CENTER FOR BRAIN-INSPIRED COMPUTING ENABLING AUTONOMOUS INTELLIGENCE

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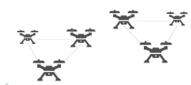




C-BRIC Vision

Enable next generation of intelligent autonomous systems

- Narrow the orders-of-magnitude computing efficiency gap between current computing systems and the brain
- Drive improvements in the robustness of cognitive computing systems
- Explore distributed intelligence across edge/hub/cloud and peer-to-peer networks
- Demonstrate the impact of these advances in end-to-end systems such as autonomous drones and personal robotics

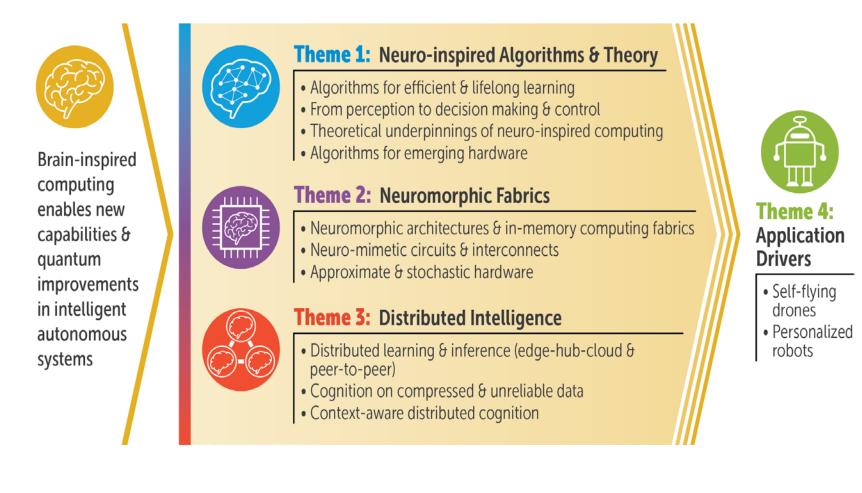








C-BRIC Organization





Theme 1: Neuro-inspired Algorithms and Theory

State-of-the-Art: Deep Neural Nets

- Largely supervised learning
- Static (one-time) learning
- Training requires global updates (Backpropagation / SGD)
- Perception (speech, images, text)
- Unknown generalization behavior
- Manually designed network topologies

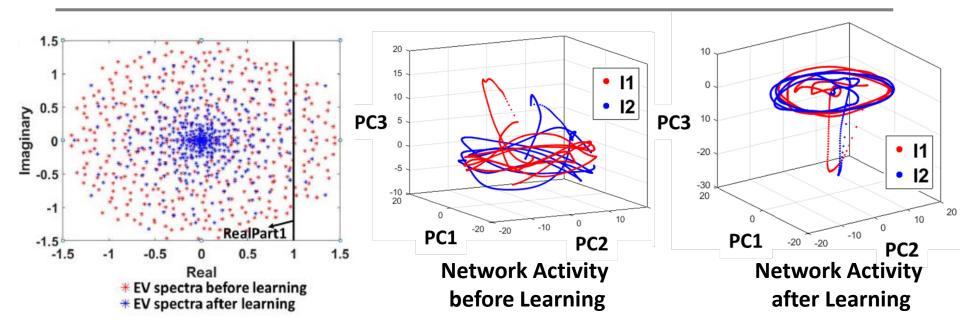
Future Neuro-Inspired Algorithms

- Computationally efficient algorithms
- Theory of neural computing from DNN to emerging models
- Learning with less data
- Incremental and lifelong learning
- Algorithms that leverage stochastic and approximate computation
- Learning and inference on emerging computing fabrics



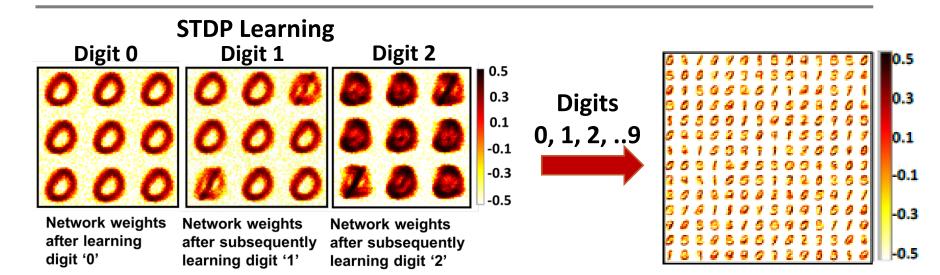
C-BRIC Theme 1

Theoretical Understanding of Learning

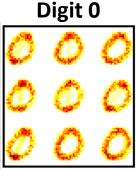


- Shrinking of the EigenValue spectral circle represents the stabilizing effect of the learning mechanism
- Understanding network behavior from Random Matrix theory and Principal Component Analysis
- Quantification of stabilizing hyperparameters from network activity

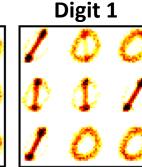
Learning to Forget with Adaptive Decay



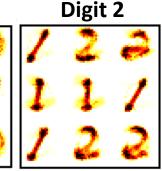
Adaptive Plasticity Learning



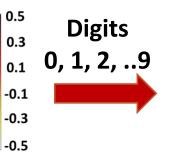
Network weights after learning digit '0'

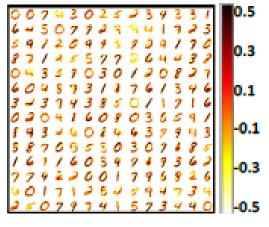


Network weights after subsequently learning digit '1'



Network weights after subsequently learning digit '2'



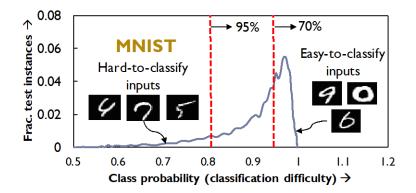




Dynamic, Variable-Effort Deep Networks

- Deep nets are fixed-effort and static
- Inputs differ greatly in their difficulty
- Mechanisms to dynamically modulate computational effort of neural nets

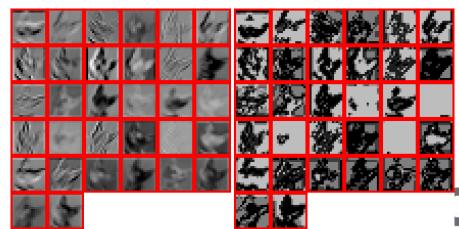
Significance-driven Feature Evaluation



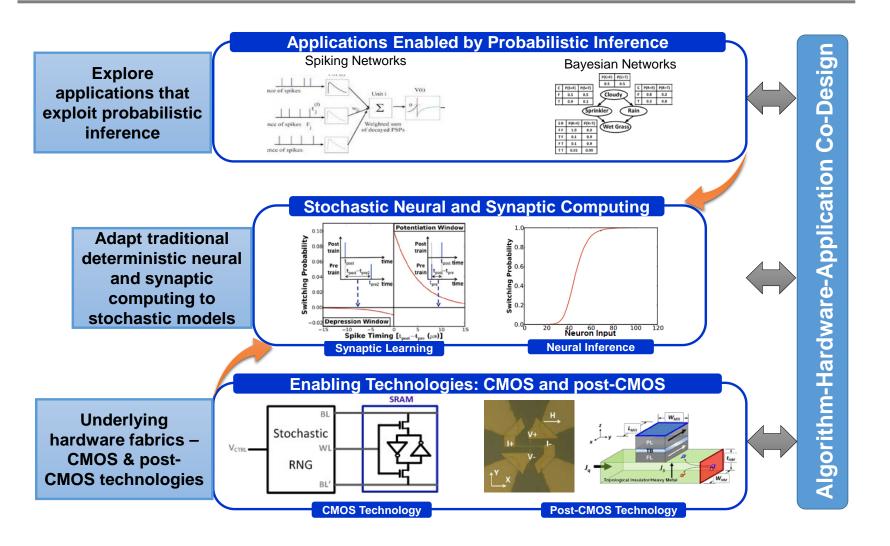
Input (CIFAR-10)



Feature maps (C1 layer) Computational effort



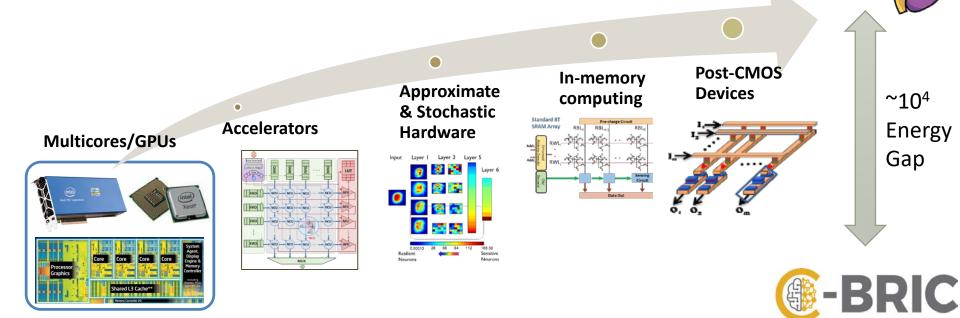
Stochastic Neurons and Synapses



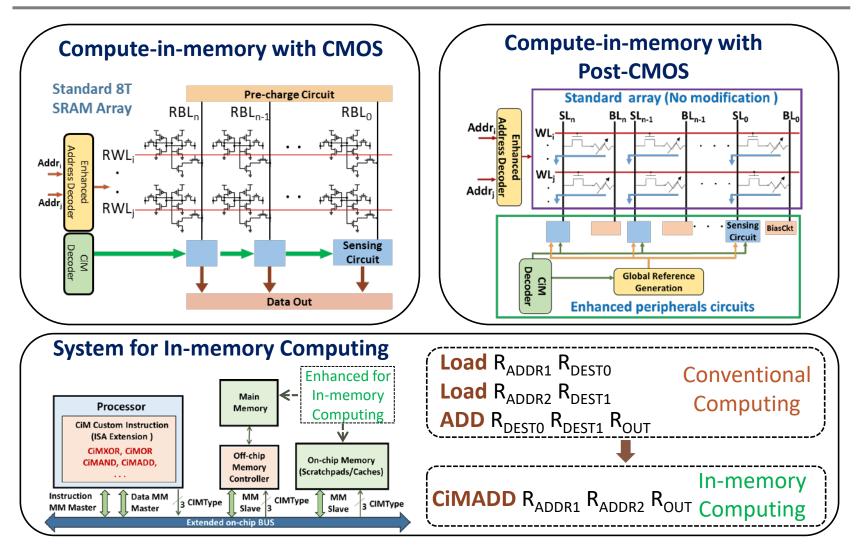


Theme 2: Neuromorphic Fabrics

- CMOS and Post-CMOS neuro-mimetic devices and interconnects
- Compute-near-memory / Compute-in-memory
- Approximate and stochastic neuronal and synaptic hardware
- Architectures that embody computing principles from the brain (sparse, irregular, event-driven, massively parallel)
- Programming and evaluation frameworks

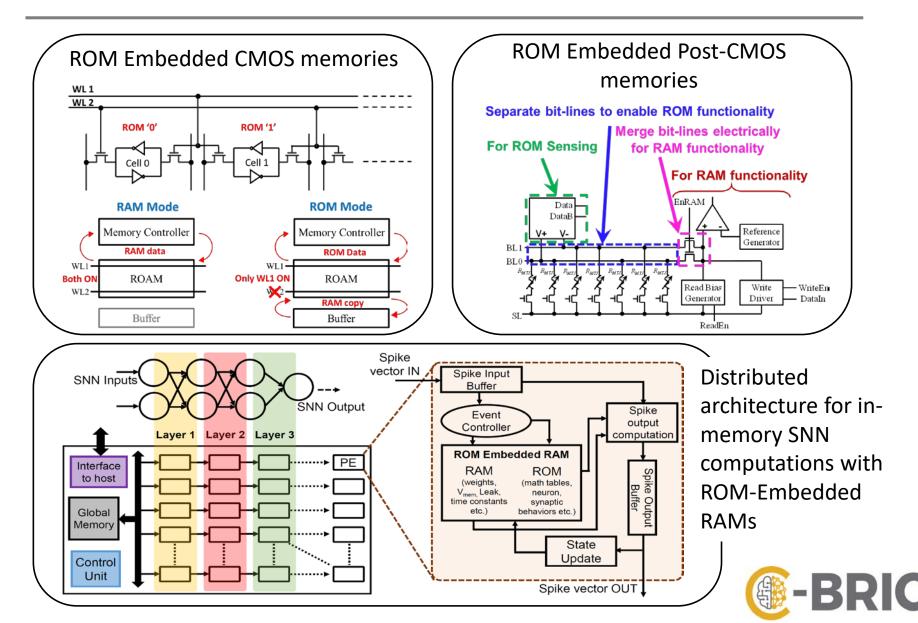


Compute-in-Memory

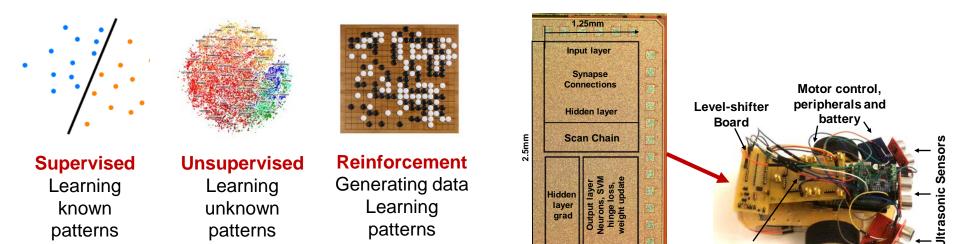




Compute-in-Memory: Embedding ROM In RAM



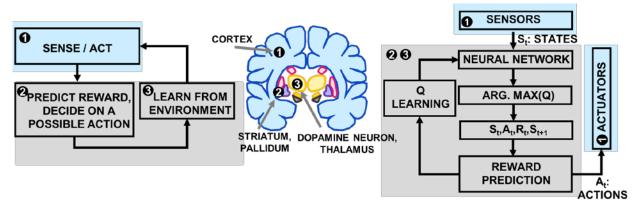
Hardware Demonstration of Autonomous Decision Making via Reinforcement Learning



patterns

patterns

patterns





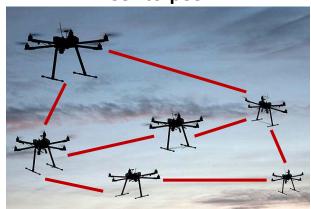
Testchip

Theme 3: Distributed Intelligence

State-of-the-Art:

Cloud-enabled Intelligence

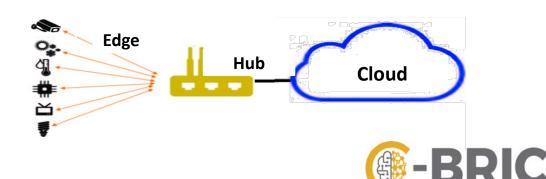
- Centralized training in cloud
- Inference entirely in cloud or entirely on edge device
- Algorithms agnostic to distributed context require high communication



C-BRIC Theme 3

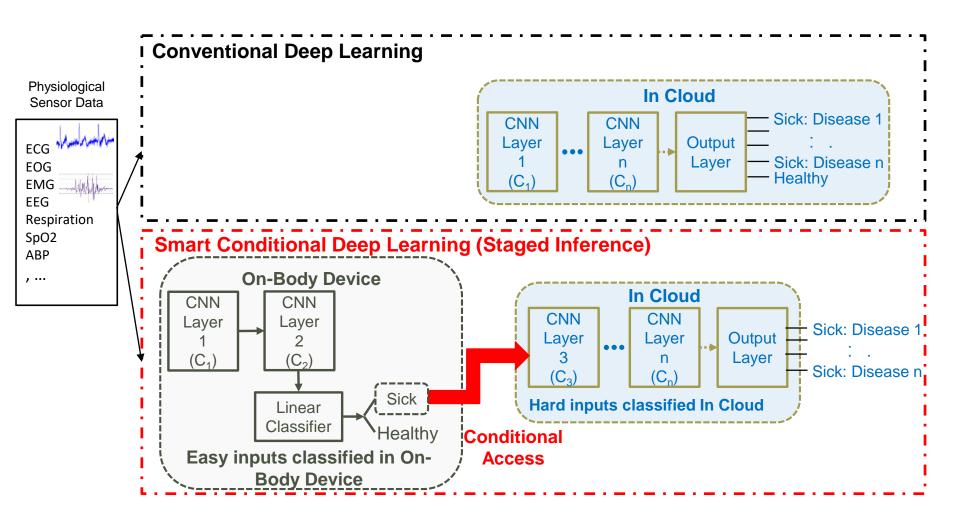
Future Distributed Intelligence

- Partitioned learning and inference
 - Algorithms for hierarchical (edge/hub/cloud) and peer-to-peer networks
- Cognition on compressed and unreliable data
 - Event-driven sensors, data fusion, learning from incomplete/ unsynchronized/noisy data
- In-sensor analytics
 - Low-complexity algorithms and hardware to enable in-sensor computing



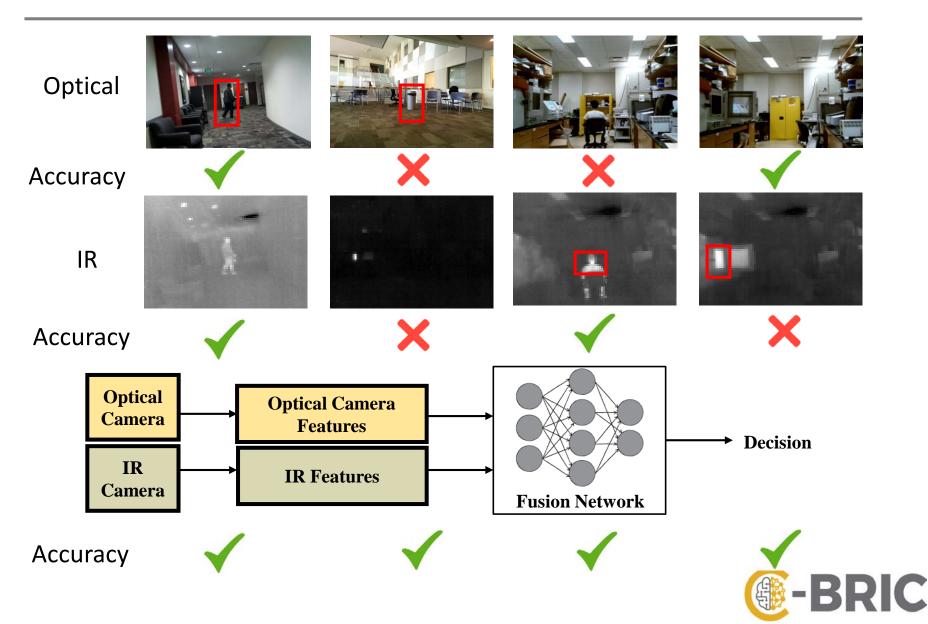
Peer-to-peer

Staged Conditional Learning/Inference





Multi-sensor Cognition in Smart Buildings



Theme 4: Application Drivers

- Autonomous drones and drone swarms
- Personal robotic assistants
- Technologies from Themes 1-3 enable new capabilities with real-time, autonomous operation





C-BRIC Universities

