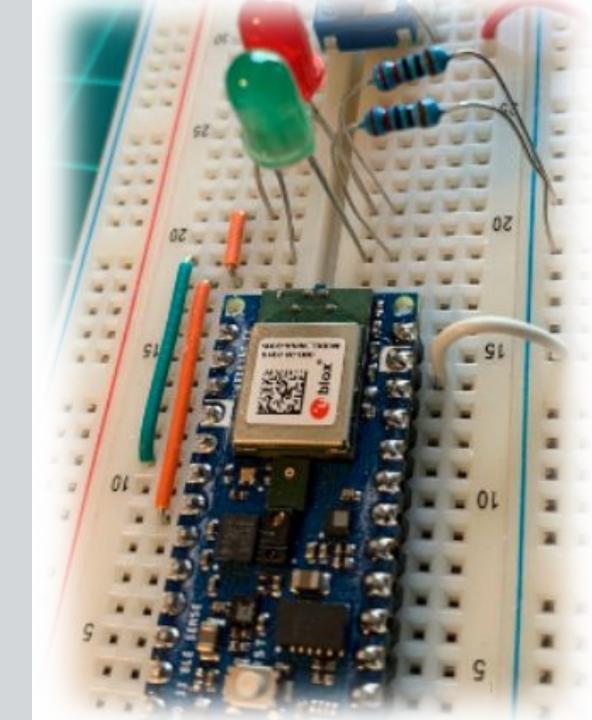
### IESTI01 - TinyML

Embedded Machine Learning

25a. Person Detection (VWW)
Application



Prof. Marcelo Rovai
UNIFEI



## Person Detection: Application Architecture

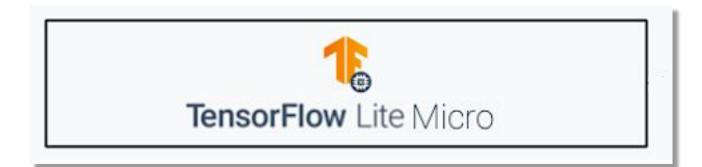




# Person Detection using Transfer Learning Model Code Walkthrough!

person\_detection.ino (Arduino IDE TFLite Example)





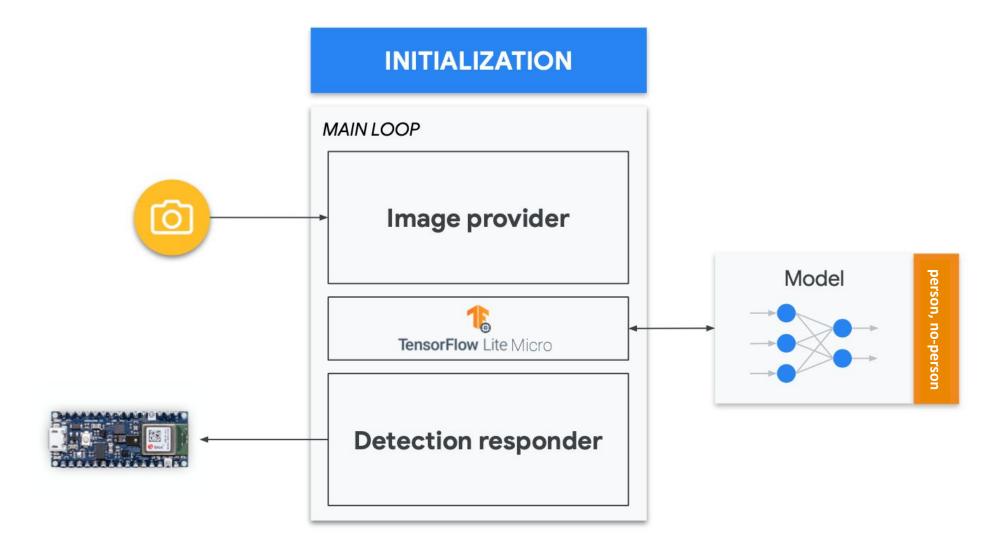
<u>TensorFlow Lite Micro - Paper</u>

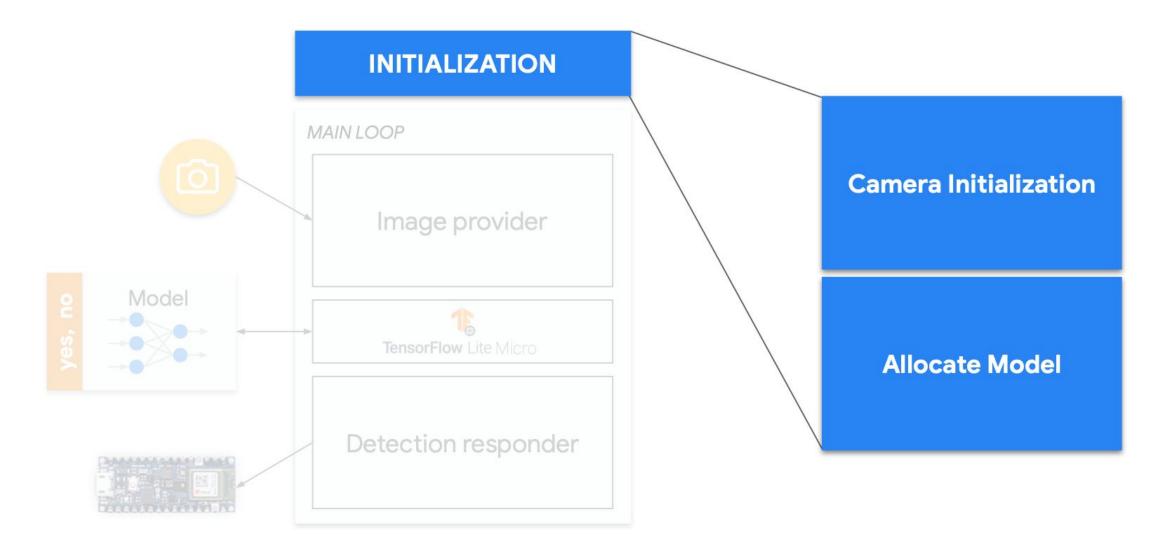


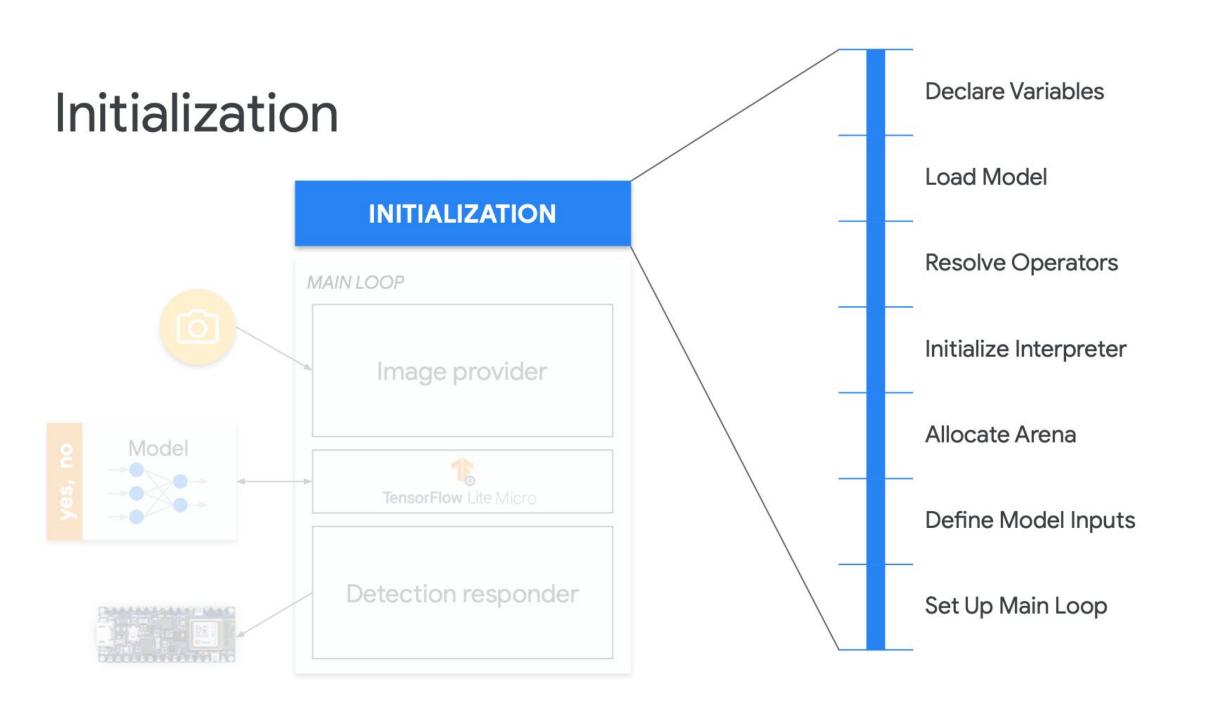
MLSys 2021: TensorFlow Lite Micro TFLM



#### Person Detection Components







```
person detection | Arduino 1.8.15
             arduino detection responder.cpp
person detection
                                   arduino image provider.cpp
                                                        arduino main.cpp
                                                                      detection responder.h
15
16 #include <TensorFlowLite.h>
17
18 #include "main_functions.h"
19
20 #include "detection_responder.h"
21 #include "image_provider.h"
22 #include "model_settings.h"
23 #include "person_detect_model_data.h"
24 #include "tensorflow/lite/micro/micro_error_reporter.h"
25 #include "tensorflow/lite/micro/micro_interpreter.h"
26 #include "tensorflow/lite/micro/micro_mutable_op_resolver.h"
27 #include "tensorflow/lite/schema/schema_generated.h"
28 #include "tensorflow/lite/version.h"
29
30 // Globals, used for compatibility with Arduino-style sketches.
31 □ namespace {
62 tflite::ErrorReporter* error_reporter = nullptr;
33 const tflite::Model* model = nullptr;
34 tflite::MicroInterpreter* interpreter = nullptr;
35 TfLiteTensor* input = nullptr;
37 // In order to use optimized tensorflow lite kernels, a signed int8_t quantized
38 // model is preferred over the legacy unsigned model format. This means that
39 // throughout this project, input images must be converted from unisgned to
40 // signed format. The easiest and quickest way to convert from unsigned to
41 // signed 8-bit integers is to subtract 128 from the unsigned value to get a
42 // signed value.
44 // An area of memory to use for input, output, and intermediate arrays
45 constexpr int kTensorArenaSize = 136 * 1024;
46 static uint8_t tensor_arena[kTensorArenaSize];
47 } // namespace
```

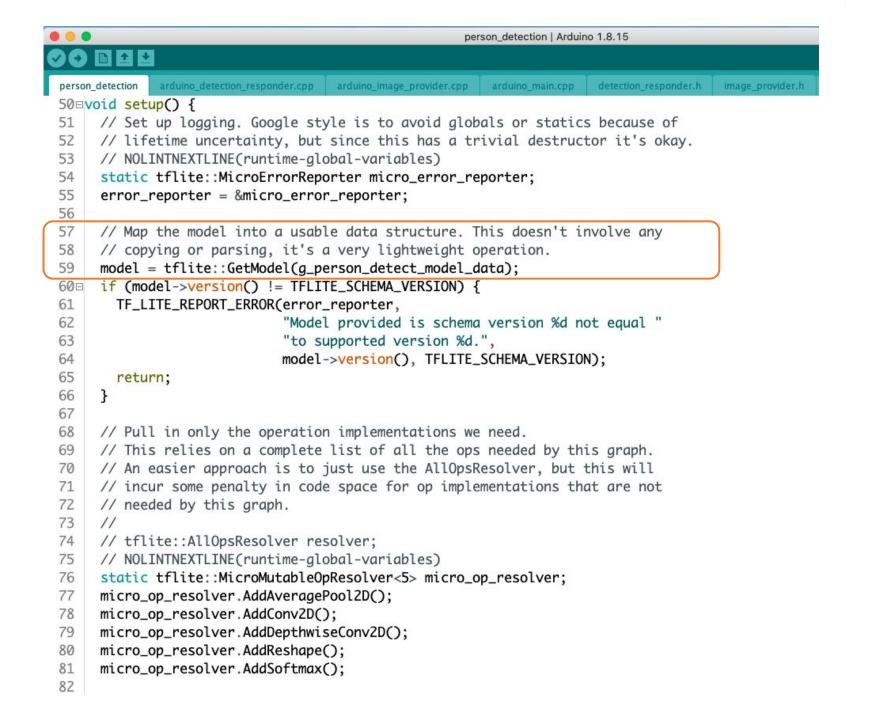
Load Model

Resolve Operators

Initialize Interpreter

Allocate Arena

Define Model Inputs



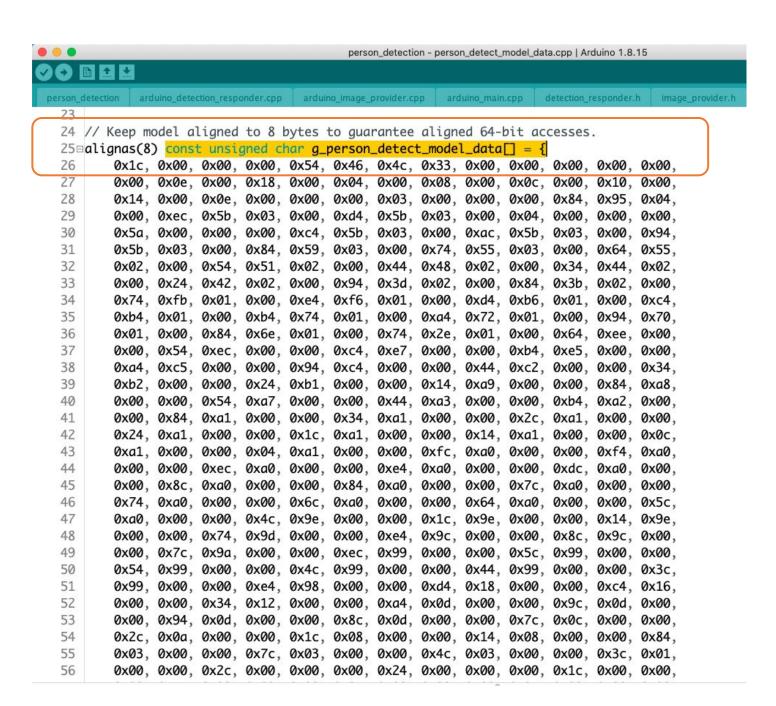
Load Model

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**Load Model** 

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Allocate Arena

Define Model Inputs

```
person_detection | Arduino 1.8.15
person detection
              arduino_detection_responder.cpp arduino_image_provider.cpp arduino_main.cpp detection_responder.h image_provider.h
    }
66
67
     // Pull in only the operation implementations we need.
     // This relies on a complete list of all the ops needed by this graph.
    // An easier approach is to just use the AllOpsResolver, but this will
     // incur some penalty in code space for op implementations that are not
     // needed by this graph.
73
     // tflite::AllOpsResolver resolver;
     // NOLINTNEXTLINE(runtime-global-variables)
     static tflite::MicroMutableOpResolver<5> micro_op_resolver;
     micro_op_resolver.AddAveragePool2D();
     micro_op_resolver.AddConv2D();
     micro_op_resolver.AddDepthwiseConv2D();
     micro_op_resolver.AddReshape();
81
     micro_op_resolver.AddSoftmax();
     // Build an interpreter to run the model with.
     // NOLINTNEXTLINE(runtime-global-variables)
     static tflite::MicroInterpreter static_interpreter(
          model, micro_op_resolver, tensor_arena, kTensorArenaSize, error_reporter);
86
      interpreter = &static_interpreter;
88
     // Allocate memory from the tensor_arena for the model's tensors.
     TfLiteStatus allocate_status = interpreter->AllocateTensors();
91 if (allocate_status != kTfLite0k) {
        TF_LITE_REPORT_ERROR(error_reporter, "AllocateTensors() failed");
92
93
        return;
94
      }
95
96
      // Get information about the memory area to use for the model's input.
      input = interpreter->input(0);
98 }
99
```

Load Model

**Resolve Operators** 

Initialize Interpreter

Allocate Arena

Define Model Inputs

Load Model

**Resolve Operators** 

Initialize Interpreter

Allocate Arena

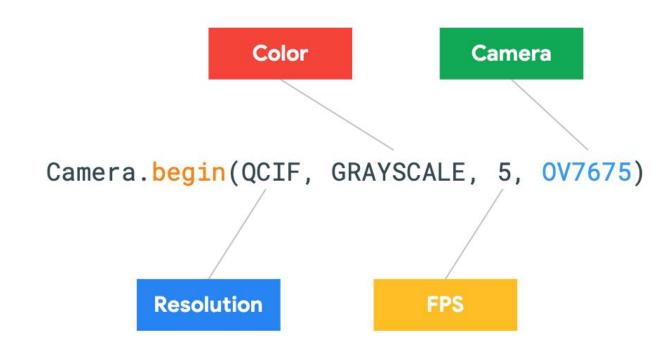
Define Model Inputs

Camera Initialization

Allocate Model

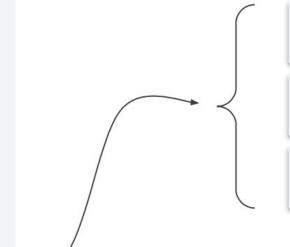
**Camera Initialization** 

Allocate Model



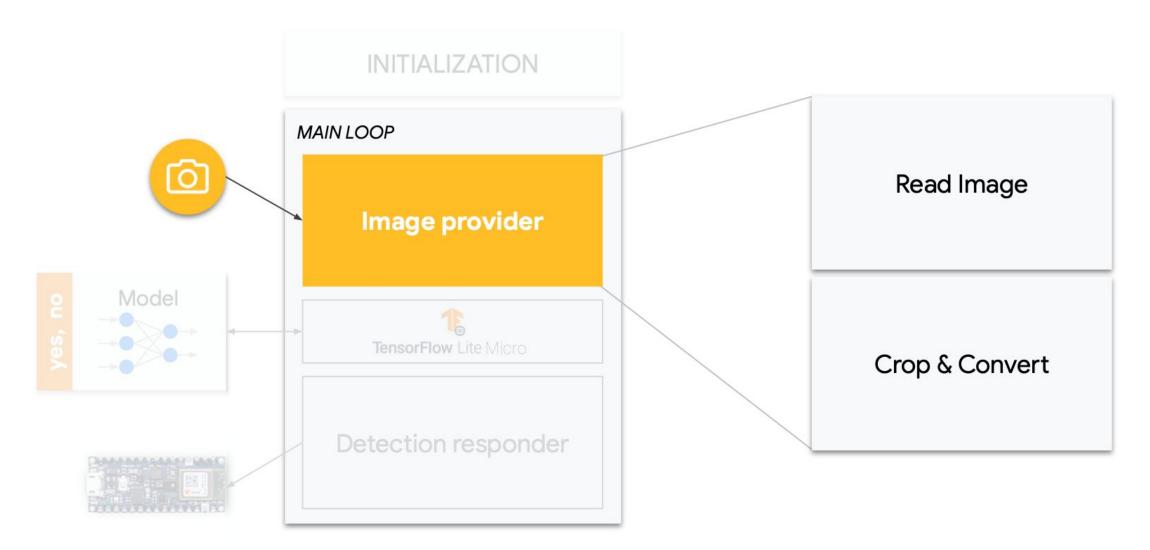
Camera Initialization

**Allocate Model** 



**Initialize Interpreter** 

**Define Model Inputs** 



**Read Image** 

**Crop & Convert** 





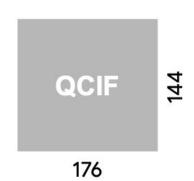


176

**Read Image** 

Crop & Convert



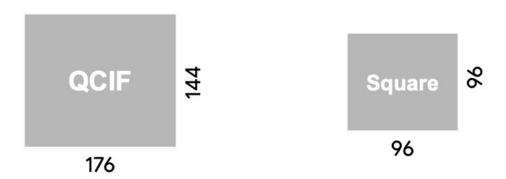




// Read camera data
Camera.readFrame(data);

Read Image

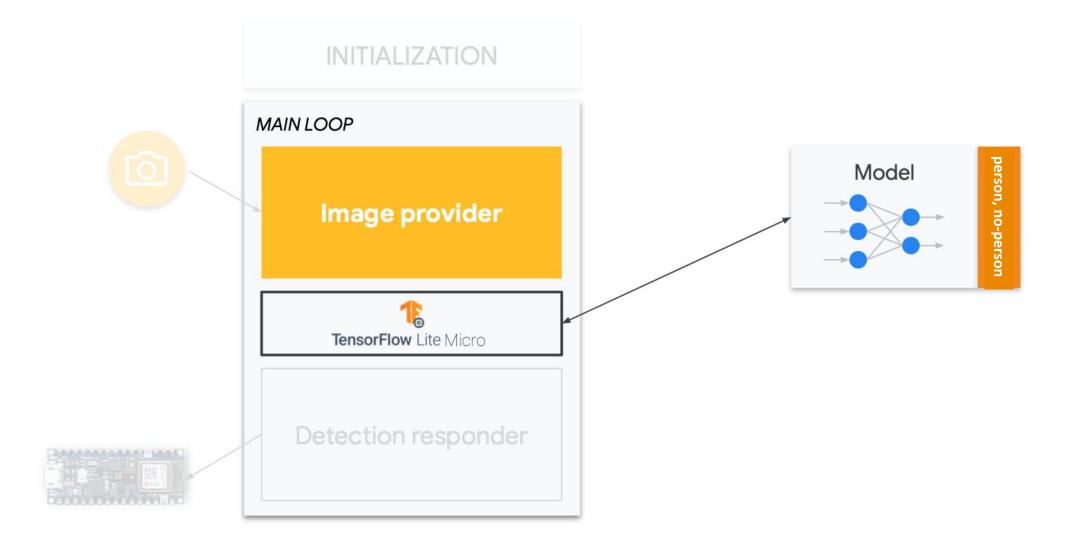
**Crop & Convert** 



```
int min_x = (176 - 96) / 2;
int min_y = (144 - 96) / 2;
int index = 0;

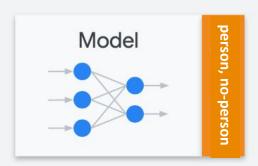
// Crop 96x96 image. This lowers FOV, ideally we should downsample
for (int y = min_y; y < min_y + 96; y++) {
   for (int x = min_x; x < min_x + 96; x++) {
      image_data[index++] = static_cast<int8_t>(data[(y * 176) + x] - 128);
      // convert TF input image to signed 8-bit
   }
}
```

### Interpreter + Model



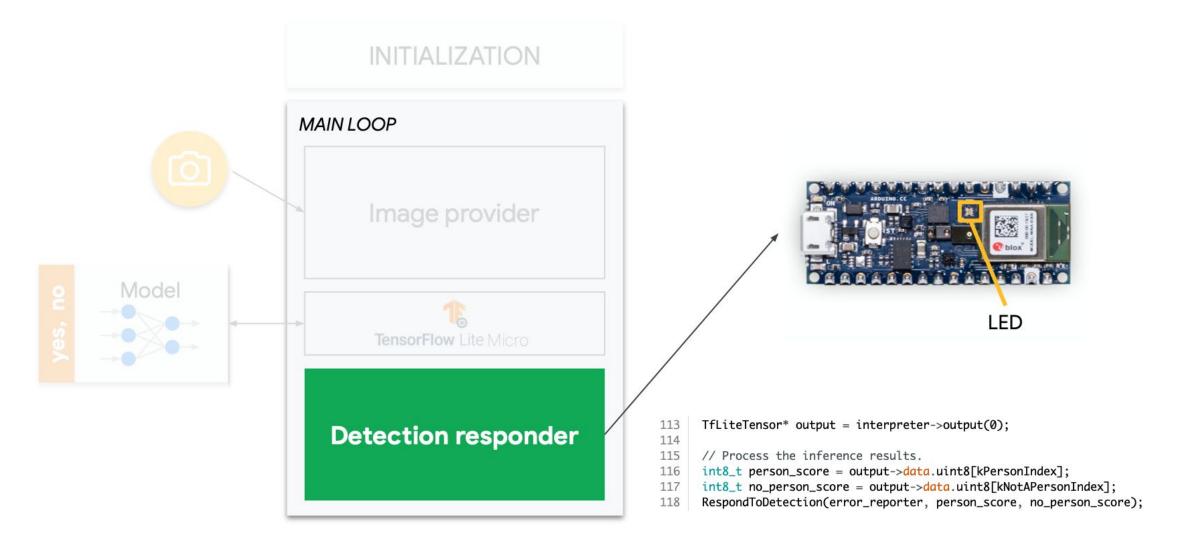
# Interpreter + Model





```
kTfLite0k != vww_interpreter->Invoke()
```

### Post-processing

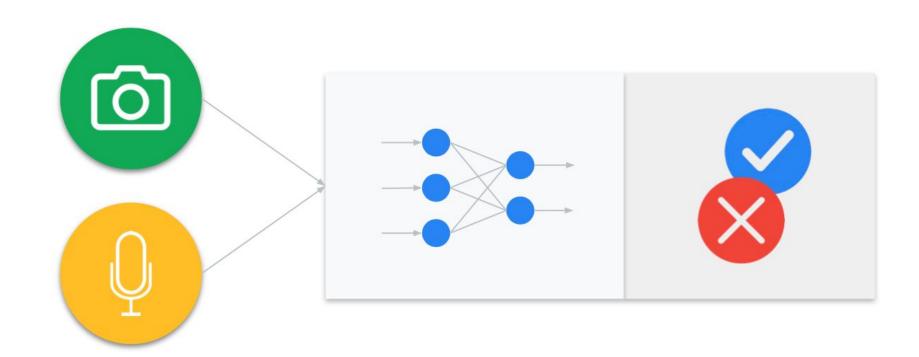


**Detection responder** 

```
if (person_score > no_person_score) {
    digitalWrite(LEDG, LOW);
    digitalWrite(LEDR, HIGH);
} else {
    digitalWrite(LEDG, HIGH);
    digitalWrite(LEDR, LOW);
}
```

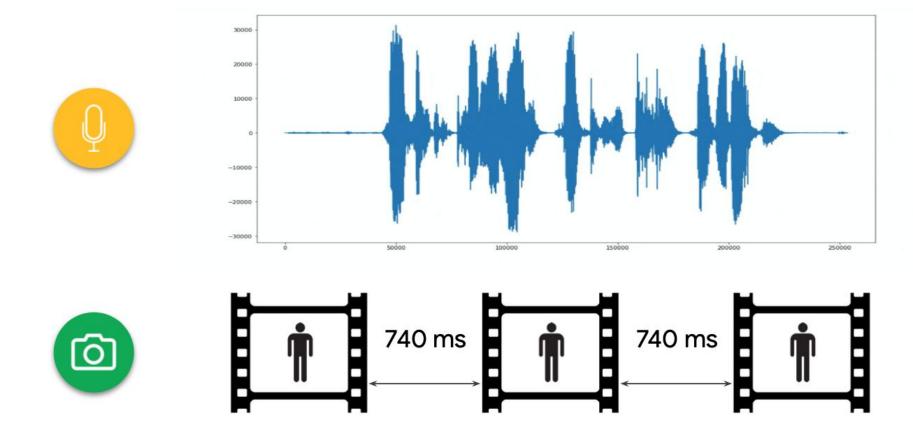
# Person Detection: Multi-Tenancy

#### MultiModal



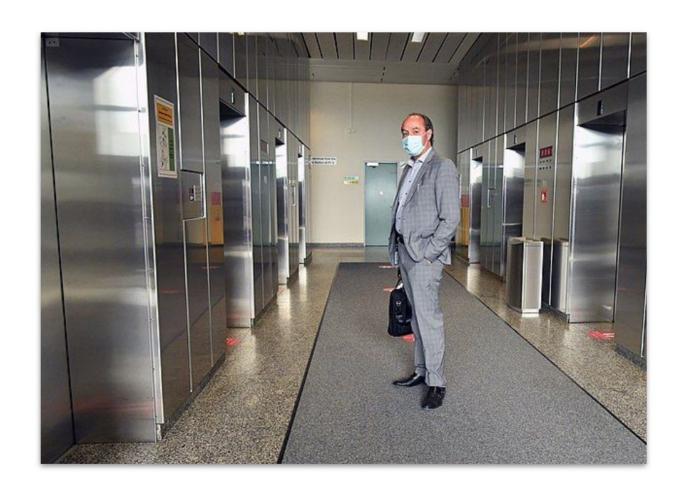
#### MultiModal ML Workflow



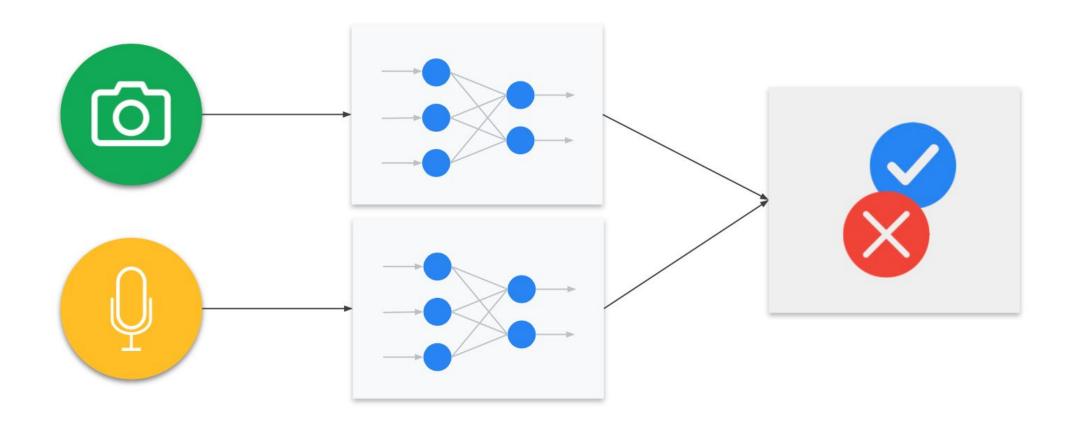


### **Example Person Detection Application**

- Contact-free elevator control that enforces mask wearing
- Requires both keyword spotting and mask detection



### MultiTenant



#### **MultiTenant** ML Workflow



same

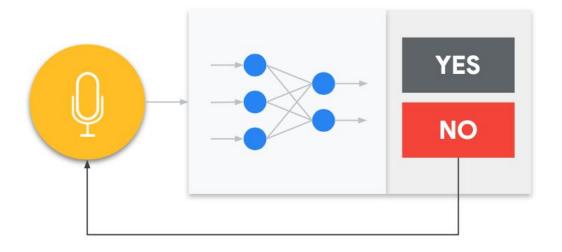




#### MultiTenant ML Workflow

Collect Preprocess Design a Model Train a Evaluate Convert Deploy Make Inferences

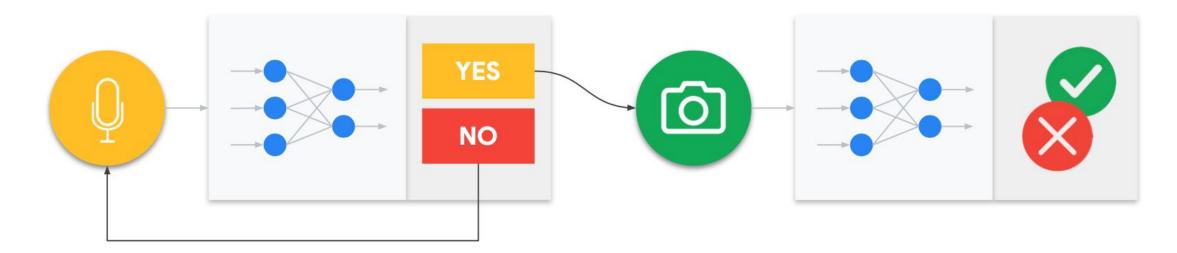
#### Cascade Multi Tenant



#### **MultiTenant** ML Workflow

Collect Preprocess Design a Model Train a Evaluate Convert Deploy Make Inferences

#### Cascade Multi Tenant



### Reading Material

#### 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: <u>"TinyML" by Pete Warden, Daniel Situnayake</u>
- Deploy textbook <u>"TinyML Cookbook" by Gian Marco Iodice</u>

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.

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## Thanks

