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Miniature dreams can come true...

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Making Edge AI a Reality



A Complete Edge ML Company Delivering **Deep Learning** Solutions for **Always-On** Devices

Syntiant Complete ML solution

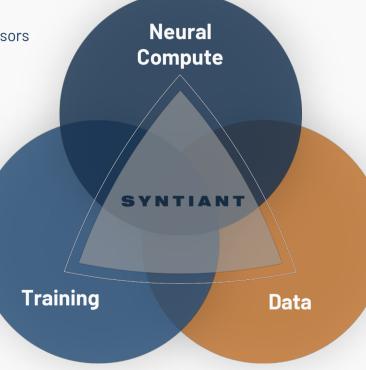
Neural Compute

Syntiant Neural Decision Processors vs. current MCUs:

- ✓ 100x efficiency improvement
- ✓ 30x increase in throughput
- \checkmark 1/2 the die size

Training

Syntiant Training Pipeline delivers turn-key models for the mass market



Easy to use

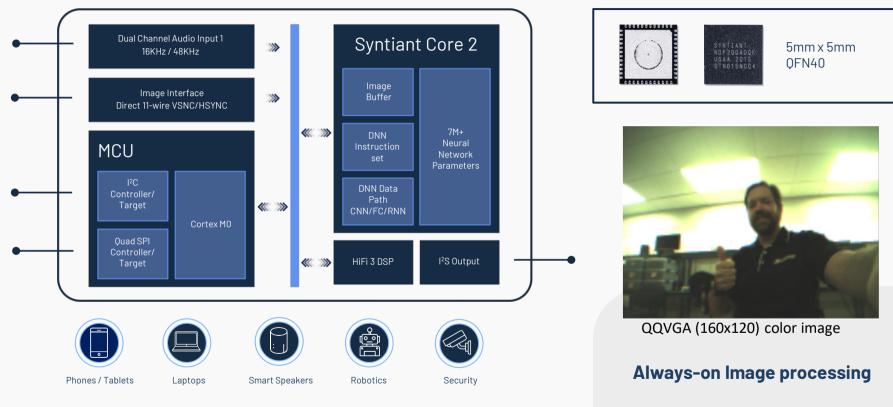
All the elements are required to go to market

Data

Syntiant Data Platform automates the ingestion, labelling, aligning, cleaning, and synthetic data generation to turn raw data into training data sets

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NDP200 Image, Voice & Sensor Neural Decision Processor



MobileNet + sensor in < 1 mW

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Syntiant Silicon: Neural Decision Processor (NDP)



At-Memory Compute

Tightly coupled memory and MAC functions to minimize data movement



Sustained High MAC Utilization

- Syntiant architecture executes NN every clock cycle achieving over 80% utilization for common NN
- Assures maximum usage of operations per • second



Native Neural Network Processing

- No need for intermediary compilers
- Networks trained in TensorFlow, etc are deployed directly to the NDPs.

Syntiant Core 2 Performance @ 32MHz Comparison = Arm A53 **Svntiant** Arm **Syntiant** @1GHz Core 2 A53 **Advantage** Inferences per 0.0471 ~30x 1.346 **Million Cycles** at **1%** µJ per 166 16131 ~100X Inference of the energy Identical MobileNet V10.25 int8 network on A53 & NDP200 (Syntiant Core 2)



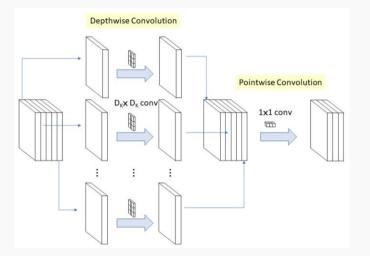
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tinyMLPerf-style test mechanics -- single, identical input vector

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If an ARM A53 powered device has a battery life of 3.5 days, the NDP200 powered device has a 1-year battery life running the same neural network.

MobileNetV1



- Scalable image network architecture for reducing model complexity
 - Depthwise convolutions break complexity
 - \circ Width Multiplier α , resolution Multiplier ρ

Width Multiplier	ImageNet	Million	Million
	Accuracy	Mult-Adds	Parameters
1.0 MobileNet-224	70.6%	569	4.2
0.75 MobileNet-224	68.4%	325	2.6
0.5 MobileNet-224	63.7%	149	1.3
0.25 MobileNet-224	50.6%	41	0.5

Different Values of Width Multiplier a

- 0.25 MobileNetV1 can easily fit in NDP200
- Under the 640 kB params support
 - Depthwise convolutions break complexity
 - Width Multiplier α , resolution Multiplier ρ

https://towardsdatascience.com/review-mobilenetv1-depthwise-separable-convolution-light-weight-model-a382df364b69 https://arxiv.org/pdf/1704.04861.pdf

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MobileNetV1Style Optimizations

Table 12. Face attribute classification using the MobileNet architecture. Each row corresponds to a different hyper-parameter setting (width multiplier α and image resolution).

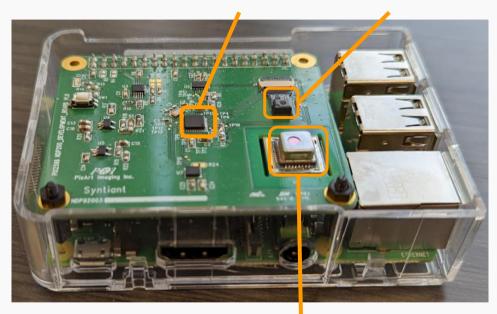
Width Multiplier /	Mean	Million	Million
Resolution	AP	Mult-Adds	Parameters
1.0 MobileNet-224	88.7%	568	3.2
0.5 MobileNet-224	88.1%	149	0.8
0.25 MobileNet-224	87.2%	45	0.2
1.0 MobileNet-128	88.1%	185	3.2
0.5 MobileNet-128	87 7%	48	0.8
0.25 MobileNet-128	86.4%	15	0.2
Baseline	86.9%	1600	7.5

Face Attribute Classification

- NDP200 is a flexible architecture that can support tuning the networks
- Example of aggressive model compression in Google's Mobilenet paper - Face attribute classification
 - Shrink to 200k params still with good performance
- Tons of work to tailor optimal solutions for the NDP200 platform
 - o MobileNet V2
 - Single-shot multibox detection (SSD)

NDP200 Development Platform

- Jointly developed image development platform with PixArt
 - Raspberry Pi form factor
 - NDP200 for ML solutions
 - 320x240 image sensor (PAG7920)
 - 8x8 thermal sensor (PAF9701)
 - 6-axis accel/gyro BMI sensor (BMI160)
- Enables both capture and model development for Computer Vision (CV) ML model solutions



NDP200

PAG7920

PAF9701

NDP200 Person Detection Demo

- Trained up a MobileNetV10.25 person detection
 - Input size 320x240 grayscale
 - First layer was additional decimation to 240x120 image
 - 59-layer Neural network on Core 2
- Used person/non-person images from MS COCO for demo network
 - OpenCV to decimate + grayscale images to match image sensor
- Seeing 90% accuracy on the demo model
 - 4-5 hours of training





NDP200 - More than just Object detection

- Can NDP200 do more than just object detection?
- Can it observe and act on the images presented to it?
- Goal: teach NDP200 to play D00M!



VIZDOOM allows developing AI bots that play **DOOM** using the visual information (the screen buffer). It is primarily intended for research in machine visual learning, and deep reinforcement learning, in particular.

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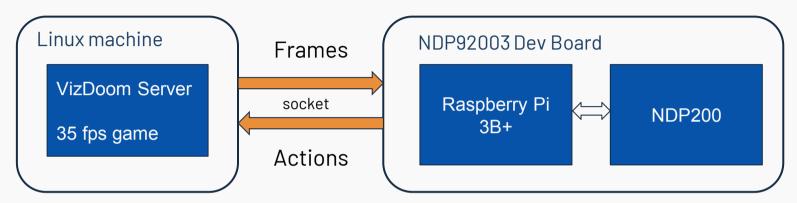
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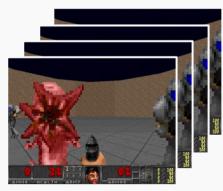


- Using the VizDoom platform, able to train an NDP200 using reinforcement learning to play "DOOM"
 - Reward is killing a monster, only limited time + ammo
 - Training on Syntiant's TDK environment (Linux), inference running on the NDP200
- Convert frames into 64x64 grayscale (4 frames in time series) to the network
 - o 3 Convolutional layers into 2 Dense 606k params

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Running VizDoom on the NDP200







5 layer Deep Neural Network 606k params 1 mW @ 35 fps

15,000x improvement over my 15W brain

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NDP200 Platform - Reinforcement Learning

- Start with no knowledge of the game mechanics
- Play thousands of games with rewards for killing monsters
- Learns to shoot and turn around 360 looking for approaching monsters



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NDP200 Platform - Reinforcement Learning

- Deadly corridor expands to 7 actions (motion)
- CANNOT get past without killing all the monsters
- Roughly 3-days of training, almost the same network
 - o 8 output classes



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Conclusion

- NDP200 is ready for Computer Vision (CV) applications at the edge in < 1 mW
 - Always-on domain
- NDP200 Development platform is ready to explore CV and sensor-fusion applications
 - Supports data collection (to help calibrate images), and real inference for development
- More than just simple object-detection, NDP200 can use ML to observe and control in the real world



Thank you



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