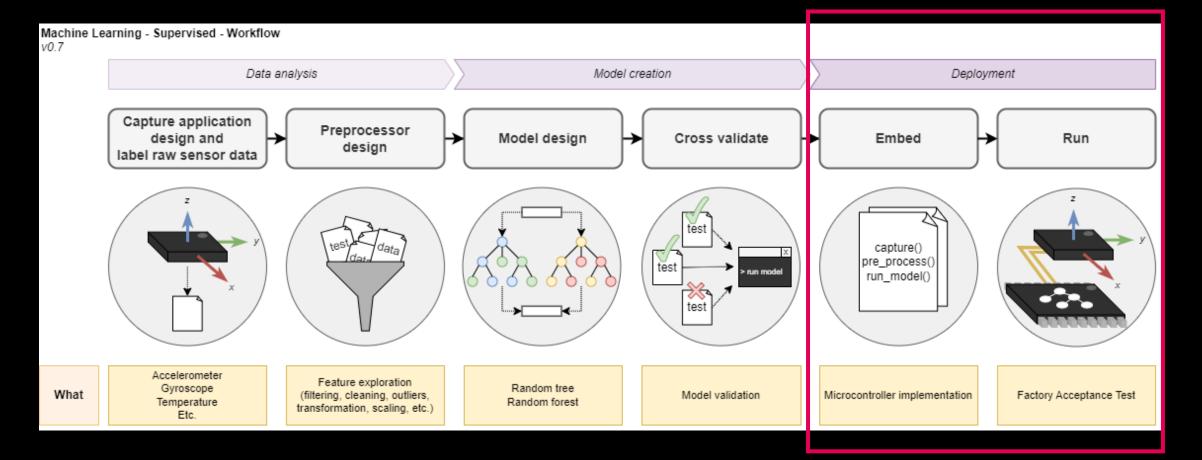
S6-ESE-AI

DEPLOYMENT

JEROEN VEEN HUGO ARENDS



WORKFLOW





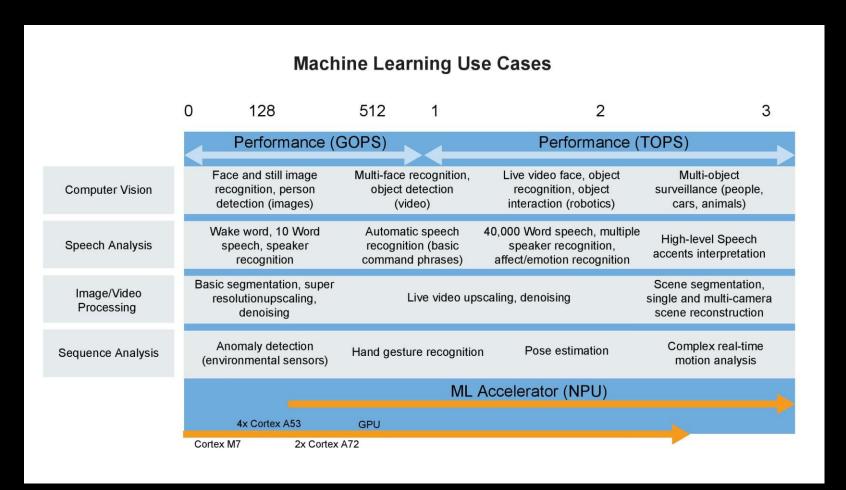
AGENDA

- Code generation
- Reducing computational complexity
- Accelerators
- Platforms
- User Acceptance Testing

REDUCING COMPUTATIONAL COMPLEXITY

- Quantization
- Pruning of trees
- Separable convolutions

EDGE ML USE CASES





HW ACCELERATORS

- Get started with the Dev Board Micro | Coral
- Al Accelerator Hailo-8 M.2 Al Module | Superior Edge Performance
- Jetson TX2 for next-level performance | NVIDIA

ML FRAMEWORKS

TinyML

Focuses on ultra-low power and small footprint ML, ideal for microcontroller-based systems

Tensorflow Lite

Well-suited for mobile and edge devices, with a broad range of model support, e.g. TensorFlow.

PyTorch Mobile

Good for deploying PyTorch models on mobile devices, with a growing ecosystem and support.

OpenVINO

Optimized for deploying on Intel hardware, including CPUs, GPUs, and VPUs.

NVIDIA Jetson

Targeted at edge computing, leveraging NVIDIA GPUs for high-performance inference.

Caffe2

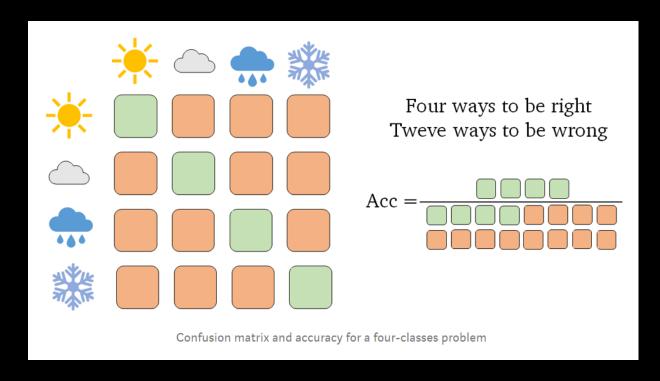
Known for its speed and efficiency, but with a smaller community

CRITERIA FOR COMPARING ML FRAMEWORKS

- Performance
- Speed
- Resource efficiency
- Ease of use and integration
- Platform and hardware compatibility
- Community and ecosystem
- Security and privacy
- Compatibility and support
- Update frequency and maintenance

USER ACCEPTANCE TESTING

- Determine confusion matrix
- Compute classification report



```
>>> from sklearn.metrics import classification report
>>> y true = [0, 1, 2, 2, 2]
>>> y pred = [0, 0, 2, 2, 1]
>>> target names = ['class 0', 'class 1', 'class 2']
>>> print(classification report(y true, y pred, target names=target names))
              precision
                          recall f1-score support
    class 0
    class 1
                   0.00
                            0.00
                                      0.00
    class 2
                  1.00
                            0.67
                                      0.80
                                      0.60
    accuracy
   macro avg
                   0.50
                            0.56
                                      0.49
weighted avg
                   0.70
                             0.60
                                      0.61
>>> y pred = [1, 1, 0]
>>> y_true = [1, 1, 1]
>>> print(classification report(y true, y pred, labels=[1, 2, 3]))
                          recall f1-score support
              precision
                  1.00
                            0.67
                                      0.80
                                                   3
                   0.00
                            0.00
                                      0.00
                   0.00
                            0.00
                                      0.00
   micro avg
                  1.00
                            0.67
                                      0.80
                                                   3
   macro avg
                  0.33
                            0.22
                                      0.27
weighted avg
                  1.00
                             0.67
                                      0.80
                                                   3
```

GENERALIZATION

- Test the model with new, unseen data to assess its generalization capability.
- Use relevant metrics (like accuracy, precision, recall) to evaluate performance.

Input Process Output



./data/model/dtc_model.gz



./model_embedding/ code_generator_dtc2c.py



./data/model_embedding/ dtc_model.c



./data/model/dtc_train_bunch.csv

./data/model/*.png
HAN UNIVERSITY

OF APPLIED SCIENCES



./data/model_embedding/dtc_model.c

```
typedef enum
{
    left_right = 0,
    stationary = 1,
    up_down = 2,
}dtc_t;
```

Enumerated type containing each label

./data/model_embedding/dtc_model.c

```
\brief Decision tree classifier
 Decision tree classifier based on
 the following input characteristics:
   BLOCK_SIZE: 100
   BLOCK_TYPE: BLOCK
  \return dtc_t
      left_right
       stationary
       up_down
*/
```

Documented function



./data/model_embedding/dtc_model.c

```
dtc_t dtc(const float y_out_fir_rescale_variance,
    const float x_out_fir_rescale_variance)
    dtc_t ret;
    if(y_out_fir_rescale_variance <= 0.000847f)</pre>
         ret = stationary;
    else // y_out_fir_rescale_variance > 0.000847f
        if(x_out_fir_rescale_variance <= 0.033878f)</pre>
             ret = up_down;
```

Function implementation

```
main.d
```

main.c – BLOCK example

```
Handle the data as soon as new data is available
  mma8451 accelerometer Output Data Rate (ODR) is set to 100 Hz
if(mma8451_ready_flag)
    // Set initial timestamp
   ms1 = ms;
    // Clear the flag
   mma8451 ready flag = false;
      Reads the data in three global variables: x_out_mg, y_out_mg and
    // z_out_mg
   mma8451_read();
```

```
main.c – BLOCK example
```

```
// TODO Implement filter function as required by the application.
```

```
// Filter accelerometer data
x_out_mg = fir(x_out_mg, fir_coefs, fir_x, N_FIR);
y_out_mg = fir(y_out_mg, fir_coefs, fir_y, N_FIR);
z_out_mg = fir(z_out_mg, fir_coefs, fir_z, N_FIR);
```

```
main.c – BLOCK example
```

```
// TODO Implement normalization function as required by the
// application.

// Scale accelerometer data
const float from[2] = {-1000.0f, 1000.0f};
const float to[2] = {-1.0f, 1.0f};

x_out_mg = rescale(x_out_mg, from, to);
y_out_mg = rescale(y_out_mg, from, to);
z_out_mg = rescale(z_out_mg, from, to);
```

```
main.c – BLOCK example
```

```
// TODO Finish this example by designing an ML model and implement
// the generated C code.

// Add accelerometer data to the buffer
buffer_x_out[n] = x_out_mg;
buffer_y_out[n] = y_out_mg;
buffer_z_out[n] = z_out_mg;
n++;
```

```
main.c – BLOCK example
```

```
// Buffer full?
if(n >= N_BUFFER)
    // Reset buffer counter for next block
    n = 0;
    // Calculate features by using feature functions
    float x_out_var = variance(buffer_x_out, N_BUFFER);
    float y_out_var = variance(buffer_y_out, N_BUFFER);
       Calculate label by using the generated Decision Tree
    // Classifier
    dtc_t label = dtc(x_out_var, y_out_var);
```

```
main.c – BLOCK example
```

```
char *label_str = "";
// Use the calculated label for further processing
if(label == stationary)
       label_str = "stationary";
else if(label == up_down)
       label_str = "up_down";
else if(label == left_right)
       label_str = "left_right";
```

```
main.c - BLOCK example

// Set final timestamp
ms2 = ms;

// Print duration and label
printf("%d,%d,%s\n",
ms1,
ms2,
label_str);
```

main.c – SLIDING example

```
Handle the data as soon as new data is available
  mma8451 accelerometer Output Data Rate (ODR) is set to 100 Hz
if(mma8451_ready_flag)
    // Set initial timestamp
   ms1 = ms;
    // Clear the flag
   mma8451 ready flag = false;
      Reads the data in three global variables: x_out_mg, y_out_mg and
    // z_out_mg
   mma8451_read();
```

```
main.c – SLIDING example
```

```
// TODO Implement filter function as required by the application.
```

```
// Filter accelerometer data
x_out_mg = fir(x_out_mg, fir_coefs, fir_x, N_FIR);
y_out_mg = fir(y_out_mg, fir_coefs, fir_y, N_FIR);
z_out_mg = fir(z_out_mg, fir_coefs, fir_z, N_FIR);
```

```
main.c – SLIDING example
```

```
// TODO Implement normalization function as required by the
// application.

// Scale accelerometer data
const float from[2] = {-1000.0f, 1000.0f};
const float to[2] = {-1.0f, 1.0f};

x_out_mg = rescale(x_out_mg, from, to);
y_out_mg = rescale(y_out_mg, from, to);
z_out_mg = rescale(z_out_mg, from, to);
```



main.c – SLIDING example

```
// TODO Finish this example by designing an ML model and implement
// the generated C code.

// Buffer full?
if(n >= (N_BUFFER-1))
{
    // Remove first data item in the buffer and move all others one
    // position.
    memmove(&buffer_x_out[0], &buffer_x_out[1], sizeof(float) * (N_BUFFER-1));
    memmove(&buffer_y_out[0], &buffer_y_out[1], sizeof(float) * (N_BUFFER-1));
    memmove(&buffer_z_out[0], &buffer_z_out[1], sizeof(float) * (N_BUFFER-1));
}
```

```
main.c – SLIDING example
```

```
// Add accelerometer data to the buffer
buffer_x_out[n] = x_out_mg;
buffer_y_out[n] = y_out_mg;
buffer_z_out[n] = z_out_mg;
```

```
main.c – SLIDING example
```

```
// Buffer full?
if(n >= N_BUFFER)
{
    // Set counter at the end of the buffer
    n = N_BUFFER-1;

    // Calculate features by using feature functions
    float y_out_var = variance(buffer_y_out, N_BUFFER);

    // Calculate label by using the generated Decision Tree
    // Classifier
    dtc_t label = dtc(y_out_var);
```

```
main.c – SLIDING example
```

```
char *label_str = "";
// Use the calculated label for further processing
if(label == stationary)
       label_str = "stationary";
else if(label == up_down)
       label_str = "up_down";
else if(label == left_right)
       label_str = "left_right";
```

```
main.c – SLIDING example
```

Process Output Input



./../lib/filters.c

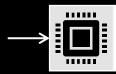


./../lib/normalizations.c

/../lib/normalizations.h

./data/model_embedding/ dtc_model.c

main.c



Source file in microcontroller IDE project

Factory acceptance test

Input Process Output

Microcontroller development board



./capturing/data_logger.py

Terminal window showing printed output



>_ Terminal

```
Press CTRL+C to quit
Opened COM13 @ 115200bps
139577,139579, stationary
140553,140555,stationary
141529,141531,up_down
142505,142507,up_down
143481,143483,left_right
144457,144459,left_right
145433,145435,left_right
147385,147387,stationary
148361,148363, stationary
Stop app
Closed COM13
```

Input Process Output

Microcontroller development board



Terminal window showing timestamp inspected output

>_ Terminal

```
Press CTRL+C to quit
Opened COM13 @ 115200bps
[ 62] average: 2.0ms | min: 1.0ms | max: 2.0ms
```