

# Introducing Akida

**NEURAL PROCESSING UNIT** 

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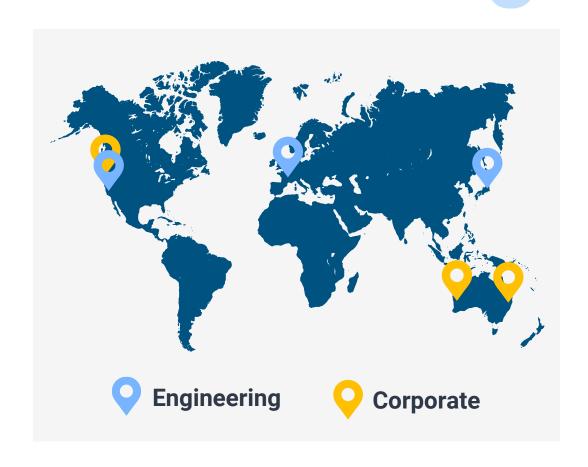
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## **About BrainChip**

BrainChip is a global technology company that has developed Akida, a revolutionary advanced neural networking processor that brings artificial intelligence to the edge in a way that existing technologies are not capable. The solution is high-performance, small, ultra-low power and enables a wide array of edge capabilities that include local training, learning and inference.

Biologically inspired, digitally engineered



## Akida NPU EDGE AI Technology

# Platform Akida

Object Detection/classification MobileNet/CFAR10 Etc Keyword Vibration other Cyber-Security

Data analytics

Vision
Camera frames
or DVS Events

Lidar 3D Point clouds Audio Data Internet Packets Data

Other Sensor Data Time series multivariate Data

Data to events Converter or Native events Data

CNN/DNN's
Inference
In Events Domain on EDGE devices

Native SNN
Learning and Inference
In Events Domain on EDGE devices

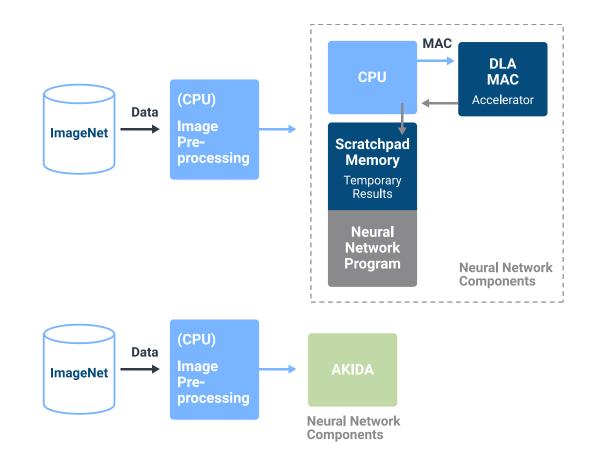
Configurable and Scalable
Spiking Events Based Neural Network Processing Technology

## Why Use Spiking Event Based Processing

- Spiking Events are non-zero activations
  - We only process Events
- \* Events are inherently sparse leading to reduced number of operations
- \* Quantizing weights and activations to 1,2 or 4 bits reduces memory requirements
- \* Each Layer computations are done on allocated NPUs and all NPUs run in parallel
  - \* All intermediate results stored on chip memory: Eliminates off chip memory access
  - NPUs communicate over mesh network : No need for external host CPU
  - \* All layers of the Neural Network run in parallel : Akida Runs the entire network
- \* Akida allows inference and incremental learning on Edge Devices within a power, size and computation budget

## Akida Architectural Differentiation

- Standard Deep Learning Accelerators (DLA)
  - \* Perform a MAC function on arrays of numbers.
  - \* The neural network is executed on the host CPU.
  - Intermediate results are stored in host CPU memory.
- \* Akida Event Based Neural network Processing
  - \* Runs the entire network on NPU cores
- Only the preprocessing part is the same in both technologies



## Akida NSoC: A Perfect solution for Edge Al Devices

- \* Single low-power, hardware platform to run:
  - Conventional CNN/DNN inference algorithms
    - \* On-chip incremental learning at EDGE devices
  - \* Native SNNs with event-based learning algorithm
  - \* Entire Multi-layer Neural Network runs on AKIDA Fabric
    - Preprocessed data in and Classified data result out in last layer
- \* Uses well-known Tensorflow Machine Learning ecosystem.
- \* Akida Technology offered as a NSoC Chip or Customized IP

## Akida NSoC Chip: Complete Al Edge Solution

Single Platform for CNN inference or Native SNN with on chip learning for EDGE Devices

#### **Data input Interfaces**

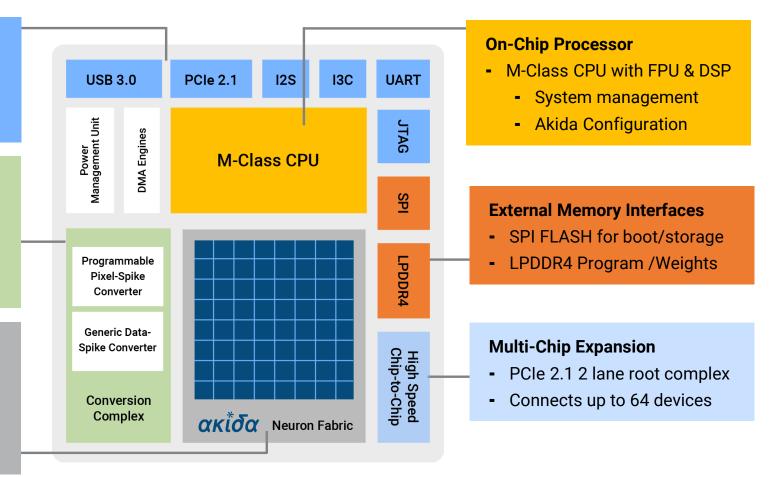
- PCI-Express 2.1 x2 Lane Endpoint
- USB 3.0 Endpoint
- I3C, I2S, UART, JTAG

#### **Conversion Complex**

- Pixel-Spike Converter
- SW Data-Spike Encoder
  - Any multivariable digital data
  - Sound, pressure, temp others

#### Flexible Akida Neuron Fabric

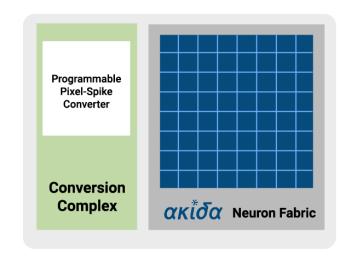
- Implements 80 NPUs
- All Digital logic with SRAM (8MB)
- Also Available as Licensed IP Core
- First Implementation : TSMC 28nm.



## Akida: Configurable Intellectual Property Blocks:

- Configurable number of NPUs
- \* Programmable Data to Spike Converter
- Event Based Processing Engines running on a single clock
- Configurable on-chip SRAM memory
- Runs full Native SNN in Hardware
- Runs full converted CNN/DNNs in Hardware
- \* On chip communication Mesh network
- Process technology independent platform
- Network Size customized to application needs

#### IP 80 Cores Single Converter

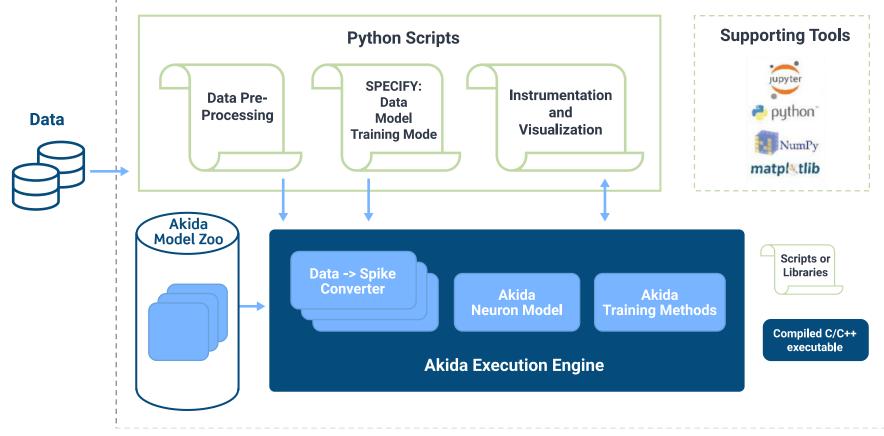


#### IP 16 Cores Single Converter



## Akida

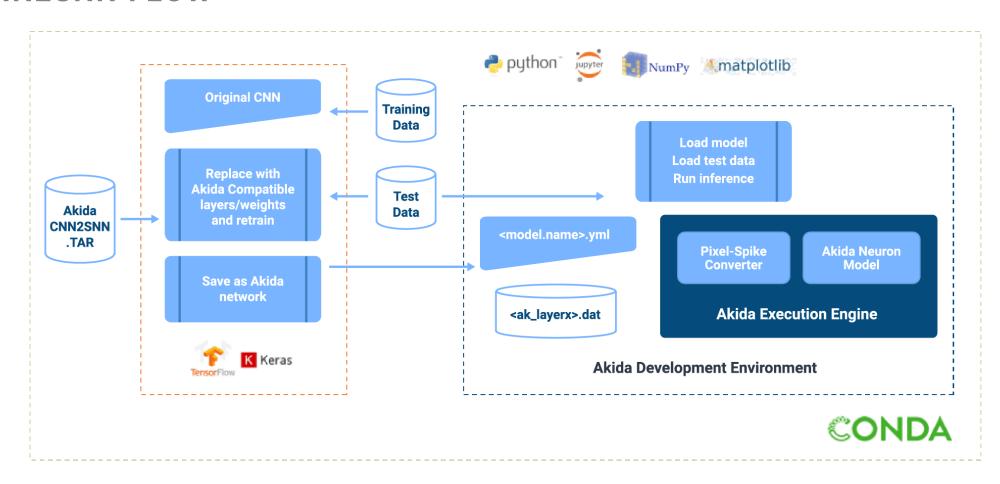
## **CHIP SIMULATOR**





## Akida

#### **CNN2SNN FLOW**



## Akida: Mapping CNN to SNN for Inference

#### \* MobileNet V1

- 30 layers with many MAC operations
- \* Entire network mapped onto separate Akida NPUs (4,208,224 parameters)
- \* All layers run in parallel storing results within the distributed SRAM memory on chip
- Performance from the chip simulator

\* Speed: 80 FPS

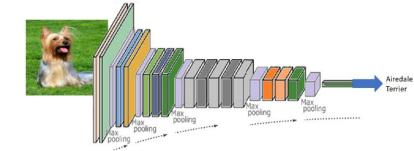
Power: 434 mw in 28nm

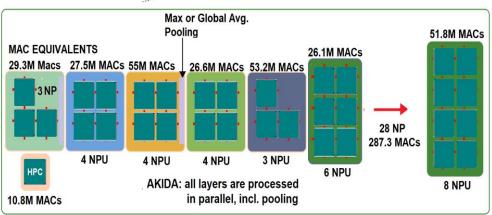
Performance: 184 FPS/W

\* 30 FPS @ 157 mw in 28 nm

#### \* Power Performance Scales with Process node

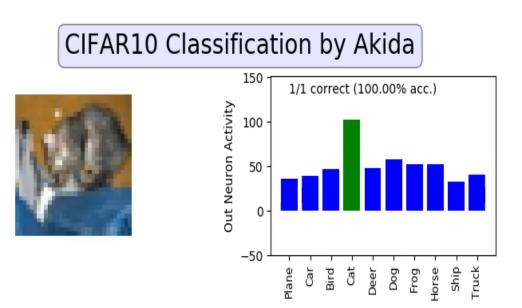
All power estimates are for Akida NPUs only

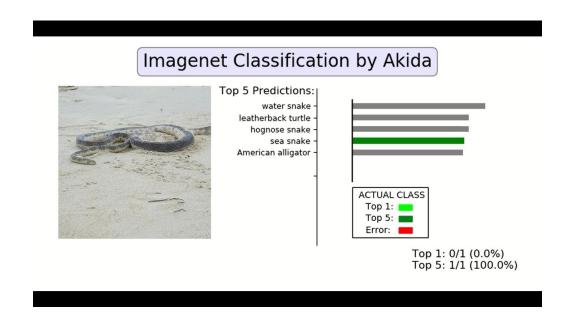




## Akida: Edge Inference Simulated Demo

## MobileNet and VGG-16 ported to AKIDA

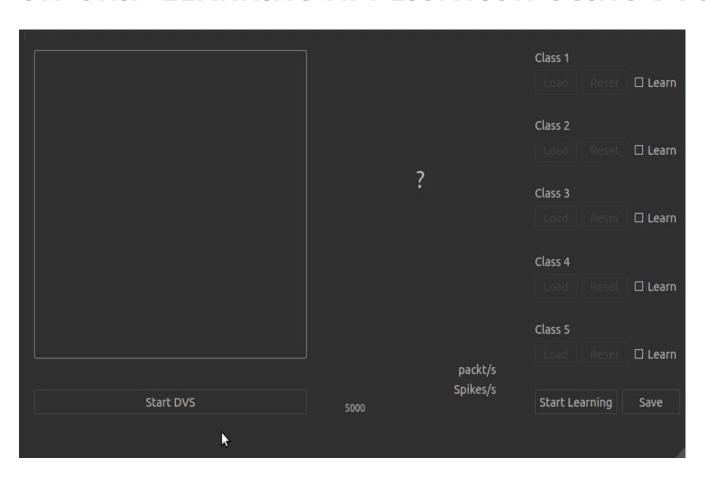




What does this mean? Any visual classification task that is done by a MobileNet or VGG DNN can be ported to AKIDA

## Native learning on-chip

#### ON-CHIP LEARNING APPLICATION USING DVS CAMERA

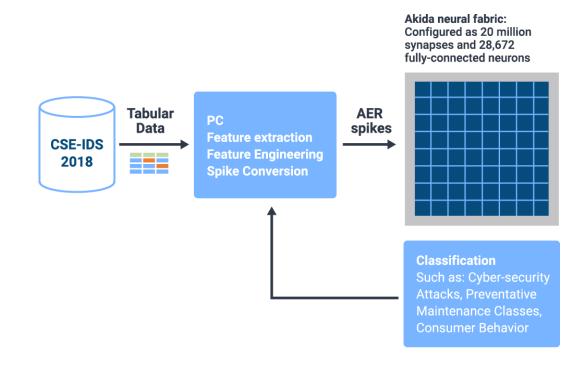


- Akida trained on hand-gestures using event-based camera (DVS)
- Real time learning and detection of patterns
- Uses our proprietary unsupervised learning rule

layer	filters	size	input	output
	(pre- processing)		128x128x1	64x64x1
conv	4	3x3 /1	64x64x1	64x64x4
max pool		3x3	64x64x4	22x22x4
conv	64 x 5 classes	7x7 /4	22x22x4	22x22x(64x5)
global max			22x22x(64x5)	1x1x(64x5)
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## **Akida Native SNN Mode**

- Example network is a cyber-security classification task using the CSE-IDS-2018 database (220.8 Gb)
- \* 15 classes (including 'normal')
- \* Training time is 2.2 hours, single epoch, including preprocessing on a simple i7 laptop with no GPU
- \* Automatic labeling of active neurons extracted from database
- \* Accuracy: 98% on test set
- Power: 20 mW at 30,000 inferences/s or 120 mW at 160,256 inferences/s
- 1 inference is one set of data points converted to spikes



## Akida Summary - A Holistic Approach to Edge Al

- Changing the game for Edge AI
  - \* Run Full network on the HW efficiently
- Game changing FPS/W
  - \* Parallel execution, lower power and real time performance
- Lower minimum power
  - Avoid external memory accesses, reduce bit computation
- \* Fewer TOPS required for delivering the FPS
- Lower memory requirements, power and size
  - Quantization to 1,2,4-bit weights

