Hybrid Intelligent Systems

Lecture 17 Future of HIS and AGI. Emotions, motivation, consciousness

Can an intelligence create another intelligence more intelligent than itself?

- Conditions
 - Same or similar to mind principles of thinking and learning
 - More speed of action
 - More volume and accessible of memory
 - Available of evolution

What is AI similar to our mind

- Similarity of learning
- Similarity of perception
- Similarity of memory
- Similarity of thinking
 - Motivation
 - Emotions
 - Goals
 - Consciousness

Developing a consistent view on emotion-oriented computing. Marc Schroder and Roddy Cowie



Equipping a Lifelike Animated Agent with a Mind L. Chen, K. Bechkoum, and G. Clapworthy IVA 2001, LNAI 2190, pp. 72-85, 2001.



Components of motivation

Motivation Category	Interpretation for an agent	Weights
Physiological	Drinking	W phys
Safety	Avoid collision & hostile agents	W safe
Affiliation	Cooperation, social activity, Helping others	W_{aff}
Achievement	Achieve its own goals	Wach
Self-Actualisation	Learning & creative activity	W _{icam}

Table 1. Static motivation hierarchy

Model of motivation

SoM_{it} =
$$W_{catei} \left(\sum_{k} D_{kt} W_k + \sum_{j} I_{jt} W_j \right)$$
.

name <motiName> static prio < one of the five category weights> participants <names of participants> incentives <set of (first-order) formula or variables> drives <set of drive formula or variables> drive change <temporal function > motivation strength < a calculation function based on Equation (2)>

Fig. 2. Abstract description of a motivation

Algorithm of motivation arbitration

- 1) Set the motivational profile of the agent.
- 2) Specify the initial situation such as internal drives, incentives, goals and objects. Then at each time step:
- 3) Compute the strength of all motivations and rank them according to their magnitudes.
- 4) Choose the first motivation Fm.
- 5) Compare the Fm with the current active motivation. Make the motivation with bigger magnitude active.
- 6) Map the active motivation to its corresponding goal g. (in Goal Generation module)
- 7a) Pursue the active goal (in the Behavior Engine)
- 7b) Continue the loop from the step 3 while the agent proceeds to step 6 and 7a.

Abstract description of goal

- goal_name < goalName>
- goal_descr < a set of formula >
- participants <name of involved objects or agents>
- importance < integer >
- preconditions <a set of formula>
- commitment status < one of active, successful, rejected and failure>
- plan < controller>

Emotion representation

- emotion_name < emName>
- type <one of the 24 types>
- intensity < integer >
- causes < set of formula>
- directional <the bearer of the emotion effect>



Fig. 5. Agent architecture

Algorithm of motivation arbitration with emotions

- 1) Set the motivational profile of an agent.
- 2) Specify the initial situation such as internal drives, goals and objects.
- At each time step:
- 3) Implement cognitive appraisal and generate the current emotional state.
- 4) Compute the strength of all motivations taking into account emotion influence and rank them according to their magnitudes.
- 5) Choose the first motivation Fm.
- 6) Compare the Fm with the current active motivation. Make the motivation with bigger magnitude active.
- 7) Map the active motivation to its corresponding goal g. (in Goal Generation module)
- 8a) Pursue the active goal (in the Behavior Engine)
- 8b) Continue the loop from the step 3 while the agent proceeds to step 7 and 8a.



Fig. 6. Structure of emotion-expressive actions

The interface agent in virtual environment



Table 2. Motivation modelling of interface agents

Motivation	Motivations	Drives	Stimuli	Goals
category				
Physiological	Drinking	Thirst	Water	Godrinking
Safety	Collision avoidance	Security	Obstacles	CollisionAvoida
-				nce
Affiliation	Greeting	Sociability	Agents	Greeting
	Provide information	cooperation	Upon request	ForwardInfo
Achievement	Achieving its own tasks	Accomplish	Task-oriented	Goto(pos) or
		tasks		Do(thing)
Self-	Self-amusing	Boring	None	Selfamusing
Actualisation	Navigation & exploration	Curiosity	Unknown area	Navigation
	Learning,			
	creative activity			

Motivated Agents.

Kathryn Kasmarik, William Uther, Mary-Lou Maher,



The Mechanism of Thought, **Robert Hecht-Nielsen** IJCNN-2006, Vancouver, BC, Canada, pp. 1146-1153 symbol (neuron collection) representing color red unidirectional symbol neuron collection-to-neuron (neuron collection knowledge link collection) representing word apple cortical module representing words describing mental Cerebral Cortex world objects Knowledge links are formed between cortical module meaningfully co-occurring symbols, representing colors essentially as describing mental postulated by Hebb

world objects

Theory of confabulation

- COGNITIVE WORLD OBJECT REPRESENTATION
 - Each module is used to represent one *attribute* that an *object* (visual, auditory, conceptual, abstract, motor process, thought process, plan, etc.) of the cognitive mental universe may possess. This representation takes the form of the selection of a single *symbol* from among a set of (typically) thousands of symbols implemented by the module for describing its object attribute
- KNOWLEDGE LINKS
 - Each knowledge link connects the neuron collection representing one symbol (termed the source symbol of the link) to neurons representing a second symbol (termed the target symbol of the link – usually a symbol in a different lexicon from that of the source symbol)
- CONFABULATION
 - each module is also responsible for carrying out the *confabulation* operation (a 'winners-take-all' competition process among the symbols of the module)
- THE ORIGIN OF BEHAVIOR
 - every time a confabulation operation being carried out by any module yields a *definitive conclusion* (namely, one symbol – not multiple symbols or no symbol), a set of *action* (movement process and/or thought process) *commands* associated from that particular conclusion symbol are immediately launched



Confabulation – the lexicon-level view. Confabulation is a winnerstake-all competition between the symbols of a module, based upon each symbol's total knowledge link *input excitation*

The conclusion-action principle



Confabulation theory

- Fundamental underlying mathematical operation of cognition is to find that symbol ε of the answer module which maximizes *cogency* p(αβγδ|ε).
- Cogency is the probability of the assumed facts being true, given an assumption that the symbol ε is true.
- In other words, confabulation theory claims that each decision making process involved in cognition is selection of that conclusion which is most supportive of the employed assumed facts being actually true

Principles of action of intelligent systems

A.V. Gavrilov, IST-2003, Novosibirsk and in <u>http://www.mind-</u> <u>consciousness-language.com</u> (2007)

• The principle of associative recall

When interacting with the environment, the intelligent system stores the associations between different images (stimuli, signals, signs, actions, etc.) which it uses to plan and execute its behavior. The associations originate from a recall process (reading associations from memory) of images based on their fragments. The associations that are involved during the storing process are reinforced. When sufficiently consolidated, the associations can be used to denote the relations (attributive, cause and effect, case, etc.) between entities

The principle of concentration and economy of resources

 Intelligent systems (natural or artificial) have a mechanism that selects (recognition) and activates the information resources (neurons, neural ensembles, frames, rules, etc.) that are essential to the solution of an actual task by the intelligent system, and that deactivates the resources that are not essential to the solution of an actual task

The principle of uncertainty

 The more precise the recognition of the structure of an image, i.e. its internal semantics, the less accurate the recognition of its interaction with other images, i.e. its external semantics

• The principle of unity in fuzzy reasoning and certain other operations

The basis of reasoning lies in operating with fuzzy images by means of a process of associative recall of images (see Principle No. 1). At the end of the process, a choice of certain operations is carried out (recalling of it): it is therefore possible to associate the successful choice (the solved task) with the focusing of attention, the start of operation as programs of operation motor neurons, etc. Thus the selected operation is involved as a tag in the further process of reasoning