Edge Computing Lab

Class: TY-AIEC

School of Computing, MIT Art Design Technology University

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Experiment No. 9

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Introduction

The Object Detection using Camera on Edge Computing Devices

Objective: Build a project to detect an object using Edge Computing

Tasks:

- Generate the dataset for customized object
- Configure Edge Impulse for Object Detection
- Building and Training a Model
- Deploy on Edge Computing Device

Introduction

Edge Impulse is a development platform for machine learning on edge devices, targeted at developers who want to create intelligent device solutions. The "Camera "sensor reading equivalent in Edge Impulse would typically involve creating a simple machine learning model that can run on an edge device, like classifying sensor data or recognizing a basic pattern.

Materials Required

Nano BLE Sense Board

Theory

GPIO (General Purpose Input/Output) pins on the Raspberry Pi are used for interfacing with other electronic components. BCM numbering refers to the pin numbers in the Broadcom SOC channel, which is a more consistent way to refer to the GPIO pins across different versions of the

Here's a high-level overview of steps you'd follow to create a "Hello World" project on Edge Impulse:

Steps to Configure the Edge Impulse:

- 1. Create an Account and New Project:
 - Sign up for an Edge Impulse account.
 - Create a new project from the dashboard.

2. Connect a Device:

- You can use a supported development board or your smartphone as a sensor device.
- Follow the instructions to connect your device to your Edge Impulse project.

3. Collect Data:

- Use the Edge Impulse mobile app or the Web interface to collect data from the onboard sensors.
- For a "Hello World" project, you could collect accelerometer data, for instance.

4. Create an Impulse:

- Go to the 'Create impulse' page.
- Add a processing block (e.g., time-series data) and a learning block (e.g., classification).
- Save the impulse, which defines the machine learning pipeline.

5. Design a Neural Network:

- Navigate to the 'NN Classifier' under the 'Learning blocks'.
- Design a simple neural network. Edge Impulse provides a default architecture that works well for most basic tasks.

6. Train the Model:

• Click on the 'Start training' button to train your machine learning model with the collected data.

7. Test the Model:

 Once the model is trained, you can test its performance with new data in the 'Model Testing' tab.

8. Deploy the Model:

- Go to the 'Deployment' tab.
- Select the deployment method that suits your edge device (e.g., Arduino library, WebAssembly, container, etc.).

• Follow the instructions to deploy the model to your device.

9. Run Inference:

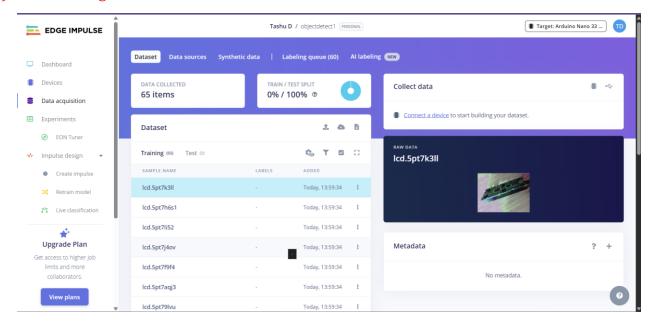
• With the model deployed, run inference on the edge device to see it classifying data in real-time.

10. Monitor:

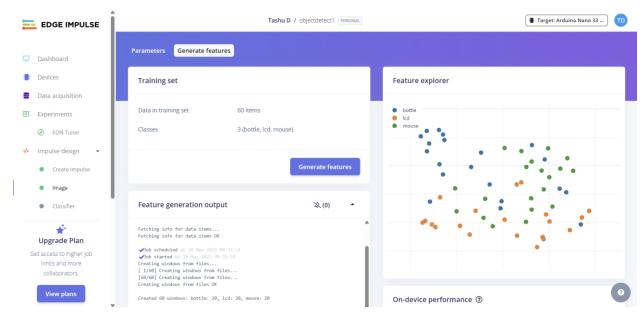
• You can monitor the performance of your device through the Edge Impulse studio.

Paste your Edge Impulse project's Results:

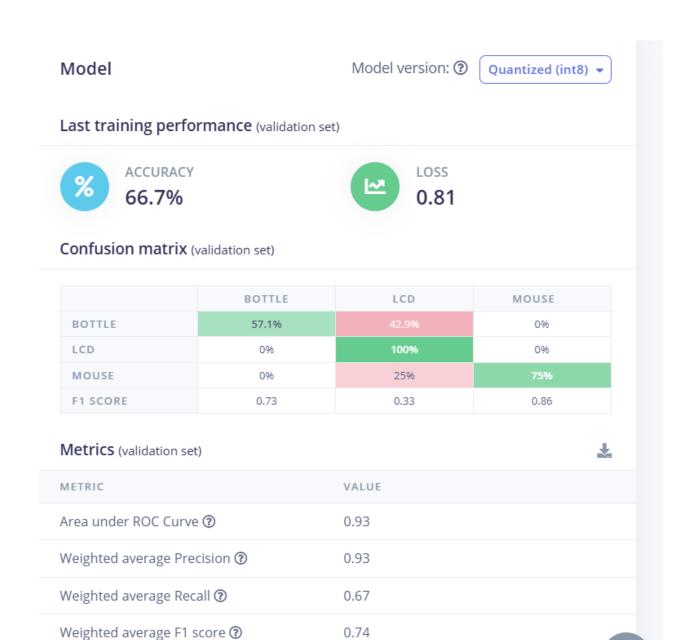
1) Dataset Image



2) Feature extraction - Image

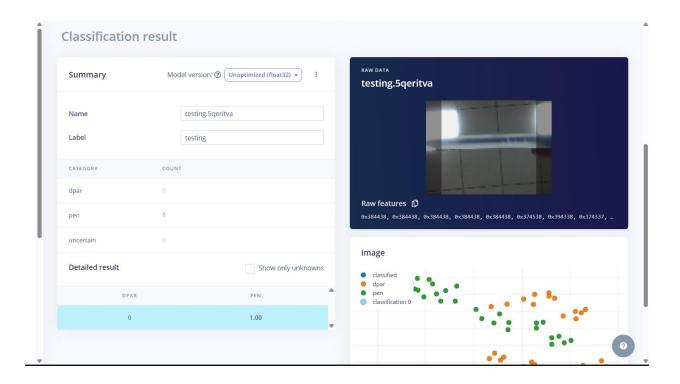


3) Accuracy / Loss - Confusion Matrix - image



Data explorer (full training set) ?

4) Validation Result - Image



5) Copy the code of Arduino Sketch



```
Arduino Nano 33 BLE
                   // Print the prediction results (classification)
   227
            #else
                       ei_printf("Predictions:\r\n");
for (uintl6_t i = 0; i < EI_CLASSIFIER_LABEL_COUNT; i++) {
    ei_printf(" %s: ", ei_classifier_inferencing_categories[i]);
    ei_printf("%.5fk\r\n", result.classification[i].value);</pre>
   235
            // Print anomaly result (if it exists)
#if EI_CLASSIFIER HAS_ANOMALY
| ei_printf("Anomaly prediction: %.3f\r\n", result.anomaly);
#endif
            #if EI_CLASSIFIER_MAS_VISUAL_ANOMALY
    ei_printf("visual anomalies:\r\n");
    for (uint3z t i = 0; i result.visual_ad_count; i++) {
        ei_impulse_result_bounding_box_t bb = result.visual_ad_grid_cells[i];
        if (bb.value == 0) {
   242
    243
                                   continue;
    247
                              }
ei_printf(" %s (%f) [ x: %u, y: %u, width: %u, height: %u ]\r\n",
bb.label,
bb.value,
    250
    251
252
253
                                         bb.x,
bb.y,
bb.width,
                                        bb height);
    254
    255
256
257
                       while (ei get serial available() > 0) {
259 if (ei get serial
o_ble33_sense_camera|Arduino|DE 2.3.6
dit Sketch Tools Help
Arduino Nano 33 BLE
             // Extends the OV767X library function. Reads buf_rows VGA rows from the
             // image sensor
    742
    743
             void OV7675::readBuf()
                  uint32_t ulPin = 33; // P1.xx set of GPIO is in 'pin' 32 and above
NRF_GPIO_Type * port;
    747
                  port = nrf gpio pin port decode(&ulPin);
                  for (int i = 0; i < buf_rows; i++) {
                       // rising edge indicates start of line
while ((*hrefPort & hrefMask) == 0); // wait for HIGH
    755
756
757
                       for (int col = 0; col < bytes_per_row; col++) {</pre>
                            // rising edges clock each data byte
while ((*pclkPort & pclkMask) != 0); // wait for LOW
                             uint32_t in = port->IN; // read all bits in parallel
                             in >>= 2; // place bits 0 and 1 at the "bottom" of the register
    762
    763
764
765
                           in %= 0x1f03; // isolate the 8 bits we care about in |= (in >> 6); // combine the upper 6 and lower 2 bits
    766
                            raw buf[offset++] = in:
                            while ((*pclkPort & pclkMask) == 0); // wait for HIGH
    770
                       while ((*hrefPort & hrefMask) != 0); // wait for LOW
            } /* OV7675::readBuf() */
```

6) Screen shot of Arduino Terminal - Result