



Machine Learning in Microcontrollers

DAY 5 : Deploying Machine Learning Models

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THE SPEAKER



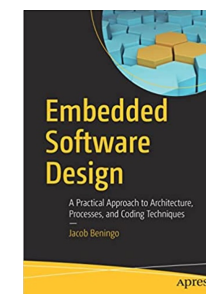
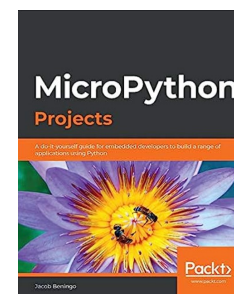
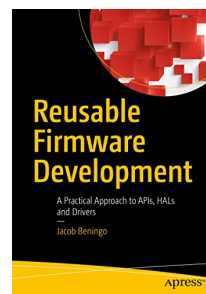
Jacob Beningo

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Beningo Embedded Group - President

Focus: Embedded Software Consulting and Training

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Course Sessions

- AI and ML for Microcontrollers
- Writing Embedded Software with ChatGPT and Open.AI
- Tools for Machine Learning in Microcontrollers
- Training a Model for the STM32
- **Deploying Machine Learning Models**

1

Preparing the Model for Export

Exporting an STM32 Binary

 Data acquisition

 Impulse design

 Create impulse

 Spectral features

 NN Classifier


 Anomaly detection

 Retrain model

 Live classification

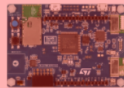
 Model testing

 Versioning

 Deployment

Build firmware

Or get a ready-to-go binary for your development board that includes your impulse.



ST IoT Discovery Kit



Arduino Nano 33 BLE Sense



Eta Compute ECM3532 AI Sensor



SiLabs Thunderboard Sense 2



Himax WE-I Plus



Nordic nRF52840 DK + IKS02A1



Nordic nRF5340 DK + IKS02A1



Linux boards

Exporting an STM32 AI pack

EDGE IMPULSE

- Dashboard
- Devices
- Data acquisition
- Impulse design
 - Create impulse
 - Spectral features
 - NN Classifier
 - Anomaly detection
- Retrain model
- Live classification
- Model testing
- Versioning
- Deployment**

DEPLOYMENT (BENINGO-PROJECT-1)

Deploy your impulse

You can deploy your impulse to any device. This makes the model run without an internet connection, minimizes latency, and runs with minimal power consumption. [Read more.](#)

Create library

Turn your impulse into optimized source code that you can run on any device.



C++ library



Arduino library



Cube.MX CMSIS-PACK



WebAssembly



TensorRT library

Build firmware

Or get a ready-to-go binary for your development board that includes your impulse.

Apply Optimization(s)

Available optimizations for NN Classifier

<p>Quantized (int8) ★</p> <p>Currently selected</p> <p>This optimization is recommended for best performance.</p>	<p>RAM USAGE 1.5K</p> <p>ROM USAGE 15.4K</p>	<p>LATENCY 1 ms</p> <p>ACCURACY 95.68%</p>	<p>CONFUSION MATRIX ?</p> <table border="1"> <tr><td>87.8</td><td>7.0</td><td>0</td><td>0</td><td>5.2</td></tr> <tr><td>0</td><td>99.2</td><td>0</td><td>0</td><td>0.8</td></tr> <tr><td>0</td><td>0</td><td>100</td><td>0</td><td>0</td></tr> <tr><td>-</td><td>-</td><td>-</td><td>-</td><td>-</td></tr> </table>	87.8	7.0	0	0	5.2	0	99.2	0	0	0.8	0	0	100	0	0	-	-	-	-	-
87.8	7.0	0	0	5.2																			
0	99.2	0	0	0.8																			
0	0	100	0	0																			
-	-	-	-	-																			
<p>Unoptimized (float32)</p> <p>Click to select</p>	<p>RAM USAGE 1.5K</p> <p>ROM USAGE 17.7K</p>	<p>LATENCY 1 ms</p> <p>ACCURACY 95.88%</p>	<p>CONFUSION MATRIX ?</p> <table border="1"> <tr><td>87.6</td><td>6.6</td><td>0</td><td>0</td><td>5.8</td></tr> <tr><td>0</td><td>100</td><td>0</td><td>0</td><td>0</td></tr> <tr><td>0</td><td>0</td><td>100</td><td>0</td><td>0</td></tr> <tr><td>-</td><td>-</td><td>-</td><td>-</td><td>-</td></tr> </table>	87.6	6.6	0	0	5.8	0	100	0	0	0	0	0	100	0	0	-	-	-	-	-
87.6	6.6	0	0	5.8																			
0	100	0	0	0																			
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Estimate for Cortex-M4F 80MHz (ST IoT Discovery Kit)

Build

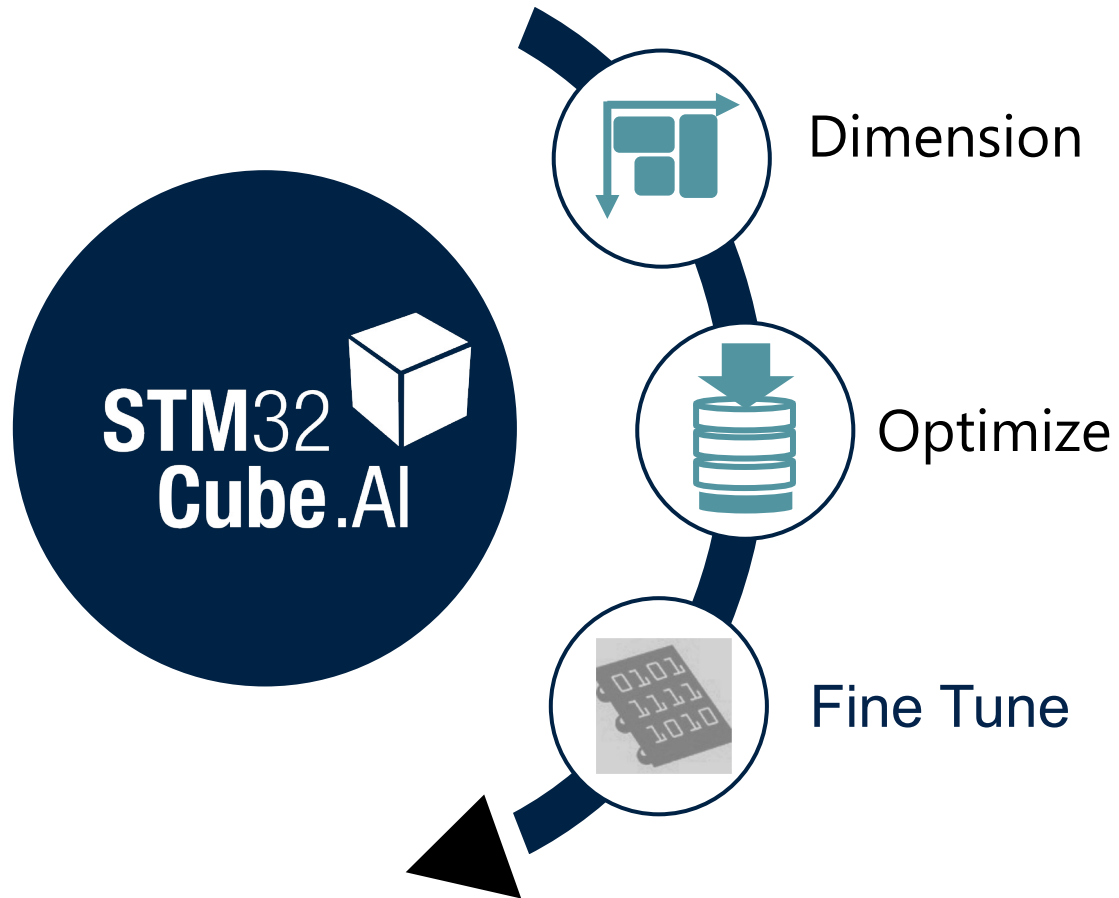
What method do you prefer for testing?

- Using the prebuilt binary
- Using the pack
- C++ library
- other

2

Importing the Model

STM32Cube.AI Overview



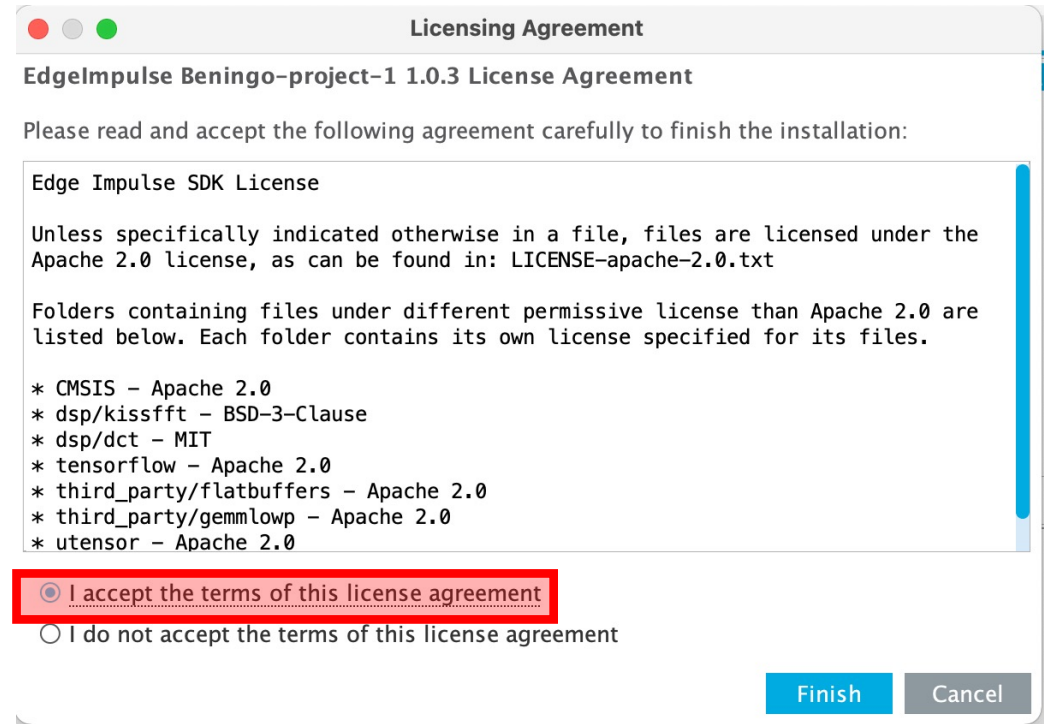
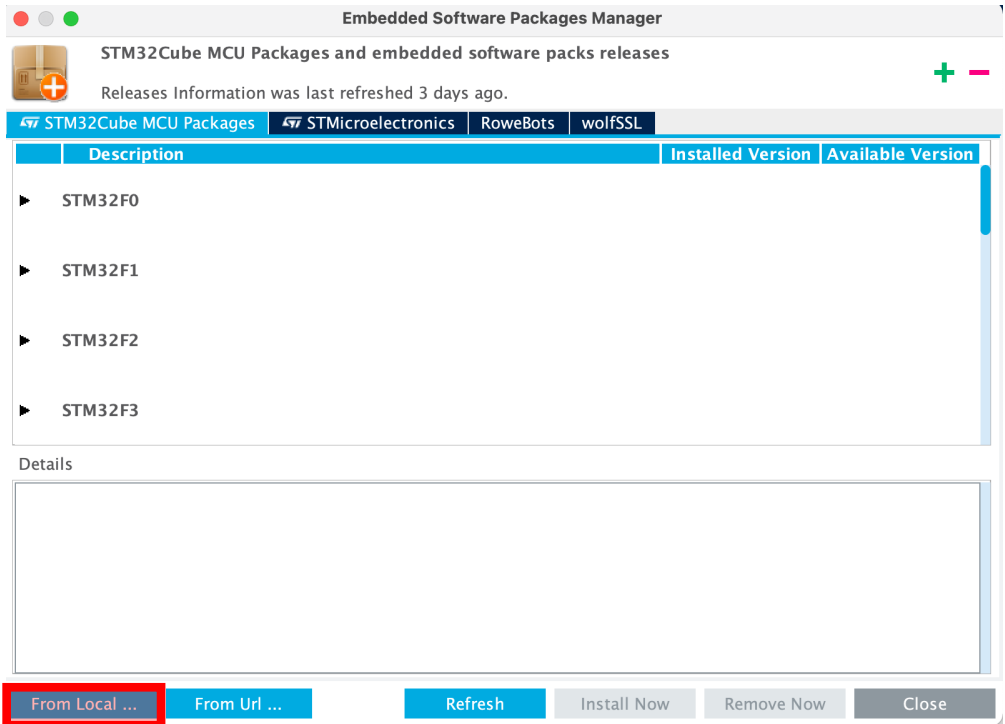
- ✓ Quickly assess model footprint requirements
- ✓ Select and configure MCU in STM32CubeMX
- ✓ Review model layers in STM32Cube.AI

- ✓ Generate C-code for pre-trained model
- ✓ Support quantized models to reduce RAM, flash and latency with minimal loss of accuracy
- ✓ Use light run-time libraries
- ✓ Optimize for performance

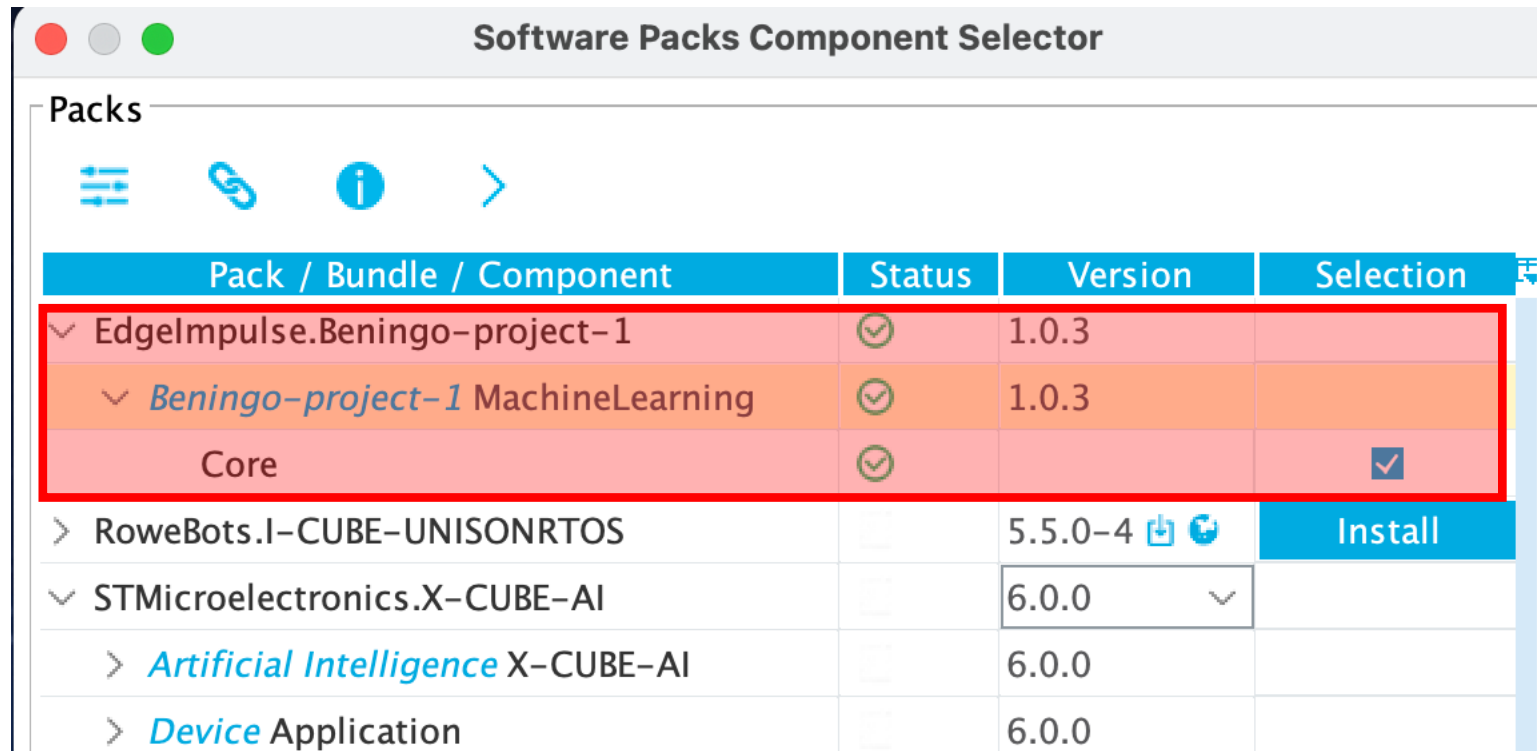
- ✓ Optimize memory allocation
- ✓ Fine control of weight mapping
- ✓ Split between internal and external memory
- ✓ Update model without full FW update




Install Pack into STM32Mx Project

Help -> Manage Embedded Software Packages

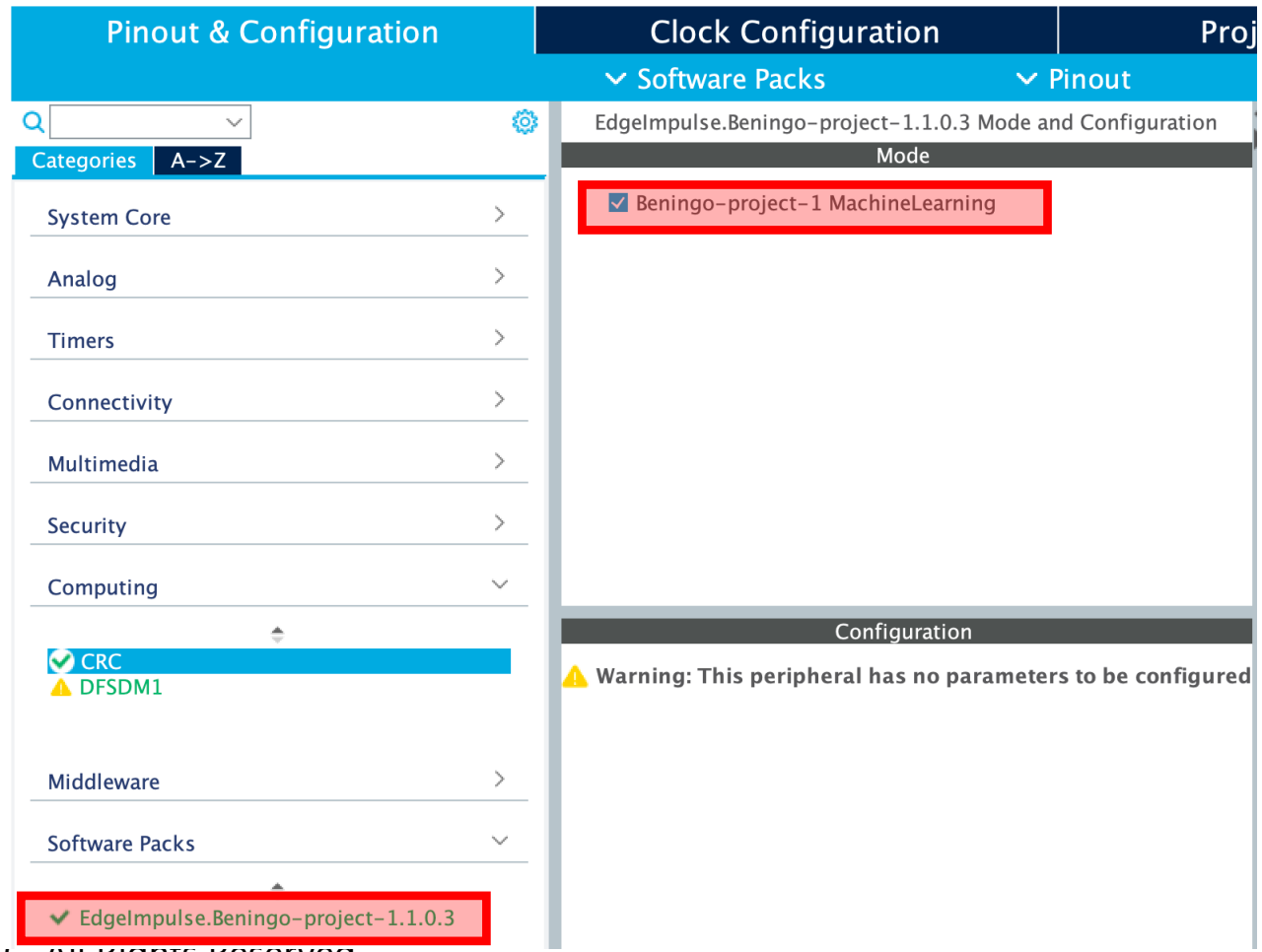


Install Pack into STM32Mx Project



Pack / Bundle / Component	Status	Version	Selection
EdgImpulse.Beningo-project-1	✓	1.0.3	
Beningo-project-1 MachineLearning	✓	1.0.3	
Core	✓		✓
> RoweBots.I-CUBE-UNISONRTOS		5.5.0-4  	Install
STMicroelectronics.X-CUBE-AI		6.0.0 	
> Artificial Intelligence X-CUBE-AI		6.0.0	
> Device Application		6.0.0	

Install Pack into STM32Mx Project



The screenshot shows the STM32CubeIDE interface with the following elements:

- Pinout & Configuration** tab is active.
- Software Packs** section is expanded, showing a search bar and a list of categories: System Core, Analog, Timers, Connectivity, Multimedia, Security, Computing, CRC, DFSDM1, Middleware, and Software Packs.
- The **Software Packs** category is expanded, showing the pack **EdgImpulse.Beningo-project-1.1.0.3** highlighted with a red box.
- The **Mode** section is expanded, showing the pack **Beningo-project-1 MachineLearning** checked with a red box.
- The **Configuration** section shows a warning: **Warning: This peripheral has no parameters to be configured**.

Install Pack into STM32Mx Project

Home > STM32L475VGTx - B-L475E-IOT01A1 > STM32_ML_Gesture.ioc - Project Manager > **GENERATE CODE**

Pinout & Configuration | Clock Configuration | **Project Manager** | Tools

Project

Code Generator

Project Settings

Project Name
STM32_ML_Gesture

Project Location
/Users/beningo

Application Structure
Advanced Do not generate the main()

Toolchain Folder Location
/Users/beningo/STM32_ML_Gesture/

Toolchain / IDE
STM32CubeIDE Generate Under Root

Modify, Build, Deploy
<https://bit.ly/32ESC3N>

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Running the Model

Running the Model

In a terminal, run the command: `edge-impulse-run-impulse`

```
Starting inferencing in 2 seconds...
Sampling... Storing in file name: /fs/device-classification.116
Predictions (DSP: 14 ms., Classification: 1 ms., Anomaly: 1 ms.):
Circle: 0.99609
Updown: 0.00000
Wave: 0.00000
anomaly score: -0.026
Finished inferencing, raw data is stored in '/fs/device-classification.116'. Use AT+UPLOADFILE to send back to Edge Impulse.
```

```
Starting inferencing in 2 seconds...
Sampling... Storing in file name: /fs/device-classification.121
Predictions (DSP: 15 ms., Classification: 0 ms., Anomaly: 2 ms.):
Circle: 0.00000
Updown: 0.00000
Wave: 0.99609
anomaly score: -0.132
Finished inferencing, raw data is stored in '/fs/device-classification.121'. Use AT+UPLOADFILE to send back to Edge Impulse.
```

Running the Model

```
Starting inferencing in 2 seconds...
Sampling... Storing in file name: /fs/device-classification.118
Predictions (DSP: 15 ms., Classification: 0 ms., Anomaly: 2 ms.):
  Circle: 0.01172
  Updown: 0.98828
  Wave: 0.00000
  anomaly score: -0.141
Finished inferencing, raw data is stored in '/fs/device-classification.118'. Use AT+UPLOADFILE to send back to Edge Impulse.
Starting inferencing in 2 seconds...
Sampling... Storing in file name: /fs/device-classification.119
Predictions (DSP: 14 ms., Classification: 1 ms., Anomaly: 1 ms.):
  Circle: 0.21094
  Updown: 0.78906
  Wave: 0.00000
  anomaly score: -0.164
Finished inferencing, raw data is stored in '/fs/device-classification.119'. Use AT+UPLOADFILE to send back to Edge Impulse.
```

What methods can be used to improve classification ?

- Running average on the output
- Monitor the anomaly value
- Set a minimum classification percentage
- All the above
- Other

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Going Further



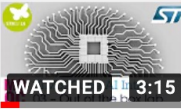

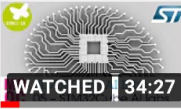
Next Steps

- Connect the output to a PWM LED channel
- Setup a DAC and drive an output voltage
- Configure the rate at which the inference runs (frequency control)
- Try and compare the Keras model behavior
- Improve the training model to provide a more accurate sine wave

Going Further

<https://bit.ly/3nf99EZ>

 FP-AI-SENSING1	 People activity recognition Audio scene classification
 FP-AI-NANOEDG1	 Condition-based monitoring
 FP-AI-VISION1	 Person presence detection Food classification
 FP-AI-FACEREC1	 Face recognition

1	 WATCHED 10:23	Introduction to STM32Cube.AI - 1 Marketing introduction STMicroelectronics
2	 WATCHED 5:24	Introduction to STM32Cube.AI - 2 Theory of AI STMicroelectronics
3	 WATCHED 3:15	Introduction to STM32Cube.AI - 3 Out of the box lab STMicroelectronics
4	 WATCHED 9:06	Introduction to STM32Cube.AI - 4 NN Model creation using Keras STMicroelectronics
5	 WATCHED 34:27	Introduction to STM32Cube.AI - 5 STM32Cube.AI labs STMicroelectronics

AI and ML Resources

- [Jacob's AI Blogs](#)
- [Jacob's CEC courses](#)
- [Jacob's ML Blogs](#)
- Embedded Bytes Newsletter
 - <http://bit.ly/1BAHYXm>

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