







## Knowledge Representation and Reasoning for Unmanned Aerial Vehicle Intelligence

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### **i**∃ Outline

- Intelligent UAV Systems: technologies and limitations
- Semantic Web Technologies for UAVs
  - Non-standard inference services
- Case studies
  - Context-aware UAV Systems
  - On-board Intelligent Hazard Detection
- Conclusion and future work





## UAV Systems

#### Unmanned Aerial Vehicles (UAV), a.k.a. "drones"

- Ground Control Station (GCS) for remote control of flight and operational parameters, mission planning and sensors monitoring and management
- Autopilot: on-board electronic subsystem autonomously driving the UAV according to received commands from GCS

#### **Traditional applications**

- Surveillance
- Search and rescue
- Precision farming

#### New applications in logistics and smart cities

- Increasing miniaturization and integration of micro-controllers, programmable processing units and sensors
- UAV Internet of Things Cloud convergence







## Intelligent UAV Systems

Data analysis and decision traditionally implemented in GCS

- Continuous communication between GCS and the UAV → high energy consumption
- Communication latency unsuitable for real-time applications

#### Embedding Artificial Intelligence (AI) into UAV systems

- Enhance perceptive capabilities with autonomous decision-making
- UAV swarms as context-aware self-coordinating teams

#### AI and Machine Learning (ML) algorithms available locally to UAVs







## Semantic Web Technologies for UAVs

#### Semantic Web of Everything vision



- Annotate data, objects, phenomenons and events with metadata to make their semantics machine understandable
- Adoption of standard Semantic Web languages for broad interoperability
- Reasoning tasks on pervasive & embedded devices to infer additional knowledge
- Non-standard inference services for on-the-fly query processing
- Logic-based explainability of results
- Tight computational resource and energy constraints

#### Web Ontology Language (OWL)

- Modelling complex and structured knowledge
- OWL Knowledge Base (KB) composed by:
  - Terminological Box (TBox, a.k.a. ontology): classes and relationships in the knowledge domain
  - Assertion Box (ABox): assertions concerning individuals of a particular problem within the domain



**Tiny-ME embedded reasoning engine** [Ruta *et al.*, JWS, 73, 2022]

### **OO UAV Context Awareness**

UAV systems can be made **situation-aware** and **self-adapting**, monitoring

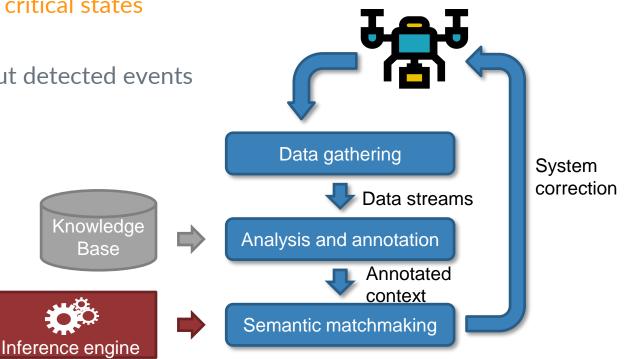
- Their internal state (kinematics, processing resource usage, previous results, ...)
- Sensory and environment contextual features (wind speed/direction, lighting, camera settings, ...)

#### On-board reasoning enables real-time autonomous context management

- Determine the proximity of the current situation to critical states
- Dynamically adapt to avoid the critical state
- Improve operational efficiency and confidence about detected events

#### **Information modeling**

- General description of internal and external features in terms of an ontology model
- Set of pre-defined individuals describing critical scenarios
- Annotation of the current state as a new individual



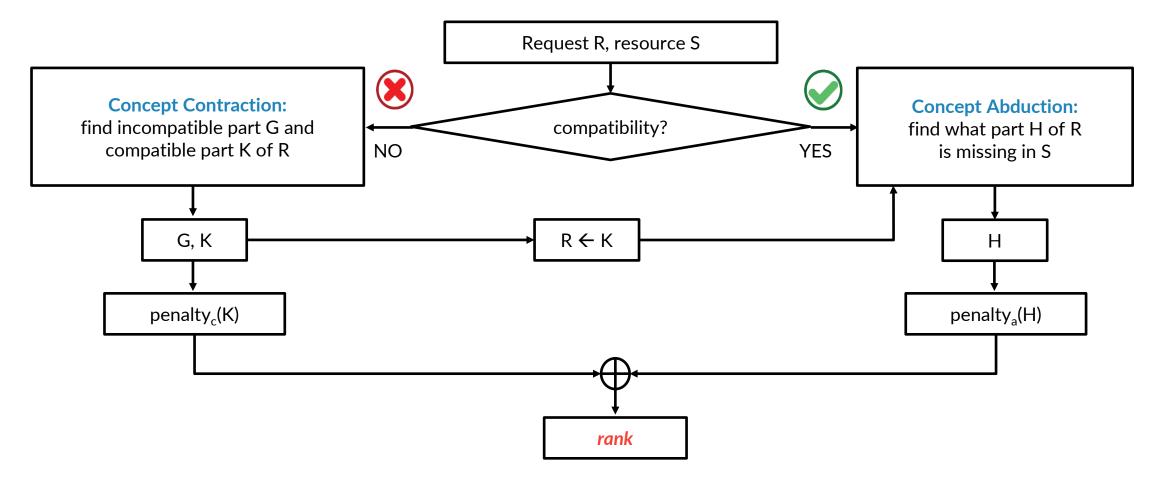


*Bilenchi et al.*, Knowledge Representation and Reasoning for Unmanned Aerial Vehicle Intelligence 7<sup>th</sup> Italian Workshop on Embedded Systems (IWES 2022)



## Semantic Matchmaking

Semantic Matchmaking exploits Concept Contraction and Concept Abduction nonstandard inference services



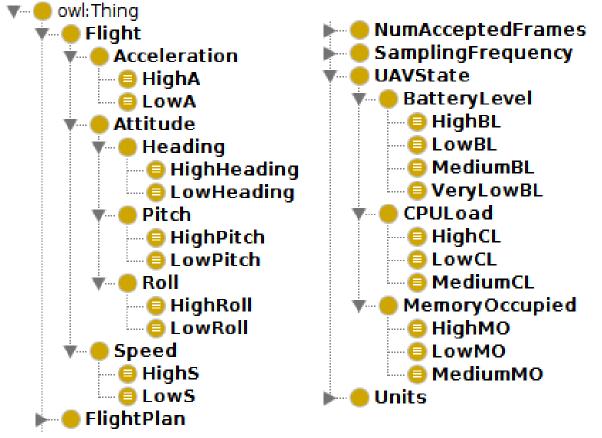
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## O UAV Context Awareness: case study

Ontology for UAV crowd detection with nadiral camera and frame-based image analysis

Classes describe the levels of the parameters managed by the UAV



Current and critical scenarios described with semantic expressions based on the ontology classes

Current:	(HighRoll) a	nd (	(LowPitch) and (LowA)		
	and (LowS) a	nd	(LowHeading) and		
	(MediumBL) a	nd	(HighCL)	and	(HighMO)

The current scenario is compared with every critical scenario in the Knowledge Base, *e.g.*:

```
Crit1: (Roll) and (Pitch) and (Heading) and
(Acceleration) and (BatteryLevel) and
(Speed) and (HighCL) and (HighMO)
```

Semantic Matchmaking detects a critical match threshold due to high CPU load and high occupied memory contextual features

Processing of the next frame is skipped

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## 

Real-time identification of environmental risk levels and handling

#### **Information modeling**

- UAV on-board sensors and actuators
- Detectable substances and their dangerous levels
- Critical atmosphere conditions for the considered substances

#### **Periodic task**

- Collect values from on-board sensors and annotate them
- Exploit the Tiny-ME reasoning engine to build a class expression for the current scenario
- Semantic matchmaking between the newly created individual and risk conditions in the KB

#### **Reference platform**

- 3DR Iris Plus UAV
- Pixhawk 1 embedded flight controller

32-bit STM32F427 Cortex® M4 core, 168 MHz/256 KB RAM/2 MB Flash

Apache NuttX OS

Politecnico di Bari



## **△** On-board Hazard Detection

Detection of flammable and explosive substances according to Directive 2014/34/UE

The current atmospheric condition is expressed through a semantic expression

MediumConcentration Methane and HighOxygenConcentration\_Methane and LowVentilation\_Methane

It is compared with every individual describing critical atmospheric conditions

Explosive\_Methane ≡ HighConcentration\_Methane and HighOxygenConcentration\_Methane and LowVentilation\_Methane Flammable\_Methane ≡ MediumConcentration\_Methane and HighOxygenConcentration\_Methane and LowVentilation\_Methane

Semantic Matchmaking detects *Flammable\_Methane* as the nearest individual

- A flammable atmospheric condition is detected
- An alert related to the methane substance is raised





### Conclusion & future work

- Integration of Knowledge Representation and Reasoning in Unmanned Aerial Vehicles
  - Miniaturization of electronic components allows advanced autonomous applications in UAV systems
  - Semantic Web of Everything approaches and tools can grant better efficiency and explainability w.r.t. conventional Machine Learning techniques
  - Case studies concerning context-awareness and advanced hazard detection

#### • Future work

- Systematic performance evaluation on relevant UAV platforms
- Expanding the scale and scope of applications
- Knowledge-based information fusion within UAV swarms and with vehicular networks and urban infrastructures









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