MMAC Day 2024

# Infinite-valued logics and applications

Master's research topics in Logic and Computation

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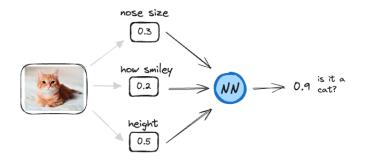
## Logic and applications...

#### goals:

- automated symbolic reasoning
- find trade-offs between expressivity and computational feasibility

## going beyond classical logic

- 2-valued : classical logic, dealing with sharp concepts
- infinite-valued : fuzzy logics, dealing with vague concepts
- motivation/applications : neuro-symbolic approaches to AI



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# Łukasiewicz logic $\in$ Fuzzy logics $\subseteq$ Infinite-valued logics

Fuzzy logics: connectives are interpreted as functions  $f : [0, 1]^n \rightarrow [0, 1]$ 

Semantics for Łukasiewicz logic				
	falsum	implication	negation	
	$\perp = 0$	$x \rightarrow y = \min\{0, 1 - (x - y)\}$	$\neg x = 1 - x$	

Deductive system for Łukasiewicz logic (Hilbert style axiomatization)

#### FACTS:

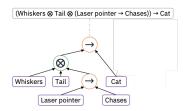
- Same computational complexity as propositional classical logic
- Piecewise linear NNs can be represented by formulas of Łukasiewicz logic
- Many other fuzzy logics have been considered, as Gödel or Product logics

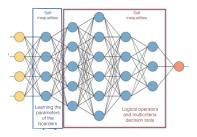
Gödel implicationProduct implication $x \to y = \begin{cases} 1 & \text{if } x \le y \\ y & \text{if } x > y \end{cases}$  $x \to y = \begin{cases} 1 & \text{if } x \le y \\ \frac{x}{y} & \frac{y}{y} < \frac{y}{y}$ 

Infinitely-valued logics and applications

## Motivation and examples of applications

#### Fuzzy logics in neuro-symbolic approaches to AI





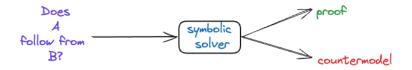
• Symbolically represent NN and their properties minimize NN, analyse NN behavior

 Multi-criteria layers using Łukasiewicz or other fuzzy connectives design inherently interpretable NN, reduce number of learning parameters

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#### T1 deductive systems

- focus on symbolic solvers allowing effective proof-search
- · extract explanations on why properties hold or not
- use, improve and compare with existing solvers



- T2 approximation strategies ...
- T3 extensions ...

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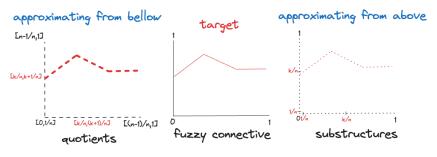
## Topics and goals...

T1 deductive systems ....

#### T2 approximation strategies

Find sequence of more tractable logics  $\mathcal{L}_1, \mathcal{L}_2, \dots$  approximating  $\mathcal{L}_{target}$ 

- from above  $\mathcal{L}_1 \supseteq \mathcal{L}_2 \supseteq \ldots \supseteq \mathcal{L}_{target}$  focus on counter-model generation
- from bellow  $\mathcal{L}_1 \subseteq \mathcal{L}_2 \subseteq \ldots \subseteq \mathcal{L}_{target}$  focus on proofs generation

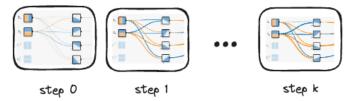


E.g. finitization techniques

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- T1 deductive systems ...
- T2 approximation strategies ...
- T3 extensions
  - Combine infinite-valued logics with other logics
  - Control interaction to find tractable fragments



E.g. combine with temporal logic to study the evolution of NN and its properties during the learning process

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