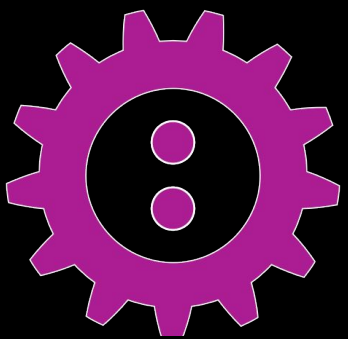


#FOSDEM



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



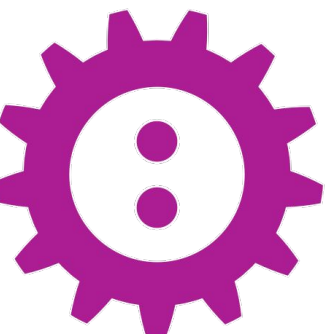
RTONE
IoT Makers

@alexis0duque



#FOSDEM #IoT #Tensorflow #AI

@alexis0duque



Who am I?

Alexis DUQUE

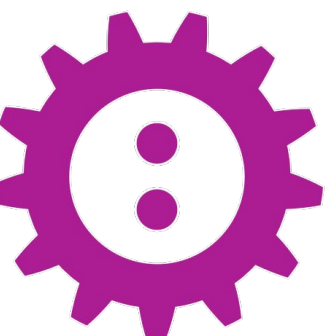
Director of Research & Development



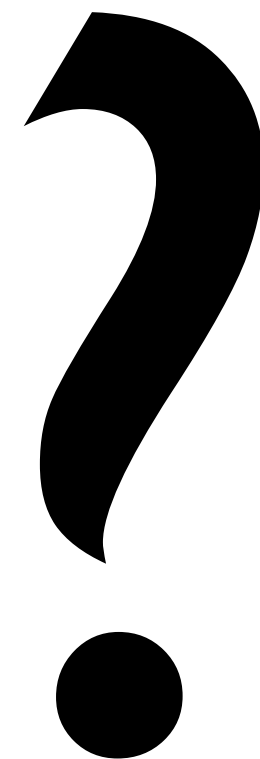
-  @alexis0duque
-  alexisduque
-  alexisd@rtone.fr
-  alexisduque.me
-  <https://goo.gl/oNUWu6>



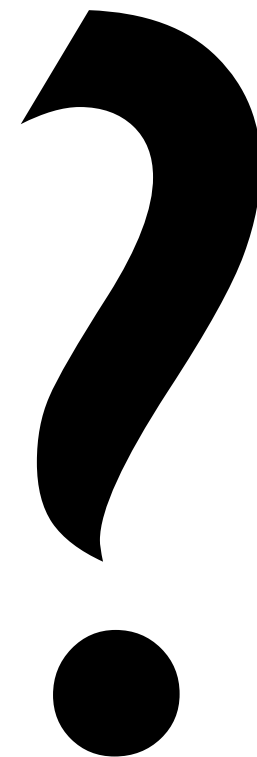
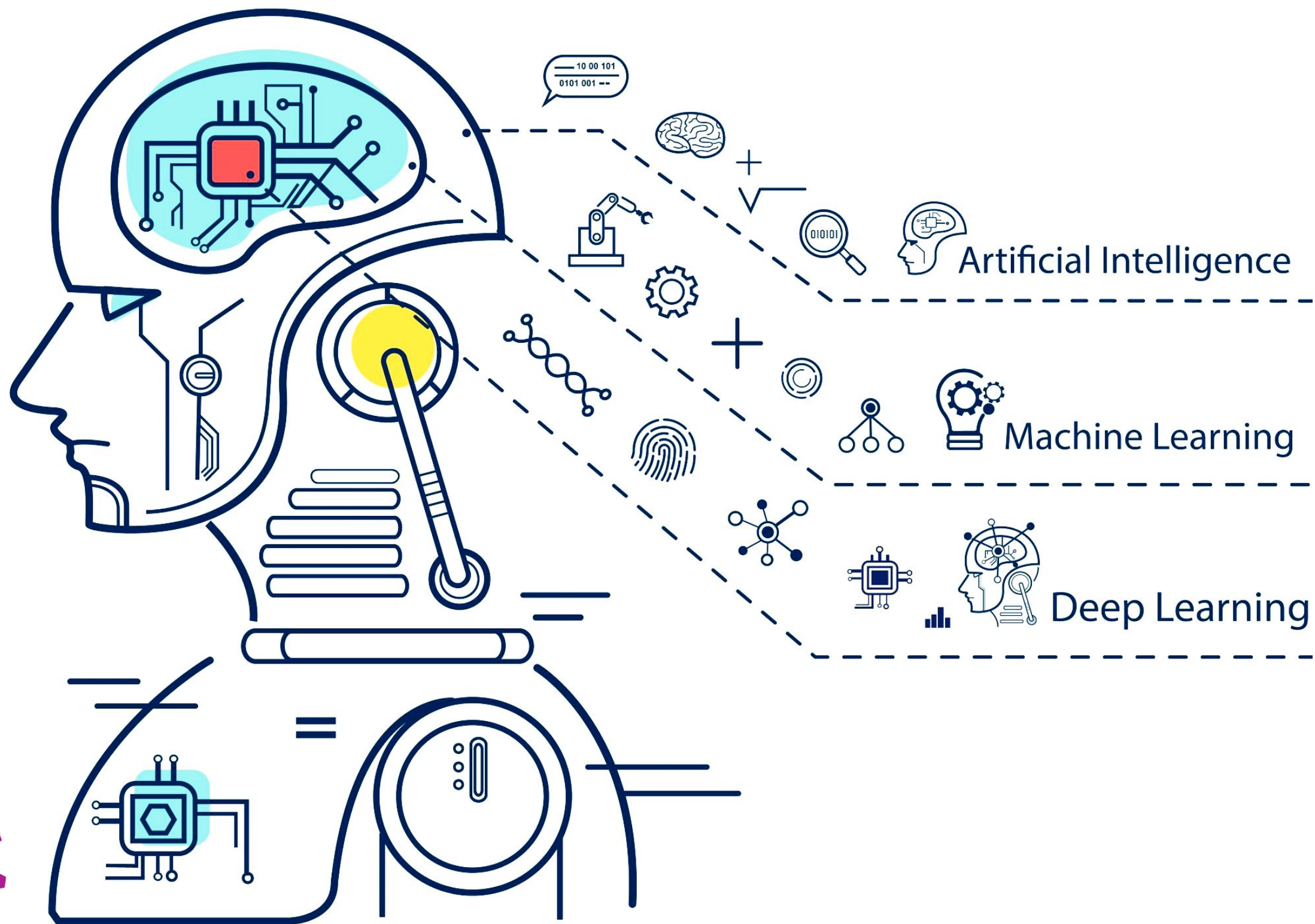
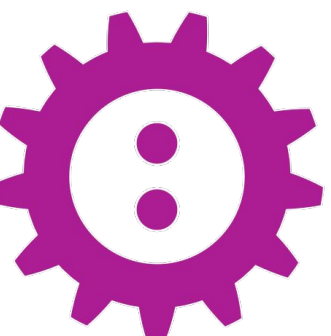
FOSDEM



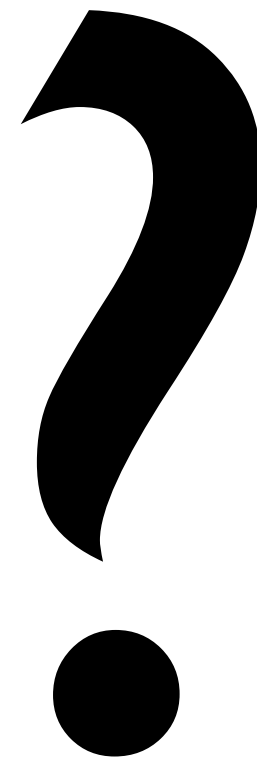
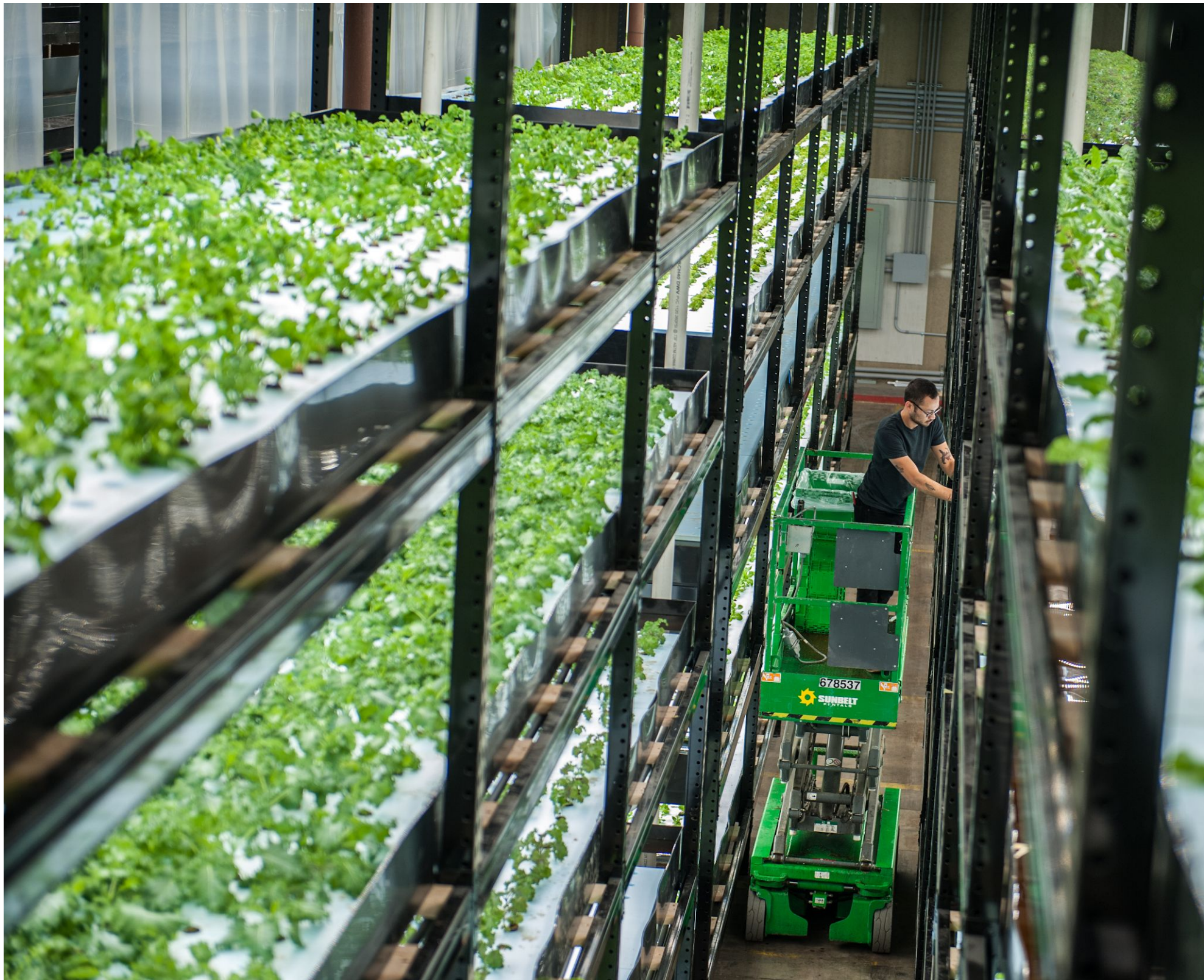
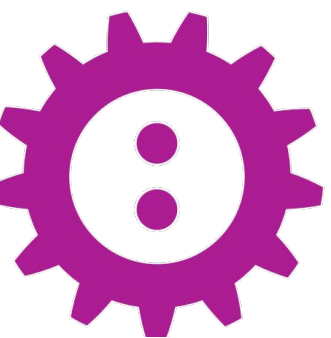
TensorFlow Lite



FOSDEM

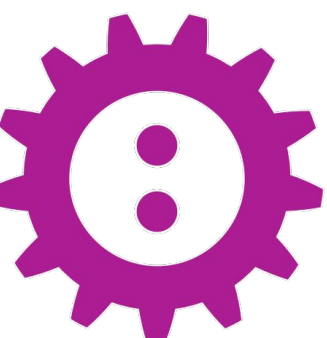


FOSDEM



#FOSDEM #IoT #Tensorflow #AI

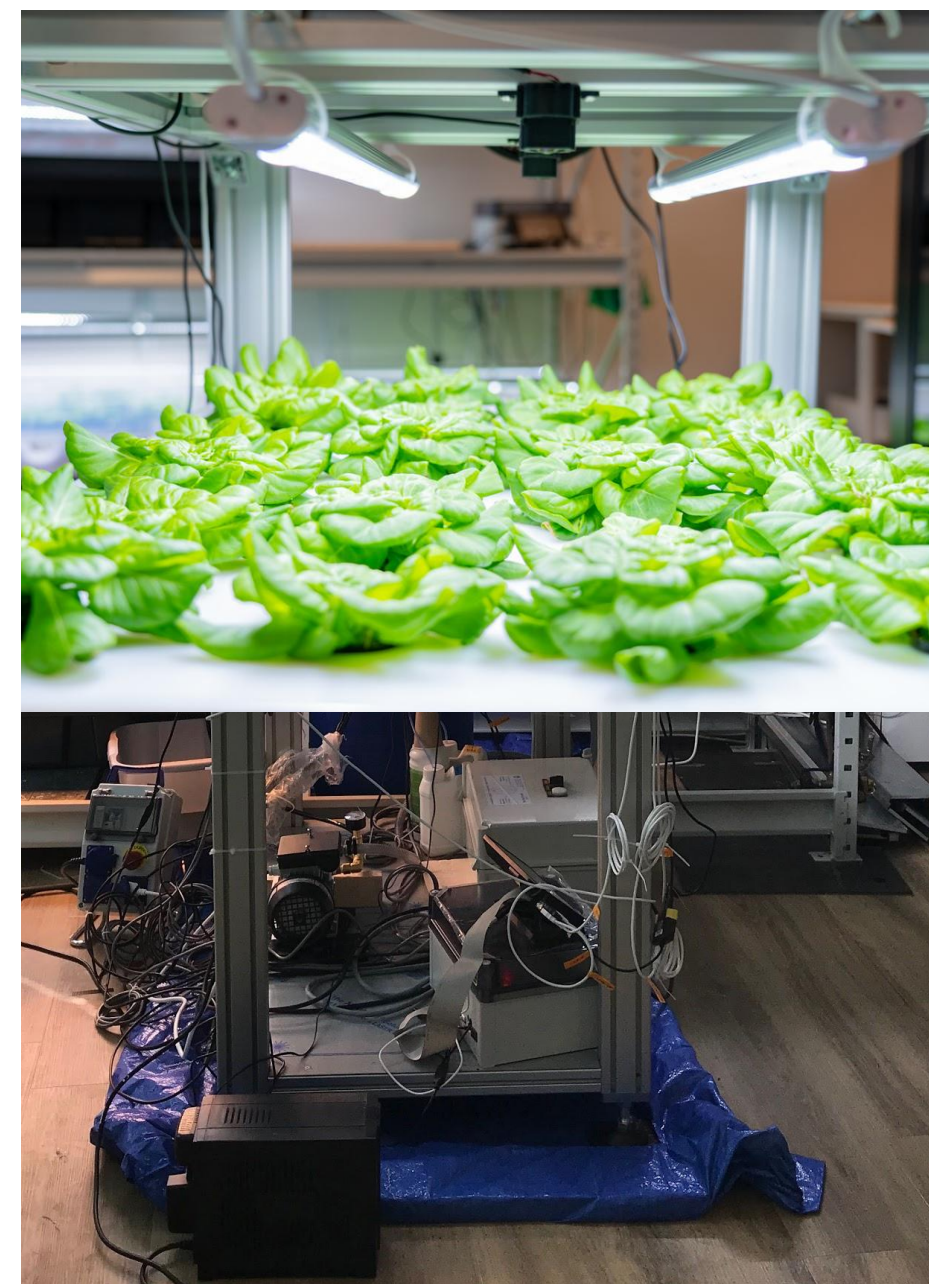
@alexis0duque



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

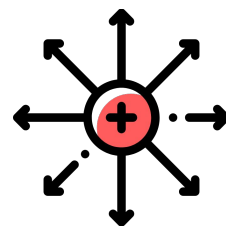


System Architecture



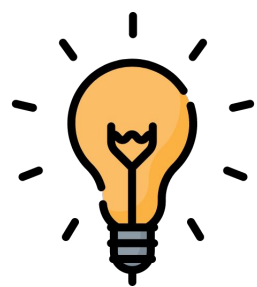
avg **co2** (double, in ppm)

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



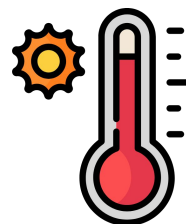
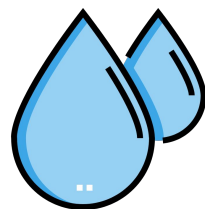
avg **electrical conductivity** (double, in $\mu\text{S}/\text{cm}$)

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



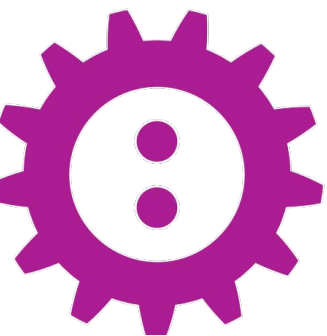
avg **PPFD** (Photosynthetic Photon Flux Density, double, in $\mu\text{mol}/\text{m}^2/\text{s}$)

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



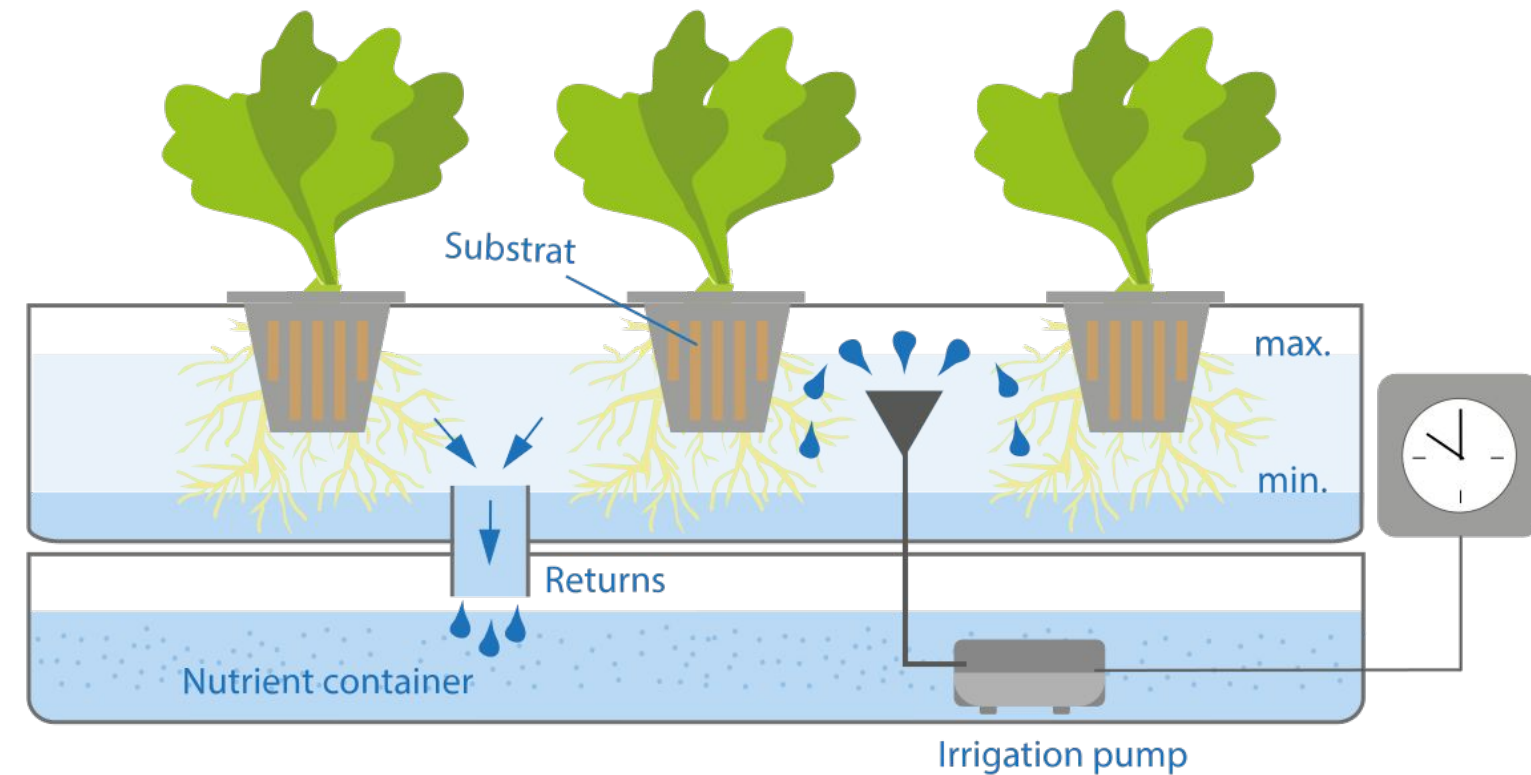
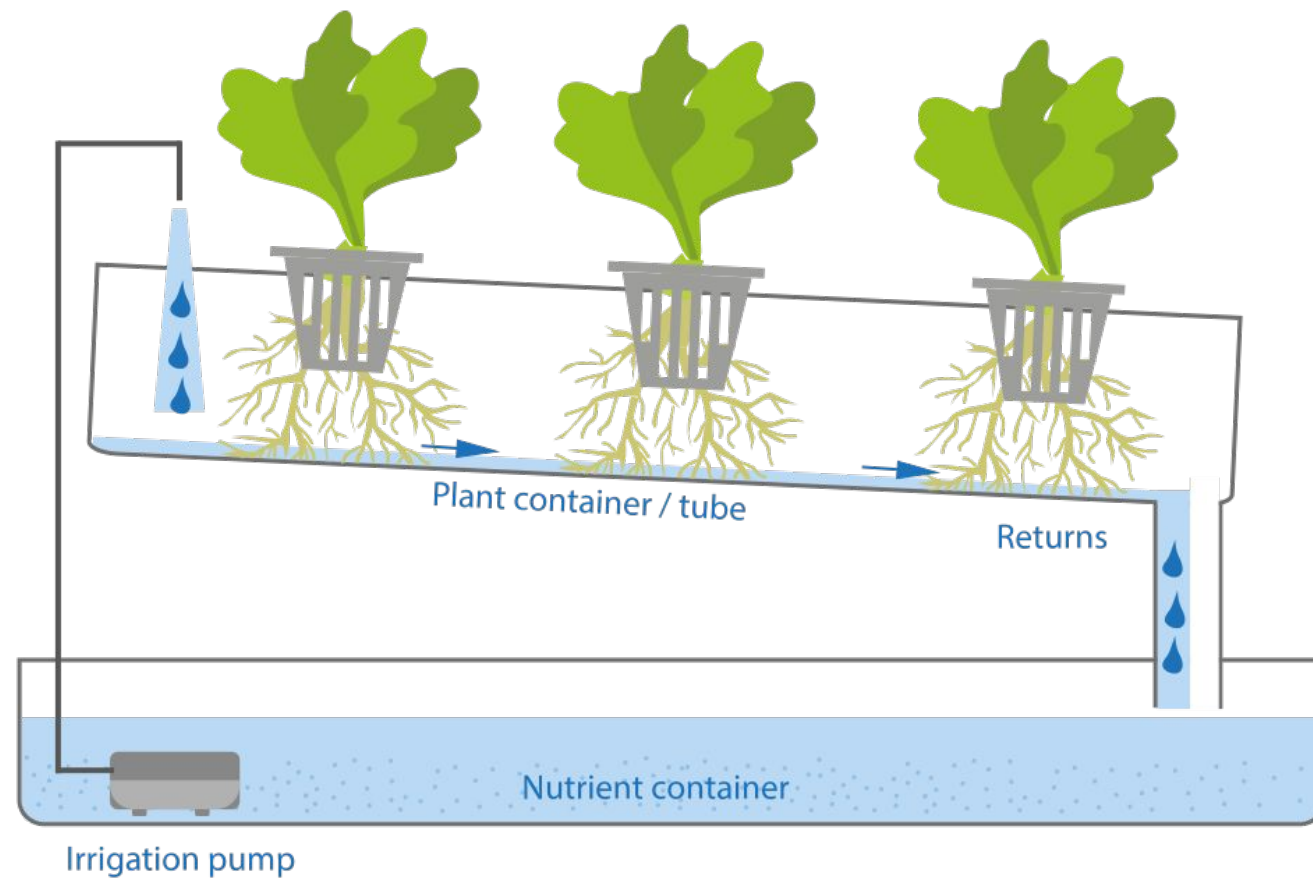
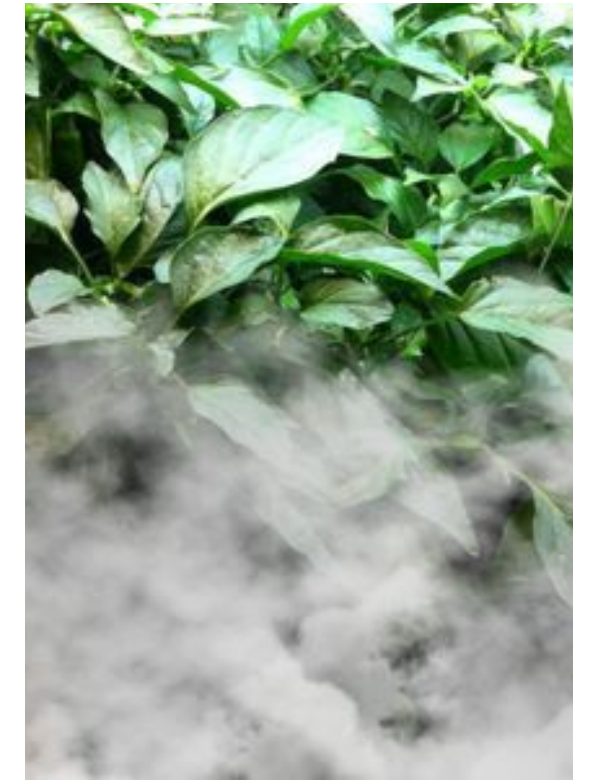
average **humidity** (double, in %)

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



Irrigation System

- HPA
- Nebulization
- NFT
- Ebb & Flow



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

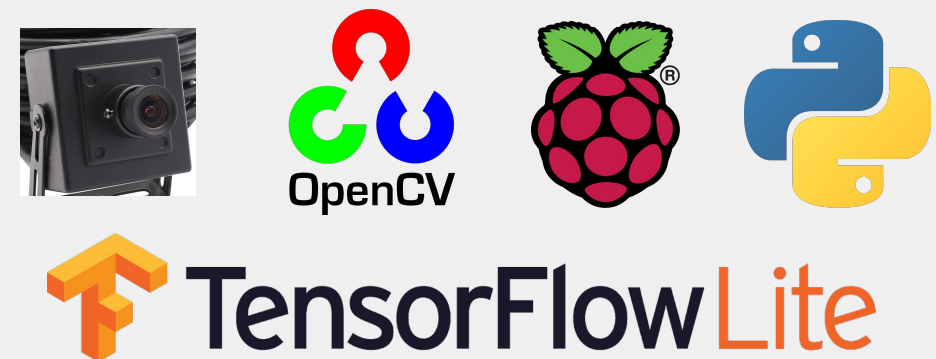
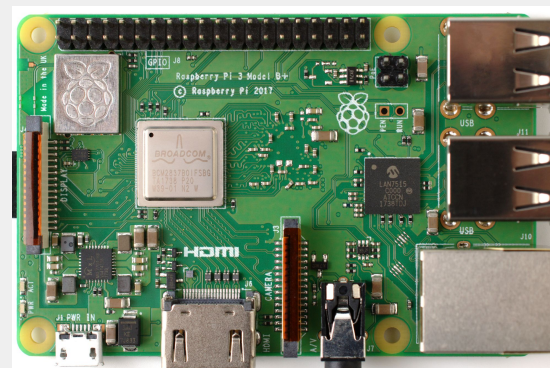
System Architecture

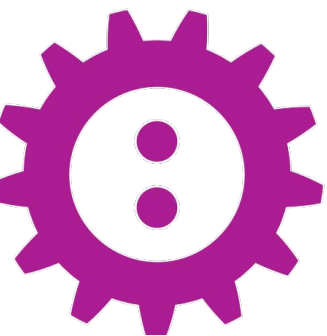


Sensors

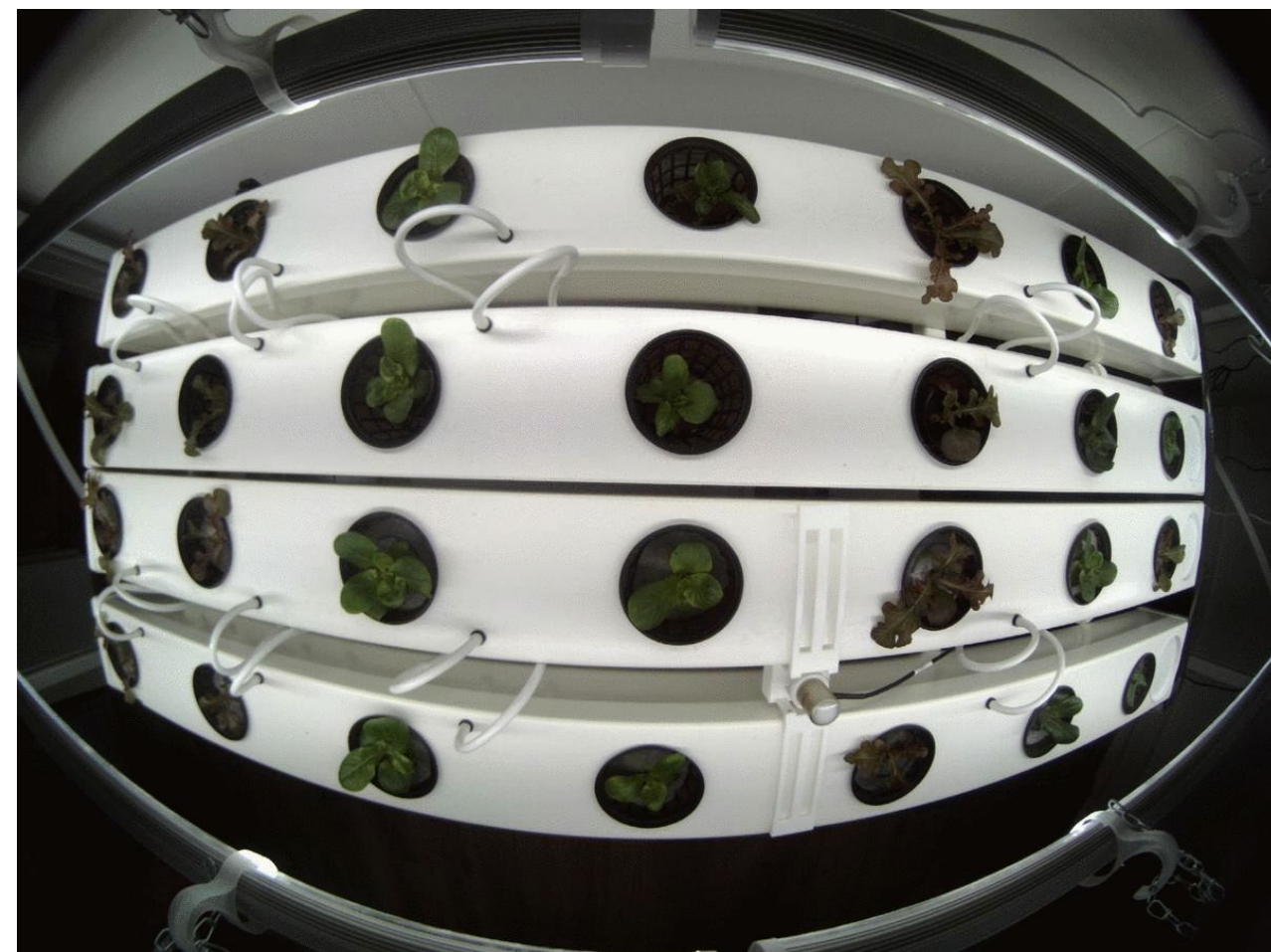
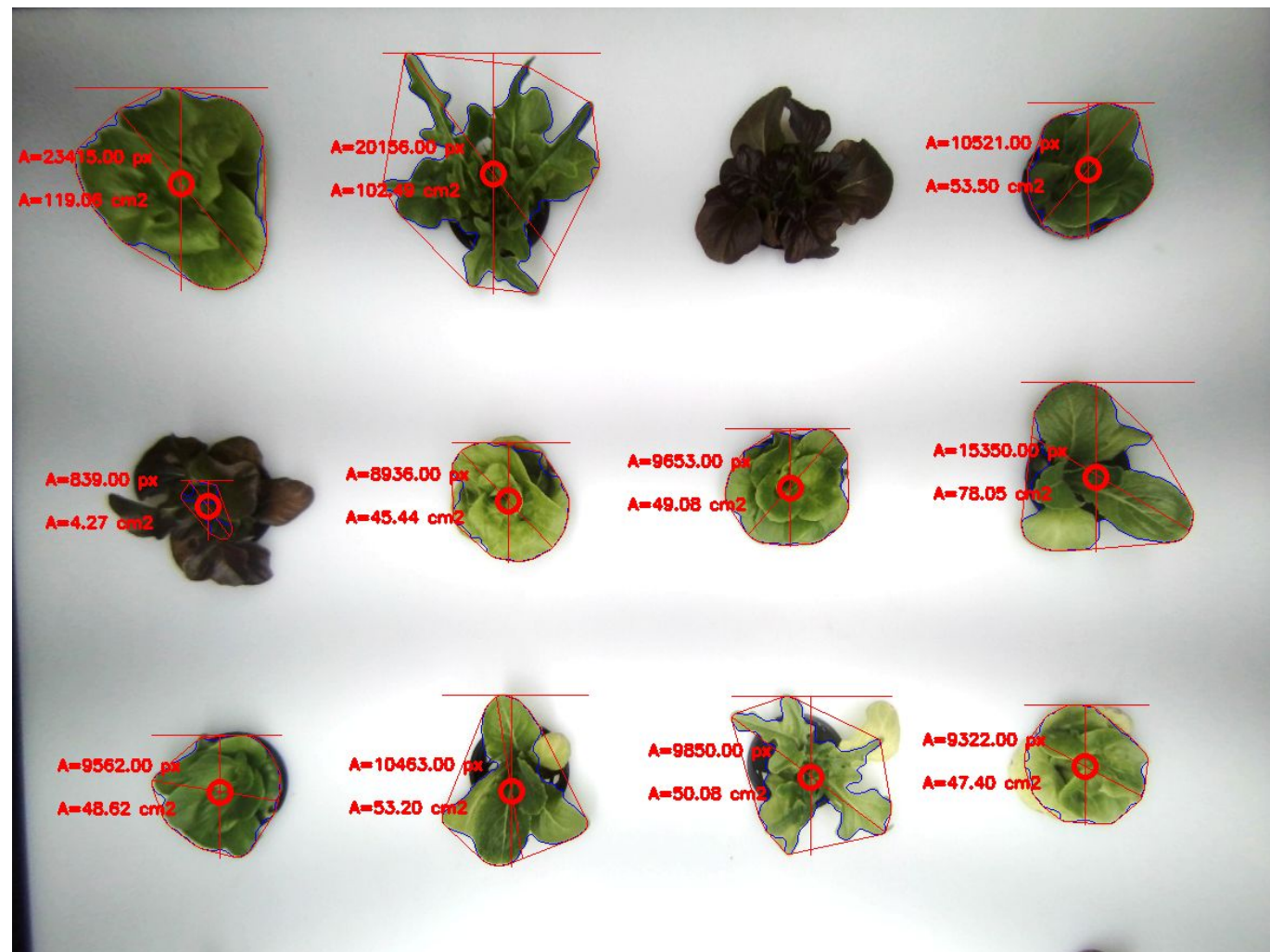


Motherboard

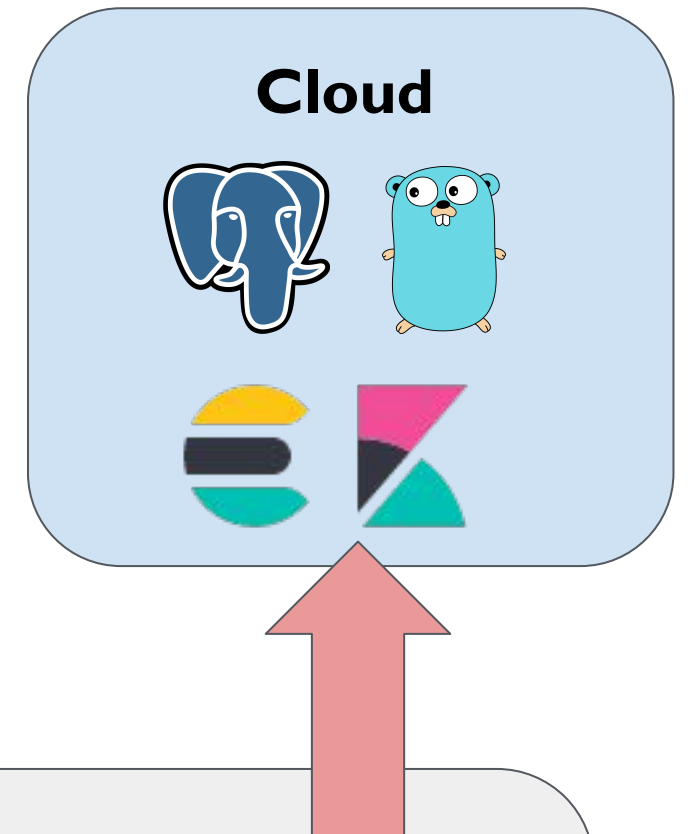
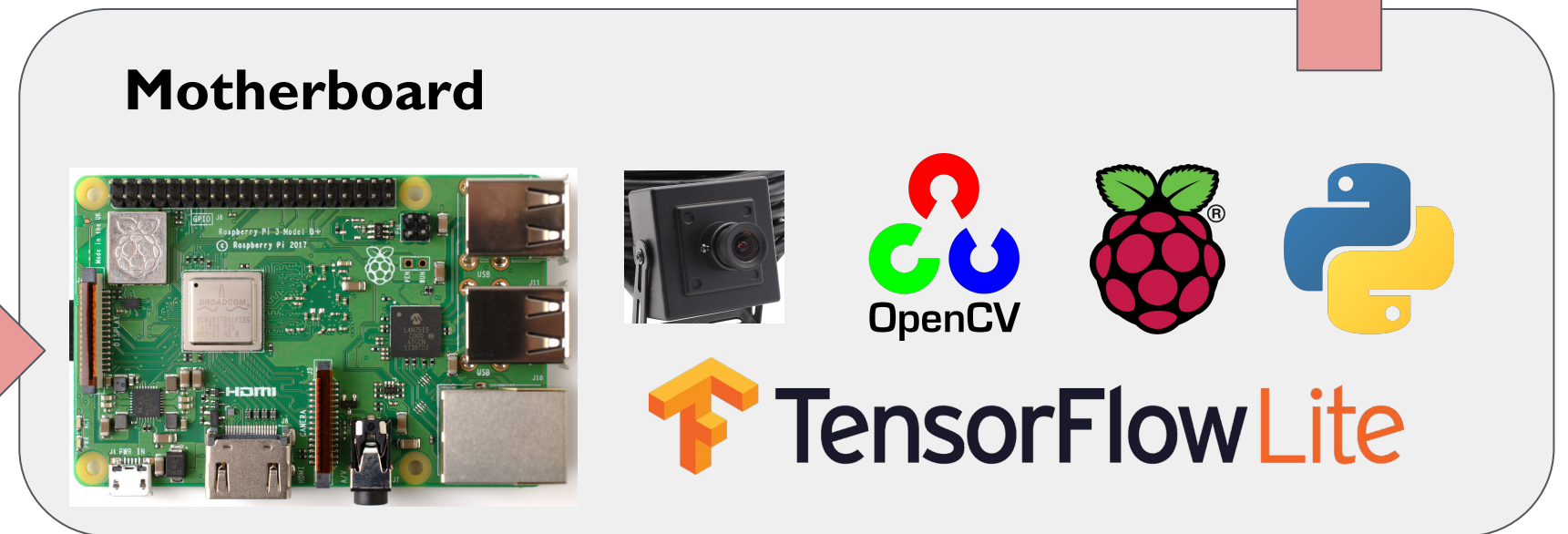


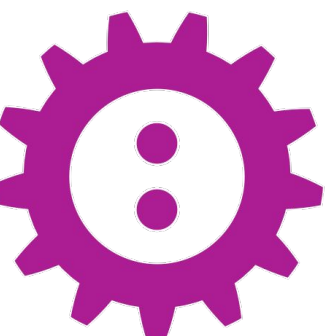


Computer Vision



System Architecture





5 - Unit 5 - 2 shelves Opened - EBB Flow

Activity Sensors Edition



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

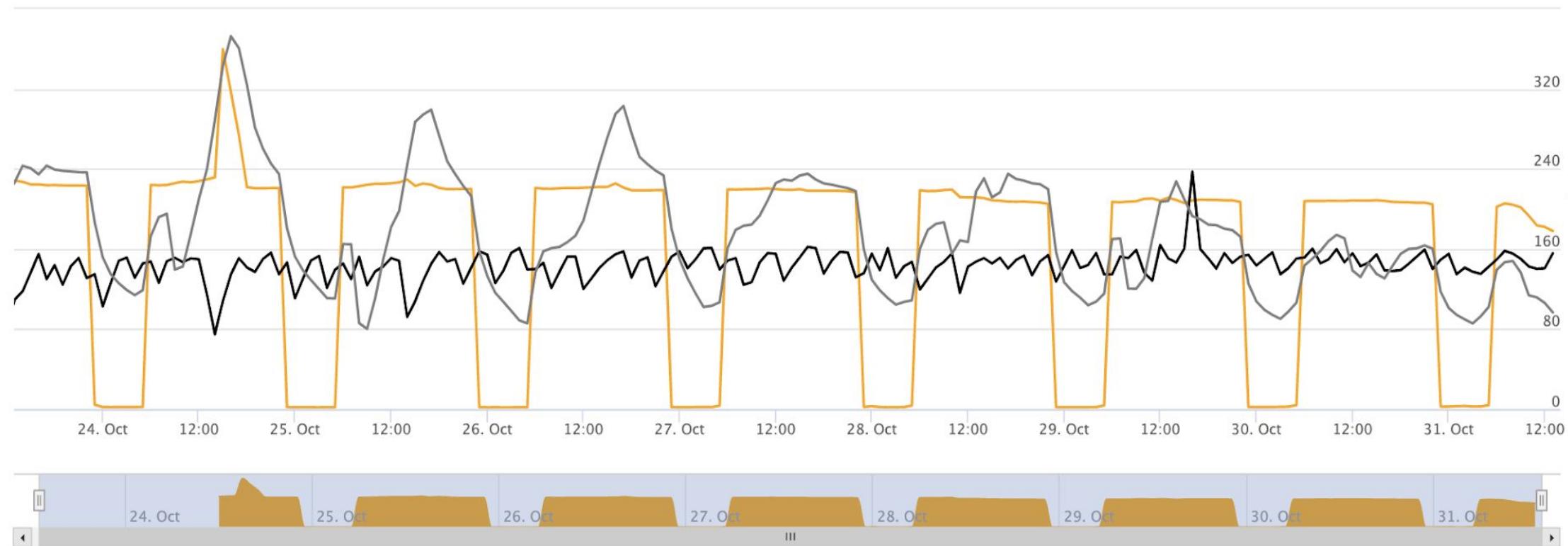
End date

10/31/2019, 2:56 PM

Displayed: average value per hour

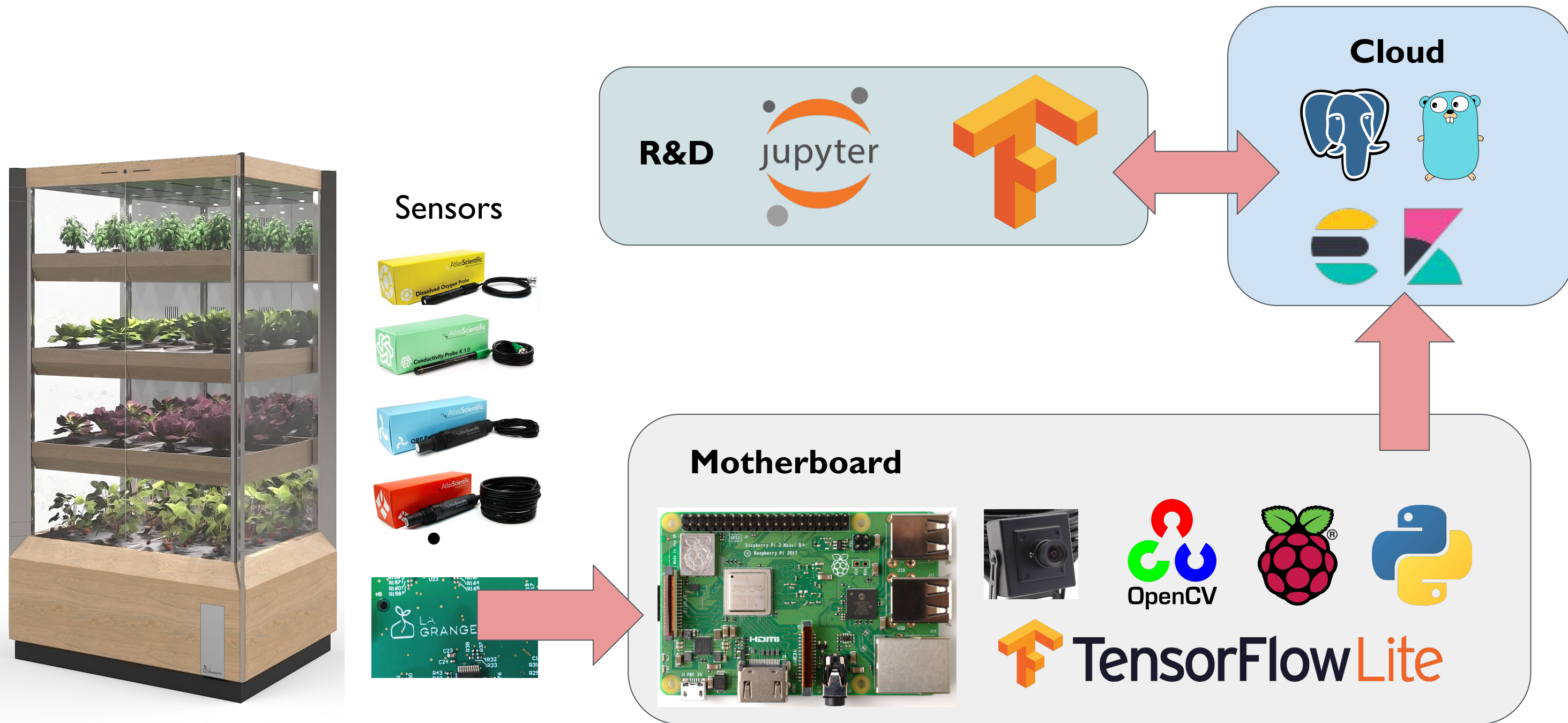
Zoom hour day week All

From Oct 23, 2019 To Oct 31, 2019



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

System Architecture



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

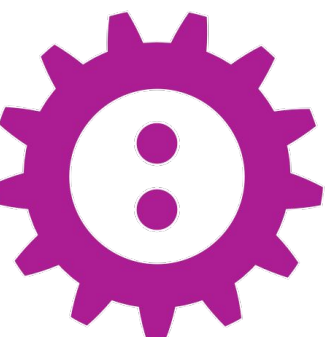
By Martin Finucane Globe Staff, April 3, 2019, 2:18 p.m.



[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.

FOSDEM



Why Edge Computing?

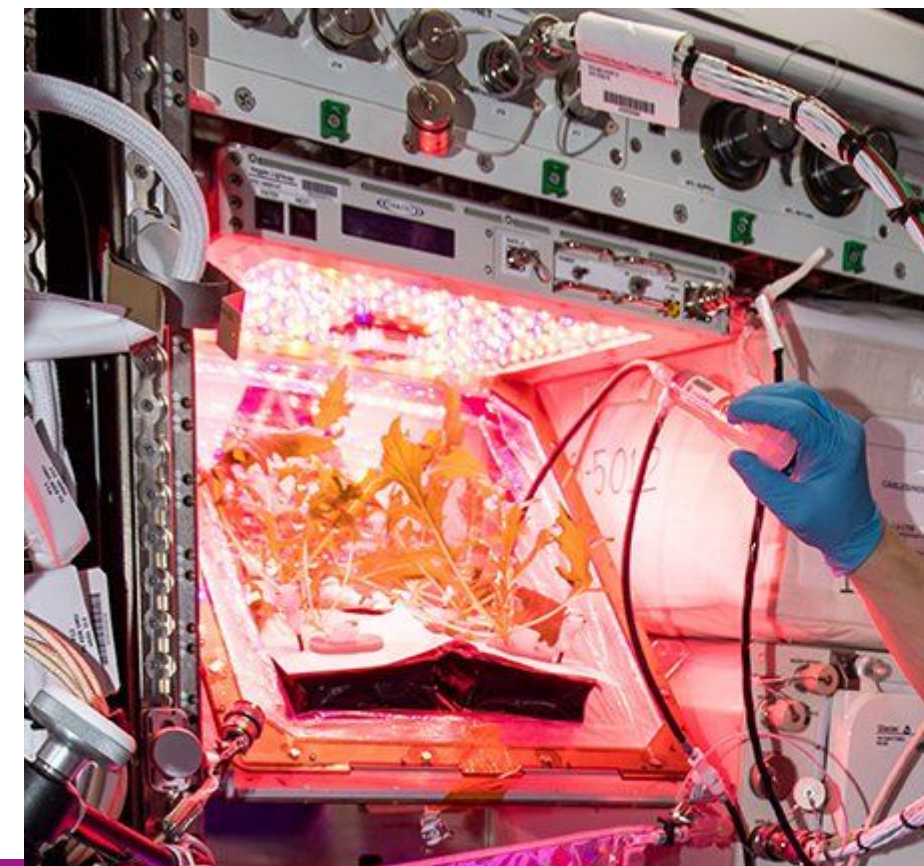
Infrastructure and cloud cost

Scalability

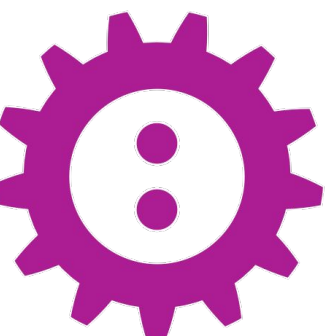
Must work in the field, without internet

Network Latency

Privacy



FOSDEM



Tensorflow



TensorFlow

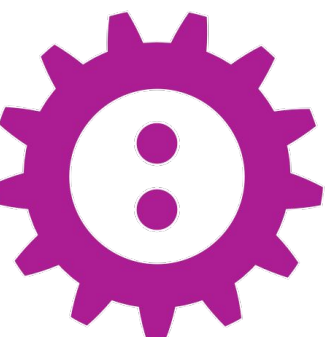
Open source library created by Google

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 Lite (TF-Lite)



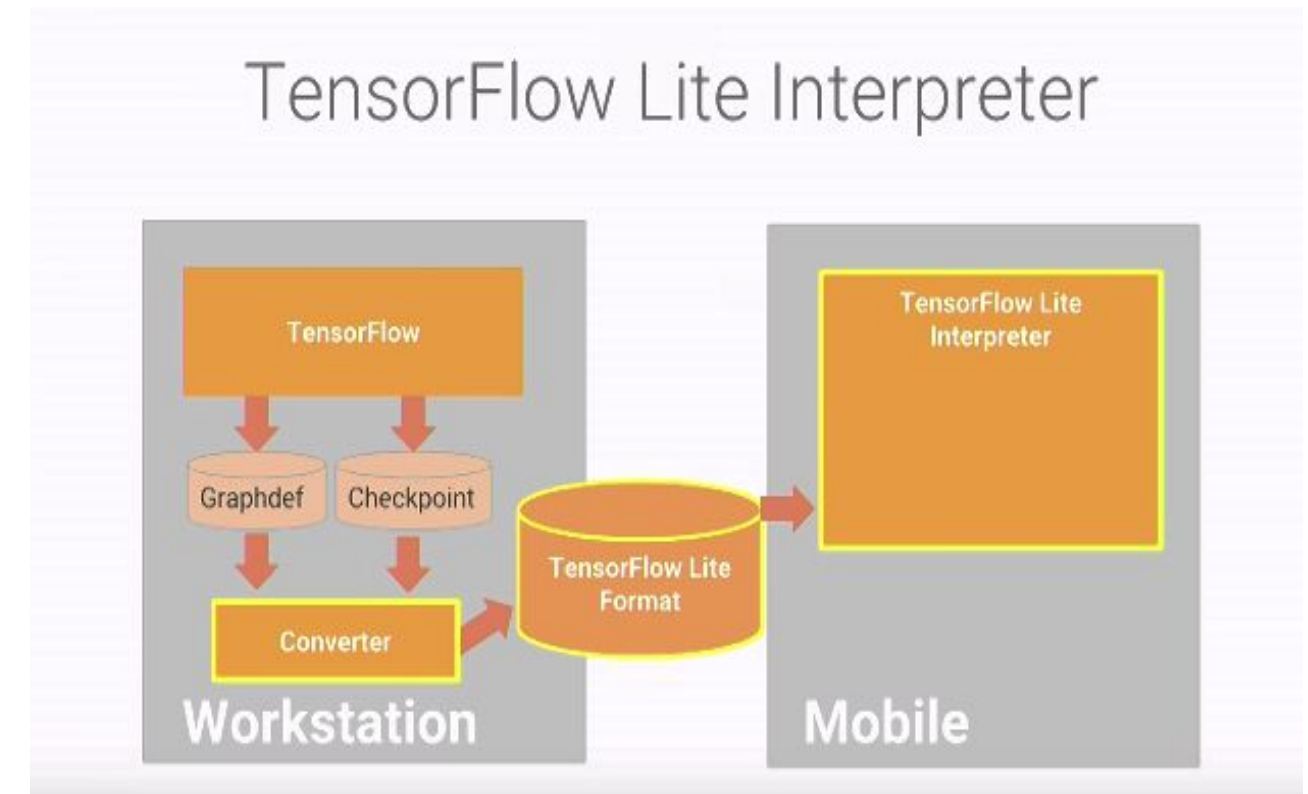
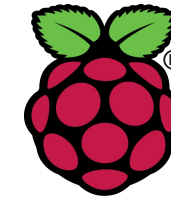
Converter + Interpreter

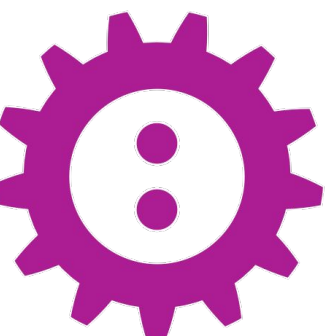
Tensorflow vs TF-Lite ?

- smaller model size
- faster inference
- mobile, embedded, MCU

but

- no training
- model is frozen => no re-training
- no transfer learning





Tensorflow Lite (TF-Lite)

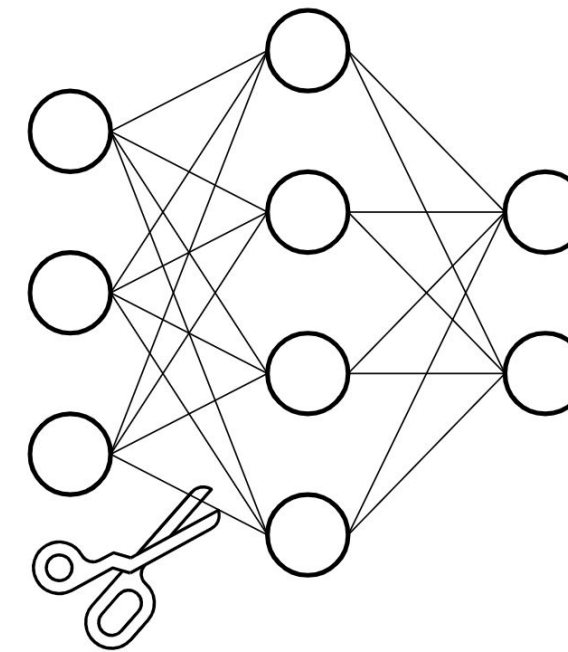


Optimization

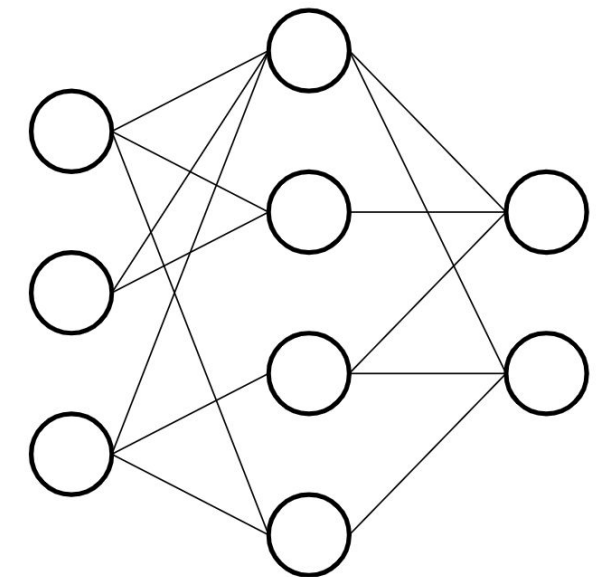
- Pruning
- Post Training Quantization

Delegate to offload execution

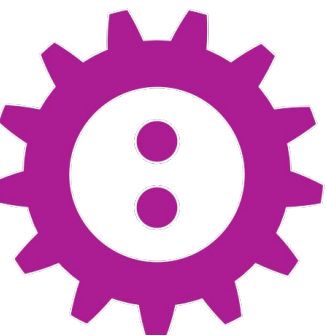
- GPU, TPU, DSP



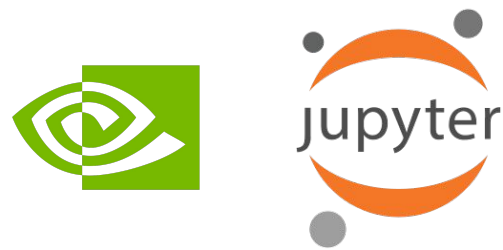
Before pruning



After pruning



ML Workflow with Tensorflow Lite



Import your dataset

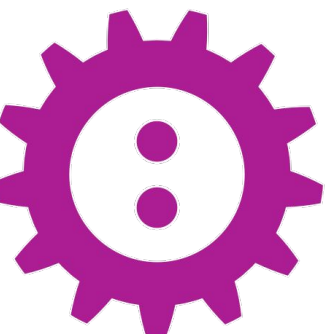
Work on data: preprocessing, normalization, features selection

Build your model with TensorFlow

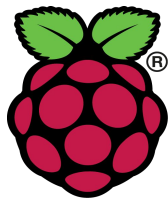
Train your model

Export and convert to .tflite





ML Workflow with Tensorflow Lite



TensorFlow Lite

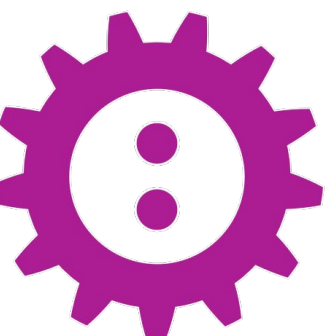
Load your model (or grab one in github.com/tensorflow/models)

Preprocess input data

Allocate Memory

Run inference

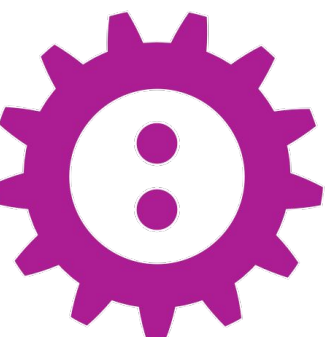
Interpret output



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
```

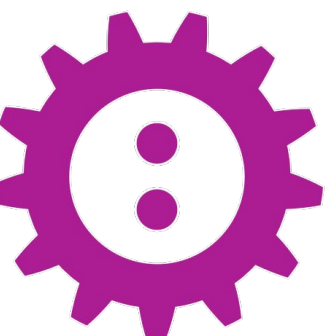
Setup on your RPI

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**



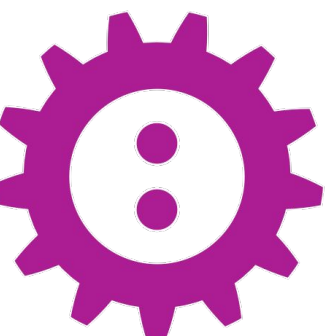
<https://github.com/PINTO0309/Tensorflowlite-bin>



Setup on your RPI

Tensorflow Lite Interpreter

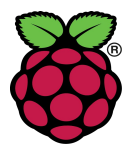
```
$ sudo apt install swig libjpeg-dev zlib1g-dev python3-dev  
python3-numpy unzip  
$ wget  
https://github.com/PINT00309/TensorflowLite-bin/raw/master/2  
.0.0/tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl  
$ pip install --upgrade  
tflite_runtime-2.0.0-cp37-cp37m-linux_armv7l.whl
```

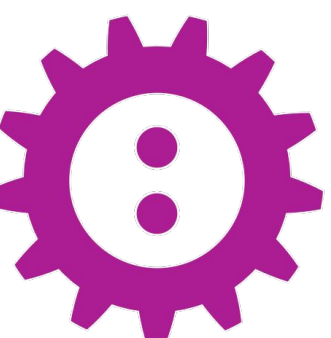
Demo



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 Limitations

Reinforcement Learning

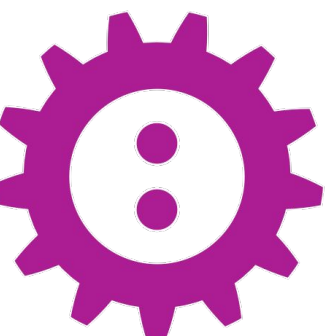
Transfer Learning

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>

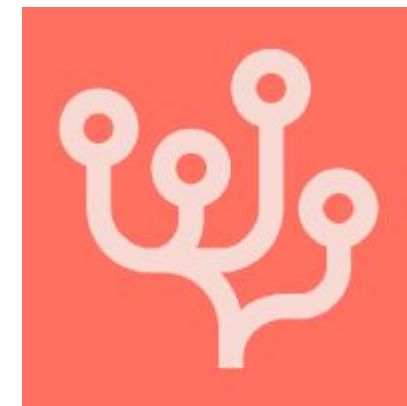
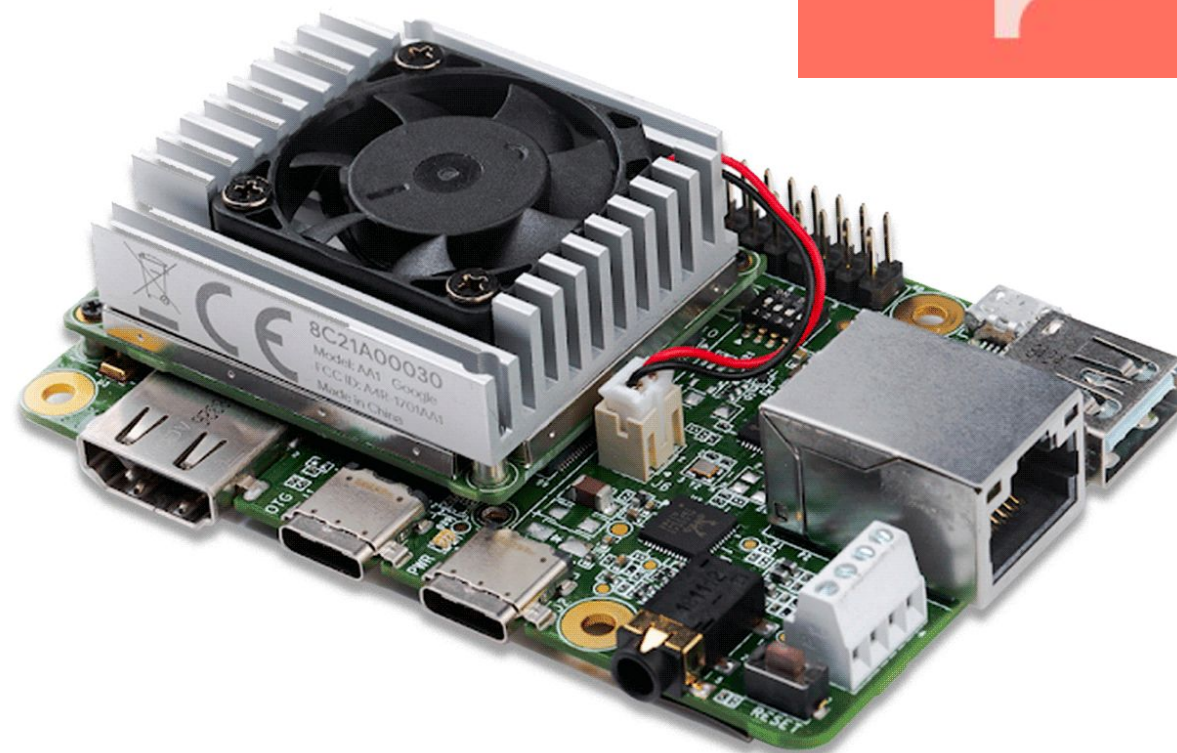


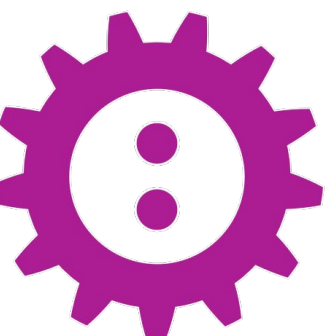
Further Work

Training at the Edge

Transfer Learning

Federated Learning





Summary

Build, Train, Optimize, Convert on laptop

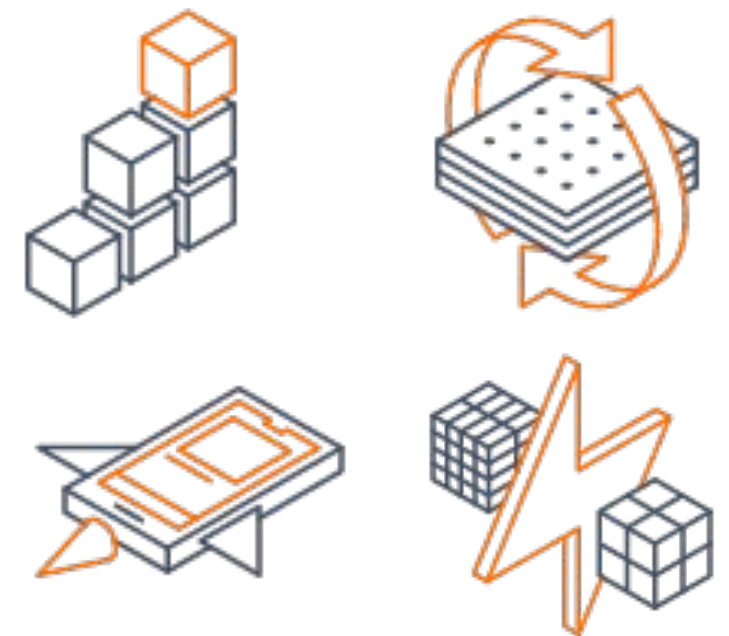
Deploy, Infer on device

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.

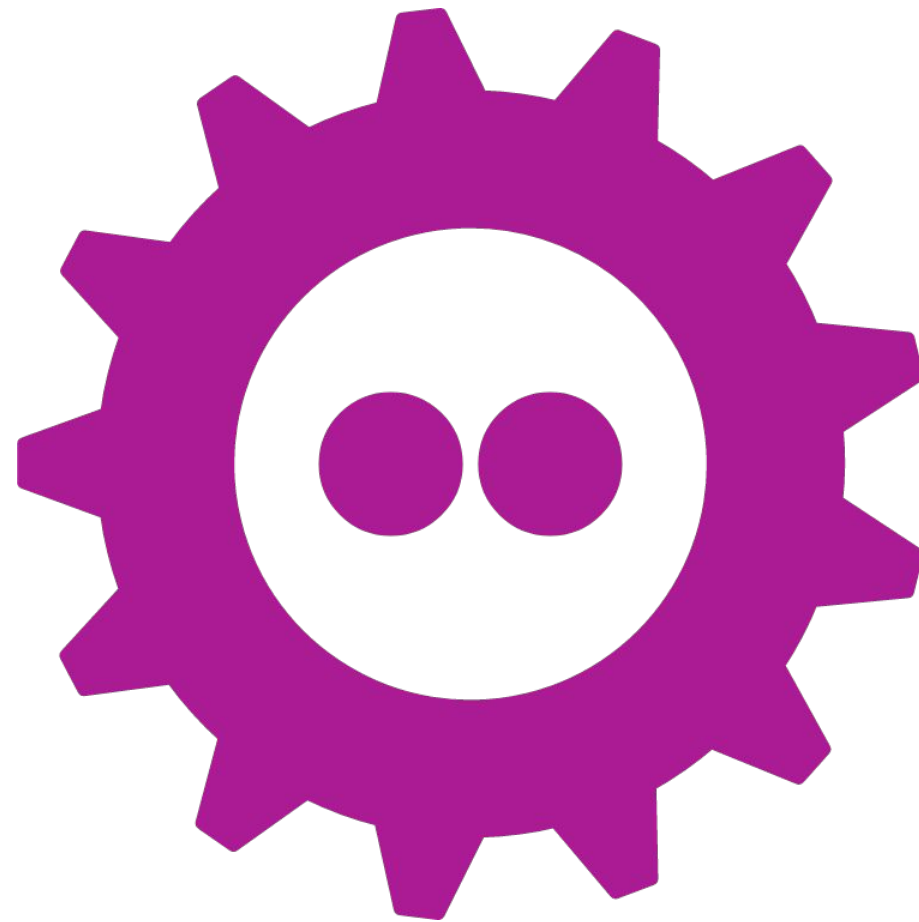


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Thanks!



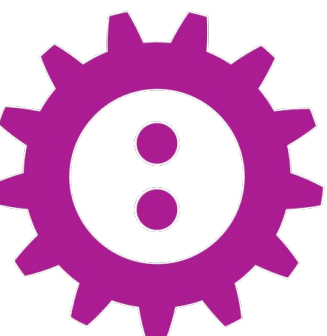
frama.link/rtone-iot-tflite



frama.link/rtone-jobs

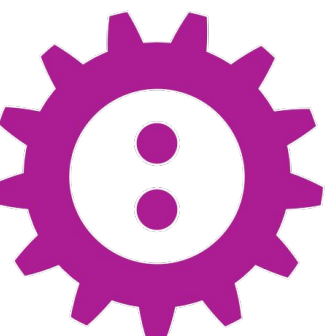


RTONE
IOT MAKERS



#FOSDEM #IoT #Tensorflow #AI

@alexis0duque



References

<https://medium.com/tensorflow/how-to-get-started-with-machine-learning-on-arduino-7daf95b4157>

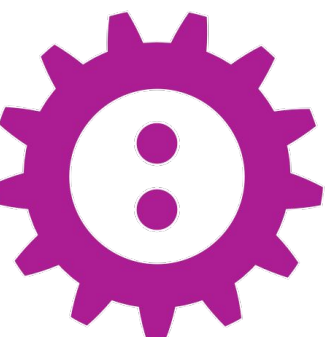
<https://www.tensorflow.org/lite>

<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>

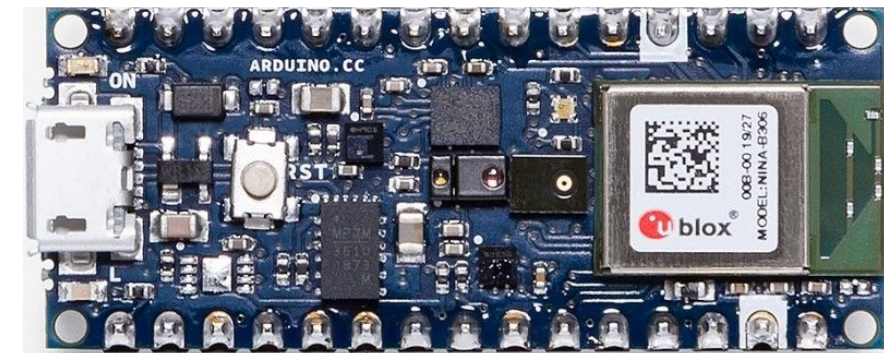


TFLite on Microcontrollers

Inference on Cortex-M microcontroller

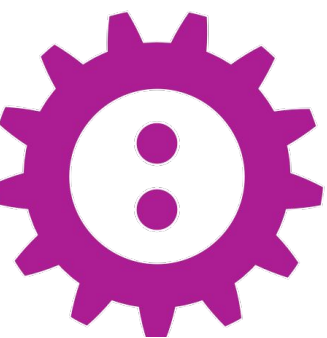
Only some operations are supported

Enough for hotword, gesture and speech recognition

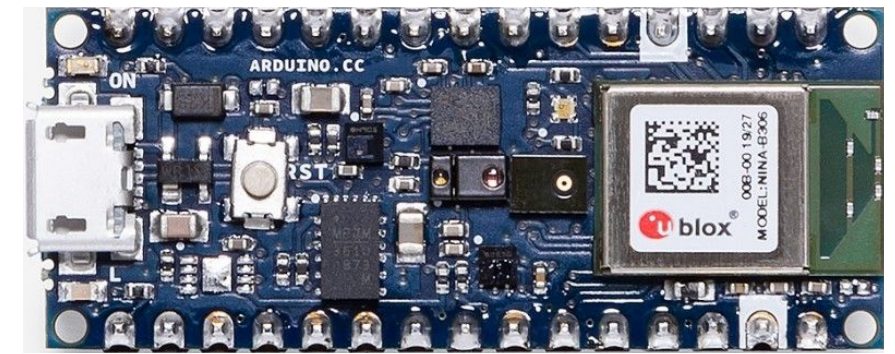


Arduino Nano 33





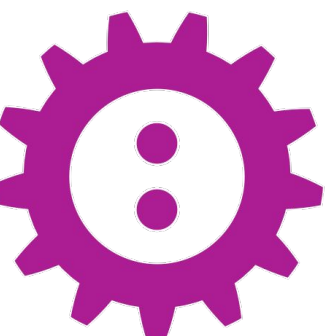
TFLite on Microcontrollers



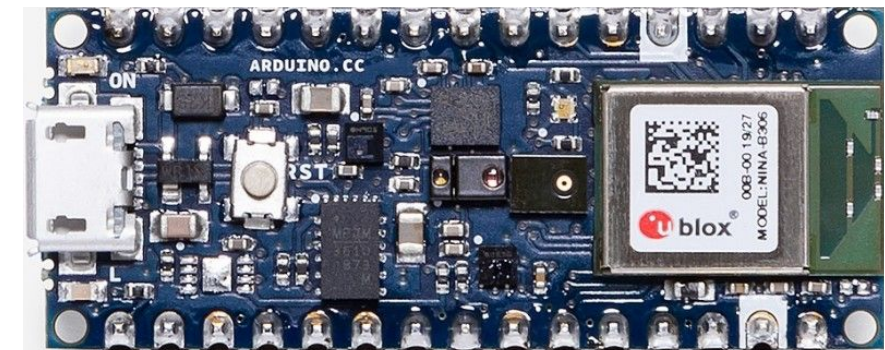
C++ API

```
#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;
```

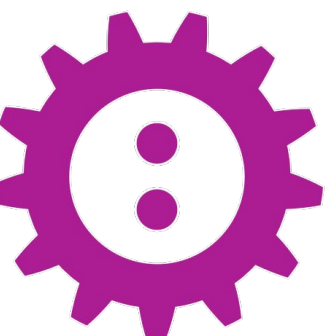
TFLite on Microcontrollers



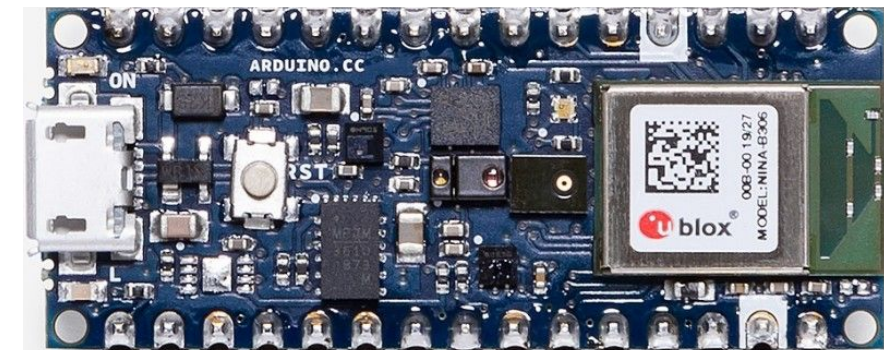
```
// Finding the min value for your model may require tests!
constexpr int kTensorArenaSize = 2 * 1024;
uint8_t tensor_arena[kTensorArenaSize];

// 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];
```