

International Workshop on Embedded Artificial Intelligence Devices, Systems, and Industrial Applications (EAI)



Milan, Italy 19 September 2022



An Embedding Workflow for Tiny Neural Networks on Arm Cortex-M0(+) Cores

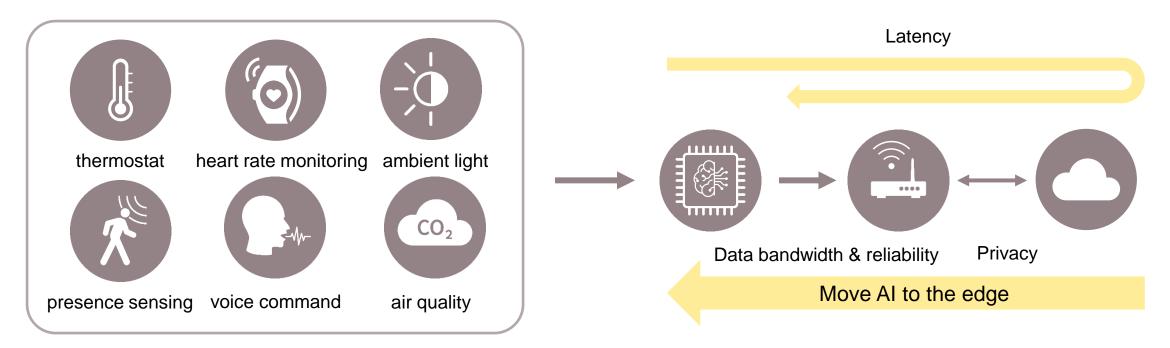
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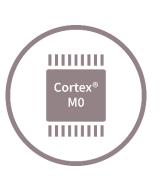




Motivation: Al-enabled IoT Edge Devices

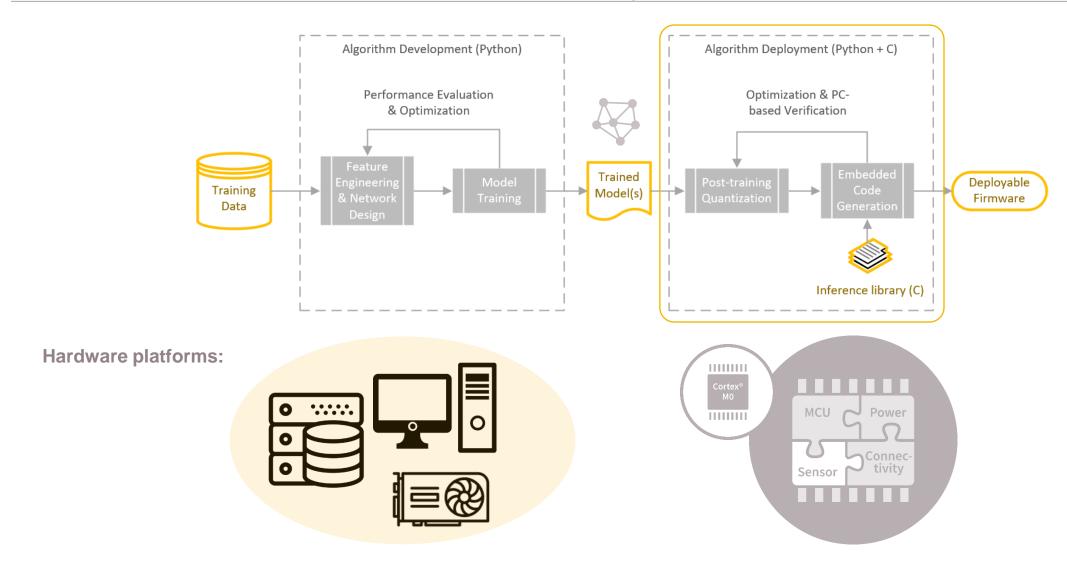


- > Always-on and very likely battery-powered: strict power and cost limit, e.g., 1 mW and 1 Euro
- > Time series analysis: performance could be improved with fully-connected and/or recurrent networks
- Arm Cortex-M0(+)
 - widely used for sensor control
 - 32-bit performance at an 8-bit price point
- Challenges
 - very limited computational resource and memory footprint, typically 16 or 32 kB Flash and 4 or 8 kB SRAM
 - no operation system, so the C code needs to run on bare metal



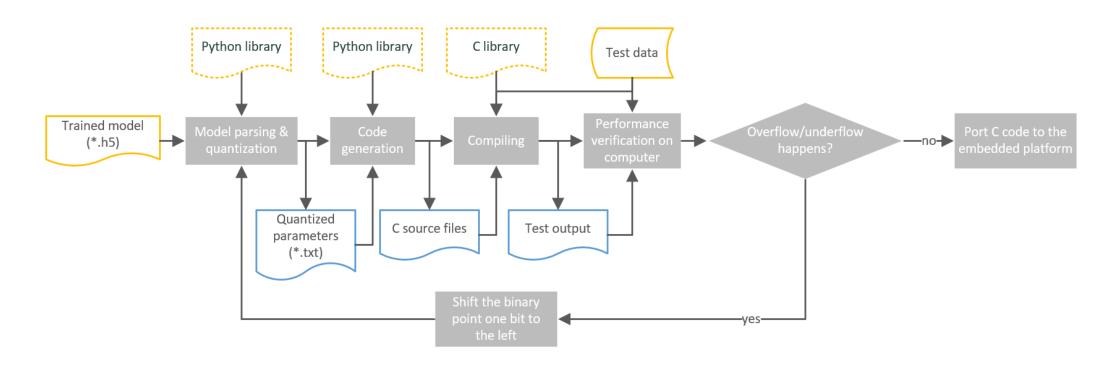


Development and Deployment of Sensor Algorithms



Embedding workflow

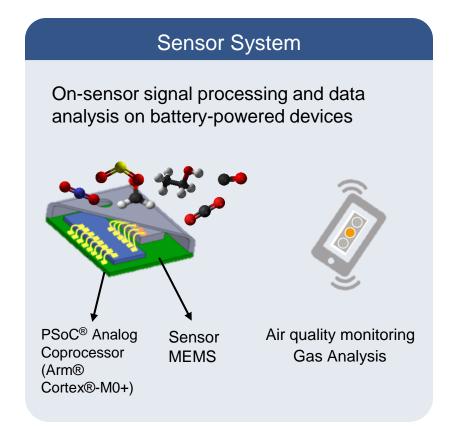




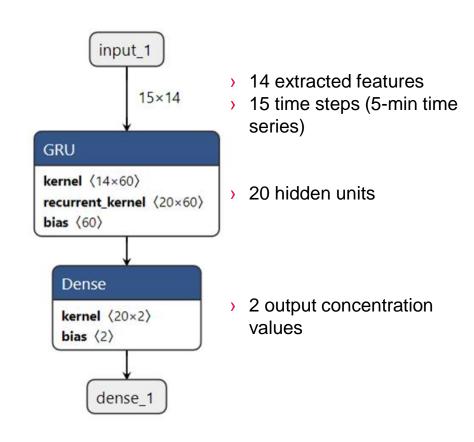
- > Reusable Python and C library for network quantization and code generation
- > Flexible layer combination, typically dense and GRU (gated recurrent unit) for time series analysis
- Customizable bit shift for each layer









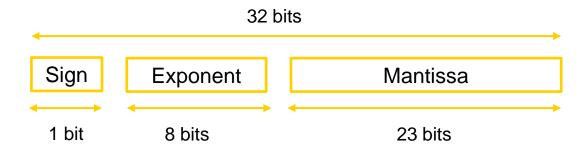




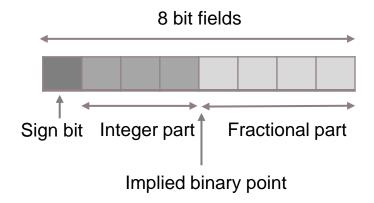


Example: Model Parsing & Quantization

Single-precision floating-point representation



> 8-bit fixed-point representation



configuration.txt

```
3
GRU 15 14 20 tanh sigmoid
Dense 20 2 relu
14 2
```

quantized_parameters.txt

```
5
6
--- GRU layer ---
<quantized weights>
<quantized biases>
<quantized weights>
<quantized biases>
<quantized weights>
<quantized weights>
<quantized biases>
--- Dense layer ---
<quantized weights>
<quantized weights>
<quantized biases>
```





Example: C library and Code Generation

Files in the C project

```
activation_functions.c
activation_functions.h
fixed_point_operations.c
fixed_point_operations.h
nn_helpers.c
nn_helpers.h
nn_layers.c
nn_layers.h
softmax_functions.c
softmax_functions.c
nn_inference_env.c
nn_inference_env.h
main.c
```

nn_inference_env.c

nn inference env.h

```
typedef struct {
    const gru param 8bit t *layer1;
    const dense param 8bit t *layer2;
    const uint8 t n features;
    const uint8 t n out;
} model param t;
static const model param t model params = {
    &(const gru param 8bit t)
    &(const dense param 8bit t)
        (const int8 t *[])
            (const int8 t []) {16, 6},
            (const int8 t []) \{-46, -49\},
            (const int8 t []) {68, 79},
            (const int8 t []) {74, 95}
        (const int8 t []) {8, -26},
       ACTIVATION SIGMOID,
        // dim out, shift
        2, 6
    14, 2
```

nn_layers.h

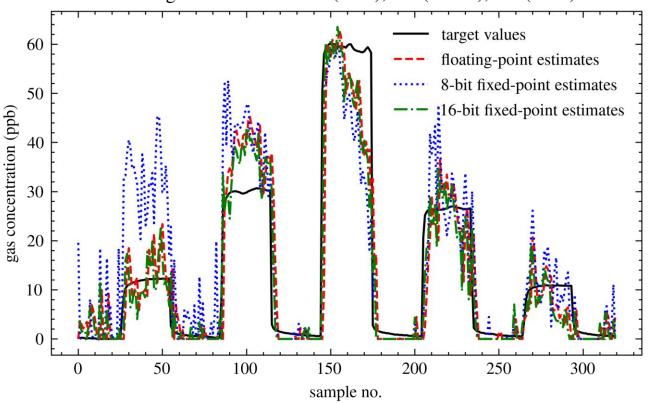
```
typedef struct {
    const int8 t **weights;
   const int8 t *biases;
   const uint8 t activation;
    const uint8 t dim out;
   const uint8 t shift;
} dense param 8bit t;
typedef struct {
   const int8 t **W z;
   const int8 t **W r;
   const int8 t **W h;
    const int8 t **U z;
    const int8 t **U r;
   const int8 t **U h;
    const int8 t *bias z;
    const int8 t *bias r;
    const int8 t *bias h;
    const uint8 t activation;
    const uint8 t recurrent activation;
   const uint8 t n timesteps;
   const uint8 t n units;
    const uint8 t shift;
}gru param 8bit t;
```



Example: Performance Verification







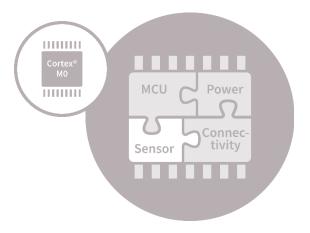
Network implementation	MAE (ppb)	Flash (kB)	SRAM (kB)
Floating-point	4.5	33.5	1.8
16-bit fixed-point	3.9	16.7	1.2
8-bit fixed-point	6.8	8.4	0.8

- 16-bit implementation: 49.3% memory reduction with little change in performance
- 8-bit implementation: 73.9% memory saved with more noisy and less accurate concentration estimates

Summary



- Arm Cortex-M0(+) for AI-enabled IoT edge devices
 - Power-efficient and cost-effective
 - For many applications, more value could be added without additional material cost
- > End-to-end workflow for network quantization and code generation
 - Reusable Python and C library
 - Software architecture supporting flexible layer combinations
 - Bit shifts customizable for each layer
- Example project: low-cost environmental sensing
 - Up to 73.9% of the memory footprint was reduced with only a small sacrifice in performance





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