Semantic Interoperability Architecture for Smart Spaces

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Abstract

The advancements in the smart space research bring us closer and closer to a future, in which the living standards of people are greatly enhanced. Smart spaces are composed of heterogeneous networks of embedded devices that communicate, perform computations and actuate. From a system point of view, a smart space is a specific type of ubiquitous system. Like any ubiquitous system, a smart space is usually heterogeneous, i.e. it is made up of embedded devices with various hardware, software and communication platforms. There is need for information exchange and interoperability between these embedded devices and their services. However, embedded devices in a smart space vary in terms of their sensing, actuation and control capabilities, as well as their resources of processing, storage, and networking. For an interoperability solution, this paper proposes a semantic interoperability architecture for smart spaces. The description details of an ontology model and interoperability interactions between smart objects are explained. An example implementation of a heterogeneous smart lighting system as a smart space using the proposed semantic interoperability architecture is discussed.

Keywords: Smart spaces, Semantic interoperability architecture, Smart lighting

1. Introduction

In smart spaces [1], the hardware and software of an individual object can differ from another object. They may utilize different communication technologies; for example, power line technologies such as X10 [2] and LonWorks [3], wireless technologies such as IEEE 802.15.4, wired technologies such as IEEE 802.3, and CAT5 wires for audio, video or data communication. On the other hand, middleware such as Jini [4], HAVi [5] and UPnP [6] may need to connect the components of a smart space. These communication technologies and services supported by a middleware in smart spaces are causing a problem of heterogeneity even at layers above the network stack. Heterogeneity for smart spaces provides system interconnectivity with multiple entities to achieve joint execution of services. Heterogeneity at higher levels ultimately aims at semantic interoperability, which proves to be difficult due to the isolation of its components and the requirement for joint execution of services. This problem is generating great demand for a semantic interoperability solution.

Semantic interoperability depends on the contexts shared by its objects, which may utilize
different hardware and software platforms as well as different communication protocols, forming a heterogeneous network. In this respect, the contexts of an object are represented in semantics for the common understanding between objects. In linguistics, semantics refer to the ‘study of meaning’. It is used for understanding human expression by the basic elements of a language. These elements are referred to as symbols, for example, a character string that is used to (collectively) designate a meaning. The meaning is to be derived from the relations between symbols, concepts and things in the real world. What we express a concept in symbols is possible to express in different symbols by others, but the meaning of the symbols help to understand the concept. This means that any two or more symbols can have the same meaning in the real world. For example, ‘human’ in the English language is known as ‘menselijk’ in the Dutch language and humain in French but the meaning of the respective symbols collectively is same in all three languages.

Semantics are constructed with syntax and pragmatics. In linguistics, the syntax defines the rules, principles and processes by which (potentially meaningful) sentences are constructed in a language. The pragmatics is the study of relationships between a sentences made using such syntax and their meanings. Several researchers in linguistics and programing language fields are working on semantics [7–9]. In this paper, we study the meaning of contexts, which is referred to as semantics. Contexts are represented in semantics in a way that allows certain types of relationships to be discovered by ontologies. In other words, ontologies capture semantics in a way that allows relationships to become visible. There are two standards, which govern instructions of ontologies: Resource Description Framework (RDF) [10] and Web Ontology Language (OWL). Smart spaces ontologies created using these standards help to develop semantic interoperability between objects in smart spaces. Semantic interoperability means that a source object and a receiver object can interact in a meaningful manner through shared semantics. All contexts in a smart space have some meaning and it is necessary to make the mapping in between. Semantic interoperability can be established between two objects if the mapping between contexts, more specifically the semantics corresponding to these contexts, define an onto mapping (surjective) between the objects. The contexts $C$ represent semantics $S$ in a domain of smart spaces, where $C$ is a set of contexts and $S$ is the set of semantics.

**Definition 1 (Semantic interoperability):** Consider a function $f$ (of onto mapping) that maps the elements of a set $C$ to elements of a set $S$, $f : C \rightarrow S$. Semantic interoperability is given by

$$s = f(c).$$

By definition of the onto mapping $f$ maps every element $c \in C$ to a corresponding element $s \in S$ and multiple elements of $C$ may correspond to the same element of $S$.

For example, consider the set of context $c_a$ involves temperature and light sensors; and $c_b$ involves temperature, light and humidity sensors. Semantic interoperability established either when the unique elements of $c_a$ and $c_b$ represent the unique semantics $s_a$ and $s_b$; i.e., $f_A(c_a) = s_a$ and $f_B(c_b) = s_b$ where $\forall c \in C \land \forall s \in S$ or when the intersection of $c_a$ and $c_b$ represents the unique $s_c$, i.e., $f_C(c_a \cap c_b) = s_c$ where $\forall c \in C \land \forall s \in S$. This means when the measured temperature in the same activity area of $c_a$ and $c_b$ must have a single semantic representation for semantic interoperability whereas the measured temperature at different activity areas must represent the unique semantic for each context.

Semantics may be dynamic and may change over time. Semantic interoperability must be able to modify the processing of ever changing semantics. The changes in semantics also must follow the conditions defined in Definition 1. This requires a smart space architecture to have compatible mechanisms for representing, modifying and updating semantics in order to produce meaningful and valid results. In this paper, we adopt the baseline architectures i.e., Smart-M3 form high capacity smart objects (HCSOs) that have plenty of resources and OSAS (Open Service Architecture for Sensors) [11][12] for low capacity smart objects (LCSOs) that are resource poor. We propose a semantic interoperability architecture for exchanging contexts between objects. Several other researchers are working on designing a semantic interoperability architecture [13][15]. In the following subsections, we briefly explain our baseline architectures for the proposed semantic interoperability architecture of Smart-M3.

Smart-M3 is a blackboard architecture and provides a solution for information interoperability in smart spaces. Smart-M3 introduces three main components of a smart space, by which the information can be stored, removed and browsed: Knowledge Processor (KP), Semantic Information Broker (SIB), and Smart Space Access Protocol (SSAP). A KP is an entity that produces or consumes information in Smart-M3. SSAP is a protocol that defines a simple set of primitives to insert, remove
and access data in the SIB, and it can be used on top of transport technologies such as TCP/IP and Network on Terminal Architecture (NoTA) [16]. A KP can use SSAP for several transactions such as join, leave, insert, remove, update, query, subscribe, and unsubscribe to manage information in a smart space. Since SSAP is transport layer independent, the same SIB can serve KPs over several transport technologies at the same time.

The KP-SIB interface architecture is shown in Figure 1. The KP functionality is split into three layers: KP Interface (KPI), KP Daemon and SIB access transport technology. The KPI provides mappings of SSAP to different programming languages and frameworks. It also handles the SSAP operations on the KP side and interfaces to a KP Daemon. The KP Daemon provides a transport-independent view of the available SIBs to the KPI. It will map each SIB access to the correct transport in a SIB access transport technology. The SIB access transport technology is made up of different transport processes, which connect to the KP Daemon. Currently, NoTA SIB access and TCP SIB access transports have been implemented. The SIB is composed of components of transport services and the SIB daemon. The transport services such as NoTA SIB service and TCP SIB service are responsible for receiving messages from appropriate transports and parsing the messages. The SIB daemon is composed of three logical layers inside: Request Handling, Graph Operations and Persistent Store. The request handling will handle the SIB operations from KPs that are received by the transport services, perform the necessary graph operations and send the confirmation back to the KPs. All requests are handled in separate threads. The graph operations are responsible for translating the operations on graphs (sets of semantics) to semantics operations that can be stored at a persistent store. The persistent store is responsible for storing the semantics that make up a graph, and performing read, write and delete operations on them.

2. Semantic Interoperability Architecture

The proposed Semantic Interoperability Architecture (SIA) for smart spaces (based on our baseline architectures Smart-M3 and OSAS) in a blackboard architecture style is shown in Figure 2. It consists of four basic physical components: semantic producers and consumers, semantic repository, semantic gateway, and LCSOs. The definitions of these components are given in Table 1. All these components of semantic based smart space are integrated by an ontology model and operated by the KP interface.

A semantic integration approach is developed for sharing information among LCSOs and HCSOs. There have mainly been two approaches to integration of LCSOs and the Internet so far: (i) Pure TCP/IP solutions, in which LCSOs implement the TCP/IP stack (or a compatible set of protocols such as 6LoWPAN [17, 18] in 802.15.4 networks); and (ii) gateway solutions,
in which a particular node acts as an application layer gateway (e.g., base station), for the purpose of making the lower layer protocols from both networks (e.g. TCP/IP and IEEE 802.15.4) transparent and routing the information from one point to another. In the first approach, sensor nodes are able to address Internet nodes directly. However, this approach does not scale well since each and every LCSO in smart spaces would need to exchange information with the SIB very frequently. In the second approach, Internet hosts and LCSOs are able to address each other and exchange information without establishing a direct connection. Therefore, we propose a semantic gateway that runs an OSAS framework for installation of subscriptions and exchanging services among LCSOs. It also helps to connect with HCSOs and translates the information needed to be exchanged.

A semantic repository, i.e., a SIB, is the semantics repository of a smart space. It provides a service interface for KPs i.e., KPI. The KPs can access semantics within a smart space by connecting to a SIB. The KPs can insert new semantics into a smart space and remove obsolete or otherwise undesirable semantics from a smart space. The semantic repository also reacts to the changes in the semantics content through subscriptions in a smart space. For example, the subscriptions are persistent queries that notify the subscribing KP every time the query results change. Therefore, this allows the KPs to react appropriately to the changes in smart spaces. The complete SIB interface is a union of the consumer and producer interfaces.

Since there are several KPs connected to a smart space, the KPs cannot expect the inserted semantics to remain unchanged until the time they read it again. Also, the insertion and removal of the information may leave the content of the smart space in an inconsistent state with regards to any ontology. This is prevented by an internal ontology enforcement process, or fixed by an internal ontology repair process, which are run automatically by a SIB. This ontology repair and enforcement are additional features, which are integrated in the SIB by Smart-M3.

An ontology model is required for the ever-changing domains in smart spaces, which provides a flexible and up-to-date description of the domains. Therefore, an ontology model can play the role of a flexible unit in smart spaces to achieve semantic interoperability. An ontology model has a semantic reasoner component to provide the decision rules and is supported by the semantic reasoner component and the SSAP protocol in the semantic interoperability architecture.

The semantic reasoner provides a language to translate information into semantics, where the information is specified using the RDF vocabulary description language. RDF provides a common framework for expressing the information of resources so that it can be exchanged between objects without loss of meaning. RDF is based on the idea of identifying things using Web identifiers (called Uniform Resource Identifiers, or URIs), and describing resources in terms of simple properties and property values. This enables RDF to represent simple statements about resources as a graph of nodes (representing resources) and arcs (representing resource properties and values). Resources may be divided into groups called classes. The members of a class are known as instances of the class. Classes are themselves resources. They are often identified by RDF URI references and are described using RDF properties. The structure of the RDF representation shares the meaning of contexts between objects. For this purpose, the classes and their relationships are represented, e.g., using RDF triples. An RDF triple consists of subject, predicate and object, represented by {Subject; Predicate; Object}, where the subject and predicate are the resources and their properties defined by a URI while the object is a literal or a resource. The generic RDF triple graph with an example is shown in Figure 3. For instance, the information “light has intensity of 500 lux” is represented as semantics (i.e., a set of triples) given by {("Light", "hasIntensity", "500"), ("Light", "hasUnit", "lux")}. RDF also allows advanced structures such as classification and containment to the features supported by XML, similar to object-oriented programming languages. Classification is

### Table 1. Description of SIA components

<table>
<thead>
<tr>
<th>Name</th>
<th>Definition</th>
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<tbody>
<tr>
<td>Semantic producer</td>
<td>Objects with capability of producing semantics by translating information of a surrounding physical environment into semantics of the smart space.</td>
</tr>
<tr>
<td>Semantic consumer</td>
<td>Objects with capability of consuming semantics received from the smart space.</td>
</tr>
<tr>
<td>Semantic repository</td>
<td>An entity with capability of storing semantics and providing an ontology model for reasoning.</td>
</tr>
<tr>
<td>Semantic gateway</td>
<td>An entity to work as an interpreter of translating physical readings into semantics and vise-versa.</td>
</tr>
<tr>
<td>Smart space access protocol</td>
<td>Provide transactions facility through the sets of instructions defined in the protocol such as query, subscription between smart space entities.</td>
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realized for a resource through specifying an  <rdf:type> property whose corresponding value is  <rdfs:Class> from the RDF schema. Instances of a specific class can then be defined by creating a type-tag that points to the corresponding resource. The example of the RDF format is shown in Table 2.

The update in the RDF graph is done by a semantic reasoner, where the semantic reasoner defines a logical reasoning based on the decisions rules. The decision rules are defined by SmartM3 and used directly in this paper. The SIB has a reasoning engine based on reasoning with OWL. The reasoning engine has RDF schema [19] which constitutes the classes and its properties.

The SSAP transactions are used in the semantic interoperability architecture through SSAP protocol to communicate between KPs and SIB. The SSAP protocol has in total 18 different message types. These messages are used to control the transactions: JOIN, INSERT, REMOVE, UPDATE, SUBSCRIBE, UNSUBSCRIBE and QUERY [20].

3. Semantic Interoperability Interactions

We have considered a general scenario based on two smart spaces consisting different SIBs called as Smart Space A (SS-A) and Smart Space B (SS-B) with SIBA and SIBB respectively as shown in Figure 4. This general scenario explains the semantic interoperability interactions for exchanging information between two smart spaces. SS-A consists of many KPs such as KP_{A1}...KP_{An} and a gateway KP (GWKP) for LCSOs. In this smart space, a KP shares information with other KPs that are joined to the same SIB_A. Therefore, KP_{A1}...KP_{An} to join the SIB_A make a local smart space for all KPs in this smart space. The communication technology for all KPs is SSAP protocol over Internet and for LCSOs, it is based on IEEE 802.15.4 communication standard. SS-B consist of KP_{B1}...KP_{Bn} and similarly make a local smart space with all KPs in this smart space. If a KP from one smart space need to share information globally then it requires a Smart Space Aggregator KP (SSA-KP) to connect both smart spaces. The SSA-KP is a KP and is used to aggregate the information between these two smart spaces, i.e., connect SIB_A and SIB_B. Therefore any KP such as KP_{A1} can exchange information with KP_{B1} through SSA-KP.

The interactions of KPs, GWKP, SIB-A, SIB-B and SSA-KP are shown in Figure 5. All KPs, GWKP, and SSA-KP need to join a SIB to be present in a smart space. Here, all KP-A and GWKP are connected to SIB_A and all KP-B are connected to SIB_B. The SSA-KP is connected to both SIBs to aggregate information from SIB_A and SIB_B. The respective SIB to a particular KP gives confirmation after successfully joining the SIB. Therefore, two transactions are established in joining transaction i.e., joining request (Join-req) by a KP and joining confirmation (Join-cnfr) by a SIB. A producer KP needs to first insert information (In-Info) at SIB and then update the information (Up-Info) periodically. In this case study, we have two producer KPs such as GWKP and KP-A. Therefore, the producer KPs has initial transaction (In-Info) and then the numbers of further transactions are based on an updating period. A consumer KP needs to subscribe (Sub-Info) for particular information in smart spaces and get updates accordingly from SIB (Up-Info). Similar to the insert transaction for the producer
KPs, the consumer KPs need to subscribe once and wait updates from SIB. Then the SIB updates the subscribed information periodically. In this case study, KP-B in SS-B behaves like a producer and consumer KPs both. In addition to the GWKP, it needs to install address (Ins-Add), services (Ins-Ser), and subscriptions (Ins-Sub) on all LCSOs such as Sensor Node (SN) and Actuator Node (AN). The SN and AN report the information to the GWKP based on the subscription periods established for the services. Further this information is stored at SIB to exchange with other KPs. Finally, all KPs and GWKP can leave the smart space by a leave request (Leave-req) at SIB. Therefore, the smart objects are able to interact and share information through SSAP transactions.

4. Conclusions and Future Work

This paper proposes a semantic interoperability architecture and its components. The major components of the semantic interoperability architecture are semantic repository, semantic gateway, LCSOs, smart space aggregator, semantic producers and semantic consumers. Semantic interoperability is established by an ontology model and its interoperability interactions are explained. The smart lighting applications using the proposed semantic interoperability architecture is experimented in [12, 21]. We focused on sensor nodes and their services for gathering information from the environment in the implementation of the proposed solution. Examples of such information are temperature, humidity, light intensity and motion data. We considered information exchanged for a single application domain i.e., smart lighting. These experiments are possible to extend in terms of network traffic and multiple applications for a performance analysis.

The tradeoff between performance and scalability is not investigated yet within the context of the proposed semantic interoperability architecture for smart spaces. This investigation is considered as future work. The objective of this investigation will give the scope of using number of SIBs in a single smart space to maximize the performance for large scale smart space installations (e.g., for smart cities).

Conflict of Interest

No potential conflict of Interest relevant to this article was reported.

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