

ACETONE : Un outil open source de génération de code C prédictible à partir de modèles de réseaux de neurones

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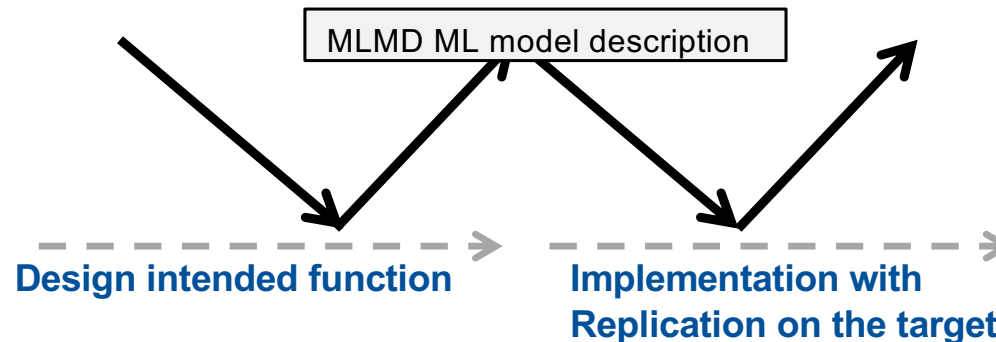
Focus Technique ANITI 2025

- **ML certification**
- ACETONE
- Integration ACETONE in AIDGE

Capture applicable regulatory requirements



- Documents
 - EASA concept paper
 - EUROCAE/SAE ED 324-ARP 6983
- Focus: (supervised) off-line trained ML models
- Outcome
 - “W Shape” Development Lifecycle
 - Semantic preservation of the trained ML model in the implementation
 - SW Implementation compliant with DO 178



Iryna de Albuquerque Silva (ANITI PhD 2021 – 2024) “Certifiable and efficient implementation of neural networks on embedded safety-critical real-time systems.”

- Developed an approach that enables the implementation of the inference model in compliance with avionics requirements
 - precise description of the inference model’s operations (math level)
 - semantics preservation and traceability between the inference model description and the final executable
 - timing predictability of the implementation (WCET estimation with OTAWA)
 - efficient usage of the available resources
- **ACETONE “Avionics C code generator for Neural Networks”**

<https://hal.science/tel-05018032v1>

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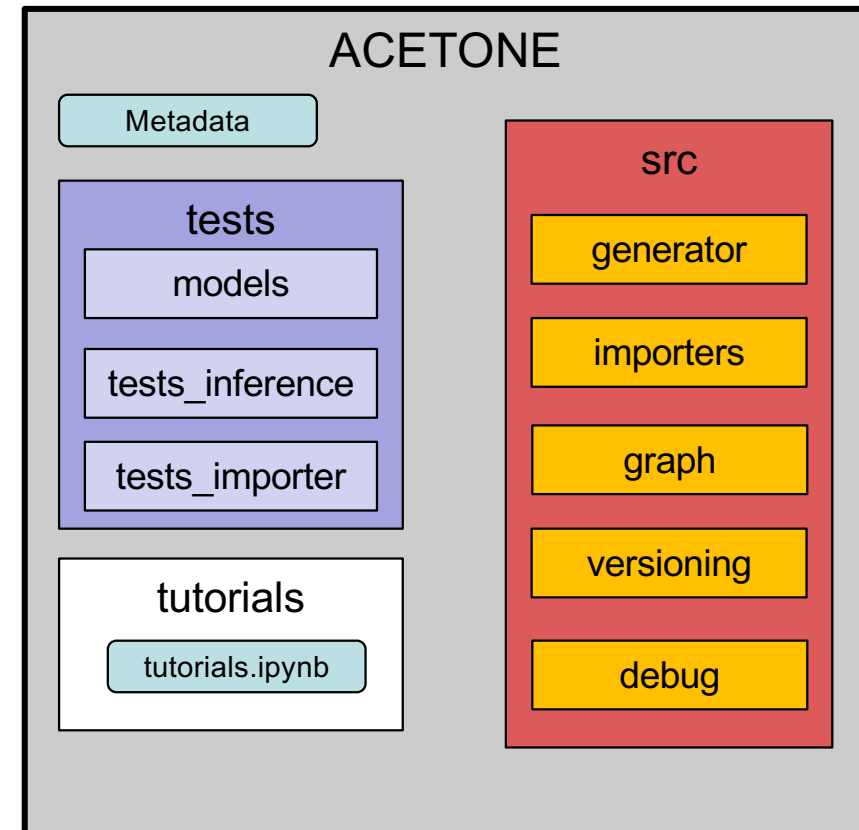
ACETONE repository



<https://github.com/onera/acetone>

Python package

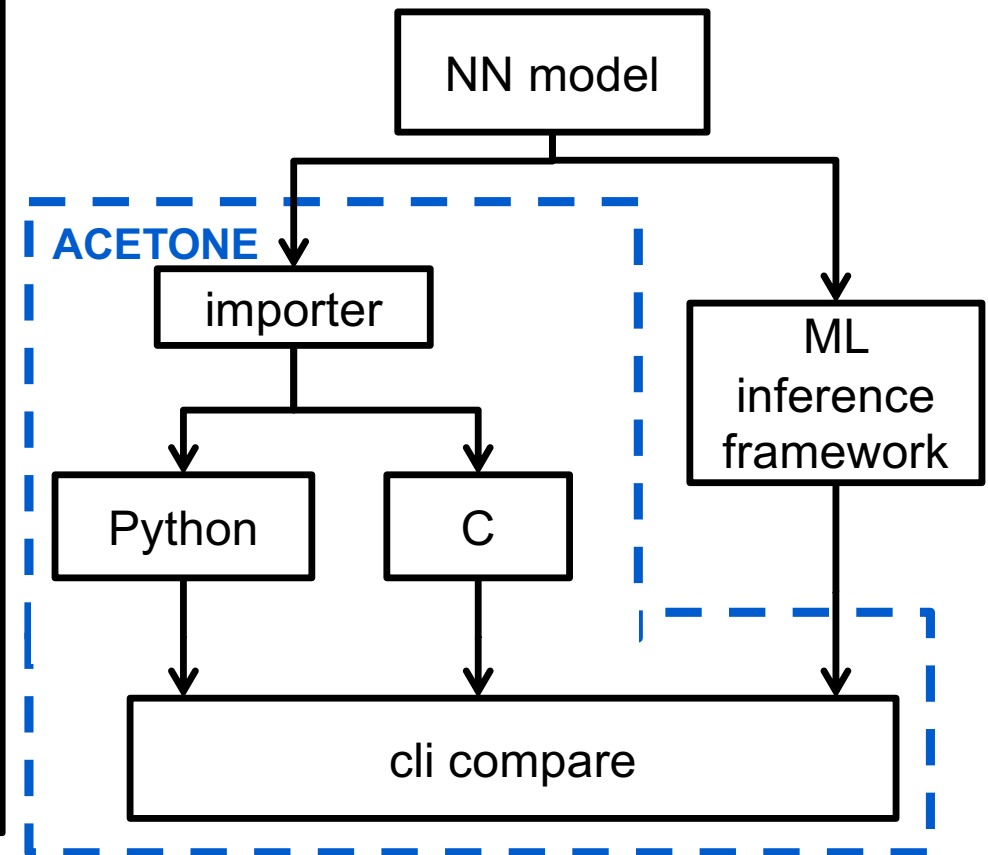
- Source code:
 - Layers description and code generator
 - Formats importers
 - Graph interpreter
 - Version handler
 - Debug tools
- Unit tests:
 - Model bank
 - Unit tests
- Tutorials



Tutorial 1



- Importers
 - *NNet*: initial format for fully connected NN, used by Reluplex git
 - *ONNX*: Widely used in industry
 - *H5*: Keras format
 - *JSON*: home made format (inspired from Keras)
- Test with ACETONE
 - ACETONE Python's reference execution
 - Training framework (e.g. ORT, Keras ...)
- **Objective:**
 - **Check semantic preservation**



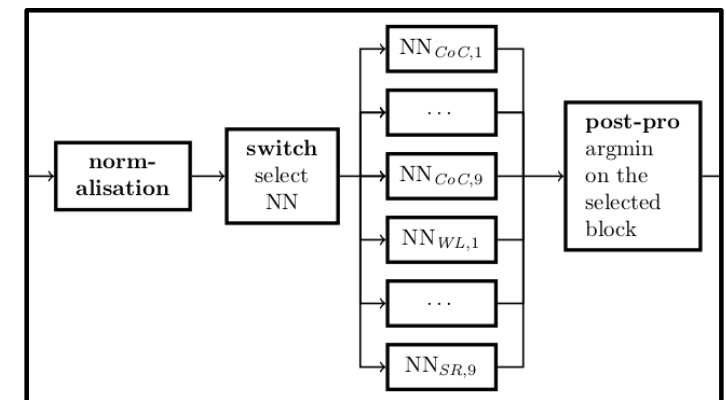
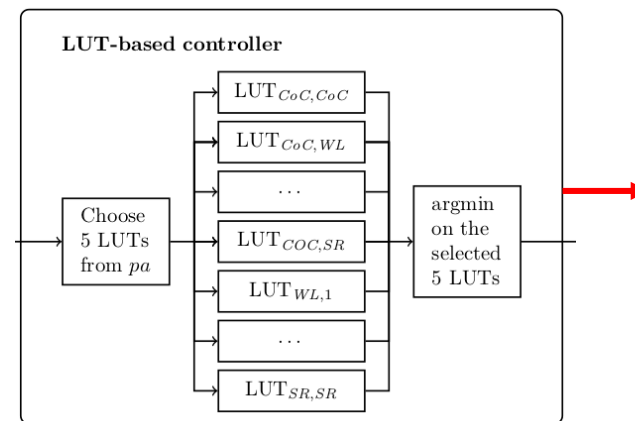
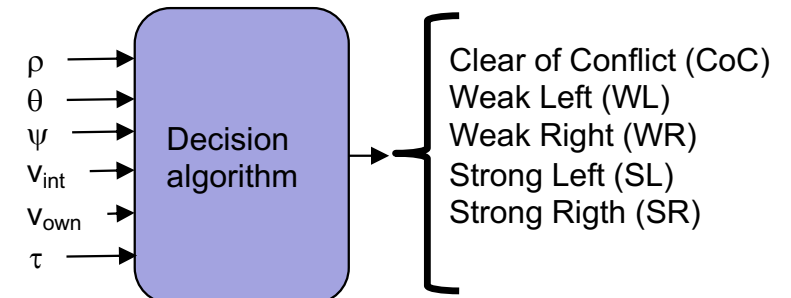
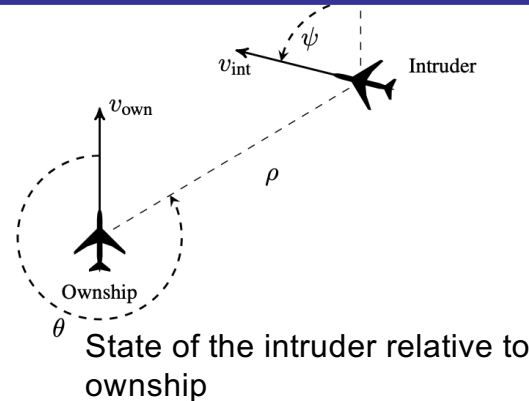
Use case for tutorial 1 – ACAS Xu



Avoidance Collision System for vertical and horizontal cooperative and non-cooperative avoidance (Multi-Intruders)

Intended function “any intruder should not enter in the ownship envelope”

- Interest: Gain in memory footprint (from 4Go to 3Mo)

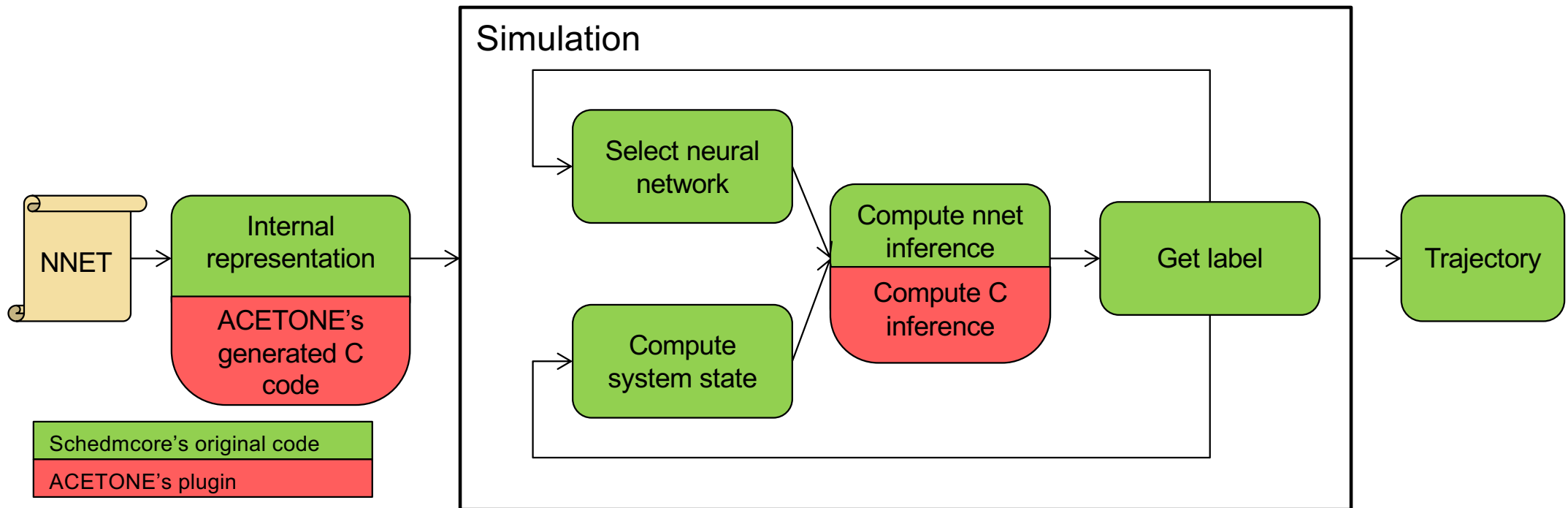


45 NN (fully connected NN with 6 layers of 50 neurons each)

Tutorial 1: simulation



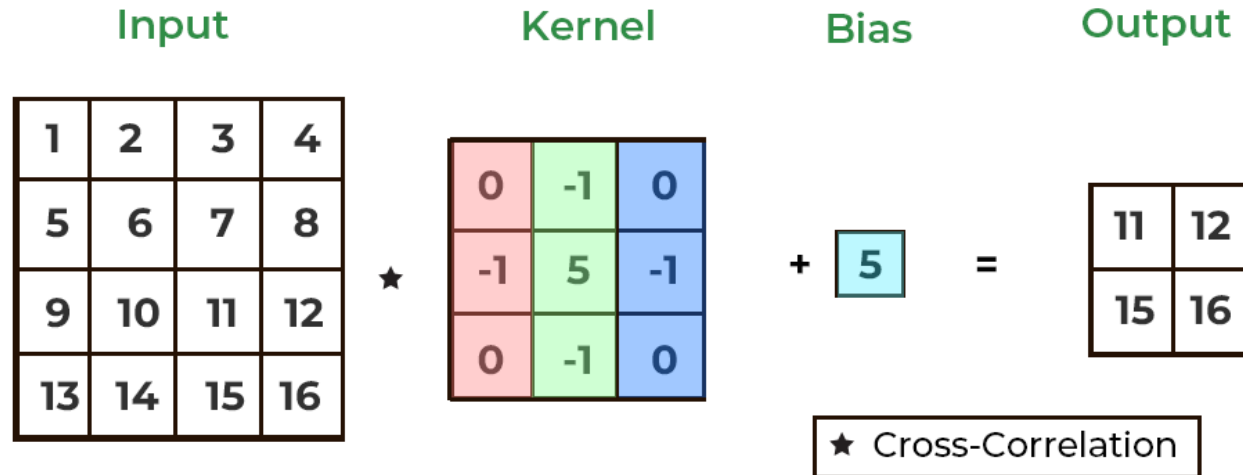
- Python Simulator developed by Arthur Clavière (PhD Collins)
https://svn.onera.fr/schedmcore/branches/ACAS_CaseStudy/
- Extended to link with C code generated by ACETONE



Tutorial 2: convolution implementation(s)



- Reminder on convolution



- Naïve implementation 6 loops
- Often more efficient: matrix based multiplication (GEMM)
- Objective: optimisation (better use of resources)**

Figure excerpt from <https://www.geeksforgeeks.org/apply-a-2d-convolution-operation-in-pytorch/>

Tutorials 3&4



- Tutorial 3: debug mode
 - Automatic adding of outputs within the neural network graph
- Tutorial 4: advanced use of ACETONE
 - Adding a new variant of convolution
 - Use of debug mode to check whether the new implementation preserves the semantic

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- ACETONE
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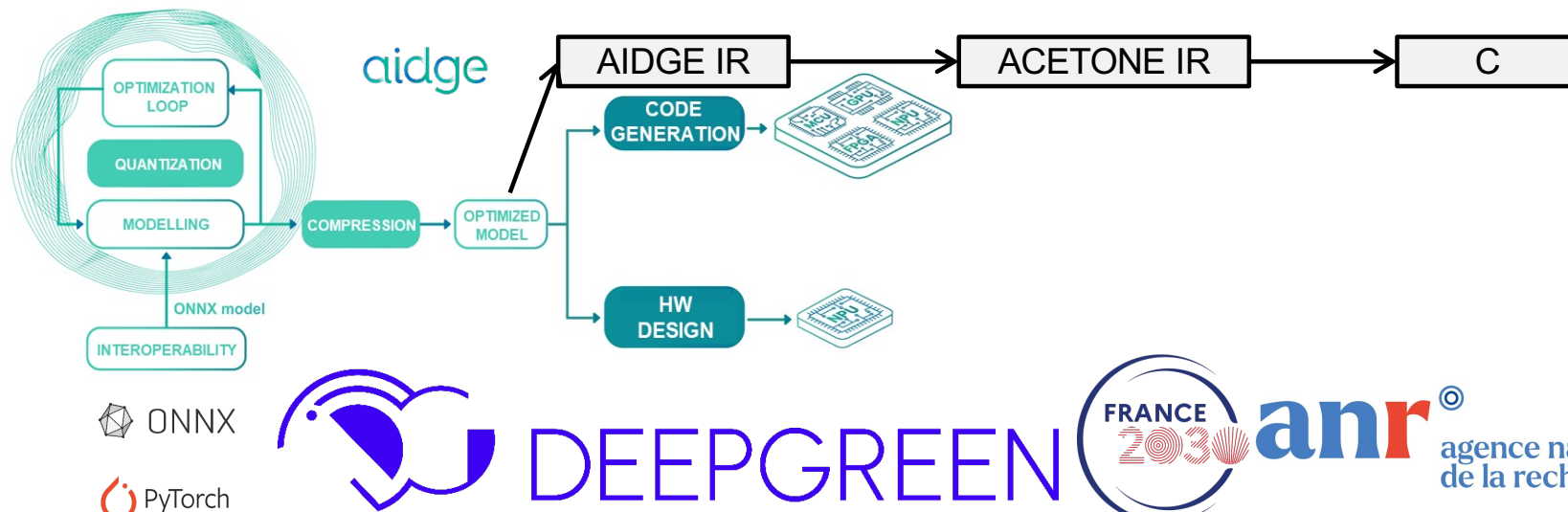
AIDGE – ACETONE as a C back-end



AIDGE Embedded AI Deep Learning Platform

- Developed by CEA
- Code generation back-end: OpenMP, OpenCL, Cuda, cuDNN and TensorRT

<https://eclipse.dev/aidge/>



ONNX

PyTorch



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Question?



Thank you for your attention