Towards Cognitive AI Systems: A Survey and Prospective on Neuro-Symbolic AI



Neuro-Symbolic Al Algorithms

23

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CURRENT AI SYSTEMS

Current Al Systems

- ✓ Features:
 - Deep neural network
 Extensive dataset
 Large computing power
- Challenges:

This work:

Unsustainability
Lack robustness/explainability
Limited human-Al collaboration

NEURO-SYMBOLIC AI SYSTEMS

- **An Emerging Al Paradigm: Neuro-Symbolic Al**
- ✓ <u>Features</u>: neuro + symbolic + probabilistic Neuro: scalable, flexible, handle inconsistency Symbolic: interpretable, explainable Probabilistic: robust to uncertainty
- ✓ Advantages:

Hardware: Characterize performance and computational operators of neuro-symbolic Al

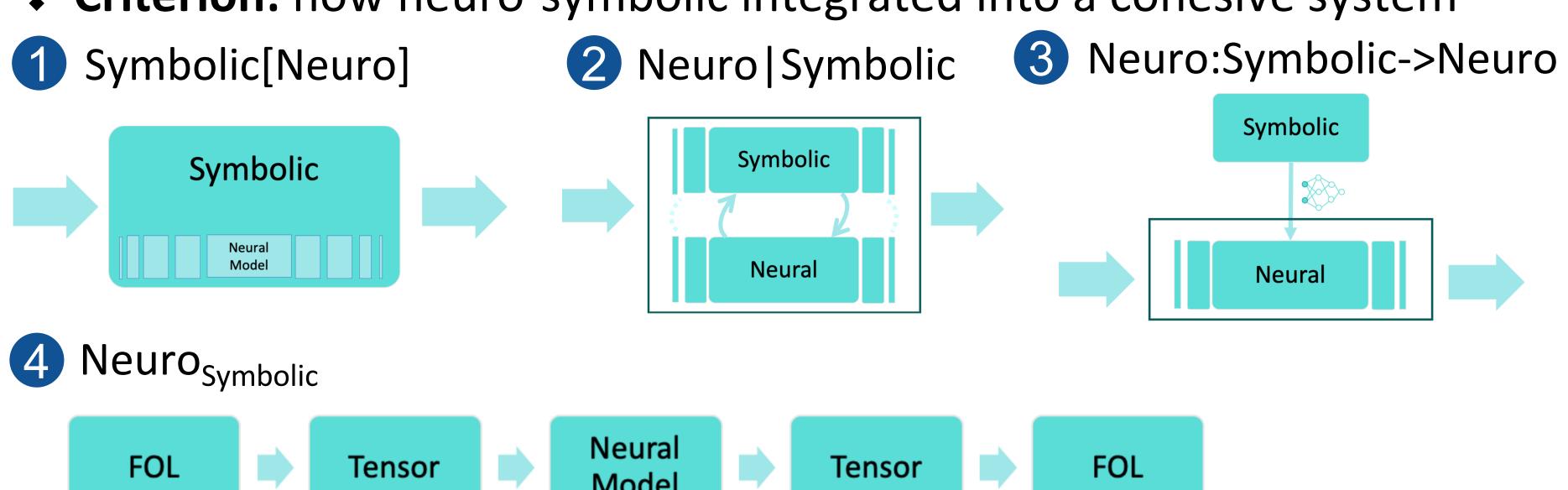
Improve efficiency, robustness, explainability
Human-like reasoning capabilities
Collaborative human-Al applications

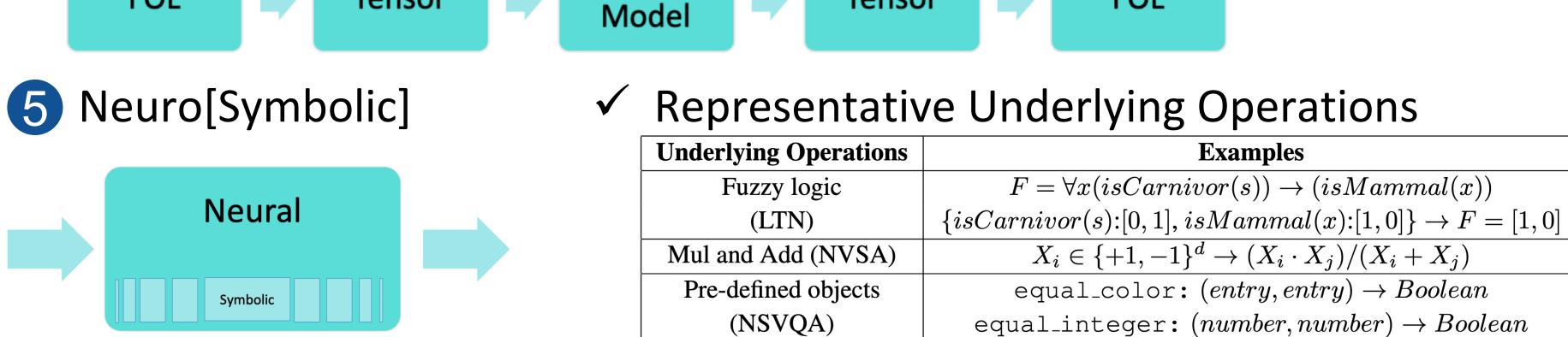
Neural Network Symbolic Probabilistic Scalable Interpretable Robust to Flexible Explainable uncertainty Handle inconsistency Data-efficient **Hardware Compute Platforms** Domain-Specific **Research Opportunities** Challenges Building more cognitive datasets/testbeds Limited evaluation tasks Ad-hoc model design (2) Unifying neuro-symbolic-prob models 3 Developing efficient software frameworks Limited extensibility 4 Benchmarking diverse NSAI workloads Large #algos and #HW Heterog. compute kernels 5 Designing cognitive architectures

NEURO-SYMBOLIC AI CATEGORIES

Criterion: how neuro-symbolic integrated into a cohesive system

Algorithm: Systematic review of recent neuro-symbolic Al algorithms





NEURO-SYMBOLIC AI ALGORITHMS

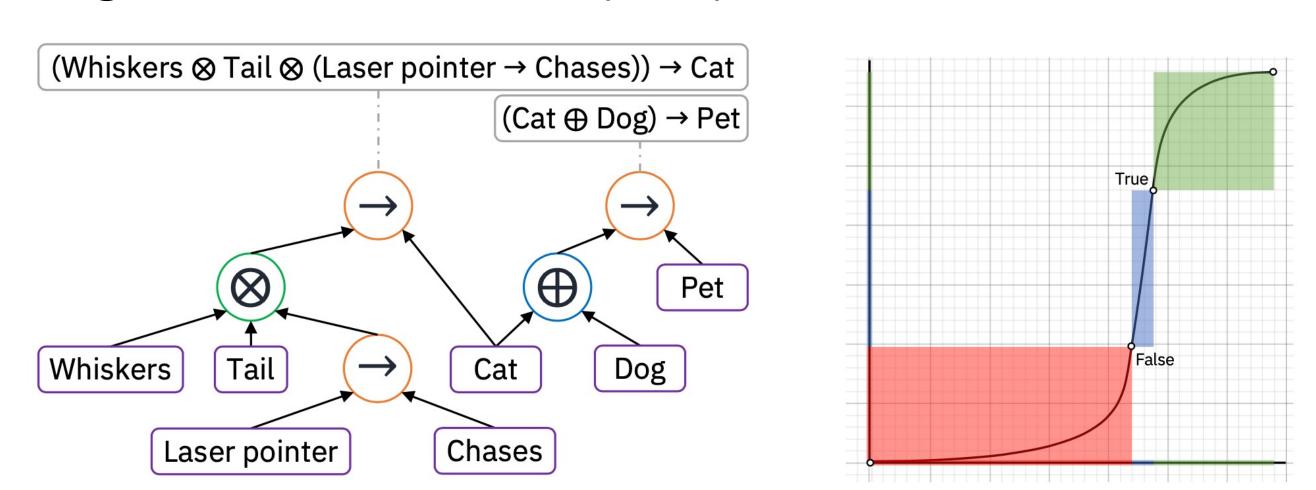
(1)

Category	NSAI Algorithm	Underlying Operation	If Vector
Symbolic[Neuro]	AlphaGo (Silver et al., 2017)	NN, MCTS	Vector
Neuro Symbolic	NVSA (Hersche et al., 2023)	NN, mul, add	Vector
	NeuPSL (Pryor et al., 2022)	NN, fuzzy logic	Vector
	NSCL (Mao et al., 2019)	NN, add, mul, div, log	Vector
	NeurASP (Yang et al., 2020)	NN, logic rules	Non-Vector
	ABL (Dai et al., 2019)	NN, logic rules	Non-Vector
	NSVQA (Yi et al., 2018)	NN, pre-defined objects	Non-Vector
Neuro:Symbolic →Neuro	LNN (Riegel et al., 2020)	NN, fuzzy logic	Vector
	Symbolic Mathematics	NN	Vector
	(Lample & Charton, 2019)		
	Differentiable ILP	NN, fuzzy logic	Vector
	(Evans & Grefenstette, 2018)		
NeuroSymbolic	LTN (Badreddine et al., 2022)	NN, fuzzy logic	Vector
	Deep ontology networks	NN	Vector
	(Hohenecker & Lukas, 2020)		
Neuro[Symbolic]	GNN+attention	NN, SpMM, SDDMM	Vector
	(Lamb et al., 2020)		
	NLM (Dong et al., 2019)	NN, permutation	Vector

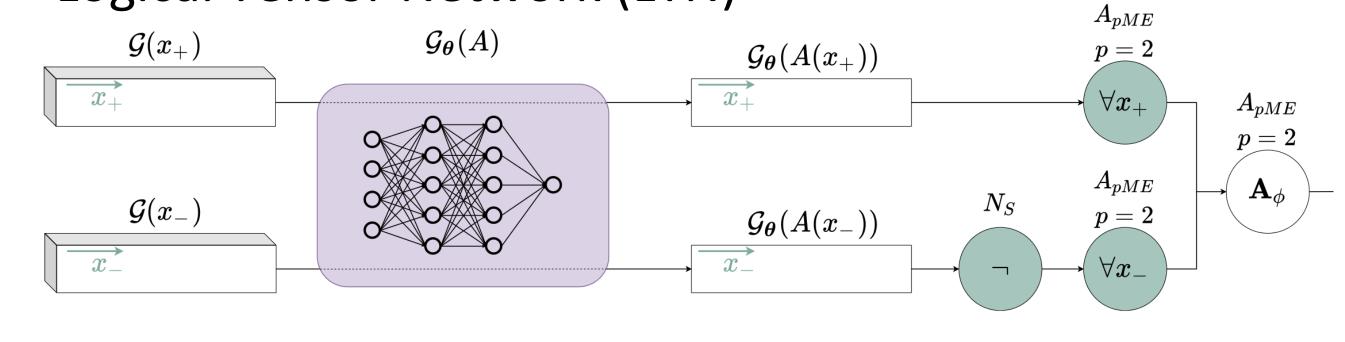
NEURO-SYMBOLIC AI SYSTEM PROFILING

Selected Neuro-Symbolic Al Models

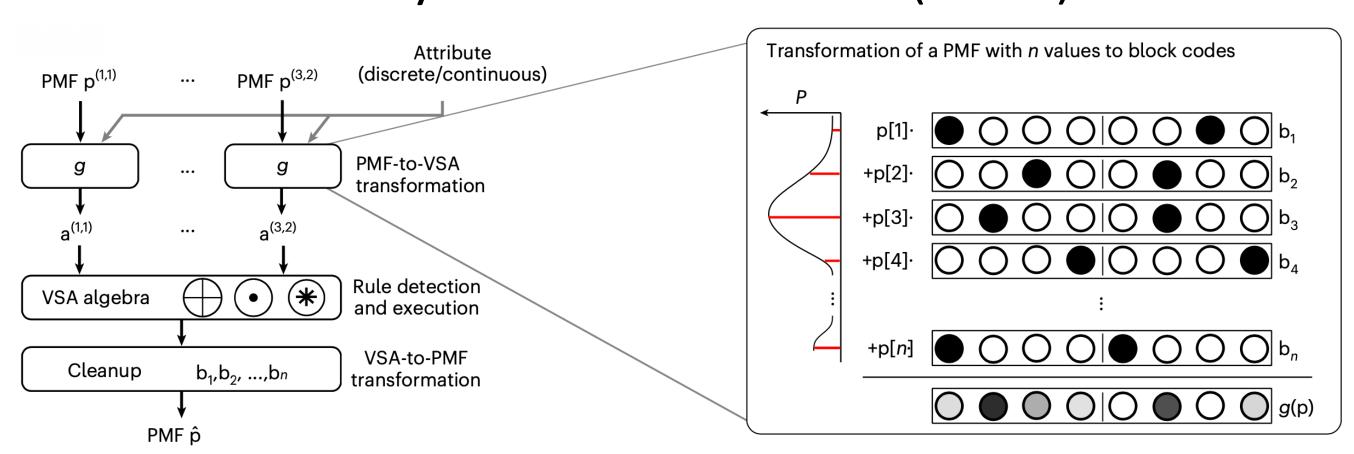
✓ Logical Neural Network (LNN)



✓ Logical Tensor Network (LTN)



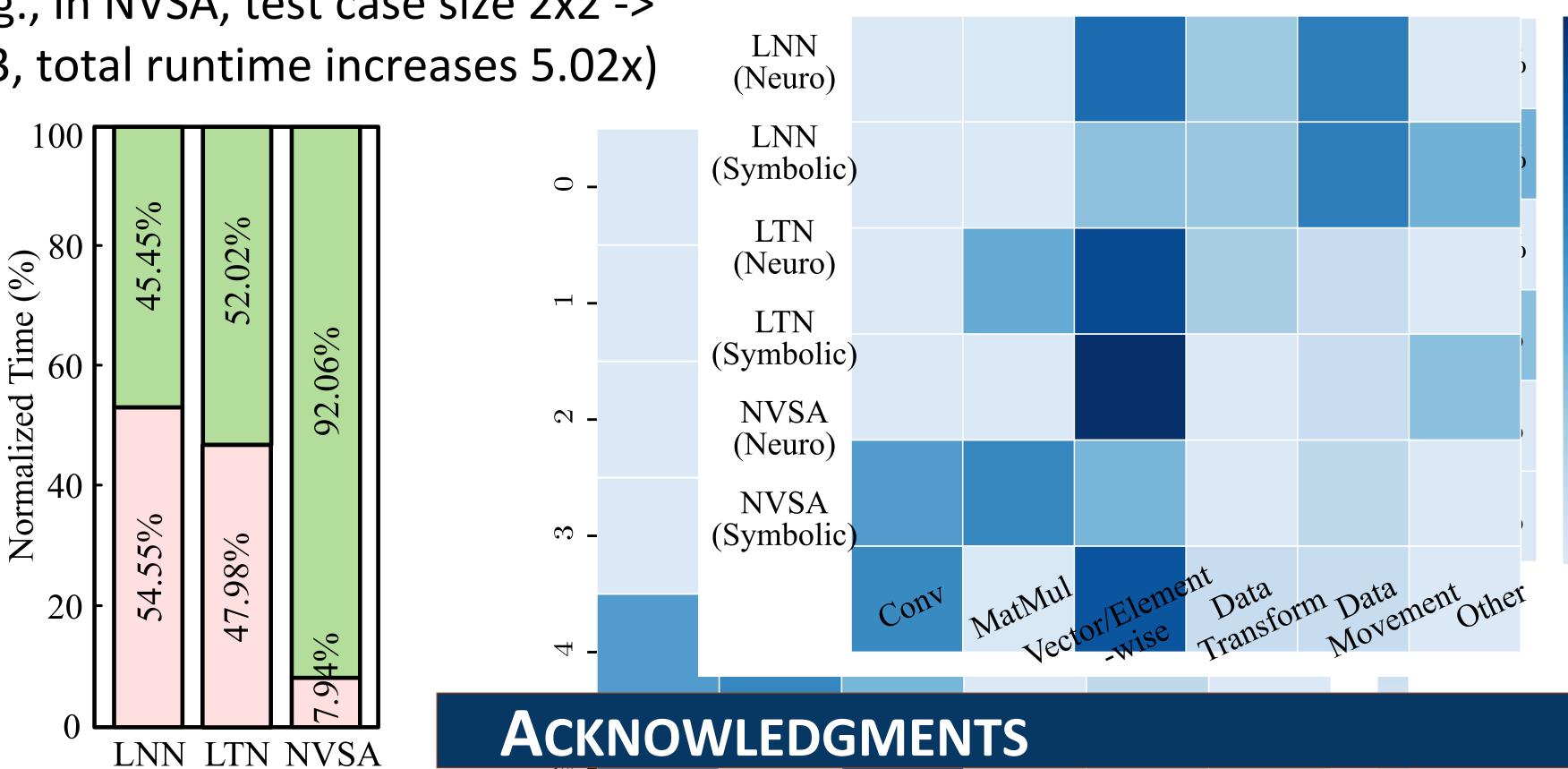
✓ Neuro-Vector-symbolic Architecture (NVSA)



- **Runtime Analysis**
- ✓ Methodology: PyTorch Profiler
- ✓ <u>Hardware</u>: Intel Xeon 4114 CPU, Nvidia RTX 2080 Ti GPU
- **Observations:**
- ✓ Symbolic workloads are not negligible and may become system bottleneck.
- ✓ Symbolic workloads in critical path.
 - Potential scalability issue (e.g., in NVSA, test case size 2x2 -> 3x3, total runtime increases 5.02x)

Neuro Symb.

- Workload Analysis
- ✓ Methodology: six operator categories
- **Observations:**
- ✓ Neuro workload is dominated by MatMul and vector/element-wise operations.
- ✓ Symbolic workload is dominated by vector/element-wise, logic operations.
- ✓ Data movement, complex control flow, low operational intensity.



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