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► **To cite this version:**

José Manuel Negrete Ramírez, José Manuel Negrete Ramírez, Philippe Roose, Marc Dalmau, Yudith Cardinale, et al.. A DSL-based Approach for Detecting Activities of Daily Living by Means of the AGGIR Variables. *Sensors*, 2021, 21 (16), pp.5674. 10.3390/s21165674 . hal-03346690

**HAL Id: hal-03346690**

**<https://univ-pau.hal.science/hal-03346690>**

Submitted on 16 Sep 2021

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Article

# A DSL-based Approach for Detecting Activities of Daily Living by Means of the AGGIR Variables

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Version July 13, 2021 submitted to Journal Not Specified

**Abstract:** In this paper, we propose a framework to study the AGGIR (Autonomie Gérontologique et Groupe Iso Ressources – Autonomy Gerontology Iso-Resources Groups) grid model, with the aim of assessing the level of independency of elderly people, in accordance with their capabilities of performing daily activities, as well as interacting with their environments. In order to model the Activities of Daily Living (ADL), we extend a previously proposed Domain Specific Language (DSL), by defining new operators to deal with constraints related to time and location of activities and event recognition. The proposed framework aims at providing an analysis tool regarding the performance of elderly/disabled people within a home environment by means of data recovered from sensors using a smart-home simulator environment. We perform an evaluation of our framework in several scenarios, considering five of the AGGIR variables (i.e., feeding, dressing, toileting, elimination, and transfers), as well as health-care devices for tracking occurrence of elderly activities. Results demonstrate the accuracy of the proposed framework to manage the tracked records correctly and thus generate the appropriate event information related to the ADL.

**Keywords:** Domain Specific Language; Feature-oriented Programming; Pervasive Computing; Pervasive Health Systems and Services; AGGIR grid.

## 1. Introduction

According the United Nations Department of Economic and Social Affairs (UN DESA)<sup>1</sup>, population is aging faster than ever before. Due to this increase, reducing medical costs and improving quality of care service [1] have become in recent years a need for new care delivery mechanisms and structures [2]. Personal Sensor Networks (PSN) and Body Sensor Networks (BSN) in smart environments have become a viable alternative to traditional healthcare solutions. PSN are used to detect human daily activities and measure conditions within the environment. BSN are used to monitor vital signs and health conditions by measuring physiological parameters.

Several approaches propose different frameworks focusing on identifying the Activities of Daily Living (ADL) that require monitoring [3–12]. However, there is still a lack of ADL to be analysed, whereas others consider a subset of specific activities; but most importantly, they are not based on a specific tool such as the AGGIR (Autonomie Gérontologique et Groupe Iso Ressources – Autonomy

<sup>1</sup> <https://www.un.org/development/desa/publications/world-population-prospects-the-2017-revision.html>

28 Gerontology Iso-Resources Groups) grid [13]. The AGGIR grid is an autonomy assessment tool used in  
29 France to measure the indenpendncy level of elderly people.

30 In this context, in a previous work [14], we have presented a DSL which allows the representation of  
31 atomic activities composing the AGGIR grid in the form of a plot, providing a history file for detecting  
32 abnormal behavior of the inhabitants of the monitored house. Afterward, we developed a framework in  
33 which the DSL is integrated [15], with the aim to achieve the monitoring of a person within a smart  
34 home environment to identify the ADL performed by the inhabitants, by means of collection and  
35 analysis of data obtained from sensors located in the environment over a certain period of time.

36 In this paper, we extend our previous works by introducing the identification of complex activities  
37 and providing support for simulation and visualization. To do so, the DSL is extended by defining  
38 operators, which deal with constraints related to time, location, and event recognition. Moreover, this  
39 DSL is able to categorise a set of activities based on the constants of the AGGIR grid, which classifies  
40 autonomy levels to various environmental factors affecting the activities and social life of a person.  
41 Thus, more AGGIR activities can be detected, such as alimentation, dressing, toileting, elimination,  
42 and transfers. The framework is improved by adding support for more type of sensors and the capacity  
43 of performing configuration of the environments and visual simulations of the behaviour of people  
44 in such environments. Thus, the framework provides an analysis tool regarding the performance of  
45 elderly/disabled people within a home environment by means of data recovered from sensors using  
46 the iCASA simulator. In turn, the framework provides a general approach to detect ADL, relate  
47 them with the AGGIR variables, and determine the independence of elder people in their homes. To  
48 evaluate our approach, we pick five of the AGGIR variables (i.e., feeding, dressing, toileting, elimination,  
49 and transfers) and evaluate their testability in many scenarios, by means of records representing the  
50 occurrence of activities or unexpected behavior of the elderly. Furthermore, a health-care device use  
51 case concerning the employment of medical equipment for tracking blood glucose is presented. Results  
52 demonstrate the accuracy of our framework to manage the obtained records correctly and thus generate  
53 the appropriate event information.

54 The remainder of this paper is organized as follows. The AGGIR grid model as well as the  
55 definition and characteristics regarding different types of DSL are explained in Section 2. Related  
56 studies are presented in Section 3. The main features of the proposed DSL, as well as a brief review  
57 of operators regarding time, location, and event constraints is described in Section 4. Our framework  
58 proposal is explained in Section 5. The details of the experiments and the discussion about obtained  
59 results are given in Section 6. Finally, conclusions and future work are given in Section 7.

## 60 2. The AGGIR grid model and DSL

61 In this section, the AGGIR grid model, a tool introduced in January 2002 to expand senior benefits,  
62 is briefly described. Additionally, a review of the different types of DSL in any context (logic languages,  
63 textual languages, and graphical languages) is presented.

### 64 2.1. AGGIR Model

65 The AGGIR grid is an autonomy assessment tool used in France to measure the indenpendency  
66 level of elderly people. The AGGIR grid model is a six-level dependence scale (GIR1 to GIR6), that  
67 can be defined based on a set of seventeen three-state variables. Each variable can have one of these  
68 values: **A**, for complete dependency; **B**, for partial dependency; and **C**, for complete independency. The  
69 variables are classified into two groups: discriminatory and illustrative variables, as shown in Table 1.

70 Since DSL are recognized as effective tools to increase the productivity and quality of software  
71 development, in this work we propose a DSL in order to express the situations related to the AGGIR  
72 variables that respond to the ADL performed by people with a physical or mental disability, support for  
73 elderly, diseases connected to aging. The DSL allows to express situations related to the maintenance of  
74 people at home [16]. Subsequently, the definition and characteristics regarding several DSL categories  
75 are presented.

**Table 1.** AGGIR variables

<b>Discriminatory Variables</b>	<b>Illustrative variables</b>
Coherence	Management
Location	Cooking
Toileting	Housekeeping
Dressing	Transportation
Self-feeding / Alimentation	Purchases
Elimination	Medical treatment
Transfers	Leisure activities
Indoor movement	
Outdoor movement	
Distant communication	

## 2.2. Domain Specific Languages (DSL)

A DSL is a programming language, which helps developers to define concepts in terminology belonging to a particular domain [17–20]. DSL are conceived as small textual languages which are able to take certain narrow parts of programming and make them “easier to understand and therefore quicker to write, quicker to modify, and less likely to breed bugs”. Their main goal is to improve developers’ productivity as well as reducing the communication gap in software development between programmers and domain experts [21]. Moreover, DSL are considered as increasingly popular software development techniques, which use concepts from the problem domain rather than the solution domain [22].

DSL are separated into two categories based on their design: internal DSL and external DSL [21]. An internal DSL (also called embedded DSL) provides a general-purpose host language to solve domain-specific problems (originally based on the LISP programming language) [23]. Due to the fact that an external DSL is independent from any other language, they need their own infrastructures like parsers, linkers, compilers, or interpreters [24]. External DSL represent 50% of the DSL [25].

Moreover, patterns [17] and design guidelines [26] have been proposed for the development of DSL. These guidelines are classified by: purpose, realization, content, concrete, and abstract syntax; as shown on Table 2. According to the concrete language, purpose, and domain; some guidelines might be contradicting or irrelevant [26], so they should be considered as proposals when designing a DSL [27]. The most common approach is to write requirements using natural language [28]. This can be used for different purposes to describe any kind of requirements and functional specifications for information systems [29].

## 3. State of the Art

Several DSL have been developed for different areas of application, e.g., expert rules, business rules, configuration rules. A systematic mapping study is presented in [25].

The conceptual DSL described in [30] expresses how business logic can be translated by means of automated code execution for the creation of logic-based smart contracts. Moreover, two execution modes for smart contracts are highlighted: lazy execution (initiated by an actor, either manually or scheduled); and eager execution (fully automated transactions).

IoTDSL, a prototype DSL meant to allow end users drive the IoT devices installed in their homes relying on a high-level rule-based language is presented in [31]; in order to achieve definition and manipulation of devices which are deployed in home environments. Users are able to describe and combine event-based semantics as well as structural configurations in a declarative manner; with a high-level representations of devices. The orchestration of events is then analysed to a component in charge of translating high-level rules into a complex event processing facility dedicated to evaluate

**Table 2.** Categories of DSL design guidelines proposed by [Karsai et al. \[26\]](#).

Category	Description	Guidelines
Language purpose	Goal of the DSL	1. Identify language uses early 2. Ask questions 3. Make your language consistent
Language realization	Implementation of the DSL	4. Decide carefully whether to use graphical or textual realization 5. Compose existing languages where possible 6. Reuse existing language definitions 7. Reuse existing type systems
Language content	Elements of the DSL	8. Reflect only the necessary domain concepts 9. Keep it simple 10. Avoid unnecessary generality 11. Limit the number of language elements 12. Avoid conceptual redundancy 13. Avoid inefficient language elements
Concrete syntax	Readable (external) representation of the DSL	14. Adopt existing notations domain experts use 15. Use descriptive notations 16. Make elements distinguishable 17. Use syntactic sugar appropriately 18. Permit comments 19. Provide organizational structures for models 20. Balance compactness and comprehensibility 21. Use the same style everywhere 22. Identify usage conventions
Abstract Syntax	Internal representation of the DSL	23. Align abstract and concrete syntax 24. Prefer layout which does not affect translation from concrete to abstract syntax 25. Enable modularity 26. Introduce interfaces

109 runtime events. Moreover, the proposed prototype relies upon a textual syntax; and simulation code  
110 can be generated to play configurations defined by the user.

111 A framework is proposed in [32] regarding complex event reasoning concerning process, integration,  
112 and provision of reasoning in order to address decision support to supply chain planners whenever  
113 information about disruption events (such as process failure) is incomplete or uncertain, by using logical  
114 and probabilistic reasoning approaches.

115 A DSL called ReqDL for describing requirements and capturing bidirectional traceability data,  
116 concerning system modeling elements is introduced in [28]. Additionally, a generation algorithm for  
117 independent trace models is presented; as well as concrete and abstract syntax in terms of grammar  
118 and metamodel. Also, three types of operators are proposed to describe requirements and capture trace  
119 data: attributes, description, model elements, and trace links. Moreover, a working example is also  
120 provided.

121 A graphical DSL provides domain experts with an intuitive and user-friendly tool to define the  
122 complex event processing domains of interest for which critical situations in real time need to be  
123 detected [33]. Some recent work on this research field are presented.

124 Dolphin, an extensible programming language implemented as a Groovy DSL for autonomous  
125 vehicle networks is described in [34]. It is designed to express an orchestrated execution of tasks defined  
126 for multiple vehicles dynamically available in a network. Additionally, the built-in operators include  
127 support for composing tasks in several forms, i.e., instance according to concurrent, sequential, or  
128 event-based task flow, partially inspired by process calculi approaches. Moreover, the integration of  
129 the aforementioned DSL with an open-source toolchain for autonomous vehicles is described, in which,  
130 users are able to edit and run Dolphin programs through a custom window, embedded in the overall  
131 GUI environment for editing and monitoring. Also, results from field tests using unmanned underwater  
132 vehicles and unmanned aerial vehicles are presented.

133 Authors in [35] introduce a DSL named SimulateIoT, a model-driven development approach aiming  
134 to define, generate code, and deploy the simulation of IoT systems, thus achieving their design as well  
135 as their deployment by means of a domain metamodel, a graphical concrete syntax, and a model to  
136 text transformation.

137 All these works demonstrate the utility of different types of DSL in several domains. With the  
138 purpose to support the identification of ADL performed at home by elderly/disabled people, many  
139 works have proposed the development of DSL. In the following section, we dedicate on the revision of  
140 existing DSL in the context of both description and recognition of ADL. Furthermore, a brief review of  
141 operators regarding time, location, and event constraints is introduced.

### 142 3.1. DSL for the detection of ADL

143 Authors in [36] introduce a smart city oriented infrastructure for collecting and managing data  
144 related to behavior patterns concerning elderly people and their daily activities, both in indoor and  
145 outdoor environments. Validation considering both low-level and high-level use cases was carried out.  
146 Part of the grammar employed regarding user motility and indoor-outdoor localization is presented:  
147 *BODY\_STATE\_START/ BODY\_STATE\_STOP* for indicating the change of a particular body  
148 state of the subject (i.e. still, walking, sleeping, etc.); *POI\_ENTER/POI\_EXIT* for managing the  
149 location type and/or the GPS coordinates. Furthermore, the user-environment interaction is described  
150 by activities such as *FURNITURE\_OPEN/ CLOSED* for sensing contact, motion and vibration  
151 on furniture and tools. Furthermore, *state\_type, location\_type* are defined in order to identify the  
152 performance of activities.

153 Authors in [37] propose a DSL for model-driven development of activity-oriented context-aware  
154 applications, in order to facilitate the development by improving the efficiency of developers. The  
155 concept model of such applications is analyzed, in order to design abstract syntax and concrete syntax.  
156 Implemented tools include the development environment as well as the generation of Java code and  
157 ontology specification. A case study and evaluation concerning a smart meeting room was introduced  
158 to demonstrate the utilities of the proposed approach.

159 An ambient assisted living architecture for the monitoring of elderly people is presented in [38].  
160 The system is able to collect all information coming from the heterogeneous sensors located in the  
161 indoor environment, such as environmental parameters or biomedical information, and to forward it  
162 towards a remote service. The proposed system is able to continuously monitoring the elderly locomotor  
163 activity. Additionally, the system can trigger specific events when dangerous situations occur (e.g., fall  
164 detection). The feasibility of the proposed architecture was demonstrated by validation functional tests  
165 implementing a real supporting the elderly subject during his daily life activities.

166 Authors in [39] presented a study based on a model-driven solution for a top-down approach for  
167 rapid design and prototyping of an ambient assisted living capable of detecting the behaviour of elderly  
168 persons in their home, acquiring data through a sensor tag wristband that sends data to a smartphone  
169 application through bluetooth low energy protocol. The list of the detected low-level activities are:  
170 body state, indoor home monitoring, presence in indoor places, presence in outdoor places, smartphone  
171 usage, usage of home appliances, interaction with transportation and ambient parameters. Moreover, an  
172 example of grammar for low-level activities for the moving status concerning the *start* and *stop* moving  
173 actions .It includes *START\_MOVING* and *STOP\_MOVING*. Also, it is possible to calculate the  
174 duration of body state status along with the *STILL\_TIME* value. Such a duration can be computed  
175 and treated as a measure.

176 A DSL for processing online events targeting binary sensors is presented in [40]. The approach  
177 is limited to handling binary sensors, due to the fact that the proposed operators are defined on the  
178 domain of boolean signals. The aforementioned language provides a primitive notion of state, modeled  
179 as a boolean signal over time, and allows the generation of complex conditions on different states using  
180 signal operators. Additionally, an interpreter for the language was implemented and applied it to  
181 rewrite a set of real ambient assisted living services.

182 A tool-based methodology, aimed to track ADL of elder adults, supporting replicable research is  
183 proposed in [41]; it allows processing sensor data with the purpose of defining a ruled-based monitoring  
184 process regarding the detection of the abovementioned activities. Moreover, it is intended to assist

185 professional caregivers by providing functional awarenesses by means of a graphical tool; by taking  
 186 advantage of user specific information and abstracting both complex and typical situations.

187 An approach with the objective to evaluate a DSL focused on assistive services by professional  
 188 caregivers is presented in [42]. Such an end-user language enables the formulation of assistive services by  
 189 employing a domain-specific terminology. Also, features concerning the representation of circumstances  
 190 where assistance needs to be provisioned are introduced; in addition to identifying the necessary  
 191 interventions to be considered if assistance is required within context-aware systems. The DSL is only  
 192 dedicated to the detection of contexts.

193 The research in [43] addresses a DSL conceived to support users with disabilities throughout the  
 194 performance of a semi/automated cooking process. Furthermore, a graphical user interface (GUI) is  
 195 furnished in order to access a set of instructions located within a cloud-based repository regarding  
 196 meal preparation including microwave cooking directions. Additionally, it relies on a barcode scanner,  
 197 a touchscreen, a set of speakers; as well as a set of environmental sensors. Audio and visual alerts  
 198 regarding safety concerns are also provided.

199 Table 3 summarizes and evaluates the most recent DSL approaches regarding the services they  
 200 provide, the categories of monitored daily activities, sensing types, and geriatric models.

**Table 3.** Comparison among the most recent DSL approaches regarding ADL's

Work	Services	Activities	Sensing	Sensing Type	Geriatric models
[39]	Movement tracking	IADL	Single	BSN	No
[37]	Activity-oriented context-aware applications, energy	ADL	Multi	PSN	No
[38]	Wellness determination, fall detection, movement tracking, energy	IADL	Multi	PSN, BSN	No
[36]	Detection of elderly behaviour patterns, movement tracking	ADL, IADL,	Multi	PSN, BSN	No
[40]	ADL estimation, energy	ADL	Multi	PSN	No
[41]	ADL estimation, energy	ADL	Multi	PSN	No
[42]	ADL estimation, energy	ADL	Multi	PSN	No
[43]	Assisting meal preparation	IADL	Multi	PSN	No

201 All these works consider the correct identification of the activities that require analysis of ADL.  
 202 However, there is still a lack of ADL to be detected, while others only focus on a subset of specific  
 203 activities; but most importantly, they are not based on a specific tool, such as the AGGIR grid.

### 204 3.2. Temporal, Localization, and Events Operators: A Review

205 In this section, we present a review of different operators used to describe time and location  
 206 relations among events. Actions and activities that can be identified in a space are called events. These  
 207 events can be tagged with time and location, that in turns make events be related according to the  
 208 moment and location in which they occur.

209 According to study presented in [44], in order for a system to make sensible decisions, it has to be  
 210 aware of where the users are and have been during some period of time. Spatial and Temporal logic  
 211 is a well established area of Artificial Intelligence (AI) [45], which has been applied to represent and  
 212 reason on spatial and temporal features and constraints of context and situations [46]. For instances,  
 213 temporal knowledge on human activities can be specified by means of the temporal operators `ANDlater`  
 214 and `ANDsim` in Event-Condition-Action rules introduced in [47]. Authors in [48] apply Allen's Temporal  
 215 Logic [49] (see Table 4) to describe, constrain, and reason on temporal sequences in dealing temporal  
 216 and spatial knowledge in smart homes as well.

217 The event calculus of Shanahan [50], infers what is true from information that expresses what  
 218 and when something occurs (actions) and what happens after those actions. Shanahan uses the action  
 219 concept instead of event, but it makes no difference between the two. The notion of fluent is used to  
 220 express anything that can change over time, as seen on Table 5.

**Table 4.** Allen's temporal operators

Relation	Symbol	Symbol for inverse	Illustration
X before Y	<	> (after)	
X equals Y	=	= (equals)	
X meets Y	m	mi (met by)	
X overlaps Y	o	oi (overlapped by)	
X during Y	s	di (contains)	
X starts Y	d	si (started by)	
X finishes Y	f	fi (finished by)	

**Table 5.** Shanahan's event calculus formulas

Formula	Meaning
Initiates( $\alpha, \beta, \tau$ )	Fluent $\beta$ starts to hold after action $\alpha$ at time $\tau$
Terminates( $\alpha, \beta, \tau$ )	Fluent $\beta$ ceases to hold after action $\alpha$ at time $\tau$
InitiallyP( $\beta$ ):	Fluent $\beta$ holds from time 0
$\tau_1 \leq \tau_2$	Time point $\tau_1$ is before time point $\tau_2$
Happens( $\alpha, \tau$ )	Action $\alpha$ occurs at time $\tau$
HoldsAt( $\beta, \tau$ )	Fluent $\beta$ holds at time $\tau$
Clipped( $\tau_1, \beta, \tau_2$ )	Fluent $\beta$ is terminated between times $\tau_1$ and $\tau_2$

221 Some others works worth revising are shown on Table 6, where temporal, spatial, and event-related  
 222 operators are proposed in order to provide a significant contribution to a solution of these problems. We  
 223 consider the temporal operators proposed by Allen [49], as well as operators such as "*Before, During,*  
 224 *After, Begin, End, startTime, endTime*". With regard to location operators, we employ "*Inside, Outside,*  
 225 *Joint, nextTo*", among others. We also utilise conventional mathematic operators such as "*greater than,*  
 226 *lower than*", to indicate event conditions based on attributes.

**Table 6.** Spatial, temporal and event operators

	Temporal	Location	Event
Freksa [51]	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Ribarić and Dalbello Bašić [52]	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Randell <i>et al.</i> [53]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Galton [54]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Randell <i>et al.</i> [55]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Kim <i>et al.</i> [56]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Pecora <i>et al.</i> [57]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Bruno <i>et al.</i> [58]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Patkos <i>et al.</i> [59]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Santos <i>et al.</i> [60]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Furze and Bennett [61]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
Angsuchotmetee <i>et al.</i> [62]	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>

## 227 4. Domain Specific Language

228 In a previous work we developed a DSL [14] in order to express situations related to the AGGIR  
 229 variables that respond to the activities performed by people with a physical or mental disability to  
 230 determine their independence at home. Afterward, we developed a framework [15], in which the DSL is  
 231 integrated, intended to supervise an occupant within a smart home environment aiming to recognise  
 232 several ADL carried out by the household residents over a certain period of time. In this paper, we  
 233 extend the DSL by introducing the identification of complex activities. In this context, an atomic event  
 234 is defined as an event that can be detected using one reading from a sensor; whereas a complex event is



235 considered as an activity performed by an inhabitant that is detected by several sensors at the same  
236 time.

237 For carrying out such identification, sensors to identify the ADL are specified (Section 4.1 and  
238 Section 4.2); as well as the orchestration for atomic (Section 4.4.1) and complex events detection  
239 (Section 4.4.2). Moreover, the features that the extended DSL provides are the recovery of data from  
240 health related sensors and home-related sensors defined and described through the DSL in order to  
241 register operations regarding a specific date or a time range starting from a particular date or between  
242 two-time points, in addition to the identification of both atomic and complex events.

#### 243 4.1. Smart Home Sensors

244 In the context of this work, the purpose of Wireless Sensor Networks (WSN) is to detect a user's  
245 vital signs, activities, and surroundings. To achieve this goal, a set of sensors is needed to measure  
246 vital signs from the inhabitant. Moreover, smart home sensors need to be deployed to monitor the  
247 surroundings in a home environment; so it is possible to identify the activities associated with the  
248 AGGIR constants, considering physiological contexts and environmental contexts.

249 Thus, for developing the proposed DSL, it is necessary to acknowledge which sensors must be  
250 employed for this matter. With regard to the identification of ADL by means of several sensors for  
251 providing non intrusive monitoring, a group of sensors that can describe such activities are proposed.  
252 The aforementioned sensors are enlisted in Table 7, where some examples of their use are proposed.  
253 Additionally data attributes and data types for each sensor are indicated.

**Table 7.** Smart Home Sensors

Sensor type	Attributes	Data type	Examples
Electromagnetic sensor	On/Off	Boolean	Cooker/stove; oven; light; switch
Proximity sensor	On/Off	Double	Sink
Capacitive sensor	On/Off	Boolean	Kitchen; counter; chair
Magnetic sensor	Open/Close	Boolean	Refrigerator door; cupboard doors
Presence Sensor	On/Off	Boolean	Room occupancy

254 Moreover, relevant data for each performed task identified through the data gathered by sensors,  
255 are considered in Table 8, in order to obtain information to achieve the identification of the AGGIR  
256 constants.

**Table 8.** Additional data for recognition of activities

Extra data type for each task	Data type
startTime	Date / String
endTime	Date / String
Duration	Integer / Double
Location	String
Day	String / Integer

#### 257 4.2. Health Related Sensors

258 As previously mentioned, health care issues within the aging population represent a problem that  
259 needs to be addressed in the proposed DSL. Regarding this aim, a group of commercial medical sensors  
260 are suggested; such sensors are usually considered as non-intrusive, replaceable, and most of them are  
261 low cost. Medical sensors are mainly used in the medical field for the objective of pervasive healthcare.  
262 Some of the conventional medical sensors for physiology measurement are listed in Table 9. Moreover,  
263 each of the devices has specific requirements in terms of parameters, units and collected data.

**Table 9.** Medical sensors [63]

Device	Parameters	Units	Collected Data	
			Data Type	Range
Pulse and Oxygen in Blood (SPO2)	Pulse: 25-250 SPO2 : 35-100	ppm %	Integer	25-250 ppm
			Integer	35-100 %
Electrocardiogram (ECG)	Heart rate ECG signal	BPM (beats per minute) Volts	Integer	0-200 bpm
			Real	0/5 V
Airflow	Respiratory rate Breathing intensity	PPM (peaks per minute) Volts	Integer	0-60 PPM
			Real	0-3.3 V
Blood Pressure Sensor	Systolic pressure Diastolic pressure Pulse	mm Hg mm Hg ppm	Integer	0-300 mmHg
			Integer	0-300 mmHg
			Integer	30~200 ppm
Glucometer	Glucose Glucose	mmol/L mg/dL	Real	
			Real	
Body Temperature Sensor (BTS)	Body Temperature	Degree Celsius (°C)	Real	0-50 °C
Electromyography (EMG)	Muscle rate Muscle signal	CPM (contractions per minute) Volts	Integer	0-60 cpm
			Real	0-5 V
Spirometer (for breathing measuring)	Volume Air flow	l l/min	Real	0.01~9.99 L
			Integer	50~900 L/min
Galvanic skin response (GSR)	Conductance Resistance Voltage	Siemens Ohms Volts	Real	0-20 Siemens
			Real	10K-100KOhms
			Real	0-5V
Body Position	Body position Acceleration	Human body position G	String	5 positions
			Real	
Snore	Snore rate Snore signal	SPM (Snores per minute) Volts	Integer	0-60 spm
			Real	0-5 V

#### 264 4.3. Main characteristics of the proposed DSL

265 Due to the fact that understanding the process under which the proposed DSL works, it is  
 266 relevant to highlight its most remarkable characteristics: (i) expressing situations related to the AGGIR  
 267 variables that respond to the activities performed by the elderly/handicapped people at home; (ii) the  
 268 representation of attribute-based event conditions; (iii) the representation of spatio-temporal event  
 269 conditions. Those are defined by temporal operators, such as *Before*, *During*, *After*. To this end, the  
 270 representation of the DSL relies on a GUI, in order to describe complex events.

271 **DSL interface.** We propose a graphical DSL rather than a textual language, to make it easy to  
 272 handle, since the user might not be necessarily a person who is acknowledged in the programming field.  
 273 Data recovery from sensors is crucial for the detection of ADL. For this matter, the proposed DSL plays  
 274 an important role by representing the gathering of information related to the sensed environments; as  
 275 well as returning graphical results by means of the aforementioned GUI. Additionally, the information  
 276 present in the results can be saved as textual specifications in order to be reused, and thus facilitating  
 277 to find the analysed data without performing any manipulation once again. This functionality allows its  
 278 reusability by different context-aware applications, thus, not being limited only to the AGGIR variables.

279 **DSL operations.** The textual specifications generated by DSL GUI can be used to describe  
 280 events from an operation perspective. Such as an operation is in charge of measuring one of the following  
 281 options: a value, a maximum or minimum value, an average of a set of values, or graphically displaying  
 282 measured values within a time domain. These operations are exposed as a programming interface:  
 283 (i) the **Value** operation: which returns the value (*s*) of the device; (ii) the **Maximum** operation: which  
 284 returns the maximum value of the value (*s*) of the device; (iii) the **Minimum** operation: that returns the  
 285 minimum of the value (*s*) of the device; (iv) the **Average** operation: which returns the average of the  
 286 value (*s*) of the device; and (v) the **Graphic** operation: that returns an icon with the value (*s*) of the  
 287 device.

288 Due to the fact that events within a smart home environment are sensed during a specific time  
 289 range, multiple possibilities to describe time using the DSL are considered. This includes determining a  
 290 specific date or description of a time range starting from a particular date or between two-time points  
 291 representing the beginning and end of the event. The specified time points can be either a date for  
 292 a day or hours within a day. The **Value** operation can be calculated for the values: (i) on a specific  
 293 date; (ii) between two hours; (iii) between two dates; (iv) from a specific date. The operations **Maximum**,

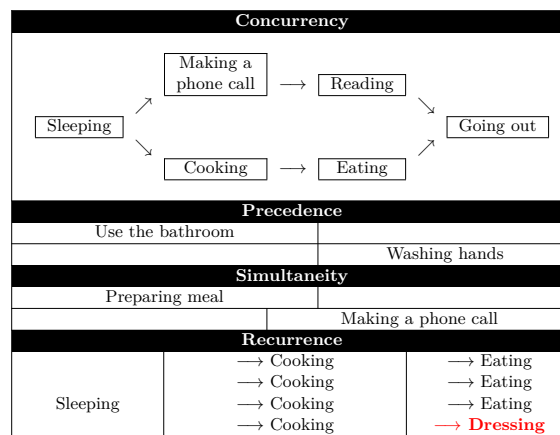
294 **Minimum, Average, Graphic**, are be calculated for the values: (i) between two dates; (ii) from a date;  
 295 and (iii) total (all available values for the device). After selecting the operation as well as the desired  
 296 calculation means, it only remains to specify the date or the time according to what has been previously  
 297 selected.

298 Since in the proposed DSL, the need to represent activities performed by the elderly or handicapped  
 299 people within a home environment is present; it is imperative to be aware when such an activity takes  
 300 place, as well as if there exists repetition or periodicity during the interval of its performance. By  
 301 implementing such a methodology, the detection of anomalies on the behaviour of the inhabitants is  
 302 intended, so coherence of the ADL can be ensured.

303 **Spatio-temporal event conditions.** In order to deal with temporal constraints, spatio-temporal  
 304 event conditions are also represented by the DSL. Those are defined by temporal operators such as  
 305 "*Before, During, After, Begin, End*", and those defined using spatial operators i.e., "*Inside, Outside,*  
 306 *Joint*". The proposed criteria to determine the parameters for the example of the precedent paragraph,  
 307 are: time and space, and within the time dimension. Also, it is relevant to consider:

- 308 i) **Concurrency:** for recognizing activities which take place simultaneously, but they do not necessarily  
 309 require the user's interaction at the same time. That is, activities that have been started but not  
 310 yet ended by the inhabitant [64].  
 311 ii) **Precedence:** for establishing a logical order of the activities, e.g., going to the bathroom and then  
 312 washing hands.  
 313 iii) **Simultaneity:** for identifying which activity takes most of the time from the user, when multitasking  
 314 capabilities might be present, e.g., preparing meal and calling on the phone; or watching television  
 315 while eating.  
 316 iv) **Recurrence:** for determining a logical sequences of situations. In the case where there is a  
 317 recurrence of an activity, it is essential to define at what this activity is carried out regularly.

Some examples of these four latter are presented on Fig. 1.



**Figure 1.** Criteria proposed over the time dimension

318

#### 319 4.4. Representation of the AGGIR variables

320 So far, the main characteristics of the proposed DSL have been introduced. In this section, the  
 321 explanation regarding the orchestration for both atomic and complex event detection is provided.

##### 322 4.4.1. Atomic Event Detection

323 In order to recognize an atomic event, it is required to evaluate a given sensor reading against  
 324 a predefined condition from a simple activity. In the proposed approach, an atomic event is simply  
 325 defined as an **Activity**, and each **Activity** represents one of different tasks composing every AGGIR  
 326 variable. Table 10 shows how each **Activity** is associated to data obtained from sensors located

327 within the smart-home environment concerning three of the AGGIR variables: toileting, dressing, and  
 328 transfers. For the toileting AGGIR variable, three atomic events extracted from sensor readings are  
 329 considered: the bathroom door sensor identifies the **Activity** regarding opening/closing the door; the  
 330 toilet flush sensor indicates the **Activity** of activating the toilet flush; and the washbasin proximity  
 331 sensor describes the **Activity** of washing hands.

**Table 10.** Orchestration of activities

Elimination	Dressing	Transfers
<b>Activity = <math>\sum(+)</math> Sensor data</b>		
Bathroom door sensor ( <i>On-Off</i> / <b>Boolean</b> )	Wardrobe door sensor ( <i>On-Off</i> / <b>Boolean</b> )	Bed capacitive sensor ( <i>On-Off</i> / <b>Boolean</b> )
Toilet flush sensor ( <i>On-Off</i> / <b>Boolean</b> )	Wardrobe proximity sensor ( <i>On-Off</i> / <b>Double</b> )	Chair capacitive sensor ( <i>On-Off</i> / <b>Boolean</b> )
Washbasin proximity sensor ( <i>On-Off</i> / <b>Double</b> )		

#### 332 4.4.2. Complex Event Detection

333 Subsequent to the description of atomic event detection, it is important to emphasize how complex  
 334 event detection takes place in the proposed approach. For this subject, parting from readings from  
 335 sensors defining a simple activity; a complex event can be considered as a sum of multiple events  
 336 connected together representing each one of the AGGIR grid variables. That is, an event that  
 337 summarizes, represents, or denotes a set of atomic events (**Activities**). In the proposed approach, a  
 338 complex event is stated clearly as a **Situation**.

339 In order to illustrate the composition of a **Situation**, Table 11 indicates how three of the AGGIR  
 340 variables are composed: (i) the toileting **Situation** is composed by the **Activities**: *open bathroom*  
 341 *door, use of toilet flush, wash hands*; (ii) the dressing **Situation** is composed by the **Activities**: *open*  
 342 *wardrobe door, spend time changing clothes, close wardrobe door*; and (iii) the Transfers **Situation**  
 343 is composed by the **Activities**: *lying down, sitting down, getting up*. Furthermore, concerning the  
 344 achievement of each one of the aforementioned variables, the activities conforming such variables must  
 345 be detected by following a specific sequence: one activity must be finished before the next one can be  
 346 performed. That is, the precedence criteria must be considered in order for the variable to be completed;  
 347 as pictured on Fig. 1.

**Table 11.** Orchestration of the AGGIR variables

<b>AGGIR Variable = Situation</b>		
Elimination	Dressing	Transfers
<b>Situation = <math>\sum(+)</math> Activities</b>		
Open bathroom door	Open wardrobe door	Lying down
Use of toilet flush	Spend time changing clothes	Sitting down
Wash hands	Close wardrobe door	Getting up
Close bathroom door		

348 After describing the mechanism for orchestrating complex events; next section describes the  
 349 proposed operators to represent complex events in the DSL.

#### 350 4.4.3. Proposed operators

351 *Allen's temporal operators.* In order to deal with constraints regarding the time dimension, it is  
 352 necessary to express conditions regarding temporal relations between sensor states, which are relevant  
 353 for describing activities. To this end, the DSL takes advantage of the criteria over the time dimension  
 354 precised on Section 4.3 and illustrated on Fig. 1: concurrency, precedence, simultaneity, recurrence. For  
 355 example, the event outlining that shower is "on" during presence in the bathroom; detects an activity

356 described between two sensors. For this purpose, a useful tool for representing temporal conditions  
 357 between two time intervals are the Allen temporal relations [49] (Table 4).

358 With the aim to establish finite time intervals, temporal bounds to the Allen's temporal relations  
 359 were applied; e.g. the constraint  $X$  *MEETS*  $[t_1, t_2][0, \infty)$   $Y$  implies that event  $Y$  should be *met by*  
 360 event  $X$ , that the start time of  $X$  must occur between  $t_1$  and  $t_2$ ; and that the end time of  $X$  should  
 361 occur right before the beginning of  $Y$ .

362 Figures 2 to 6 present an example of how temporal conditions can be used to model activity  
 363 recognition in our approach, alongside with location based operators; which are presented in the next  
 364 paragraphs.

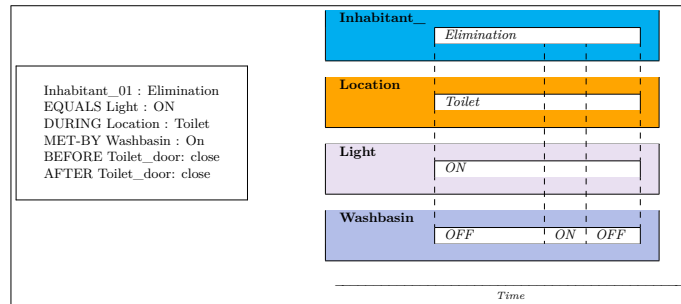


Figure 2. Elimination

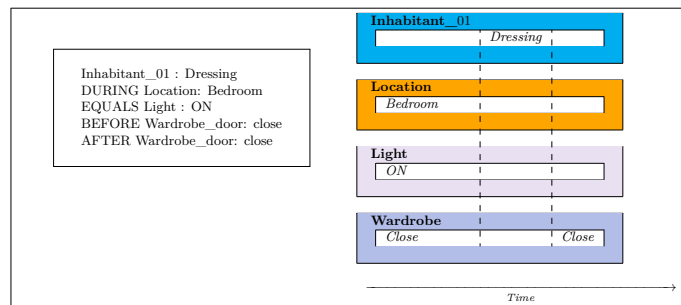


Figure 3. Dressing

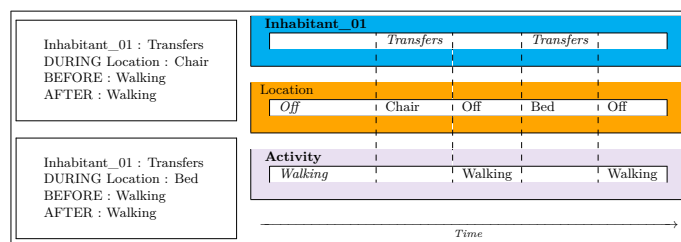


Figure 4. Transfers

365 *Location based operators.* Such operators are in charge of helping determine the position of the  
 366 inhabitant within a home environment; as well as the description of ADL through sensor readings. To  
 367 this effect, the proposed approach considers location based operators: "Inside", "Outside" and "Joint";  
 368 in order to locate in space the ADL performed for the inhabitant. The description regarding each one  
 369 of the proposed location based operators are presented on Table 12.

370 *Event based operators.* Additionally, with regard to determination of planned tasks that should be  
 371 carried out by the inhabitant, such as taking medications; event based operators are important for this  
 372 matter. Those operators are presented on Table 13.

373 Next section is dedicated to the presentation of the framework proposal for the validation and  
 374 experimentation of the DSL.

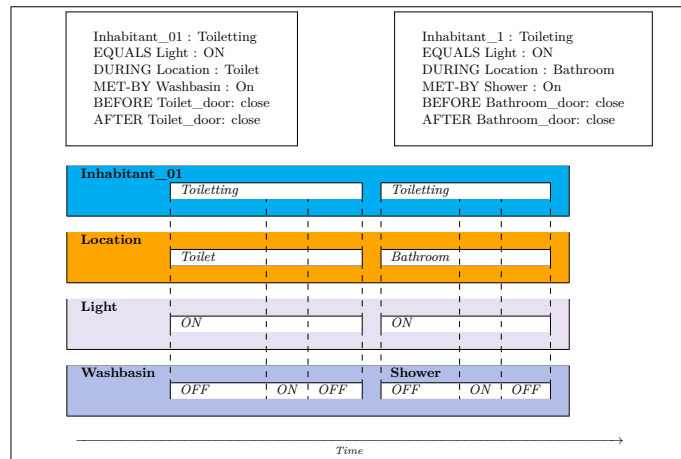


Figure 5. Toileting

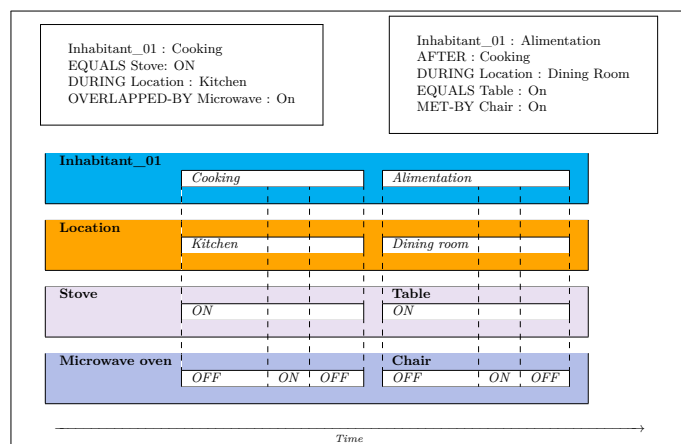


Figure 6. Alimentation

Table 12. Location based operators

Operator	Description	Data type
<i>Inside</i>	Returns true if the user is inside a given area.	String
<i>Outside</i>	Returns true if the user is outside a given area.	String
<i>Joint</i>	Returns true if the user is in two locations at the same time	String

Table 13. Event based operators

Operator	Description	Data type
<i>Planned task (medical measurements)</i>	Returns true if the user performs a planned task	String
<i>Unplanned tasks (medical measurements)</i>	Returns true if the user performs an unplanned task	String

## 375 5. A Framework to Evaluate the AGGIR Variables: Our Approach

376 Subsequently to the proposed DSL, a framework is introduced to fulfill the requirements for  
 377 processing complex events regarding the AGGIR grid model generated by data recovered from sensors  
 378 within a smart home environment; in order to evaluate the level of independency of elderly people,  
 379 according to their capabilities of performing activities and interact with their environments over the  
 380 time.

### 381 5.1. Framework Modules

382 The main purpose of the presented framework is to encourage users who are not necessarily  
 383 acknowledged in the programming field to be able to define events according to the AGGIR grid  
 384 variables; by providing a high-level abstraction, which makes it easy and intuitive by means of concepts  
 385 that are close to the final users.

386 The proposed framework is composed by three main modules, as shown on Fig. 7: (i) the simulator  
 387 module; (ii) the descriptor module; and (iii) the analyzer module. All of them are supported by the  
 388 DSL. A brief description of each one is provided in the next paragraphs.

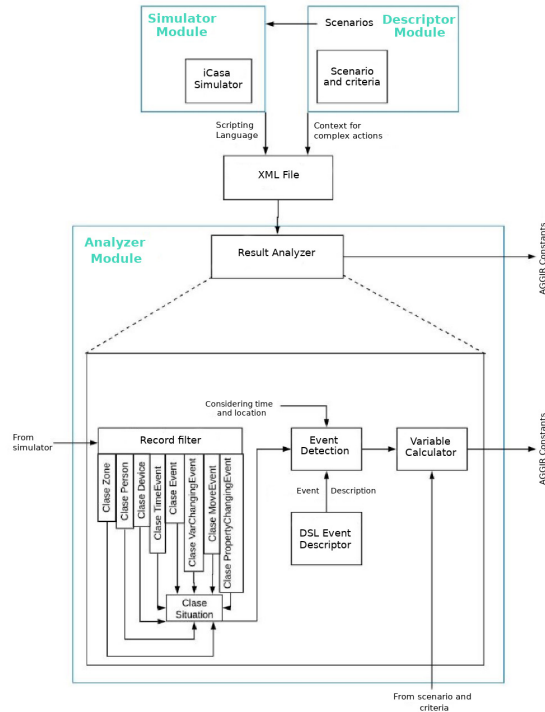


Figure 7. General Architecture of the proposed framework

389 **Simulator module:** In order to describe the elderly’s ADL, smart-home environment scenarios  
 390 need to be parametrized. For this purpose, the activities performed by the subject are carried out and  
 391 information is recovered from sensors located within the smart home environment. As the Simulator  
 392 module, iCASA [65] was integrated to the proposed framework, which is a smart home simulator which  
 393 allows to have control over: time, environment, inhabitants, devices, a GUI, scripting facilities (for the  
 394 environment), and notification facilities [66], as enlisted on Table 14. iCASA is used in order to set up  
 395 a simulated scenario.

Table 14. iCASA facilities

Time	Possibility to slow down, speed up or stop time during the simulation. Simulation of long-term actions such as energy consumption to skip to important actions.
Environment	Definition of different zones in a house. An administration interface to modify different physical properties (temperature, luminosity, etc.) of the different zones, is provided.
Inhabitants	Insertion or removal inhabitants from the environment. Inhabitants, can move from zone to zone, may be carrying physical devices.
Devices	Devices can be simulated or real. At any time, the user can add or remove new simulated devices and modify their localisation in the rooms.
A graphical user interface	The interface displays a map of the house and the localisation of the different devices. It allows to create and configure devices, create and move physical users and watch their actual configurations.
Scripting facilities	Support to the scripts writing to control the environment. Scripts provide a convenient way to test the applications under reproducible conditions.
Notifications facilities	iCASA is event-based and is able to notify subscribers of any modifications in the environment.

396 **Descriptor module:** For the purpose of helping users to interact with the framework, the  
 397 Descriptor module provides a GUI, in order to describe all the criteria over the time dimension, location,  
 398 and events, as well as the required scenarios for the simulation. Also, sensors to identify the ADL have  
 399 to be specified through the GUI. A group of sensors that can describe such activities are introduced on  
 400 Table 7.

401 Moreover, with regard to the scenario specification concerning a certain period of time, the  
 402 descriptor module is in charge of the definition of the parameters with respect to location of both ADL  
 403 and sensors, type of sensors, time, and events that need to be identified: (i) Location Map, which  
 404 is applied to refer to the representation of the environment by means of the house plan where the  
 405 implementation of the sensor network takes place; (ii) Sensor network, that is focused on defining  
 406 information related to the sensor network environment infrastructure; to this end, relevant data should  
 407 be managed, such as: the inventory of sensors located within the smart-home environment, as well as  
 408 the location where each sensor is implemented; (iii) Sensor reading, due to the fact that information  
 409 retrieved from sensors are in a raw format, such data need to be organized for the purpose of easing their  
 410 retrieval and interpretation during the event detection process; (iv) Event condition, since conditions  
 411 are established for triggering the detection of events, such conditions must be defined by the user; and  
 412 (v) Event occurrence, once the event conditions have been provided, the event occurrence must be  
 413 managed in accordance with such aforementioned conditions.

414 **Analyzer module:** In order to recover data from the Simulator module and the Descriptor  
 415 module; the Analyzer module is proposed to carry out such a task. Once all the necessary data is  
 416 collected, the analyzer module analyzes them in order to classify them and evaluate if the AGGIR  
 417 variables of the case study have been carried out to completion. The Analyzer module consists of: (i)  
 418 a record filter; (ii) an event detector; and (iii) a variable calculator. Such components are described  
 419 hereafter.

420 *Record Filter:* Once the simulation has been conducted, in order to manage the obtained data, the  
 421 record filter of the Analyzer module separates the recover data into a series of records. Each record is  
 422 responsible of the description of actions collected by the sensor network located within the smart-home  
 423 environment. Therefore, each register is conformed by data files, i.e., the time when the action takes  
 424 place (corresponding to the simulator clock), the sensor ID, and the sensed value. Furthermore, because  
 425 of the retrieval of such values is generated in raw data format, it is necessary to translate them by means  
 426 of the Analyzer module with the aim of achieving the identification of events, as well as calculation of  
 427 values related to the AGGIR grid variables.

428 With the aim of performing a filtering of all entries, the first task of the analyzer module is to  
 429 classify such inputs in a correct manner. For this matter, the generation of “filters” was modeled in the  
 430 present approach as classes. These latter were introduced in order to determine the type of elements  
 431 that integrate the analyzed set of instructions, by giving each element the appropriate attributes for the  
 432 modeling of raw data. To this extent, the proposed generated classes for data classification are enlisted  
 433 in Table 15. Subsequently, in order to detect the performed activities, the *Event detector* is presented.

**Table 15.** Classes for data classification

Class name	Function
<i>Zone</i>	Modelling the simulator zones; in charge of identifiers, measurements, variables and their values.
<i>Device</i>	Managing several aspects concerning the simulator devices; i.e. identifiers, properties, values spatial location.
<i>Person</i>	Controlling the inhabitants within the smart-home environment; such as the person identifiers and their location with respect to the simulation context.
<i>TimeEvent</i>	Modelling the instructions concerning the time dimension: generating real time data parting from time raw data, allowing time management during the analysis.
<i>Event</i>	Considered as a basic event; modelling situations not requiring additional data.
<i>MoveEvent</i>	Modelling events regarding movement within a specific zone in the smart-home area, either a person or a device. Subclass of the <i>Event</i> class.
<i>VarChangingEvent</i>	Modelling events concerning modifications to designated variables to a specific zone, i.e. temperature change. Subclass of the <i>Event</i> class.
<i>PropertyChangingEvent</i>	Modelling events that generate modifications with regard to intrinsic properties regarding the simulation devices.
<i>Situation</i>	Modelling complex situations. Complex situations are a result of the interaction of several atomic events which allows to identify abnormal behaviour

434 *Event detector:* After all relevant data from PSN and BSN have been gathered, the identification  
 435 of events must be carried out. To this extent, the Event detector relies on the introduced DSL to



436 achieve such identification, by managing information grouped in multiple records which are extracted  
437 from several sensors

438 Aiming to separate each reading detected by several devices, the Event detector retrieves information  
439 in terms of the scripting language. Such events are identified as low-level actions instances; e.g., if a  
440 presence sensor is off in room X, and after a period of time another presence sensor is on in room Y; it  
441 can be inferred that the action "walk" was carried out.

442 Regarding the recognition of malfunctions, the detection of abnormal situations is carried out at  
443 the level of identification of events and situations. For this matter, throughout the analysis of low-level  
444 events (i.e., Situation class instances) collected from device readings from the simulator module script, it  
445 was possible to extract a set of anomalies to be detected by the aforementioned module. Such anomalies  
446 are enlisted hereafter in Table 16.

**Table 16.** Anomalies identified by the event detector module

Sensor/Activity	Detected problem	Issue for the inhabitant
FloodSensor	Leak or break in pipelines	Flood within the home environment
DimmerLight / BinaryLight	Exceeded MAX time ON ON at a wrong time	High energy cost; neglecting of ADL's; location, coherence.
Heater/Cooler	Exceeded MAX time ON is ON when no needed	High cost, neglect; abnormal thermal sensation; location problem Neglecting of household tasks, location, coherence
Not leaving room	Exceeded MAX time ON	Housekeeping, alimentation, elimination, leisure activities.
Light/Presence	Inhabitant stays still	Accident in ZONE, location, coherent behaviour, transfer.
CO/CO <sub>2</sub> sensor	High CO/CO <sub>2</sub> concentration	Household tasks; i.e. impossibility to check on the status on pipelines
Door sensor	Door <i>LEFT OPEN</i>	Location, coherence, household maintaining.
Siren	Resident must leave the house	Maintenance of home.
Toilet sensors: Door toilet flush, washbasin	Irregular urination time.	Toileting, elimination, personal hygiene, transfers.
Wardrobe door	Not changing clothes	Dressing, coherence, toileting
Motion sensor	Wandering around at unusual time	Location, coherence
Temperature/Motion	Abandoning kitchen while cooking	Location, coherence, household tasks.

447 Each one of the aforementioned anomalies is related to the AGGIR grid variables. The detection  
448 of any of these issues may return a negative value for the associated variable, that is to say, the inability  
449 to satisfactorily carry out the AGGIR variable in question.

450 *Variable calculator:* Regarding the achievement of the AGGIR variables, criteria consisting on  
451 accomplishing a determined number of events during a specific time lapse related to the evaluated AGGIR  
452 variable must be fulfilled; e.g., in order to validate if the dressing AGGIR variable is consummated,  
453 AGGIR dressing event conditions such as moving towards the wardrobe at least twice a day must be  
454 met. For this purpose, although every AGGIR variable is based on three major states, which identify  
455 that the elderly inhabitant possesses the ability to perform the ADL conforming a specific variable,  
456 whether it is completely, partially, or not existent at all; this study covers a proposal which only  
457 relies on two cases out of the three abovementioned options; meaning that either the inhabitant is  
458 in complete possession of the skills concerning the performance of the ADL composing the evaluated  
459 AGGIR variable or simply not.

## 460 5.2. Complex Event Detection

461 Provided the main components of the aforementioned framework proposal, the description of the  
462 process for achieving the detection of complex events is then referred.

463 Due to the fact that ADL take place within a domestic surroundings, sensors for detecting such  
464 ADL are placed within a smart-home environment. For this matter, such an environment is defined  
465 and configured through the Simulator module. Then, the detection process is triggered by the readings  
466 obtained from the sensors located within the smart home environment simulator.

467 In order to obtain information to achieve the identification of the AGGIR variables after every  
468 activity has been carried out to completion, some relevant data have to be considered, e.g., *startTime*,  
469 *endTime*; as shown in Table 8. All these data are specified in the Descriptor module. After the  
470 simulation took place, a history log file is created, which considers the logical aspect of situations, i.e.,

471 if the inhabitant is eating and its noon, it might seem logical, but the fact that the inhabitant is eating  
 472 in the toilet is not logical.

473 As a result, the history log file is necessary to deduct the possible activities which will be performed  
 474 in the future, and to find a relation among them to assure a coherent behaviour. Such a file will permit  
 475 to obtain information whether it is from long or short periods of time; making possible the identification  
 476 of complex situations inside the home environment, such as feeding, toileting, transfers; where data  
 477 collected through the timeline of activities is useful to determine if the behaviour of the inhabitant can  
 478 be considered as normal.

## 479 6. Experimental Evaluation: Use Case Description

480 In order to demonstrate the efficacy and suitability of our proposed approach, a set of experiments  
 481 with regard to the detection of the AGGIR grid variables are performed. For the purpose of evaluating  
 482 the introduced DSL, four of the AGGIR variables (i.e., dressing, toileting, transfers, and feeding) are  
 483 picked with the aim of determining their testability in many scenarios, by means of records representing  
 484 the occurrence of ADL performed the elderly inhabitant within a one-week period. To detect either  
 485 the achievement or absence of the AGGIR variables by means of the DSL, we follow a methodological  
 486 process for the experimental simulations, consisting on the following steps:

487 **Step 1.** In order to prepare the scenario for an elderly indoor daily routine over the course of  
 488 one week, the first step consists on the specification of the the criteria over the time and location of  
 489 activities, type and location of sensors, and events that have to be detected; in this specific case, all  
 490 activities related to the dressing, toileting, feeding, and transfer AGGIR variables. For this matter,  
 491 the aforementioned actions are specified and described through the GUI of the Descriptor module. To  
 492 this end, the primary source of information used to generate the scenario is the schedule proposed  
 493 by the work of [9], where the behaviour of an inhabitant living in a genuine household environment  
 494 is simulated based on the daily routine of an elderly resident, as shown in Fig. 8. However, in order  
 495 to make the scenario more suitable for the simulation, many modifications take place. Furthermore,  
 496 for the purpose of making the scenario more suitable for the simulation, the XML format is used to  
 497 describe the scenario programmatically, given that such files allow to define events chronologically; thus  
 498 facilitating the processing carried out by the Simulator module. A brief example of the above-mentioned  
 499 files provided in Fig. 9.

	SUNDAY	MONDAY	TUESDAY	WEDNESDAY	THURSDAY	FRIDAY	SATURDAY
0 H - 6 H	23:00 H - 7H Sleeping	23:00 H - 6:30 H Sleeping	22:30 H - 7H Sleeping	23:00 H - 7:00 H Sleeping	23:00 H - 7:00 H Sleeping	22:30 H - 6:30 H Sleeping	22:30 H - 7:00 H Sleeping
6 H		6:30 H - 7:30 H				6:30 H - 7:30 H	
7 H	7-8 H Toileting	Toileting 7 H - Bathing 7:30 H - 8:30 H	7-8 H Toileting 7:30 H - Bathing	7-8 H Toileting 7:30 H - Bathing	7-8 H Toileting 7:30 H - Bathing	Toileting 7 H - Bathing 7:30 H Eating	7-8 H Toileting 7:30 H - Bathing
8 H	8-9 H Eating	8 H - Cooking 8 H - Eating 8:30 - 10:30	8 H Eating 8:30 H - 10:30 H	8 H Cooking 8:30 H Eating	8 H Cooking 8:30 H Eating	8 H - 10 H Watching TV	8 H - 10 H 8:30 H Eating
9 H	9-10:30 Going Out	Resting	Watching TV	9 H - 11 H Resting	9 H - 11 H Resting		9 H - 10:30 H Going Out
10 H	10:30 H-Toileting	10:30 H-Toileting	10:30 H - 12:00 H			10 H - 11:30 H Going Out	10:30 H Toileting

Figure 8. Schedule employed in the simulation

500 **Step 2.** Once the information provided by the Descriptor module is established; the Simulator  
 501 module, i.e., the iCASA framework, executes the simulation according to such precised information.

502 **Step 3.** After the simulation is performed, the record filter component of the Analyzer module  
 503 organizes the resulting data into a set of records, each of which represents an action captured by a sensor  
 504 within the smart home environment (Fig. 7). To this end, each record consists of data fields, such as  
 505 the time it occurred (according to the simulator clock), the sensor ID, and the sensed value. Moreover,  
 506 due to the fact that such values are raw data, they must be interpreted by the Analyzer module in order  
 507 to generate the detection of events and then calculate the values of the AGGIR variables. Subsequently,  
 508 having collected all the results, the next step is the detection of events. For this purpose, an event is  
 509 considered as a composite act that is described by data provided by several sensors; that is to say, an

XML File Syntax	
Attribute	Description
startdate	Starting date of script execution
factor	Speed factor for execution

```
1 <behavior startdate="27/06/2017-00:00:00" factor="1440">
```

---

Zone creation	
Attribute	Description
id	Zone's Identifier
leftX	Left X coordinate value
topY	Top Y coordinate value
bottomZ	Bottom Z coordinate value
X-Length	Zone X-Length
Y-Length	Zone Y-Length
Z-Length	Zone Z-Length

```
...
7 <create-zone id="bedroom" leftX="55" topY="370" X-Length="259" Y-Length="210" />
8 <create-zone id="bathroom" leftX="55" topY="20" X-Length="260" Y-Length="350" />
9 <create-zone id="Toilet" leftX="143" topY="299" X-Length="45" Y-Length="45" />
```

---

Device creation	
Attribute	Description
id	Device's Identifier
type	Device's type

```
...
28 <create-device id="Pres-A1010" type="iCasa.PresenceSensor" />
29 <create-device id="Pres-A1011" type="iCasa.PresenceSensor" />
30 <create-device id="Pres-A1012" type="iCasa.PresenceSensor" />
```

---

Person displacement	
Attribute	Description
personId	Person's Identifier
zoneId	Zone's identifier

```
...
43 <move-device-zone deviceId="Pres-A1009" zoneId="Chairs" />
44 <move-device-zone deviceId="Pres-A1010" zoneId="Oven" />
45 <move-device-zone deviceId="Pres-A1011" zoneId="washingbasin" />
```

---

Person creation	
Attribute	Description
id	Person's Identifier
type	Person's type

```
51 <create-person id="Paul" type="Grandfather" />
```

---

Person displacement	
Attribute	Description
personId	Person's Identifier
zoneId	Zone's identifier

```
...
58 <move-person-zone personId="Paul" zoneId="bedroom" />
59 <move-person-zone personId="Paul" zoneId="bathroom" />
60 <move-person-zone personId="Paul" zoneId="Toilet" />
```

---

Simulation delay	
Attribute	Description
value	Time in (virtual) units
unit	Time unit, value it can be h (hours), m (minutes, default), s (seconds)

```
...
85 <move-person-zone personId="Paul" zoneId="Sofa" />
86 <delay value="1" unit="h"/>
87 <delay value="30" unit="m"/>
```

---

```
...
```

**Figure 9.** Excerpt of an XML format file scenario employed by the simulator module

510 event relies on more than one record in raw data. Afterwards, the DSL is in charge of providing an  
511 accurate and unified description of the events. Once the event detection has occurred, the aim of the  
512 variable calculator within the Analyzer module is indicating whether an AGGIR variable has been  
513 accomplished or not.

514 In order to test the whole approach, we show its application to several scenarios, as we present in  
515 the following section.

### 516 *6.1. Experimental Results*

517 We designed two scenarios for the experimental evaluation. First scenario, consisting on an ideal  
518 week case scenario, in which all the ADL related to a specific AGGIR variable performed by the  
519 inhabitant are conducted successfully. Second scenario, with regard to the generation of a failure on the  
520 daily routine of the inhabitant, by randomly dropping some of the ADL conforming a specific AGGIR  
521 variable. Moreover, in order to generate the required simulations, the aforementioned scenarios consist  
522 on two sets of inputs managed by the proposed framework. Each one of such scenarios is conformed by  
523 seven days (one-week period).

524 In what follows, the main characteristics of the simulation scenarios concerning the identification  
525 of ADL as well as the AGGIR grid variables are described. In addition, since it is required to evaluate  
526 the ability of the proposed approach to deal with larger time domains, as well as several records and  
527 events; the scalability of the introduced framework is also illustrated.

#### 528 *Toileting, Dressing, and Transfer AGGIR Variables*

529 Conditions regarding the achievement of a certain number of events within a period of time must  
530 be accomplished for the evaluation of the AGGIR variables. In order to detect the toileting AGGIR  
531 variable the following conditions must be met: the inhabitant must use the toilet, and wash his hands  
532 after eating or using the toilet, for at least three times a day. To calculate the transfer variable, namely  
533 the ability of a person to perform the basic movements of his daily routine, such as rising from bed,  
534 sitting down, and standing up from a chair; it is considered that there must be at least three sitting  
535 events a day, either taking place in the living room or in the kitchen, and at least rising from the  
536 bed one time per day. While the verification of the dressing variable relies on dressing events such as  
537 approaching the wardrobe at least twice a day.

538 In this sense, we represent all these variables as follows. The conditions for the achievement of  
539 the toileting AGGIR variable are: (i) the inhabitant must use the toilet; (ii) after, the chain of the  
540 toilet flush must be pulled; (iii) after using the toilet, the inhabitant should have his hands washed.  
541 In order to analyze such a variable, the records created by the presence sensor located within the  
542 bathroom area, alongside with the records issued from the use of both the toilet flush sensor and the  
543 washbasin proximity sensor, must be examined. To identify the AGGIR dressing variable, the rules for  
544 the detection of ADL composing such a variable are: (i) the inhabitant must get close to the wardrobe  
545 area; (ii) the inhabitant must spend time changing clothes. To this end, data originated by the wardrobe  
546 door sensor, as well as the wardrobe proximity sensor, must be gathered. For the purpose of recognizing  
547 the AGGIR transfer variable, the constituent events regarding the aforementioned variable must occur:  
548 (i) getting up from bed; (ii) taking a seat; (iii) standing up from a chair; not necessarily in that order.  
549 In this matter, capacitive sensors located in both bed and chair/armchair are responsible for the data  
550 collection.

551 In order to illustrate the three AGGIR variables outlined above, Fig. 10 provides a description of  
552 all activities related to the three considered AGGIR variables, as well as sensors related to each one.

553 With a view to perform the evaluation of the AGGIR variables, all the simulating and processing  
554 operations are performed by means of two sets of inputs for the introduced framework, in order to  
555 generate two different scenarios. For this matter, the week is separated into seven days, each of which  
556 is represented by means of one simulation file within the smart home environment simulator.

AGGIR Variable = Situation		
Toileting	Dressing	Transfers
Situation = $\sum(+)$ Activities		
Open bathroom door	Open wardrobe door	Lying down
Use of toilet flush	Spend time changing clothes	Sitting down
Wash hands	Close wardrobe door	Getting up
Close bathroom door		
Activity = $\sum(+)$ Sensor data		
Bathroom door sensor (On-Off / Boolean)	Wardrobe door sensor (On-Off / Boolean)	Bed capacitive sensor (On-Off / Boolean)
Toilet flush sensor (On-Off / Boolean)	Wardrobe proximity sensor (On-Off / Double)	Chair capacitive sensor (On-Off / Boolean)
Washbasin proximity sensor (On-Off / Double)		

**Figure 10.** Orchestration of activities / AGGIR Variables

557 To this effect, the first scenario is simulated of an ideal week case scenario, meaning that all the  
 558 ADL were performed with no impediment by the inhabitant throughout the entire week, by means of  
 559 a simulated sensor network within a house environment for monitoring the ADL carried out by the  
 560 elderly resident. Results obtained from the simulation are presented in Table 17, in which, the number  
 561 of detected activities per day related to each one of the considered AGGIR variables are shown. The  
 562 check mark means that the criteria for determining that the person is independent (i.e., minimum  
 563 number of activities related to each variable) has been reached.

564 For the purpose of generating a malfunctioning on the developed criteria, some of the ADL related  
 565 to the evaluated AGGIR variables were randomly excluded in the second scenario during the seven-day  
 566 period of analysis. Once the simulation was conducted, the obtained results demonstrate that the  
 567 detection of complex events regarding either the achievement or the absence of ADL were carried out  
 568 successfully, according to the pre-established conditions, as enlisted on Table 18.

**Table 17.** Results of the first scenario.

		Transfer	Toileting	Dressing
No. of events	MON	☑ 7	☑ 8	☑ 3
	TUE	☑ 6	☑ 8	☑ 3
	WED	☑ 7	☑ 11	☑ 2
	THU	☑ 7	☑ 10	☑ 2
	FRI	☑ 8	☑ 8	☑ 3
	SAT	☑ 8	☑ 6	☑ 4
	SUN	☑ 6	☑ 9	☑ 3
	<b>Total</b>	☑ 49	☑ 59	☑ 19

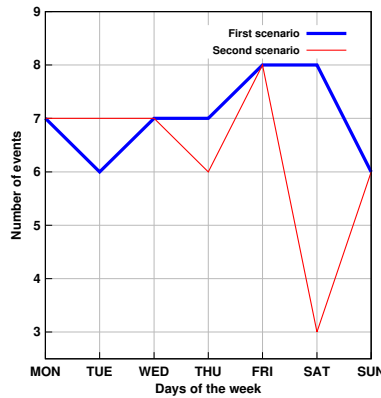
**Table 18.** Results of the second scenario.

		Transfer	Toileting	Dressing
No. of events	MON	☑ 7	☑ 7	☑ 3
	TUE	☑ 7	☑ 7	☑ 0
	WED	☑ 7	☑ 11	☑ 2
	THU	☑ 6	☑ 11	☑ 3
	FRI	☑ 8	☑ 8	☑ 3
	SAT	☑ 3	☑ 6	☑ 4
	SUN	☑ 6	☑ 6	☑ 2
	<b>Total</b>	☑ 44	☑ 56	☑ 16

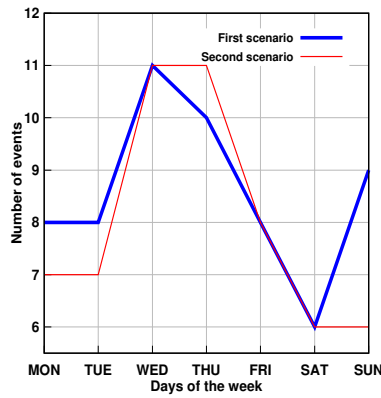
569 Furthermore, the proposed method succeeded in identifying three simulated problems within a  
 570 week. The days were randomly chosen. Fig. 11 and Fig. 12 show that despite the number of records in  
 571 the second scenario related to personal transfer and toileting did not change significantly compared to  
 572 the first analyzed scenario, however a problem with the dressing variable was detected, as illustrated  
 573 on Fig. 13. This, in turn, reflects how the DSL event descriptor can perform a smart analysis of events.

#### 574 *AGGIR alimentation variable*

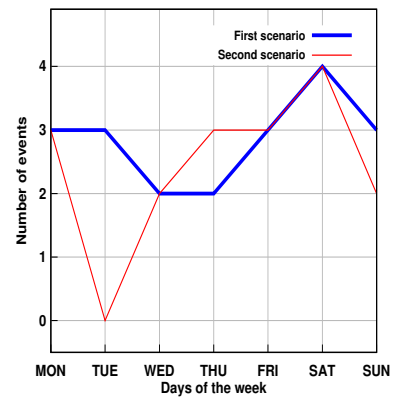
575 With the aim of evaluating the alimentation AGGIR variable, for the first scenario, the ADL  
 576 related to cooking must be performed by the inhabitant before the alimentation AGGIR variable takes  
 577 place. That is to say, the accomplishment of preconditioned events conforming the cooking activity is  
 578 necessary; more particularly: (i) open/close the refrigerator door; (ii) open/close the kitchen cabinet  
 579 door; (iii) use of the stove/burners; (iv) use of the traditional/microwave oven; not necessarily in that  
 580 order. Whereas, as for the second scenario, some of the abovementioned events concerning the cooking  
 581 task were dropped. For this matter, information collected from sensors located within the kitchen  
 582 area must be analyzed. To this end, data originated by both the refrigerator and kitchen cabinet door  
 583 sensors, the stove magnetic sensor, as well as the oven / microwave oven sensor, must be gathered.



**Figure 11.** AGGIR  
Transfer events



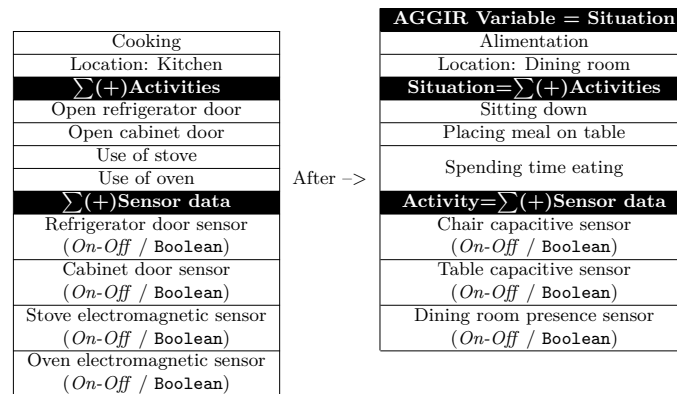
**Figure 12.** AGGIR  
Toileting events



**Figure 13.** AGGIR  
Dressing events

584 Nonetheless, it is important to remark that according to the AGGIR grid, cooking is not conceived  
 585 as a discriminatory variable, neither as an evaluation criterion for the alimentation AGGIR variable;  
 586 which only requires the elderly being capable of serving and eating a prepared meal. For this reason, in  
 587 our approach, cooking is considered as an ADL which can trigger the Alimentation AGGIR variable,  
 588 not meaning to contradict the pre-established evaluation criteria, which can also be used for detecting  
 589 the aforementioned variable in this work.

590 Once the cooking task is achieved, additionally to the first scenario, events defining the alimentation  
 591 AGGIR variable must be identified, such as: (i) sitting within the dining room area; (ii) placing meal  
 592 on the table; (iii) spending time consuming food. Moreover, with the aim of identifying the occurrence  
 593 of the abovementioned events, data from sensors located within the dining room area must be collected,  
 594 such as: chair capacitive sensor, table capacitive sensor, dinning room presence sensor. With respect to  
 595 the second scenario, some of the ADL conforming the alimentation AGGIR variable, were arbitrarily  
 596 left out. Additionally, in order to assure a coherent self-feeding behaviour, both scenarios must occur  
 597 at least three times a day. The orchestration of activities regarding the alimentation AGGIR variable is  
 598 illustrated on Fig. 14.



**Figure 14.** Orchestration of the AGGIR Alimentation Variable

599 Having completed the simulations of both scenarios, results concerning the first scenario proves  
 600 that, indeed, the detection of complex events conforming the cooking tasks, as well as those from  
 601 the alimentation AGGIR variable were carried out to completion successfully, in accord with to the  
 602 predetermined input conditions, as enlisted on Table 19, in which, check marks indicate that the  
 603 minimum number of detected events related to the aforementioned variable have been performed  
 604 satisfactorily.

605 Moreover, as for the second scenario, due to the fact that some of the activities were not taken  
 606 in account for the analysis of both the cooking activity and the alimentation AGGIR variable; the

607 proposed framework also succeeded in identifying either the concerned ADL was performed or was not  
 608 completed according to initial parameters, as presented on Table 20.

Table 19. Results of the first scenario

		Cooking	Alimentation
No. of events	MON	✓ 12	✓ 7
	TUE	✓ 11	✓ 9
	WED	✓ 9	✓ 9
	THU	✓ 11	✓ 8
	FRI	✓ 10	✓ 8
	SAT	✓ 10	✓ 9
	SUN	✓ 9	✓ 7
	<b>Total</b>	✓ 72	✓ 57

Table 20. Results of the second scenario

		Cooking	Alimentation
No. of events	MON	☒ 7	✓ 9
	TUE	✓ 10	☒ 6
	WED	✓ 11	✓ 8
	THU	☒ 5	✓ 8
	FRI	✓ 9	☒ 3
	SAT	☒ 7	☒ 5
	SUN	✓ 10	✓ 9
	<b>Total</b>	☒ 59	☒ 48

609 Furthermore, in order to illustrate the number of events related to the cooking task, as well as for  
 610 the alimentation AGGIR variable, Fig. 15 and Fig. 16 introduce by means of a graphical representation  
 611 the events detected for both scenarios. Additionally, it can be observed for each case that the second  
 612 scenario present a significant threshold with respect to the first scenario in some of the simulation  
 613 weekdays, which implies that both the cooking activity and the alimentation AGGIR variable were  
 614 not achieved particularly on those days; and hence, during the evaluated week concerning the second  
 615 scenario.

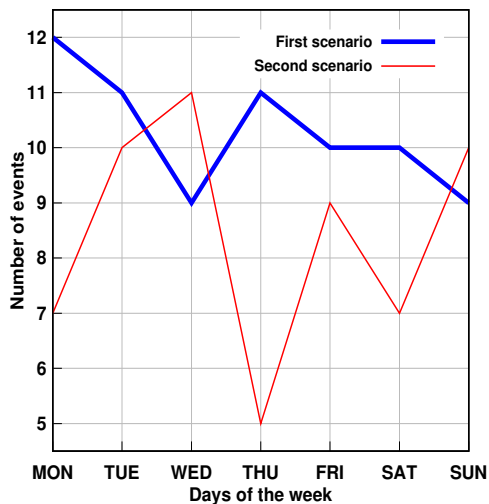


Figure 15. Cooking events

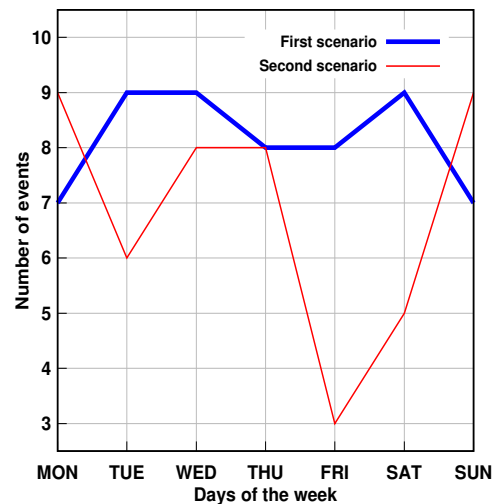


Figure 16. AGGIR Alimentation events

#### 616 Health-care device use case

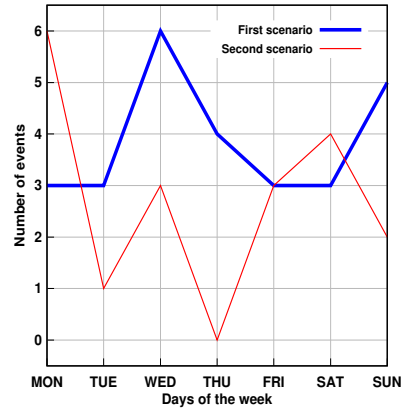
617 In order to prove the flexibility of the proposed framework, a health-care device use case is  
 618 introduced. This case considers a resident in charge of taking glucose level measurements by means of  
 619 a glucometer, performing a type of measurement other than those listed in the AGGIR grid. We design  
 620 two scenarios, in which to ensure continuous measuring, it is necessary for the house resident to perform  
 621 such measurements at least three times a day. To this extent, the first simulation scenario consists  
 622 of a week during which the person was completely responsible of employing a glucometer device for  
 623 measuring his own glucose levels, meaning that the completion of a planned medical task was achieved.  
 624 The second scenario was thought to randomly exclude some of the measurements throughout several  
 625 days of the week. All the simulations and operations are carried out by the proposed framework by  
 626 means of records corresponding to each scenario. Results obtained from the simulations corresponding  
 627 to both scenarios, are presented in Table 21.

628 With regard to the simulation results, it can be observed in Fig. 17, that the events occurred  
 629 during the simulation week for the first scenario were satisfactorily detected, which implies that the

630 measurement of the glucose level were achieved in accordance with the preset criteria, meaning that the  
 631 house resident is in charge of planned tasks in a coherent manner. Additionally, information obtained  
 632 from the second scenario shows that the inhabitant did not succeed in taking charge of a planned task,  
 633 due to the fact that the number of identified events per day were not performed coherently (three to  
 634 six times a day); more precisely, tuesday, thursday, and sunday.

**Table 21.** Results of the first scenario

		First scenario	Second scenario
No. of events	MON	☑ 3	☑ 6
	TUE	☑ 3	☒ 1
	WED	☑ 6	☑ 3
	THU	☑ 4	☒ 0
	FRI	☑ 3	☑ 3
	SAT	☑ 3	☑ 4
	SUN	☑ 5	☒ 2
	Total	☑ 27	☒ 19

**Figure 17.** Glucose device events

635 This latter can be interpreted as the efficiency of the framework to perform the recognition of  
 636 planned tasks concerning measurements carried out by the user by means of medical devices.

637 We also perform experiments to evaluate the framework scalability and performance. In what  
 638 follows, we describe the results regarding these experiments.

### 639 6.2. Framework scalability

640 To validate the execution of the framework under event processing conditions regarding both  
 641 larger time domains and higher workload cases, the scalability of the framework for a period of time  
 642 longer than one week is tested. For this matter, a time lapse of three months was simulated in order  
 643 to generate the corresponding records. Next, the proposed criteria were applied with the intent to  
 644 create the events, and thus calculate the values of three AGGIR variables: toileting, transfers, and  
 645 dressing. In addition, with a view to allow flexibility when processing records and generating events,  
 646 the assumption that each month is conformed by thirty days was considered. By doing so, there is  
 647 a possibility to perform the required simulations by employing an approximate number of days per  
 648 month; i.e., 28, 29, 30, and 31 days (Fig. 18).

649 Finally, to make sure that the algorithm is scalable, the elapsed time during the calculation of  
 650 the three abovementioned AGGIR variables for simulation time periods for one day, one week, two  
 651 weeks, three weeks, a month, two months, and three months; were monitored. Table 22 summarizes the  
 652 elapsed time for each case. The values displayed include the time spent for each simulation, as well as  
 653 results analysis, events generation, and finally the calculation of the AGGIR variable.

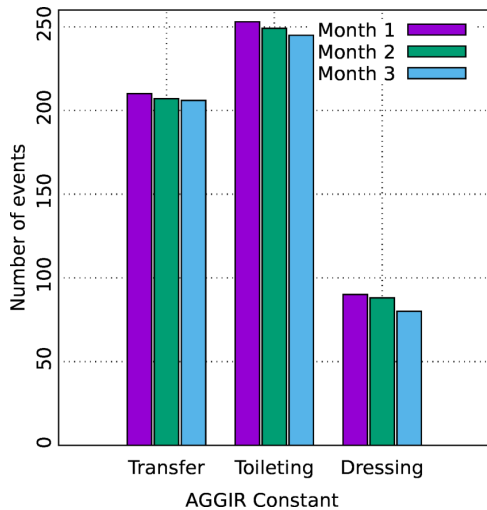
**Table 22.** Real time elapsed to calculate the constants in seven different simulation time ranges (*Day, Week, Month*).

Time period	1D	1W	2W	3W	1M	2M	3M
No. of days	1	7	14	21	30	60	90
Time (s)	62	434	867	1301	1863	3713	5590

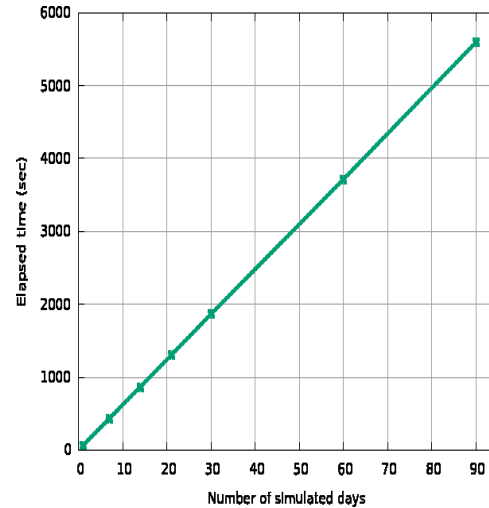
654 In order to present the results thus achieved, Fig. 19 shows the performance of the algorithm with  
 655 simulating scenarios carried out with different time ranges. As a result, the obtained diagram is linear,  
 656 which corresponds to the complexity of the algorithm of the Analyzer module, represented by  $O(n)$ ,  
 657 where  $n$  defines the number of events, implying that the framework as a whole, is scalable.

658 With the aim of applying the above-mentioned steps, several use cases are provided; namely, the  
 659 alimentation, toileting, cooking and transfer AGGIR variables; as well as a description regarding an use





**Figure 18.** Three-month algorithm scalability



**Figure 19.** Analysis of Algorithm performance

660 case of a medical device for measuring blood glucose level. Moreover, results obtained show that the  
 661 proposed framework succeeding on detecting the events and therefore the AGGIR variables according to  
 662 preestablished conditions present in the experimentation scenarios.

## 663 7. Conclusions

664 In recent years, it has been reported that the population sector targeted as elderly is increasing  
 665 more rapidly than in the previous decades. Besides that, age-related health issues as well as different  
 666 chronic diseases start to become apparent at this stage of life. Moreover, various surveys indicate that  
 667 elderly people are willing to live independently in their own homes as long as possible. In order to assess  
 668 the abovementioned problem, PSNs and BSNs in smart environments represent an option regarding  
 669 healthcare solutions for monitoring elderly people at home. However, despite their popularity, the  
 670 absence of a monitoring approach with regard to a dependency evaluation model such as the AGGIR  
 671 grid variables has never been treated. In this context, the core contributions of this research work  
 672 are: (i) a state-of-the-art related to the proposed subject and future circumstances with regard to  
 673 aging population, as well as the capabilities of technology for assisting the maintenance and monitoring  
 674 the health needs of people, are reviewed; (ii) a DSL relying on the AGGIR grid evaluation model  
 675 for assessing the performance of ADL of the elderly/handicapped at home; (iii) a framework tool for  
 676 validating the proposed DSL in terms of its capacity of processing complex events; this framework offers  
 677 a smart-home simulator called iCASA for carrying out evaluation and experimentation and a parser in  
 678 charge of processing the data issued from the DSL aimed to create the appropriate instructions for the  
 679 iCASA platform.

680 We are currently working on proposing the identification of a three-state variable approach for the  
 681 AGGIR grid. Due to fact that the current version of the introduced DSL is only able to recognise two  
 682 of the three values (**A** for complete dependency and **C** for complete independency) of the AGGIR grid  
 683 variables, the proposal of an improved version of the DSL comprehending the three states of identification  
 684 (including **B** for partial dependency), is one of the possible directions of the future work. For the  
 685 purpose of assessing AGGIR variables that can be considered as more elaborate, such as coherence,  
 686 further experimentation is required to evaluate crucial activities describing the concerned variable.  
 687 Even though the presented approach considers the evaluation of the AGGIR discriminatory variables;  
 688 with the goal of covering the whole spectrum of the AGGIR variables, the integration of the AGGIR  
 689 illustrative variables is one of the next steps to be taken. So far, the perspective introduced in this  
 690 research relies on the AGGIR grid variables. In order to expand the coverage of the suggested approach,  
 691 the inclusion of other dependency evaluation models (e.g., the Functional Autonomy Measurement

692 System – SMAF model), represents a field of opportunity to broaden the scope of the present work.  
 693 Besides the presence of the house residents, visitors can represent a factor for change regarding the  
 694 household dynamics. For this matter, the adaptability of the proposed approach must be determined by  
 695 tests concerning unknown users to the smart-home environment; as well as the impact that their visit  
 696 may cause to the actual inhabitants. Despite the fact of having employed a smart-home simulator, as  
 697 well as a schedule describing the actual activities that were carried out by the user; the application of  
 698 the present work for real-time use cases is recommended. To this extent, monitoring of ADL within an  
 699 actual smart-home environment would be ideal for testing the present approach in a real-time scenario.

700 **Author Contributions:** J.-M. Negrete Ramirez and Y. Cardinale conceived and designed the experiments;  
 701 Research for analyzing the proposed model, analyzing the experiments and revising the proposed model; E.  
 702 Silva performed the experiments; P. Roose, M. Dalmau and Y. Cardinale supervised the project; J.-M. Negrete  
 703 Ramirez and Y. Cardinale wrote the paper.

704 **Acknowledgments:** The authors would like to acknowledge the the Institut National des Sciences Appliquées  
 705 Centre Val de Loire and the Laboratoire d'Informatique Fondamentale d'Orléans for supporting the publication  
 706 of the research herein presented.

707 **Conflicts of Interest:** The authors declare no conflict of interest.

## 708 Abbreviations

709 The following abbreviations are used in this manuscript:

710	ADL	Activities of Daily Living
	AGGIR	Autonomy Gerontology Iso-Resources Groups
	AI	Artificial Intelligence
	BPM	Beats per minute
	BSN	Body Sensor Networks
	BTS	Body Temperature Sensor
	CPM	Contractions per minute
	DSL	Domain Specific Language
711	ECG	Electrocardiogram
	EMG	Electromyography
	GSR	Galvanic skin response
	GUI	Graphical User Interface
	IoT	Internet of things
	PPM	Peaks per minute
	PSN	Personal Sensor Networks
	SMAF	Functional Autonomy Measurement System
	SPM	Snores per minute

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