We propose a modification to the Service Discovery Protocol in Bluetooth in a mobile environment where devices like PDAs, cell phones or laptops communicate together, to provide and share various services. Within the environment, resources discovery is performed exploiting technologies and techniques for knowledge representation developed for the Semantic Web, which have been adapted to cope with the highly flexible structure of ad-hoc networks in ubiquitous computing. The proposed approach employs the standard Bluetooth stack, using the original UUID payload to carry semantically annotated data. The framework is motivated and presented in a museum case study.

1. Introduction

The growth in the diffusion of wireless-enabled handheld devices provides the necessary infrastructure for creating ad-hoc environments for ubiquitous computing. They are based on short range, low power technologies such as Bluetooth [4], which grant the peer to peer interaction among hosts. In such a mobile infrastructure there is one or more devices providing and using services. Since an ad-hoc network is a very unpredictable environment, Service Discovery (SD) becomes an essential feature. In an ubiquitous context, information and descriptions about services are often unavailable because the location of mobile devices could change continuously [5]. A flexible service
search system is desirable, based on wireless network infrastructure and able to overcome difficulties due to the host mobility. Existing service discovery methods use a syntactic matching which is largely inefficient in ad-hoc environments where a common service interface is unavailable. Here we need to submit articulate requests and to receive adequate answers [6]. In this paper we present an ontology-based environment where a semantic service search mechanism is used to obtain advanced discovery features, providing automated services to hosts participating to the ad-hoc network.

To achieve this goal, we borrow from the Semantic Web vision, both ideas and technologies, and export them in small wireless networks based on the Bluetooth technology. In a semantic-enabled Web, each resource is annotated using OWL-DL [1], an XML-based language, whose formal semantics is mapped to a formal logic language. There is a close relation between OWL-DL and Description Logics (DLs) [3] semantics, which allows the use of DLs reasoners in order to infer new information from the one stated in the annotation itself. We use some results and approaches we experienced both in semantic-enabled marketplaces [7] and semantic web services orchestration [10] to extend the Bluetooth Service Discovery Protocol (SDP) in order to provide semantic-based services to the users within the piconet.

The original Bluetooth standard uses SDP at the application layer. It is based on a 128 bit Universally Unique Identifier (UUID), where each UUID is associated to a single service class. Therefore in a Service Discovery Request the device specifies the requested service UUID and any host in the piconet, holding the service identified by the same UUID, replies to the requester. Then a communication can be established.

Obviously a code-based SDP can handle only exact matches. Yet, if we want to search and retrieve resources whose description cannot be classified within a rigid schema (e.g. the description of paintings in a museum or goods in a shopping mall [2]) a more flexible SDP is needed. Such a SDP must be able also to cope with non-exact matches and provide a ranked list of discovered resources, computing a distance between each retrieved resource and the requested one after a matchmaking process. Actually, a set-value based approach could satisfy, at a first glance, similar SDP requirements but imagine the following simple example related to a museum. Suppose you are looking for paintings whose subject is a portrait, \( R = \{\text{portrait}\} \), and in the museum there is a collection of self-portraits as offered resources, \( O = \{\text{self-portrait}\} \); \( O \) does not match \( R \) and nothing is known about their relations. Hence, no resource will be retrieved to answer \( R \). To overcome similar, and more complex, problems we need to model the meaning of the terms and their relations. That is, a representation of their semantics is needed. The SDP efficiency and flexibility can then be increased by exploiting knowledge representation techniques.

The rest of the paper is structured as follows: next section presents the framework and our approach; in section 3 we introduce our system and present its features and behaviour. In section 4 we comment on related work, and finally we present our conclusions in section 5.
2. Framework and approach

As pointed out in [2], Bluetooth SDP is largely inefficient when it comes to complex requests. In fact SD on Bluetooth only allows exact match discovery of uniquely identified services. This is a restriction in view of the transmission capabilities increase, devised in new drafts of the standard. A more advanced usage of service discovery protocol is desirable, associating semantic descriptions to the services rather than simple numeric identifiers. For such purpose knowledge representation (KR) technologies and techniques can be exploited and adapted to the ubiquitous environments. In the framework we present, after the wireless client has been identified within the piconet, it is able to share and retrieve information from other hosts.

In a typical configuration, a user contacts the zone service provider (hotspot) and submits a query about his/her interests. The server identifies clients able to share services and replies with found services, possibly ranked in a list according to their degree of correspondence to the request. The zone server classifies services contents by means of an ontology. In addition users submit requests in a DL language.

Then the hotspot collects the descriptions of the offered resources (modelled using DLs) and computes the matchmaking rank between the request and them. The provided result is a ranked list of offered resources potentially matching the user request.

It should be noticed that DL-based systems usually only provide two basic reasoning services:

- **Concept Satisfiability**: given a ontology $\mathcal{T}$ and a concept $C$, does there exist at least one model of $\mathcal{T}$ assigning a nonempty extension to $C$?

- **Subsumption**: given a ontology $\mathcal{T}$ and two concepts $C$ and $D$, is $C$ more general than $D$ in any model of $\mathcal{T}$?

In a semantically-enabled resource retrieval scenario, where a matchmaking process between a request $C$ and each available resource $D$ is needed, using subsumption it is possible to establish if $D$ is more specific than the request $C$, $D \sqsubseteq C$. If the previous relation holds, then the retrieved resource completely satisfies the request, i.e., an exact match occurs. With Concept Satisfiability the discovery of incompatible resources with respect to the request can be performed. If $C \sqcap D$ is not satisfiable w.r.t. the ontology $\mathcal{T}$, $D$ is not compatible with the request.

Although subsumption and concept satisfiability are very useful in several scenarios of resource discovery, as a matter of fact exact matches are usually
rare in a semantically-enabled matchmaking process and the above services result inadequate.

Typically, both $C \sqcap D$ is satisfiable and $\tau \not\sqcup D \sqsubseteq C$, that is $D$ is compatible with $C$ but it does not completely satisfy it. Then there is the need to go beyond subsumption and concept satisfiability to manage these frequent situations. A metric is needed to establish "how much" the resource $D$ is compatible with the request $C$ or, equivalently, "how much" it is not specified in $D$ in order to completely satisfy $C$, that is to make the subsumption relation $D \sqsubseteq C$ true. In [9] the rankPotential algorithm was presented, such that, given a set of $\mathcal{ALN}$ axioms $\tau$ and two $\mathcal{ALN}$ concepts $C$ and $D$ both satisfiable in $\tau$, it computes a semantic distance of $C$ from $D$ with respect to the ontology $\tau$. Notice that we write the distance of $C$ from $D$ rather than the distance between $C$ and $D$ because of the non-symmetric behaviour of rankPotential (see [9] for further details). In fact the relation we wish to reach here is $D \sqsubseteq C$ rather than $D \equiv C$.

With the aid of rankPotential it is also possible to compute a complex concept depth with respect to the taxonomy represented by the axioms set $\tau$. In fact, if $C \equiv T$ then $\text{rankPotential}(C,D) = \text{rankPotential}(T,D)$ represents the distance of $D$ from $T$, i.e., the most generic concept in the ontology. Notice that such distance is not trivially the depth of a node in a tree for at least two main reasons:

- An $\mathcal{ALN}$ ontology, typically, is not a simple terms taxonomy tree, i.e., it does not contain only IS-A relations between two atomic concepts and can be better represented as a labelled oriented graph.

- An $\mathcal{ALN}$ complex concept is the conjunction of atomic concepts and role expressions. The value returned by rankPotential$(T,D)$ here represents how specific is a complex concept expression $D$ with respect to an ontology $\tau$.

Notice that since the proposed approach is fully-compliant with Semantic Web technologies, the user exploits the same semantic enabled descriptions s/he may use in other semantic web compliant systems. That is, there is no need for different customized resource descriptions and modelling if the user uses different applications either on the web, or in mobile systems. The syntax and formal semantics of the descriptions is unique with respect to the reference ontology and can be shared among different environments.

In what follows we detail the discovery and matchmaking process with respect to a case study deployed in a museum. Suppose you are in a museum, which uses ontologies shared on the web to describe its resources. Such resources and related descriptions are also available on the museum web site. We associate the 128 bit UUID of the original Bluetooth standard to each specific ontology and we call this identifier OUUID (Ontology Universally Unique Identifier). In this way, a generic client request can be done by means of OUUID by hotspot thanks to other invisible ubiquitous hosts in the environment.
With respect to the implementation of matchmaking and ontology support features, we have inserted *Semantic Service Discovery* function into the stack with some light modifications of the discovery protocol. SDP uses a simple request/response method for data exchange between SDP client and SDP server [11].

Since the communication is referred to the peer layers of the protocol stack, each transaction is represented by one request Protocol Data Unit (PDU) and another PDU as response. In every SDP PDU, we have an header containing the identifier of the PDU, the identifier of the transaction and the length of the next PDU parameters field. If the SDP request needs more than a single PDU (this case is frequent enough if we use semantic service discovery) SDP server generates a partial response and SDP client waits for next part of the complete response. By adding to the original **SDP_SERVICESEARCHPDU** (request and response) and **SDP_SERVICEATTRIBUTEPDU** (request and response) the new **SDP_SEMANTICSERVICESEARCHPDU**, we insert into SDP the further semantic service search function.

No modifications are made to the original structure of transactions, but simply we differently use the SDP framework. Hence our approach allows to reuse UUID function within Bluetooth, without troubling the original standard, and furthermore it implements an advanced P2P exchange information mechanism, where users are peer clients in the piconet and can be both service requesters and possible service suppliers.
3. The system

In order to explain the approach and the rationale behind it, we refer to a scenario such as the one pictured in Figure 1. The purpose of such piconet is to find, within the ad-hoc network, resources requested by users (with a generic Bluetooth enabled device) searching among on-line available ones. Each resource in the environment owns an URI and exposes its OWL description. The hotspot is endowed with a MatchMaker (in our system we adapt the MAMAS reasoner [9]), which carries out the matchmaking process between each offered service in the piconet and the requested one measuring a "semantic distance". When a user becomes a member of the ad-hoc network, s/he is able to ask for a specific service/resource (by submitting a semantic based description). After receiving a user request, the system collects resources available in the area and performs the matchmaking process between their OWL-based descriptions and the user request. The matchmaking results are then ranked and returned to the user.

The generic steps from piconet establishment to response providing for a service request are detailed in the following:

- Hotspot inquiring and piconet establishment;

- The user sends a service request \( (R) \) to the hotspot.
- The hotspot broadcasts the OUUID provided by the user together with the request \((\text{OUUID}_R)\) to the piconet clients and selects the ones matching \(\text{OUUID}_R\);

- The hotspot sends a service request to the piconet clients with \(\text{OUUID}_R\);

- Each client replies with the corresponding OWL description of each resource it shares, which is classified with the previously selected \(\text{OUUID}_R\);

- The hotspot performs the matchmaking process between \(R\) and both each received OWL description and the ones related to the resources cached within the hotspot itself. Taking into account the matchmaking results, all the resources are ranked with respect to \(R\);

- The hotspot replies to the user.

Each resource retrieval session starts after the submission from master to slaves of the ontology identifier (OUUID) in order to select possible hosts suitable for requesting services. Each host processes the incoming request at SDP layer and it provides the hotspot with OWL descriptions of its shared resource. During matchmaking step the hotspot uses the semantic-based service MAMAS internally and computes the ranked list of the retrieved resource descriptions with respect to the request.

A simple use case will explain the system behaviour. Let us suppose we are visiting an art gallery, which classifies its work of art in a local knowledge base with an ontology (with OUUID=2017) represented in Figure 2.

\[
\begin{align*}
\text{Painting} & \sqsubseteq \text{WorkOfArt} \\
\text{Sculpture} & \sqsubseteq \text{WorkOfArt} \\
\text{Dadaism} & \sqsubseteq \text{Style} \\
\text{Futurism} & \sqsubseteq \text{Style} \\
\text{Symbolism} & \sqsubseteq \text{Style} \\
\text{Nature} & \sqsubseteq \text{Subject} \\
\text{Portrait} & \sqsubseteq \text{Subject} \\
\text{Scenery} & \sqsubseteq \text{Subject} \\
\text{PaintingTechnique} & \sqsubseteq \text{Technique} \\
\text{Bust} & \sqsubseteq \text{Sculpture} \\
\text{Statue} & \sqsubseteq \text{Sculpture} \\
\text{Landscape} & \sqsubseteq \text{Nature} \\
\text{Seascape} & \sqsubseteq \text{Nature} \\
\text{Naked} & \sqsubseteq \text{Portrait} \\
\text{SelfPortrait} & \sqsubseteq \text{Portrait} \\
\text{Cityscape} & \sqsubseteq \text{Scenery} \\
\text{Landscape} & \sqsubseteq \text{Scenery} \\
\text{Seascape} & \sqsubseteq \text{Scenery} \\
\text{Distemper} & \sqsubseteq \text{PaintingTechnique} \\
\text{Oil} & \sqsubseteq \text{PaintingTechnique} \\
\text{Pastel} & \sqsubseteq \text{PaintingTechnique} \\
\text{Watercolors} & \sqsubseteq \text{PaintingTechnique} \\
\text{Sculpture} & \sqsubseteq \neg \text{Painting} \\
\text{Style} & \sqsubseteq \neg \text{Subject} \\
\text{Statue} & \sqsubseteq \neg \text{Bust} \\
\text{Watercolors} & \sqsubseteq \neg \text{Oil}
\end{align*}
\]
For the sake of simplicity, in the example ontology in Figure 2 only subclass and disjoint relations are represented. Let us also suppose that a generic visitor (User2) is visiting the same museum and in particular the same room we are in. User2 is interested in dadaist style. S/He has previously downloaded on his/her mobile device a document file with more information concerning dadaist oils on canvas from the knowledge base of the consortium the art gallery is associated with. This KB classifies its contents by means of the previous OUUID=2017 ontology. Let us imagine that the hotspot KB is populated with the following individuals:

- LN League of nations classified as:
  - Painting ∩ ∀hasStyle.Dadaism ∩ ∀hasTechnique.Pastel
- FU Forme uniche nella continuità di spazio classified as:
  - Sculpture ∩ ∀hasStyle.Futurism
- MSV Mont Saint Victor classified as:
  - Painting ∩ ∀hasTechnique.Oil
- SP Self-portrait classified as:
  - WorkOfArt ∩ ∀hasSubject.SelfPortrait ∩ ∀hasStyle.Symbolism
- WCH Woman combing her hair classified as:
  - Statue ∩ ∀hasSubject.Naked

Moreover the document file on the User2 PDA is a specific document on painting Broyeuse de chocolat no.2 classified as: Painting ∩ ∀hasStyle.Dadaism ∩ ∀hasTechnique.Oil (we will indicate such resource with BC). Hence we can submit to the hotspot a semantic request for more info on oil dadaist paintings with landscape as subject.

The request will be formulated in DLs as \[ D = \text{Painting} \cap \forall\text{hasStyle.Dadaism} \cap \forall\text{hasTechnique.Oil} \cap \forall\text{hasSubject.Landscape} \] with respect to the ontology identified by the OUUID=2017.
The hotspot will search for OUUID=2017 and will find the User2 client plus the others hosts already known. In fact User2 is in the range of the hotspot and s/he exposes a service classified by the same ontology managed by the matchmaker. Then User2 sends the OWL description of his resource to the hotspot which calls the matchmaker module for rank computation.

In table 1 matchmaking results are presented. The second column shows whether the resource is compatible or not with D and, in case, the rankPotential computed result.

<table>
<thead>
<tr>
<th>resource name</th>
<th>compatibility (y/n)</th>
<th>score</th>
</tr>
</thead>
<tbody>
<tr>
<td>BC</td>
<td>Y</td>
<td>3</td>
</tr>
<tr>
<td>LN</td>
<td>Y</td>
<td>4</td>
</tr>
<tr>
<td>MSV</td>
<td>Y</td>
<td>4</td>
</tr>
<tr>
<td>SP</td>
<td>Y</td>
<td>9</td>
</tr>
<tr>
<td>WCH</td>
<td>N</td>
<td>-</td>
</tr>
<tr>
<td>FU</td>
<td>N</td>
<td>-</td>
</tr>
</tbody>
</table>

Table 1: matchmaking results

After having shared his resource file, User2 will receive a connection request from our PDA with our connection parameters and he can decide to establish a communication session with us.

Notice that the semantic distance of BC from D is the smaller one, then the system will recommend to the user this resource. Hence the ranked list returned by the hotspot is a strict indication for the user about best available services in the piconet in order of relevance. Nevertheless a user can choice or not a resource according to his personal preferences and his initial purposes.

4. Related Work

There is a widespread request for an increase of discovery features in wireless contexts like Bluetooth piconets. In [2] the need for discovery mechanisms more powerful than those of the original standard, inadequate for modern ubiquitous scenarios, was clearly pointed out for the first time. In recent years dynamic distributed systems have been developed adopting various technologies and for different purposes.

In [6] a Jini-based distributed agent framework is used in a hybrid agent-oriented/service-oriented approach, whereas in [13] semantic user profiles are introduced to increase matching level of services.
In [12] an example of collaborative environment is presented, where ontologies are used to infer new information about mobile users profile. We believe that existing semantic service discovery architectures have some limitations which make them unsuitable for a widespread use in ad-hoc networks.

In a mobile environment, where subjects on-line are continuously in evolution, modelling really peer to peer interaction calls for a common vocabulary to classify semantic descriptions of services. In fact two or more clients in the piconet who want share information must have a common way for describing them. Existing service discovery systems do not support a well defined common ontology infrastructure. In fact architectures like Jini allow to "capture" the ontology among services by means of mechanisms like Java classes which are difficult to be widely adapted. This limitation, as admitted in [6] and in [5], is due to the lack of shared ontology support.

In [5] it is assumed that a client request is described by means of the same ontology a service uses for describing itself. This assumption is fundamental because it restricts the discovery only to services classified in the same class. In [5] there is no mention of the technique to obtain this objective. Here we proposed a simple method for ontology matching prior to service discovery. The preliminary ontology matching grants a quick restriction of the available services only to those semantically suitable with the request.

Semantic service discovery via matchmaking in the Bluetooth framework was investigated in [2]. Also in this paper the issue of somehow ranking and proposing approximate matches in the absence of exact matches was discussed, but as in the previous papers no formal framework was given. Instead, a logical formulation is expected to allow devising correct algorithms to classify and rank matches.

In [13] a mobile environment is presented where semantic services are matched against semantic user profiles. In it, if there is no intersection between user interests and service offers, we conclude the user is not interested in the service. A complete and integrated solution for matching degree determination is absent.

Chen et al. in [6] present a hybrid approach, agent/service oriented, to perform dynamic service discovery in mobile environments based on Bluetooth-like devices. For such purpose they make use of Jini platform and enrich them with a distributed agent layer. In fact Jini Lookup Service does not solve some important service discovery problems. Therefore the provided framework may require too large computational resources to be easily adapted to a really mobile scenario. The agent software layer should perform semantic level service discovery or inexact matching, but it is computationally heavy to run on an handheld device. Hence, as admitted by the authors, a proxy agent which resides in a computer on the wired side is needed. The mobile devices are responsible only for rendering GUI display. Furthermore there is no mention to the solution of inexact matching problem. No formal methods to determine approximate matches are outlined. The system is strictly client/server. It does
not allow to implement a real P2P scenario. The sharing of resources managed by a network client with other mobile hosts is not expected, then it can be obtained only by loading shared services into a local database and by registering them into Jini Lookup Service. This is a big restriction because it makes impossible a direct communication among two or more peers in the ad-hoc network bypassing Jini Lookup or any broker agent.

The spontaneous and occasional collaboration among mobile users is investigated in [12]. Here a collaborative context is described where a matchmaking service communicates with a localization service which discovers all the MAC addresses of the mobile devices in the environment. Matchmaking service compares the user profiles associated to those MAC. Thus we do not have a close integration between discovery phase and matching phase. A merging of the proposed matchmaking system in a complex semantic service discovery architecture is still lacking.

5. Conclusions and Future Work

In this paper we have exploited KR techniques and technologies to enrich the capabilities of the SDP in Bluetooth. Adding information modelled using languages with a well-defined formal semantics, in particular DLs, we have made the SDP able to manage also non-exact match between the requested service (resource) and the offered ones. In our approach, a semantic layer has been added to the existing standard SDP for Bluetooth in order to perform a semantic-based matchmaking process. The system we present adopts these enhanced features and uses them to provide various user-oriented services which benefit of the KR approach. Under development is a Natural Language based querying process facilitator [8] involving both the hotspot and the mobile clients.

6. References


