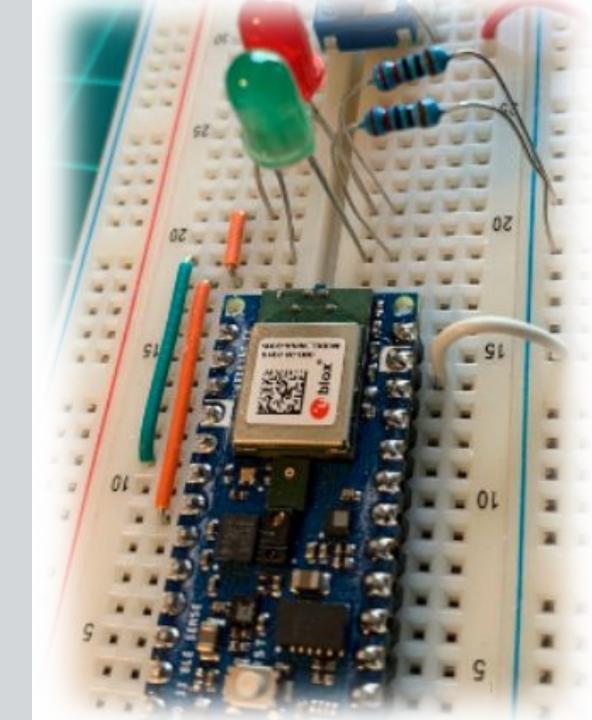
IESTI01 - TinyML

Embedded Machine Learning

26.a 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)





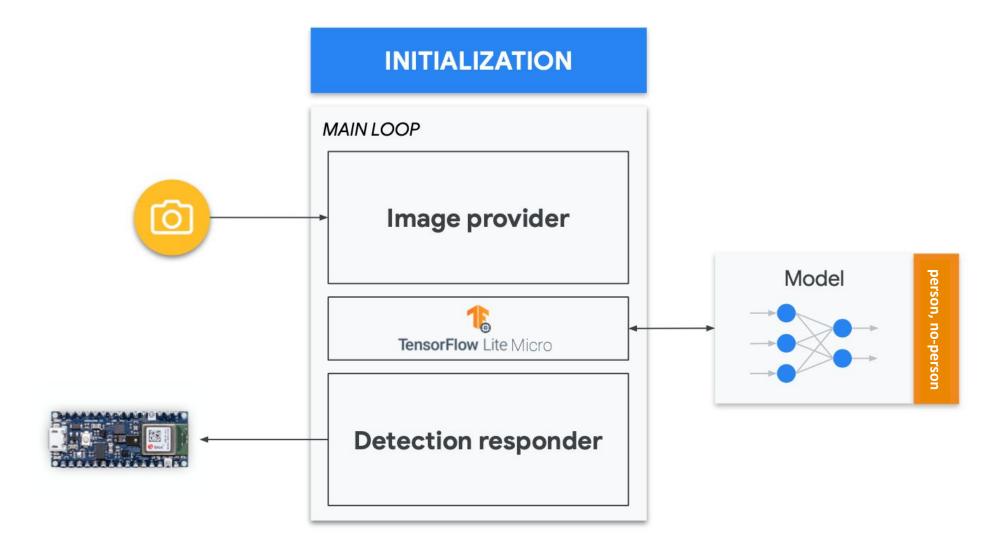
<u>TensorFlow Lite Micro - Paper</u>

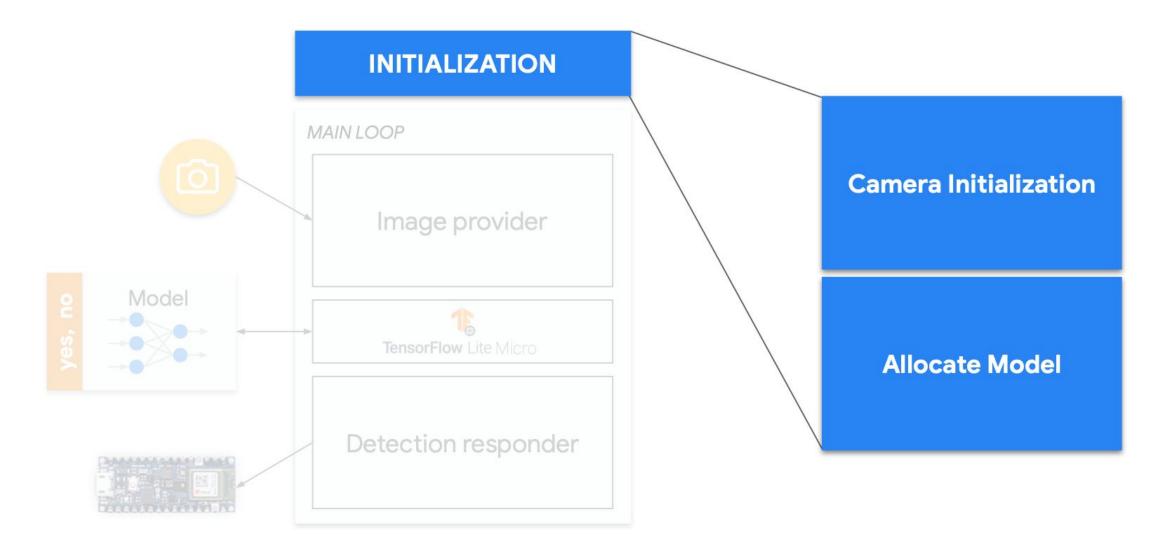


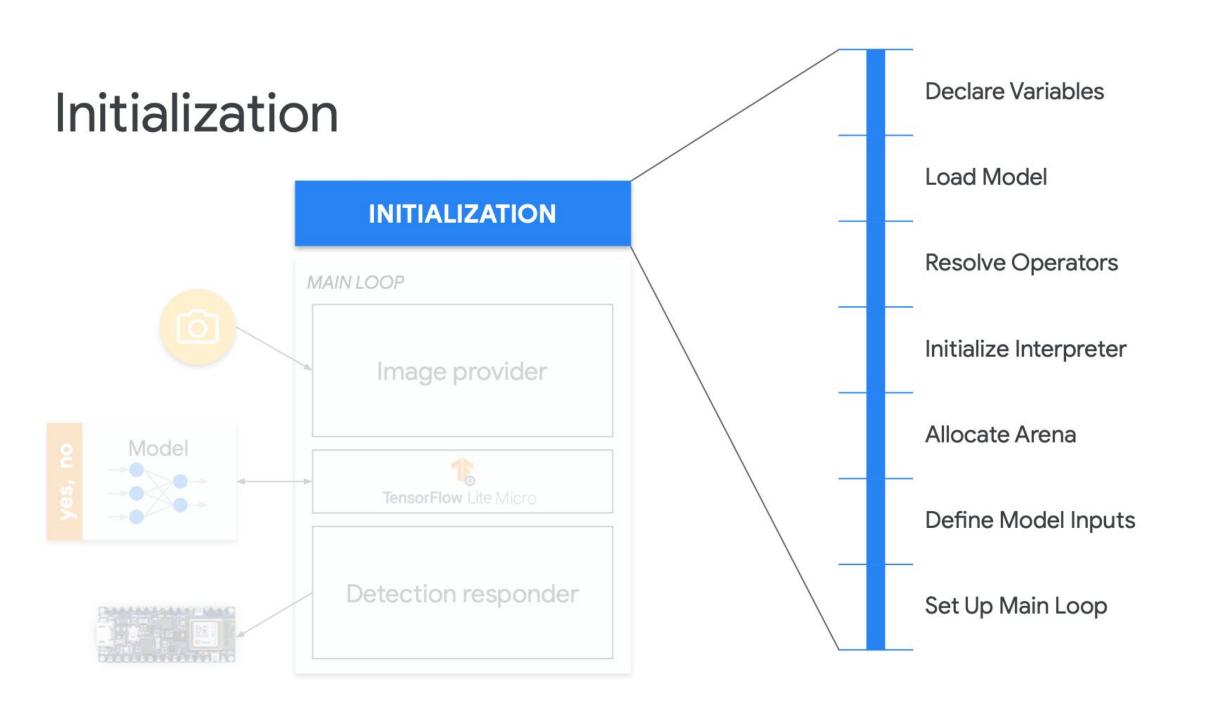
MLSys 2021: TensorFlow Lite Micro TFLM



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
```

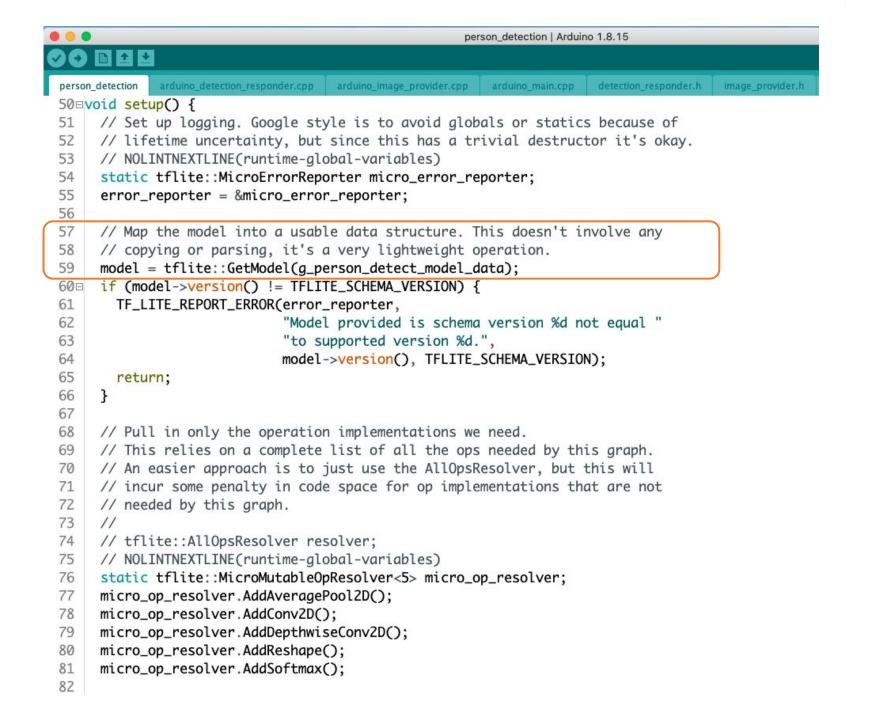
Load Model

Resolve Operators

Initialize Interpreter

Allocate Arena

Define Model Inputs



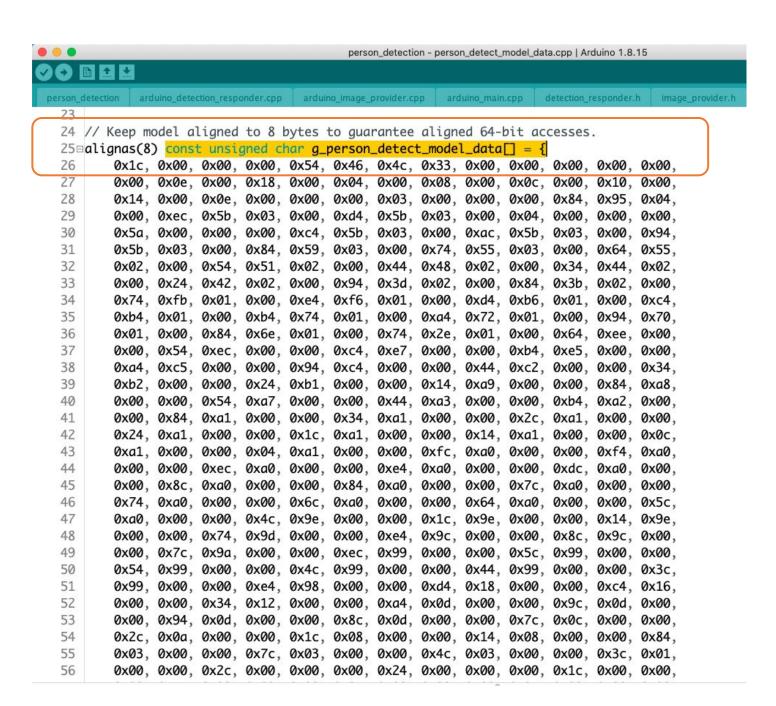
Load Model

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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

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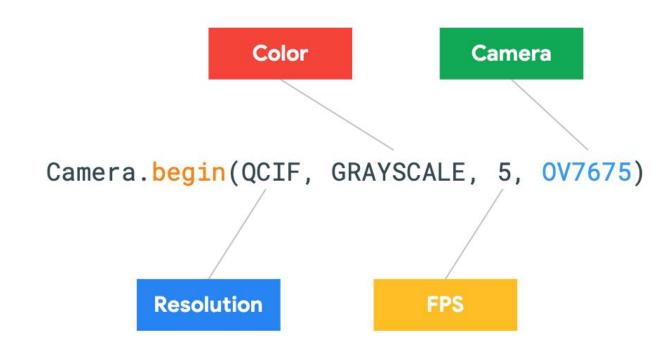
Define Model Inputs

Camera Initialization

Allocate Model

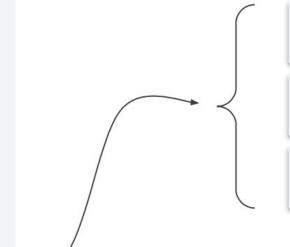
Camera Initialization

Allocate Model



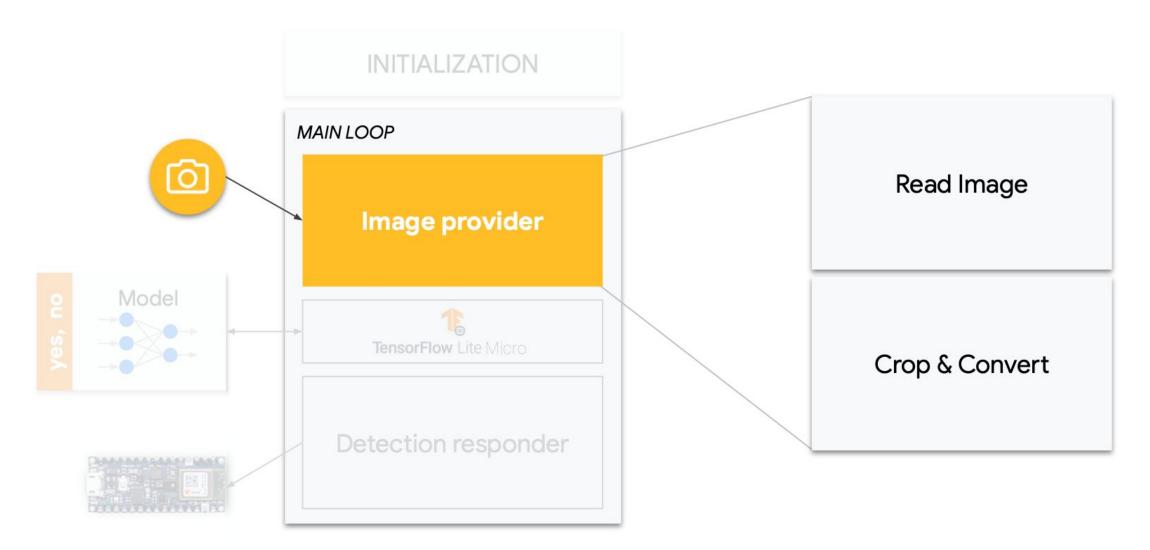
Camera Initialization

Allocate Model



Initialize Interpreter

Define Model Inputs



Read Image

Crop & Convert





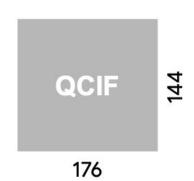


176

Read Image

Crop & Convert



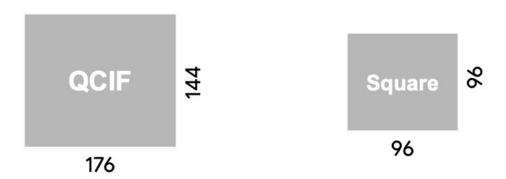




// Read camera data
Camera.readFrame(data);

Read Image

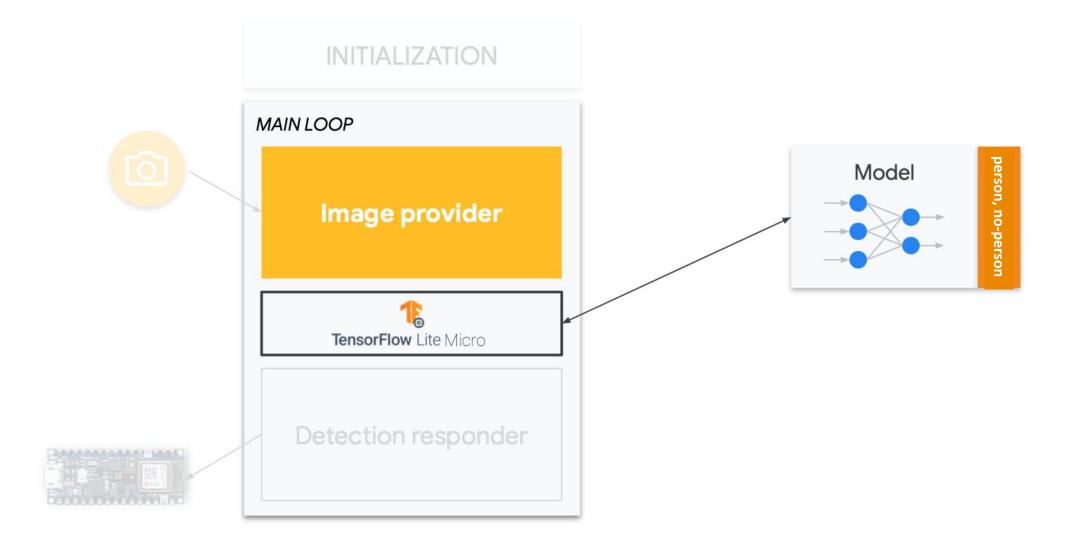
Crop & Convert



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

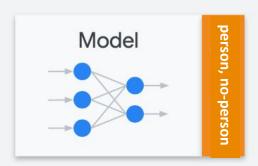
// Crop 96x96 image. This lowers FOV, ideally we should downsample
for (int y = min_y; y < min_y + 96; y++) {
   for (int x = min_x; x < min_x + 96; x++) {
      image_data[index++] = static_cast<int8_t>(data[(y * 176) + x] - 128);
      // convert TF input image to signed 8-bit
   }
}
```

Interpreter + Model



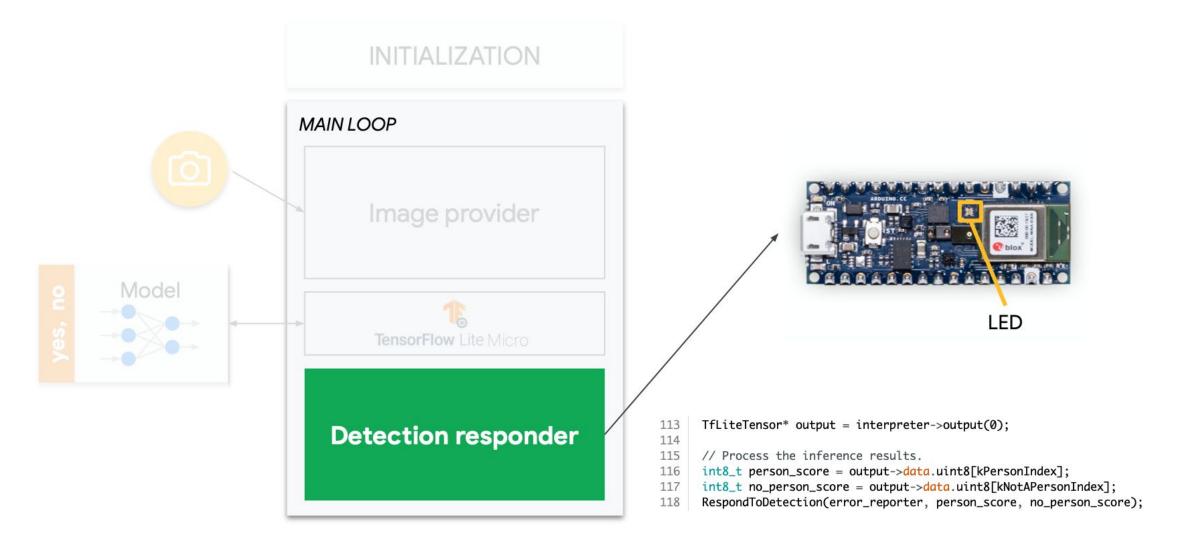
Interpreter + Model





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

Post-processing

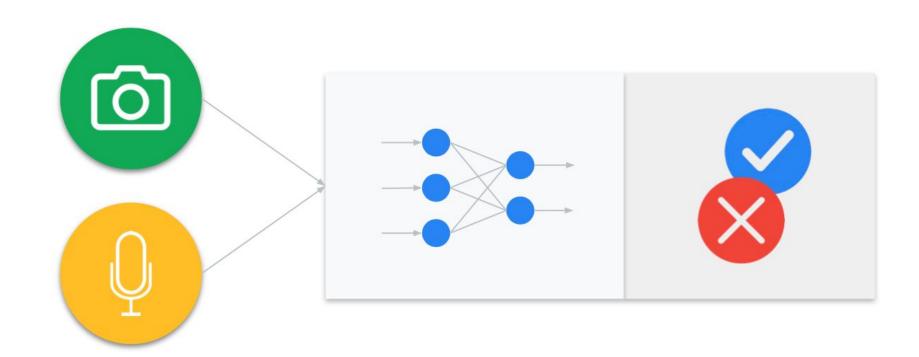


Detection responder

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

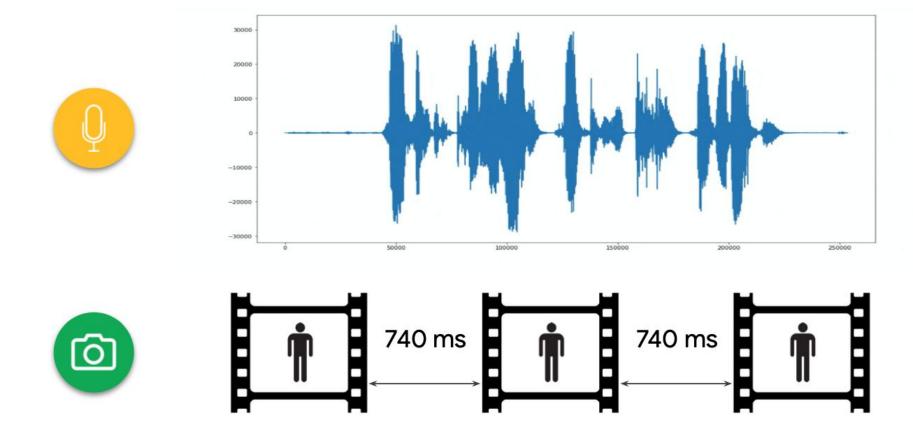
Person Detection: Multi-Tenancy

MultiModal



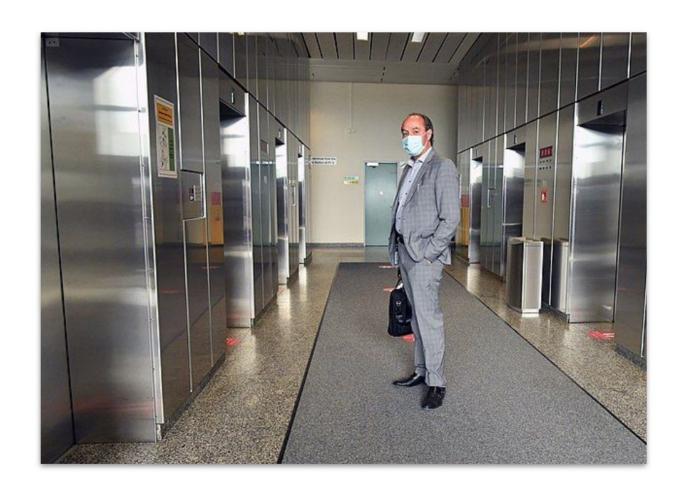
MultiModal ML Workflow



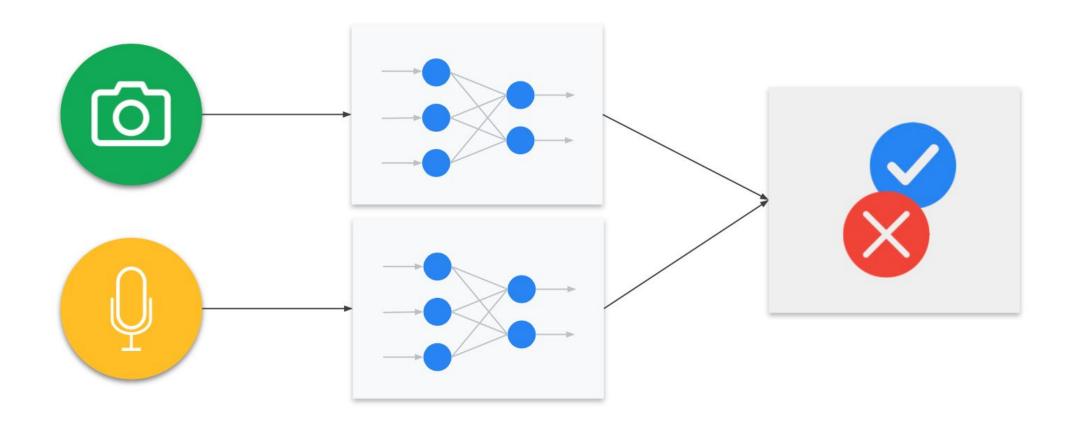


Example Person Detection Application

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



MultiTenant



MultiTenant ML Workflow



same

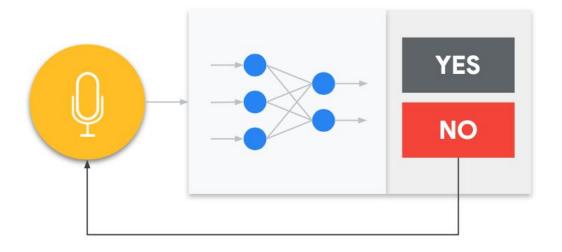




MultiTenant ML Workflow

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

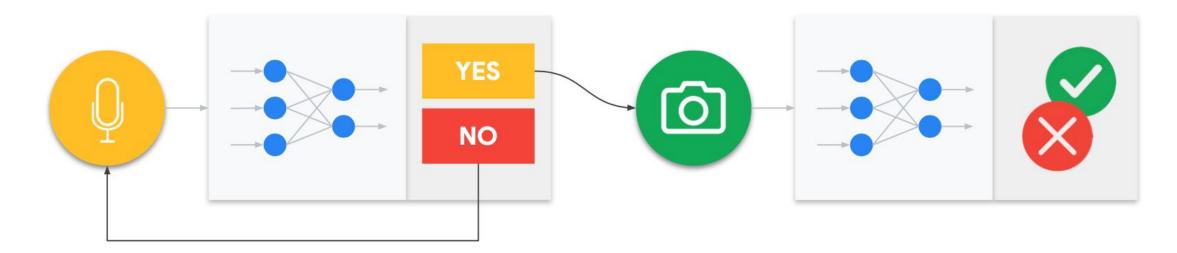
Cascade Multi Tenant



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)
- <u>Text Book: "TinyML" by Pete Warden, Daniel Situnayake</u>

I want to thank <u>Shawn Hymel</u> and Edge Impulse, <u>Pete Warden</u> and <u>Laurence</u> <u>Moroney</u> from Google, and especially Harvard professor <u>Vijay Janapa Reddi</u>, Ph.D. student <u>Brian Plancher</u> and their staff for preparing the excellent material on TinyML that is the basis of this course at UNIFEI.

The IESTI01 course is part of the <u>TinyML4D</u>, an initiative to make TinyML education available to everyone globally.

Thanks And stay safe!

