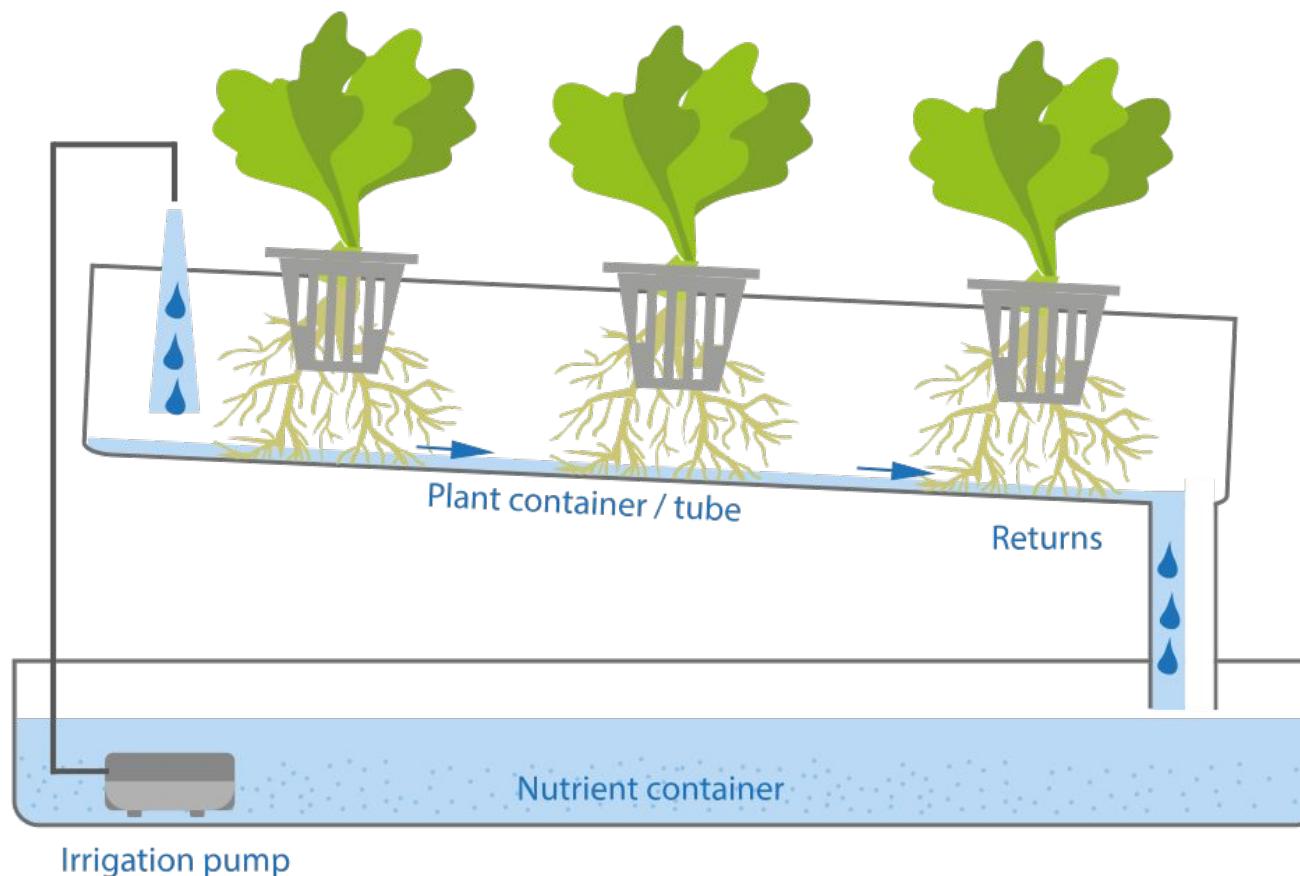
Sensors





Irrigation System

- HPA
- Nebulization
- NFT
- Ebb & Flow



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



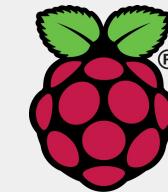
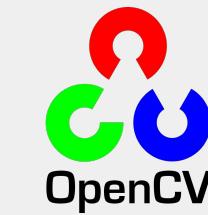
System Architecture



Sensors



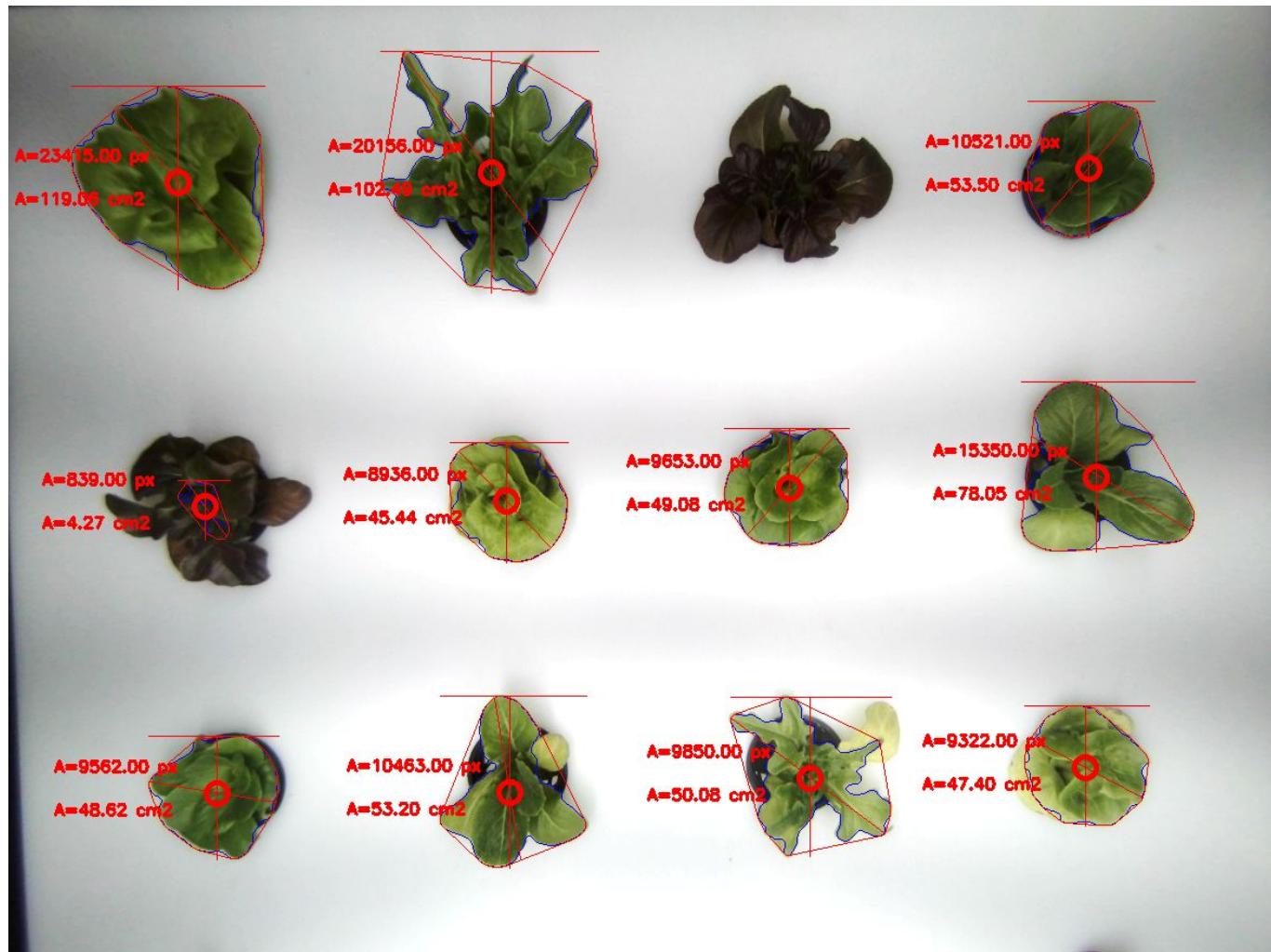
Motherboard



TensorFlowLite



Computer Vision

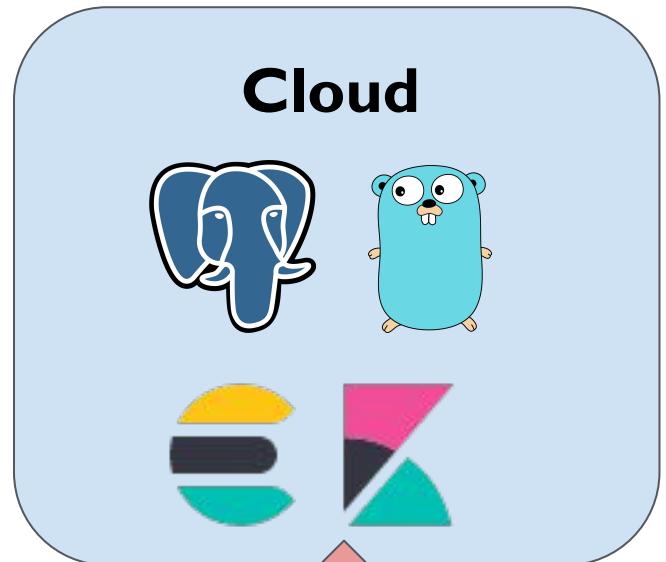
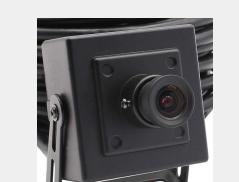




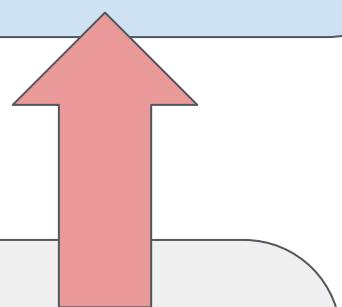
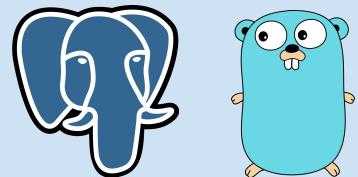
System Architecture



Motherboard



Cloud



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

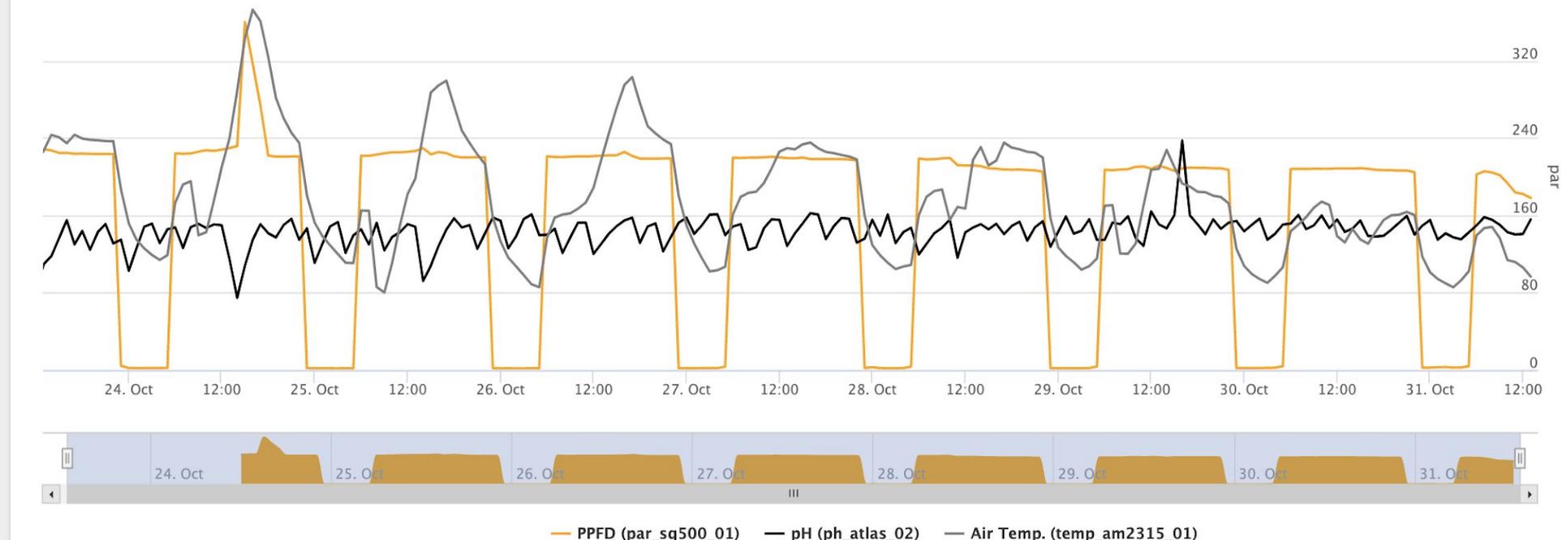
Start date
10/24/2019, 2:56 PM

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



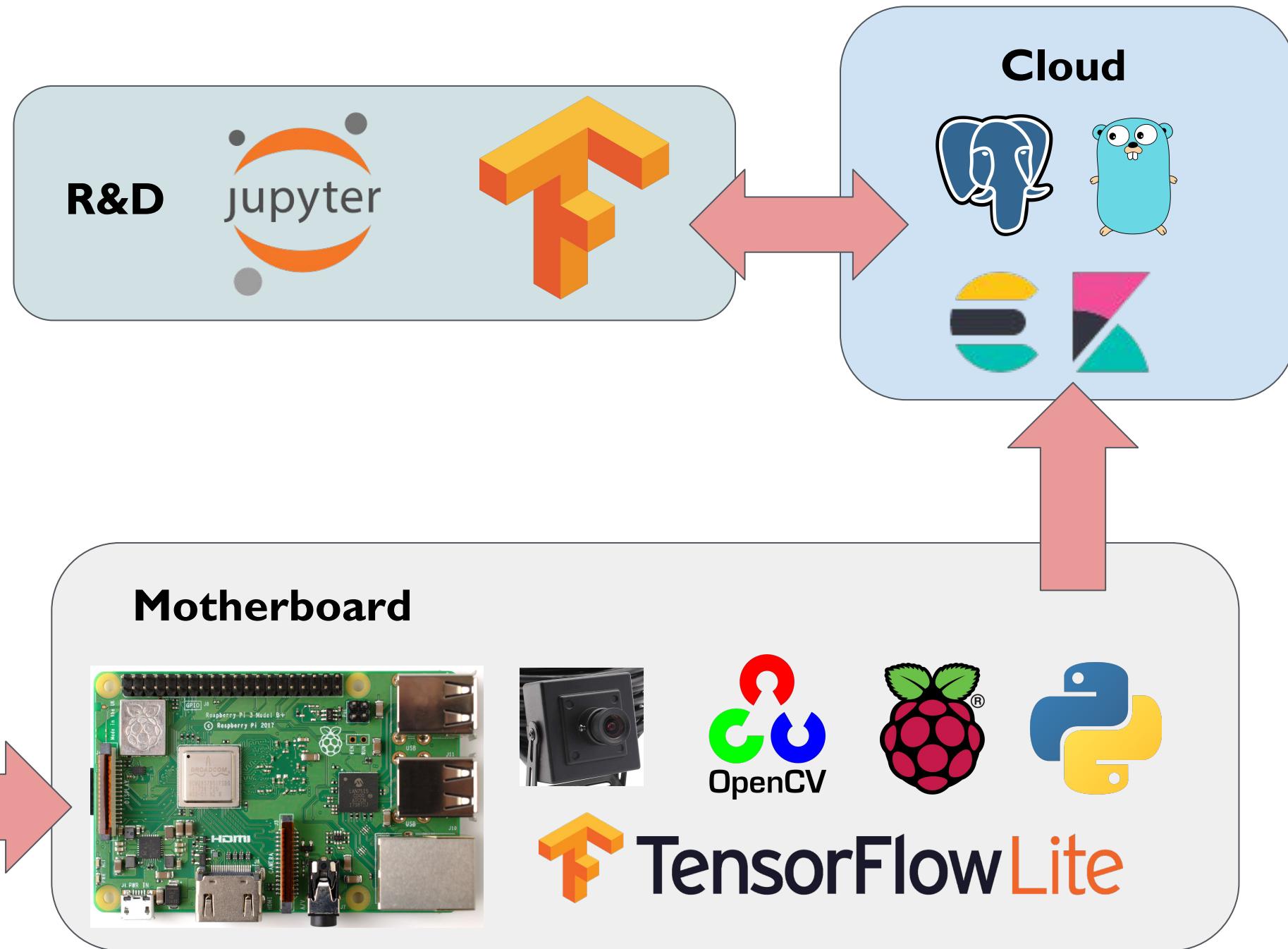
— PPFD (par_sq500_01) — pH (ph_atlas_02) — Air Temp. (temp_am2315_01)



System Architecture



Sensors



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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?

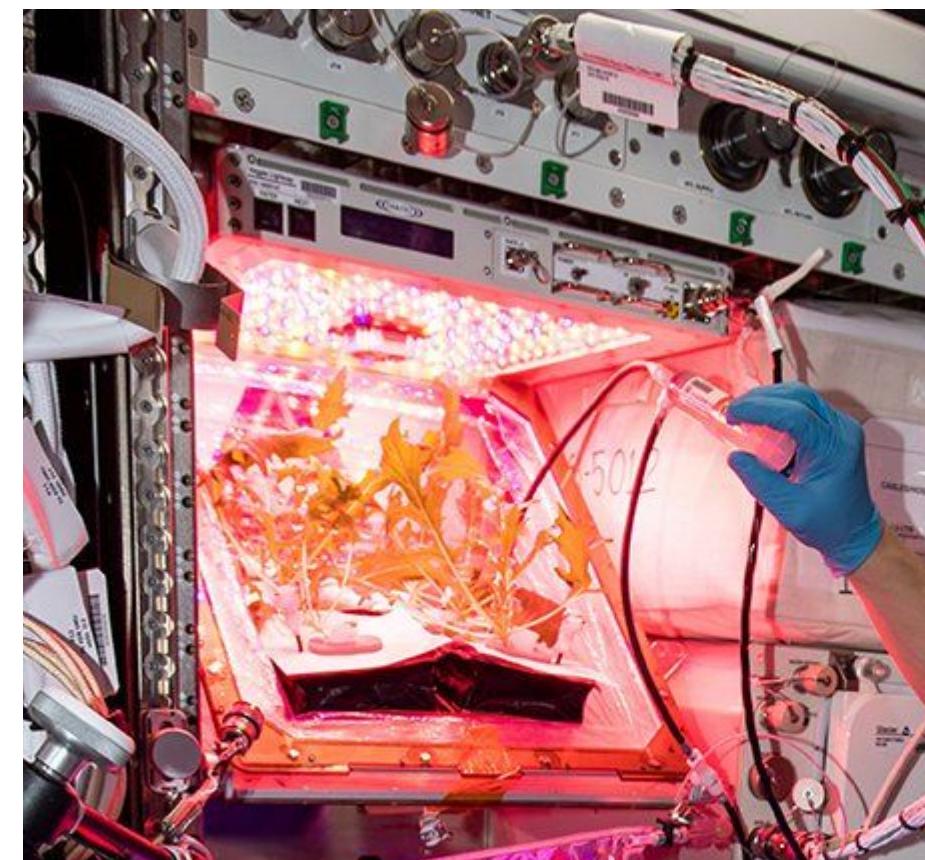
Infrastructure and cloud cost

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

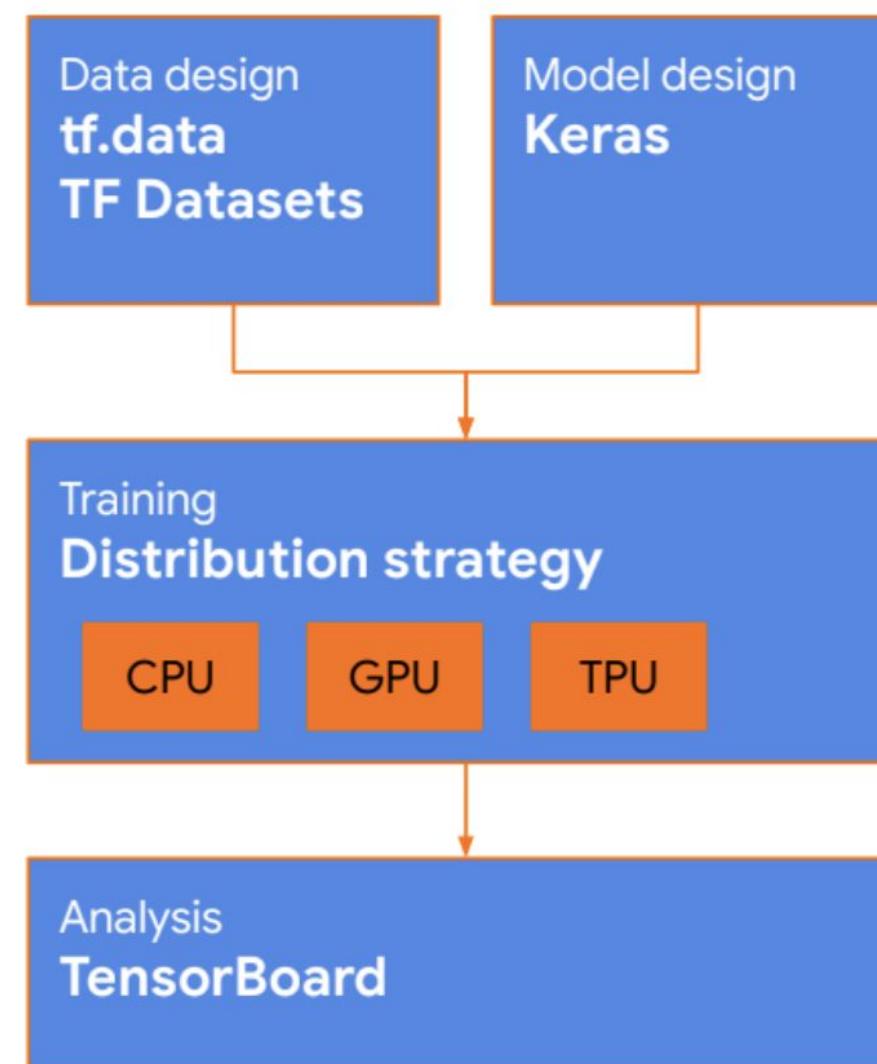


Tensorflow



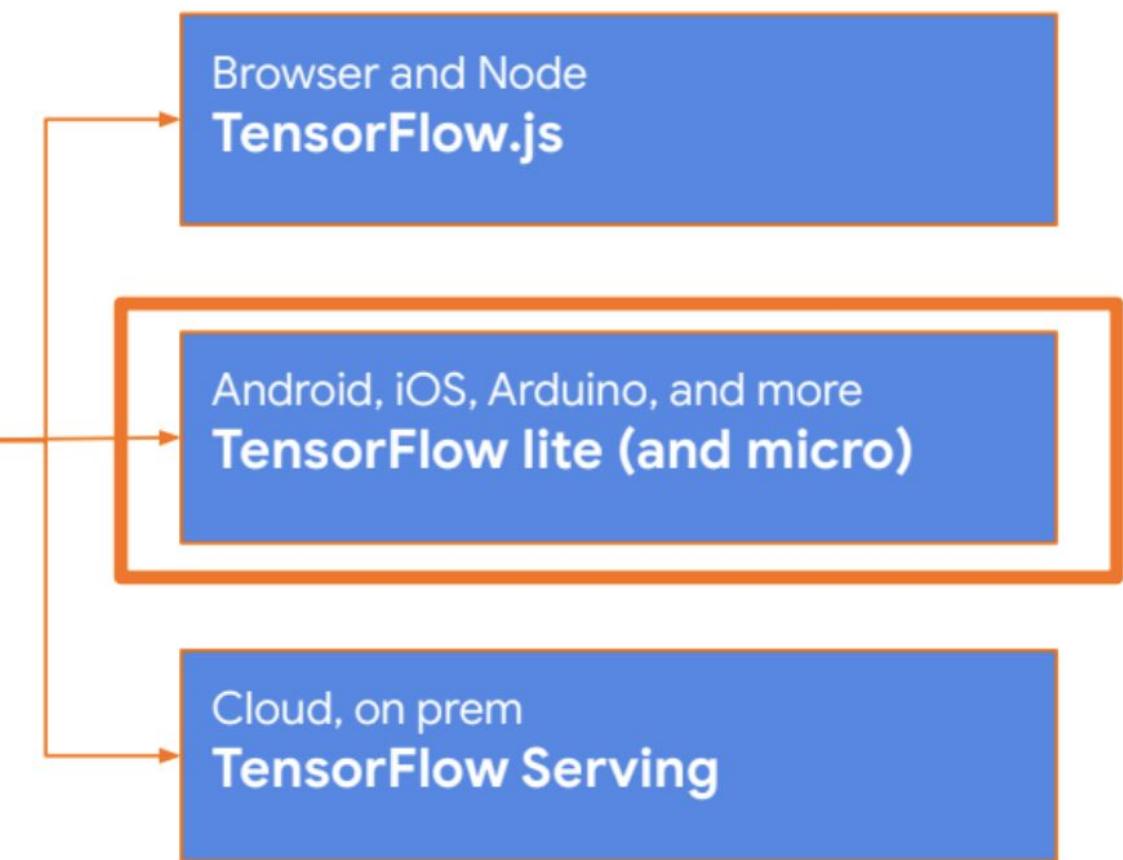
TensorFlow

Training



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

Deployment





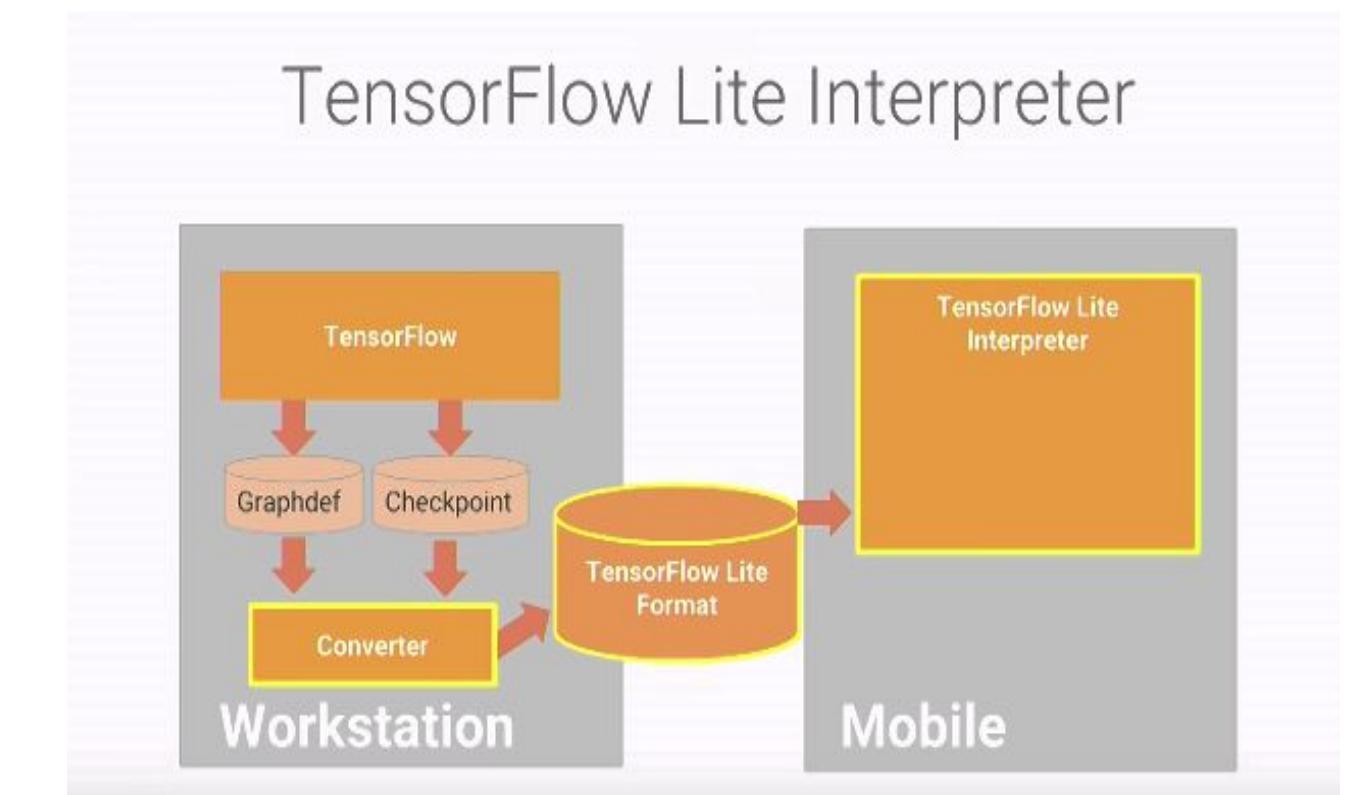
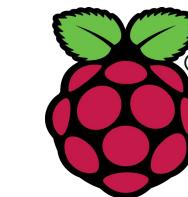
Tensorflow Lite (TF-Lite)



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)

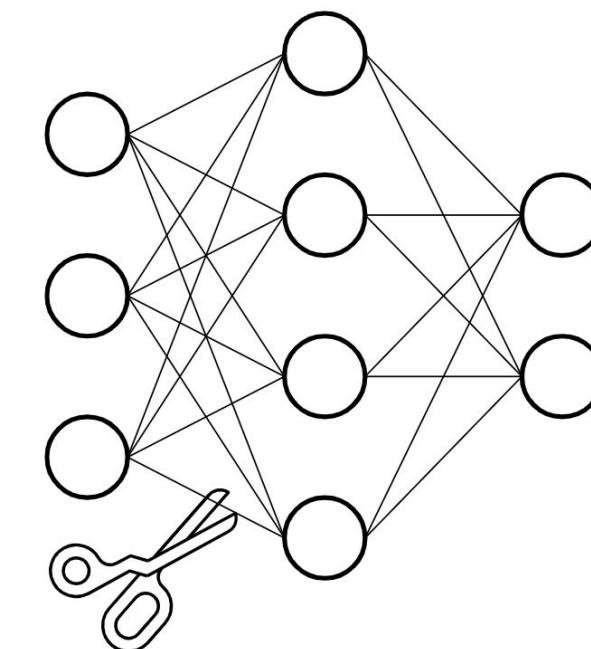


Optimization

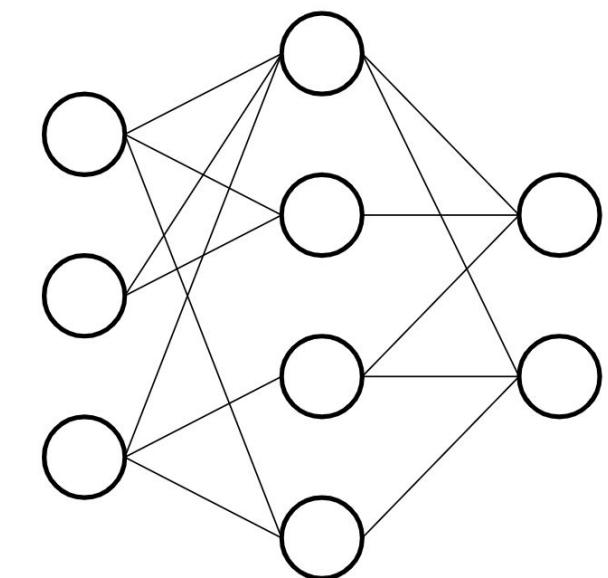
- Pruning
- Post Training Quantization

Delegate to offload execution

- GPU, TPU, DSP



Before pruning



After pruning



ML Workflow with Tensorflow Lite



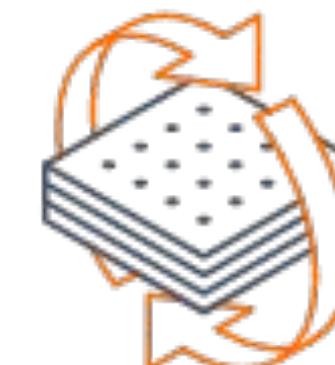
Import your dataset

Work on data: preprocessing, normalization, features selection

Build your model with Tensorflow

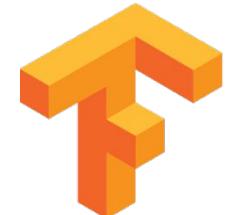
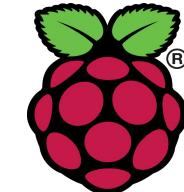
Train your model

Export and convert to .tflite





ML Workflow with Tensorflow Lite



TensorFlow Lite

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

Preprocess input data

Allocate Memory

Run inference

Interpret output

TM

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-cp37-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://storage.googleapis.com/tflite-repo/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.  
=====>] 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

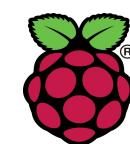
```
sudo apt-get install -y libhdf5-dev libc-ares-dev libeigen3-dev  
pip install keras_applications==1.0.8 --no-deps  
pip install keras_preprocessing==1.1.0 --no-deps  
pip install h5py==2.9.0  
sudo apt-get install -y openmpi-bin libopenmpi-dev  
sudo apt-get install -y libatlas-base-dev  
pip install 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

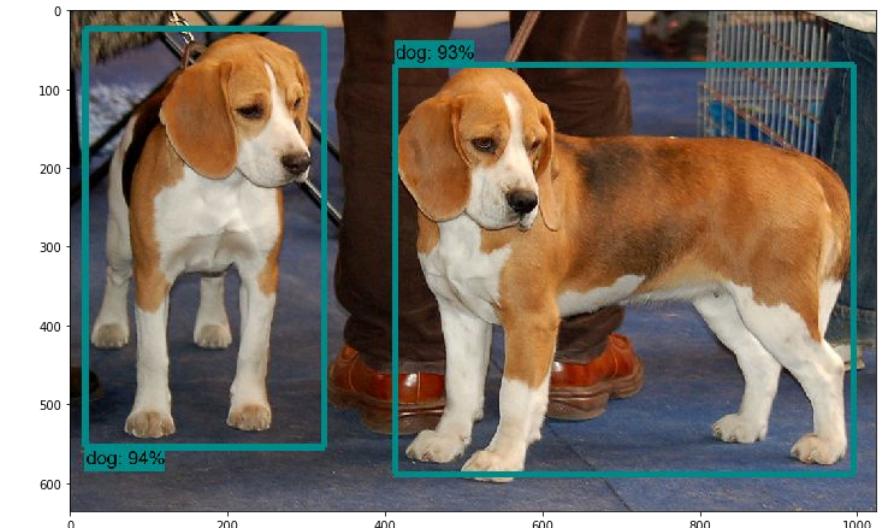
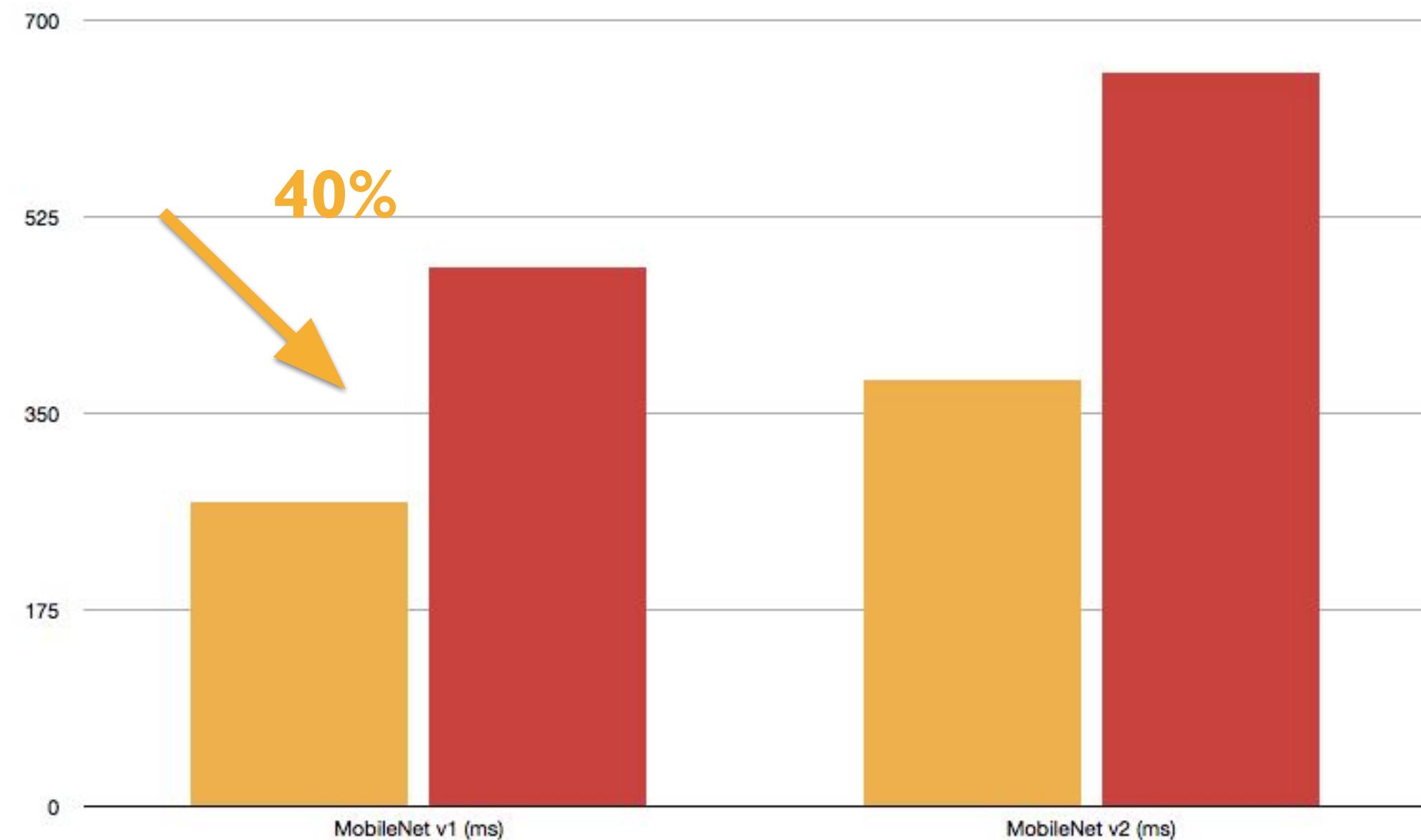


Part 2 - Deploy your tflite model on RPI and run inference



Tensorflow Lite Benchmark

Inference Time

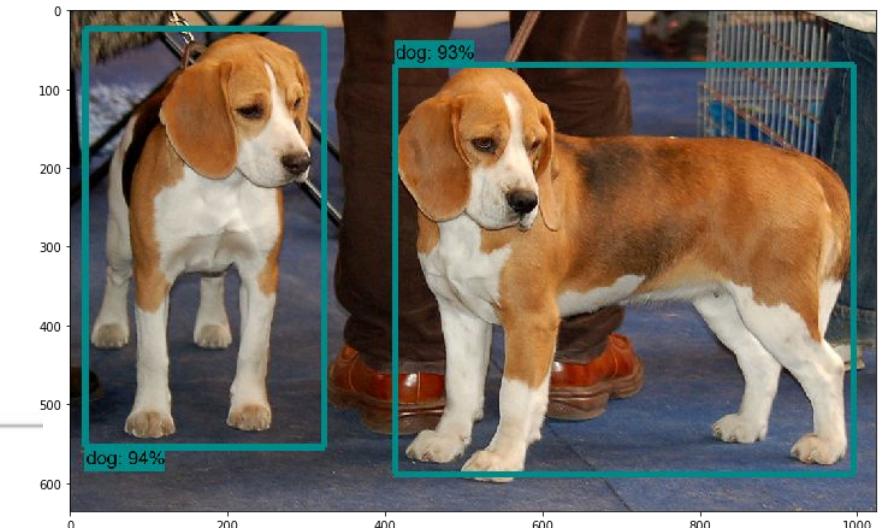
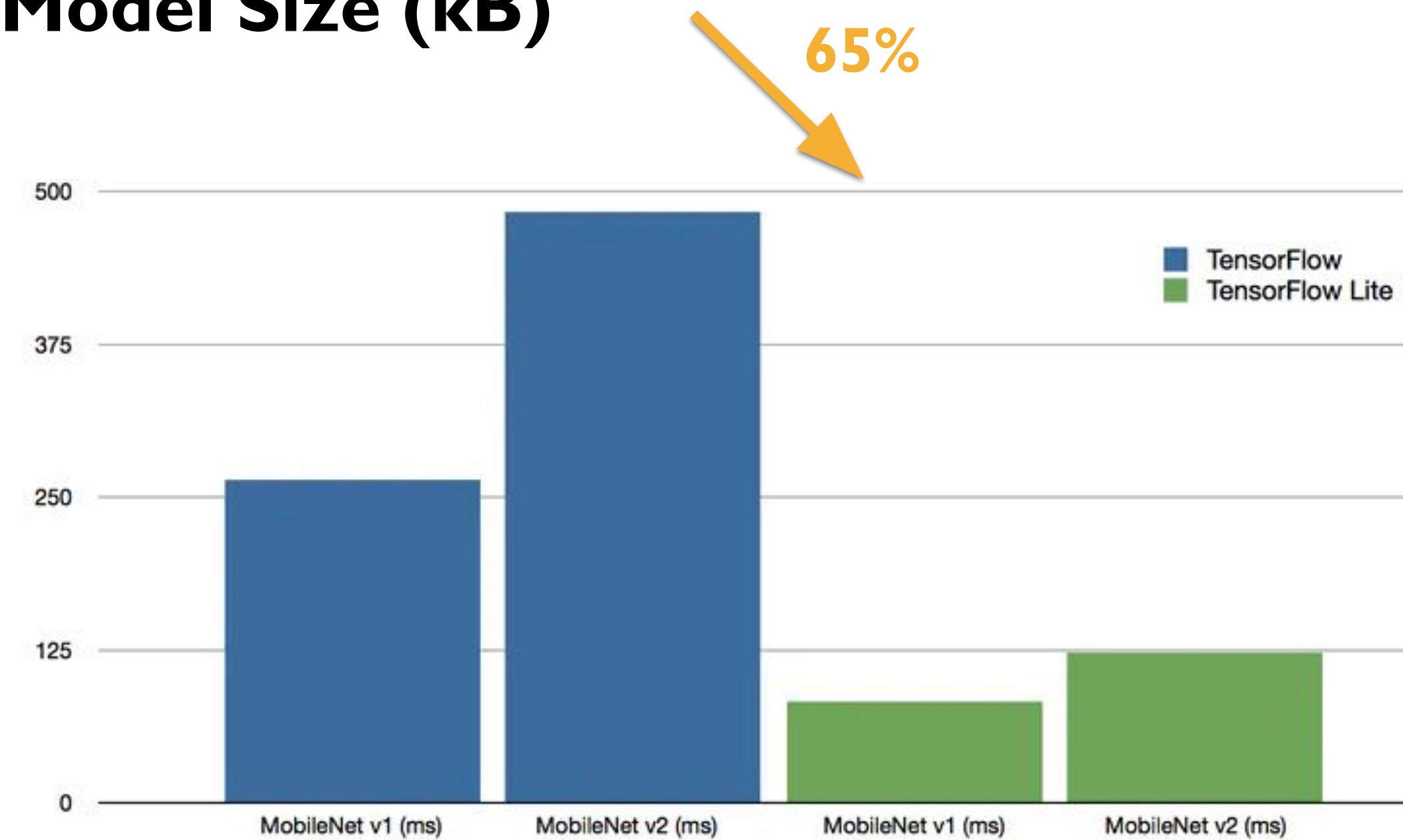


Source: Alasdair Allan



Tensorflow Lite Benchmark

Model Size (kB)



Source: Alasdair Allan



Tensorflow Lite Limitations

Reinforcement Learning

Transfer Learning

Recurrent Neural Network (RNN) like LSTM

Operation Compatibility

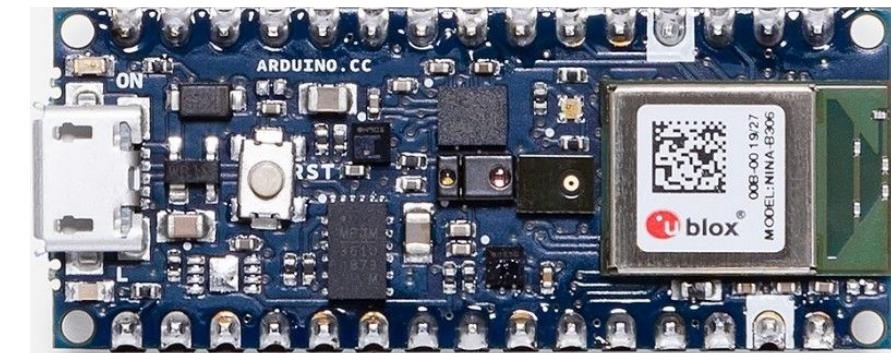
https://www.tensorflow.org/lite/guide/ops_compatibility

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

TM



TFLite on Microcontrollers



Inference on Cortex-M microcontroller

Arduino Nano 33

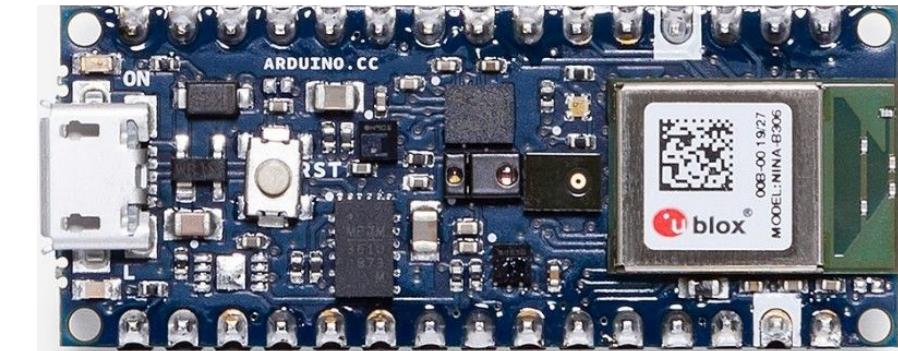
Only some operations are supported

Enough for hotword, gesture and speech
recognition





TFLite on Microcontrollers



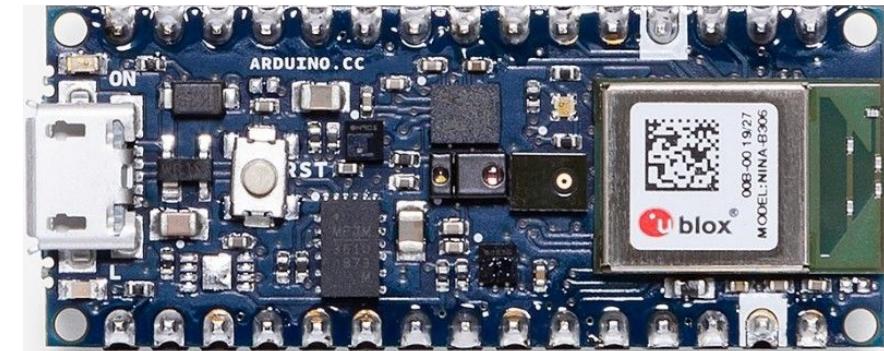
C++ API

```
#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



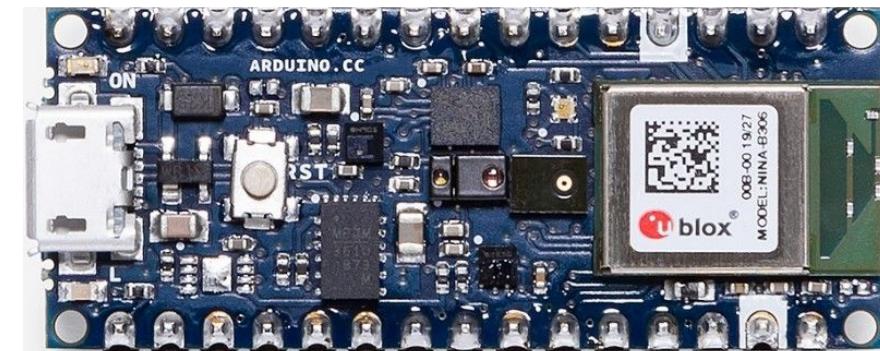
```
// Finding the min value for your model may require tests!
constexpr int kTensorArenaSize = 2 * 1024;
uint8_t tensor_arena[kTensorArenaSize];

// 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

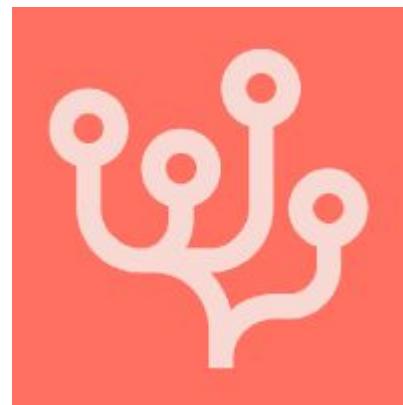
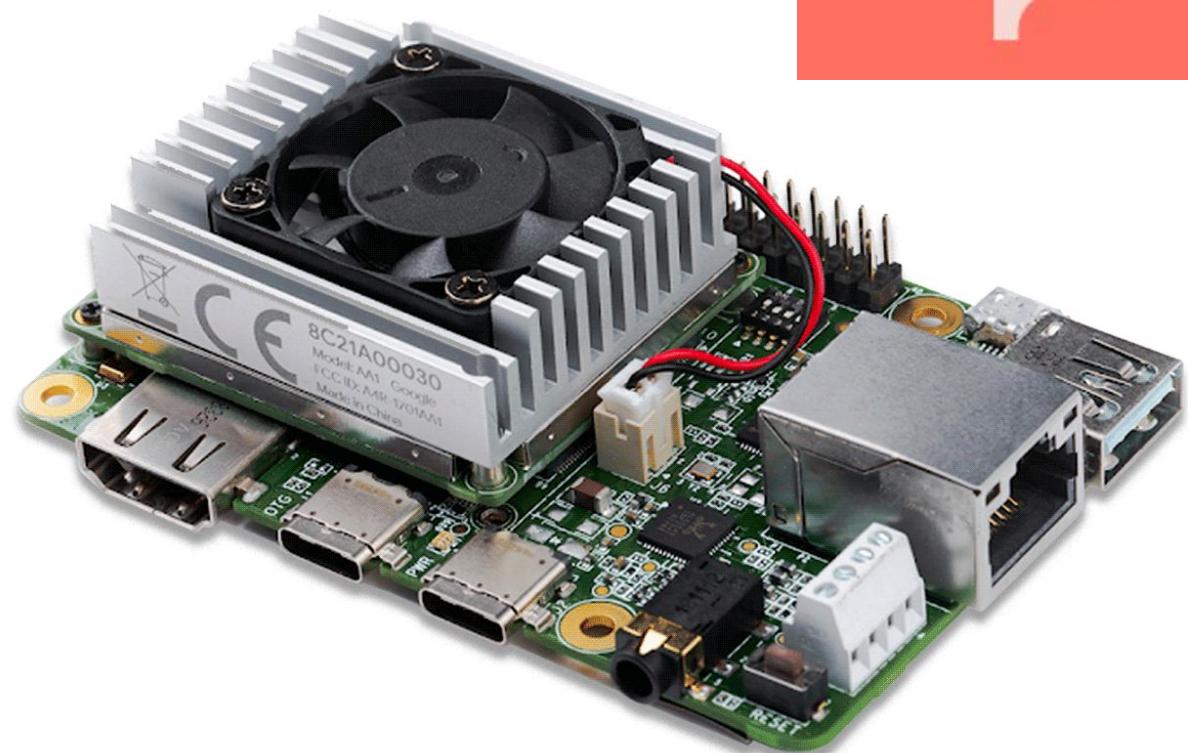


```
// Allocate memory for the model's tensors.  
TfLiteStatus allocate_status = interpreter->AllocateTensors();  
// Obtain pointers to the model's input and output tensors.  
input = interpreter->input(0);  
output = interpreter->output(0);  
// Feed the interpreter with the input value  
float x_val = 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

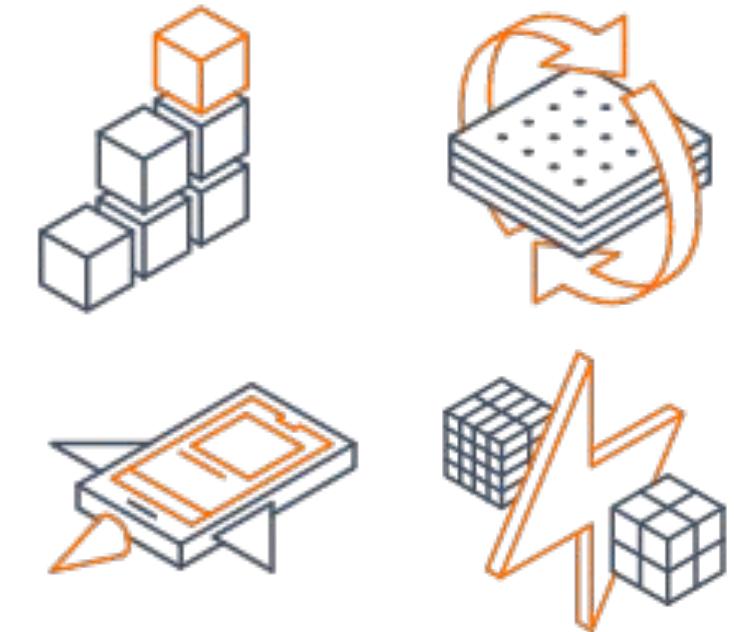
Deploy, Infer on device

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.

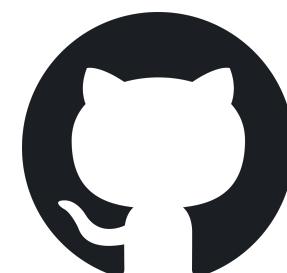




Thanks!



frama.link/tflite-Devoxx



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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>