

CARISMA: Context-Aware Reflective Middleware System for Mobile Applications

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Abstract—Mobile devices, such as mobile phones and personal digital assistants, have gained wide-spread popularity. These devices will increasingly be networked, thus enabling the construction of distributed applications that have to adapt to changes in context, such as variations in network bandwidth, battery power, connectivity, reachability of services and hosts, etc. In this paper, we describe CARISMA, a mobile computing middleware which exploits the principle of reflection to enhance the construction of adaptive and context-aware mobile applications. The middleware provides software engineers with primitives to describe how context changes should be handled using policies. These policies may conflict. We classify the different types of conflicts that may arise in mobile computing and argue that conflicts cannot be resolved statically at the time applications are designed, but, rather, need to be resolved at execution time. We demonstrate a method by which policy conflicts can be handled; this method uses a microeconomic approach that relies on a particular type of sealed-bid auction. We describe how this method is implemented in the CARISMA middleware architecture and sketch a distributed context-aware application for mobile devices to illustrate how the method works in practice. We show, by way of a systematic performance evaluation, that conflict resolution does not imply undue overheads, before comparing our research to related work and concluding the paper.

Index Terms—Middleware, mobile computing, reflection, context-awareness, conflict resolution, game theory, quality of service.

1 INTRODUCTION

MOBILE computing devices, such as palmtop computers, mobile phones, personal digital assistants (PDAs), and digital cameras have gained wide-spread popularity. These devices will increasingly be networked and software development kits are available that can be used by third parties to develop applications [1].

Even though devices and networking capabilities are becoming increasingly powerful, the design of mobile applications will continue to be constrained by physical limitations. Mobile devices will continue to be battery-dependent and users are likely to be reluctant to carry heavy-weight devices. Wide-area networking capabilities will continue to be based on communication with base-stations, with fluctuations in bandwidth depending on physical location. In order to provide acceptable quality of service to their users and, consequently, improve user satisfaction of the system, applications have to be *context-aware* [2] and able to adapt to context changes, such as variations in network bandwidth, exhaustion of battery power, or reachability of services on other devices. This would require application engineers, for example, to periodically query heterogeneous physical sensors, in order to get updated context information, to detect context configurations of interest to the application, and adapt

accordingly; however, doing so would be extremely tedious and error-prone.

In order to ease the development of context-aware applications, middleware layered between the network operating system and the application have to provide application engineers with powerful abstractions and mechanisms that relieve them from dealing with low-level details. For example, applications must be able to specify, in a uniform way, which resources they are interested and which behaviors to adopt in particular contexts. The middleware, then, maintains, on behalf of the applications, updated context information, detects changes of interest to them and reacts accordingly.

In the past decade, the development of distributed applications for wired systems has been greatly enhanced by middleware systems that succeeded in facilitating the communication between distributed components. Traditional middleware systems (e.g., CORBA, Java/RMI, MQSeries) provide application engineers with communication abstractions that relieve them from dealing with, for example, the location of distributed components, network failures and hardware heterogeneity (e.g., for marshaling/unmarshaling of parameters). These middleware are based on the principle of transparency [3], [4]: Implementation details are hidden from both users and application designers and are encapsulated inside the middleware itself, so that the distributed system appears to application developers as a single integrated computing facility.

Although having proven successful in supporting the construction of traditional distributed systems, we argue that transparency cannot be used as the guiding principle to develop the new abstractions and mechanisms needed by mobile computing middleware to foster the development of

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context-aware applications. By providing transparency, middleware must take decisions on behalf of the application; this is inevitably done using built-in mechanisms and policies that cater for the common case rather than the high levels of dynamicity and heterogeneity intrinsic in mobile environments. Applications, instead, may have valuable information that could enable the middleware to execute more efficiently in different contexts.

We argue that *reflection* [5] offers significant advantages for building mobile computing middleware. A reflective system may modify its own behavior by means of inspection (i.e., the internal behavior of the system is exposed) and/or adaptation. (i.e., the internal behavior of a system can be dynamically changed). For example, applications may dynamically alter the set of resources that middleware monitors on their behalf, the context configurations they are interested into, and the behaviors they want to adhere to. However, while doing so, applications may introduce ambiguities, contradictions, and other logical inconsistencies. For example, contradictory behaviors may be requested by the same application to react to a particular context change, or cooperating applications may not agree on a common behavior to be applied. We refer to these inconsistencies as *conflicts*.

The novel contribution of this paper is the design, formalization, and evaluation of new abstractions and mechanisms that, embedded in a mobile computing middleware software layer, facilitate the development of context-aware applications. In particular, we exploit the principle of reflection to achieve dynamic adaptation to context changes: We offer applications an abstraction of the middleware as a dynamically customizable service provider, where customization takes place by means of metadata, which encode middleware behavior to answer application service requests in various contexts. Through reflection, the metainformation can be changed and, therefore, the middleware behavior tuned, with the risk, however, of incurring conflicts. We have designed a microeconomic approach to *conflict resolution* that relies on a particular type of sealed-bid auction. Our approach treats a distributed mobile system as an “economy,” where applications compete to have the middleware deliver the quality-of-service they desire. The mobile computing middleware plays the role of an auctioneer, collecting bids from applications and delivering services with the QoS requested by the successful one. We show why our auctioning mechanism is particularly useful in a mobile setting and that it achieves fair conflict resolution.

The remainder of the paper is structured as follows: Section 2 describes our reflective middleware model; Section 3 introduces the issue of conflicts in a mobile setting, provides a classification of the types of conflicts we deal with, and illustrates them using an example of context-aware application. In Section 4, we formalize the microeconomic mechanism we propose to solve these conflicts and illustrate a comprehensive example, based on the application described before, to clarify how the model works in practice. Section 5 evaluates our model in terms of

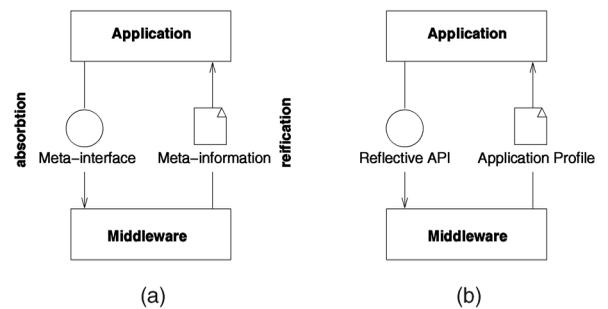


Fig. 1. The reflective process.

usability and performance; Section 6 compares our approach to related work and, finally, Section 7 concludes the paper and identifies possible future work.

2 THE REFLECTIVE MODEL

Mobile applications execute in an extremely dynamic context: Location changes all the time while moving around with our portable device and, so, the services and devices in reach; local resource availability varies quickly as well, such as memory availability, bandwidth, and battery power. In order to provide reasonable quality-of-service to their users, applications have to be context-aware.

We argue that *reflection* [5] is a powerful means to build mobile computing middleware that supports the development of context-aware applications. The key to the approach is to make some aspects of the internal representation of the middleware explicit and, hence, accessible from the application, through a process called *reification*. Applications are then allowed to dynamically inspect middleware behavior (introspection) and also to dynamically change it (adaptation) by means of a metainterface that enables runtime modification of the internal representation previously made explicit. The process where some aspects of the system are altered or overridden is called *absorption*. The whole process is depicted in Fig. 1a.

CARISMA, a project carried out at University College London, is a middleware model that exploits reflection to enable context-aware interactions between mobile applications. In our model [6], the middleware is in charge of maintaining a valid representation of the execution context by directly interacting with the underlying network operating system. By context, we mean everything that can influence the behavior of an application from resources within the device, such as memory, battery power, screen size, and processing power, to resources outside the physical device, such as bandwidth, network connection, location, and other hosts within reach, to application-defined resources, such as user activity and mood.

Applications may require some services to be delivered in different ways (using different policies) when requested in different context. For example, a messaging application may wish to send messages in plain text when bandwidth is high, while exchanging compressed messages when bandwidth is low.

```

serviceList ::= service serviceList |  $\epsilon$ 
service ::= sname policyList
policyList ::= policy policyList | policy
policy ::= pname contextList
contextList ::= context contextList | context
context ::= resourceList
resourceList ::= resource resourceList |  $\epsilon$ 
resource ::= rname oname valueList
valueList ::= value valueList |  $\epsilon$ 

```

Fig. 2. Application profile abstract syntax. $sname \in S$, $pname \in P$, $rname \in R$, where $S, P, R \subset \Sigma^*$, respectively, are the sets of all valid service/policy/resource names over our alphabet Σ . $value \in V$, where V is the set of all possible values of resources in R (e.g., IP addresses for hosts in reach, etc.). $oname \in O$, where O is the set of all valid operator names that can be applied to values of monitorable resources (e.g., *equals*, *lessThan*).

To enhance the development of context-aware applications, CARISMA provides application engineers with an abstraction of the middleware as a customizable service provider. In particular, the behavior of the middleware with respect to a specific application is described as a set of associations between the *services* that the middleware customises, the *policies* that can be applied to deliver the services, and the *context configurations* that must hold in order for a policy to be applied. In the example above, an association is defined between the “messaging” service, the “plain message” policy, and a context where the resource “bandwidth” is high, and another one between the same “messaging” service, the “compressed message” policy, and a context where “bandwidth” is low. The behavior of the middleware with respect to a particular application is reified in what we call an *application profile*, as shown in Fig. 1b.¹ Figs. 2 and 3 show respectively the profile abstract syntax and an example of a customized service encoded using this syntax.

Profiles are passed down to the middleware; each time a service is invoked, the middleware consults the profile of the application that requests it, queries the status of the resources of interest to the application itself, as declared in the profile, and determines which policy can be applied in the current context, thus relieving the application from performing these steps. Our model assumes that the behavior of the middleware with respect to a particular service is determined, at any time, by one and only one policy; that is, a service cannot be delivered using a combination of different policies. More policies can logically be combined; for example, the messaging service can be provided with a “compressed message” policy, that is a logical combination of two separate policies, “compress” and “send.” However, we regard the combined policy as a new one and, in the profile, we will refer to this new policy, not to the sequential execution of two distinct policies.

1. Our reflective middleware model assumes a single user for each mobile device, though there may be many applications running simultaneously on that device, hence, in our model, on the same middleware instance (this assumption is reasonable for portable devices, such as PDAs and mobile phones).

```

messagingService
  plainMsg
    bandwidth > 40%
  compressedMsg
    bandwidth < 40%

```

Fig. 3. Customization of the messaging service.

As both the user needs and the context change quite frequently (e.g., due to movement of the device to a different location), we cannot expect application designers to foresee all possible configurations. Through a reflective API (Fig. 1b), applications can dynamically inspect the content of their profile (i.e., the current configuration) and alter it by adding, deleting, and updating the associations previously encoded. As the behavior of the middleware is dictated by the associations encoded in the application profiles, changing this information means dynamically affecting middleware behavior (i.e., reconfiguration of the system). If we consider once again the messaging example, an application may add an association to its profile requesting the execution of the messaging service using an “encrypted message” policy, in order to achieve privacy of information, when both battery and bandwidth availability are high.

A default profile exists for every application, where each service that the middleware delivers (to that application) is associated to exactly one policy, regardless of context. It is up to the application to decide whether and when to exploit the power of reflection to alter the information here encoded, that is, to customize middleware behavior in order to achieve better quality of service.

So far, we have focused our discussion on the interaction between middleware and applications, leaving the end-users of the system behind the scene. As Fig. 4 illustrates, the middleware provides applications a reflective API (i.e., metainterface) that they can exploit to inspect and alter middleware behavior, as encoded in application profiles. The target users of our middleware model are therefore application developers. In customizing middleware behavior, however, end-user requirements and expectations must be taken into consideration. We therefore expect applications built on top of CARISMA to provide end-users with a user interface through which end-user preferences can be captured and used by applications to encode profiles.

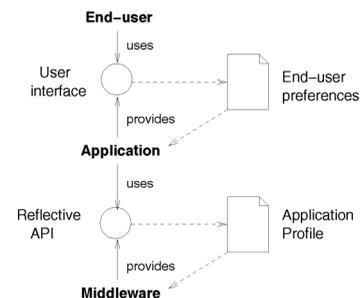


Fig. 4. Roles and responsibilities in the reflective process.

In this paper, we are mainly concerned with the interaction between middleware and applications, thus leaving the issue of gathering user preferences and synthesizing them in application profiles for future work; however, we will provide in Section 5.2 some insights on the complexity of doing so and on the amount of work required from the user to teach the system to behave according to his/her own expectations.

3 DEALING WITH CONFLICTS

The model presented above allows applications to control the behavior of the middleware based on current user needs and context. This is achieved by means of application profiles that can be dynamically changed through a reflective API. Although a middleware based on this model supports the development of context-aware applications, it also opens the door to conflicts. In our model, a conflict exists when different policies can be used in the same context to deliver a service, so that the middleware does not know which one to apply (note that, we made the assumption that a service can be delivered using only one policy at a time). Reflection gives applications the “intelligence” that transparency takes away in traditional middleware systems. Applications, however, may not be smart enough to cope with the new power, and may produce profiles that lead to conflicts. In particular, when setting up application profiles, the following two basic kinds of conflicts may be created.

Intraprofile conflict: A conflict exists inside the profile of an application running on a particular device. This class identifies conflicts that are *local* to a middleware instance.

Interprofile conflict: A conflict exists between the profiles of applications running on different devices. This class identifies conflicts that are *distributed* among various middleware instances. A particular case of interprofile conflict happens when applications run on the same device (i.e., on the same middleware instance); we refer to this situation as an *N-on-1* (i.e., *N* applications on one device) *conflict*.

In order to understand how these types of conflicts arise, we sketch a conference application that is representative of the class of context-aware mobile applications that would benefit from our reflective middleware model; we then discuss the requirements that a conflict resolution mechanism must meet, before presenting the details of the mechanism we have designed. At this stage, we are not interested in implementation details (in particular, in the language used to encode profiles); we therefore use the abstract syntax illustrated before to discuss the following examples.

3.1 Conference Application

Let us imagine a researcher Alice traveling to a conference with her own PDA. When arriving at the conference location, she is provided with a Conference Application to be installed on her portable device that, based on a wireless network infrastructure, allows attendees to access the proceedings electronically, browse through the technical

```

talkReminder
  soundAlert
    location = outdoor
  vibraAlert
    location = conferenceRoom
  silentAlert
    userFocus = on

```

Fig. 5. Example of local (intraprofile) conflict.

program, select the talks they wish to attend and be alerted of the selected ones 10 minutes before they start, and exchange messages with other attendees. These services may have to be delivered in different ways when requested in different contexts, in order to meet the user’s needs. Let us consider, in particular, the talk reminder service and the messaging service; through our reflective middleware model, Alice’s preferences can be taken into account and used to generate the following associations.

Reminder of the next talk. The reminder functionality of the system can capture user attention through one of the following policies: `soundAlert`, particularly useful to capture user attention in noisy and open air places; `vibraAlert`, to capture user attention without disturbing anybody else (e.g., while attending a talk); and `silentAlert`, to remind the user of the next talk through a blinking message, for example, while the user is actively using the portable device. The `talkReminder` is an example of local service, as it does not require the cooperation of any other party. Let us consider, for example, the encoding shown in Fig. 5 and let us assume that the service is requested when Alice is attending a talk (i.e., `location = conferenceRoom`) and using her PDA to take notes at the same time (i.e., `userFocus = on`). The middleware checks which policy should be applied and determines that more than one policy suits the current context (i.e., `vibraAlert` and `silentAlert`). As we made the assumption that each service is delivered using one and only one policy at a time, the middleware is unable to choose which of the context-suitable policies to apply.² This is an example of *intraprofile conflict*.

Exchange of messages. Attendees can exchange messages using any of the following policies: `charMsg` that delivers one character at a time, `plainMsg` to exchange messages in plain, `compressedMsg` to exchange compressed messages, and `encryptedMsg` to send encrypted messages.

The `messagingService` is an example of peer-to-peer service, where any number of peers may participate in the delivery of the service. In order for the service to be delivered, all the communicating peers have to agree on a common policy to be applied. Let us assume, for example, that Alice, Bob, and Claire are willing to exchange messages; let us also assume that their profiles are the ones illustrated in Fig. 6. Note that no context information is

2. Note that, by removing this assumption, we do not avoid the issue of conflicts, we just need to formulate it under different terms. In particular, conflicts would appear as different *sets* of policies enabled at the same time; in this case, we should consider whether the order in which policies appear in a profile is relevant, or whether their execution is commutative.

```

% Alice
messagingService
  plainMsg
    battery < 40%
  encryptedMsg
    battery > 40%

% Bob
messagingService
  plainMsg

% Claire
messagingService
  plainMsg
    bandwidth > 50%
  compressedMsg
    bandwidth < 50%

```

Fig. 6. Example of distributed (interprofile) conflict.

associated to the `plainMsg` policy of Bob's profile: This means that this policy is always enabled regardless of current context. At any time, users may change their preferences through the user interface that the conference application provides; the application, in turn, dynamically updates the meta-information encoded in their profiles, in order to take the new preferences into account.

If the `messagingService` is requested when battery availability is below 40 percent on Alice's PDA and Claire's bandwidth is greater than 50 percent, they all agree on the `plainMsg` policy to be applied; but, what if Alice's battery is greater than 40 percent, or if Claire's bandwidth is lower than 50 percent? This is an example of *interprofile conflict*.

3.2 Requirements

Whenever a service that incorporates a conflict, either intra- or interprofile, is requested, a conflict resolution mechanism has to be run to solve the conflict and find out which policy to use to deliver the service, otherwise applications cannot execute. In designing such mechanism, the following requirements have to be considered.

Dynamicity. Neither intra- nor interprofile conflicts can be detected and resolved statically; that is, at the time the profile is written by the application and passed down to the middleware. In case of intraprofile conflict, a possible static approach would require us to check whether there is any intersection between the different contexts of the policies associated to each service. Due to the complex nature of context (the number of monitored resources may be large), a static conflict analysis would produce an explosion in the context information that must be checked, and would require a consumption of resources (especially in terms of battery, memory, and processing power) that portable devices cannot bear. Providing the conflict resolution as an external service on a powerful machine that is contacted on-demand is not feasible either, as this would require persistent connectivity that in mobile settings cannot be taken for granted. As for interprofile conflicts, the situation is even worse; mobile devices connect opportunistically and sporadically. We cannot foresee which devices are going to be encountered and, even so, we cannot assume that all of them will be connected and in reach at the time a profile is modified; this means that the middleware cannot statically check whether the new configuration is conflict-free. Even assuming that this distributed check could be statically performed, it would not be worth the effort, as we would find many more potential conflicts than what we would

actually need, as conflicts manifest themselves only with respect to the particular context in which the service is requested and the profiles of the participating peers. As a consequence, a dynamic solution is needed: Conflicts may exist inside or among profiles, but both applications and middleware can live with these conflicts until a service which incorporates a conflict is invoked.

Simplicity. The conflict resolution mechanism must be simple in the sense that it must not consume resources that are already scarce on a mobile device. Only a low computation and communication overhead should be imposed, even if this may occasionally prevent from an optimal solution to the conflict to be found.

Customization. Middleware cannot choose how to solve conflicts independently of the applications that requested the conflicting service, as only the applications know how much they value the execution of the various policies. On one hand, we do not want applications to be questioned each time a conflict is detected; that is, middleware should be in charge of carrying out the conflict resolution process in an automatic way as much as possible. On the other hand, it must be possible for the applications to customize the conflict resolution mechanism, thus influencing which policy is chosen and applied and which others are discarded.

In the following section, we formally describe a conflict resolution mechanism that meets these requirements.

4 MICROECONOMIC MECHANISM

When applications participating in the delivery of a service cannot agree on which policy must be applied, a *dynamic* conflict resolution scheme is necessary to resolve the dispute. The conflict resolution mechanism we propose is based on microeconomic techniques [7]. The motivating idea is that a mobile distributed system can be seen as an *economy*, where a set of *consumers* must make a collective choice over a set of alternative *goods*. Goods represent the various policies that can be used to deliver a service (not the resources needed to apply a policy); for example, policies "plainMsg," "encryptedMsg," and "compressedMsg" are the goods associated to service "messagingService." Consumers are applications seeking to achieve their own goals, that is, to have the middleware delivering a service using the policy that provides the best quality of service, according to application-specific preferences.

Simple schemes include, for example, priority assignment or per capita distribution. However, those do not suit situations where participation in exchange of goods is voluntary on the part of all parties (i.e., the applications), so that action requires a consensus and mutual perception of benefit. A better scheme would use an *auction protocol*. Auctions allow parties to make decisions independently, on the basis of private state, revealing only offers and acceptance of the offers made by others. Applications may vary greatly in their preferences and decision processes. An auction permits greater degrees of heterogeneity than simpler schemes.

The question we have to answer next is which auction protocol to use. This is known in microeconomic theory as a

mechanism design problem [8]. A *protocol*, or mechanism, consists of a set of rules that govern interactions, by which agents (i.e., our applications) will come to an agreement. It constraints the deals that can be made, as well as the offers that are allowed. We argue that the auction protocol we have designed [9] can be successfully applied in a mobile setting, where the requirements listed in Section 3.2 must be satisfied.

4.1 The Protocol

The rules of our auction are very simple: Given a setting with N agents that must make a collective choice from a set of P possible alternatives, each agent submits a single sealed bid for each element in P . The auctioneer collects the bids and selects the alternative in P that maximizes social welfare, that is, the alternative with the highest sum of bids received. Each agent then pays the auctioneer an amount of money that is proportional to the bid they placed on the winning alternative.

In our scenario, the role of the auctioneer is played by the middleware, which we assume is a trusted entity whose code and behavior cannot be interfered with. Applications are the agents, and the good they are competing for is the execution of the policy they value most, among a set of alternatives that correspond to the policies that can be applied in a particular context to deliver a service. As previously said, the aim of the middleware is not to select the policy that received the highest bid (i.e., the one that maximizes the selling price), but, rather, the policy that satisfies the largest number of applications involved in the conflict. In our scenarios, in fact, applications are participating in the delivery of the same service, rather than competing for it (i.e., the service will be delivered to all of them, not only to one or some of them). In these collaborative, or at least compromise, scenarios, a solution that satisfies the total benefit of all the applications is preferred to one that maximizes the revenue of a single one.

Our auction has been inspired by traditional sealed bid auctions (e.g., first-price and second-price sealed bid auction [10]). Unlike ascending bid auctions, such as the standard English auction [11], where the auctioneer, adopting a possibly long iterative process, continuously raises the price of the good until only one bidder is willing to meet the price called, sealed bid auctions consist of a one-step bid that cuts down the computation and communication costs when the auction is distributed over space and time, as in our mobile setting. This meets our requirement of *simplicity*. We will show in Section 4.2 how customization is met by our auctioning mechanism.

In the following, we formalize the steps of our auctioning mechanism. We do not discuss here how coordination among different middleware instances takes place; details about the algorithm that implements this coordination can be found in Section 5.1. To avoid confusion between an application (which may exist on different devices) and an application instance (which runs on a particular device), we will identify an application instance and the device it is executing on as a “peer.” Peers are partners in the communication process. We call PEER the set of all possible peers. Under these assumptions, the auctioning process can be formalized as follows.

\mathcal{F}	:	$service \rightarrow E \rightarrow \wp(P)$
$\mathcal{F}[\![sn\ pList]\!]_e$	=	$\mathcal{F}[\![pList]\!]_e$
$\mathcal{F}[\![policy\ pList]\!]_e$	=	$\mathcal{F}[\![policy]\!]_e \cup \mathcal{F}[\![pList]\!]_e$
$\mathcal{F}[\![pn\ cList]\!]_e$	=	$\{pn\}$ if $valid[\![cList]\!]_e = \top$ \emptyset if $valid[\![cList]\!]_e = \perp$
$valid$:	$cList \rightarrow E \rightarrow bool$
$valid[\![context\ cList]\!]_e$	=	$valid[\![context]\!]_e \vee valid[\![cList]\!]_e$
$valid[\![context]\!]_e$	=	$valid[\![rList]\!]_e$
$valid[\![resource\ rList]\!]_e$	=	$valid[\![resource]\!]_e \wedge valid[\![rList]\!]_e$
$valid[\![rn\ on\ vList]\!]_e$	=	$eval((rn, on, vList), e)$
$valid[\![\epsilon]\!]_e$	=	\top

Fig. 7. Computation of locally enabled policies. Given a service specification $service$, the semantic function \mathcal{F} evaluates, in current context $e \in E$, the set of locally enabled policies. $E = \wp(R \times V)$ represents the set of all possible execution contexts (e.g., $\{(Memory, 8), (Battery, 4)\}$); $eval$ is a Boolean function that returns *true* if the value of resource rn in the execution context e is among the values obtained by applying the operator on to $vList$ (e.g., $eval((Memory, inBetween, [2, 7]), \{(Memory, 6)\}) = \top$).

Initialization. As part of an initialization process, for every peer $peer_i$, $i \in [1, N]$, a utility function $u_i : P \rightarrow \mathbb{R}^+$ that represents the user’s goals (e.g., minimization of consumption of resources, maximization of quality of service, etc.) can be determined. Peers use their utility function to specify how much they value the use of a policy $p_j \in P$ during an auction, that is, $u_i(p_j) = u_{i,j}$. Each peer is also assigned a quota q_i by the middleware. The quota q_i represents the maximum amount of money that $peer_i$ can bid during a bidding process; that is, the bid placed by peer $peer_i$ on policy p_j is a number $b_{i,j} = \min\{u_{i,j}, q_i\}$.

Service Request. Whenever an application requires the middleware to execute a service, a command like the one illustrated below is issued:

```
command ::= sname peerList
peerList ::= peer peerList | peer
```

being $sname \in S$ the name of the requested service and $peerList$ the set of peers involved in the service execution.

Assuming that service $sname$ requires the cooperation of $n \leq N$ peers, each peer (or, better, the middleware instance operating on the device of the peer) computes P_i as the set of policies that the above running application instance A_i has associated to service $sname$ in its profile and that can be applied in the current context (i.e., according to current resource availability). More formally, P_i can be defined as follows:

$$P_i = \mathcal{F}[\![serv(sname, peer_i)]\!]_{\mathcal{E}nv(peer_i)},$$

where \mathcal{F} is the semantic function defined in Fig. 7, $serv : S \times PEER \rightarrow service$ a function that, given a service name and a peer, returns the corresponding service specification, and $\mathcal{E}nv : PEER \rightarrow E$ a function that computes the current execution environment of a peer.

Computation of the solution set. Middleware instances then cooperate to compute the *solution set* P^* , that is, the set of policies that all peers involved in the execution of the service have agreed upon:

$$\begin{array}{l}
\mathcal{I} : S \rightarrow \wp(\text{PEER}) \rightarrow \wp(\text{P}) \\
\mathcal{I}[\![sn]\!]_{\{peer\ pList\}} = \mathcal{I}[\![sn]\!]_{\{peer\}} \cap \mathcal{I}[\![sn]\!]_{\{pList\}} \\
\mathcal{I}[\![sn]\!]_{\{peer\}} = \mathcal{F}[\![serv(sn, peer)]\!]_{\mathcal{E}_{nv(peer)}}
\end{array}$$

Fig. 8. Computation of the solution set. Given a service name sn , the semantic function \mathcal{I} computes the set of policies that all peers involved in the service delivery agree.

$$P^* = \mathcal{I}[\![sname]\!]_{\{peer_1, \dots, peer_n\}},$$

where \mathcal{I} is the semantic function described in Fig. 8.

If the cardinality of P^* is zero, that is, the solution set is empty, a conflict exists that cannot be solved, as peers do not agree on a common policy to be applied; the conflict resolution process is terminated with a failure and peers are notified. If the cardinality is exactly 1, there is an agreement on the policy to apply (i.e., there is no conflict). Finally, if the cardinality is greater than 1, there is a conflict that can be resolved using one of the policies in P^* . In this case, the auctioning process proceeds as below, to decide which of these policies should be applied.

Computation of Bids. For every peer $peer_i$ participating in the communication process and for every agreed policy $p_j \in P^*$, $j \in [1, m]$, a bid $b_{i,j}$ is computed, based on the peer utility function u_i and quota q_i . Unlike “human” auctions, we make the assumption that all peers participating in a bidding process bid a price, that is, they cannot refuse to bid. Middleware instances of bidding peers exchange the bids they have received, ending up with a merged set of tuples \mathcal{B}^* specifying how much each peer values the use of each agreed policy:

$$B^* = \mathcal{B}[\![\{p_1, \dots, p_m\}]\!]_{\{peer_1, \dots, peer_n\}},$$

where \mathcal{B} is the semantic function shown in Fig. 9.

Election of the Winner. From the set B^* , middleware instances participating in the conflict resolution process select the winning policy p_j as the one with the highest sum of the bids placed:

$$p_j = \mathcal{W}[\![B^*]\!],$$

where \mathcal{W} is the semantic function defined in Fig. 10; as shown there, each peer also pays an amount of money that is proportional to the “added” benefit obtained by applying the winning policy over the other peers. To understand how the payment is split, let us consider three peers x , y and z ,

$$\begin{array}{l}
\mathcal{B} : \wp(\text{P}) \rightarrow \wp(\text{PEER}) \rightarrow \wp(\text{P} \times \text{PEER} \times \mathbb{R}^+) \\
\mathcal{B}[\![\{p_1, \dots, p_m\}]\!]_{\{peer\ pList\}} = \mathcal{B}[\![\{p_1, \dots, p_m\}]\!]_{\{peer\}} \cup \\
\mathcal{B}[\![\{p_1, \dots, p_m\}]\!]_{\{pList\}} \\
\mathcal{B}[\![\{p_1, \dots, p_m\}]\!]_{\{peer\}} = \bigcup_{j=1}^m \{(p_j, peer, \\
\min\{q_{peer}, u_{peer,j}\})\} \\
\mathcal{B}[\![\{p}\]\!]_{\{pList\}} = \{(p, -, 0)\} \text{ No conflict} \\
\mathcal{B}[\![\emptyset]\!]_{\{pList\}} = \emptyset \text{ No agreement}
\end{array}$$

Fig. 9. Computation of bids. Given the set of agreed policies, and the list of participating peers, the semantic function \mathcal{B} associates a bid to each couple (policy, peer).

$$\begin{array}{l}
\mathcal{W} : \wp(\text{P} \times \text{PEER} \times \mathbb{R}^+) \rightarrow \text{P} \\
\mathcal{W}[\![\{(p_j, peer_i, b_{i,j})\}]\!] = p_j \mid p_j \in \{\pi_1(p_j, peer_i, b_{i,j}), \\
\forall i \in [1, n], j \in [1, m]\}] \\
\wedge \sum_{i=1}^n \pi_3(p_j, peer_i, b_{i,j}) = \\
\max_{j \in [1, m]} \sum_{i=1}^n \pi_3(p_j, peer_i, b_{i,j}) \\
\wedge \text{pay}(q_{mw}(i), f_i, q_i), \forall i \in [1, n] \\
\mathcal{W}[\![\{(p, -, 0)\}]\!] = p \text{ No conflict} \\
\mathcal{W}[\![\emptyset]\!] = \epsilon \text{ No agreement} \\
f_i = \begin{cases} 0 \text{ if } \forall k \in [1, n] \\ \pi_3(p_j, peer_k, b_{k,j}) = \max_{j \in [1, m]} \pi_3(p_j, peer_k, b_{k,j}) \\ \sum_{l \in \{s \mid s \in [1, n] \wedge b_{s,j} \leq b_{i,j}\}} \\ \frac{b_{l,j} - \max(\{b_{s,j} \mid b_{s,j} < b_{l,j}, s \in [1, n]\} \cup \{b_{min,j}\})}{\#\{b_{s,j} \mid b_{s,j} \geq b_{l,j}, s \in [1, n]\} * \#\{b_{s,j} \mid b_{s,j} = b_{l,j}, s \in [1, n]\}}, \\ b_{min,j} = \min\{b_{i,j}, i \in [1, n]\} \text{ otherwise} \end{cases}
\end{array}$$

Fig. 10. Election of the winning policy. $\pi_i(a_1, a_2, \dots, a_n) = a_i$ projects a tuple onto the i th value; $\#\{a_1, a_2, \dots, a_n\} = n$ computes the cardinality of a set; $q_{mw}(i)$ retrieves the quota of the middleware on top of which peer $peer_i$ is executing; $\text{pay}(q_1, x, q_2) = (q_1 + x, q_2 - x)$ both increases the middleware quota q_1 , and decreases the peer quota q_2 , of the specified amount x .

who bid $b_x < b_y < b_z$, respectively, on a winning policy p . Applying p gives an equal benefit of b_x to each peer; moreover, y and z share an added benefit of $b_y - b_x$ over x , and z enjoys an extra benefit equal to $b_z - b_y$ over both x and y . Our payment scheme demands that x , y , and z pay 0, $(b_y - b_x)/2$, and $(b_y - b_x)/2 + (b_z - b_y)/1$, respectively. Note that, if the winning policy is the one that has been valued most by all peers (i.e., $b_x = \max_i b_{i,x}$, $b_y = \max_i b_{i,y}$, $b_z = \max_i b_{i,z}$), then no payment is demanded, as there was no real conflict to be solved. Note also that, in case of intraprofile conflicts, the payment is always zero, as the winning policy is never “imposed” on anyone, that is, there is no added benefit over anyone. The rationale for this payment scheme is that applications are not paying for the resources they use when applying a policy, but, rather, for the (added) quality-of-service level they get from it. We assume that ties are broken by selecting a policy randomly (i.e., a k -way tie is decided by flipping a “ k -sided coin,” where each policy is chosen with probability $1/k$).

If a service $sname$ is requested which requires the cooperation of a set of peer $peerList$, then the whole conflict resolution mechanism can be summarized as follows:

$$\mathcal{G} : \text{command} \rightarrow \text{P}$$

$$\mathcal{G}[\![sname\ peerList]\!] = \mathcal{W}[\![\mathcal{B}[\![\mathcal{I}[\![sname]\!]_{\{peerList\}}]\!]_{\{peerList\}}]\!].$$

A service request may then produce one the following two results:

- $\mathcal{G}[\![sname\ peerList]\!] = pname$: service $sname$ is delivered using policy $pname$ (either because all peers agreed on the policy, or because $pname$ was the policy selected during a conflict resolution process);
- $\mathcal{G}[\![sname\ peerList]\!] = \epsilon$: the service request fails as no policy can be found that is both agreed on by all peers and valid in the current context.

$ufunction$	$::=$	$addendList$
$addendList$	$::=$	$addend\ addendList \mid addend$
$addend$	$::=$	$cb_name\ weight$

Fig. 11. Utility function abstract syntax.

The auctioning mechanism has been described in the general situation where there are different applications running on different hosts (interprofile conflict). N -on-1 conflicts are detected and solved in the same way as interprofile conflicts. However, as the application instances involved are running on the same host (i.e., in our model, on the same middleware instance), no communication overhead is required and the solution set P^* can be computed locally. Intraprofile conflicts can be considered a degeneration of interprofile conflicts, where the number n of bidders is 1 and the solution set coincides with P_1 (i.e., the set of policies that can be applied in the current execution context according to $peer_1$ application profile). The auction proceeds as described above, selecting the policy that maximizes $peer_1$ utility, without communication costs.

Once a conflict has been detected and resolved using the auctioning mechanism presented above, no further action is taken. The conflict cannot be removed as it is usually not local to a profile but distributed among the profiles of different peers. If the peers involved change, or if the context changes, there may be no conflict at all. Also, we assume that each auction is carried out in isolation: We cannot assume that next time the same conflict arises, the winning policy will be the same one, as the result depends on current peer quotas, utility functions, and application profiles. Therefore, each conflict resolution process stands alone.

There are few questions that need to be answered about the process described above; in particular, how is an utility function defined, and how is the quota managed by middleware? We answer these questions in the following sections.

4.2 Utility Function

Whenever a conflict is detected, either inside a profile (intraprofile conflict) or among various profiles (interprofile conflict), user goals, such as privacy of information for the messaging service, must be taken into account. In other words, users should be allowed to influence the conflict resolution process operated by the middleware as they are the only ones who know what their goals are at the moment, and how different outcomes are valued.

Utility functions serve this purpose. A utility function u_i translates user goals with respect to peer $peer_i$ into a value $u_{i,j}$, that represents the price the user is currently willing to pay to have policy p_j applied, that is, to see its goals fulfilled. The following holds:

$$u_{i,j} \geq 0, \quad \forall i \in [1, n], \quad j \in [1, m].$$

As in “human” auctions, values cannot be negative; a value $u_{i,j} = 0$ means that policy p_j is not relevant to peer $peer_i$, that is, applying p_j does not give any benefit to $peer_i$ (this is

	U	$:$	$ufunction \rightarrow PSPEC \rightarrow \mathbb{R}^+$
$U[addend\ addList]_{ps}$	$=$	$U[addend]_{ps} + U[addList]_{ps}$	
$U[cb_name\ weight]_{ps}$	$=$	$\frac{int(S[cb_name]_{ps}) * int(weight)}{LMAX * WMAX * RQMAX}$	

Fig. 12. Semantics of utility function. $S : (R \cup Q) \rightarrow PSPEC \rightarrow level$ is a function that, given a resource/benefit name cb_name and a policy specification ps , fetches the $level$ associated to cb_name in ps (if the utility function tries to retrieve a value for a resource/benefit that does not appear in the policy specification, the returned value is 0). int is a function that given a literal in $\{ '1', \dots, 'MAX' \}$, returns the corresponding integer value in $[1, MAX]$; $LMAX * WMAX * RQMAX$ is the maximum bid an application can place, being $RQMAX$ the maximum number of resources/benefits of interest to an application.

a plausible “machine” representation of a “human” refuse to bid).

Utility functions vary *dynamically* to reflect changes in the user goals; however, the value they return is computed over *static* policy specifications which estimate the *consumption* of resources that applying the policy entails and the *benefits* it gives in terms of quality of service. If $R \subset \Sigma^*$ defines the set of resource names that the middleware monitors and $Q \subset \Sigma^*$ the set of benefits achieved by applying policies in P , then a policy specification can be described as a domain set $PSPEC = \wp(\{R \cup Q\} \times level)$, being $level ::= '1' \dots 'LMAX'$ an estimate of resource consumption/benefit achieved which the policy developers compute before delivering the policy.

The abstract syntax of a utility function is given in Fig. 11, where $cb_name \in (R \cup Q)$ is a name that uniquely identifies a resource or benefit inside a policy specification, and $weight ::= '1' \dots 'WMAX'$ is a factor that represents the importance the user associates to a particular resource/benefit (the higher the weight, the more important the resource/benefit). Although we consider the issue of generating weights that represent user needs as faithfully as possible a matter of future research, we will give a flavor of how these numbers can be obtained from users and directly used in our system in Section 5.2.

Whenever a peer $peer_i$ is involved in a bidding process, its utility function is retrieved and used to find the peer utility value $u_{i,j}$ for each conflicting policy p_j . The semantics of a utility function is presented in Fig. 12. As shown, each value is normalized to vary in a range $[0, 1]$, so that different bids can be compared effectively and money fairly redistributed (see Section 5.2).

As stated before, while policy specifications are fixed, utility function specifications change over time, as they have to reflect current user needs. This implies that the reflective mechanism of our middleware has to allow dynamic modification of utility function specifications. This allows our conflict resolution scheme to fulfill our second requirement, that is, *customization*.

Note that, to avoid incompatibility among the prices bid during a conflict resolution process, utility functions are locked at the beginning of an auction, and cannot be modified until the auction finishes. Thus, applications cannot “cheat” and associate high bids to the policies they value most, while bidding zero for the others, to increase the chances to have the policy they value most finally

Time / Action	q_1	$\bar{q}(1)$	q_2	$\bar{q}(2)$
t_0 / Start	3	0	3	0
t_1 / Bid	2.1	0.9	2.7	0.3
t_2 / Bid	1.2	1.8	2.4	0.6
t_3 / Bid			2.1	0.9
t_4 / Bid			1.8	1.2
t_5 / Bid			1.5	1.5
t_6 / Bid			1.2	1.8
t_7 / Redistribution	2.1	0.9	2.7	0.3

Fig. 13. Example of quota redistribution (with $t_7 - t_0 = t$).

applied, as this would require applications to change the weights of their utility functions during the auction.

4.3 Quota Allocation

When describing the rules of our mechanism (see Section 4.1), we specified that each peer $peer_i$ is allowed to bid a value $b_{i,j}$ for policy p_j , given that this value is lower than its current quota q_i . We now explain how this quota is managed.

Whenever an application instance A_i is started, an initial quota $q_i = q_{init}$ is awarded. Each time A_i participates to an auction, its quota is decreased by an amount equal to $f_i \in [0, 1]$, as defined in Fig. 10. A_i 's underlying middleware instance collects A_i payments and stores them in a wallet $\bar{q}(i)$. We assume that there is no flow of money from one middleware instance to another (i.e., each application instance pays its underlying middleware instance). Moreover, we assume that there is no explicit utility transfer among applications (e.g., no money can be transferred to a peer to compensate for a disadvantageous agreement).

Every t time units, each middleware instance redistributes the money it has collected in its wallets $\bar{q}(i)$, $i \in [1, n]$, to the various application instances A_i , $i \in [1, n]$. The amount of money each application instance gets back is in direct relation to the number of interactions it has been involved during the last t time units and in inverse relation to the amount of money it bid. We define an interaction as a service request which incorporates an interprofile conflict (intraprofile conflicts are excluded from the quota recharging as no flow of money occurs).

In particular, if we indicate with $N_t(i)$ the number of interactions in which application instance A_i was involved in the last t time units, then the recharging process is carried out as described below:

$$q_i = q_i + \left(\bar{q}(i) - \frac{\bar{q}(i)}{N_t(i)} \right); \quad \bar{q}(i) = \frac{\bar{q}(i)}{N_t(i)},$$

where $\bar{q}(i)$ is the money currently stored by the middleware in the wallet associated to A_i , and q_i A_i current quota.

This quota redistribution scheme discourages dictatorial interactions: If an application instance bids very high in a few interactions, "imposing" its preferred policy over the others, then only a very low amount of money is returned during a recharging process. The only way to get money back from the middleware is to participate in other interactions in a more cooperative fashion (i.e., by bidding lower and interacting more). For example, let us assume

```
soundAlert: { (battery, 6), (privacy, 1), (focus, 8) }
vibraAlert: { (battery, 10), (privacy, 7), (focus, 8) }
silentAlert: { (battery, 1), (privacy, 10),
              (focus, 2) }
```

Fig. 14. Policy specifications.

that at time t_0 , two application instances A_1 and A_2 started and awarded the same quota $q_i = 3$, $i \in \{1, 2\}$. During the following t time units, they are involved in a number of interactions that cost them altogether the same amount of money; however, while A_1 bid aggressively, paying a lot of money in few interactions, A_2 was more cooperative, paying low amounts in many interactions. As a result, our quota redistribution scheme returns money to A_2 faster than to A_1 (see Fig. 13).

The approach to quota redistribution that we have described could be defined as "conservative:" At any time, an application instance A_i has got the same amount of money, although split differently between its current quota q_i and the corresponding middleware wallet $\bar{q}(i)$. In other words:

$$\bar{q}(i) + q_i = q_{max},$$

where q_{max} is a fixed amount that is the same for any application. At time t_0 when an application instance A_i is started, different choices of q_{init} and $\bar{q}(i)$ are possible. In particular, any assignment that complies with the following equations is acceptable:

$$\forall \alpha \in [0, 1] \left\{ \begin{array}{l} q_{init} = \alpha \cdot q_{max} \\ \bar{q}(i) = (1 - \alpha) \cdot q_{max} \end{array} \right.$$

Setting $\alpha = 1$ favors newly started application instances, while setting $\alpha = 0$ favors applications that have been executing for a long while. The differences among these possibilities disappear while time passes. It is beyond the scope of this paper to investigate the optimal choice for q_{init} , q_{max} , t , and α .

This concludes the discussion about our auctioning approach to the conflict resolution problem. In the following section, we illustrate how this mechanism can be instantiated and used to solve conflicts.

4.4 Conference Application

In this Section, we present examples of intra- and interprofile conflicts that may occur in the conference application and show how our auctioning mechanism is used to resolve them.

4.5 Intraprofile Conflict: Talk Reminder

Let us assume that the talk reminder service can be delivered using one of the following policies: a `soundAlert` policy, a `vibraAlert` policy, and a `silentAlert` policy. Each of these policies requires different amounts of resources to be used (in particular, battery), and achieves a different quality of service (in terms of focusing and privacy). The corresponding policy specifications are shown in Fig. 14.

Whenever a talk reminder has to be delivered, the application profile is consulted to find out which policy to

talkReminder			
soundAlert			
location = outdoor	battery		2
vibraAlert	privacy		10
location = indoor	focus		10
silentAlert			
location = indoor			
battery < 15%			
(a)		(b)	

Fig. 15. (a) Application profile. (b) Utility function. $peer_1$ aims to maximize privacy and focusing, without too much interest in battery consumption.

apply. Let us assume that the application profile is the one illustrated in Fig. 15a, and that the talk reminder service is invoked when the user is attending a talk (i.e., `location = indoor`), and battery is lower than 15 percent, so that both `vibraAlert` and `silentAlert` are enabled (intraprofile conflict). Note that, although it could be argued that such a conflict would not exist if the profile were properly written (i.e., if a line containing `battery > 15%` were added to the context of the `vibraAlert` policy), avoiding context overlaps is not so easy. When the number of resources associated to a context increases, chances of making mistakes and of writing profiles with context overlaps increase quickly. As already argued, a static conflict analysis would be unmanageable on portable devices, and therefore a dynamic solution is needed. We now illustrate how our dynamic conflict resolution mechanism works effectively to solve this conflict, assuming that the utility function is the one illustrated in Fig. 15b.

Computation of the solution set. First, the solution set P^* is computed; as only one peer is involved, P^* coincides with P_1 :

$$\mathcal{I}[\text{talkReminder}]_{\{peer_1\}} = \{\text{vibraAlert}, \text{silentAlert}\}.$$

Computation of bids. High weights associated to resources in utility function specifications mean that the user aims at sparing resources; however, policy specifications estimate the amount of resources consumed, not spared. In order to give higher scores (i.e., higher bid prices) to the policies that reduce resource consumption, we therefore need to compute the value: $LMAX - \text{expected consumption}$. For example, if we assume $LMAX = 10$, $WMAX = 10$, and $RQMAX = 5$ (i.e., battery, bandwidth, focusing, availability and privacy), then:

$$u_{peer_1}(\text{vibraAlert}) = \frac{(10 - 10) * 2 + 7 * 10 + 8 * 10}{10 * 10 * 5} = 0.3.$$

Assuming that the peer quota $q_{peer_1} > 1$ (i.e., the bid is not constrained by current quota, as each bid $b_{1,j} \in [0, 1]$), we obtain:

$$\mathcal{B}[\{\text{vibraAlert}, \text{silentAlert}\}]_{\{peer_1\}} = \{(\text{vibraAlert}, peer_1, 0.3), (\text{silentAlert}, peer_1, 0.276)\}.$$

Election of the winner. As only one peer is involved in an intraprofile conflict, maximizing social welfare coincides

charMsg: { (battery, 4), (bandwidth, 10), (availability, 10) }
plainMsg: { (battery, 3), (bandwidth, 6), (availability, 7) }
compressedMsg: { (battery, 5), (bandwidth, 4), (availability, 5) }
encryptedMsg: { (battery, 6), (bandwidth, 7), (availability, 4), (privacy, 10) }

Fig. 16. Policy specifications.

with maximizing individual utility. The winning policy is the one that $peer_1$ valued most and no quota adjustment is needed.

$$\mathcal{W}[\mathcal{B}[\{\text{vibraAlert}, \text{silentAlert}\}]_{\{peer_1\}}] = \text{vibraAlert}.$$

4.6 Interprofile Conflict: Messaging

Peers can exchange messages using one of the following policies: a `charMsg` policy, a `plainMsg` policy, a `compressedMsg` policy, and an `encryptedMsg` policy. Again, each of these policies requires different amounts of resources (in particular, battery and bandwidth), and achieves a different quality of service (in terms of availability and privacy of the message). The corresponding policy specifications are shown in Fig. 16.

Let us suppose that three peers $peer_1$, $peer_2$, and $peer_3$ are in reach of each other and want to start a chat. In order to do so, they have to agree on a common policy to be applied to exchange messages. During the lifetime of the chat, the policy used may change to adapt to new context configurations where the currently used policy is no longer suitable. However, when this happens, all the chatting peers must agree on the new policy to use.

The peers' application profiles are represented in Fig. 17. Note that $peer_3$ leaves the `plainMsg` policy always enabled: This is a good way to reduce the risk of ending a conflict resolution process with a failure because no agreed policy could be found. However, this increases the risk of conflicts and, consequently, the time used to resolve them

% peer 1	% peer 3
messagingService	messagingService
charMsg	plainMsg
bandwidth > 70%	
plainMsg	compressedMsg
bandwidth < 70%	bandwidth < 40%
compressedMsg	encryptedMsg
bandwidth < 35%	battery > 60%
encryptedMsg	
battery > 50%	
% peer 2	
messagingService	
plainMsg	
battery < 50%	
compressedMsg	
bandwidth < 40%	

Fig. 17. Application profiles.

% peer 1		% peer 2	% peer 3
battery	4	battery	7
bandwidth	3	bandwidth	9
availability	10		privacy 10

Fig. 18. Utility functions. $peer_1$ aims to maximize availability without wasting resources, $peer_2$ aims to minimize resource consumption, and $peer_3$ aims to maximize privacy.

(which is anyway rather low, as it will be shown in Section 5.1). It is up to the application to decide which strategy is best.

Assuming that the utility functions are the ones shown in Fig. 18 and that the current execution context enables the following sets of policies:

$$\begin{aligned} P_1 &= \{\text{plainMsg}, \text{compressedMsg}, \text{encryptedMsg}\} \\ P_2 &= \{\text{plainMsg}, \text{compressedMsg}\} \\ P_3 &= \{\text{plainMsg}, \text{compressedMsg}, \text{encryptedMsg}\} \end{aligned}$$

for peers $peer_1$, $peer_2$, and $peer_3$, respectively, then the conflict resolution process proceeds as described below.

Computation of the solution set. First, the solution set P^* , that is, the set of commonly agreed policies is computed:

$$\begin{aligned} I[\text{messagingService}]_{\{peer_1, peer_2, peer_3\}} &= P_1 \cap P_2 \cap P_3 \\ &= \{\text{plainMsg}, \text{compressedMsg}\}. \end{aligned}$$

Computation of bids. Assuming, as before, $LMAX = 10$, $WMAX = 10$, $RQMAX = 5$, and that each peer has a quota $q_{peer_i} > 1$, we obtain:

$$\begin{aligned} \mathcal{B}[\{\text{plainMsg}, \text{compressedMsg}\}]_{\{peer_1, peer_2, peer_3\}} &= \\ &\{(\text{plainMsg}, peer_1, 0.22), (\text{compressedMsg}, peer_1, 0.176), \\ &(\text{plainMsg}, peer_2, 0.17), (\text{compressedMsg}, peer_2, 0.178), \\ &(\text{plainMsg}, peer_3, 0), (\text{compressedMsg}, peer_3, 0)\}. \end{aligned}$$

Election of the winner. Bids received for each policy in the solution set are added and the policy that maximizes the sum (i.e., social welfare) is selected.

	plainMsg		compressedMsg	
$peer_1$	0.22	+	0.176	+
$peer_2$	0.17	+	0.178	+
$peer_3$	0		0	
	0.39		0.354	

$$\mathcal{W}[\mathcal{B}[\{\text{plainMsg}, \text{compressedMsg}\}]_{\{peer_1, peer_2, peer_3\}}] = \text{plainMsg}.$$

Finally, each peer quota is adjusted in the following way:

$$\begin{aligned} q_1 &= q_1 - \frac{0.22 - 0.17}{1} - \frac{0.17 - 0}{2} - \frac{0}{3} \\ q_2 &= q_2 - \frac{0.17 - 0}{2} - \frac{0}{3} \\ q_3 &= q_3 - \frac{0}{3}. \end{aligned}$$

5 EVALUATION

In this section, we evaluate our approach in terms of performance and usability.

5.1 Performance

The performance of CARISMA have been measured based on our current implementation: The middleware has been implemented in Java using jdk 1.4.1, while application profiles and utility functions have been encoded using the eXtensible Markup Language (XML) (their grammar has been defined in two associated XML Schema available at <http://www.cs.ucl.ac.uk/staff/l.capra/schema>). We chose to use XML as we believe this language may enhance context-aware and user-driven interactions between middleware and applications, supporting a representation of information that can be both easily manipulated by machines, and readily understood by humans. Also, XML related technologies, in particular, DOM and XPath, and available XML parsers have considerably reduced the development time. Communication takes place via a simple message passing mechanism that we have implemented. The middleware platform currently requires only 110Kb of persistent storage and less than 800Kb of memory (without considering the memory required by the Java Virtual Machine and XML parser).

We performed tests on Dell Latitude laptops equipped with 128MB RAM, Intel Pentium II processors rated at 300MHz, and connected in an ad hoc network using Cisco Aironet 340 10Mbps wireless cards. The operating system used was Microsoft's Windows2000 and the Java Virtual Machine version was 1.4.1. We believe these machines are well-suited to estimate the performance of our middleware, as they do not outperform the currently available portable devices (e.g., the Sony Ericsson P800 mobile phone is equipped with 12Mb internal storage, plus external memory stick, and ARM9 200MHz processor; the HP iPAQ Pocket PC h5450 is equipped with 64Mb RAM and Intel 400MHz processor; COMPAQ Tablet PC TC1000 is already extremely powerful, with 256MB RAM minimum and 1GHz processor).

In order to estimate the scalability of CARISMA in terms of the number of devices involved in the delivery of a service, the number of (possibly conflicting) policies associated to each service, the number of contexts for each policy, and the number of resources in each context, we have implemented a benchmark that allowed us to tune each of these parameters independently. The charts shown in this section represent the average of the results obtained over 20 service requests.

In the remainder of this section, we first consider a simple local service, i.e., a service that involves a single device. In this simple scenario, we illustrate the basic overheads introduced by reflection, by context-awareness while varying number of contexts and resources associated to each policy, and by the conflict resolution mechanism. We then move to a distributed setting and we analyze the performance of CARISMA while varying the number of devices involved in a service request.

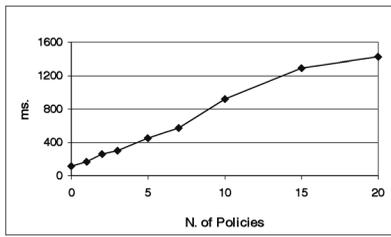


Fig. 19. Impact of reflection.

5.1.1 Impact of Reflection

Fig. 19 illustrates the overhead introduced by reflection over a basic mechanism where a service is statically associated to a policy. This lower bound is represented in the picture by the intersection of the curve with the Y axis. As the picture shows, the overhead (in milliseconds) is more or less linear in the number of policies associated to the service, and is kept below 1 second even when 10 policies are associated to the same service. This overhead includes also the evaluation of a simple context configuration made of one context with one resource associated to each policy (these associations are necessary to avoid conflicts).

5.1.2 Impact of Context-Awareness

Fixing the maximum number of policies associated to a service to 10, Fig. 20 shows the impact of context-awareness on performance. We assume here that, whenever a service is invoked, the current value of each resource is already available (i.e., the middleware is probing the physical sensors at regular intervals to keep updated context information). The performance results we are discussing do not consider the time necessary to initialize a sensor and to process the information gathered through it; we believe this approach is plausible, as sensors may greatly vary in nature and, therefore, may introduce overheads of different orders of magnitude (e.g., knowing the amount of battery left requires much less time than gathering and processing location information).

As shown, managing profiles with 10 or more policies, each with five or more contexts associated, and 10 or more resources for each of these contexts, represents a scalability limit in the performance of CARISMA. This is due to the fact that the number of comparisons between the current context and the associations encoded in the profile grows

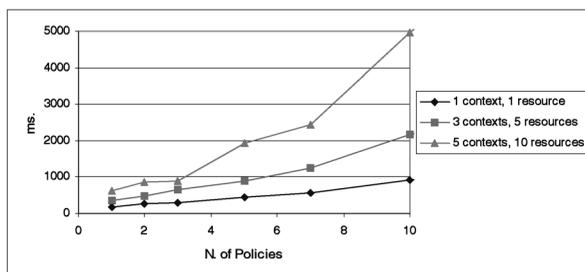


Fig. 20. Impact of context-awareness.

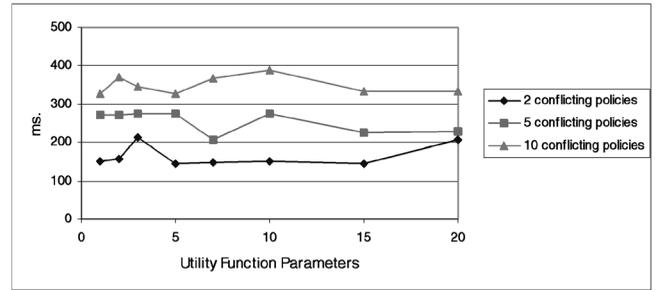


Fig. 21. Impact of utility function parameters.

exponentially with the number of contexts and resources. In our experience with the conference application, however, this scalability limit was never reached, as having five policies associated to three contexts with five resources each, already represented the maximum level of adaptation we needed (i.e., the worst-case scenario); in this case, the average amount of time to request a local service is still below one second.

5.1.3 Impact of Conflicts

As the following two figures show, the conflict resolution process has a minor impact on the performance of CARISMA. First (see Fig. 21), the number of utility function parameters does not influence the performance of a service request at all. Also, this chart depicts the performance of a local service request where no context is associated to the policies (i.e., they are always enabled); comparing this chart with the one shown in Fig. 19, we can conclude that the conflict resolution mechanism introduces a much lower overhead than the simplest case of context-awareness. In fact, it takes about 900ms to determine which policy to apply out of 10, in case a very simple, mutually exclusive (i.e., no conflict) context is provided (one context with one resource), while it takes less than 400ms in case no context is provided and the conflict resolution procedure has to be executed.

Second, the conflict resolution mechanism adds a negligible overhead over the standard mechanism (inclusive of context-awareness), as depicted in Fig. 22. In case each policy is associated with the same number of contexts and resources, the overhead introduced by the conflicts resolution mechanism is almost constant and in the order of 200ms.

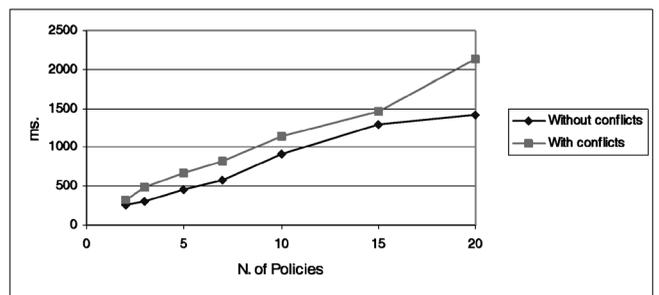


Fig. 22. Impact of conflict resolution mechanism.

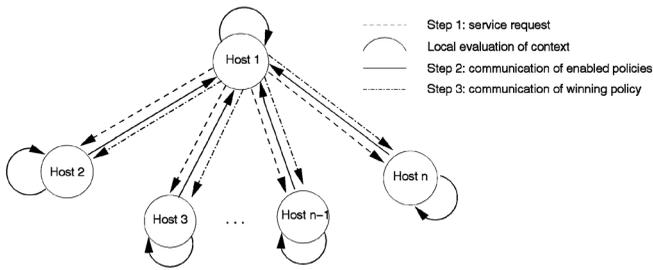


Fig. 23. Communication protocol.

We can conclude that a good strategy in developing applications on top of CARISMA is to associate only minimal context configurations to the policies and have the auction mechanism solve potential conflicts.

5.1.4 Impact of Distribution

The last chart shows the performance of CARISMA in answering a service request for two plausible profile configurations, while varying the number of devices involved in the delivery of the service. These results have been computed considering an implementation of the auction protocol that is based on the 3-step communication protocol shown in Fig. 23.

This protocol tries to maximize performance by parallelizing the context evaluation step among the n peers involved in a service execution. Whenever an application running on Host 1 needs a service that requires the cooperation of $n - 1$ other applications running on as many hosts, these steps are followed.

Step 1: First, Host 1 sends out a service request message to each of the $n - 1$ peers involved in the service execution. At this point, all the n peers evaluate their local context in parallel and find out the locally enabled policies (i.e., the sets P_i , for $i \in [1, n]$). Moreover, although no conflict has been detected yet, they compute a bid for each of these policies.

Step 2: The $n - 1$ peers communicate their own P_i and corresponding bids back to the requesting host, which now computes the solution set P^* . If no conflict is found, the precomputation of the bids was a waste of time and resources, but if, on the contrary, a conflict is detected, two additional communication steps are saved (i.e., to ask the $n - 1$ peers to bid for the policies in P^* and to communicate these bids back to the requesting host). As the time taken by the computation of the bids is negligible (i.e., few milliseconds) compared to the time taken by two additional communication steps, the precomputation proves to be worthwhile.

Step 3: Once the winning policy has been selected and the payments have been computed, the requesting host sends this information to the $n - 1$ participating peers and the service can be finally delivered.

Individual failures of participating peers taking place during the protocol execution do not compromise its success, as long as there are at least $0 < m < n$ peers connected until the end of the process (the minimum number of connected peers, m , is application dependent). However, if the requesting peer fails, the entire service request is aborted.

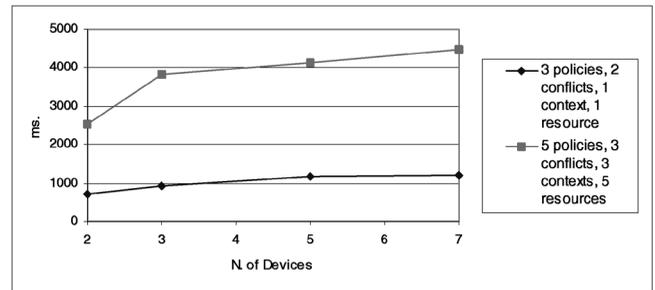


Fig. 24. Impact of conflicts in a distributed setting.

As shown in Fig. 24, the overhead tends to be constant, and does not increase considerably while increasing the number of devices involved. The results shown here do not consider peer failures; in case failures are taken into account, the overhead depends on the timeout values used before acknowledging a peer is no more in reach.

5.2 Usability

To estimate the usability of our middleware, we have assigned a student the task of implementing the conference application on top of CARISMA. From the student's report, it emerged that the most difficult task was to decide which nonfunctional parameters the user could tune, and how to map them into application profiles, while using the abstractions and mechanisms provided by CARISMA turned out to be rather straightforward.

The student decided to allow the end-user of the system to tune the importance he/she assigned to both nonfunctional requirements (i.e., availability of information, accuracy, and privacy), and to local resources (i.e., memory, battery, and bandwidth), by means of the customization windows illustrated in Fig. 25. The effort required from the end-user of the system was rather limited: When his/her preferences were changing, all he/she had to do was to input the new preferences through these windows.

Based on these preferences, the student implemented a synthesizing algorithm to write both application profiles and utility functions. Application profiles were encoding associations with 2/3 policies per service, each with 1/2 contexts made of 2/3 resources. Utility functions simply listed the importance users assigned to customizable parameters (e.g., in the picture above, memory=4, battery=9, bandwidth=0, availability=8, accuracy=3, privacy=0).

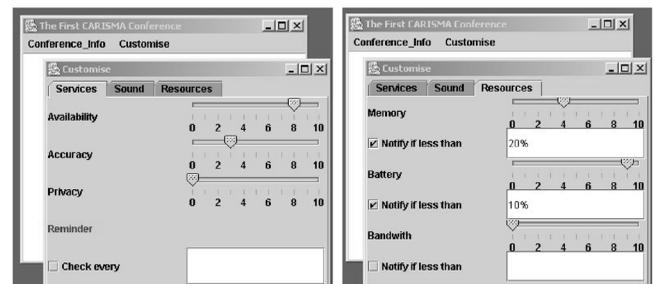


Fig. 25. Conference application customization.

Estimating the end-user efforts in teaching the system to behave according to his/her own preferences, strongly depends on the level of adaptation the application permits: the higher the number of parameters that are subject to customization, the finer the level of adaptation the system may achieve. However, this would have an impact on the amount of human effort required, as well as on the complexity of the synthesizing algorithm that application engineers have to come up with. Note that these issues are not intrinsic in our middleware model, but apply, in general, to scenarios where adaptation to changing context and user requirements is needed; in developing context-aware applications, therefore, granularity of adaptation has to be traded against human effort. Further research is needed in this direction to estimate where the equilibrium lies.

6 RELATED WORK

Providing a detailed review of the state of the art in the area of context-awareness, reflection, and mobile computing is beyond the scope of this paper. A critical literature review in the area of reflection and context-awareness (and, more generally, in middleware for mobile computing), can be found in [12]. In this paper, we compare our work with what various research communities have done, as far as conflict resolution is concerned.

The operating systems community has studied the issue of conflicts in a distributed environment, where conflicts manifest themselves as processes competing for shared resources. Microeconomic techniques, and auctions in particular, have been explored; in [13], a market-like bidding mechanism is described which assigns tasks to processors that have given the lowest estimated completion time; similar techniques have been used to manage network traffic [14] and allocation of storage space [15]. We have demonstrated that game theory can also be successfully used to resolve QoS conflicts that arise in the mobile setting; however, the nature of conflicts is fundamentally different, thus requiring different conflict resolution algorithms. In particular, resource conflicts happening at the operating system level represent competitive situations where only one competitor obtains the resources, leaving all the others without them. In our case, instead, collaboration characterizes the nature of the auction better: peers participating in the delivery of a service will *all* get the good (the delivery of the service), but with varying degrees of satisfaction. Traditional auctions cannot be applied in this setting, and we had to come up with a novel mechanism to deal with these conflicts.

Despite the extensive research that has been carried out within the mobile middleware community, the issue of QoS conflicts has attracted little attention. On one hand, many systems do not support dynamic adaptation of middleware behavior and, thus, they avoid the problem of conflicts a priori. On the other hand, systems which exploit reflection to improve flexibility and allow dynamic reconfigurability of the middleware [16], [17] generally target a stationary distributed environment, where context changes (and, consequently, adaptation of middleware behavior) are

much less frequent than in a mobile setting, so that the problem of conflicts is less pressing. Data conflicts have been investigated more extensively instead: In order to maximize data availability in mobile settings, where sudden disconnections may happen frequently, even for long periods of time, systems such as Coda [18], Bayou [19], and Xmiddle [20] give users access to replicas. They differ in the way they ensure that replicas move towards eventual consistency, that is, in the mechanisms they provide to detect and remove conflicts that naturally arise in mobile systems. Data conflicts, however, are fundamentally different from the QoS conflicts we treat and, therefore, these solutions can hardly be applied; in particular, interprofile conflicts are not intrinsic in any profile, but manifest themselves only in relation to (some) other profiles and in particular contexts and, therefore, cannot be removed, but only dynamically solved.

The software engineering community has investigated the issue of conflicts too. Software development environments [21], [22] have devised mechanisms for specifying consistency constraints between artifacts. They are able to detect static violations of these constraints and resolve them automatically (e.g., by propagating changes to dependent documents). Inconsistencies are often found in requirements documents, indicating conflicts between the different stakeholders involved. Requirements management methods and tools therefore include inconsistency detection and resolution mechanisms. The KAOS method [23] uses a goal-oriented approach to decompose requirements and formalizes them using a temporal logic. Conflicts are detected by reasoning about the temporal logic formulae and conflict resolution strategies [24] can be applied so that requirement conflicts are not come down to design. Other requirements engineering approaches [25] leave inconsistencies in specifications and use an appropriate logic to continue reasoning, even in the presence of an inconsistency. These approaches, however, are of limited use in a mobile setting where the nature of conflicts is such that they cannot be detected statically at the time an application is designed but, instead, they can only be detected and resolved at runtime. Also, they must be resolved, otherwise applications cannot execute.

Our work is more closely related to approaches that monitor requirements and assumptions during the execution of systems. Fickas and Feather's approach towards requirements monitoring [26] uses a Formal Language for Expressing Assumptions (FLEA). FLEA is supported by a CLISP-based runtime environment, which can alert requirement violations to the user. For mobile systems, however, this is insufficient and a more proactive approach to resolving conflicts is required. Robinson and Pawlowski [27] have developed a so-called "requirements dialog Metamodel," which supports not only the definition and monitoring of goals, but also the reestablishment of a dialog goal in case of a goal failure. Goal monitoring is performed actively, so that violations are detected immediately; however, this requires a consumption of resources that hand-held devices cannot bear.

In the Distributed Artificial Intelligence (DAI) community, game theory [7] has been extensively applied to treat negotiation issues. Negotiation mechanisms have been used both to assign tasks to agents, to allocate resources, and to decide which problem solving tasks to undertake (e.g., [28], [29]). These scenarios typically involve a group of agents operating in a shared environment. Each agent has its own private goal; a negotiation process is put in place that, through a sequence of offers and counter-offers, explores the chance for agents of achieving their (possibly conflicting) goals, at the lowest cost. Despite similarities with our scenario, there are a number of assumptions that differentiate our work from previous results obtained in the DAI community. In particular, in DAI the quality of the result is valued much more than the cost of achieving it; as a consequence, negotiation mechanisms are usually iterative processes which carry on until an (optimal) agreement is reached. In a mobile setting, instead, resource constraints call for simple conflict resolution mechanisms that do not waste (scarce) resources. Moreover, the nature of goals is fundamentally different. In DAI, a goal can be seen as a task composed of atomic operations that the negotiation mechanism is able to assign to different agents; in our setting, goals are rather indivisible units that suggest the quality of service levels that applications are wishing to achieve to the middleware.

Also relevant to our work is the research on quality of service provision in a mobile computing environment [30]. QoS requirements are defined by all applications and a negotiation mechanism is put in place to reach an agreement between all parties; as a result of context changes, a dynamic renegotiation of the contract may be necessary. The approaches we have analyzed usually target a specific domain (e.g., multimedia applications over broadband cellular networks), mainly focusing on bandwidth allocation [31]. Moreover, applications have a rather limited way of influencing the policies that are chosen to meet QoS requirements. Our middleware aims at being general and uses reflection to give applications the power to influence the way adaptation is achieved. This may lead to disagreements among applications to reach the quality-of-service level they wish.

7 CONCLUSION AND FUTURE WORK

The increasing popularity of portable devices and recent advances in wireless networking technologies are facilitating the engineering of new classes of applications, which present challenging problems to designers. To accommodate the new requirements of mobility and, in particular, the need for context-awareness and adaptation, middleware platforms for mobile computing must be capable of both deployment-time configurability and runtime reconfigurability.

In this paper, we have described CARISMA, a mobile computing middleware that exploits reflective techniques to enable mobile application designers to address these requirements. Besides enabling dynamic adaptation to context, reflection may also cause conflicts. We have

demonstrated how CARISMA uses microeconomic techniques effectively in order to solve conflicts that arise in the mobile setting. In particular, we have modeled a mobile distributed system as an economy, where applications compete to have a common service delivered according to their preferred quality-of-service level; in this economy, the middleware plays the role of an auctioneer, collecting bids from applications and selecting the policy that maximizes social welfare. This approach is particularly suited in the mobile setting as it meets the requirements of dynamicity, simplicity, and customizability that are typical of this environment.

Future improvements and extensions of CARISMA span towards different directions. Despite being a very powerful means, reflection enables adaptability and flexibility only in those contexts that middleware designers have considered likely to be unstable at design time. However, in a mobile ad hoc setting, mobile hosts cannot forecast all the possible contexts they are going to encounter and, therefore, which protocols (i.e., behaviors) they are going to need; new behaviors may be delivered from time to time to cope with unforeseen context configurations and new application needs. Moreover, only a minimum set of behaviors can be stored on a device, so to avoid wasting memory. A future direction of research is to combine reflection with mobile code techniques to overcome this limitation, thus allowing applications to download new protocols either from a service provider or from other peers in reach only when needed [32].

To accommodate dynamicity requirements, services and policies are installed and uninstalled on the fly; moreover, different application needs result in different system configurations, that vary over time. The changing interactions among distributed services and policies may alter the semantics of the applications built on top of our reflective middleware. The development of safe customizable middleware becomes, therefore, an issue. A first step towards the definition of a formal semantics for specifying and reasoning about the properties of, and interactions among, middleware components can be found in [33]. These principles have been used, for example, in [34] to manage changes in large-scale distributed systems while ensuring application QoS requirements. The principles they use are based on a two-level architecture, where the application, at the base level, interacts with the middleware, at the metalevel, via middleware-defined core services that are then used to initiate other activities. The similarity of this approach with our architecture makes us think that similar principles could be investigated to develop a formal semantics of composition within our reflective middleware framework.

Another direction of research that is worth mentioning is service discovery. Traditional naming and trading service discovery techniques developed for fixed distributed systems cannot be successfully applied in mobile settings, where intermittent rather than continuous network connection is the norm. However, service discovery for mobile settings has not yet gained significant attention. Two

notable exceptions are the Jini specification [35] and the work by Handorean and Roman [36]. A disadvantage common to both approaches is that they do not take quality of service requirements into account when deciding which service to use. We believe that QoS-aware service discovery would fit naturally in our framework, where application needs are made explicit and used to decide how a service should be delivered in current context. Currently, these needs are taken into account only locally; a future direction of research would be to make use of this information to discover services available in an entire ad hoc network that would deliver the user the best QoS, according to current user-specific requirements.

Last but not least, a study that puts together middleware practitioners, HCI experts, and requirement elicitation experts is necessary to estimate the amount of work required from application engineers to develop context-aware applications and from end-users to learn how to use these systems.

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