

NMDL lunch seminar

Feature Extracting Properties of sRBM

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2. HOG

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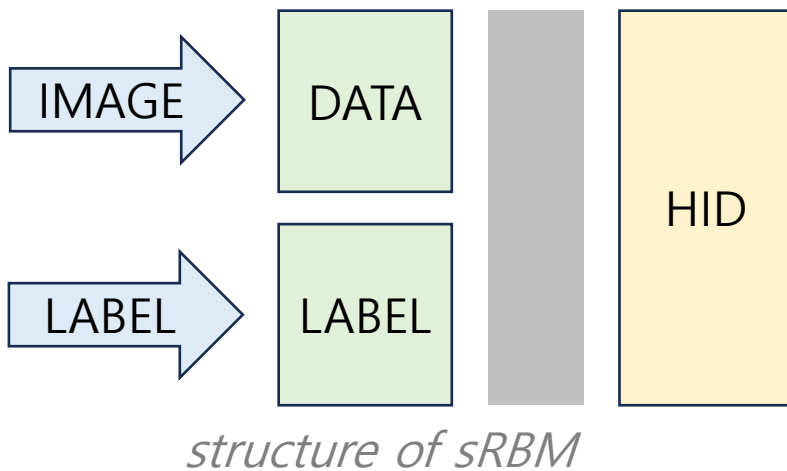
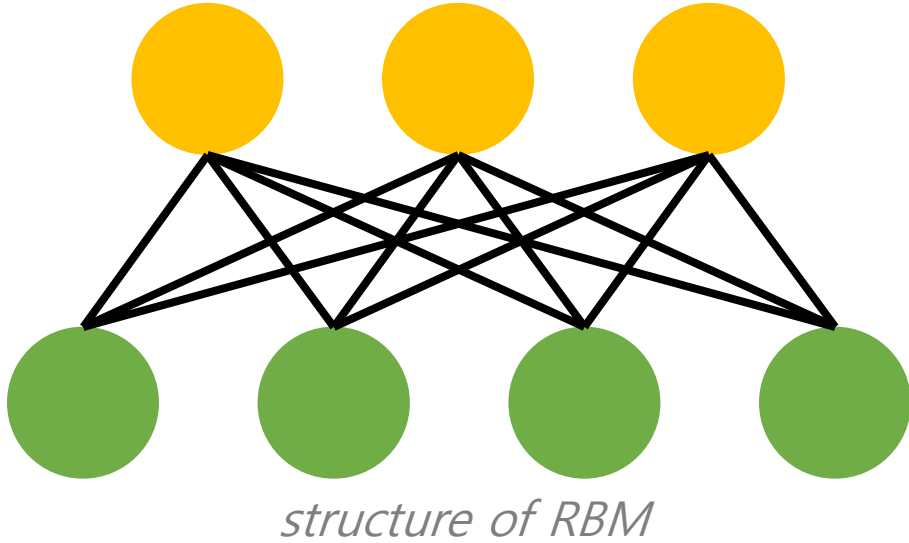
- Training from generated features

5. Future Work

- Connecting Neuromorphic computing and AI

1. Outline

Extracting features from sRBM



What is the goal of RBM?

- Input data $\approx$ Reconstruction
- **Extract Latent factor at hidden layer**

How is "RBM" employed in ANN?

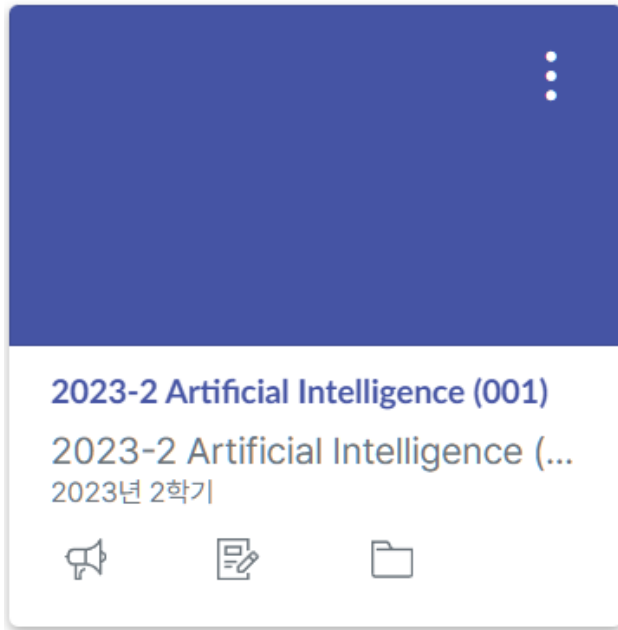
- pretraining DBN
- AutoEncoder

How to extract features from sRBM?

Count Spike!

1. Outline

Extracting features from sRBM



- Math for AI
- **Computer Vision**
- Machine Learning
- Deep Learning
- Natural Language Processing

2. HOG

Feature extractor for ANN

face recognition via HOG



Feature Matters!

2. HOG

Feature extractor for ANN

Recap: 2023-S intern

3. CNN Transfer-learning

Midterm goal: Achieve accuracy 90%

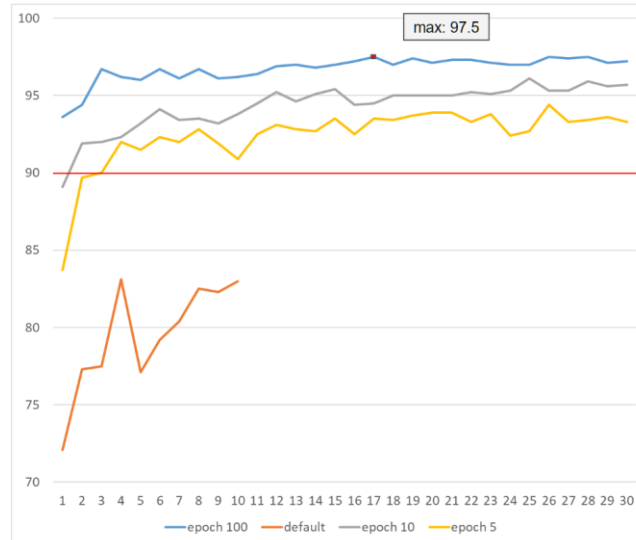
Achieved the goal!

max accuracy of 97.5 at epoch 17 with:

- $w_{ij_init_sigma} = 1.06 \rightarrow 0.3426$
- MNIST threshold = 0
- random walk step = 0.053
- label neuron number = 40 $\rightarrow$ 120
- hidden neuron number = 824 $\rightarrow$ 888
- CNN-sRBM (Mode 2)

Accuracy was already 93.6 at epoch 1
Accuracy also remains above 97 in later epochs

pre-trained CNN model of 5, 10, 100 epoch
(accuracy ~ 90%, 96%, >99%)

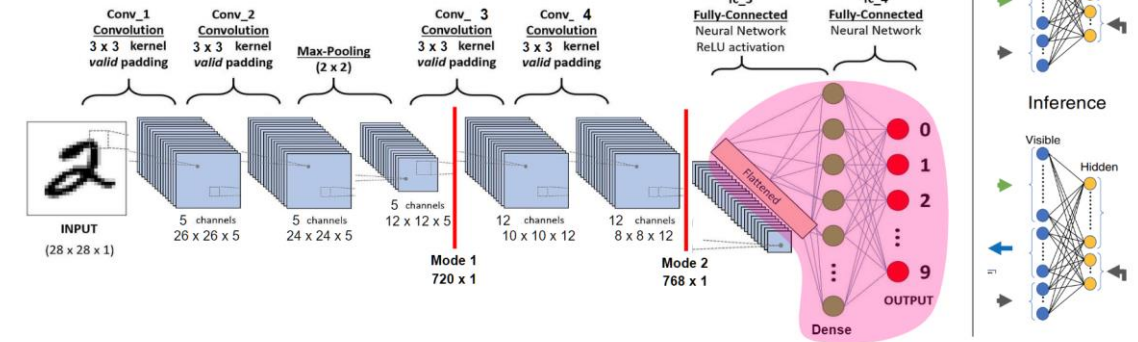


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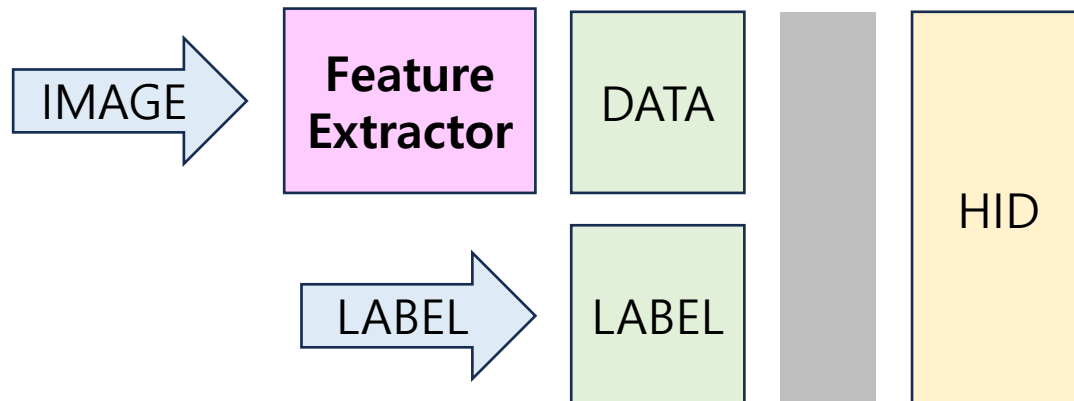
3. CNN Transfer-learning

hypothesis:

- pink part of CNN model(Flattened - Dense - Output) is the equivalent of sRBM.



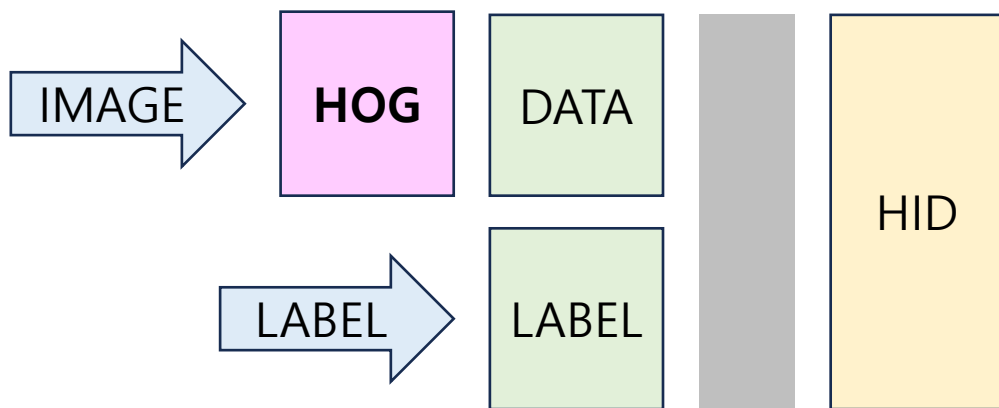
14



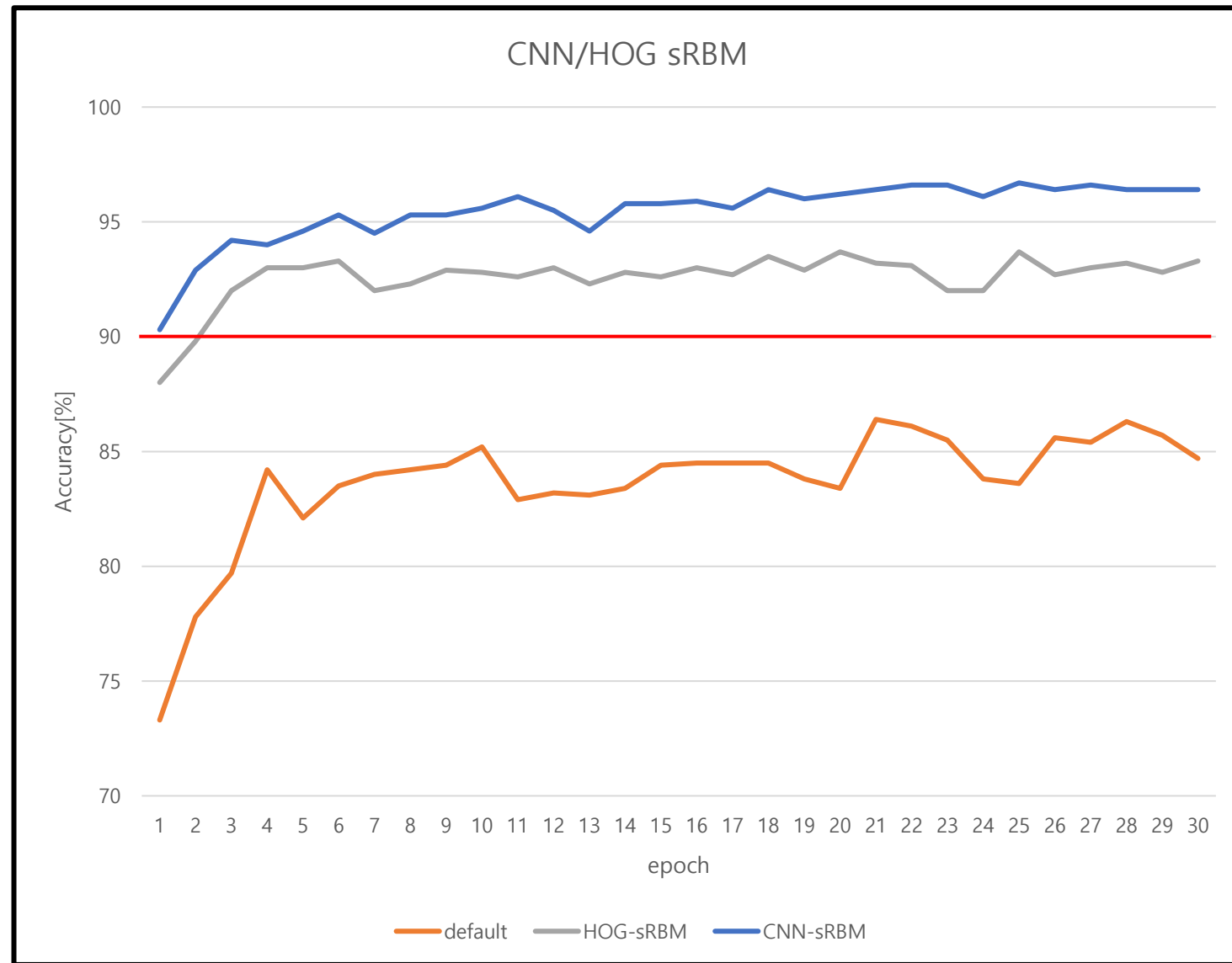
Trained sRBM with better feature
Feature Extractor : CNN

2. HOG

Feature extractor for ANN



Again, Feature Matters!

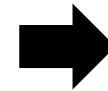


3. Spike Counting

Reconstruction vs Feature Extraction

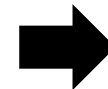
What to count?

Hidden

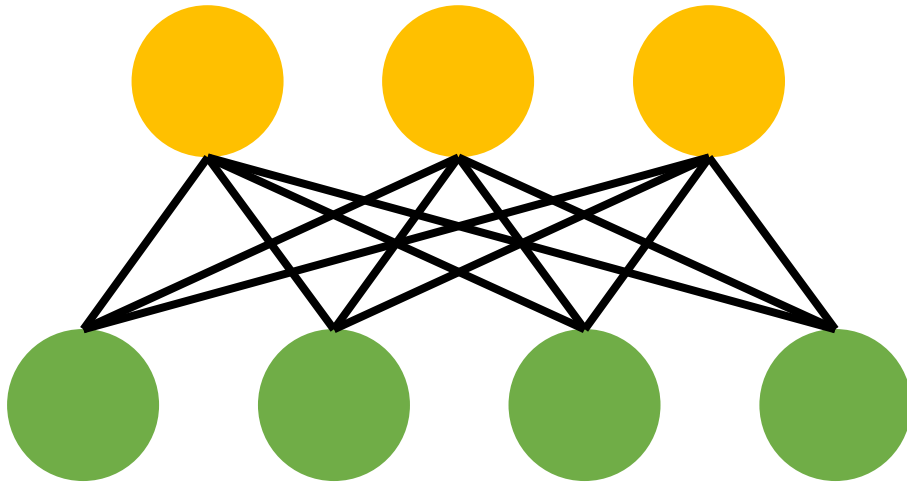


Feature extraction

Visible



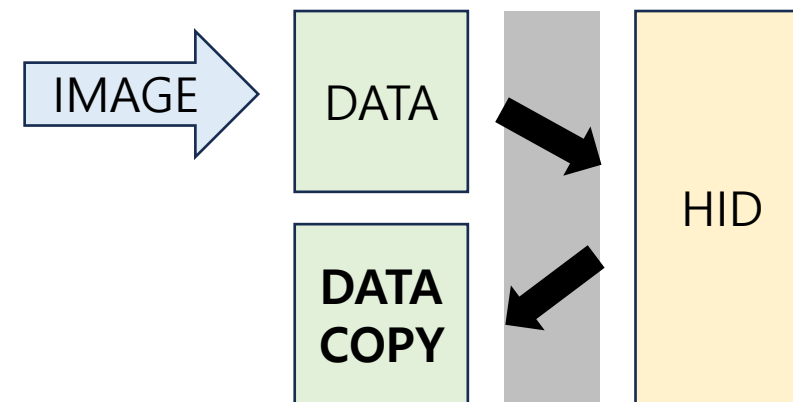
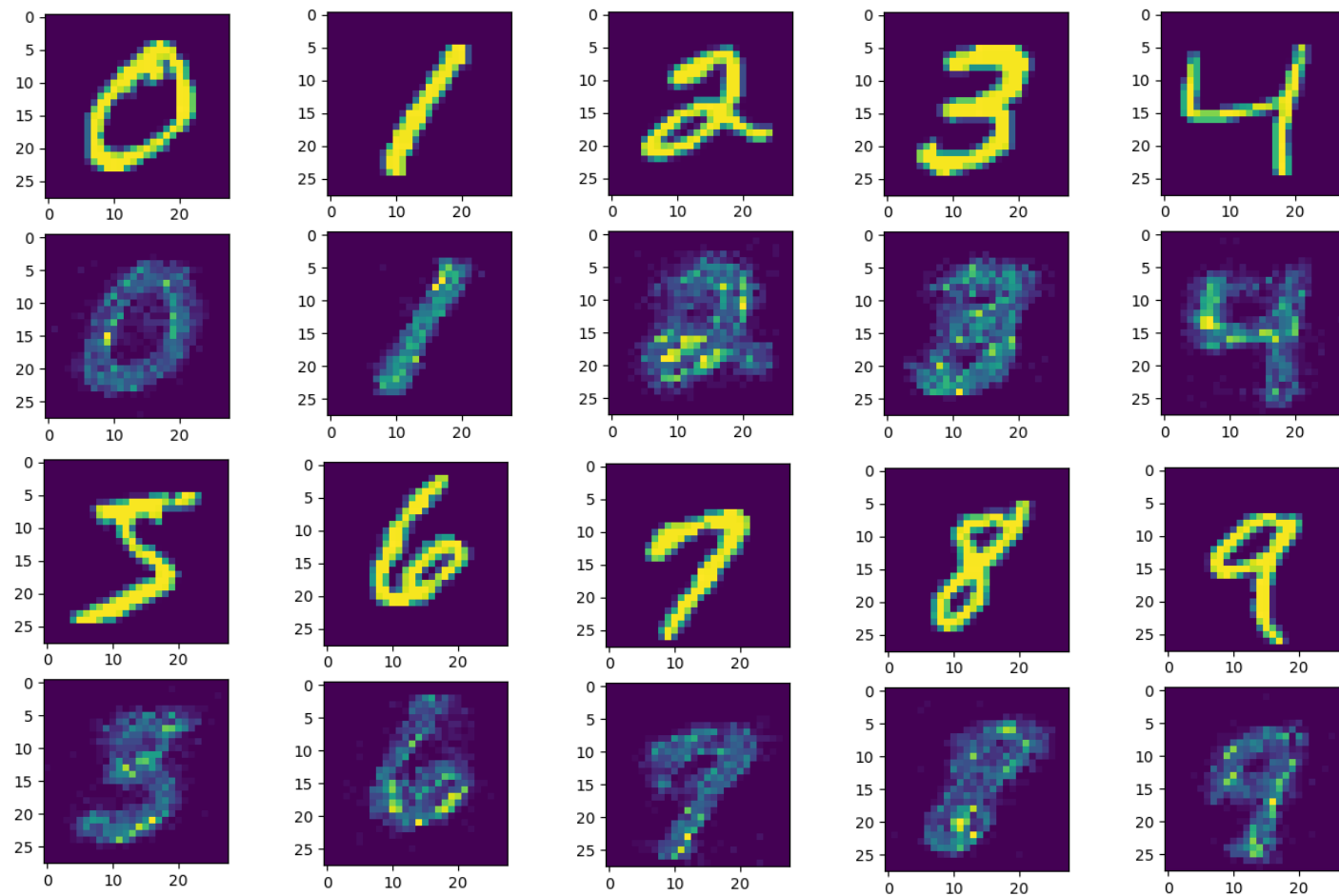
Reconstruction



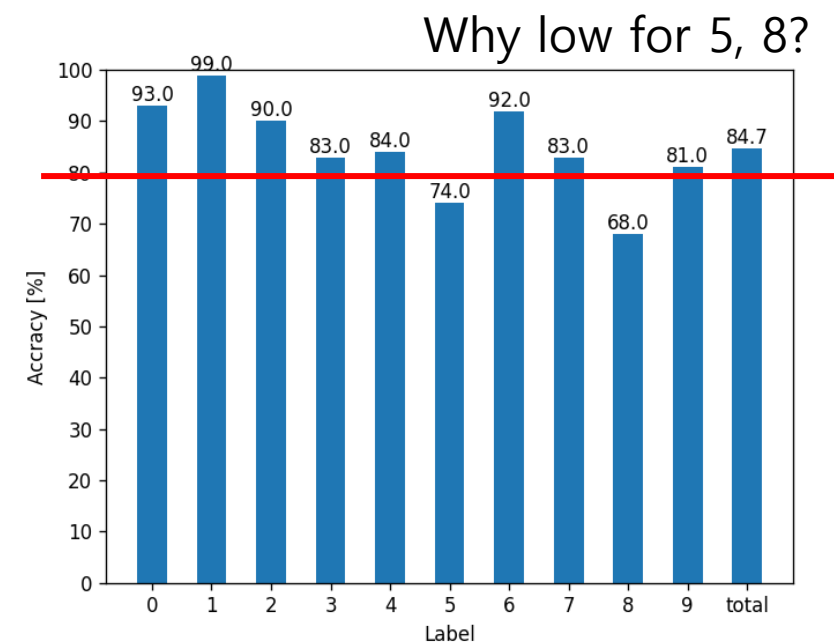
structure of RBM

3. Spike Counting

Reconstruction vs Feature Extraction



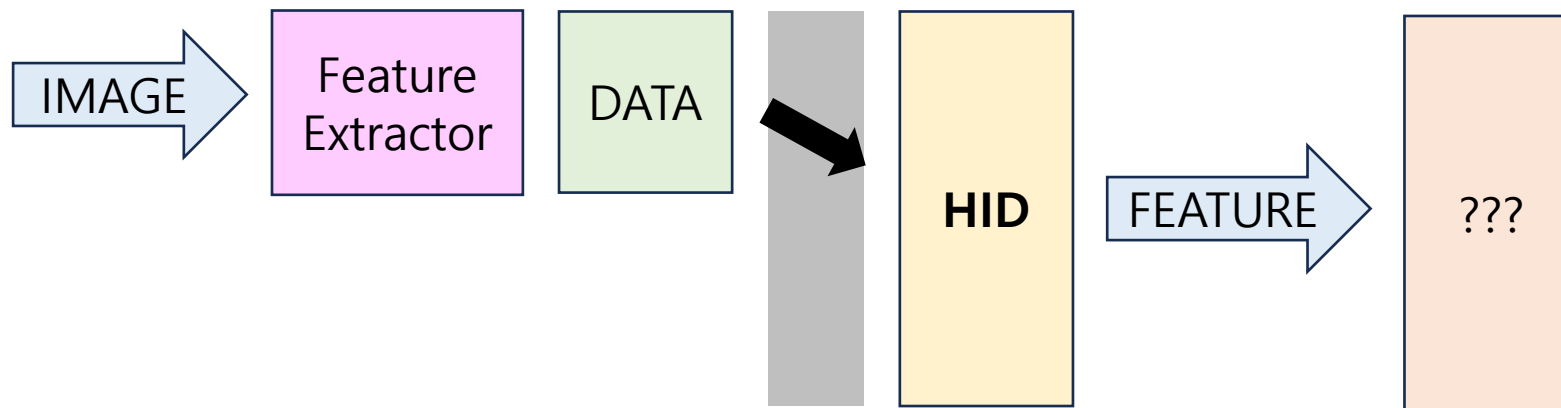
sRBM actually “sees” image like this!



Reconstruction vs Feature Extraction

Where, How to use the feature?

- SVM (Support Vector Machine)
- MLP (Multi-Layer Perceptron)
- new sRBM (Deep-sRBM)



3. Spike Counting

Feature Extraction

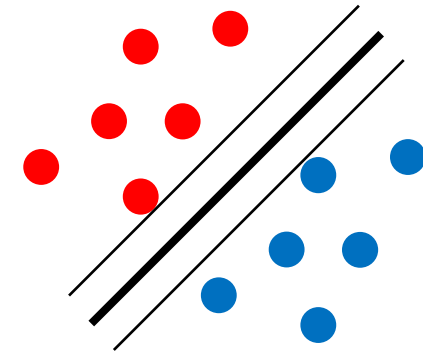
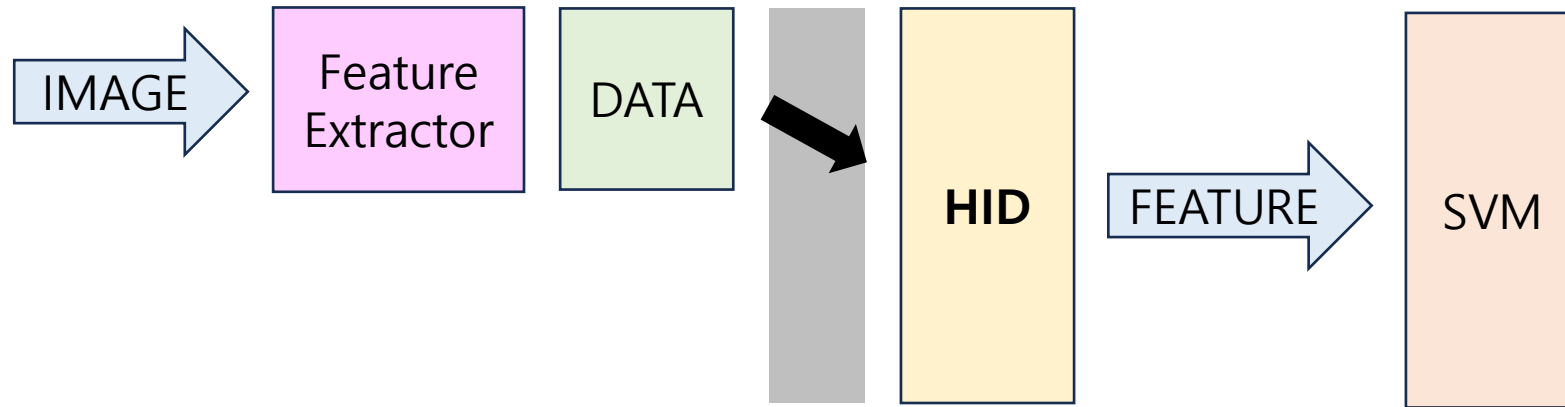
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[illegible]

Some of them never fires...

3. Spike Counting

Reconstruction vs Feature Extraction

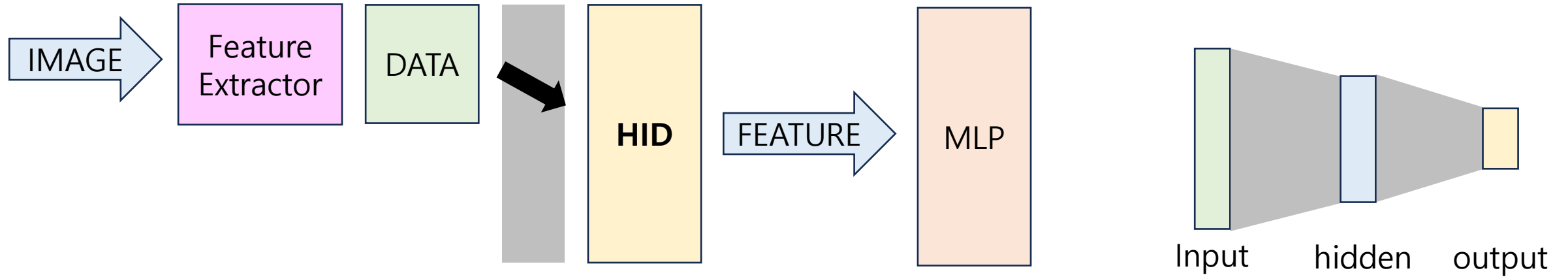


SVM (Train 5000 / Test 1000) [%]

feature extractor \ data	Default	HOG	CNN (10epoch, 0.9817)
MNIST	88.7	96.2	97.8
sRBM Feature	91.3	95.8	

3. Spike Counting

Reconstruction vs Feature Extraction

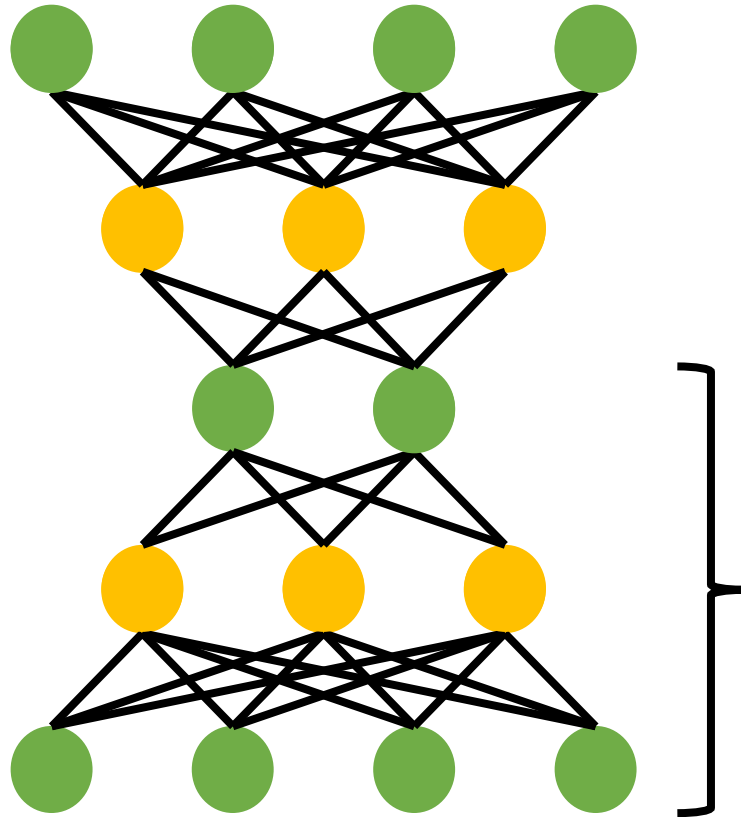


MLP (Train 5000 / Test 1000) [%]

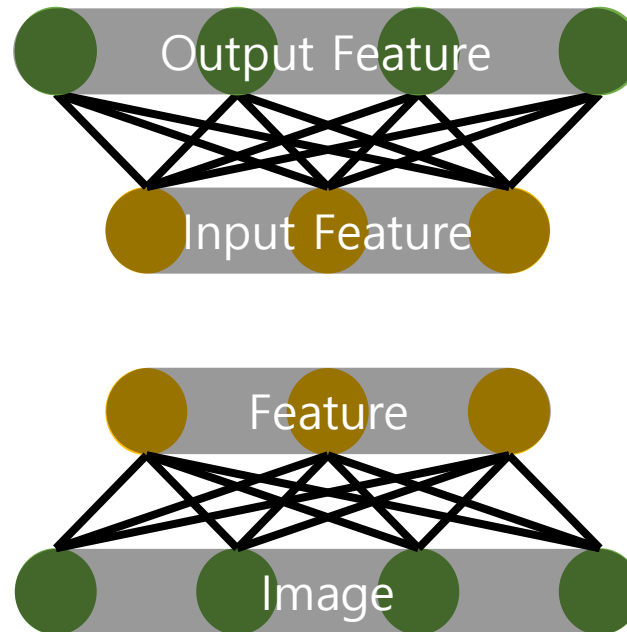
feature extractor \ data	Default	HOG	CNN (10epoch, 0.9817)
MNIST	93.2	97	98.17
sRBM Feature	90.8	96.4	97.3

4. Deep-sRBM

Learning from generated features



structure of DBN/Autoencoder



Recap

What is the goal of RBM?

- Input data $\approx$ Reconstruction
- **Extract Latent factor at hidden layer**

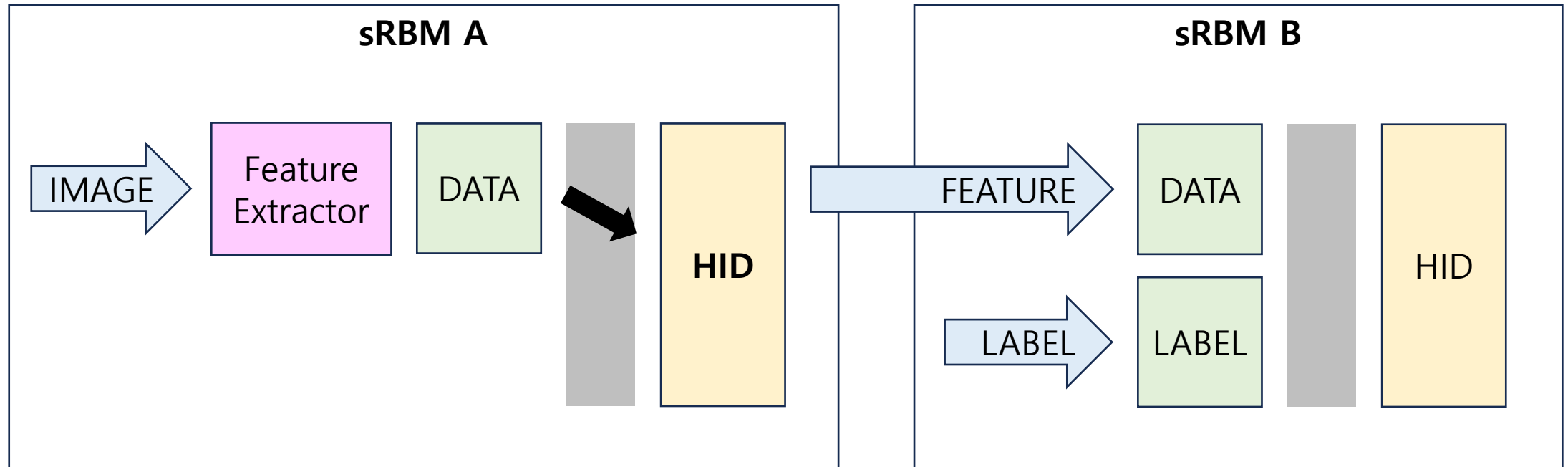
How is "RBM" employed in ANN?

- pretraining DBN
- AutoEncoder

How to extract features from sRBM? Count Spike!

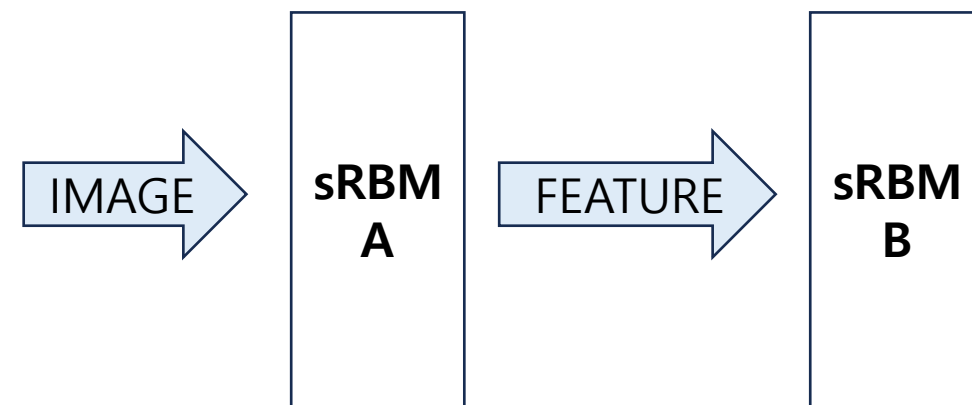
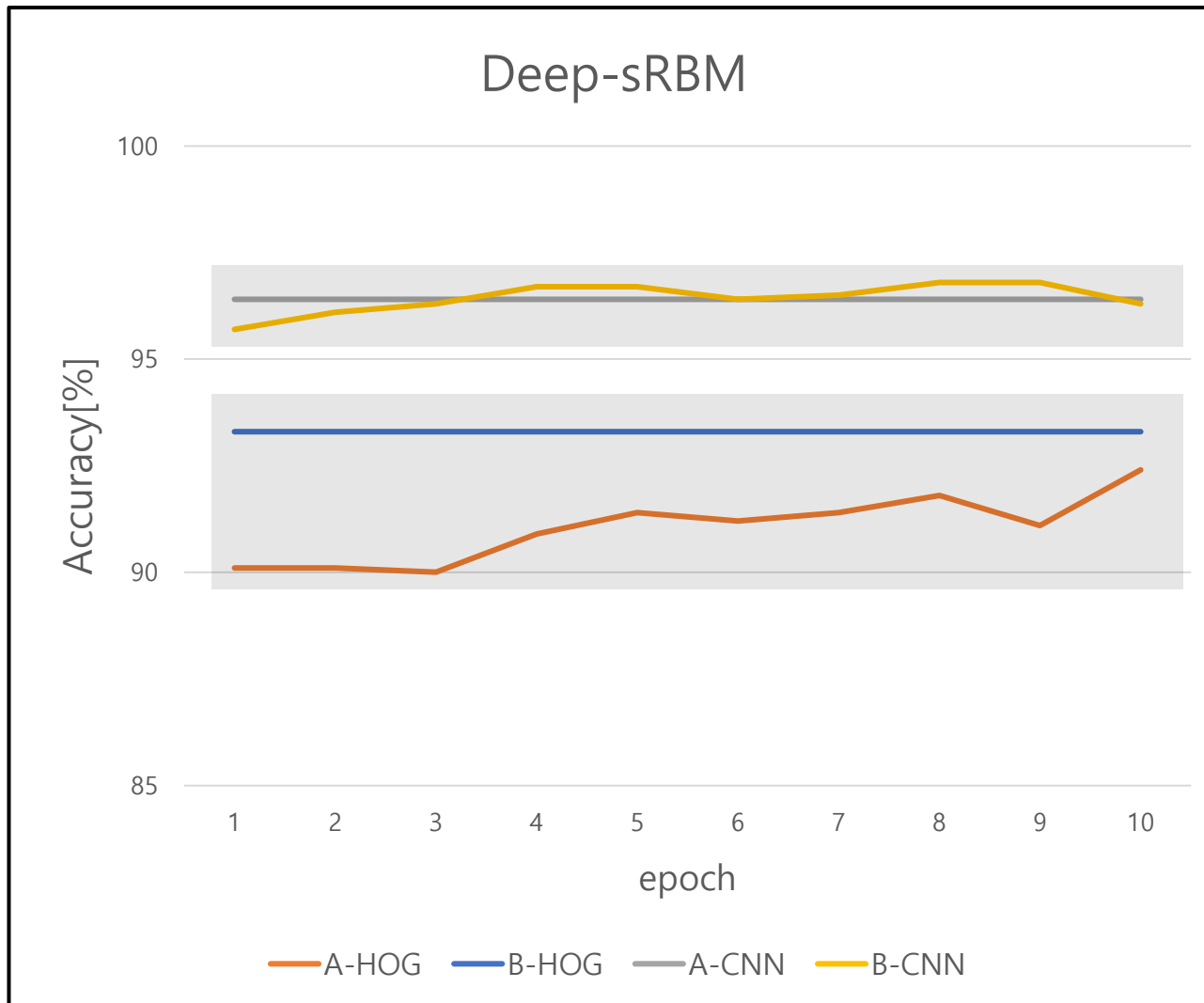
4. Deep-sRBM

Learning from generated features



4. Deep-sRBM

Learning from generated features



feature extractor \ sRBM		
	A (freeze)	B
HOG	93.3	92.4
CNN	96.4	96.8

[%]

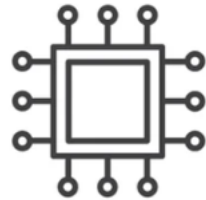
5. Future Work

Connecting Neuromorphic computing and AI

Neuromorphic Simulation



Neuromorphic?



Simulation?



5. Future Work

Connecting Neuromorphic computing and AI



Computing Platform

A new computing processor that mimics the human brain

Neuromorphic Processor

With AI, machine learning, and blockchain demanding greater computation, we are engaged in fundamental breakthroughs. One such advance is through the development of neuromorphic processors inspired by the human brain. The goal of this approach is to emulate the brain's dynamic learning capability and power efficiency.

At SAIT, we are actively researching near/in memory computing, asynchronous spiking neural networks, and other concepts to create a brain-like processor. Our interests include brain-inspired learning and inference algorithms, low-power mixed signal computing architectures, and new synaptic memories.

SAIT collaborates with the Business Units to develop and verify technologies using the latest logic and memory fabrication.

Thank you