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## Software Engineering for Autonomous Systems

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# Smart Olive Grove Manager (SOGM)

An Autonomic System for Precision Agriculture

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Github project link: <https://github.com/ocraton/smart-olive-grove>



# 1. Introduction: Autonomic Computing in Agriculture

Precision agriculture requires systems capable of adapting to rapidly changing environmental conditions without constant human intervention. The **Smart Olive Grove Manager (SOGM)** is an autonomic system designed to manage the critical operations of an olive grove: optimized irrigation, pest control, and frost protection.

The system is built upon the **MAPE-K** (Monitor-Analyze-Plan-Execute + Knowledge) reference model. Unlike traditional static automation, SOGM features a **Dynamic, Data-Driven Architecture** where control logic is injected at runtime via a centralized Configuration Service. This ensures high flexibility, decoupling, and resilience, adhering to the principles of Self-Adaptive Software Systems.

## 2. Adaptation Goals and Priorities

The system must fulfill conflicting objectives (e.g., treating pests vs. avoiding chemical drift due to wind). To manage this, we defined a strict hierarchy of goals based on **Numerical Priority**:

1. **Safety (Storm Protection)** - *Priority: 10 (Critical)*
  - **Goal:** Protect the physical infrastructure and prevent resource waste during extreme weather events.
  - **Constraint:** If wind speed exceeds safety limits, all water/chemical emitters must be shut down immediately.
2. **Crop Survival (Frost Protection)** - *Priority: 8 (High)*
  - **Goal:** Prevent irreversible damage to the trees due to freezing temperatures.
  - **Constraint:** Must react proactively to rapid temperature drops before  $0^{\circ}\text{C}$  is reached.
3. **Crop Health (Pest Control)** - *Priority: 5 (Medium)*
  - **Goal:** Mitigate pest infestations (e.g., Olive Fruit Fly) when population thresholds are exceeded.
  - **Constraint:** Chemical nebulization is subject to wind conditions to prevent pollution, unless the infestation persists for too long (Override Logic).
4. **Resource Optimization (Hydration)** - *Priority: 1 (Low)*
  - **Goal:** Maintain optimal soil humidity for production.
  - **Constraint:** Irrigation is the baseline operation but is interruptible by any higher-priority event.



### 3. System Architecture

The architecture follows the **External Approach** pattern, where the Autonomic Manager is distinct from the Managed Resource. The system is fully containerized using Docker to ensure portability and scalability.

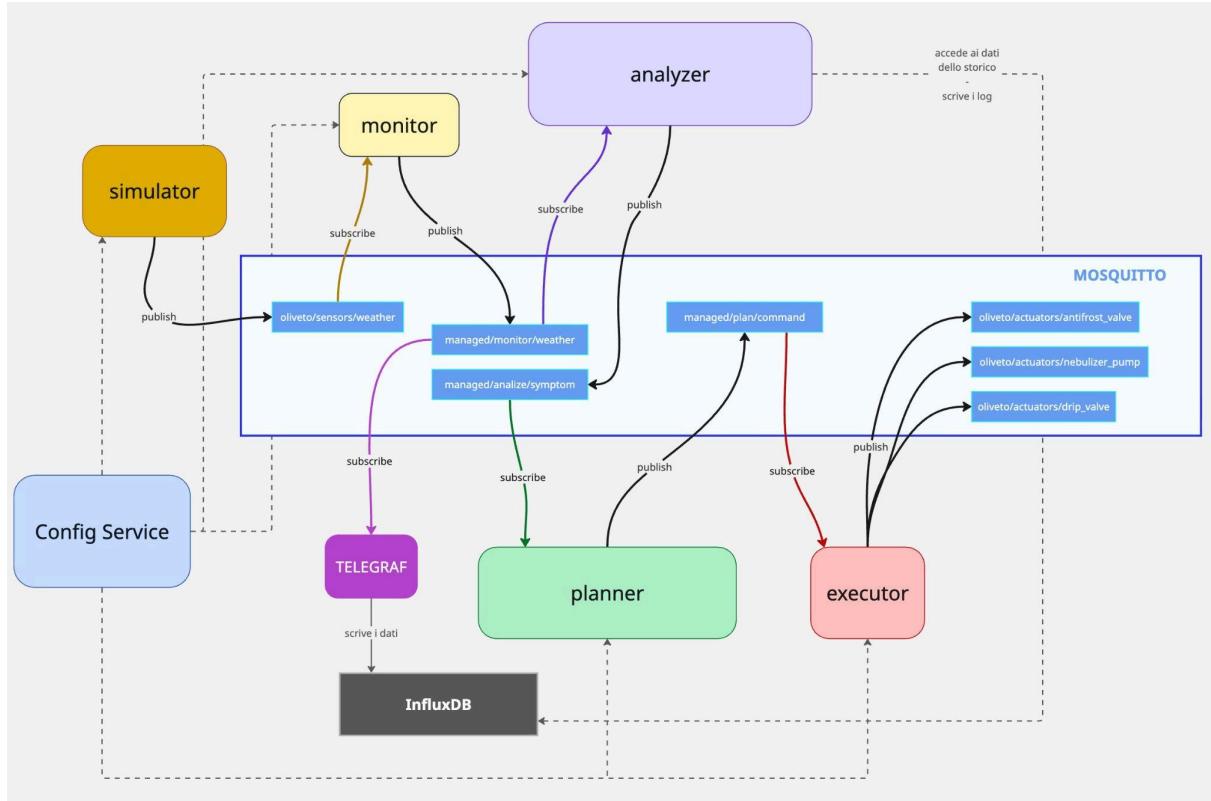


Fig 1: Runtime Architecture of the SOGM system. Solid lines represent reactive MQTT data flows; dashed lines represent initialization via config service and historical knowledge access.

#### 3.1 Component Description

- **Managed Resource (Simulator)**: A Node.js digital twin simulating environmental physics (Soil, Weather, Pests) and actuators. It is agnostic and learns its topology from the Config Service.
- **Config Service (Single Source of Truth)**: An HTTP service that provides system topology and control policies to all microservices at boot.
- **Monitor**: Collects and standardizes sensor data via MQTT.
- **Analyzer**: The intelligence engine. Evaluates logical rules and performs stateful analysis (e.g., checking historical wind data on InfluxDB).
- **Planner**: The decision maker. Resolves conflicts using the Priority Hierarchy (e.g.,  $P_{\{Storm\}} > P_{\{Pest\}}$ ).
- **Executor**: Translates logical plans into physical actuator commands.



- **Knowledge Base:** Composed of **InfluxDB** (Time-series history) and **Mosquitto** (Current state bus).

## 4. Self-Adaptation Taxonomy

Following the taxonomy for self-adaptive systems, SOGM is classified as follows:

- **Time:**
  - *Reactive*: Immediate response to humidity thresholds or storm winds.
  - *Proactive*: Predictive activation of anti-frost systems based on temperature drop rates ( $dT/dt$ ).
- **Reason:**
  - *Change in Context*: Adaptation driven by environmental variables (Weather, Pests).
- **Technique:**
  - *Parameter Adjustment*: Modifying actuator states (ON/OFF).
  - *Structure*: The system supports dynamic addition of sensors via the Config Service.
- **Control Approach:**
  - *Centralized with Heterarchical Logic*: A single Autonomic Manager executes multiple parallel control loops defined in the configuration.

## 5. Adaptation Logic and Scenarios

The system's intelligence is defined by **Control Loops**. Below is the formalization of the logic used in the Analyzer and Planner.

### 5.1 Scenario A: Hydration Maintenance (Reactive)

**Goal:** Maintain soil humidity ( $H$ ) above a minimum threshold ( $H_{\{min\}}$ ).

**Logic Formulation:**

$$Action_{irrigate}(t) = \begin{cases} ON & \text{if } H(t) < H_{min} \\ OFF & \text{if } H(t) \geq H_{target} \end{cases}$$

**Implementation:** A simple threshold check. In our tests,  $H_{\{min\}} = 30\%$ . The system employs a hysteresis cycle with  $H_{\{target\}} = 35\%$  to prevent actuator flickering.



## 5.2 Scenario B: Smart Pest Control (Conflict Resolution)

**Goal:** Treat pests ( $P$ ) only if safe.

**Variables:**  $P_{\{count\}}$  (Trap count),  $W_{\{speed\}}$  (Wind speed),  $T_{\{delay\}}$  (Time elapsed since first attempt).

**Logic Formulation:**

The activation function  $f_{nebulizer}$  depends on a composite condition:

$$C_{infestation} = P_{count} > Threshold_{pest}$$

$$C_{safety} = W_{speed} < Threshold_{wind-safe}$$

$$C_{override} = T_{delay} > 30 \text{ min}$$

$$Action_{nebulizer} = C_{infestation} \wedge (C_{safety} \vee C_{override})$$

**Behavior:** If  $C_{infestation}$  is true but  $C_{safety}$  is false (high wind), the system queries InfluxDB. If the high wind persists longer than  $T_{delay}$ ,  $C_{override}$  becomes true, forcing the activation.

## 5.3 Scenario C: Frost Protection (Predictive)

**Goal:** Prevent freezing using proactive trend analysis.

**Logic Formulation:**

We define the temperature drop rate  $\Delta T_{rate}$  over a time window  $\Delta t$ :

$$\Delta T_{rate} = \frac{T(t - \Delta t) - T(t)}{\Delta t}$$

The protection triggers if the temperature is critical OR if it is dropping too fast:

$$Action_{antifrost} = (T(t) \leq 0) \vee (\Delta T_{rate} > 2.0^{\circ}\text{C}/h)$$

This predictive capability allows the system to act *before* the critical  $0^{\circ}\text{C}$  threshold is reached.



## 6. Visualization and Validation

The system's status is monitored via a **Grafana Dashboard**, connected to the InfluxDB Knowledge Base. The following sections demonstrate the validation of the three key scenarios, comparing the theoretical logic with the actual runtime behavior captured from the dashboard.

### 6.1 Validation of Scenario A: Hydration Maintenance

**Goal:** Reactive maintenance of soil humidity.

**Evidence:**

The graph below shows the system reaction when soil humidity drops below the threshold.

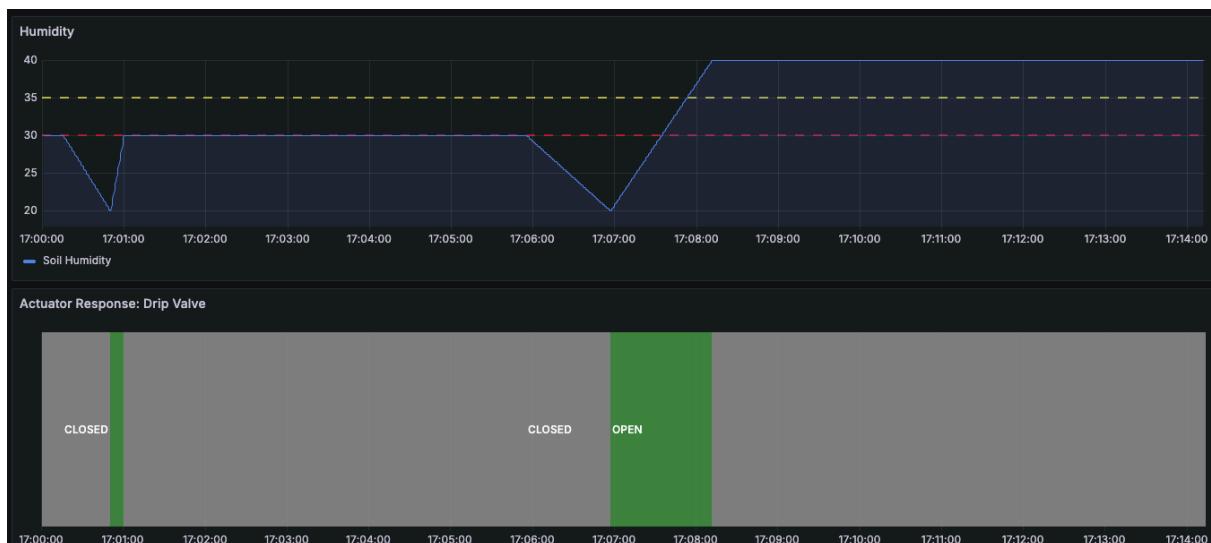


Fig 2: Hydration Loop response. The system detects humidity < 30% and activates the valve.

**Logic Applied:**

The activation follows the simple reactive rule defined in the configuration:

$$Action_{irrigate}(t) = \begin{cases} ON & \text{if } H(t) < 30\% \\ OFF & \text{if } H(t) \geq 35\% \end{cases}$$



## 6.2 Validation of Scenario B: Conflict Resolution & Temporal Override

**Goal:** Manage conflicting objectives (Crop Health vs. Environmental Safety) using a stateful priority system.

### Logic Applied:

The Planner resolves the conflict using a composite condition that includes a time-based override:

$$Action_{nebulizer} = (P_{count} > 50) \wedge ((W_{speed} < 15) \vee (T_{delay} > 30\text{min}))$$

With  $T_{\{max\}} = 30\text{min}$ . This logic ensures that safety is prioritized initially, but critical crop preservation takes precedence if the adverse condition persists too long.

### Evidence 1: Safety Constraint Enforcement (Wait State)

In the initial phase, a critical pest infestation ( $P_{\{count\}} > 50$ ) coincides with unsafe wind conditions ( $W_{\{speed\}} > 15 \text{ km/h}$ ).

As shown in **Fig 3a**, the **Nebulizer** remains **OFF** (Gray/Empty state). The system detects the conflict and enters a "Waiting State" to prevent chemical drift.



*Fig 3a: Safety Lock. The system postpones the treatment due to high wind speeds, prioritizing safety over immediate pest control.*

### Evidence 2: Critical Override Activation (Stateful Behavior)



**Fig 3b** captures the system behavior when the unsafe wind condition persists beyond the defined threshold ( $T_{\{delay\}} > T_{\{max\}}$ ) with  $T_{\{max\}} = 30 \text{ min}$ .

Although the wind speed (Blue Line) remains in the critical zone ( $> 15 \text{ km/h}$ ), the **Nebulizer** switches to **ON** (Green State). The Analyzer queried the **InfluxDB Knowledge Base**, calculated the delay duration, and triggered the override to prevent irreversible crop damage.



*Fig 3b: Temporal Override. The system forces the activation despite the unsafe wind constraint (visible particularly in the initial segment of the graph, where Wind > 15 km/h and Nebulizer is ON), as the infestation condition persisted longer than the allowed safety window (`EXT_INFLUX_DELAY`).*

### 6.3 Validation of Scenario C: Frost Protection (Predictive)

**Goal:** Proactive infrastructure protection.



## Evidence:

The dashboard captures a rapid drop in temperature. The system calculates the drop rate ( $dT/dt$ ) and activates the **Anti-frost Emitter** before the freezing point is reached, demonstrating proactive behavior.



Fig 4: Predictive Frost Protection. Activation occurs based on the negative temperature trend.

## Logic Applied:

The Analyzer evaluates the trend using a sliding window on the time-series database:

$$Action_{antifrost} = (T(t) \leq 0) \vee \left( \frac{\Delta T}{\Delta t} > 2.0^{\circ}\text{C}/h \right)$$

## 7. Future Expansions

To further enhance SOGM, the following expansions are proposed:

1. **AI-Based Pest Detection:** Replace the simulated "Trap Count" with a Computer Vision module that analyzes images from real traps to count insects automatically.
2. **Energy Management:** Introduce a new control loop to manage the energy consumption of the pumps, prioritizing activation during hours of high solar production (if photovoltaic panels are available).
3. **Decentralized Executor:** Move the Executor logic directly to Edge devices (e.g., ESP32) to reduce latency and dependence on the central server.
4. **Automated Unit Testing Strategy:** Developing robust autonomic systems requires verifying logic without relying on full runtime simulations. A future expansion involves introducing a **Unit Testing Framework** (e.g., Jest or Mocha) to test the *Analyzer* logic in isolation.
  - **Mocking Methodology:** We can inject synthetic JSON messages (simulating Monitor data) directly into the Analyzer's evaluation functions.



- **Assertion:** The test asserts that for a specific input (e.g.,  $Wind=20$ ,  $Pests=60$ ,  $Time=0$ ), the Analyzer produces exactly the expected plan (e.g.,  $Action=NO\_OP$ ), ensuring logical correctness and preventing regressions in the adaptation rules.