



CentraleSupélec

# Agent and Multi-Agent Systems: architectures and reasoning

## Concepts and Definitions

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27.09.2021

CentraleSupélec- SAFRAN AI Training

## **Organization of the course**

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# Organization

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## Dates

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27/09/2021 Course 1: Agent and MAS: Concepts (Course+PW)

28/09/2021 Course 2 : Multi-Agent based Simulation (Course+PW)

29/09/2021 Course 3 : Interaction mechanisms: models and Implementation (Course+PW)

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## Artificial Intelligence (AI)

” Designing and building machines that do things that would require intelligence if performed by humans” [ Marvin Minsky].

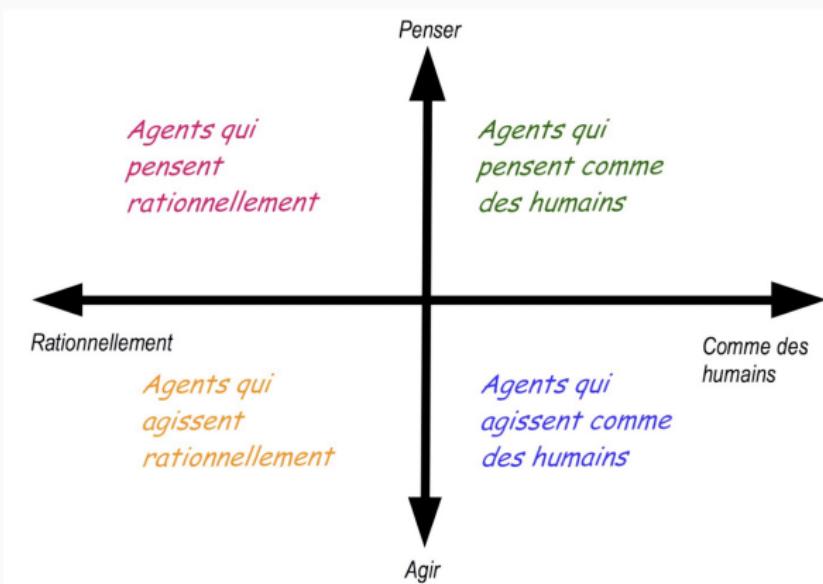


Source: <https://humanoïdes.fr/robot-appris-jouer-échecs-tout-seul/>

# Artificial Intelligence: several definitions!

Penser comme des humains	Penser rationnellement
<p><i>The exciting new effort to make computers think... machine with minds, in the full and literal sense.(Haugeland,1985)</i></p> <p><i>The automation of activities that we associate with human thinking, activities such as decision-making, problem sloving, learning.(Bellman, 1978)</i></p>	<p><i>The study of mental faculties through the use of computational models (Charniak and Mc Dermott, 1985)</i></p> <p><i>The study of the computation that make it possible to perceive, reason and act(Winston, 1992)</i></p>
Agir comme des humains	Agir rationnellement
<p><i>The art of creating machines that perform functions that require intelligence when performed by humans (Kurzweil, 1990)</i></p> <p><i>The study of how to make computers do things at which, at the moment, people are better(Rich and Knight, 1991)</i></p>	<p><i>Computational intelligence is the study of the design of intelligent agents(Poole et al, 1998)</i></p> <p><i>AI... is concerned with intelligence behavior in artifacts(Nilsson, 1998)</i></p>

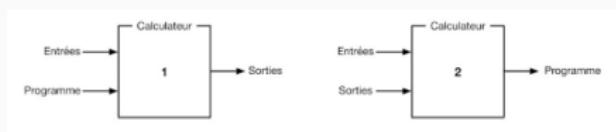
# AI approaches



# Two major opposing approaches<sup>1</sup>

## Two different assumptions

- Human reasoning and knowledge are complex: knowledge **implicit** in data
  - Numerical or data-centric AI - Connectionist approaches - Learning from data
  - Exploitation of **past experience** represented by annotated data, by building calibrated predictive models from them.
- human reasoning can be captured, even if partially incomplete: **explicit** representation of knowledge.
  - **Symbolic or formal knowledge centered IA** - Based on the modeling of logical reasoning, on the **formalisms of representation of knowledge and reasoning**.



<sup>1</sup> D. Cardon et al. La revanche des neurones - <https://hal.archives-ouvertes.fr/hal-02005537/document>

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# **A brief history of Multiagent Systems**

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# A brief history of Multiagent Systems

- Multiagent Systems are a part of Computer Science, at the frontier between: *software engineering*, *distributed computing* and *artificial intelligence*.
- They appeared in the 1990s when the research and engineering questions of these three subfields came to similar solutions to deal with different problems.

## MAS as seen by a **software engineer**

- Software engineering: how to maximise the speed, the reusability, the readability, the security of code, ... ?
- *From Object to Service*: a piece of code, possibly a thread, should be provided with some machine-understandable interface that could allow other code to use it without a priori knowledge of its operational content.
- Software engineers rely on **interaction protocols** to define how the different services should be assembled, how the information has to be exchanged between them, so as to produce some expected result.

## MAS as seen by a **software engineer**

- Multiagent systems: a series of platforms to support the development of such service-oriented software, with coordination entities capable to interpreting protocols.
- for a software engineer's point of view, what matters in multiagent systems is the possibility to **assemble separate components or threads by designing well-formed interaction protocols**.

# MAS as seen by a distributed systems engineer

- The core idea: different processes are located on different physical systems (all connected by some network infrastructure) and interact with each other using some message passing mechanism.
- the agents (or processes) can run on one single computer. What matters is that they have *independent memory spaces and runtimes*.  
~~ cannot rely on synchronous interactions

```
1 def compute(x,y,z):  
2     ... computation here ...  
3     return result  
4  
5 def calling_function():  
6     ... beginning of code ...  
7     res = compute(1,3,'hello')  
8     ... end of code ...
```

## MAS as seen by a distributed systems engineer

On the contrary, DS consider situation in which several problems can occur:

- The called process can not be running, or even not be accessible in the system;
- The called process can not answer, whatever the reason;
- If an answer arrives, there is no guarantee on the delay before it arrives;

All these problems assume a perfectly functioning infrastructure

↔ programming a piece of software by taking into account the fact that the other half of the code is run in a completely asynchronous manner.

## Consequences !

- Agents (processes) send message but don't wait for the answer: they must carry on their activity while a possible answer is computed by another agent;
- Agents must use timeouts to deal with the absence of answers.

## Consequences !

- Agents (processes) send message but don't wait for the answer: they must carry on their activity while a possible answer is computed by another agent;
- Agents must use timeouts to deal with the absence of answers.

~~ For a distributed system engineer, what matters in multiagent systems is **the capacity of agent to deal with asynchronous interactions and possible errors**.

## MAS as seen by an Artificial Intelligence engineer

- Artificial Intelligence is about having machine solve problems (through computation) that human beings solve with their intelligence.
- A subfiled of AI: Planning
  - design of actions and changes
  - solve problems that consist in finding a sequence of actions to go from a given state to another.
- The STanford Research Institute Problem Solver (STRIPS) is one of the first (and most famous) planner.

# MAS as seen by an Artificial Intelligence engineer

```
1 Initial state: At(A), Level(low), BoxAt(C), BananasAt(B)
2 Goal state:      Have(bananas)
3
4 Actions:
5           // move from X to Y
6           _Move(X, Y)_
7           Preconditions: At(X), Level(low)
8           Postconditions: not At(X), At(Y)
9
10          // climb up on the box
11          _ClimbUp(Location)_
12          Preconditions: At(Location), BoxAt(Location), Level(low)
13          Postconditions: Level(high), not Level(low)
14
15          // climb down from the box
16          _ClimbDown(Location)_
17          Preconditions: At(Location), BoxAt(Location), Level(high)
18          Postconditions: Level(low), not Level(high)
19
20          // move monkey and box from X to Y
21          _MoveBox(X, Y)_
22          Preconditions: At(X), BoxAt(X), Level(low)
23          Postconditions: BoxAt(Y), not BoxAt(X), At(Y), not At(X)
24
25          // take the bananas
26          _TakeBananas(Location)_
27          Preconditions: At(Location), BananasAt(Location), Level(high)
28          Postconditions: Have(bananas)
```

A solution: Move(A,C), MoveBox(C,B), ClimbUp(B), TakeBananas(B) (automatically is an NP-hard problem).

# MAS as seen by an Artificial Intelligence engineer

**Collective Intelligence:** two new approaches emerged in the domain:

- **BDI (Belief, Desire, Intention):** a new logic for modeling actions and changes and simulating human-like behavior
    - ~~ programming systems that *simulate* human decision making for action selection based on its perception of the environment
  - **Social Animals Behaviour.** The solution does not arise from the behaviour of one single individual, but it *emerges* from the interactions between the individuals.
    - ~~ programming systems that interact with each other to compute a solution.
- ~~ For an AI engineer, what matters in multi-agent systems is **the capacity of agents to solve problems in a distributed manner, using interactions and automated reasoning.**

## To be remembered

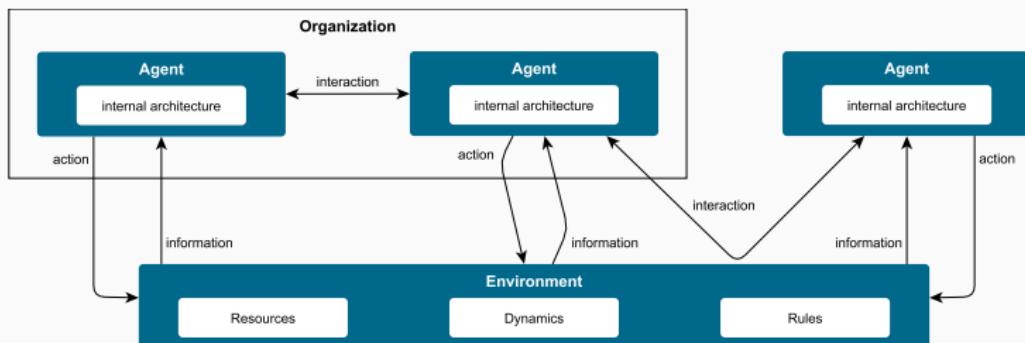
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An agent is :

- a **software process** (or part of a process, or possibly attached to some physical components e.g. in robotics...)
- with some **encapsulated data** (i.e. other agents/processes can't modify directly this data),
- capable of providing a predefined set of **services** (its actions)
- and capable of **reasoning** about its actions and the other agents' actions and to **coordinate** with the other agents so as to participate in a **collective problem solving**.

## To be remembered

- A MAS is a group of agents that share a **common environment** and that act in a **distributed manner** (the agents runtime are supposed to be asynchronous) to solve a problem for a **user**.
- A MAS always has a predefined purpose, and it is important to define the expected outcomes, should it be a solution to a concrete problem (e.g. a path in a TSP problem) or a set of observable properties in a multi-agent based simulation.



# Some Practical Application of MAS

## MAS Applied to Fault Detection

- SurferLab is a joint research lab formed by Bombardier Transport and the university of Valencienne, France.
- MAS is used for diagnosis of trains with the Jade platform.
- The multi-agent systems is capable to detect faulty cars so as to correct them faster<sup>2</sup>.

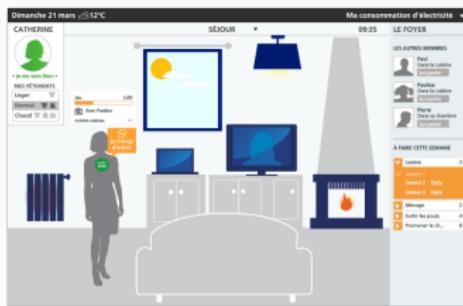


<sup>2</sup>Antoine Le Mortellec, Joffrey Clarhaut, Yves Sallez, Thierry Berger, Damien Trentesaux, Embedded holonic fault diagnosis of complex transportation systems, Engineering Applications of Artificial Intelligence, Volume 26, Issue 1, 2013

# Some Practical Application of MAS

## MAS Applied to the Study of Electrical Consumption

- The R&D branch of EDF has been working on a MAS for the simulation of electrical consumption in households. This applied research project is called SMACH<sup>3</sup>.
  - The MAS platform is used for the generation of realistic load curves in answer to prospective studies: market research & pricing, new production means, new consumption habits, etc.



<sup>3</sup><https://www.youtube.com/watch?v=Hyc7XX4vEjw&t=17s>

# Some Practical Application of MAS

## MAS Applied to Taxi Fleet Management

- The company Magenta Technology has developed in 2008 a MAS-based technology for the management of taxi fleets.
- This software solution is now being used by many taxi companies, such as Green Tomato Cars or Blackberry Cars, two major UK companies in London, UK.



# Some Practical Application of MAS

## MAS for Crowd Simulation in Movies

- **Massive Software** was originally developed for use in Peter Jackson's The Lord Of The Rings film trilogy.
- Massive has become the leading software for crowd related visual effects and autonomous character animation.
- It combines MultiAgent Systems with Graphical Design.



# Some Practical Application of MAS

## MASA



### L'intelligence artificielle au service des décideurs

Fondée en 1996, MASA développe des solutions logicielles innovantes. L'agilité, la réactivité et les hautes compétences techniques de MASA lui ont permis de gagner la confiance de plus de dix-sept armées dans le monde ainsi que de nombreux organismes civils.

S'appuyant sur une technologie d'Intelligence artificielle brevetée, les logiciels développés par MASA permettent la formation et l'entraînement au commandement et à la gestion de crise des décideurs militaires ou civils, la préparation d'exercices complexes avec de très nombreux paramètres, l'analyse après action ainsi que la recherche dans le domaine des équipements ou de la doctrine.

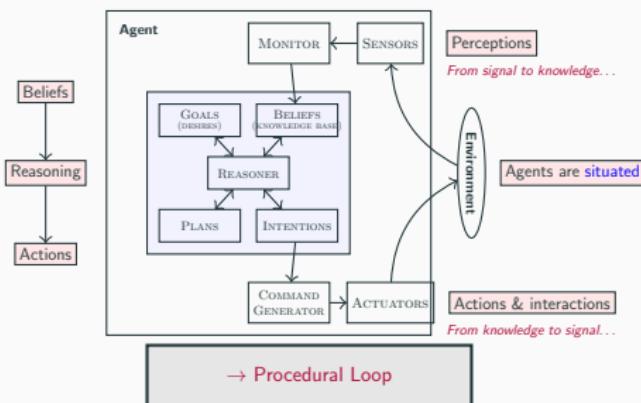
Fournisseur officiel de l'armée française, MASA développe des partenariats de longue durée avec ses clients qu'ils soient étatiques ou privés pour les aider à



## **What is an agent?**

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# PRS architecture [Georgeff and Lansky 87]



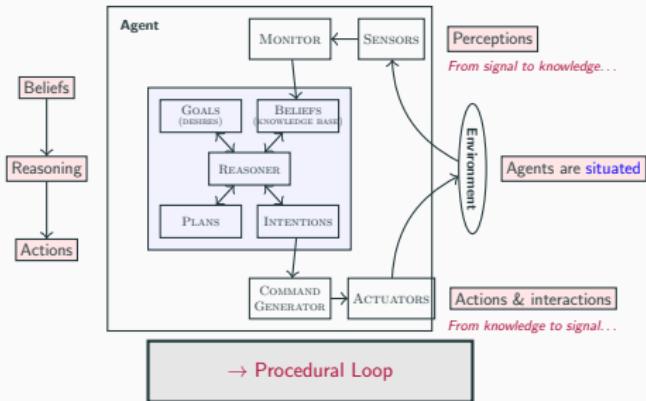
- PRS: Procedural Reasoning System
- The environment: characterized by a (partially) observable state:
  - it can be modified,
  - some of the variable values can be observed by the agents.

~> a set of observation functions and a set of modification functions.
- Agent: **sensors** (to access the observation function ) + **actuators** (to access the modification functions).

# PRS architecture [Georgeff and Lansky 87]

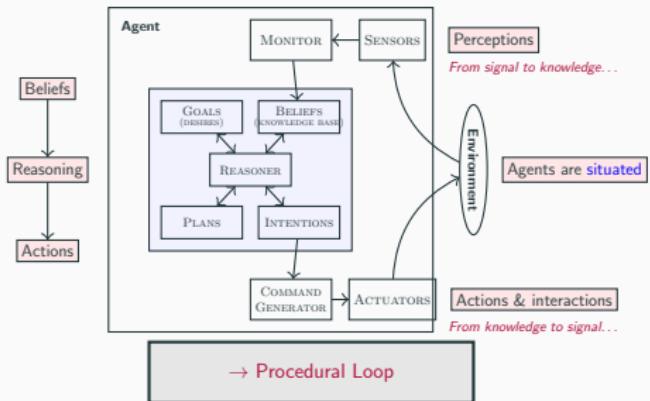
```
1 environment_variable = 0
2
3 def operation_increase_variable():
4     global environment_variable
5     environment_variable += 1
6
7 def perception_get_variable():
8     global environment_variable
9     return environment_variable
10
11 def agent(preferred_value):
12     v = perception_get_variable() # this is a sensor
13     while (v<preferred_value):
14         operation_increase_variable() # this is an actuator
15         v = perception_get_variable()
16
```

# PRS architecture [Georgeff and Lansky 87]



- **Procedural Loop:** an agent runs continuously in three steps:
  1. Perception of the environment
  2. Deliberation = selection of the action (i.e. transforms the result of the perception into internal variables, called *beliefs*. It then combines these with a *set of goals* and a *set of plans* so as to decide for some *intention* that will be turned into a *concrete action* in the environment.)
  3. Action on the environment

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  3. Action on the environment

## To keep in mind: an agent

- Is situated in an environment: it can perceive and act;
- Continuously performs a procedural loop: perception, deliberation, action.

## An agent: remark

---

There is not "the" definition of an agent

[Russel and Norvig,1995]

"An agent is anything that can be viewed as **perceiving** its **environment** through *sensors* and **acting** upon that environment through *effectors*."

## An agent: remark

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"An agent is anything that can be viewed as **perceiving** its **environment** through **sensors** and **acting** upon that environment through **effectors**."

### [Wooldridge & Jennings, 1995]

"An agent is a computer system that is **situated** in some **environment**, and that is capable of **autonomous** action in this environment in order to meet its design objectives"

...

# Agent Models

---

- **Reactive agents:** do not make use of any internal variable: they directly wire all perception to a specific action (or set of actions), hence their name: they perceive and react.

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- **Cognitive agents:**
  - add internal variables and data
  - a first type: internal variables (beliefs) to store its perception  $\rightsquigarrow$  it has a memory of what it perceived, and it can make decisions based on this memory.

## STRIPS example

Move(A,C) , MoveBox(C,B) , ClimbUp(B) , TakeBananas(B)

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  - Note that “explicit planning” is not mandatory. Agent simply checks its current state of beliefs and search for a possible action. The “plan” is then hard-wired in the code. Different levels of “reasoning” about the beliefs: simple deduction, belief revision, diagnosis, ...

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  - Note that “explicit planning” is not mandatory. Agent simply checks its current state of beliefs and search for a possible action. The “plan” is then hard-wired in the code. Different levels of “reasoning” about the beliefs: simple deduction, belief revision, diagnosis, ...
- “explicit planning”: agents that have explicit goals and that perform any sort of computation (from simple plan selection in a predefined list to complex planning algorithms such as SATPLAN)  $\rightsquigarrow$  **Rational agents**.

## To be remembered

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According to the PRS model that is at the core of agent modeling, an *agent*:

- is **situated** in the environment (it can perceive and act);
- might have some internal data called **beliefs** and these are encapsulated within the agent (they are not accessible to other agents or to the environment);
- runs with a **procedural loop** (perceive, decide for an action, act) in an **asynchronous** w.r.t other agents;
- in the decision phase, it can do any sort of reasoning, from simple reaction (**reactive agents**) to more complex reasoning on the beliefs (**cognitive agents**);
- agents that have *explicit goals* and *planning procedures* in the deliberation phase are called **rational agents**.

# Agent and MAS: Exercises

Open the notebook: `IntroductionMAS_2021_Student.ipynb`.

Let us begin with two very simple agents in an environment

```
1  from time import sleep
2
3  class Environment:
4      def act(self,message):
5          print(message)
6      def perceive(self):
7          pass
8
9  class Agent:
10     def __init__(self,name,env):
11         self.name = name
12         self.env = env
13     def procedural_loop(self):
14         while True:
15             self.env.act("Agent "+self.name+" says hello!")
16             sleep(0.1)
17
18 class Runtime:
19     def __init__(self):
20         e = Environment()
21         (Agent("Alice",e)).procedural_loop()
22         (Agent("Bob",e)).procedural_loop()
23
24 Runtime()
```

## Questions

- Why does this not behave as a multiagent system?
- What is the problem?

# Agent and MAS: Exercises

---

## Questions

- Why does this not behave as a multiagent system?
- What is the problem?

## Answers

- Running this code produces a continuous output of “Agent Alice says hello!”.
- It does not behave as a MAS because the agents do not run asynchronously.
- The reason is that the runtime never leaves the procedural loop of agent Alice.

## Questions

To overcome the above limitation, two solutions can be considered.

The first one is to write a **scheduler**, i.e. a piece of code that calls the procedural loops of all agents, one after the other.

- Modify the previous code so that Runtime creates two agents and calls a single-step procedural loop for all agents.

## Answers. Version1: with home made scheduler

```
1
2 from time import sleep
3
4 class Environment:
5     def act(self,message):
6         print(message)
7     def perceive(self):
8         pass
9
10 class Agent:
11     def __init__(self,name,env):
12         self.name = name
13         self.env = env
14     def procedural_loop(self):
15         self.env.act("Agent "+self.name+" says hello!")
16         sleep(0.1)
17
18 class Runtime:
19     def __init__(self):
20         e = Environment()
21         a = Agent("Alice",e)
22         b = Agent("Bob",e)
23         while True:
24             a.procedural_loop()
25             b.procedural_loop()
26
27 Runtime()
28
```

## Answers. Version1: with home made scheduler

1. This MAS verifies a specific property, which is that the procedural loop of all agents is performed at each time step: all agents run at the same "speed".

This is called a **synchronous MAS**. While the agent has its own runtime, interleaved with the one of the other agents (which corresponds to the specification of a MAS) these runtimes are synchronised.

2. The agents procedural loops are always invoked in the same order, which is not a valid hypothesis in a MAS. Agents should **never** use that property.

If you want to avoid this, you can simply modify the Runtime class as follows:

## Answers. Version1: with home made scheduler – Avoiding the same order

```
1  from time import sleep
2
3
4  class Environment:
5      def act(self,message):
6          print(message)
7      def perceive(self):
8          pass
9
10 class Agent:
11     def __init__(self,name,env):
12         self.name = name
13         self.env = env
14     def procedural_loop(self):
15         self.env.act("Agent "+self.name+" says hello!")
16         sleep(0.1)
17
18 class Runtime:
19     def __init__(self):
20         e = Environment()
21         agents = [Agent("Alice", e), Agent("Bob", e)]
22         while True:
23             shuffle(agents)
24             for a in agents:
25                 a.procedural_loop()
26 Runtime()
```

Note that this new version still is a synchronous MAS.

## Answers. Version2: with Threads

```
1 from time import sleep
2 from threading import Thread
3
4 class Environment:
5     def act(self, message):
6         print(message)
7
8     def perceive(self):
9         pass
10
11 class Agent(Thread):
12     def __init__(self, name, env):
13         Thread.__init__(self)
14         self.name = name
15         self.env = env
16
17     def run(self):
18         while True:
19             self.procedural_loop()
20
21     def procedural_loop(self):
22         self.env.act("Agent " + self.name + " says hello!")
23         sleep(0.1)
24
25 class Runtime:
26     def __init__(self):
27         e = Environment()
28         a = Agent("Alice", e)
29         b = Agent("Bob", e)
30         a.start()
31         b.start()
32
33 Runtime()
```

## Questions

Do we have a multiagent system yet?

## Questions

Do we have a multiagent system yet?

## Answers

- The agents in the preceding example are not really situated since the perception phase does nothing.
- They use encapsulated data (the agent's name), but we can't really call these a set of beliefs since they are not connected to the perception.
- Although they have asynchronous runtimes, with a procedural loop, we cannot really call them agents, even not reactive agents. In order to write reactive agents, we want the deliberation phase, in the procedural loop, to depend on the perceptions.

## Questions

Write a new multi-agent system in which the environment has some variable that the agents can perceive. Write two reactive agents: the first one increases the variable by a random value when it is even, the other one when it odd.

# Answers. Two agents and an environment's variable

```
1  from time import sleep
2  from random import *
3  from threading import Thread
4
5
6  class Environment:
7      v = 0
8
9      def increase(self, name):
10         x = randint(1,4)
11         print("Agent " + name + " increase the value by " + str(x))
12         self.v = self.v+x
13         print(" --> " + str(self.v))
14
15     def perceive(self):
16         return self.v
17
18 class Agent(Thread):
19     def __init__(self, name, env):
20         Thread.__init__(self)
21         self.name = name
22         self.env = env
23
24     def run(self):
25         while True:
26             self.procedural_loop()
27
28     def procedural_loop(self):
29         self.b = self.env.perceive()
30         self.act()
31         sleep(uniform(0.1,0.5))
32
```

## Answers. Two agents and an environment's variable (next)

```
1
2
3 class AgentOdd(Agent):
4     def act(self):
5         if self.b%2!=0:
6             self.env.increase(self.name)
7
8 class AgentEven(Agent):
9     def act(self):
10        if self.b % 2 == 0:
11            self.env.increase(self.name)
12
13 class Runtime:
14     def __init__(self):
15         e = Environment()
16         a = AgentOdd("Alice", e)
17         b = AgentEven("Bob", e)
18         a.start()
19         b.start()
20
21 Runtime()
22
```

# Concurrent Modifications

Writing asynchronous MAS with threads requires to consider possible concurrent modifications. In the general case, you must consider a possible interruption of the following sequence of code:

```
1 self.b = self.env.perceive()      # in Agent.procedural_loop()
2 if condition_on_b:              # in AgentX.act()
3     x = randint(1,4)            # in Environment.increase() -- 4
4     lines
5     print("Agent " + name + " increase the value by "+ str(x))
6     self.v = self.v+
7     print(" --> " + str(self.v))
```

# Concurrent Modifications

Writing asynchronous MAS with threads requires to consider possible concurrent modifications. In the general case, you must consider a possible interruption of the following sequence of code:

```
1 self.b = self.env.perceive()      # in Agent.procedural_loop()
2 if condition_on_b:              # in AgentX.act()
3     x = randint(1,4)            # in Environment.increase() -- 4
4     lines
5     print("Agent " + name + " increase the value by "+ str(x))
6     self.v = self.v+
7     print(" --> " + str(self.v))
```

## Questions

- What do you expect to be a problem? Explain why.
- What is a possible solution?

## Concurrent Modifications: Answers

---

- the scheduler interrupts the code between line 1 and 2, or even between line 2 and 3, and passes the “handle” to the other agent:  
~~ the first agent will perform an increase operation on a variable that might not be of the right parity. This is perfectly correct according to a MAS specification!

## Concurrent Modifications: Answers

---

- the scheduler interrupts the code between line 1 and 2, or even between line 2 and 3, and passes the “handle” to the other agent:
  - ~> the first agent will perform an increase operation on a variable that might not be of the right parity. This is perfectly correct according to a MAS specification!
- the scheduler stops in the middle of the Environment.increase() method.
  - ~> This is a problem because the calling agent could perfectly assume this method to be atomic, uninterruptible. This is what you expect when you perform operations in an environment.

Property of a MAS: actions in the environment must be atomic

## Concurrent Modifications: Answers

- the scheduler interrupts the code between line 1 and 2, or even between line 2 and 3, and passes the “handle” to the other agent:  
~~ the first agent will perform an increase operation on a variable that might not be of the right parity. This is perfectly correct according to a MAS specification!
- the scheduler stops in the middle of the Environment.increase() method.  
~~ This is a problem because the calling agent could perfectly assume this method to be atomic, uninterruptible. This is what you expect when you perform operations in an environment.

Property of a MAS: actions in the environment must be atomic

```
1 from threading import Lock
2 l = Lock()
3 with l:
4     ... concurrent code ...
5
```

When writing an asynchronous MAS with threads, the environment operations should be protected by some lock.

## Properties and Definitions

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# Properties of an environment [Russel & Norvig 2003]

## Properties of an environment [Russel & Norvig 2003]

- Fully observable vs. partially observable – can agents obtain complete and correct information about the state of the world?
- Deterministic vs. stochastic – Do actions have guaranteed and uniquely defined effects? Independent episodes (ML algorithms can be used to build Markov Decision Processes representations of the MAS behavior.)?
- Static vs. dynamic – Does the environment change by processes beyond agent control (e.g. ant colony simulation in which pheromones dropped by the agents in the environment need to spread and evaporate with time)?
- Discrete vs. continuous – Is the number of actions and percepts fixed and finite?

# Properties of the Agents in the MAS

## Autonomous agents

The independence of one agent's behavior (i.e. action selection) with respect to some part of the system.

1. **At the agent level**, pro-action is the first degree of autonomy (i.e. independently from any specific signal from the environment).  
e.g. an agent that plays hide and seek and that begins seeking after ending a countdown.
2. **At the interaction level**, autonomy consists in having agents that do not necessarily accept other agent's requests (ignoring messages because they are considered irrelevant; not answering immediately while it could have; agent answers but refuses the request; etc.).
3. **At the MAS level**, *there is no* autonomy: agents are not autonomous w.r.t. the system. On the contrary, they perform what they are designed for. They follow well-defined protocols so that the global task is achieved. Otherwise, there is some misconception in the system.

# Properties of the Agents in the MAS

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## Loosely vs tightly coupled?

1. **In tightly coupled MAS**, the developer has a complete view of the MAS, of possible actions performed by other agents. In other words, it can use some of this information to design action mechanism in a more efficient manner.
2. **In a loosely coupled MAS**, you must make no assumption on the design of other agents in the MAS. This means that you must define very robust behaviours to avoid deadlocks, infinite loops or unnecessary waiting times in communication.

# Properties of the Agents in the MAS

## Open MAS?

- a specific kind of system in which you assume that agents can enter and leave the system at any time during the execution.
- requires programmers to design specific mechanism to ensure that messages are not left unanswered, that some feature in the system will not disappear when not expected, etc.
- Concrete examples of open MAS are ant colonies or prey-predator simulations

# Properties of the Agents in the MAS

## Distributed systems?

1. Conceptual level: it implements distribution principles: encapsulated data, accessed via perception and action methods only, using a *procedural loop* that interleaves the perceptions and actions of agents.  
~~ [Netlogo or Gamma](#)
2. Software level: agents get separated in different threads or processes (depending on the language and implementation). These threads generally run on a single computer.  
~~ [Repast Symphony](#)
3. Physical level: specific code to connect the abstract actions and perceptions to physical operations, message passing, etc. (e.g. a network of smart sensors that exchange information with each other)  
~~ [The Java Agent Development Environment \(Jade\)](#) , [The Robot OS \(ROS\)](#)

# **The mesa platform for Python**

## **27.09.2021 (PM)**

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# The mesa platform for Python

Open Notebook: `money_model_Students.ipynb`

- Mesa is a Python framework for agent-based modeling.
- The running example will be the implementation of a simple Agent-based Computational Economics (ACE) model that simulates distributions of money, income, and wealth in society using statistical physics<sup>4</sup>.

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<sup>4</sup><https://arxiv.org/abs/cond-mat/0211175>