# tinyML. EMEA

Enabling Ultra-low Power Machine Learning at the Edge

tinyML EMEA Technical Forum 2021 Proceedings

June 7 – 10, 2021 Virtual Event





Neuton.ai

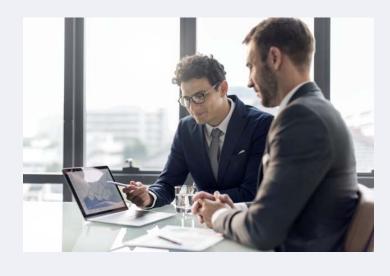
A Novel Approach to Building Exceptionally Tiny Models



#### **Our TinyML Community**







#### **TECH RESOURCES**

New to Edge Computing / Tiny ML Experienced Tech Resources

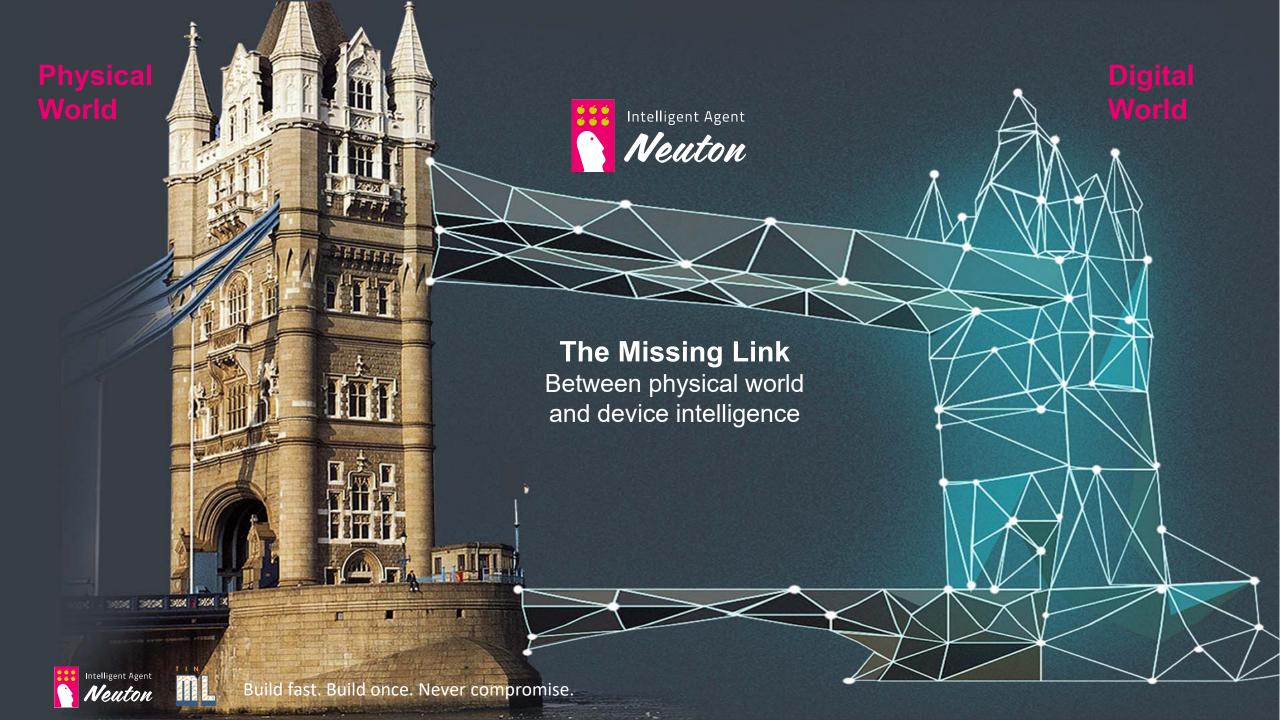
#### **DATA SCIENCE**

New to AI /ML Experienced Data Scientists

#### **BUSINESS**

Business Users Executives

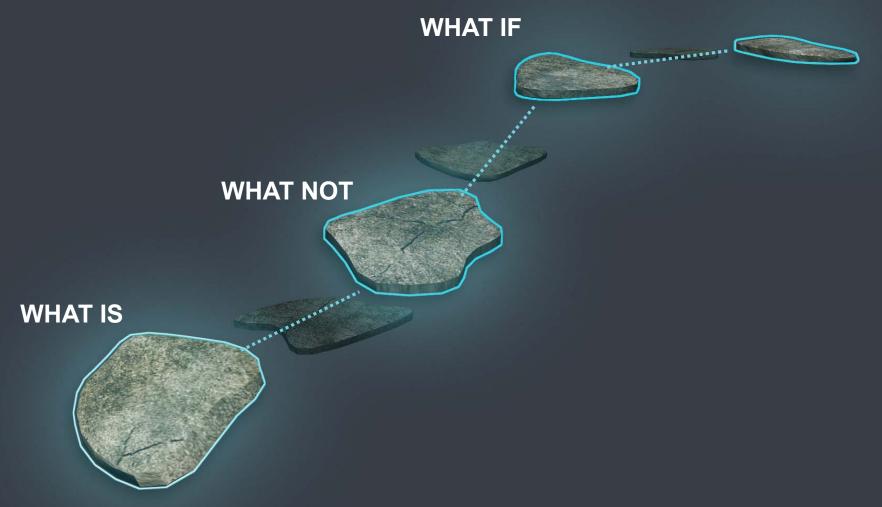






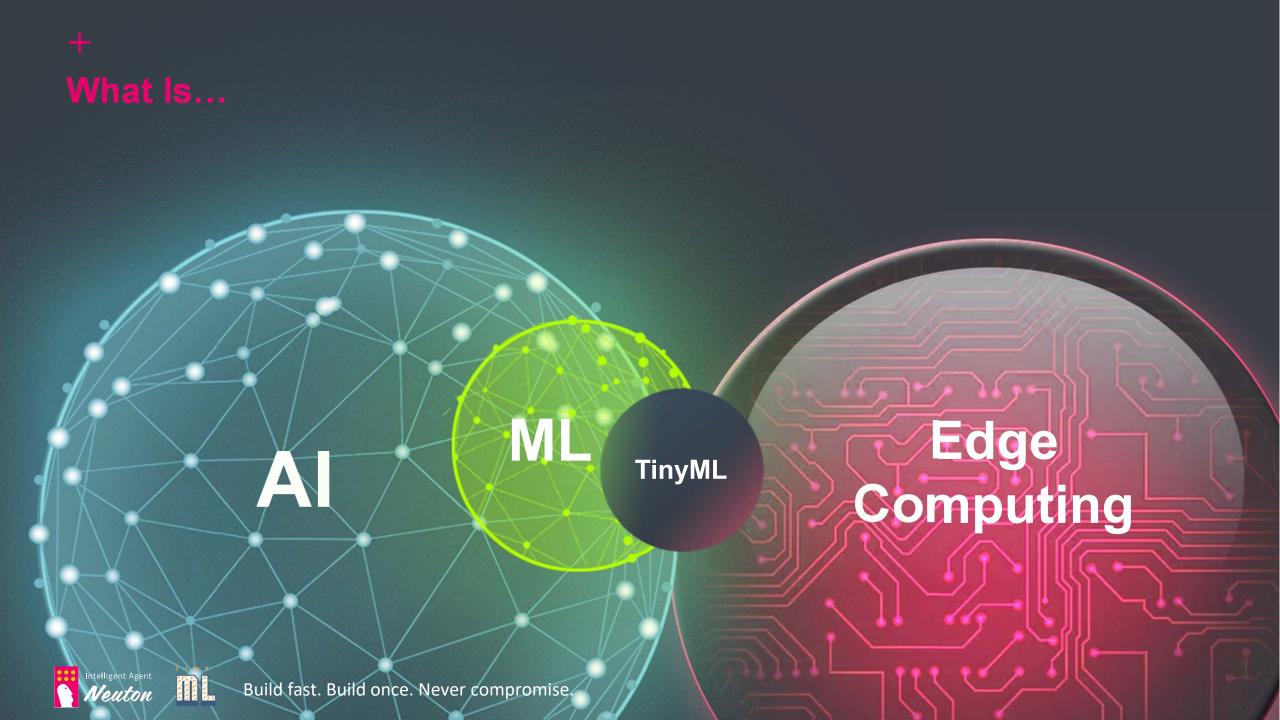
#### Our Journey Today

#### WHAT NEXT











#### What Not...

#### Main Barriers and Challenges for TinyML



Limited knowledge and availability of resources in Machine Learning and software development



The challenge of integrating large ML models into edge devices



Barriers of evaluating the quality and understanding the logic behind the model







#### What If You Could...

	build a model without having any technical expertise?	
	create a model in 3 clicks?	
	find the most optimal model in one iteration?	
	produce models up to 1,000 times smaller than TensorFlow lite?	
	run inferences up to 60% faster?	
	eliminate the need to perform compression and not compromise accuracy?	
	explain why your model makes every single prediction?	
	accelerate your time to market by 85%?	
	build fast, build once and never compromise accuracy or your business requirements?	





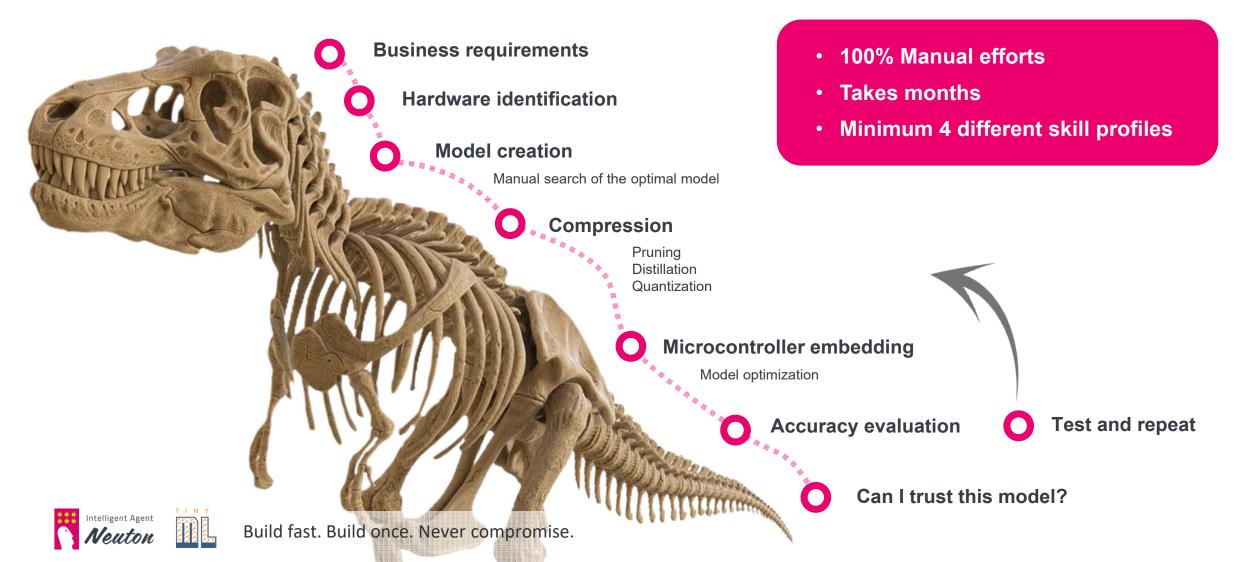
## What Next?



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#### **Traditional Approach**

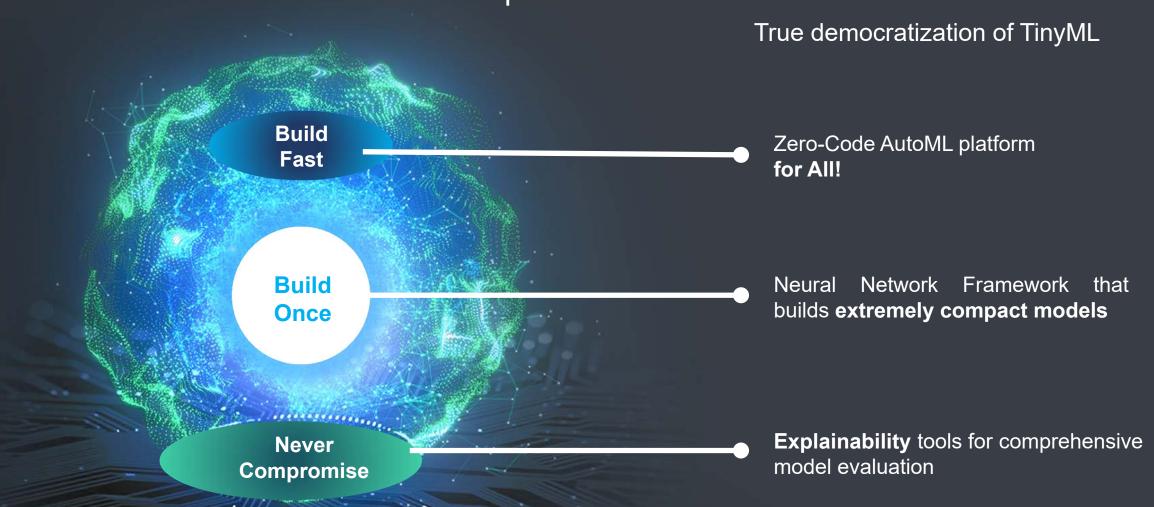
**Embedding Models to Edge Devices** 



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#### Neuton Approach

Build Fast. Build Once. Never Compromise





Build fast. Build once. Never compromise.



#### Model Size or Model Quality

# NEUTON'S MODELS ARE: up to





- Smaller in size (Kb)
- Faster inference

In comparison to TensorFlow and other algorithms

Our unique framework allows creation of a

neural network structure of:

THE MOST OPTIMAL SIZE & ACCURACY





How Do We Create Compact Models without Comprising Accuracy?











Selective approach to the connected features



Automatic neuron-by-neuron network structure growth



No manual search for neural network parameters



Unique patented global optimization algorithm



Permanent cross-validation





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## How We Got to Today Building Neural Networks



- Manual random search of too many variables:
  - Seed
  - Number on Neurons
  - Number of Layers
  - Activation Function (Sigmoid, ReLU etc)
  - Learning Rate
  - Number of epoches
  - Cross Validation Folds
  - Dropout
- Predetermined architecture (structure) defined by the researcher and the method of stochastic gradient descent
- Only neuron parameters undergo optimization the architecture remains predetermined
- Unnecessary growth of network size



#### **Our Future Today**

#### Efficient Global Optimization Algorithm

#### Our Framework uses a new efficient global optimization algorithm

- It is not based on back propagation of errors or stochastic gradient descent
- No problems of local extremes and plateaus
- Helps significantly improve each neuron's efficiency and to reduce the network's volume as a result
- Has enormous potential for parallelizing
- Allows for permanent cross-validation
- Automatic neuron-by-neuron network growth with overfitting control
- Dynamic growth of the network until it achieves its maximum generalization ability
- Learning the parameters of each neuron also allows for a significant reduction in the volume of the network







#### **Our Future Today**

Neuron by Neuron Structure Growth

#### Neuton uses neuron-by-neuron network structure growth

- Optimization occurs neuron by neuron at each iteration
- Eliminating the necessity of a Neural Network architecture search
- Creates minimum-size neural networks, with no loss of accuracy
- Neuton does not reduce the model size after creation (quantization, pruning, distillation)
- Accuracy is not compromised by compactness

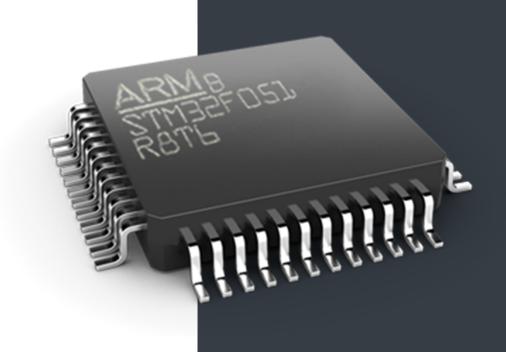


#### **Embedding into Edge Devices**

#### Neuton Meets the World

Neuton's models can be built into microcontrollers even with the following characteristics

- Energy 10s-100s mAh
- Processor < 100 MHz</li>
- Memory < 100 Kb</li>



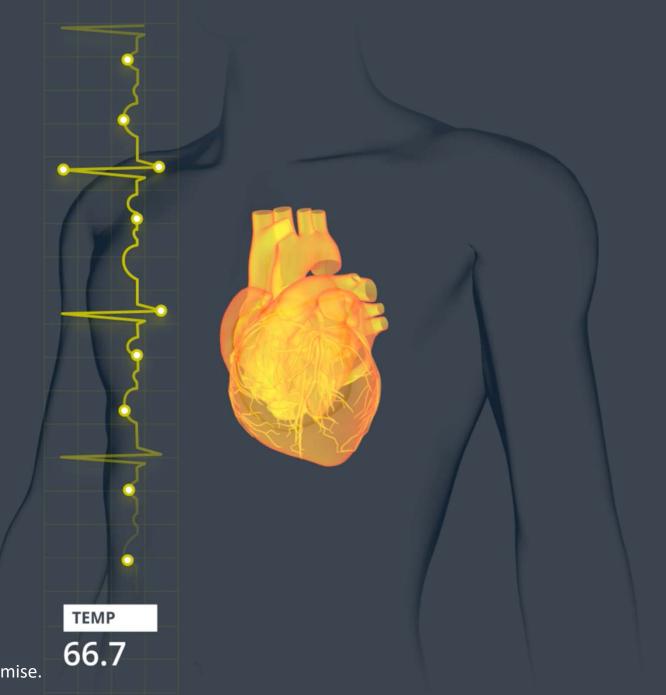




# **Embedding into a Microcontroller**Neuton in Action

Determine cardiac arrhythmias case study











#### **Neuton vs. TensorFlow Lite**

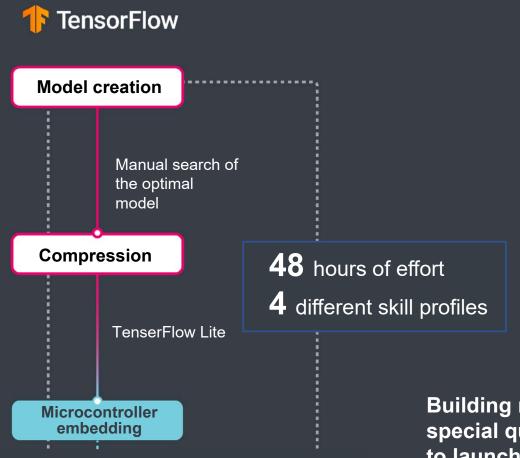
	Neuton	TensorFlow Lite	TensorFlow	Edge Impulse	
AUC	0.966	0.958	0.965	0.852	
Size, Kbytes	0.7	6.93	33.04	-	10 times smaller
Coefficients	253	N/A	2,414	N/A	
RAM Usage, Bytes	970	3,326		2,203	3.5 times less RAM utilization
Flash Usage, Bytes	24,532	56,628		47,448	
Inference time, Microseconds	1,987	4,979		6,556	2.5 times faster inference time

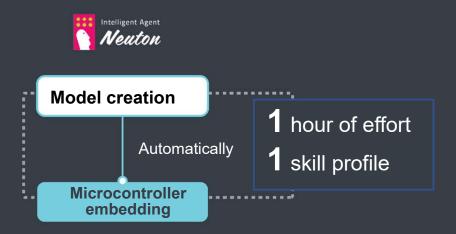




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#### Bringing the Future Today





Building models with Neuton does not require any special qualifications and requires minimal effort to launch similar projects



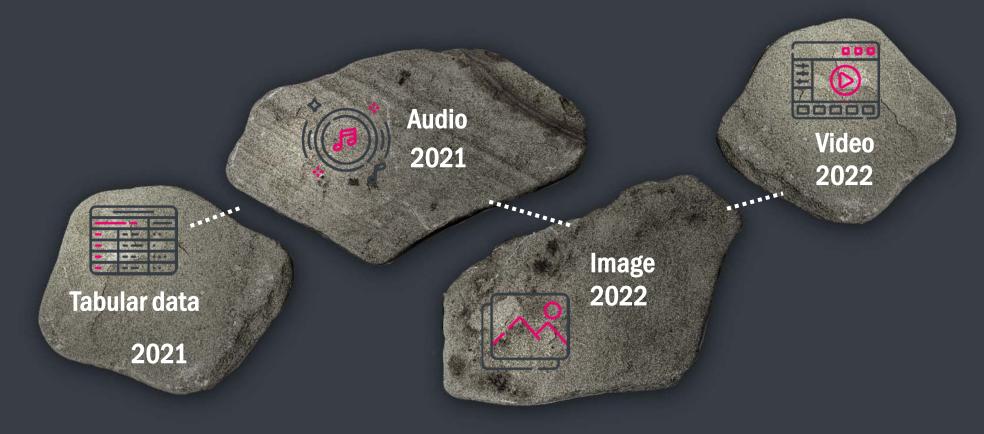


MicroOpsResolver



#### Neuton Today

Positioned for the Future



Regression, Time Series, binomial and multinomial classification including NLP





## A Model is Embedded into a Device, but What's Next?

In the context of making our devices Al driven, Explainability is essential





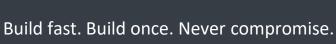
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#### The 4 W's We Should Ask Ourselves:

- 1 What does my data consist of?
- Where are the important features?
- Why does my model make certain Predictions?
- When should I consider retraining my model?







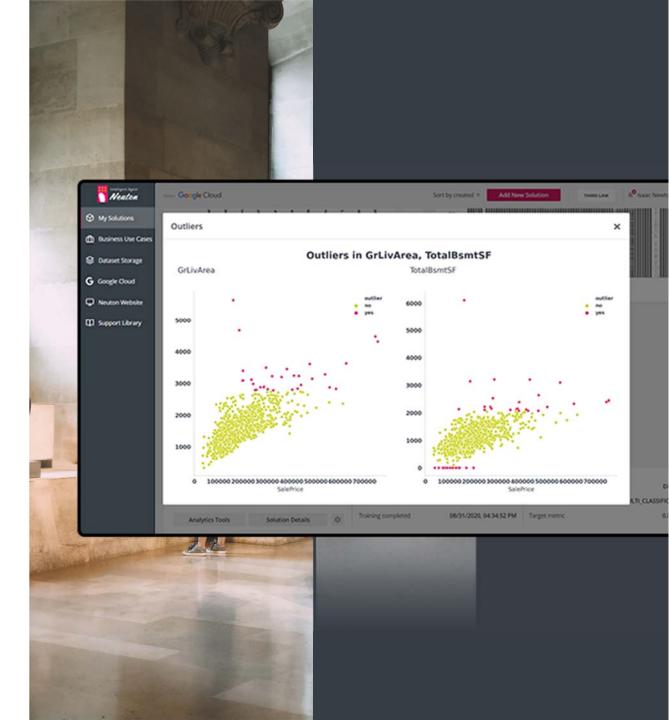




#### What

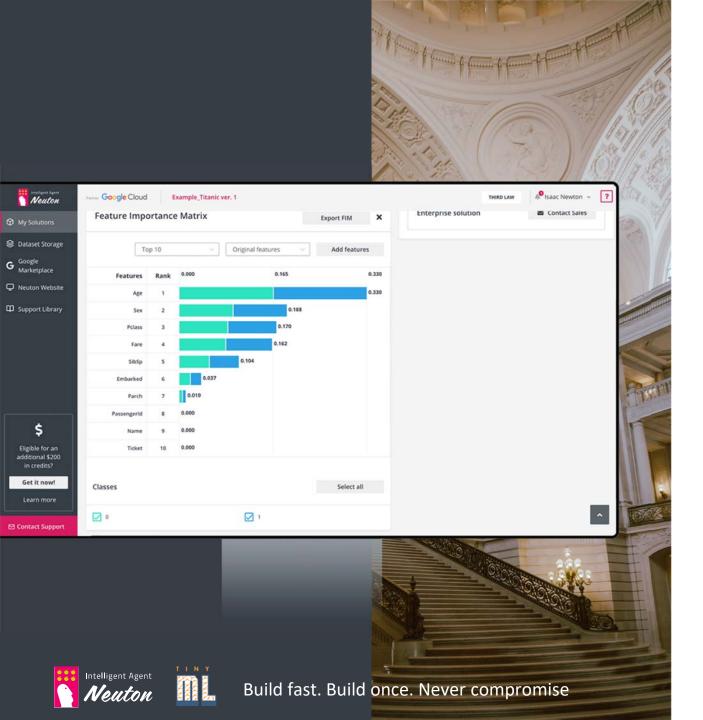
Does my Data Consist of?

Exploratory Data Analysis - graphical data analysis and the most important statistics









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#### Where

are the Important Features in my Dataset?

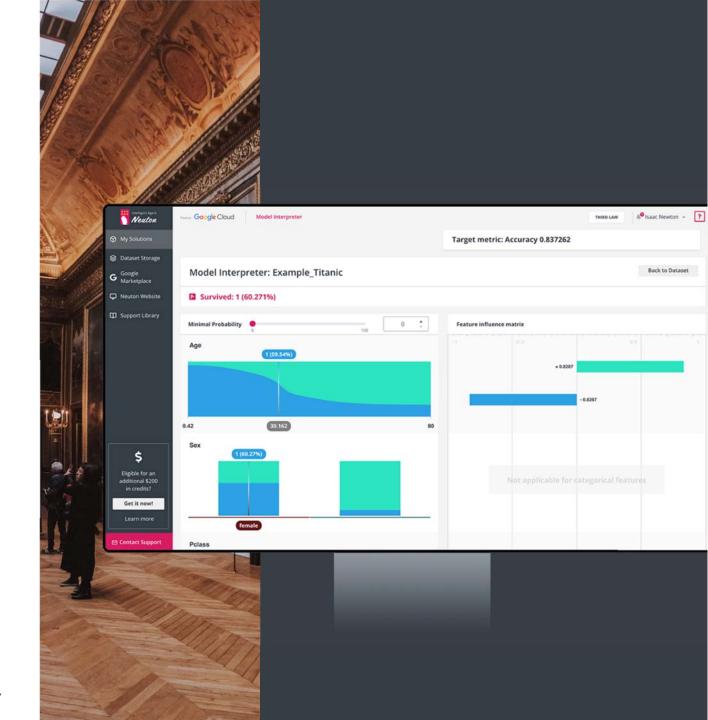
Feature Importance Matrix is a chart with features that had the most and least significant impact on the model prediction power.



#### Why

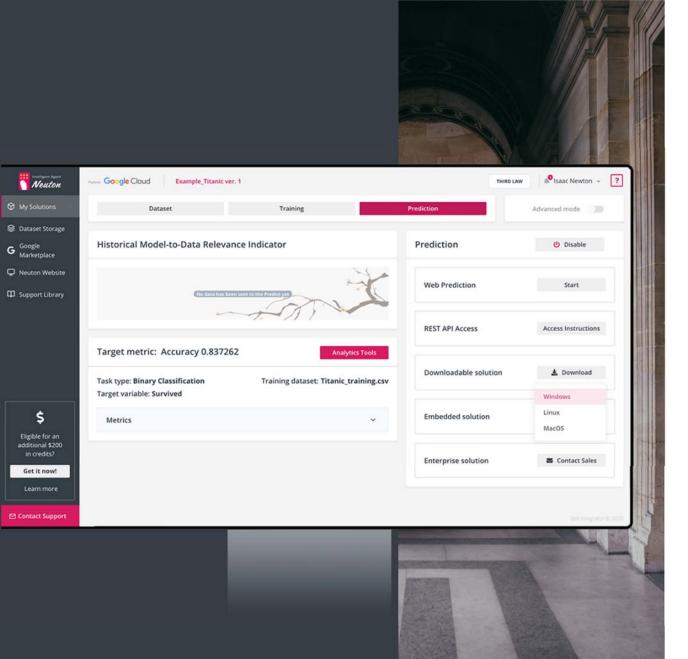
## Does my Model Make Certain Predictions?

The Model Interpreter allows you to see the logic, direction and the effects of changes in individual variables in the model.











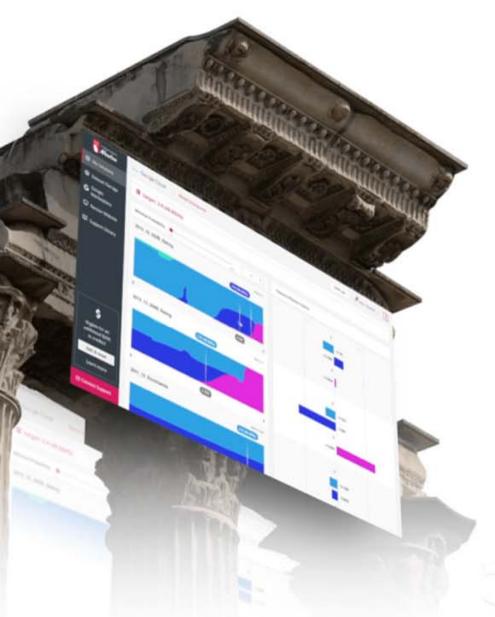
#### When

Should I Consider Retraining my Model?

Historical Model-to-data Relevance Indicator allows to manage a model lifecycle by signaling for models to retrain.



Build fast. Build once. Never compromise.



## **TinyML EMEA Technical Forum 2021**

## **Explainability Office:**

the next chapter for tinyML

Tuesday June 8, 2021 6:30 pm UTC +1



Join our Explainability office demo







#### **Your Future Today**

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- explain why your model makes every single prediction?
- accelerate your time to market by 85%?
- build fast, build once and never compromise accuracy or your business requirements?























## Build Fast. Build Once. Never Compromise.



Limited knowledge and availability of resources in Machine Learning and software development

Zero-Code AutoML platform for non-Data Scientists



The challenge of integrating large ML models into edge devices

Neural Network Framework that builds extremely compact models in one iteration



Barriers of evaluating the quality evaluating the quality and understanding the logic behind the model

**Explainability** tools for comprehensive model evaluation









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## **Premier Sponsor**



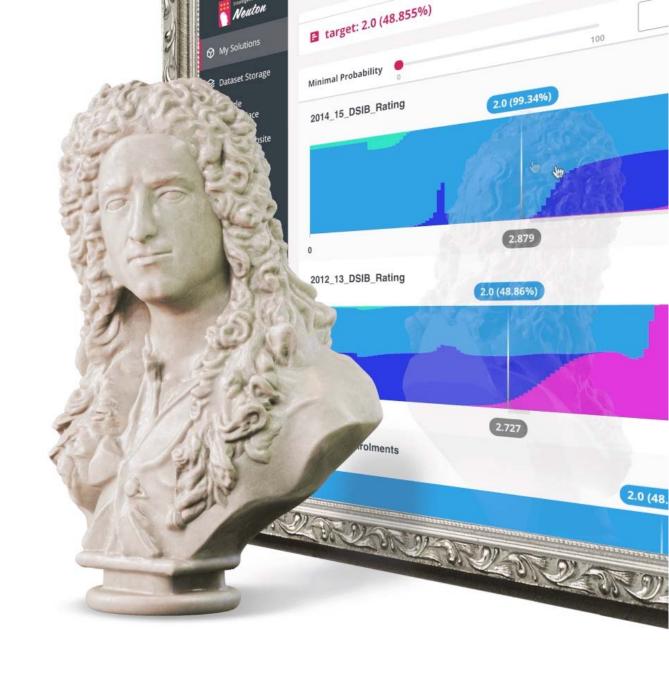
## **Automated TinyML**

Zero-code SaaS solution

Create tiny models, ready for embedding, in just a few clicks!

Compare the benchmarks of our compact models to those of TensorFlow and other leading neural network frameworks.

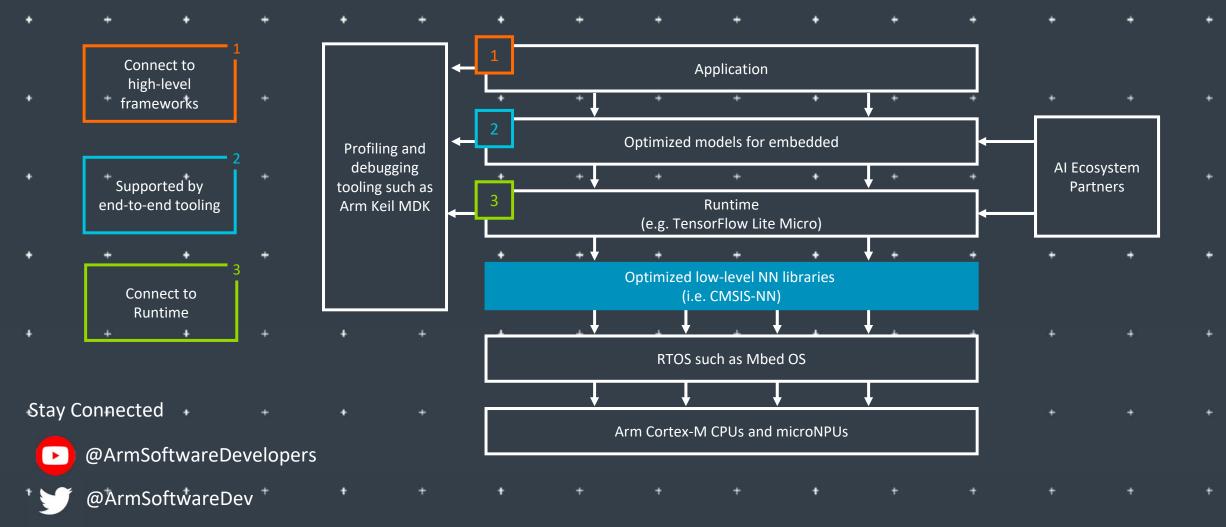
Build Fast. Build Once. Never Compromise.





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## Arm: The Software and Hardware Foundation for tinyML



Resources: developer.arm.com/solutions/machine-learning-on-arm

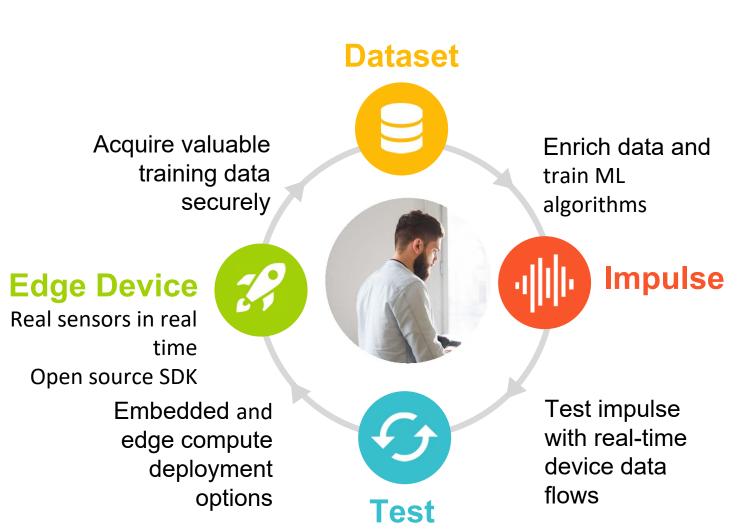


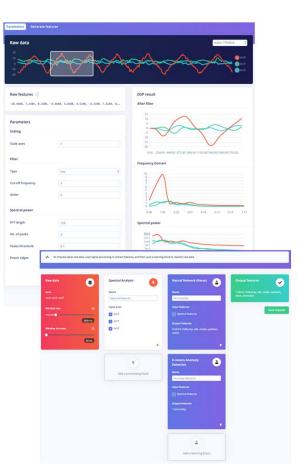
## TinyML for all developers











#### Qualcom Al research

### Advancing Al research to make efficient AI ubiquitous

#### Power efficiency

Model design, compression, quantization, algorithms, efficient hardware, software tool

#### Personalization

Continuous learning, contextual, always-on, privacy-preserved, distributed learning

#### Efficient learning

Robust learning through minimal data, unsupervised learning. on-device learning

A platform to scale Al across the industry



#### Perception

Object detection, speech recognition, contextual fusion

Reasoning

**Action** 

Reinforcement learning for decision making



Edge cloud







Mobile

IoT/IIoT







# SYNTIANT

<u>Syntiant Corp.</u> is moving artificial intelligence and machine learning from the cloud to edge devices. Syntiant's chip solutions merge deep learning with semiconductor design to produce ultra-low-power, high performance, deep neural network processors. These network processors enable always-on applications in battery-powered devices, such as smartphones, smart speakers, earbuds, hearing aids, and laptops. Syntiant's Neural Decision Processors<sup>TM</sup> offer wake word, command word, and event detection in a chip for always-on voice and sensor applications.

Founded in 2017 and headquartered in Irvine, California, the company is backed by Amazon, Applied Materials, Atlantic Bridge Capital, Bosch, Intel Capital, Microsoft, Motorola, and others. Syntiant was recently named a <a href="Maintenance-ES">CES® 2021 Best of Innovation Awards Honoree</a>, <a href="maintenance-shipped-over-10M">shipped over 10M</a> <a href="maintenance-units-worldwide">units worldwide</a>, and <a href="maintenance-units-units-worldwide">unveiled the NDP120</a> part of the NDP10x family of inference engines for low-power applications.

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# Add Advanced Sensing to your Product with Edge AI / TinyML

https://reality.ai







## Pre-built Edge Al sensing modules, plus tools to build your own

#### Reality AI solutions

Prebuilt sound recognition models for indoor and outdoor use cases

Solution for industrial anomaly detection

Pre-built automotive solution that lets cars "see with sound"

#### Reality AI Tools® software

Build prototypes, then turn them into real products

Explain ML models and relate the function to the physics

Optimize the hardware, including sensor selection and placement



## **Gold Sponsors**



Adaptive AI for the Intelligent Edge

Latentai.com

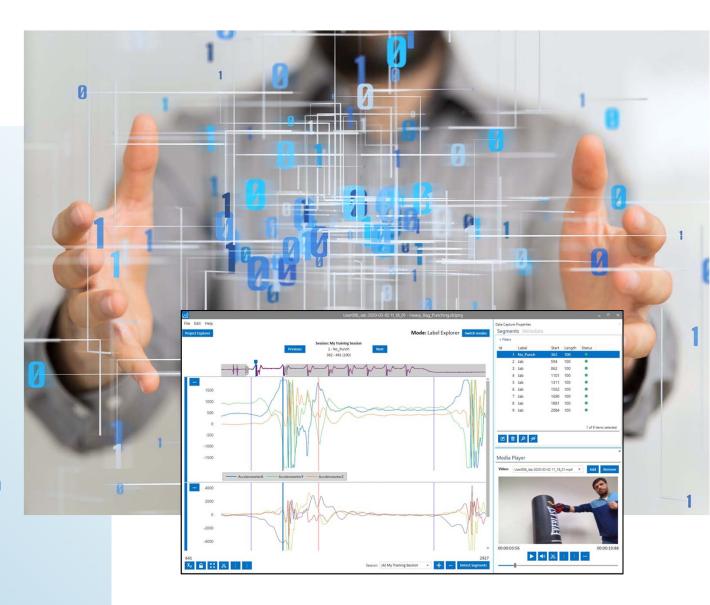


# **Build Smart IoT Sensor Devices From Data**

SensiML pioneered TinyML software tools that auto generate AI code for the intelligent edge.

- End-to-end AI workflow
- Multi-user auto-labeling of time-series data
- Code transparency and customization at each step in the pipeline

We enable the creation of productiongrade smart sensor devices.



sensiml.com



## **Silver Sponsors**

















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