



Artificial  
Intelligence  
Vlaanderen/Flanders



## Extending neurosymbolic AI to infinity and beyond

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This is fine for simple tasks  
but **dangerous** for sensitive applications

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We create expressive neurosymbolic systems  
with improved scalability

- 1 The building blocks needed to express knowledge
- 2 Approximate probabilistic inference improves scalability
- 3 Gradient estimation maintains differentiability

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light ~ cat([pr, po, pg]).  
speed ~ normal( $\mu$ ,  $\sigma$ ).
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```
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```

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```
safe :-  
    light = r, speed < 5.  
safe :-  
    light = o, speed < 20.  
safe :-  
    light = g, speed > 10.
```

Rules express **higher-order knowledge**, which defines constraints that should hold

- 1 The building blocks needed to express knowledge
- 2 **Approximate probabilistic inference improves scalability**
- 3 Gradient estimation maintains differentiability

# Approximate probabilistic inference improves scalability


Probability of satisfying a logical property over a neural distribution

$$\mathbb{P}(q) = \int \mathbb{1}_{\omega \models q} p_{\theta}(\omega) d\omega$$

The diagram illustrates the components of the equation  $\mathbb{P}(q) = \int \mathbb{1}_{\omega \models q} p_{\theta}(\omega) d\omega$ . An orange line connects the word "Probability" to the term  $\mathbb{P}(q)$ . A blue line connects the phrase "logical property" to the indicator function  $\mathbb{1}_{\omega \models q}$ . A green line connects the phrase "neural distribution" to the probability density function  $p_{\theta}(\omega)$ .

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We decomposed integration into smaller problems  
and use approximate probabilistic inference

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- 3 **Gradient estimation maintains differentiability**

Gradient estimation maintains differentiability  
to optimise neural networks through logic

$$\approx \nabla_{\theta} \mathbb{P}(q)$$

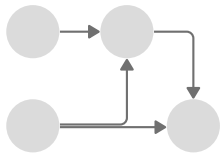
Gradients are needed to improve neural nets  
based on the provided expert knowledge



# Gradient estimation maintains differentiability to optimise neural networks through logic

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Gradients are needed to improve neural nets  
based on the provided expert knowledge



We decomposed the gradient for lower variability  
leading to improved optimisation

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expressive and scalable neurosymbolic AI

A framework that...

...can express knowledge  
ranging from finite to infinite domains

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Expressivity used in energy grid use case (Vito)

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A framework that...

...can express knowledge  
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...scales under certain conditions  
**but** different approaches are necessary

...controls neural networks through optimisation  
for tasks like detection and image generation

You can trust LLMs at least a little more in the future  
because of neurosymbolic AI