Social Learning for Social Robots

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HCT Seminar University of Sussex, 24 Nov 2006 The Adaptive Systems Research Group is a multidisciplinary group of faculty, students, and friends of the University of Hertfordshire who have a connection to research in adaptive systems and related areas: computer science, robotics, biology, mathematics, from psychology to engineering.

Areas of research:

- Artificial Life
- Socially Intelligent Agents/Robots
- ► Artificial Intelligence (Embodied)

Key Academics:

- Prof. Kerstin Dautenhahn (coordinator)
- ► Prof. Chrystopher Nehaniv
- ► Dr. René te Boekhorst
- ▶ Dr. Lola Cañamero
- Dr. Daniel Polani
- \sim 14 Postdocs/Research Assistants
- \sim 15 PhD students

One of the great challenges in bringing robots out of laboratories into real-world domestic and public environments is to address the fact that humans and robots have different modalities, and exploit this knowledge for successfully operate in the same space. This talk will present current work by the Adaptive Systems research group at the University of Hertfordshire on social interaction and social learning between humans and robots, in shared environments.

Imitation is a powerful learning tool when a number of agents interact in a social context. Robots capable of imitation (or other, simpler, forms of social learning) would allow humans to interact with them and transfer knowledge in a more adaptive and "natural" (for the humans) way.

Learning how to do an act from seeing it done. [Thorndike 1898]

Copying of a novel or otherwise improbable act or utterance, or some act for which there is clearly no instinctive tendency. [Thorpe 1963]

Something C(opy) is produced by an organism or machine, where C is similar to something else M(odel), registration of M is necessary for the production of C, and C is designed to be similar to M. [Mitchell 1987]

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Second-level: At this level the organism perceives the model and produces a copy, but is not concerned about the accuracy and the correspondence of this copy.

Third-level: Comparisons between the produced copy and the model can be made by the organism. Learning is involved and if the replication is repeated, the organism will try to better match the model.

Fourth-level: Variation into the copy is introduced by the self-aware organism, changing some aspects to conform to the model and others to extend it in some way.

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Imitation is best considered as the behaviour of an autonomous agent in relation to its environment, including other autonomous agents. The mechanisms underlying imitation are not separated from the behaviour-in-context, including the social and non-social environments, motivations, relationships among the agents, the agent's individual and learning history etc. [Dautenhahn and Nehaniv 2002]



Who to imitate?

- ▶ When to imitate?
- ▶ What to imitate?
- ► *How* to imitate?
 - ► Evaluation . . .



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Given an observed behaviour of the model. which from a given starting state leads the model through a sequence (or hierarchy [or program]) of *sub-goals* in *states*, *actions* and/or *effects*, one must find and execute a sequence of actions using one's own (**possibly dissimilar**) embodiment, which from a corresponding starting state, leads through corresponding sub-goals - in corresponding states, actions, and/or effects, while possibly responding to corresponding events. [Nehaniv and Dautenhahn 2000, 2001, 2002]



ALICE Framework



We have developed ALICE (Action Learning via Imitating Corresponding Embodiments), a generic framework for solving the correspondence problem. The ALICE framework builds up a library of actions from the repertoire of an imitator agent that can be executed to achieve corresponding actions, states and effects to those of a model agent (according to given metrics and granularity); it provides a functional architecture that informs the design of robotic systems that can learn socially from a human demonstrator. [Alissandrakis et al. 2002, 2003, 2004]













Rabit (Robotic Arm emBodiment for Imitation Testbed)

Agents are two-dimensional robotic arms with different number of rotary joints of varying length, $L = [l_1 \ l_2 \ l_3 \ ... \ l_n]$.

Action: vector describing the change of angle for each of the joints,

 $A = \left[\alpha_1 \ \alpha_2 \ \alpha_3 \ \dots \ \alpha_n \right].$

State: vector containing absolute angle for each of the joints, $S = [\sigma_1 \ \sigma_2 \ \sigma_3 \ \dots \ \sigma_n]$. Each action takes the arm from a *previous* state to a *current* state.

Effect: a directed straight line, implemented as the vector of displacement between *previous* and *current* end tips of the arm,

 $E = (\mathbf{x}_{c} - \mathbf{x}_{p}, \mathbf{y}_{c} - \mathbf{y}_{p}).$

Behaviour: a sequence of actions,

$$B = (A_1 \quad A_2 \quad \dots \quad A_n)$$





An example of cultural transmission among heterogeneous agents.









A. Alissandrakis (UH)



Using data captured from a human and given appropriate metrics and sub-goal granularity, the multi-target system can produce action command sequences that when executed by a software or hardware agent can achieve corresponding actions, states and/or effects. The corresponding actions, states and effects as demonstrated by the imitator can also be captured and used as a demonstration for another imitating agent. Differently embodied and constrained target systems in various contexts need to be supported. [Alissandrakis et al. 2005]



Given the demonstration, in a different context (here initial configuration), where on the workspace should the object be moved to?



Should the same *relative displacement* be achieved?



Should the same absolute position on the workspace be reached?



Or should the same *relative* (to the other objects) *position* be reached?

... complement the JABBERWOCKY work (towards a characterization of the space of effect metrics) with work on the formalization of body mappings via correspondence matrices [Alissandrakis et al. 2006].

Different agent bodies can be described as simplified kinematic models, comprising of a rooted acyclic connected graph of *segments*.



State

(or posture) of such a kinematic model can be defined the vector S containing the values of the degrees of freedom (DOF).

Action

can be thought as motion, or the amount of change in the DOFs required so that a posture transforms to another, and can be defined as the difference between two consecutive state vectors S and S': A = S' - S.

Effects

can be defined as changes to the body-world relationship (e.g. location) of the agent and/or to positions, orientations and states of external objects.

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A first global¹ state metric can be defined as

$$\mu_{state} = \sum_{j=1}^{n} |S_{j}^{\alpha} - S_{j}^{\beta}|,$$

where S_j^{α} and S_j^{β} are the values of the state vectors for the two agents. Similarly, a first global **action metric** can be defined as

$$\mu_{action} = \sum_{j=1}^{n} |A_j^{\alpha} - A_j^{\beta}|,$$

where A_i^{α} and A_i^{β} are the values of the action vectors for the two agents.

¹Here, 'global' implies that both agents have the *same embodiment*, i.e. the same morphology and number of corresponding DOFs.

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For two agents, demonstrator α and imitator β with *n* and *m* DOFs respectively, a $n \times m$ correspondence matrix can be defined as

$$\mathcal{C} = \begin{bmatrix} w_{1,1} & w_{1,2} & \dots & w_{1,m} \\ w_{2,1} & w_{2,2} & \dots & w_{2,m} \\ \vdots & \vdots & \ddots & \vdots \\ w_{n,1} & w_{n,2} & \dots & w_{n,m} \end{bmatrix},$$

where the $w_{i,j}$ values are real-valued weights, determining how the j^{th} DOF of the imitator β depends on the i^{th} DOF of the demonstrator α . The j^{th} column of the matrix can be thought as a vector indicating how the DOFs of the demonstrator influence the j^{th} DOF of the imitator.

By multiplying an appropriate correspondence matrix C with the state and action vectors S^{α} and A^{α} of the demonstrator respectively, two new vectors in imitator coordinates can be produced:

$$S = S^{\alpha} \times C$$
$$A = A^{\alpha} \times C$$

Combining the global metric definitions with the induced vectors gives

$$\mu_{state}^{\mathcal{C}} = \sum_{j=1}^{m} |\mathcal{S}_j - \mathcal{S}_j^{\beta}| \epsilon_j$$
$$\mu_{action}^{\mathcal{C}} = \sum_{j=1}^{m} |\mathcal{A}_j - \mathcal{A}_j^{\beta}| \epsilon_j$$

where the corrective term

$$\epsilon_j = \left\{ egin{array}{c} 0 ext{ if } \sum_{i=1}^n w_{i,j}^2 = 0 \ 1 ext{ otherwise } \end{array}
ight.$$

takes the value zero if the j^{th} column of the correspondence matrix contains only zeros (effectively omitting the imitator's j^{th} DOF).

These new μ_{state}^{C} and μ_{action}^{C} metrics are called the *induced state* and *action metrics for the linear correspondence* C.

Intuitively, the components of S and A (for such $\epsilon_j \neq 0$) can be thought as the current subgoal state and action target values.









A different "abstract human" \rightarrow human mapping.

 $\label{eq:ROSSUM} \mbox{-} RObot \mbox{Self-imitation and } S\mbox{caffolding } U\mbox{tility } M\mbox{echanism } [Saunders et al. 2006a, 2006b]$

A robot social learning architecture that uses self-imitation with task and environmental scaffolding; the architecture is based on ideas from cognitive development with examples from social animals. *Zone of Proximal Development* (ZPD) - the gap between what the child can learn unaided and what can be learnt with the help a more capable peer - this guiding and support is called **scaffolding**. Vygotsky argued² that the learner learns based on their own sensorimotor experiences, *their own activity is at the centre of the learning process* (**self-imitation**). The approach is closest to that of *Extended Ideomotor Theory*³: *Similarity* between an event *perceived* and an event *learned* from the imitators own actions - *will induce that event*.

²J.N.Wertsch (1985) - Vygotsky and the social formation of mind. ³Wolfgang Prinz (1995-2005), William James (1890)

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Social Learning



The architecture *environmental scaffolding* to select relevant attributes and *task scaffolding* to build competencies.

The robot learning process is platform independent, constructed during learning and extensible.

A *prediction mechanism* is used to inform trainer of existing competencies.

Thank You!

Please check http://homepages.feis.herts.ac.uk/~comqaa1/ and feel free to send an email to a.alissandrakis@herts.ac.uk for further information.

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