# **Greedy Agents and Interfering Humans** An artwork making humans meddle with a life in the machine

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# New Media Arts with A-Life

 The arts by autonomous machines Cellular Automaton, L-System, Evolutionary Algorithms BOIDs, Ant Colony, Art by Robots Interactive Art (Human — A-Life interaction) Affecting the generative mechanism from visitors (vice versa) Interaction with Reinforcement Learner • cf. Creation process with A-Life, ex. Interactive Evolutionary Computing

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# Reinforcement Learning

 Learning by delayed reward (and punishment) Fundamental model of human/animal learning • Thorndike, E. L. (1898), Skinner, B. F. (1953) Classic framework (late 1980s —) TD, Actor-Critic, Q-Learning • Q-Learning by a lookup table ← Dynamic Programming Sutton, R. S. & Burto, A. G. (1998, 2018)

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### Environment



# Learning Environment

• 11×6 Grid World (approx. 16:9) • Agent behaviour: Start Cell  $\rightarrow$  Goal Cell Sensation: Cell's ID it is located • Action: Up, Down, Left, and Right Reward: given only when it reaches the Goal. Forced to return back to Start. Reset when a specified steps / goals passed.

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S: Start, G: Goal, A: Agent, filled cell: Obstacle





# Learning Mechanism

Classic Q-Learning by loop-up table

 $\Delta Q(s,a) = \alpha \cdot \left( r + \gamma \max_{b \in A(s')} Q(s',b) - Q(s,a) \right)$ 

 $P(a \mid s) = \frac{\exp(Q(s, a)/T)}{\sum_{b \in A(s)} \exp(Q(s, b)/T)}$ 

• Dyna-Q: Rehearses past experiences randomly.

accelerates the propagation of Q-values.

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## nteraction

- Detecting the visitors hands on the table.
  - by Kinect Azure on the ceiling.
- Placing obstacles
  - An obstacle is growing when a hand stays for a while.
  - Interfering the agent's move.
- Affecting the Q-Values
  - Hands motion in a range of speed modifies the vectors.

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## Installation (1)

Custom built table.

- Kinect Azure above the table.
- Ultra short focus projector.
  - 1920×1080 pixels
- Three stereo headphones
  - attached to table edges, Left, Front, Right

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# Installation (2)

- Windows box computer.
  - for Kinect Azure and image processing.
- Mac mini M2
  - for learning and visualisation.
- Mac mini M2
  - for sound generation and output.
  - with multi-cannel analog audio output.

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# Visualisation

- Distribution of Q-values in the grid world Four vectors for each action in each cell. • • Vector field by interpolating those vectors. Particle flow following the vector field Colouring by speed. 320K particles, 60 FPS.
  - M1 macMini (up to 1 million particles on M2 Ultra)

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## Sonification

 Providing sounds for each visitor through a headphone. Generating sounds of neighbouring area his/her hands placed. • Mixing the sounds generated using data of particle flow + agent movement. • Up to three visitors can enjoy it simultaneously.

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# Human lives together with A-Lives

- cf. Unemi, T. and Bisig, D.: Playing Music by Conducting BOID Agents A Style of Interaction in the Life with A-Life, ALIFE IX, Boston, MA, USA, 2004.
  - Collective behaviour, but fixed rules.
- Society of learners / teachers (mutual learning)
  - Oscillation would happen among eager learners.
  - How to regulate the diversity (mixture of lazy and eager individuals) by evolution?
- Evolutionary adaptation needs much experience by a population.
  - Hard to realise it as an interactive art?

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## Outlook

 Grid World → Continuous Coordinate System • Look-up table  $\rightarrow$  Artificial Neural Network, k-NN, etc. Continuous distribution of vector field. • Multi-agent learning Social interaction among agents and humans.

→ complex relations of cooperation and conflicts.

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