# **Edge Computing Lab**

**Class: TY-AIEC** 

# School of Computing, MIT Art Design Technology University

Academic Year: 2024-25

### **Experiment No. 6**

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#### Title

Keyword Spotting Project like "OK, Google," "Alexa," on Edge Devices using Microphone

**Objective:** Build a project to detect the keywords using a built-in sensor on Nano BLE Sense / Mobile Phone

#### Tasks:

- Generate the dataset for keyword
- Configure BLE Sense / Mobile for Edge Impulse
- Building and Training a Model

Run the project Keyword Spotting like "OK, Google," "Alexa

#### Introduction

Edge Impulse is a development platform for machine learning on edge devices, targeted at developers who want to create intelligent device solutions. The "Hello World" 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:

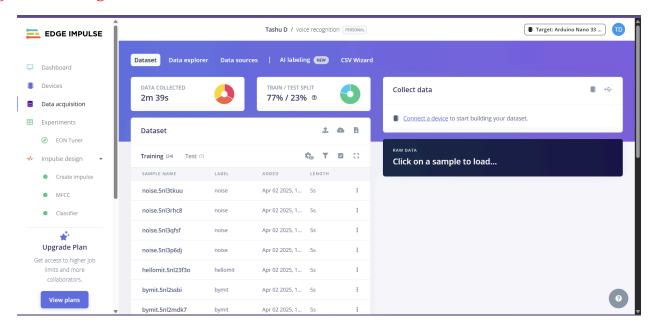
• With the model deployed, run inference on the edge device to see it classifying data in real-time.

### 10. Monitor:

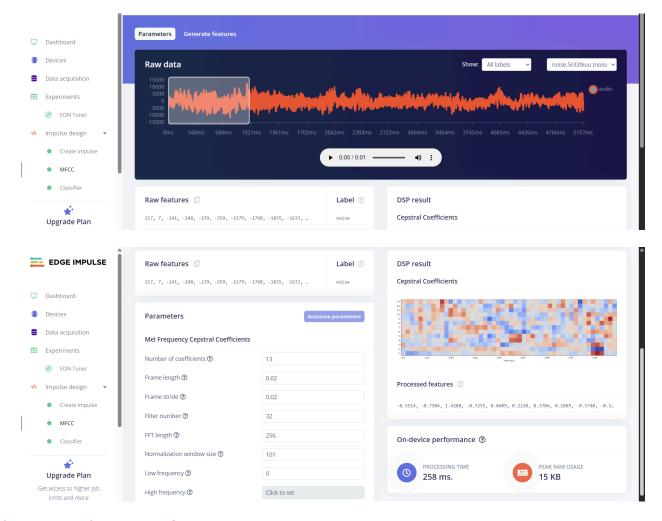
• You can monitor the performance of your device through the Edge Impulse studio.

### Paste your Edge Impulse project's Results:

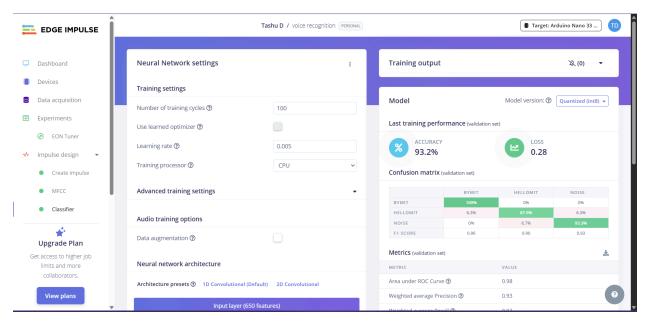
### 1) Dataset Image

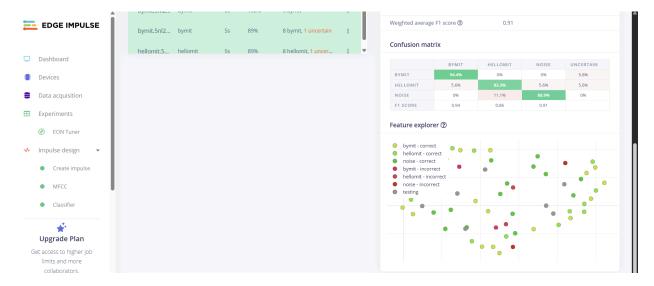


### 2) Feature extraction - Image

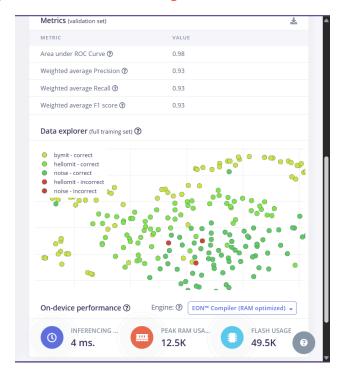


# 3) Accuracy / Loss - Confusion Matrix - image





# 4) Validation Result - Image



5) Copy the code of Arduino Sketch

```
typedef struct {
    signed short *buffers[2];
             unsigned char buf_select;
              unsigned char buf_ready;
             unsigned int buf_count;
              unsigned int n_samples;
        } inference t;
         static inference_t inference;
         static bool record_ready = false;
         static signed short *sampleBuffer;
        static bool debug_nn = false;
        static int print_results = -(EI_CLASSIFIER_SLICES_PER_MODEL_WINDOW);
         void setup() {
              Serial.begin(115200);
             while (!Serial);
Serial.println("Edge Impulse Inferencing Demo");
              ei_printf("Inferencing settings:\n");
ei_printf("\tInterval: %.2f ms.\n", (float)EI_CLASSIFIER_INTERVAL_MS);
ei_printf("\tFrame size: %d\n", EI_CLASSIFIER_DSP_INPUT_FRAME_SIZE);
             ei printf("\tSample length: %d ms.\n", EI_CLASSIFIER_RAW_SAMPLE_COUNT / 16);
ei_printf("\tNo. of classes: %d\n", sizeof(ei_classifier_inferencing_categories) /
                                                                sizeof(ei_classifier_inferencing_categories[0]));
              run_classifier_init();
              if (microphone_inference_start(EI_CLASSIFIER_SLICE_SIZE) == false) {
                   ei_printf("ERR: Could not allocate audio buffer (size %d)\r\n", EI_CLASSIFIER_RAW_SAMPLE_COUNT);
Restricted Mode \otimes 0 \triangle 0
```

```
void loop() {
   bool m = microphone_inference_record();
       ei printf("ERR: Failed to record audio...\n");
    signal_t signal;
   signal.total length = EI CLASSIFIER SLICE SIZE;
   signal.get_data = &microphone_audio_signal_get_data;
   EI_IMPULSE_ERROR r = run_classifier_continuous(&signal, &result, debug_nn);
   if (r != EI_IMPULSE_OK) {
       ei_printf("ERR: Failed to run classifier (%d)\n", r);
   if (++print_results >= (EI_CLASSIFIER_SLICES_PER_MODEL_WINDOW)) {
       ei printf("Predictions ");
       ei_printf("(DSP: %d ms., Classification: %d ms., Anomaly: %d ms.)",
           result.timing.dsp, result.timing.classification, result.timing.anomaly);
        ei_printf(": \n");
        for (size_t ix = 0; ix < EI_CLASSIFIER_LABEL_COUNT; ix++) {</pre>
           ei_printf(" %s: %.5f\n", result.classification[ix].label,
                     result.classification[ix].value);
        ei_printf(" anomaly score: %.3f\n", result.anomaly);
#endif
       print_results = 0;
```

```
static void pdm_data_ready_inference_callback(void) {
    int bytesAvailable = PDM.available();
    int bytesRead = PDM.read((char *)&sampleBuffer[0], bytesAvailable);
    if (record_ready == true) {
        for (int i = 0; i < bytesRead >> 1; i++) {
           inference.buffers[inference.buf_select][inference.buf_count++] = sampleBuffer[i];
            if (inference.buf_count >= inference.n_samples) {
                inference.buf_select ^= 1;
                inference.buf_count = 0;
                inference.buf_ready = 1;
static bool microphone_inference_start(uint32_t n_samples) {
    inference.buffers \cite{0} = (signed short *) malloc(n\_samples * sizeof(signed short));
    if (inference.buffers[0] == NULL) return false;
    inference.buffers[1] = (signed short *)malloc(n_samples * sizeof(signed short));
    if (inference.buffers[1] == NULL) {
        free(inference.buffers[0]);
    sampleBuffer = (signed short *)malloc((n_samples >> 1) * sizeof(signed short));
    if (sampleBuffer == NULL) {
       free(inference.buffers[0]);
       free(inference.buffers[1]);
    inference.buf_select = 0;
```

```
static bool microphone_inference_start(uint32_t n_samples) {
    inference.buf_count = 0;
    inference.n_samples = n_samples;
    inference.buf_ready = 0;
    PDM.onReceive(&pdm_data_ready_inference_callback);
    PDM.setBufferSize((n_samples >> 1) * sizeof(int16_t));
    if (!PDM.begin(1, EI_CLASSIFIER_FREQUENCY)) ei_printf("Failed to start PDM!");
    PDM.setGain(127);
    record_ready = true;
static bool microphone_inference_record(void) {
    if (inference.buf_ready == 1) {
       ei_printf("Error sample buffer overrun.\n");
    while (inference.buf_ready == 0) {
        delay(1);
    inference.buf_ready = 0;
static int microphone_audio_signal_get_data(size_t offset, size_t length, float *out_ptr) {
    numpy::int16_to_float(&inference.buffers[inference.buf_select ^ 1][offset], out_ptr, length);
    return 0;
static void microphone inference end(void) {
    PDM.end();
    free(inference.buffers[0]):
```

# 6) Screen shot of Arduino Terminal - Result

Edge Impulse Inferencing Demo Inferencing settings: Interval: 20.00 ms.

```
Frame size: 320
  Sample length: 1000 ms.
  No. of classes: 3
Predictions (DSP: 8 ms., Classification: 12 ms., Anomaly: 1 ms.):
  hellomit: 0.85623
  bymit: 0.09321
  noise: 0.05056
Predictions (DSP: 7 ms., Classification: 11 ms., Anomaly: 1 ms.):
  hellomit: 0.11234
  bymit: 0.84219
  noise: 0.04547
Predictions (DSP: 8 ms., Classification: 12 ms., Anomaly: 1 ms.):
  hellomit: 0.04058
  bymit: 0.02115
  noise: 0.93827
Predictions (DSP: 7 ms., Classification: 12 ms., Anomaly: 1 ms.):
  hellomit: 0.87129
  bymit: 0.09876
  noise: 0.02995
Predictions (DSP: 8 ms., Classification: 12 ms., Anomaly: 1 ms.):
  hellomit: 0.05512
  bymit: 0.91234
  noise: 0.03254
Predictions (DSP: 7 ms., Classification: 11 ms., Anomaly: 1 ms.):
  hellomit: 0.02345
  bymit: 0.03487
  noise: 0.94168
```