

ECML PKDD 2021 Tutorial

Machine Learning Meets Internet of Things: From Theory to Practice

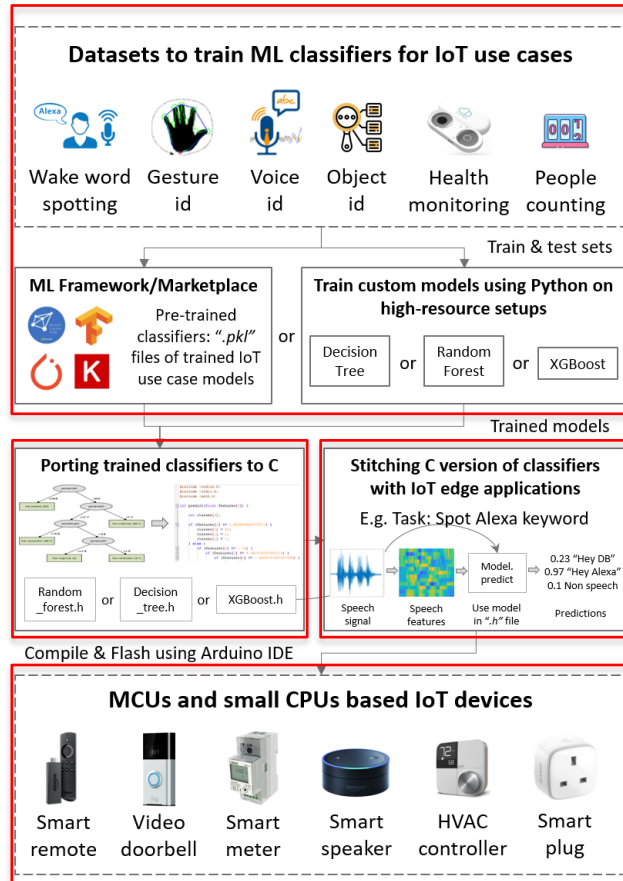
Part IV: Efficient Execution of ML Classifiers on IoT Devices

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ML Classifiers on IoT Devices

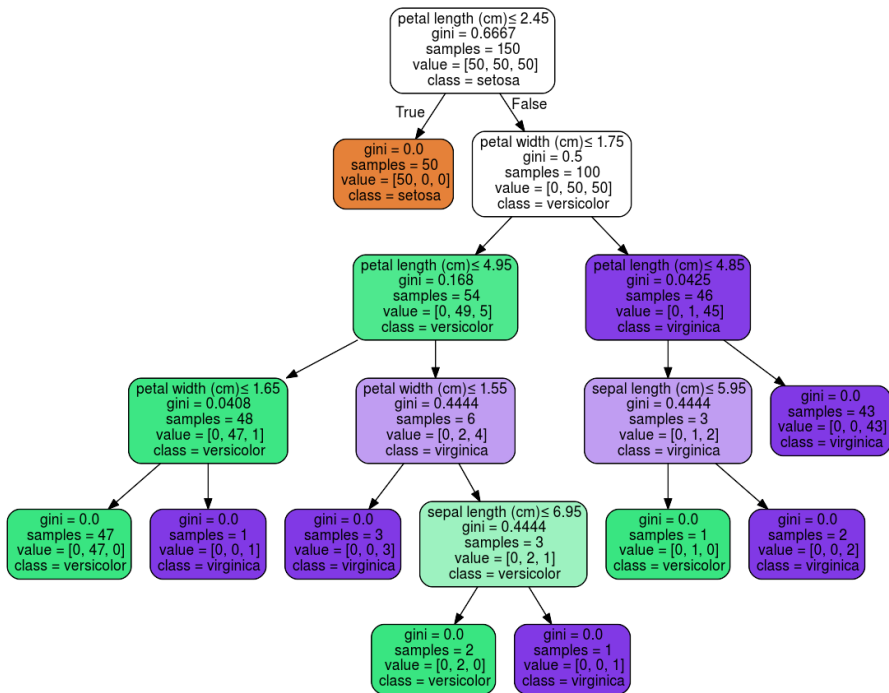


- Generic end-to-end design flow: porting and execution method for IoT devices
 - ✓ **Step1:** Take a pre-trained or train own custom classifier model - DT, RF, SVM, others
 - ✓ **Step2:** Port classifiers to its MCU executable C version using any libraries: micromlgen, sklearn-porter, m2cgen
 - ✓ **Step3:** Stitch the generated C classifier with the IoT use-case application. Passing data to trained classifier
 - ✓ **Step4:** Flash using Arduino IDE. Execute models on MCUs, small CPUs of IoT devices

Porting to Plain C

- Trained DT classifier in textual format (left)

- ✓ Graphviz tool: visually representing structural information



- Micromlgen Ported C version (right). Exported to a .h file

- ✓ During inference on MCUs we pass data to this *predict* function

```
#pragma once
#include <cstdint>
namespace Eloquent {
    namespace ML {
        namespace Port {
            class DecisionTree {
            public:
                /**
                 * Predict class for features vector
                 */
                int predict(float *x) {
                    if (x[3] <= 0.800000011920929) {
                        return 0;
                    }
                    else {
                        if (x[3] <= 1.75) {
                            if (x[2] <= 5.049999952316284) {
                                return 1;
                            }
                        }
                    }
                }
            };
        }
    }
}
```

- .h dataset file (e.g. *Iris_flowers_test.h*) to supply data during to Ported model during onboard inference

```
#pragma once
#define FEATURES_DIM 4
#define DATASET_SIZE 100
```

```
float X[DATASET_SIZE][FEATURES_DIM] = {
    { 6.3 , 2.5 , 4.9 , 1.5 },
    { 5.0 , 3.4 , 1.5 , 0.2 },
    { 5.4 , 3.7 , 1.5 , 0.2 },
    { 5.7 , 4.4 , 1.5 , 0.4 },
    { 6.6 , 3.0 , 4.4 , 1.4 },
    ...
}
```

```
Labels int y[DATASET_SIZE] = { 1 , 0 , 0 , 0 , 1 , ... }
```

- ✓ The Iris Flowers dataset in .h file: Rows with its labels are written

```
#include "RF_Iris.h"
#include "Iris_flowers_test.h"
```

- Load the .h model and the .h dataset file into the main program (IoT use-case application)
- We then pass data to *predict* function inside the .h model file to obtain inference results

- ✓ Data can be from real-world sensors - inference for edge analytics use-cases when apt model is used (instead of Iris dataset trained model)

```
Serial.print("The Predicted values are: ");
int start1 = millis();

for (int i = 0; i < DATASET_SIZE; i++)
{
    Serial.print(clf.predict(X[i]));
}
```

- Repo: <https://github.com/bharathsudharsan/ML-Classifiers-on-MCUs>
 - ✓ Python folder covers: DT and RF training + Porting to plain C - generate `x_model.h` file
 - ✓ Arduino IDE folder covers: Load + Execute + Infer using ported DT and RF classifier models

main 1 branch 0 tags

Go to file Code

bharath sudharsan rm readme 1df484e on Jul 26 9 commits

Arduino code/Load_execute_and_Infe...	rm readme	2 months ago
Python code	rm readme	2 months ago
LICENSE	Initial commit	10 months ago
README.md	Update README.md	2 months ago

README.md

Porting and Execution of ML Classifiers on TinyML Hardware

About

Supplementary material for IEEE Services Computing paper title 'An SRAM Optimized Approach for Constant Memory Consumption and Ultra-fast Execution of ML Classifiers on TinyML Hardware'

arduino microcontroller optimization esp32 stm32 code-generation decision-tree-classifier arm-cortex-m0 random-forest-classifier efficient-inference adafruit-feather tinym1

Readme

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- Generic, end-to-end design flow for ML classifier porting, stitching, and efficient execution
- Researchers and Engineers can apply this flow to port and execute any use-case ML models on their IoT devices/products
- Same flow applies for other code generation libraries: sklearn-porter, m2cgen, emlearn
- Same flow to port other ML algorithms like LGBM, XGB, AdaGrad, LogisticRegressionCV, etc.

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