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Vienna University of Technology



Institut für
Computertechnik
Institute of
Computer Technology

Artificial Recognition System

Development and Evaluation

Samer Schaat, Alexander Wendt, Matthias Jakubec,
Friedrich Gelbard, Lukas Herret, and Dietmar Dietrich

Artificial Recognition System (ARS) Project

- General-purpose model of human information processing for the usage in various artificial systems

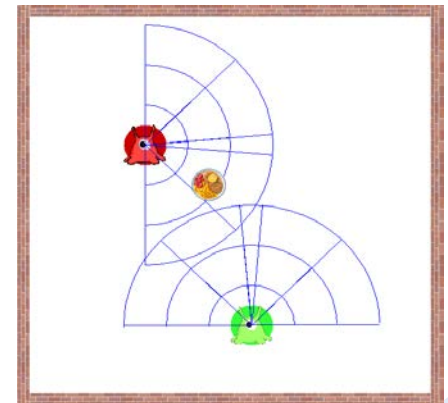
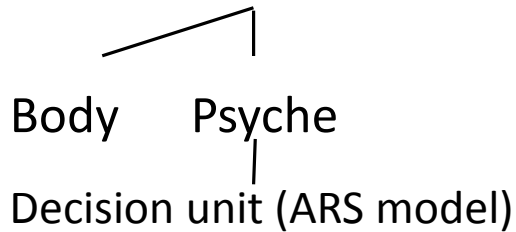


Human-Robot Interaction (Kismet)



Evacuation Simulation (ESCAPES)

- Humanoid agents in a virtual world



Key Features of the ARS Approach

- Functional model

Generative approach: describing functions not behavior → generic, flexible

- Layered description model

*Appropriate means of description for different **aspects** (neurons, neurosymbolics, psyche)*

- Holistic and unitary model

Consistent and coherent integration of basic aspects (motivation, emotion, planning...)

- Top-down approach

Concretize abstract functions incrementally, starting with psychic layer

- Bionic and interdisciplinary approach

Translate knowledge into technical models



Basic question:

How to develop and evaluate such a model?

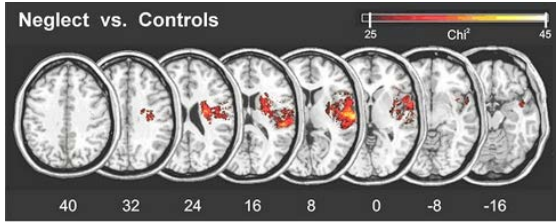
Challenges

- Restricted accessibility of mind's functioning
- Interdisciplinary understanding and knowledge translation
- Complexity in description and explanation



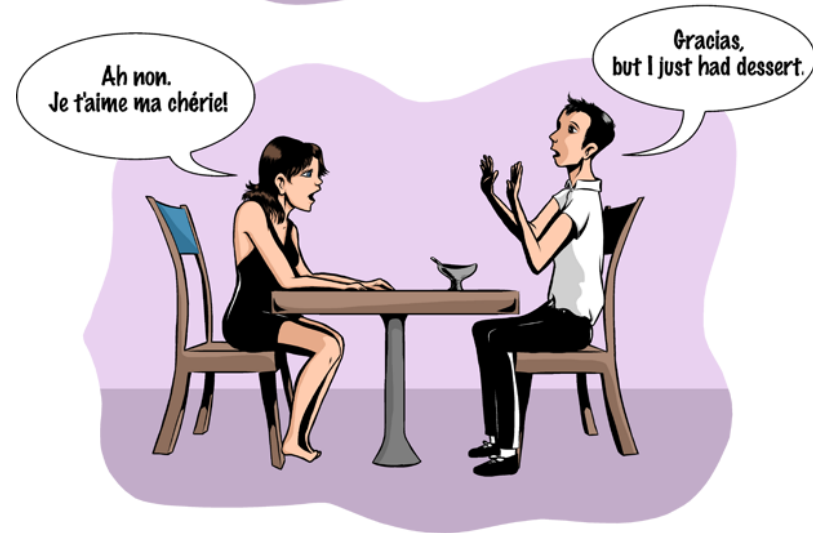
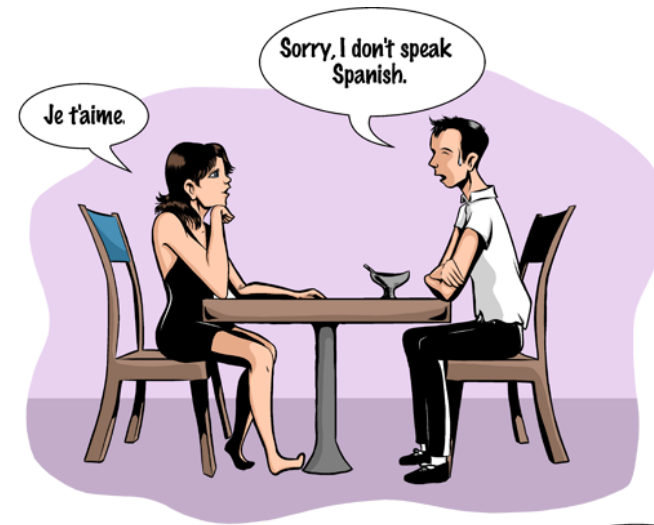
Restricted Accessibility of the Mind

- Various ways to get information about the mind's functioning
- Relevant knowledge for our objective? Right level (psyche)?
- Cannot be used directly
- Interpretation and knowledge translation required → Experts needed



Interdisciplinary Understanding

- Regular, intensive collaboration
- Different concepts, vocabulary....?



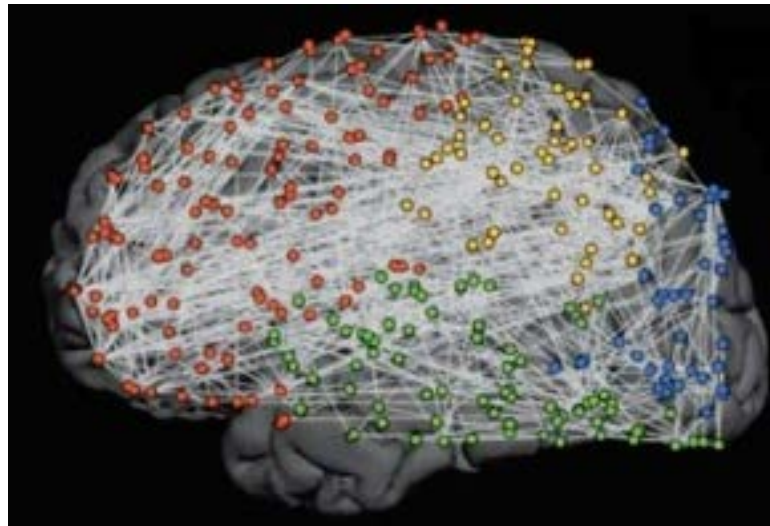
REC Dialogue Tales Misunderstanding

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Complexity and Explanation

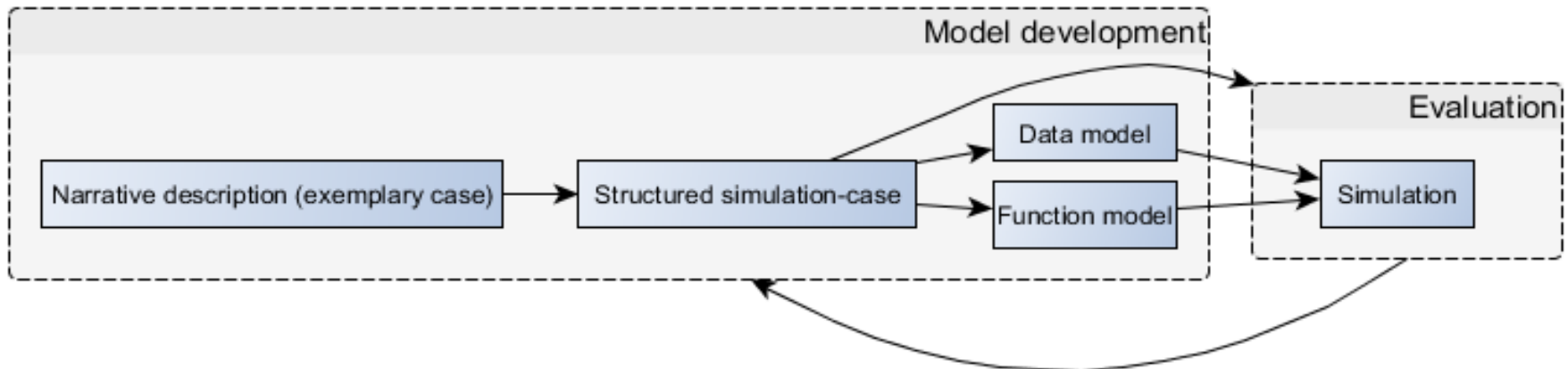
- Right level, relevant knowledge?
- Not only on neuronal level, also on psychic level
- Interplay of various factors determine behavior



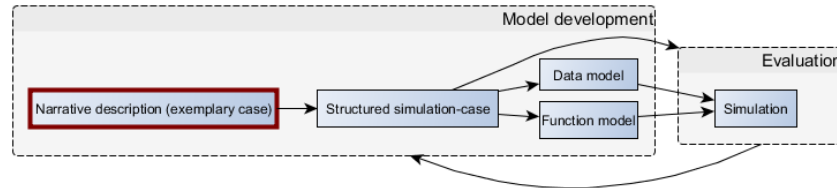
Case-driven Agent-based Simulation

Combination of

- Casuistics for interdisciplinary collaboration
- UC-based requirement analysis for deterministic structuring
- Agent-based simulation as a evaluation framework



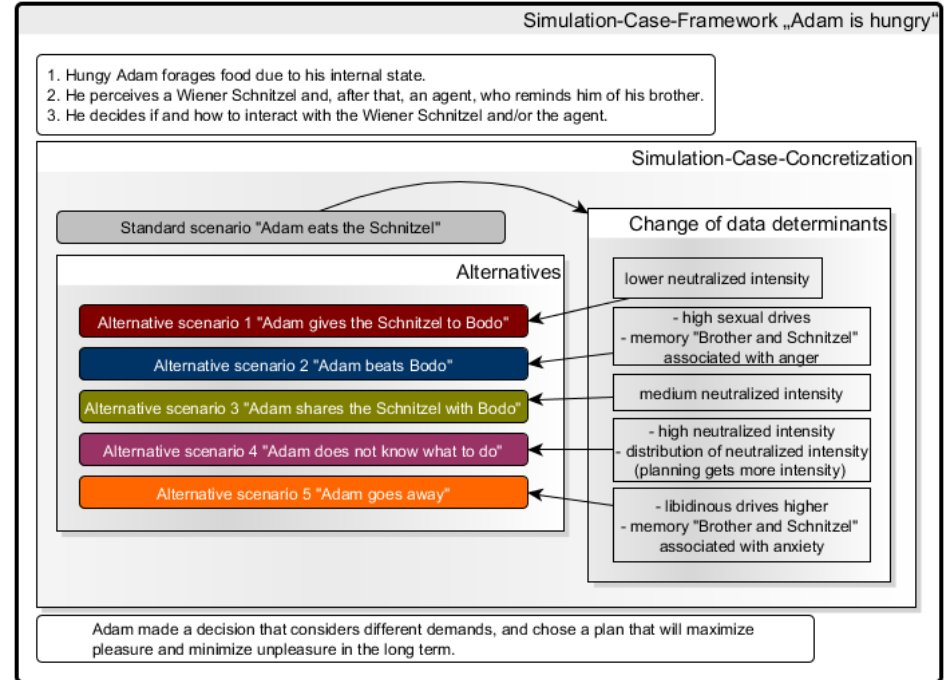
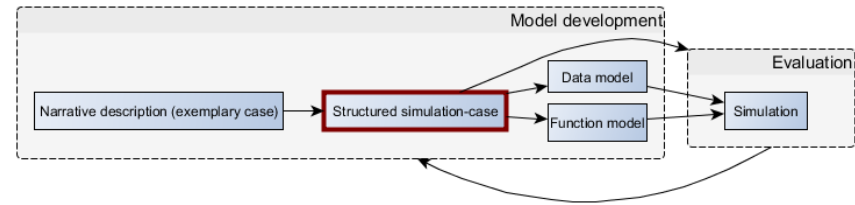
Step 1: Describe phenomena and assumptions



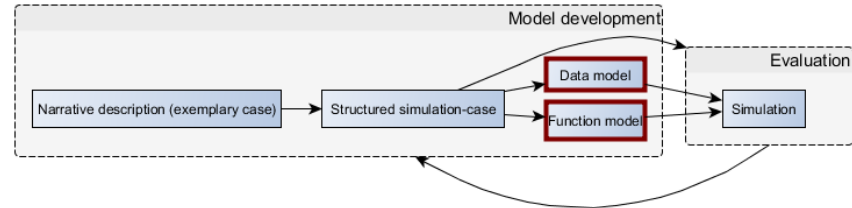
- Platform and tool for interdisciplinary collaboration
- Exemplify and discuss research question with a concrete *exemplary* case
e.g. How two hungry agents behave in front of a food source (eat, share...)
- Enables stating (and testing) concrete assumptions
(e.g. the role of emotions, drives, and norms)
- Avoids drifting into abstract discussions
- Embodies and integrate theories from different disciplines to explain behavior
State of the art, experts' interpretation of real world conditions
- But: indeterministic, gaps in assumptions, inconsistent → no direct usage

Step 2: Analysis and Structuring

- Clarify the exemplary case
 - Explication of assumptions
 - Consistent description
- Structure to deterministic description
 - Causal function description
 - Data determinants of behavior (*Memories, personality, environment, internal state*)
- Simulation-case (SC) enables
 - Requirements analysis
 - Computational model
 - Test plan for evaluation

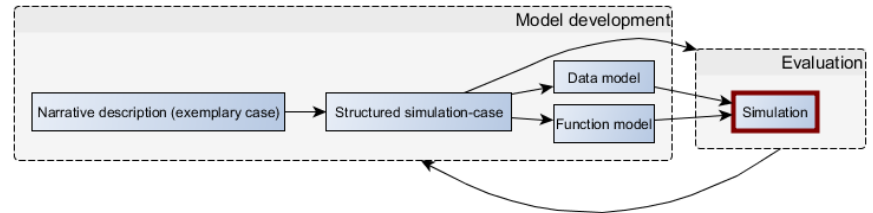


Step 3: Data and Functional Model

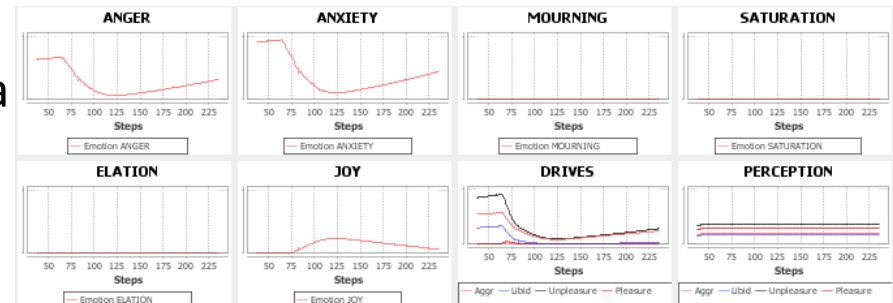


- Previous steps enable
 - Requirements statement
 - Algorithmic description of functions
 - Modelling of knowledge representation
- Specify function modules, interfaces, data
Adaption or extension?
- Implemented in MASON (Java) and Protégé (Ontology)

Step 4: Evaluation

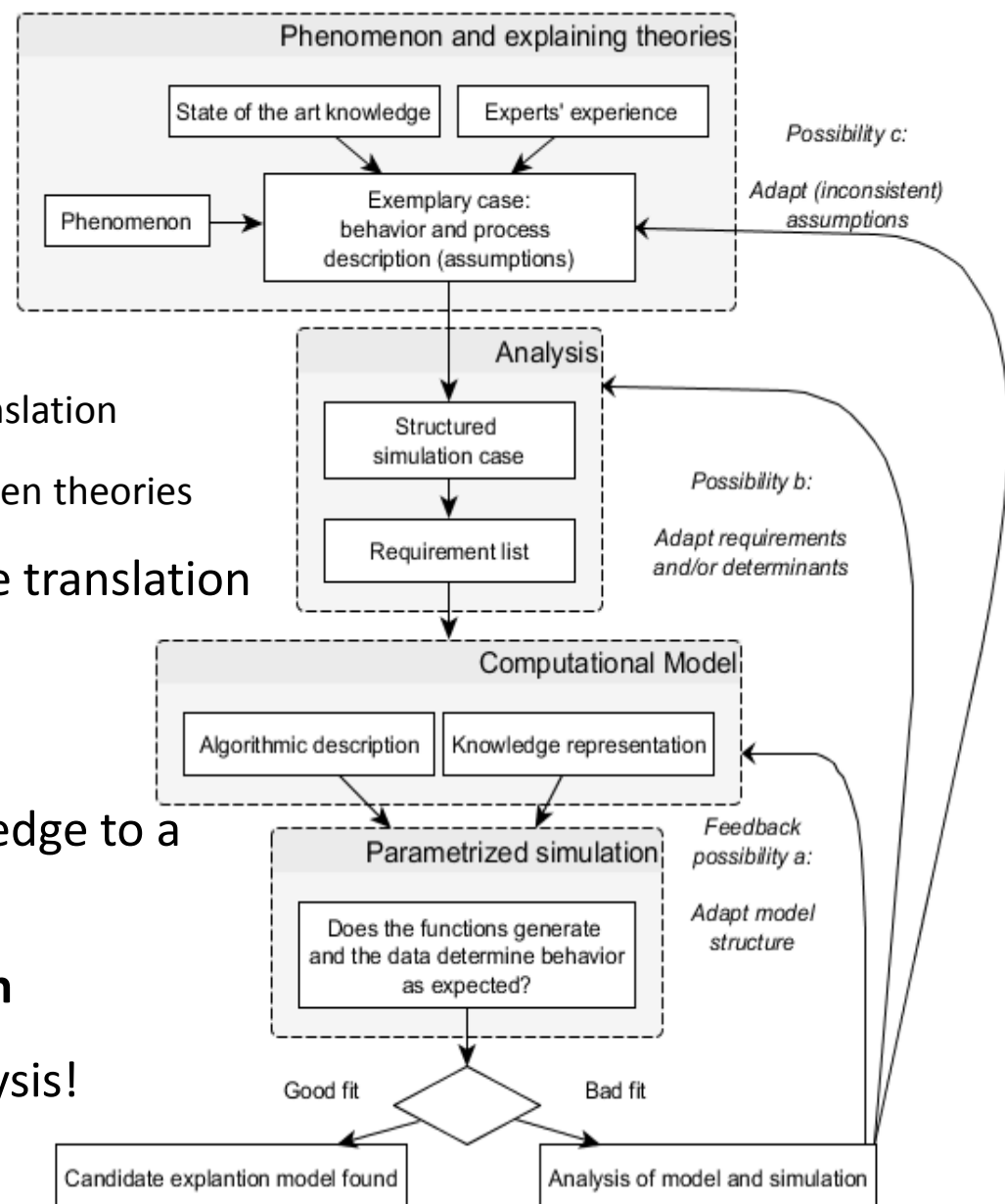


- Simulation-case as test-template → parameterize simulation according to scenarios
- Does the functions generate and data determine behavior as expected?
- How is the behavior generated?
- Test our hypotheses' predictability
 - Are the assumptions of exemplary case valid?
 - Does the interplay of specified factors (e.g. emotions, drives, norms) generate the expected behavior?
 - Does the specified data determine behavior (change)?
- Unexpected behavior or state → analysis on different levels → feedback cycles



Conclusion

- Feedback cycles
 - Possibility a, b: mistake in model translation
 - Possibility c: inconsistent in or between theories
- Bridge disciplines, test knowledge translation
- Concretize testable assumptions from other disciplines
- Structure interdisciplinary knowledge to a causal model and test plan
- SC scenarios → **model calibration**
- Stable model? → sensitivity analysis!
- Premises for model application in specific domains → Outlook





Thank you!