



11<sup>th</sup> European Workshop on Multi-Agent Systems

# Diagnosis of Unwanted Behaviours in Multi-Agent Systems



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# Outline

**Part I – Introduction**

**Part II – Relevant Works**

**Part III – Extended Spectrum-Based Fault Diagnosis  
for MAS**

**Part IV – The Cleaning Agents Example**

**Part V – Conclusion**

## Part I

# Introduction



**Multi-agent Systems (MAS) has a set of features to be deployed in real-world applications.**

**However, agents are under-explored in real deployments** (McKean et al., 2008; Pechoucek and Marík, 2008).

**One reason is the lack of reliability of MAS.**

**To overcome this, a set of works incorporate fault tolerance features and propose better testing techniques.**

- Both rely on the correct *diagnosis* of an unforeseen event.

**A diagnosis process demands for some knowledge about the system.**

- This approach better suit deterministic systems.

**Agents add novel research challenges because of their *non-determinism*.**

**A requirement to deal with agents is treat them as a *black-box component*.**

- Multi-agent diagnosis should not rely on a priori knowledge.

# Aim and Goals

## Main goal of the proposed research is

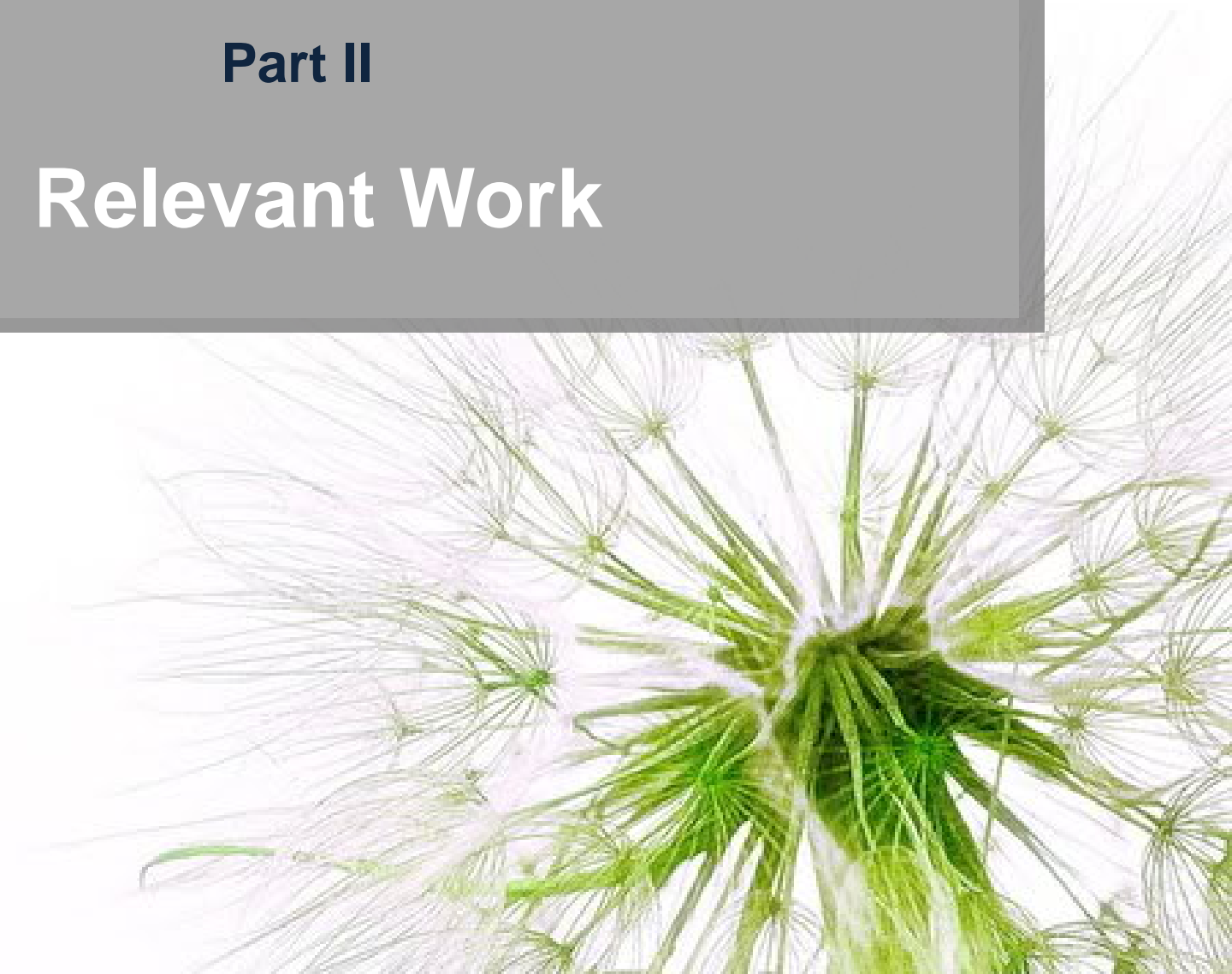
*Diagnose agents with unwanted behaviour that affect the system's performance relying on minimal a priori information.*

## Specific goals of this paper:

- To propose an Extended Spectrum-based Fault Localization approach to Multi-Agent Systems (ESFL-MAS).
- To instantiate the proposal with cleaning agents in a grid-form environment.
- To demonstrate that the ESFL-MAS is able to locate a faulty agent with minimal a priori knowledge.

## Part II

# Relevant Work



***Diagnosis process*** is triggered by a detected deviation from the desired behaviour.

Further, it locates responsible cause(s) for the undesired situation.

**The literature focus on two types of information:**

- *Fault-based*: rely on all known faults in a given domain.
- *Model-based*: depend on a system's model while any abnormality is classified as a fault.

# Lack in the Literature

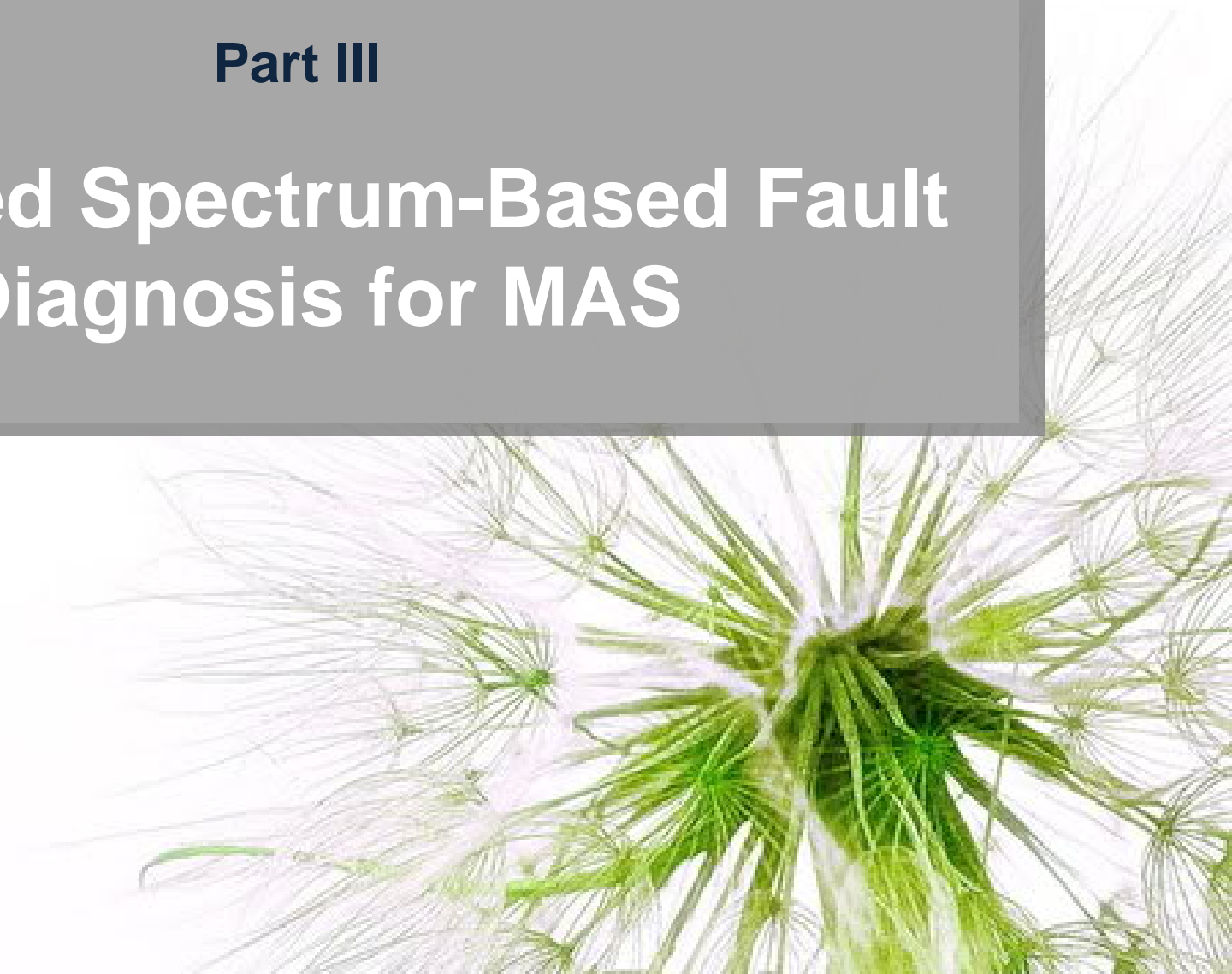
## Summary:

The assumption of a priori (either correct or faulty) knowledge is a constant in agent diagnosis endeavours.

**Our technique aims to *decouple* the diagnosis process from given information.**

## **Part III**

# **Extended Spectrum-Based Fault Diagnosis for MAS**



It is a statistical technique which analyses the software behaviour over multiples runs to find the faulty component.

- *Program spectrum* is a set of runtime profiles that gives a specific view of a software behaviour.
- *Testing results* inform the SFL whether the program behaves correctly (*passed*) or not (*failed*)

Our proposed approach aims to extend some of its characteristics to cover MAS applications.

## ***Test Suite:***

- A multi-agent system must be executed throughout several time steps to observe its behaviour.
- We must execute several rounds of the same test case to cover as many variations as possible to enhance coverage.

## ***Spectra:***

- *Metric-based spectra:* agents are assessed using a certain performance threshold respectively.

## ***Performance Assessment:***

- Measure how effectively the multi-agent system performs.

## Information of a multi-agent system $P$ analysed by the ESFL-MAS.

$$\begin{array}{c} \mathcal{T} \\ \left[ \begin{array}{c} T_1 \\ T_2 \\ \vdots \\ T_m \end{array} \right] \times C \end{array} \quad \begin{array}{c} X [T_\alpha, c_\beta] \\ \left[ \begin{array}{cccc} x_{11} & x_{21} & \cdots & x_{n1} \\ x_{12} & x_{22} & \cdots & x_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ x_{1K} & x_{2K} & \cdots & x_{nK} \end{array} \right] \\ Ag_1 \quad Ag_2 \quad \cdots \quad Ag_n \end{array} \quad \begin{array}{c} E [T_\alpha, c_\beta] \\ \left[ \begin{array}{c} e_1 \\ e_2 \\ \vdots \\ e_K \end{array} \right] \end{array}$$

- $\mathcal{T}$  is a set of test cases;
- Test case  $T_i$  considers a set of environmental variables;
- $C \in \mathbb{N}$  is the number of executions per test case;
- $K$  is the total number of time steps in a given execution;
- $X [T_\alpha, c_\beta]$ ,  $0 \leq \alpha < m$ ,  $0 \leq j < C$  is the spectrum matrix;
- $E$  is a vector of Boolean variables, where

$$e_i = \begin{cases} 0 & \text{if } p_{MAS} \geq thr \\ 1 & \text{if } p_{MAS} < thr \end{cases}$$

# The Algorithm

***Input:*** Multi-agent system, set of test cases, number of executions, and similarity coefficient

***Output:*** Diagnosis report

***begin***

    Get the number of Agents

    Initialize counter

    Run MAS and keep spectra

***for each*** test case and execution

        Get the number of time steps

***for each*** agent

            Feed the respective counter

***end***

***end***

***for each*** agent

        Calculate similarity

***end***

    Sort agents by crescent value of similarity

    Return diagnosis report

***end***

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*begin*

**if**  $x[i, j, k, l] = 1 \wedge e[i, j, k] = 1$  **then**

$a_{11}(l) \leftarrow a_{11}(l) + 1$

**else if**  $x[i, j, k, l] = 0 \wedge e[i, j, k] = 1$  **then**

$a_{01}(l) \leftarrow a_{01}(l) + 1$

**else if**  $x[i, j, k, l] = 1 \wedge e[i, j, k] = 0$  **then**

$a_{10}(l) \leftarrow a_{10}(l) + 1$

**else if**  $x[i, j, k, l] = 0 \wedge e[i, j, k] = 0$  **then**

$a_{00}(l) \leftarrow a_{00}(l) + 1$

**end**

    Calculate similarity

*end*

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***end***

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Calculate similarity

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Sort agents by crescent value of similarity

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**The technique essentially consists in identifying the agent whose column vector resembles the underperformance vector the most.**

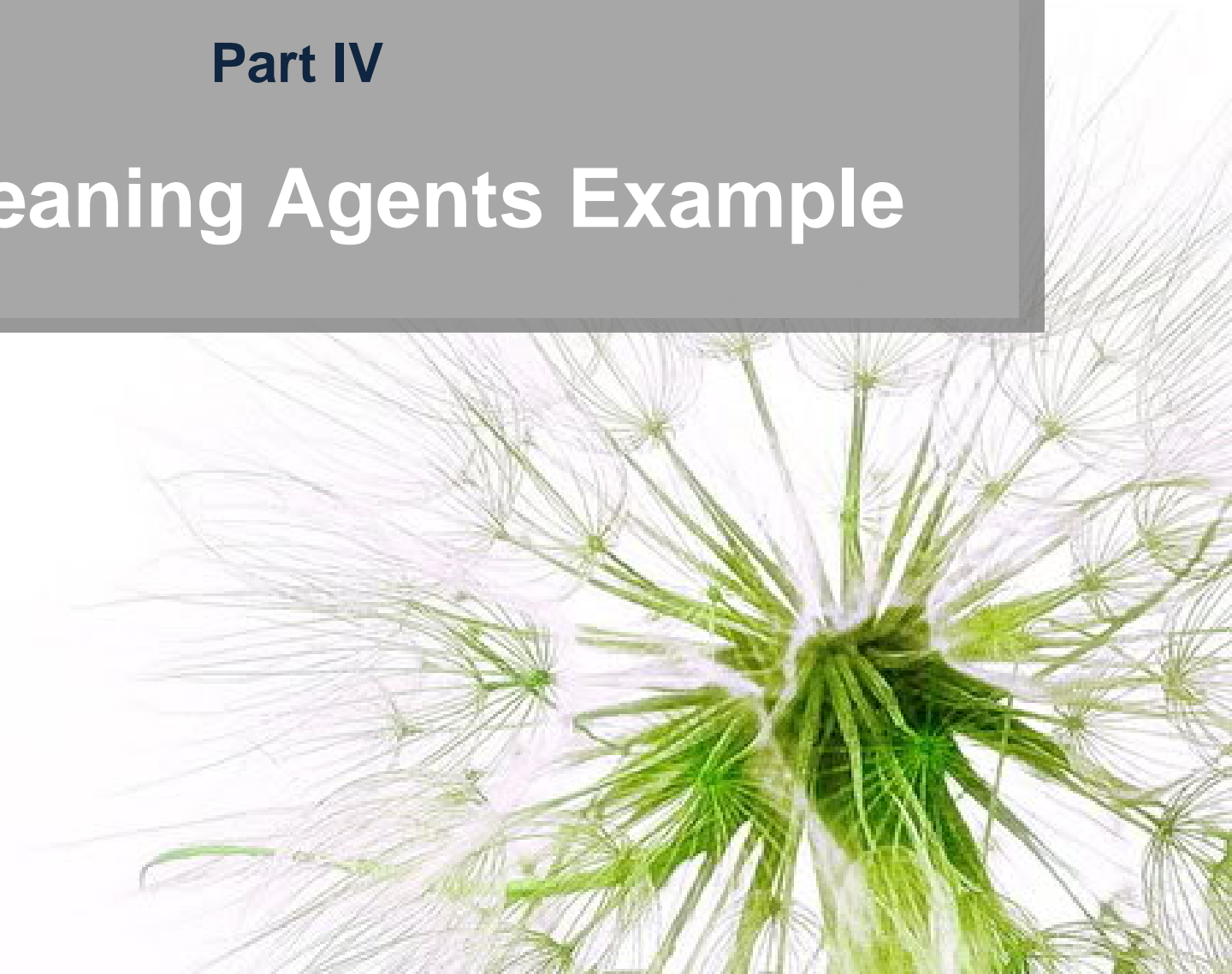
- Similarity coefficient quantifies these resemblances.

## **Must highlight:**

- It not the “silver bullet” for fault diagnosis in MAS.
- Only applicable to close multi-agent systems that work within a static environment.

## Part IV

# The Cleaning Agents Example



# Cleaning Agents Example

## Example characteristics:

- *Close* multi-agent system upon a *static* environment and at *discrete* time.
- 3 cleaner agents, 3 wastes, and 1 recycling station. One agent with a injected fault.
- Allowed actions:  
Move: *UP*, *DOWN*, *LEFT*, and *RIGHT*.  
With waste: *PICK* and *DROP*.
- Agent have a priori knowledge about the recycling station's location.
- The environment is represented as a grid.
- No communication between agents.

# Test Conditions – Part 1

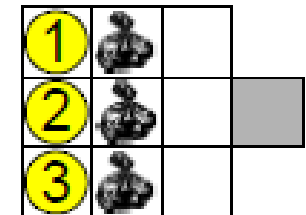
## Agent metric-based spectra:

- an agent must pick up a waste in four time steps
- an agent must drop a waste in four time steps

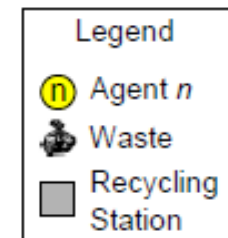
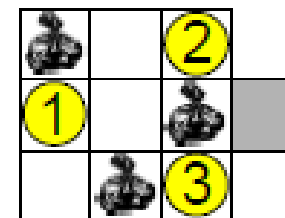
Test Case 1



Test Case 2



Test Case 3



Both are simple and domain-dependent.

We assume that external entities assess each agent

We adopted a simple metric to assess the overall performance:

- The recycling station must receive one waste each 4 time steps.

$$\begin{array}{c} X [2, 2] \\ \left[ \begin{array}{ccc} 0 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{array} \right] \\ Ag_1 \quad Ag_2 \quad Ag_3 \end{array} \quad \begin{array}{c} E [2, 2] \\ \left[ \begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{array} \right] \end{array}$$

# Similarity Coefficient

The *similarity coefficient* indicates the probability of a certain agent to be the faulty one.

For the example we use the Ochiai coefficient, given by:

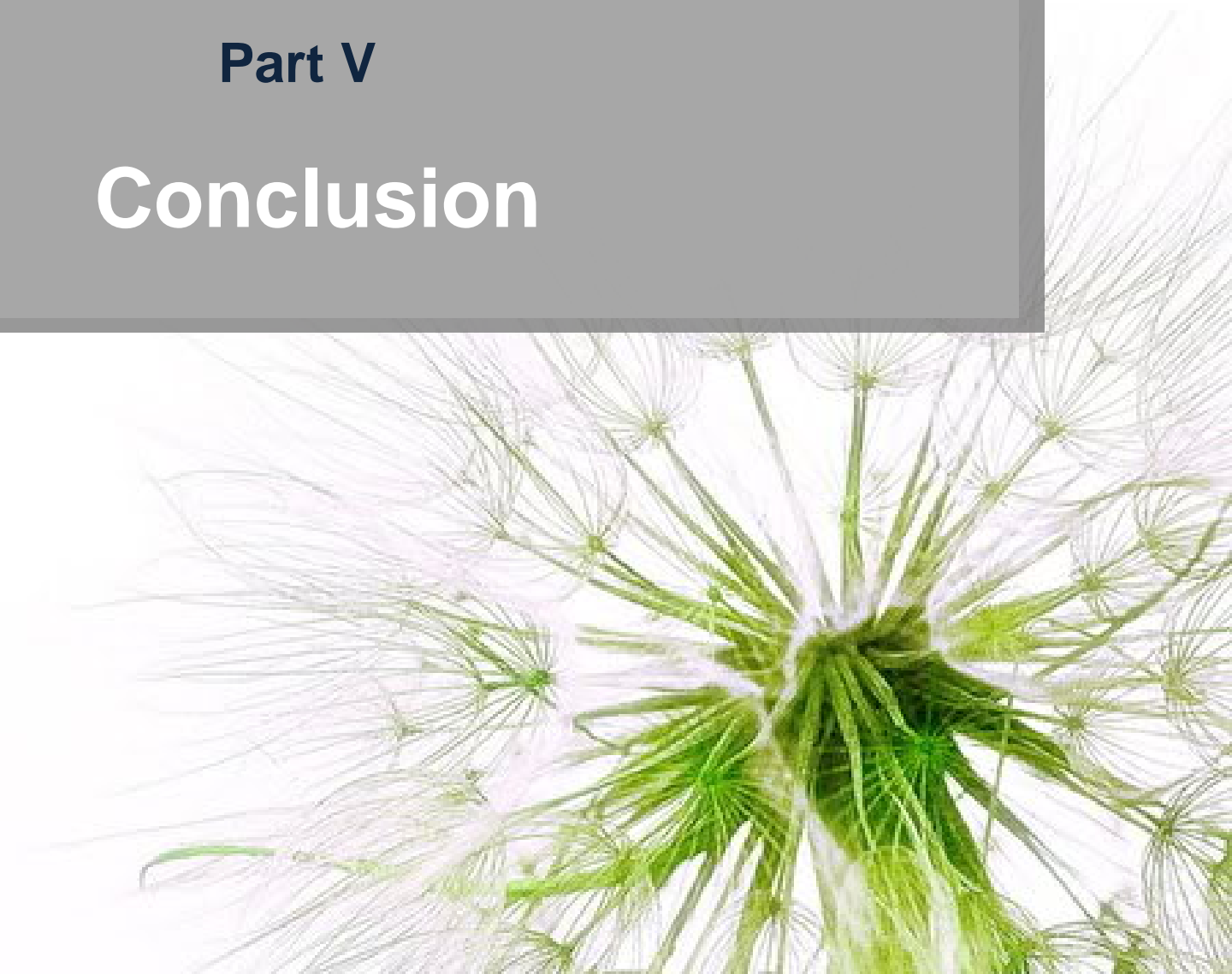
$$s_O = \frac{a_{11}}{\sqrt{(a_{11} + a_{01}) * (a_{11} + a_{10})}}$$

## Obtained Results:

- It was able to diagnose agent 2 as the faulty one.
- The  $D = \{\text{agent 2, agent 1, agent 3}\}$  with values of similarity equal to 0.304, 0.225, and 0.091 respectively.

**Part V**

# Conclusion



**Diagnosing faults in MAS is a very challenging task.**

**We propose a diagnosis technique for MAS relying on minimal given information.**

**ESFL-MAS collects run-time spectra and identify the underlying causes of a failure.**

- However this first appraisal addresses close MAS and static environments

**By the illustrate example, we conclude that ESFL-MAS was able to identify the faulty agent.**

- ESFL-MAS must be validated in more complex scenarios

# Future Research Directions

**Study the influence of similarity coefficients and metrics on the accuracy of the ESFL-MAS.**

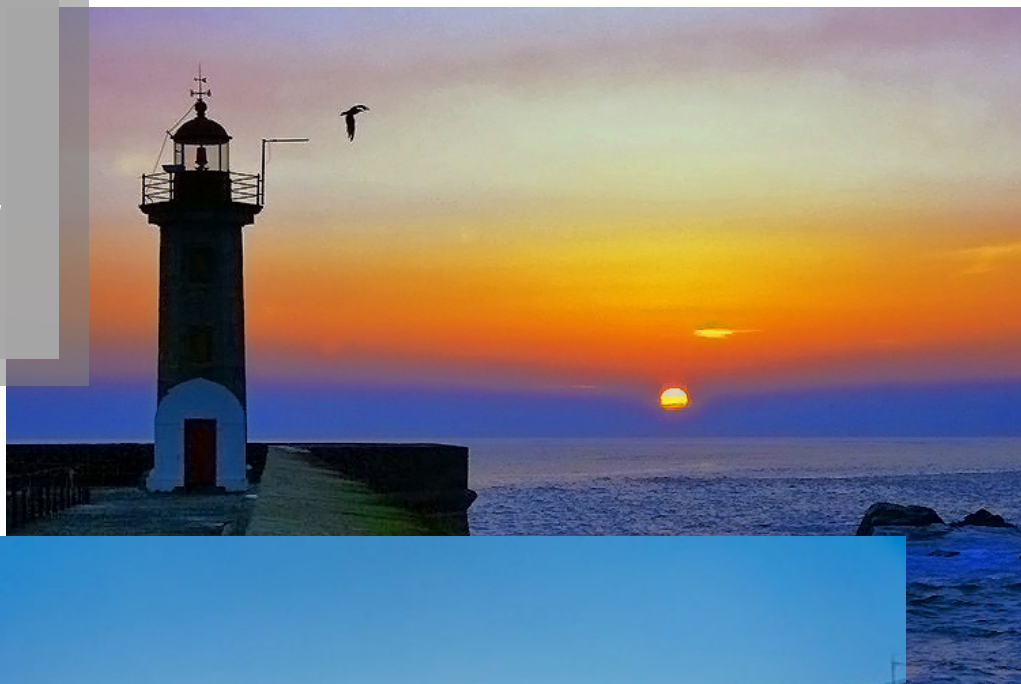
**Domain-independent metric which would be based on individual and global utilities.**

**Take into account the social interactions when diagnosing the potential faulty agents.**

**Investigate the cleaning scenario in a more realistic condition.**

- Larger environment, more agents, different test cases and executions.

# *Thank You!*



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