## **OpenNARS:**

## Autonomous Learning and Decision-Making

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Properties of the Non-Axiomatic Reasoning System (NARS)

Designed to work with insufficient knowledge and resources, featuring:

- Real-time operation
- Being open to unexpected tasks
- Learning from experience
- Dealing with uncertainty

=> High degree of autonomy

## **NARS** Overview

Inference Engine:

- Memory
- Control
- Logic

## Memory

Containing concepts:

- Named by terms
- Semantically linked
- Containing task statements

## Memory



## Control

Probabilistic premise selection:

• Favoring high priority items

In each step:

- 1. Select concept C, and task T from it
- 2. Select a belief B from neighbour concept D
- 3. Apply inference using T and B as premises
- 4. Feed results back into correct concepts

## Concept

Some concept roles:

- Represent a certain pattern in experience
- Keeping track of the evidence about the pattern it represents. (Revision)
- Allow remembering relevant preconditions, execute operation goals (Decision)
- Interact with task/event buffer (Temporal Reasoning, "Top-down Attention")

## Memory&Control: Desired Properties

- Allow often appearing patterns to stabilize:
  - Perception viewpoint: making them easier to be re-observed
  - Cognition viewpoint: Allow useful knowledge to "survive" and to be highly priorized in the right context.
- Allow mental flexibility: Concepts being composed and decomposed by the reasoning process.

## **Big Picture**



## Logic: Desired Properties

- Allow sufficient expressiveness for the encodings the system will need to form
- Ability for deductive, inductive and abductive reasoning
- Allow to deal with uncertainty

## Non-Axiomatic Logic

• Evidence-based Truth Value

Featuring:

- Deduction
- Induction (Learning-related)
- Abduction

## From evidence to truth value

Positive evidence w<sub>+</sub> Negative evidence w

Frequency: w<sub>+</sub> / (w<sub>+</sub> + w<sub>-</sub>)
Confidence (w<sub>+</sub> + w<sub>-</sub>) / ( (w<sub>+</sub> + w<sub>-</sub>) + k )

## **Truth Value**



## **Observing temporal patterns**

Example, Sequences: A, B events, derive compound event: (A,B)

## Example

#### given observation





## Predictive hypothesis generation

#### Example, Temporal Induction: A, B events, derive hypothesis: A => B

## Anticipation

Given A => B is believed, and given A is observed: predict B

Predicted event B wasn't observed? Generate negative evidence for B

allows A => B to get revised.

## Example

## Conditioning based on observed event sequence a,b,c,a,a,a,b,c,b,b,a,b,c

how => c ?

## Example

# Conditioning based on observed event sequence a,b,c,a,a,a,b,c,b,b,b,a,b,c how => c ?

a => c ? b => c? (a,b) => c?

## Procedural knowledge

Conditioning to acquire procedural knowledge:

(a,b) => c

(precondition, operation) => goal

## Microworld

2D environment, featuring:

- Agent in bird-view perspective
- Goal to capture green objects
- Goal to avoid red objects

## Microworld

Perception:

• Pixel-based 1D-retina from agent perspective

Actions:

• Rotate left, rotate right, move forward

## Input Representation

Input as events of the form:  ${pixel_i} \rightarrow [on] \% degree\%$ Operations:

$$(*, \{SELF\}) \rightarrow ^{left}$$

## Microworld





### Goal: Move to the right side Issue: Obstacles are in the way Solution: Jump over it

## Improvisation through analogy

• "Toothbrush" decision making example

## problem solving



#### problem solving Solutions: 1. Melt cup, 2. Properly re-shape cup can bebent and also a solution: 1. Melt toothbrush made ofplastic 2. Re-shape toothbrush by comparison .. after letting a properly reshaped object cool down, it can then potentially be used can bebent to unscrew the screw. plastic made ofmelted can becauses heat generatespliable can be and distant and a second second re-shaped can turn hardened can be causes cooling

Thank you!