(Based on Chapter 6 of G. Weiss, (Ed.), Multiagent Systems: A Modern Approach to Distributed Artificial Intelligence, MIT Press, 1999.)

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What is Learning?

The acquisition of new knowledge and motor and cognitive skills and the incorporation of the acquired knowledge and skills in future system activities, provided that this acquisition and incorporation is conducted by the system itself and leads to an improvement in its performance

Differencing Features

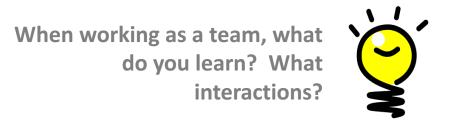
- Degree of decentralization
- Interaction-specific features
- Involvement-specific features
- Goal-specific features
- Learning method
- Learning feedback



Degree of Decentralization

- Concerning distributedness and parallelism
- One extreme is that a single agent carries out all learning activities sequentially
- The other extreme is that the learning activities are distributed over and parallelized through all agents in a MAS

Interaction-Specific Features



- Concerning interactions required for realizing a decentralized learning process
 - the level of interaction (ranging from pure observation over simple signal passing and sophisticated information exchange to complex dialogues and negotiations),
 - the persistence of interaction (ranging from short-term to long-term),
 - the frequency of interaction (ranging from low to high),
 - the pattern of interaction (ranging from completely unstructured to strictly hierarchical: peer-to-peer, broadcast, etc.), and
 - the variability of interaction (ranging from fixed to changeable) as some learning requires only minimal interaction, some maximal.

Involvement-Specific Features

- Concerning (a) the relevance of involvement and (b) role played during involvement
- Relevance: there are two extremes
 - the involvement of an agent is not a condition for goal attainment because its learning activities could be executed by another available agent as well; and
 - the learning goal could not be achieved without the involvement of exactly this agent

Role

- an agent may act as a "generalist" in so far as it performs all learning activities
- or it may act as a "specialist" learning a particular activity.



Scenarios where agents should learn to become generalists, or specialists?

Goal-Specific Features

- Concerning (a) the type of involvement that is tried to be achieved by learning and (b) the compatibility of the learning goals pursued by the agents.
- Type of Involvement
 - learning that aims at an improvement with respect to a single agent (e.g., its motor skills or inference abilities)
 - learning that aims at an improvement with respect to several agents acting as a group (e.g., their communication and negotiation abilities or their degree of coordination and coherence)
- Scenarios
 where agents
 learn to
 improve group?

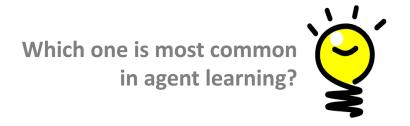
Scenarios where agents have

about pursuit games)

conflicting learning goals? (Think

- Compatibility of Learning Goals
 - Conflicting or complementary?

Learning Methods





Rote Learning

 direct implantation of knowledge and skills without requiring further inference or transformation from the learner, like primary/elementary school

Learning from Instruction and by Advice Taking

 operationalization—transformation into an internal representation and integration with prior knowledge and skills—of new information like an instruction or advice that is not directly executable by the learner

Learning from Examples and by Practice (or by Doing)

 extraction and refinement of knowledge and skills like a general concept or a standardized pattern of motion from positive and negative examples or from practical experience

Learning by Analogy

 solution-preserving information of knowledge and skills from a solved to a similar but unsolved problem

Learning by Discovery

discovering and gathering new (novel) knowledge and skills by making observations

Learning Feedback

- Supervised learning: Often offline
- Concerning how to give (what) feedback on performance level achieved so far
- Supervised Learning
 - Specifies the desired activity of the learner and the objective is to match this desired action as closely as possible
- Reinforcement Learning
 - Specifies the utility of the actual activity of the learner and the objective is to maximize this utility,
- Unsupervised Learning
 - no explicit feedback is provided and the objective is to find out useful and desired activities on the basis of trial-and-error and self-organization processes



RL a popular approach in MAS. Why? Environment too complex, can't prescribe the desired activities that would lead to goal states





Learning to communicate

- Learning is viewed as a method for reducing the load of communication among individual agents
 - communication usually is very slow and expensive, and therefore should be avoided or at least reduced whenever this is possible

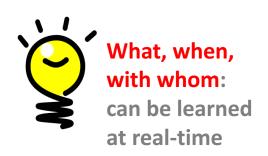
Communication as learning

- Communication is viewed as a method for exchanging information that allows agents to continue or refine their learning activities
 - learning is inherently limited in its potential effects by the information that is available to and can be processed by an agent





- What to communicate
 - e.g., what information is of interest to the others
- When to communicate
 - e.g., what efforts should an agent investigate in solving a problem before asking others for support
- With whom to communicate
 - e.g., what agent is interested in this information, what agent should be asked for support
- How to communicate
 - e.g., at what level should the agents communicate, what language and protocol should be used, should the exchange of information occur directly—point-to-point and broadcast—or via a blackboard mechanism



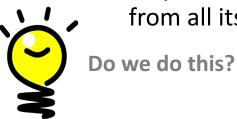


Example | Reducing Communication by Addressee Learning

 Broadcasting is costly; direct communication paths are not always known

Addressee Learning

- to reduce the communication efforts for tasks announcement by enabling the individual agents to acquire and refine knowledge about the other agents' task solving abilities
 - Suppose that agent a has previously asked agent b to help with task T_i successfully
 - Suppose that now agent a encounters a new task T_j
 - If T_i is very similar to T_j , then a can decide to approach b again, instead of asking help from all its neighbors



Example | Improving Learning by Communication



- Agents cannot be assumed to be omniscient without violating realistic assumptions. In general, agents have incomplete information about:
 - the environment in which it is embedded and the problem to be solved
 - other agents
 - the dependencies among different activities and the effects of one own's and other agents' activities on the environment and on potential future activities.

Two forms

- Learning based on low-level communication: relatively simple query-andanswer interactions for the purpose of exchanging missing pieces of information (knowledge and belief) – shared information
- Learning based on high-level communication: more complex communicative interactions like negotiation and mutual explanation for the purpose of combining and synthesizing pieces of information shared understanding

Connection to MAS?

Single agent learning: can ignore other agents, and focus on learning about the environment



Multiagent learning: more complex but often necessary in order to improve performance



Learning is powerful, adds autonomy to agents! But it is complex



Silly Question: In a 1-fish competition, each contestant can only submit one fish. Each contestant has 4 hours to fish. Each contestant is accompanied by a referee. The referee's job is to measure the fish for the contestant, accept one and only one submission, and deliver the submitted fish to the authority. The contestant with the longest fish wins the competition. What would you do as a contestant?

