

Anchoring Knowledge in Interaction:
Towards a harmonic subsymbolic/symbolic
framework and architecture of computational
cognition

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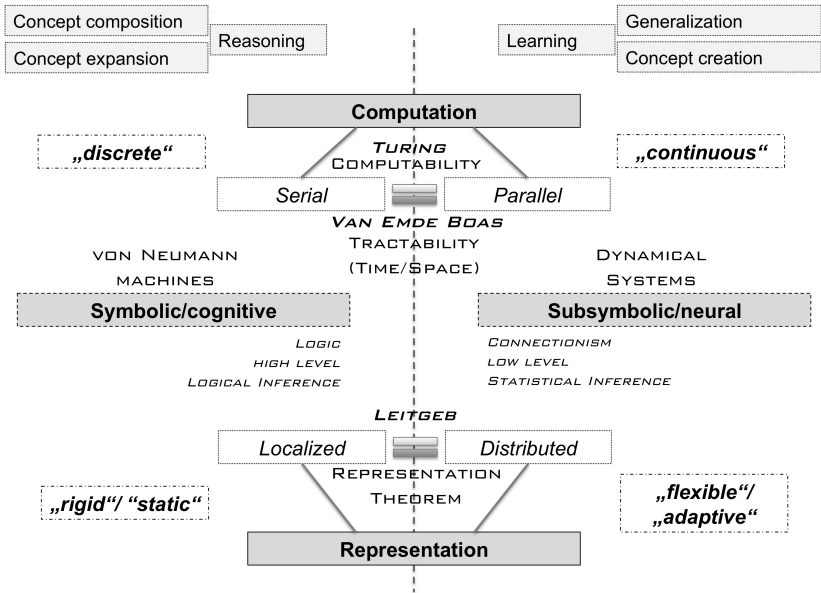
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Prelude: The Status Quo in Neural-Symbolic Integration



Overview of the Talk

- 1 A Harmonic Analogy: Coupled Layers of Knowledge from Embodied Interaction to Symbols (and Back Again)
- 2 The Core Ideas and Objectives
- 3 Steps Towards an Implementation: A Sketch of an Architecture
- 4 Going Far Beyond Multi-Level Data Fusion



Natural agents seem to rely on...

- ...enormous richness of representations (multimodal, grounded, embodied, situated).
- ...many layers of representation at different abstraction levels.
- ...dynamic re-organization of knowledge.
- ...dynamic changes or alignments of representation (e.g., in agent-agent interactions).
- ...online and bidirectional learning in real-time.
- ...flexible adaptations to changes in the environment, the task(s), the “social setting” (presence of other agents), etc.



A Harmonic Analogy

Conceptually similar situation in music:

- Different levels: Physical level (audio data), MIDI level, chord progression level, harmonic level, melodic level, rhythmic level, score level, structural level of a piece, (semantic) meta-level, etc.
- Transfer/interaction between levels:
 - Sometimes obvious mappings: MIDI to score to harmonic structure.
 - Sometimes partial or incomplete mappings: Harmonic structure to score, rhythmic to physical level.
 - Sometimes fuzzy or tentative mappings: Melody to harmony (in an idiom), physical to structural level of a piece.
 - Sometimes there are no mappings: MIDI to semantic/meta level, melodic to structural to harmonic level.

⇒ Piece of music as multi-layered, multi-representational entity with certain connections and constraints between layers (relations, mappings, etc.).



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The music/knowledge analogy:

- Changing one layer in a piece of music influences (in an obvious, partial, or fuzzy way) many (but not all) other levels.
- Multi-representational analysis can be used to learn or detect mappings between layers, novelties and correlations, to systematically unfold specific properties, or to find invariant properties.
- Envision an agent system also operating on different levels of representations:
 - Neural layer learning on the perception/motor level.
 - Anchoring layer learning elementary (semi-)symbolic representations of objects.
 - Reactive layer taking over in critical situations.
 - Deep learning layer learning on more abstract levels.
 - Symbolic layer for reasoning and planning.
 - (Higher) Symbolic layer providing core ontology.
- Some layers have obvious, some have partial, some have fuzzy, some have no mappings/relations between themselves.



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A “pre-established harmony”:

- Triggering abstract plan to move from A to B should result in corresponding motor action, classifying (on the neural level) a perceptual input as chair should activate the concept “chair” in the ontology, etc.
- Basic links might be hard-coded,...
- ...learning a new concept on the subsymbolic level should nonetheless result in a new concept entry in the ontology.

⇒ Interaction between the different layers in terms of information and conceptualizations.

⇒ Simulated or actual system operating on interacting levels in multi-representational manner should allow for mechanisms/interactions similar to music case.



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The Core Ideas and Objectives

- Developing, theoretically and practically, a conceptual framework and corresponding architecture that model an agent's knowledge, thinking, and acting as interrelated parts of a unified cognitive capacity.
- Knowledge as...
 - ...multi-layered phenomenon appearing at different levels of abstraction.
 - ...promoting interaction between levels.
 - ...influenced by interaction between agent and environment (potentially including other agents).
- Radically new paradigm in...
 - ...interaction styles: Action-centered, embodied, multi-modal.
 - ...knowledge repositories: Different levels and forms of knowledge representation, e.g., multi-modal, hybrid.
 - ...user modeling and communication through learning and adaptation.



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Scientific aims:

- Embodiment level:
 - **Learning of elementary forms of multi-modal representation from agent interaction with environment.**
 - Emphasize the importance of sensorimotor interactions as part of knowledge formation.
 - Systematic assessment of basic learning signatures in the presence of different sensorimotor experiences.
 - Recommendations for the development of cognitively-inspired formal frameworks for embodied computation.
 - Together with approaches from computational neuroscience and network-level cognitive modeling create cognitively-inspired foundations and low-level input representations for subsequent stages.



Scientific aims:

- Anchoring level:
 - Representations resulting from embodiment level may be noisy, uncertain, vague, differ in representation languages between agents, subject to changes in the environment, etc.
 - Remedy: **Expand anchoring framework in robotics to grounding not only objects, but also certain general observable properties appearing in the environment.**
 - Top-down and bottom-up anchoring during learning.
 - Dynamic introduction of new symbols for new objects and categories by repair and concept invention mechanisms.
 - Denotations of a symbol used in communication must be consistent across communicating agents.
 - Enable the establishment of analogical links across agents.



Scientific aims:

- Neural level:
 - Embodiment view provides interaction-based neural representation of knowledge not represented at conceptual level.
 - Remedy: **Specify lifting procedure producing descriptions, i.e., lifting grounded situations and agent's action patterns to more abstract (symbolic) representations.**
 - Combine neural learning with temporal knowledge representation using variations of RBM models.
 - Validate hypotheses through symbolic description of trained networks while robustly dealing with uncertainty/errors through Bayesian inference model.
 - Use conceptual spaces (Gärdenfors) to link symbolic and sub-symbolic data.
 - Additionally combine this with analogy-making and corresponding transfer mechanisms between representation systems.



Scientific aims:

- Knowledge level:
 - Lifted multi-modal representations can be error-prone, different agents possibly use distinct/incompatible languages, etc.
 - Remedy: **Develop domain-independent dynamic re-organization of knowledge based on ontology repair mechanisms, analogy, concept invention, and knowledge transfer.**
 - Enable adaptation of agent to new situations, alignment between representations across agents, reformation of knowledge entries, and generation of new knowledge.



Overall account:

- Grounding knowledge in cognitively plausible multimodal interaction paradigms.
- Lifting grounded situations into more abstract representations.
- Reasoning by analogy and concept blending at more abstract levels.
- Repair and re-organization of initial and generated abstract representations.



Five thrusts:

- 1 Cognitive Foundations of Knowledge.
- 2 Anchoring Knowledge in Perception, Action, and Interaction.
- 3 Lifting Knowledge from the Subsymbolic to the Symbolic Level.
- 4 Analogy/Blending.
- 5 Concept Formation/Reformation.



Conceptual commitments:

- *How does knowledge develop from the concrete interaction sequences to the abstract representation level?*

The crucial aspect is the lifting of grounded situations to more abstract representations.

- *How can experience be modeled?*

Experience can be explained by deep learning.

- *How is deeper understanding of a complex concept made possible?*

Theory repair makes precisely this possible.

- *To which extent do social aspects play a role?*

Analogical transfer of knowledge between agents is a central aspect concerning efficient and flexible learning and understanding.



Steps Towards an Implementation

Conceptors and similar approaches are used for abstraction levels based on interaction



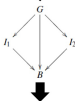
Objects can be anchored in a hierarchy of concepts



Hierarchical anchoring is the basis for lifting procedures moving knowledge from subsymbolic representations to symbolic ones



Two input theories I_1 and I_2 result in a generalized theory G and blend space B



Analogy and concept blending results in a new theory T_2 containing a new (potentially inconsistent) concept C

$$T_1 \rightarrow T_2[C]$$

Cognitive Foundations of Knowledge

Anchoring Knowledge in Perception, Action, and Interaction

Lifting Knowledge from the Sub-symbolic to the Symbolic Level

Analogy / Blending

Concept Formation / Reformation

The technical perspective of anchoring imposes constraints on perception, action, and interaction for cognitive agents

The resolved concept C' and the resolved theory $T_2' [C']$ might be fed back to update the generalized theory G and the blend B

$$G \rightarrow G'$$

$$B \rightarrow B'$$

The inconsistent concept C is resolved by reformation into a concept C' resulting in a new theory $T_2' [C']$

$$T_2[C] \rightarrow T_2'[C']$$



Data fusion: *“data fusion techniques combine data from multiple sensors and related information from associated databases to achieve improved accuracy and more specific inferences than could be achieved by the use of a single sensor alone.”*

Difference in ambition:

- Development of a cognitively-inspired combination of low-level sensing with high-level reasoning in attempt of anchoring (symbolic) knowledge in (subsymbolic) perception and (inter)action in continuous feedback loop.
- Significant step towards (re-)creation of foundation for cognitive capacities and forms of reasoning in next generation AI systems.
- Major progress towards development of computational test bench and agent model for theories from cognitive science.



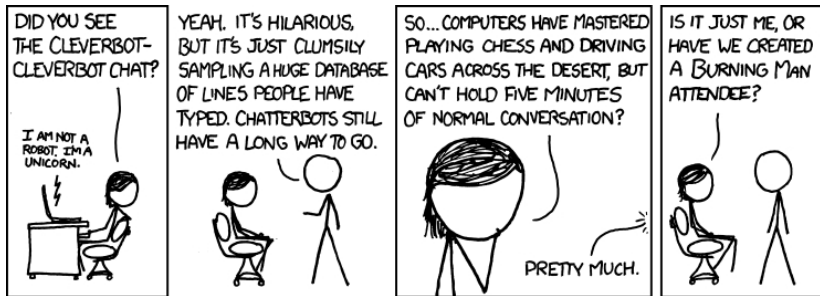
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(Definitely Not) The End



(XKCD #948)

Questions, comments, criticism, ideas,...?

Get in touch:

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