

Cognitive social evaluations for multi-context BDI agents

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IIIA - Artificial Intelligence Research Institute
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- 1 Introduction and Motivation
- 2 The Repage Model and Multicontext BDI agents
- 3 Embedding Repage in a Multi-context BDI Agent
- 4 Conclusions and Future Work

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Reputation in Multi-Agent Systems (MAS)

Control of Interactions

Approaches to control interactions among agents:

- **Security Approach:** Integrity and authenticity of messages, agent's identity validation...
- **Institutional Approach:** Institutions observe agents' behaviors, and may punish them.
- **Social Approach:** Reputation mechanisms are here.

Mechanism to control agents' behavior in MAS when:

- Agents in open environments.
- Agents may have unknown intentions.
- Agents need to interact to each other to achieve their goals.

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- Several models have appeared in literature:
 - **Centralized approach:** eBay, SPORAS, HISTOS ...
 - **Distributed approach:** Regret, AFRAS, Schillo et al., Yu and Singh...
- Among them, **Repage:**
 - Computational system based on a cognitive theory of reputation.
 - Fundamental distinction between **Image** and **Reputation**
 - **Image:** Social evaluation that is **believed** by the agent.
 - **Reputation:** Social evaluation that has been **spread** by agents.

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Current state-of-the-art work

Focus on model definitions, not in the integration with the agent's architecture.

Global Objective

Integrate Repage information into a BDI agent architecture.

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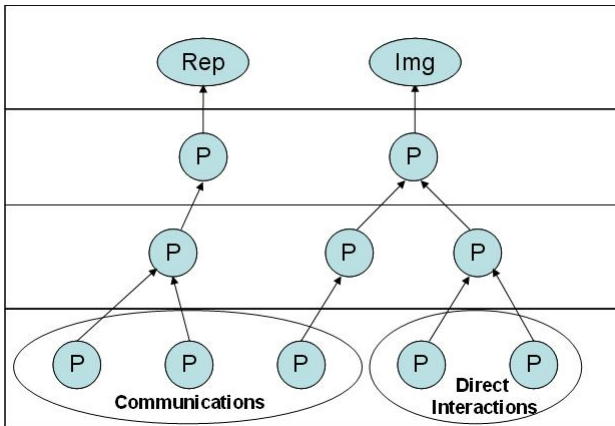
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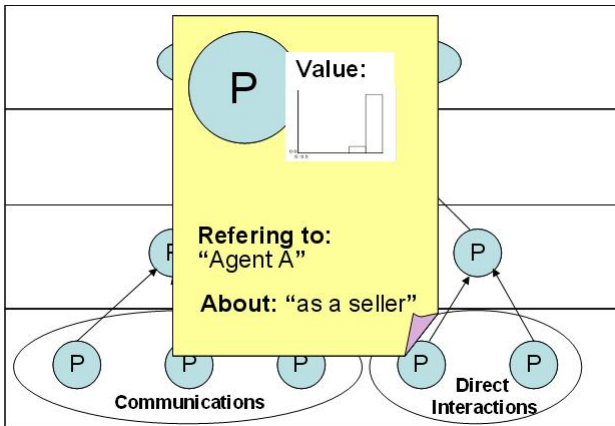
The Repage System: Modeling Image and Reputation

Building Image and Reputation



The Repage System: Modeling Image and Reputation

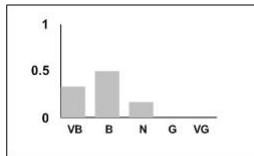
Value Representation



The Repage System: Modeling Image and Reputation

An Example

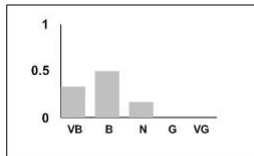
Image of A as a *seller* is



Implies that in the next direct interaction with A , agent **believes** that

- with a prob. of 0.3 it will be **very bad**
- with a prob. of 0.5 it will be **bad**
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Multi-Context Systems(MCS)

Definition

Let I be the set of context names, a MCS is formalized as

$\langle \{C_i\}_{i \in I}, \Delta_{br} \rangle$

- $C_i = \langle L_i, A_i, \Delta_i \rangle$, where L_i is a formal language with its syntax and semantics, A_i is a set of axioms and Δ_i the a set of inference rules.
- Δ_{br} is a set of bridge rules.

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Bridge rules are inference rules to exchange information between contexts:

$$\frac{C_{i_1} : \varphi_1, \dots, C_{i_n} : \varphi_n}{C_{i_x} : \varphi_x}$$

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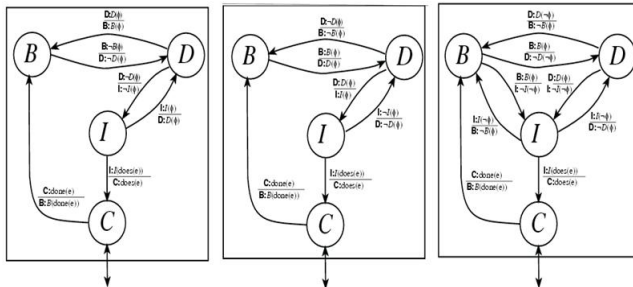
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A Multi-Context BDI Agent Specification



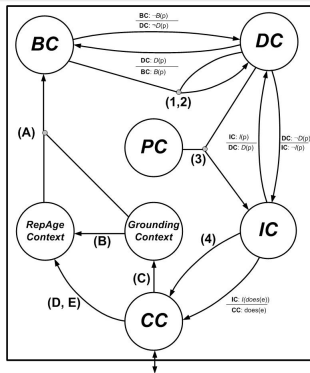
1

Relation between attitudes: strong realism, realism and weak realism.

¹Source: Parsons *et al.* Agents that reason and negotiate by arguing. *Journal of Logic and Computation*, 8(3):261-292,1998. Page 272

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Embedding Repage in a Multi-context BDI agent



2

General Features

- Lukasiewicz Logic (many-valued logic): So, predicates become fuzzy, and attitudes are graded.
- Dynamic Logic: $[\alpha]\varphi$, after executing action α , the formula φ holds.

²Based on: Casali *et al.* Graded models for BDI agents. *Proceedings of CLIMA* MA-CSIC
Pages 18-33. 2004

Belief Context

- **Crisp:** If $\varphi \in L_D$ then $\varphi \in BC$
- **B-Modal:** Modal operator B over **crisp** formulas, together with a truth constant \bar{r} where $r \in [0, 1]$. Then:
 - If $\varphi \in L_D$ then $B\varphi \in BC$
 - If $r \in [0, 1]$ then $\bar{r} \in BC$
 - If $\alpha, \beta \in BC$ then $\alpha \rightarrow_L \beta$ and $\alpha \wedge_L \beta \in BC$. Here, the connectives \rightarrow_L and \wedge_L (conjunction and implication of Lukasiewicz multi-valued logic).

Example

$(B[\text{buy}(\text{alice})](\text{haveCar} \wedge \text{hasPaid}(7500)), 1)$

$(B[\text{buy}(\text{alice})]dT(0, 1), .5)$

$(B[\text{buy}(\text{alice})]dT(3, \infty), .2)$

$(B[\text{buy}(\text{alice})]vgCar, .3)$

$(B[\text{buy}(\text{alice})]regularCar, .4)$

⋮

Desire Context

- **Generic Desires:** $(D^+\varphi, \delta)$, $(D^-\varphi, \delta)$: Level of satisfaction/disgust if φ holds is δ .
- **Realistic Desires:** $(D^+[\alpha]\varphi, \delta)$: Expected level of satisfaction of φ if α is executed is δ .

Example

$(D^+(haveCar \wedge vgCar \wedge dT(0, 1)), 0.9)$

$(D^+(haveCar \wedge goodCar \wedge dT(1, 3)), 0.7)$

$(D^-dTTime(3, \infty), 0.8)$

$(D^-(vbadCar \vee badCar \vee regularCar), 0.7)$

Intention Context

- $(I[\alpha]\varphi, \gamma)$: Trade-off between expected satisfaction level of φ after executing α , and its cost.

Desire and Intention Context

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Bridge Rules

Generating realistic desires

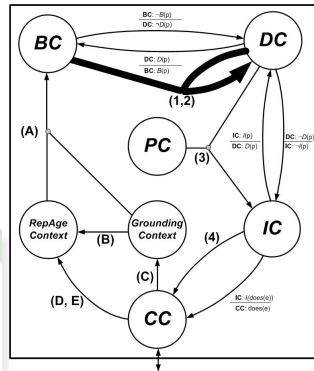
- **Generic Desire** ($D^+ \varphi, d_\varphi$)
- **Belief** about getting φ through the **action** α : ($B([\alpha]\varphi), p_\varphi$)
- **RESULT**: **Realistic desire** ($D^+[\alpha]\varphi, g(d_\varphi, p_\varphi)$)

Example

$DC : (D^- dT(3, \infty), 0.8)$

$BC : (B[buy(alice)]dT(3, \infty), 0.2)$

$DC : (D^- [buy(alice)]dT(3, \infty), 0.16)$

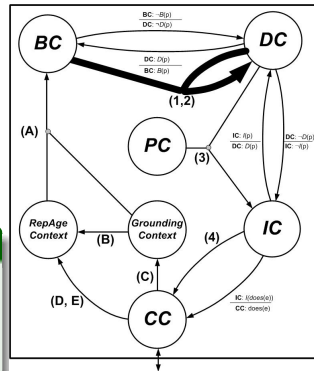


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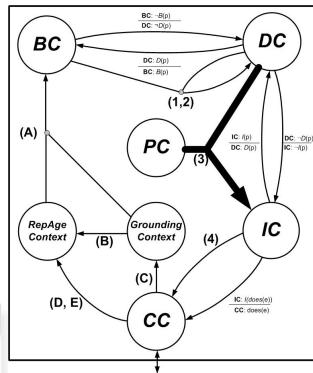
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- **A positive realistic desire** ($D^+[\alpha]\varphi, \delta$)
- **All negative realistic desires** over the same action α .
 $(D^-[\alpha]\psi_1, \delta_{\psi_1}), \dots, (D^-[\alpha]\psi_n, \delta_{\psi_n})$
- **Cost** of the action α : $action(\alpha, c)$
- **RESULT: Intention** of getting φ through α : $(I[\alpha]\varphi, f(\delta - \sum_{i=1}^n \delta_{\psi_i}, c))$

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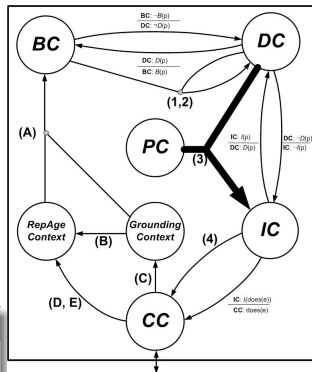
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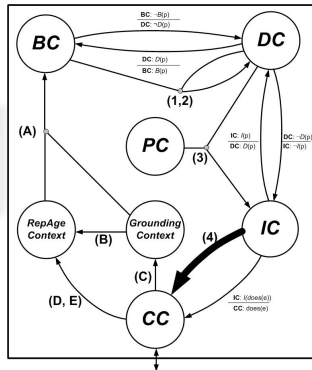
Bridge Rules

Selecting the action

$$4: \frac{IC : (I[\alpha]\varphi, i_{max})}{C : does(\alpha)}$$

Example

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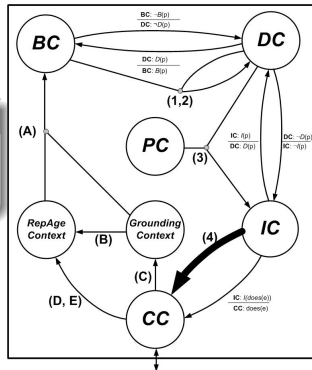
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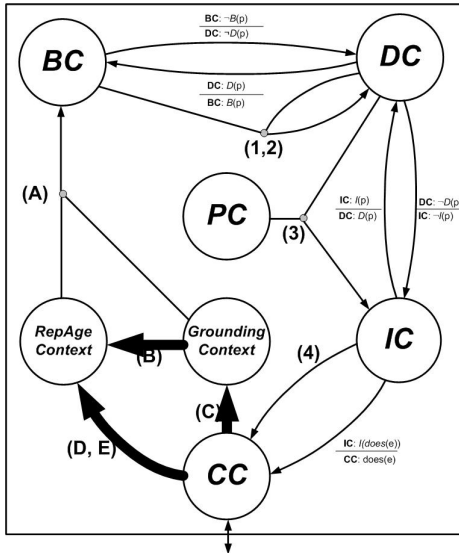
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Grounding and Repage Context

Direct Experience in RepAge

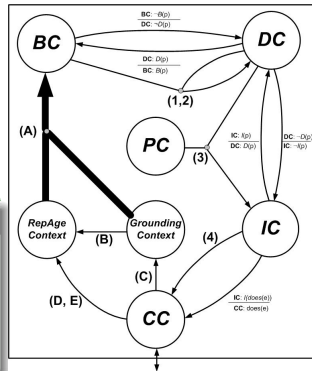


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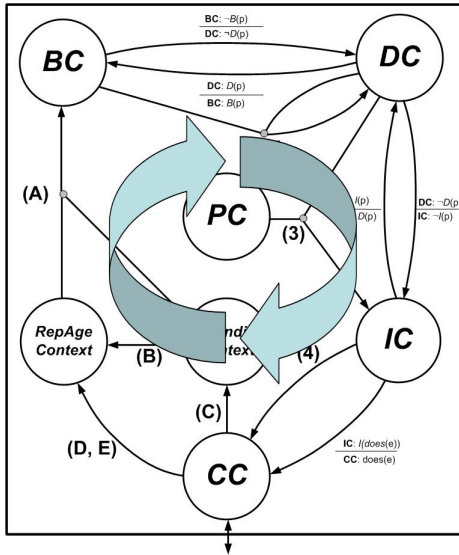
Generating beliefs from image predicates

- **Image Predicate** $Img(Ag, r, [v_1, v_2, \dots])$
- **Grounding discretization functions:**
 $\varphi_1(r), \varphi_2(r), \dots$
- **Action associated to r :** $\phi(r)$

Example

$$\frac{Img(alice, seller(dTime), [0.2, 0.3, 0.5])}{\begin{array}{l} (B[buy(alice)]dT(3, \infty), 0.2) \\ (B[buy(alice)]dT(1, 3), 0.3) \\ (B[buy(alice)]dT(0, 1), 0.5) \end{array}}$$


Ending...



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Conclusions and Future Work

- Incorporate Reputation (**done**)
- Implementation / Simulations
- Incorporate proactivity to solve cognitive dissonances.
- Argumentation.
- Ontology alignment.
- ...

Thanks for you attention!