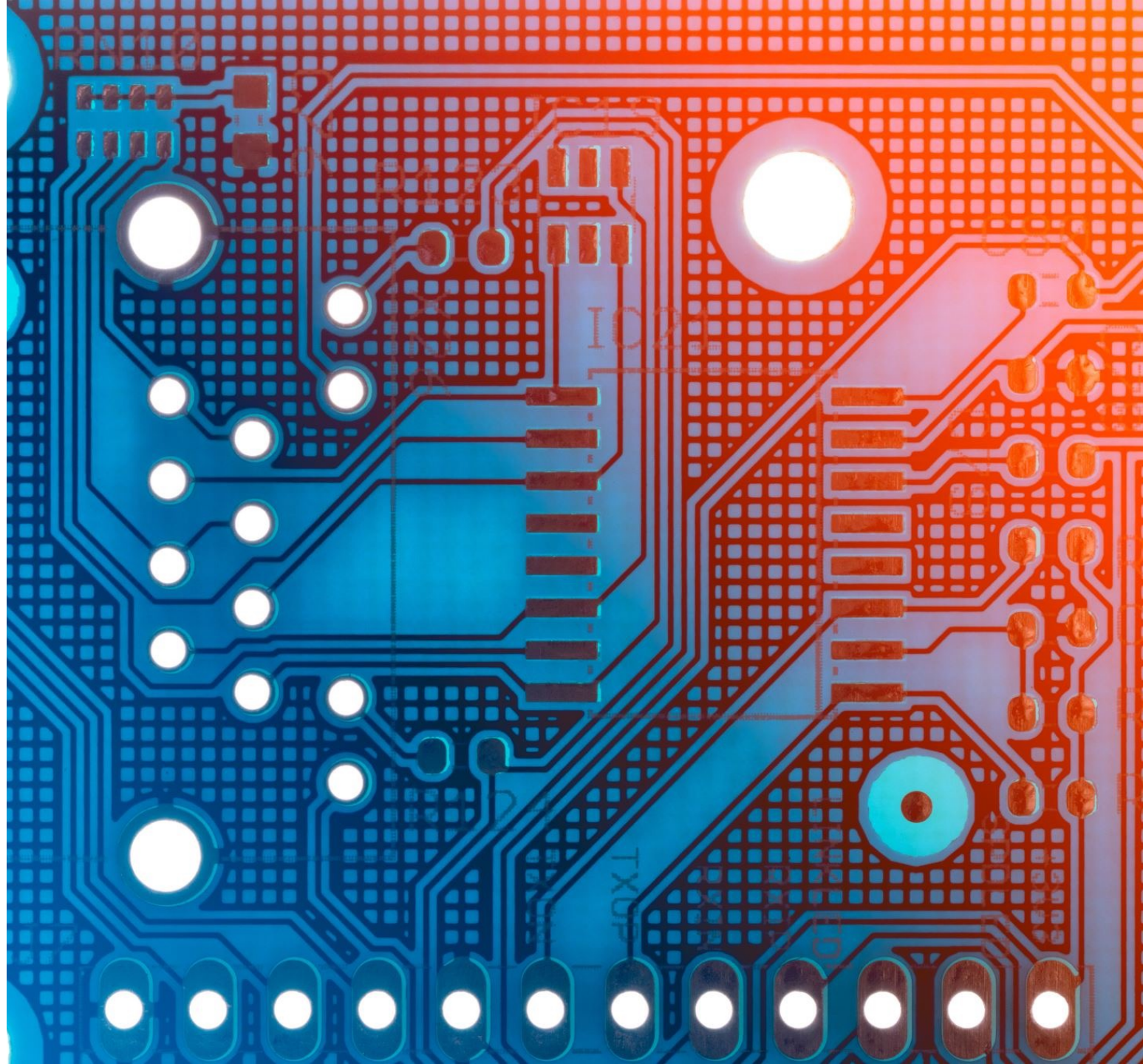




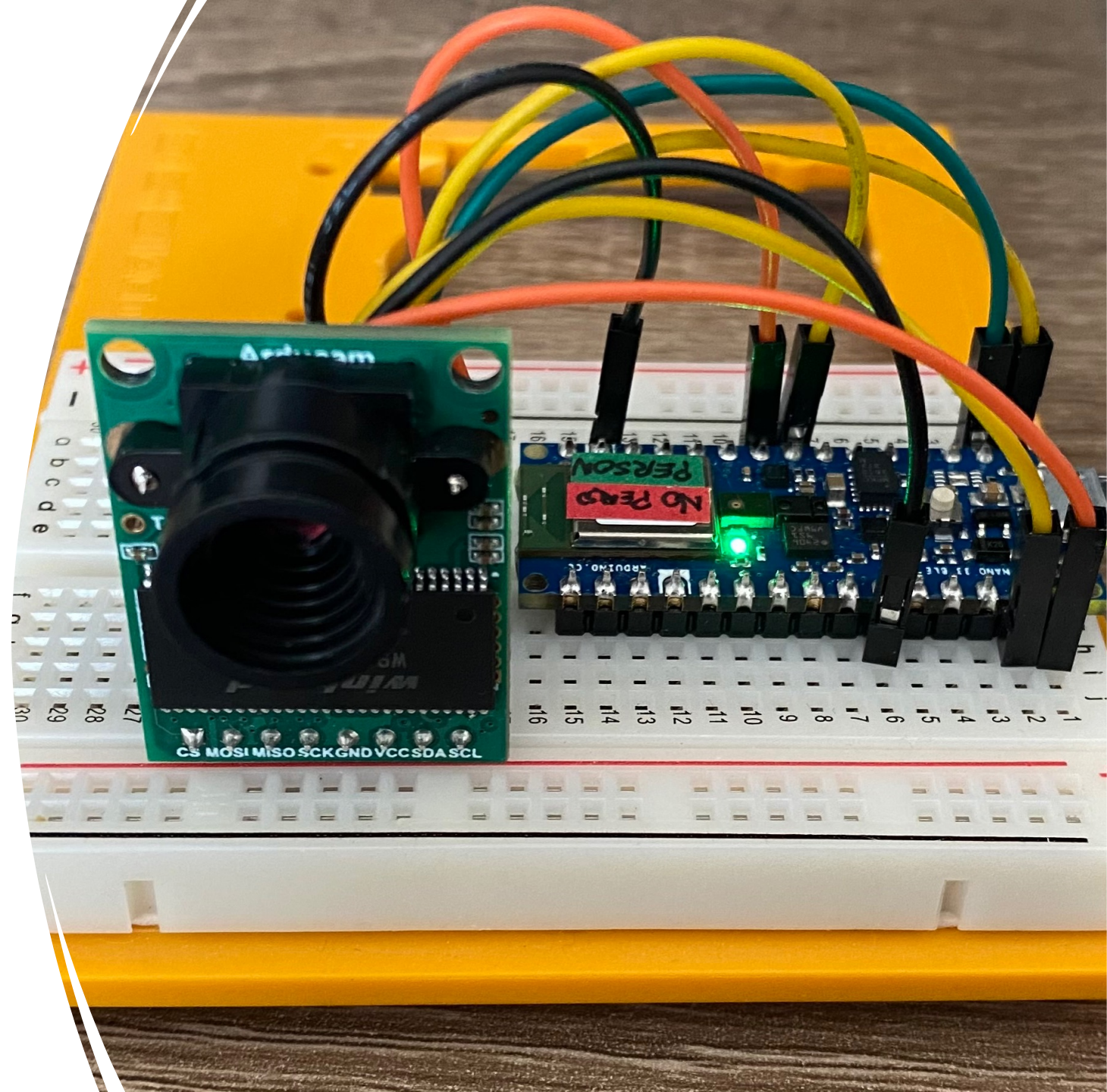
Embedded Machine Learning for Person Detection

Presented by: Ana Ferraz



Contents

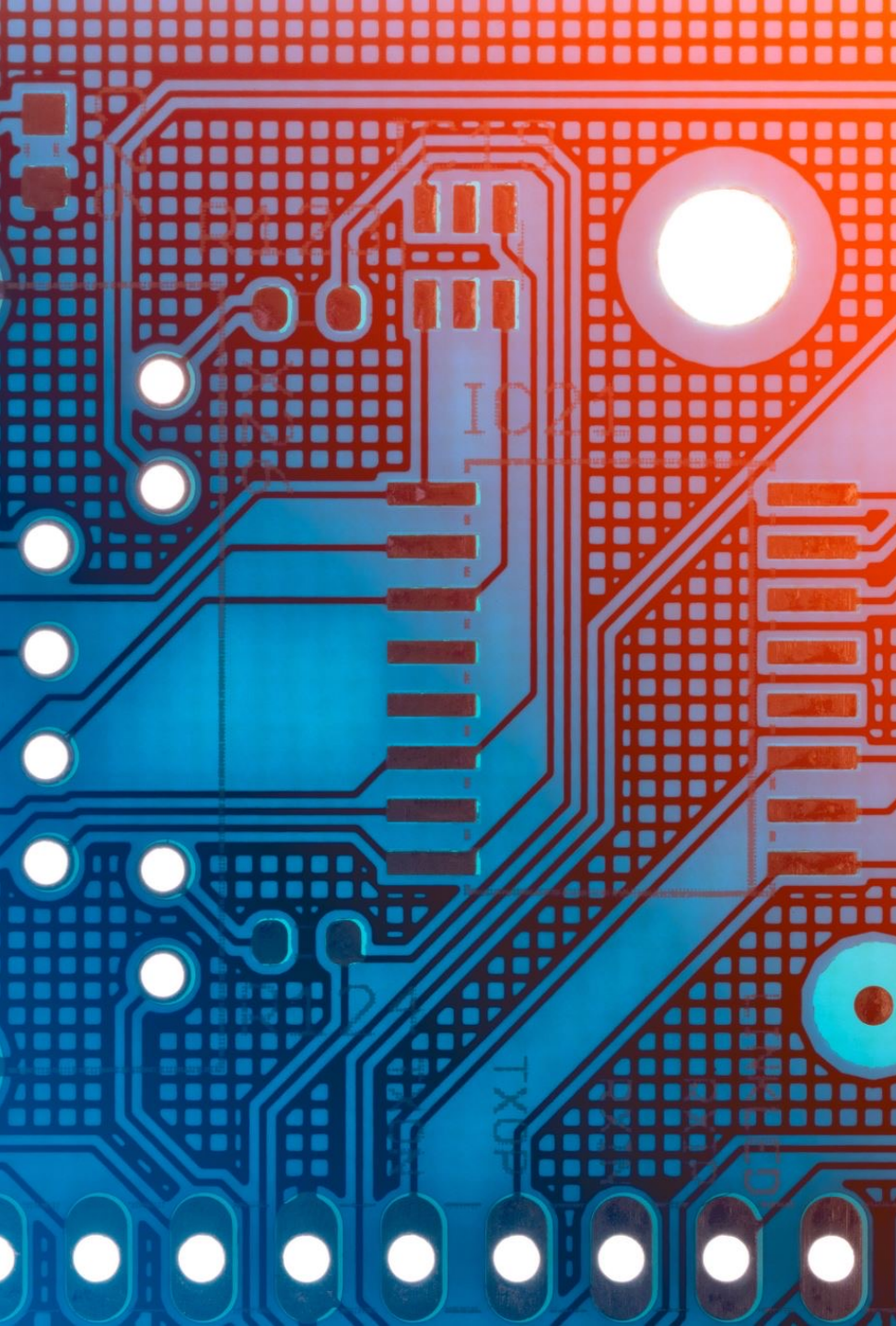
- Introduction
- Workflow
- Hardware used
- Visual Wake Words Dataset
- Application Architecture
- Person Detection
- Training a Model
- Conclusion and future work



Introduction

- TinyML – Review previous seminar
- ML at the embedded edge devices
- Embedded devices have serious constraints
- Various sensors built-in or connected
- Recent - new field





Deep Learning Workflow

- Decide the goal
- Collect a Dataset
- Design a Model Architecture
- Train the model
- Convert the Model
- Run Inference
- Evaluate and Troubleshoot

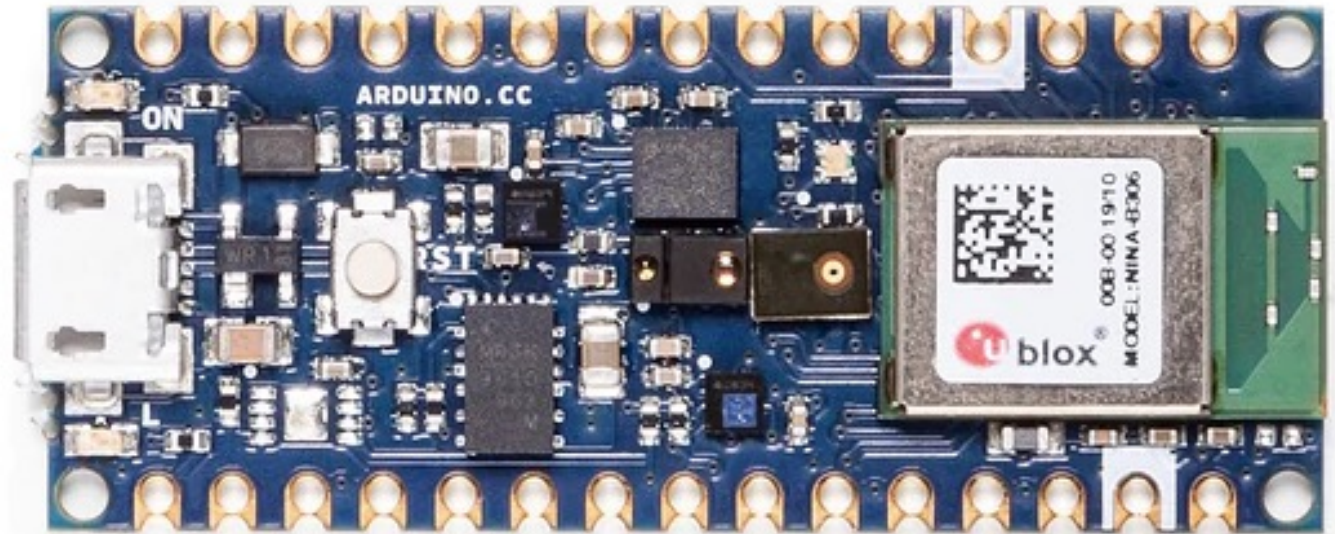


Hardware (previous presentation)

- Apollo3 ([Apollo3, 2021](#)),
- STM32F Discovery ([STM32F, 2021](#)),
- ST IoT Discovery ([ST IoT Discovery, 2021](#)),
- ECM3532 AI Sensor Neuro sensor processor (NSP) ([ECM3532, 2021](#)),
- Arduino Nano 33 BLE Sense ([Arduino Nano 33, 2021](#)),
- OpenMV Cam H7 Plus ([OpenMV, 2021](#)),
- Himax EW-I Plus ([Himax, 2021](#)),
- Thunderboard Sense 2 ([Thunderboard Sense 2, 2021](#)),
- Sony's Spresense TinyML Board ([Sony's Spresense TinyML Board, 2021](#)),
- Arduino Portenta H7 ([Arduino Portenta H7, 2021](#)),
- Raspberry Pi 4B ([Raspberry Pi 4B, 2021](#)),
- Nvidia Jetson Nano ([Nvidia Jetson Nano, 2021](#)),
- CC1352P Launchpad ([CC1352P Launchpad, 2021](#)),
- ESP-EYE ([ESP-EYE, 2021](#)),
- GAP8 ([GAP8, 2021](#)),
- GAP9 ([GAP9, 2021](#)),
- AI-deck 1.1 ([AI-deck 1.1, 2021](#)),
- Seeed Wio Terminal ([Seeed Wio Terminal, 2021](#)),
- Agora Product Development Kit ([Agora Product Development Kit, 2021](#)),
- Pico4ML BLE ([Pico4ML BLE, 2021](#)),
- MKR Video 4000 ([MKR Video 4000, 2021](#)),
- Nicla Sense ME ([Nicla Sense ME, 2021](#)),
- Nordic Semi nRF52840 DK ([Nordic Semi nRF52840 DK, 2021](#)),
- Nordic Semi Thingy:91 ([Nordic Semi Thingy:91, 2021](#)),
- XCore.ai ([XCore.ai, 2021](#)),
- FRDM-K64F ([FRDM-K64F, 2021](#)).

Hardware Board

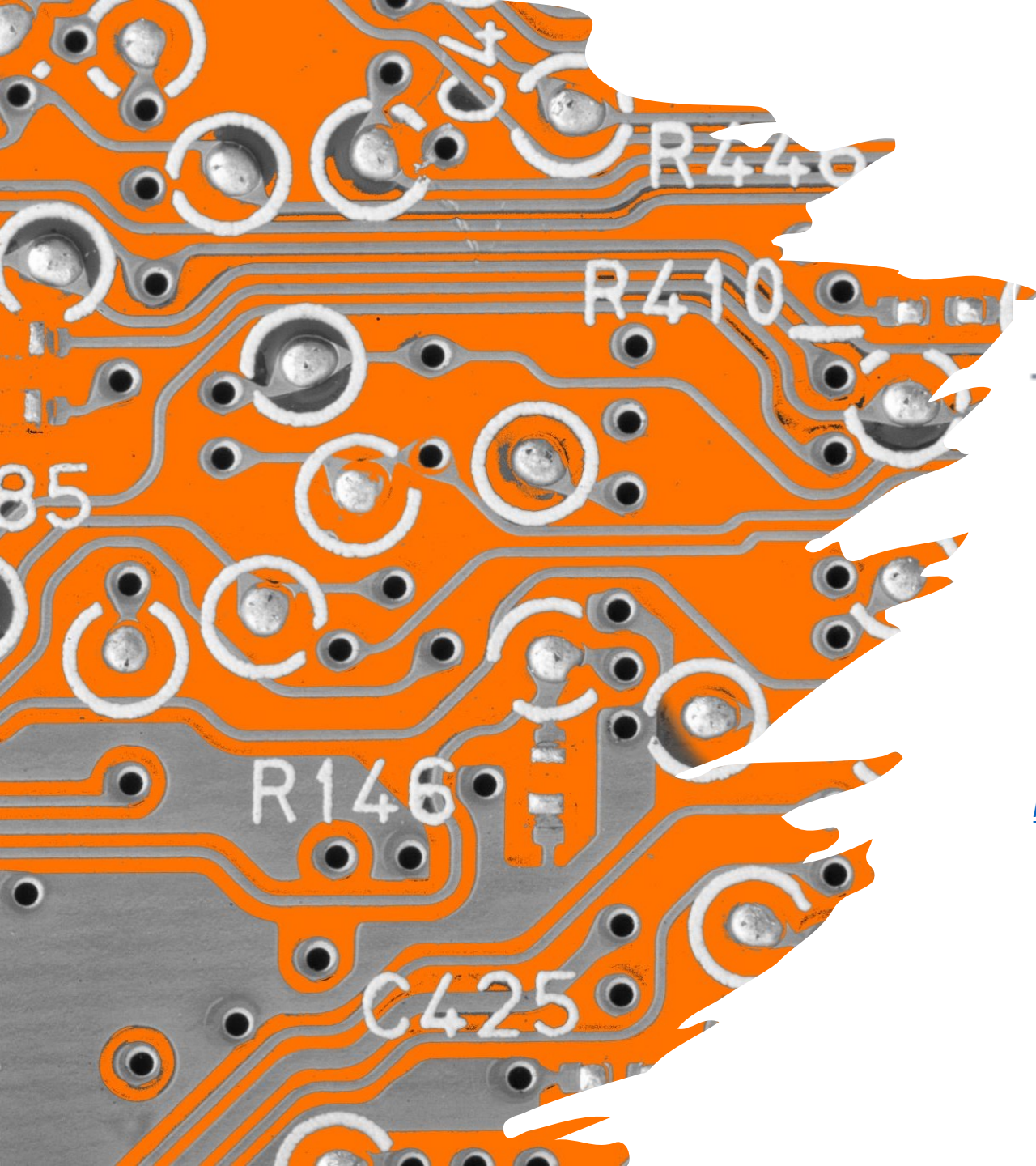
- Arduino Nano 33 BLE Sense
 - 9 axis inertial sensor
 - Humidity and Temperature
 - Barometric
 - Microphone
 - Gesture
 - Proximity, light color, intensity
- 32-bit ARM® Cortex®-M4 CPU
- 64MHz
- 1MB program memory
- SRAM 256KB



Hardware Camera

- Can be used in Arduino, Raspberry Pi, etc.
- 2 megapixels image
- SPI interface for the sensor configuration
- Output format: FAW, YUV, RGB, JPEG





Software and Libraries (previous presentation)



[TensorFlow Lite \(TFL\)](#)



[uTensor](#)



[Edge Impulse](#)



[NanoEdge AI Studio](#)



[PyTorch Mobile](#)



[Embedded Learning Library \(ELL\)](#)



[STM32Cube.AI](#)



[µTVM: MicroTVM](#)

Software Used

- Google Cloud Platform
- TensorFlow Lite – Training and Conversion
- Arduino – upload to hardware

Google
Cloud
Platform

Computer application



TensorFlow Lite



Dataset Used

- Visual Wake Words [3]
- Re-labeling COCO dataset
 - Label 1 – has at least one object bounding box
 - Label 2 – doesn't have the object bounding box
- Small bounding boxes (<0.5%) excluded



(a) 'Person'



(b) 'Not-person'

What is COCO?



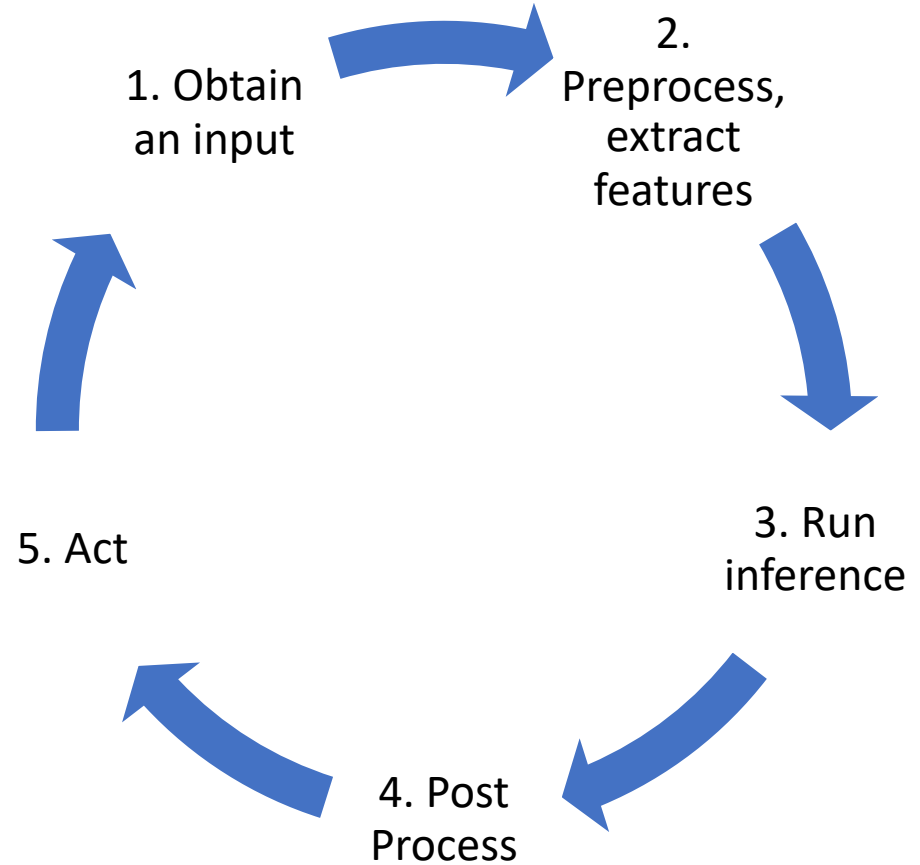
COCO is a large-scale object detection, segmentation, and captioning dataset. COCO has several features:

- ✓ Object segmentation
- ✓ Recognition in context
- ✓ Superpixel stuff segmentation
- ✓ 330K images (>200K labeled)
- ✓ 1.5 million object instances
- ✓ 80 object categories
- ✓ 91 stuff categories
- ✓ 5 captions per image
- ✓ 250,000 people with keypoints

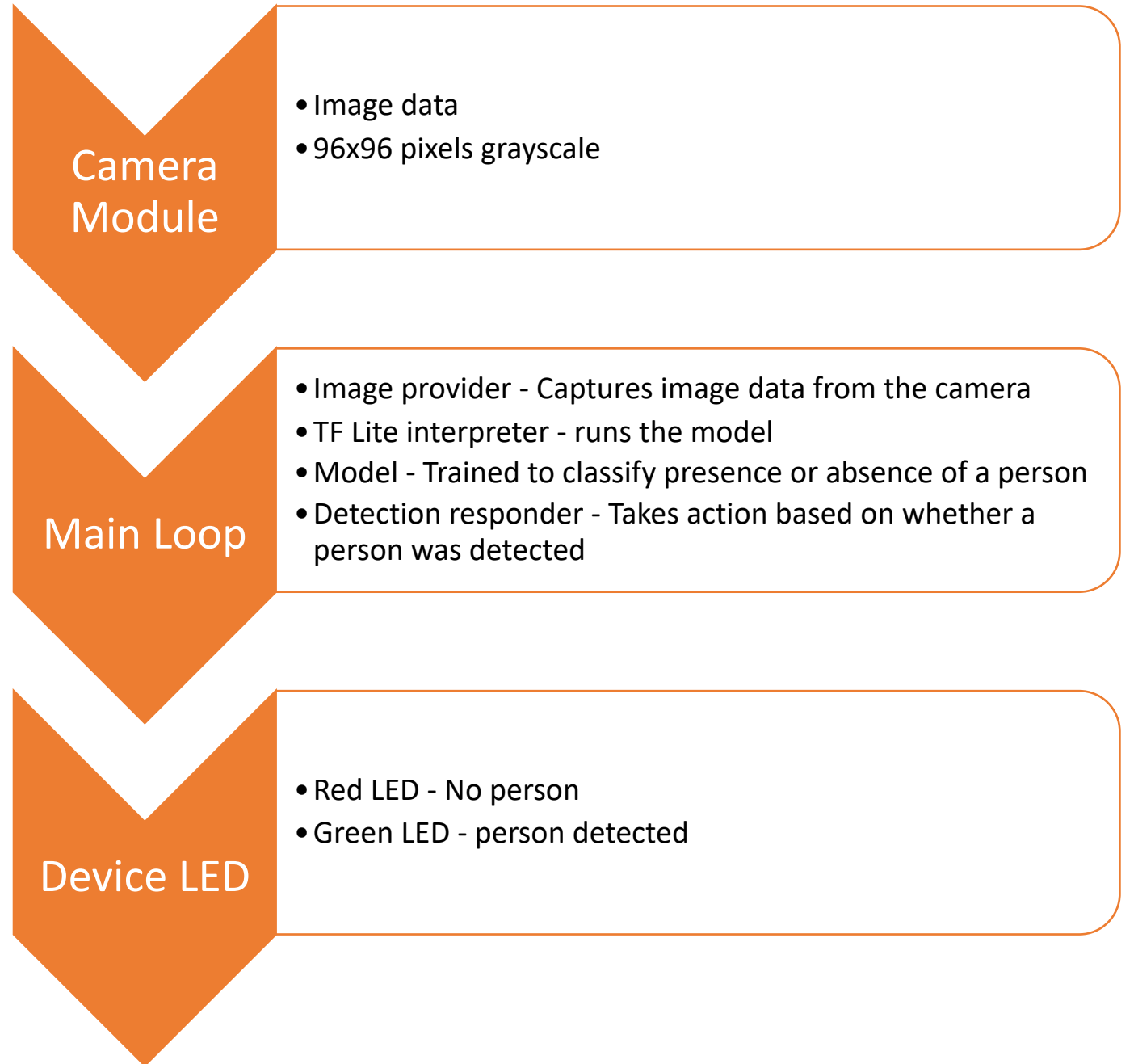
[4] <https://cocodataset.org/#home>

Application Architecture

- Obtain an input
- Preprocess the input to extract features
- Run inference
- Post process the model's output
- Use resulting information to act



Structure Person Detection Application



Main routines

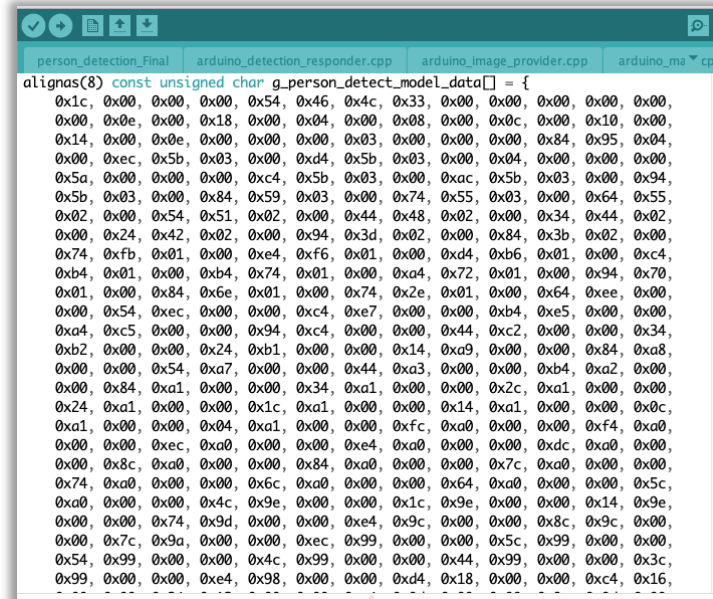
```
// The name of this function is important for Arduino compatibility.
void loop() {
  // Get image from provider.
  if (kTfLiteOk != GetImage(error_reporter, kNumCols, kNumRows, kNumChannels,
    input->data.uint8)) {
    TF_LITE_REPORT_ERROR(error_reporter, "Image capture failed.");
  }

  // Run the model on this input and make sure it succeeds.
  if (kTfLiteOk != interpreter->Invoke()) {
    TF_LITE_REPORT_ERROR(error_reporter, "Invoke failed.");
  }

  TfLiteTensor* output = interpreter->output(0);

  // Process the inference results.
  uint8_t person_score = output->data.uint8[kPersonIndex];
  uint8_t no_person_score = output->data.uint8[kNotAPersonIndex];
  RespondToDetection(error_reporter, person_score, no_person_score);
}
```

Main Loop



```
person_detection_Final  arduino_detection_responder.cpp  arduino_image_provider.cpp  arduino_ma...cpp
aligns(8) const unsigned char g_person_detect_model_data[] = {
  0x1c, 0x00, 0x00, 0x00, 0x54, 0x46, 0x4c, 0x33, 0x00, 0x00, 0x00, 0x00, 0x00,
  0x00, 0x0e, 0x00, 0x18, 0x00, 0x04, 0x00, 0x08, 0x00, 0x0c, 0x00, 0x10, 0x00,
  0x14, 0x00, 0x0e, 0x00, 0x00, 0x00, 0x03, 0x00, 0x00, 0x00, 0x84, 0x95, 0x04,
  0x00, 0xec, 0x5b, 0x03, 0x00, 0xd4, 0x5b, 0x03, 0x00, 0x04, 0x00, 0x00, 0x00,
  0x5a, 0x00, 0x00, 0x00, 0xc4, 0x5b, 0x03, 0x00, 0xac, 0x5b, 0x03, 0x00, 0x94,
  0x5b, 0x03, 0x00, 0x84, 0x59, 0x03, 0x00, 0x74, 0x55, 0x03, 0x00, 0x64, 0x55,
  0x02, 0x00, 0x54, 0x51, 0x02, 0x00, 0x44, 0x48, 0x02, 0x00, 0x34, 0x44, 0x02,
  0x00, 0x24, 0x42, 0x02, 0x00, 0x94, 0x3d, 0x02, 0x00, 0x84, 0x3b, 0x02, 0x00,
  0x74, 0xfb, 0x01, 0x00, 0xe4, 0xf6, 0x01, 0x00, 0xd4, 0xb6, 0x01, 0x00, 0xc4,
  0xb4, 0x01, 0x00, 0xb4, 0x74, 0x01, 0x00, 0xa4, 0x72, 0x01, 0x00, 0x94, 0x70,
  0x01, 0x00, 0x84, 0x6e, 0x01, 0x00, 0x74, 0x2e, 0x01, 0x00, 0x64, 0xee, 0x00,
  0x00, 0x54, 0xec, 0x00, 0x00, 0xc4, 0xe7, 0x00, 0x00, 0xb4, 0xe5, 0x00, 0x00,
  0xa4, 0xc5, 0x00, 0x00, 0x94, 0xc4, 0x00, 0x00, 0x44, 0xc2, 0x00, 0x00, 0x34,
  0xb2, 0x00, 0x00, 0x24, 0xb1, 0x00, 0x00, 0x14, 0xa9, 0x00, 0x00, 0x84, 0xa8,
  0x00, 0x00, 0x54, 0xa7, 0x00, 0x00, 0x44, 0xa3, 0x00, 0x00, 0xb4, 0xa2, 0x00,
  0x00, 0x84, 0xa1, 0x00, 0x00, 0x34, 0xa1, 0x00, 0x00, 0x2c, 0xa1, 0x00, 0x00,
  0x24, 0xa1, 0x00, 0x00, 0x1c, 0xa1, 0x00, 0x00, 0x14, 0xa1, 0x00, 0x00, 0xc,
  0xa1, 0x00, 0x00, 0x04, 0xa1, 0x00, 0x00, 0xfc, 0xa0, 0x00, 0x00, 0xf4, 0xa0,
  0x00, 0x00, 0xec, 0xa0, 0x00, 0x00, 0xe4, 0xa0, 0x00, 0x00, 0xdc, 0xa0, 0x00,
  0x00, 0x8c, 0xa0, 0x00, 0x00, 0x84, 0xa0, 0x00, 0x00, 0x7c, 0xa0, 0x00, 0x00,
  0x74, 0xa0, 0x00, 0x00, 0x6c, 0xa0, 0x00, 0x00, 0x64, 0xa0, 0x00, 0x00, 0x5c,
  0xa0, 0x00, 0x00, 0x4c, 0x9e, 0x00, 0x00, 0x1c, 0x9e, 0x00, 0x00, 0x14, 0x9e,
  0x00, 0x00, 0x74, 0x9d, 0x00, 0x00, 0xe4, 0x9c, 0x00, 0x00, 0x8c, 0x9c, 0x00,
  0x00, 0x7c, 0x9a, 0x00, 0x00, 0xec, 0x99, 0x00, 0x00, 0x5c, 0x99, 0x00, 0x00,
  0x54, 0x99, 0x00, 0x00, 0x4c, 0x99, 0x00, 0x00, 0x44, 0x99, 0x00, 0x00, 0x3c,
  0x99, 0x00, 0x00, 0xe4, 0x98, 0x00, 0x00, 0xd4, 0x18, 0x00, 0x00, 0xc4, 0x16,
```

Model data array

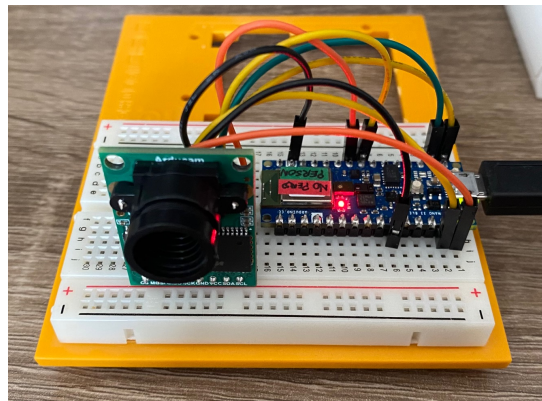
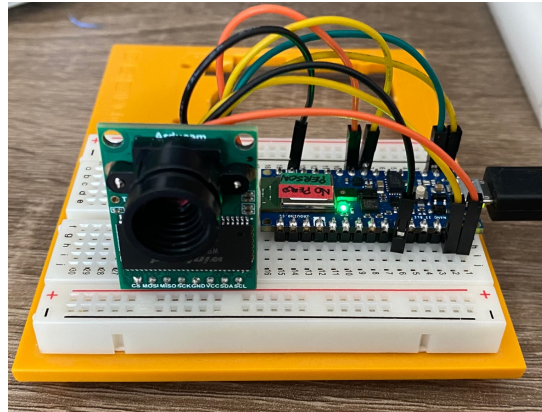
```
// Switch the person/not person LEDs off
digitalWrite(LEDG, HIGH);
digitalWrite(LEDG, HIGH);

// Flash the blue LED after every inference.
digitalWrite(LEDG, LOW);
delay(100);
digitalWrite(LEDG, HIGH);

// Switch on the green LED when a person is detected,
// the red when no person is detected
if (person_score > no_person_score) {
  digitalWrite(LEDG, LOW);
  digitalWrite(LEDG, HIGH);
} else {
  digitalWrite(LEDG, HIGH);
  digitalWrite(LEDG, LOW);
}
```

Detection responder

Deploying to Microcontrollers



Camera e Board connections

Camera Pin	Arduino Board Pin
CS	D7
MOSI	D11
MISO	D12
SCK	D13
GND	GND
VCC	3.3V
SDA	A4
SCL	A5

Deploying to Microcontrollers

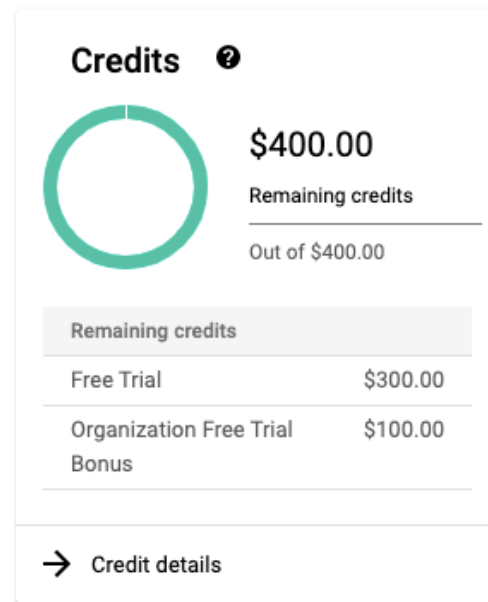
Person Score	No Person Score	Explanation
-82	+82	High confidence in No Person Score
+49	-50	High confidence in Person Score
-28	+28	Slight confidence in Person Score
+28	-28	Slight confidence in No Person Score

```
Starting capture
Image captured
Reading 2056 bytes from Arducam
Finished reading
Decoding JPEG and converting to greyscale
Image decoded and processed
Person score: -93 No person score: 93
Starting capture
Image captured
Reading 3080 bytes from Arducam
Finished reading
Decoding JPEG and converting to greyscale
Image decoded and processed
Person score: -56 No person score: 56
Starting capture
```

[Arduino Project Hub](#) link

Google Cloud Platform

- Picking a Machine
- Google Cloud Platform Instance
- Training the model for other categories



\$1,519.14 monthly estimate

That's about \$2.081 hourly

Pay for what you use: No upfront costs and per second billing

Networking cost also applies. [Learn more](#)

Machine configuration ^

Machine type *
n1-standard-4 (4 vCPUs, 15 GB RAM) ▼ ⓘ

GPU type
NVIDIA Tesla V100 ▼

Number of GPUs
1 ▼



Exporting to TensorFlow Lite

Series of commands:

- Exporting to a GraphDef Protobuf File
- Freezing the Weights
- Quantizing and Converting to TensorFlow Lite
- Converting to a C Source File

Conclusion

- TensorFlow Lite broadens the reach of ML by enabling the transfer of deep learning models into tiny embedded systems.
- The TinyML process of training simplified models in the cloud, converting the files and uploading into the embedded device poses different challenges than traditional ML.
- The hardware/software/libraries compatibility, code compilation, driver updates are also added challenges to TinyML systems.
- Trade accuracy and size of the model

References

- | | |
|-----|--|
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