

Sistemas de Multiagentes

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Multigent systems and their relation to AI?





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Multigent systems and their relation to AI?







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Some references

A Concise Introduction to Multiagent Systems and Distributed Artificial Intelligence







Main Characteristic:

Autonomy, the agents must be autonomous in their actions and decisions.







Agents' features (intelligence?):

Reactivity to the environment

Proactiviness, planning to reach a goal







Formalizing it:

Environment: $\{e_0, e_1, e_2, e_3,\}$

Agents' actions:
$$\{A_0, A_1, A_2, A_3,\}$$

Sequence for the model: $\{(e_{t=0}, A_{t=0}), (e_{t=1}, A_{t=1}), (e_{t=2}, A_{t=2}),\}$





Decisions:

The agents must select an action based on the environment:

- state of the environment e_t
- state of the agent: a_t

$$\mathsf{P}(\mathsf{A}_{t+1}) = \mathsf{f}(\mathsf{e}_0, \mathsf{e}_1, \ldots, \mathsf{e}_t, \mathsf{a}_0, \mathsf{a}_1, \ldots, \mathsf{a}_t, \mathsf{A}_1, \ldots, \mathsf{A}_t)$$





Decisions:

The agents can be

- Reactive: $P(A_{t+1}) = f(e_0, e_1, ..., e_t)$
- "Planning" with a goal: they have a payoff function that attempt to optimize.



Memory:





- Infinite memory: $P(A_{t+1}) = f(e_0, e_1, \dots, e_t, a_0, a_1, \dots, a_t, A_1, \dots, A_t)$
- Markovian: $P(A_{t+1}) = f(e_t, a_t, A_t)$







Updates:

- Asynchronous: agent by agent
- Synchronous or Parallel: all the agents update to t+1 at the same time depending on the state of the system in t

Communication





Structure of the communication network:

- All-all, meanfield
- Global variables
- Network



Communication





Structure of the communication network:





Communication



Structure of the communication network:







John von Neumann y Oskar Morgenstern (1944): The theory of games and economic behavior (zero-sum games)

John Nash (1950): Equilibrium points in n-person games

John Maynard Smith (1982): Evolution and the theory of games Evolution (biological)

William Hamilton and Robert Axelrod (1981):

Robert Axelrod (1984): The Evolution of Cooperation











Ultimatum







most famous and influential idea models in the social sciences! (~3,000,000 results on Google Scholar)







General problem:

- The origin of cooperation.
- Whenever there is a conflict between self-interest and the common good.
- You are tempted to do something, but know it would be a great mistake if everybody did the same thing.

'The origins of virtue', Matt Ridley (1996)

Nash equilibrium:

•A set of strategies (one per player) from which no player benefits by changing unilaterally

•A set of strategies such that each one of them is a **best response** (highest payoff) to the joint strategies of the rest



Ask what each player would do, *taking into account* the decision-making of the others: Each player is told the strategies of the others. Suppose then that each player asks himself or herself: "Knowing the strategies of the other players, and treating the strategies of the other players as set in stone, can I benefit by changing my strategy?"

If any player would answer "Yes", then that set of strategies is not a Nash equilibrium. But if every player prefers not to switch (or is indifferent between switching and not) then the set of strategies is a Nash equilibrium.

The largest pay-off is not necessarily achieved at the Nash equilibrium.



A unique Nash equilibrium does not exist for every game



John Maynard Smith

Evolutionary version of Game Theory:

i) Players not required to be rationalii) Player required to have a strategyiii) Multiplayer game

Strategies are not fixed, the question is dynamical: How strategies are selected on time by interaction?

Strategy:

Classical theory: players have strategy sets from where to choose their actions *Biology:* species have strategy sets from which every individual inherits one

Interactions:

Classical theory: one-shot games and iterated games *Biology:* random and repeated pairing of individuals, with strategies based on their genome and not on the past

Equilibria:

Classical theory: Nash equilibrium *Biology:* Evolutionary stable strategy (ESS)

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Playing games

Axelrod's Prisoner's Dilemma Tournaments

- In the 1980s, Axelrod organized two tournaments and invited many scientists and mathematicians to submit strategies (n-person games).
- The strategies played iterated games against one another in a round-robin fashion.
- Some strategies were quite complicated e.g., creating complex predictive models of various opponents

Conclusions on best strategies:

- Be Nice (never be first to defect)
- **Be Forgiving** (be willing to cooperate if cooperation is offered)
- Be Retaliatory (be willing to defect if others defect against you)
- Be Clear (be transparent about what your strategy is make it easy to infer)

And the winner is (Anatole Rapoport)

TIT FOR TAT: Start out by cooperating. Then at each successive round, do what the other player did on the previous round.

Simplest of all strategies and Nice, Forgiving, Retaliatory, Clear

Towards the market: minority game



W. Brian Arthur, "Inductive Reasoning and Bounded Rationality", American Economic Review **84**, 406 (1994).













(http://www.transtats.bts.gov/)

Socio technical systems

Planes and delays

- Total cost of flight delay in US in 2007 was 41B dollars.
- In the EU, the direct cost is around **2B** euros •
- Rich transport dynamics. .
- Cascading failure. ٠

25%

0%



- Aircraft Arriving Late
- Security Delay
- National Aviation
- Extreme Weather







(http://www.eurocontrol.int)

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Databases: Planes and delays

- Airline On-Time Performance Data (<u>BTS (USA), CODA Eurocontrol-EU)</u>)
- > Schedule & actual departure (arrival) times
- Origin & destination airports
- Airline id
- Tail number
- 2010 flights (USA):
- > 6,450,129 flights (74 %)
- 18 carriers
- ➢ 305 airports
- 2013 flights (EU):
- > 20,000 flights/day
- > 50 carriers
- ➢ 320 airports

Network:

- Nodes: airports
- Edges: direct flights between airports
- Node attributes: average delay per flight

Planes and delays

Clusters:

- Formed by airports in problems
 > average delay per flight > T min
- Must be connected (flight route between them)
- A group of airports connected by flights that their average delay is higher than T minutes





Planes and delays

• March 12, 2010

Average delay per delayed flight:
53.2 min





Planes and delays







Initial Conditions

- From the data...
 - Known \rightarrow when, where and the departure delay for the first flight of the sequence.
- Random initial conditions...
 - Fixed initial delay (min)
 - % of initially delayed planes





Planes and delays







Planes and delays

Data and model comparison for March 12 and April 19, 2010





Planes and delays

• With random initial conditions...



- Each day is potentially a bad day, if some initial conditions are met.
- Flight connectivity is a key factor for the rise of congestion in the network.
- Sensitivity to initial conditions.



Planes and delavs





Mobility and cities





Mobility and cities





Mobility and cities

MATSIM + Phone users agenda





Mobility and cities





Mobility and cities





Modal split



_ 24.5%	23.0%	22.5%					
- 23.9%	24.8%	25.2%					
51.6%	52.2%	52.3%					
Baseine in or toll peripheral in 9							
Private transport	Not motoris	sed 🔲 Public transpor					



Mobility and cities







Conclusions



Needs: Data & knowledge on the decision process



- Agents, communication, decision making
- Characterization
- Calibration
- Validation
- Scenario analysis