A survey on context-aware systems

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Abstract: Context-aware systems offer entirely new opportunities for application developers and for end users by gathering context data and adapting systems behaviour accordingly. Especially in combination with mobile devices these mechanisms are of high value and are used to increase usability tremendously. In this paper, we present common architecture principles of context-aware systems and derive a layered conceptual design framework to explain the different elements common to most context-aware architectures. Based on these design principles, we introduce various existing context-aware systems focusing on context-aware middleware and frameworks, which ease the development of context-aware applications. We discuss various approaches and analyse important aspects in context-aware computing on the basis of the presented systems.

Keywords: context-awareness; context framework; context middleware; sensors; context model; context ontology; context-aware services.


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1 Introduction

With the appearance and penetration of mobile devices such as notebooks, PDAs, and smart phones, pervasive (or ubiquitous) systems are becoming increasingly popular these days. The term ‘pervasive’ introduced first by Weiser (1991) refers to the seamless integration of devices into the users everyday life. Appliances should vanish into the background to make the user and his tasks the central focus rather than computing devices and technical issues. One field in the wide range of pervasive computing are the so-called context-aware (or sentient) systems. Context-aware systems are able to adapt their operations to the current context without explicit user intervention and thus aim at increasing usability and effectiveness by taking environmental context into account. Particularly when it
comes to using mobile devices, it is desirable that programs and services react specifically to their current location, time and other environment attributes and adapt their behaviour according to the changing circumstances as context data may change rapidly. The needed context information may be retrieved in a variety of ways, such as applying sensors, network information, device status, browsing user profiles and using other sources. The history of context-aware systems started when Want et al. (1992) introduced their Active Badge Location System which is considered to be one of the first context-aware applications. The infrared technology based system is able to determine a user’s current location which was used to forward phone calls to a telephone close to the user. In the middle of the 1990s, a couple of location-aware tour guides (Abowd et al., 1997; Sumi et al., 1998; Cheverst et al., 2000) emerged which provided information according to the user’s current location. While location information is by far the most frequently used attribute of context, attempts to use other context information as well have grown over the last few years as the examples in this paper will show. Hence, it is a challenging task to define the word ‘context’ and many researchers tried to find their own definition for what context actually includes. In literature the term context-aware appeared in Schilit and Theimer (1994) the first time. There the authors describe context as location, identities of nearby people, objects and changes to those objects. Such enumerations of context examples were often used in the beginning of context-aware systems research. Ryan et al. (1997) referred to context as the user’s location, environment, identity and time. Dey (1998) defines context as the user’s emotional state, focus of attention, location and orientation, date and time, as well as objects and people in the user’s environment. Another common way of defining context was the use of synonyms. Hull et al. (1997) describe context as the aspects of the current situation. These kind of definitions are often too wide. However, a good one can be found in Brown (1996). Brown defines context to be the elements of the user’s environment which the computer knows about. One of the most accurate definitions is given by Dey and Abowd (2000b). These authors refer to context as

“any information that can be used to characterize the situation of entities (i.e., whether a person, place or object) that are considered relevant to the interaction between a user and an application, including the user and the application themselves.”

One popular way to classify context instances is the distinction of different context dimensions. Prekop and Burnett (2003) and Gustavsen (2002) call these dimensions external and internal, and Hofer et al. (2002) refer to xtitaphysical and logical context. The external (physical) dimension refers to context that can be measured by hardware sensors, i.e., location, light, sound, movement, touch, temperature or air pressure, whereas the internal (logical) dimension is mostly specified by the user or captured by monitoring user interactions, i.e., the user’s goals, tasks, work context, business processes, the user’s emotional state. Most context-aware systems make use of external context factors as they provide useful data, such as location information. Furthermore, external attributes are easy to sense by using off-the-shelf sensing technologies. Virtually all systems presented in this paper apply physical context information. Examples for the use of logical data are the Watson Project (Budzik and Hammond, 2000) and the IntelliZap Project (Finkelstein et al., 2001) which support the user by providing relevant information due to information read out of opened web pages, documents, etc. When dealing with context, three entities can be distinguished (Dey and Abowd, 2001): places (rooms, buildings etc.), people (individuals, groups) and things (physical objects, computer components etc.). Each of these entities may be described by various attributes which can be classified into four categories: identity (each entity has a unique identifier), location (an entity’s position, co-location, proximity etc.), status (or activity, meaning the intrinsic properties of an entity, e.g., temperature and lightning for a room, processes running currently on a device etc.) and time (used for timestamps to accurately define situation, ordering events etc.).

This paper is structured as follows. Section 2 introduces current design principles for context-aware systems and common context models used in various context-aware systems. In Section 3, we present a comparison of existent context-aware systems and explain approaches, varieties and similarities. In Section 4, we discuss the presented approaches, and highlight advantages and disadvantages. Finally, Section 5 draws some concluding remarks and presents some future work in this area.

2 Design principles

In this section, we describe basic design principles and introduce a conceptually layered framework, to associate the functionality implemented in existent frameworks to various layers. Furthermore, we depict different context models used for representing, storing and exchanging contextual information.

2.1 Architecture

Context-aware systems can be implemented in many ways. The approach depends on special requirements and conditions such as the location of sensors (local or remote), the amount of possible users (one user or many), the available resources of the used devices (high-end-PCs or small mobile devices) or the facility of a further extension of the system. Furthermore, the method of context-data acquisition is very important when designing context-aware systems because it predefines the architectural style of the system at least to some extent. Chen (2004) presents three different approaches on how to acquire contextual information.
• **Direct sensor access.** This approach is often used in devices with sensors locally built in. The client software gathers the desired information directly from these sensors, i.e., there is no additional layer for gaining and processing sensor data. Drivers for the sensors are hardwired into the application, so this tightly coupled method is usable only in rare cases. Therefore, it is not suited for distributed systems due to its direct access nature which lacks a component capable of managing multiple concurrent sensor accesses.

• **Middleware infrastructure.** Modern software design uses methods of encapsulation to separate e.g., business logic and graphical user interfaces. The middleware based approach introduces a layered architecture to context-aware systems with the intention of hiding low-level sensing details. Compared to direct sensor access this technique eases extensibility since the client code has not be modified anymore and it simplifies the reusability of hardware dependent sensing code due to the strict encapsulation.

• **Context server.** The next logical step is to permit multiple clients access to remote data sources. This distributed approach extends the middleware based architecture by introducing an access managing remote component. Gathering sensor data is moved to this so-called context server to facilitate concurrent multiple access. Besides the reuse of sensors, the usage of a context server has the advantage of relieving clients of resource intensive operations. As probably the majority of end devices used in context-aware systems are mobile gadgets with limitations in computation power, disk space etc., this is an important aspect. In return one has to consider about appropriate protocols, network performance, quality of service parameters etc., when designing a context-aware system based on client-server architecture.

In a similar manner, Winograd (2001) describes three different context management models for coordinating multiple processes and components:

• **Widgets.** Derived from the homonymous GUI elements a widget is a software component that provides a public interface for a hardware sensor (Dey and Abowd, 2000a, 2001). Widgets hide low-level details of sensing and ease application development due to their reusability. Because of the encapsulation in widgets it is possible to exchange widgets which provide the same kind of context data (e.g., exchange a radio frequency widget by a camera widget to collect location data). Widgets are usually controlled by some kind of a widget manager. The tightly coupled widget approach increases efficiency but is not robust to component failures.

• **Networked services.** This more flexible approach, argued for example in Hong and Landay (2001), resembles the context server architecture. Instead of a global widget manager discovery techniques are used to find networked services. This service based approach is not as efficient as a widget architecture due to complex network based components but provides robustness.

• **Blackboard model.** In contrast to the process-centric view of the widget and the service-oriented model, the blackboard model represents a data-centric view. In this asymmetric approach processes post messages to a shared media, the so-called blackboard, and subscribe to it to be notified when some specified event occurs. Advantages of this model are the simplicity of adding new context sources and the easy configuration. One drawback is the need of a centralised server to host the blackboard and the lack in communication efficiency as two hops per communication are needed.

In this paper we will focus on middleware based and context-server based systems with regards to their usability in distributed systems. Many layered context-aware systems and frameworks have evolved during the last years. Most of them differ in functional range, location and naming of layers, the use of optional agents or other architectural concerns. Besides these adaptations and modifications, a common architecture in modern context-aware applications is identifiable when analysing the various design approaches.

As mentioned above, a separation of detecting and using context is necessary to improve extensibility and reusability of systems. The following layered conceptual architecture, as depicted in Figure 1, augments layers for detecting and using context by adding interpreting and reasoning functionality (Ailisto et al., 2002; Dey and Abowd, 2001).

**Figure 1** Layered conceptual framework for context-aware systems

The first layer consists of a collection of different sensors. It is notable that the word 'sensor' not only refers to sensing hardware but also to every data source which may provide usable context information. Concerning the way data is captured; sensors can be classified in three groups (Indulska and Sutton, 2003).
• **Physical sensors.** The most frequently used type of sensors are physical sensors. Many hardware sensors are available nowadays which are capable of capturing almost any physical data. Table 1 shows some examples of physical sensors (Schmidt and van Laerhoven, 2001).

• **Virtual sensors.** Virtual sensors source context data from software applications or services. For example, it is possible to determine an employee’s location not only by using tracking systems (physical sensors) but also by a virtual sensor, e.g., by browsing an electronic calendar, a travel-booking system, emails etc., for location information. Other context attributes that can be sensed by virtual sensors include, e.g., the user’s activity by checking for mouse-movement and keyboard input.

• **Logical sensors.** These sensors make use of a couple of information sources, and combine physical and virtual sensors with additional information from databases or various other sources in order to solve higher tasks. For example, a logical sensor can be constructed to detect an employee’s current position by analysing logins at desktop PCs and a database mapping of devices to location information.

### Table 1

<table>
<thead>
<tr>
<th>Type of context</th>
<th>Available sensors</th>
</tr>
</thead>
<tbody>
<tr>
<td>Light</td>
<td>Photodiodes, colour sensors, IR and UV-sensors etc.</td>
</tr>
<tr>
<td>Visual</td>
<td>Various cameras</td>
</tr>
<tr>
<td>Audio</td>
<td>Microphones</td>
</tr>
<tr>
<td>Motion, acceleration</td>
<td>Mercury switches, angular sensors, accelerometers, motion detectors, magnetic fields</td>
</tr>
<tr>
<td>Location</td>
<td>Outdoor: Global Positioning System (GPS), Global System for Mobile Communications (GSM); Indoor: Active Badge system, etc.</td>
</tr>
<tr>
<td>Touch</td>
<td>Touch sensors implemented in mobile devices</td>
</tr>
<tr>
<td>Temperature</td>
<td>Thermometers</td>
</tr>
<tr>
<td>Physical attributes</td>
<td>Biosensors to measure skin resistance, blood pressure</td>
</tr>
</tbody>
</table>

The second layer is responsible for the retrieval of raw context data. It makes use of appropriate drivers for physical sensors and APIs for virtual and logical sensors. The query functionality is often implemented in reusable software components which make low-level details of hardware access transparent by providing more abstract methods such as getPosition(). By using interfaces for components responsible for equal types of context these components become exchangeable. Therefore, it is possible, for instance, to replace a RFID system by a GPS system without any major modification in the current and upper layers.

The Preprocessing layer is not implemented in every context-aware system but may offer useful information if the raw data are too coarse grained. The preprocessing layer is responsible for reasoning and interpreting contextual information. The sensors queried in the underlying layer most often return technical data that are not appropriate to use by application designers. Hence this layer raises the results of layer two to a higher abstraction level. The transformations include extraction and quantisation operations. For example, the exact GPS position of a person might not be of value for an application but the name of the room the person is in, may be.

In context-aware systems consisting of several different context data sources, the single context atoms can be combined to high-level information in this layer. This process is also called ‘aggregation’ or ‘composition’. A single sensor value is often not important to an application, whereas combined information might be more precious and accurate. In this vein, a system is able to determine, e.g., whether a client is situated indoor or outdoor by analysing various physical data like temperature and light or whether a person is currently attending a meeting by capturing noise level and location. To make this analysis work correctly a multitude of statistical methods are involved and often some kind of training phase is required.

Obviously, this abstraction functionality could also be implemented directly by the application. But due to a couple of reasons this task should better be encapsulated and moved to the context server. The encapsulation advances the reusability and, hence, eases the development of client applications. And by making such aggregators remotely accessible the network performance increases (as clients have to send only one request to gain high-level data instead of connecting to various sensors) and limited client resources are saved.

The problem of sensing conflicts that might occur when using several data sources has to be solved in this layer as well. For example, when a system is notified about a person’s location by the coordinates of her mobile phone and by a camera spotting this person, it might be difficult to decide what information to use. Often this conflict is approached by using additional data like time stamps and resolution information.

The fourth layer, Storage and Management, organises the gathered data and offers them via a public interface to the client. Clients may gain access in two different ways, synchronous and asynchronous. In the synchronous manner the client is polling the server for changes via remote method calls. Therefore, it sends a message requesting some kind of offered data and pauses until it receives the server’s answer. The asynchronous mode works via subscriptions. Each client subscribes to specific events it is interested in. On occurrence of one of these events, the client is either simply notified or a client’s method is directly involved using a call back. In the majority of cases the asynchronous approach is more suitable due to rapid changes in the underlying context. The polling technique is more resource
intensive as context data has to be requested quite often and
the application has to prove for changes itself, using some
kind of context history.

The client is realised in the fifth layer, the Application layer. The actual reaction on different events and context-instances is implemented here. Sometimes information retrieval and application specific context management and reasoning is encapsulated in form of
agents, which communicate with the context server and act
as an additional layer between the preprocessing and the
application layer (Chen, 2004). An example for context
logic at the client side is the display on mobile devices: as a
light sensor detects bad illumination, text may be displayed
in higher colour contrast.

All the systems, we analysed in this paper implement
most of the layers of the conceptual framework presented
above.

2.2 Context models

A context model is needed to define and store context data
in a machine processable form. To develop flexible and
useable context ontologies that cover the wide range
of possible contexts is a challenging task. Strang and
Linnhoff-Popien (2004) summarised the most relevant
context modelling approaches, which are based on the data
structures used for representing and exchanging contextual
information in the respective system.

- **Key-Value models.** These models represent the simplest
data structure for context modelling. They are
frequently used in various service frameworks, where
the key-value pairs are used to describe the capabilities
of a service. Service discovery is then applied by using
matching algorithms which use these key-value pairs.

- **Markup scheme models.** All markup based models use a
hierarchical data structure consisting of markup tags
with attributes and content. Profiles represent typical
markup-scheme models. Typical examples for such
profiles are the Composite Capabilities/Preference
Profile (CC/PP) (W3C, 2004a) and User Agent Profile
(UAProf) (Wapforum, 2001), which are encoded in
RDF/S. Various other examples can be found in Strang

- **Graphical models.** The Unified Modelling Language
(UML) is also suitable for modelling context. Various
approaches exist where contextual aspects are modelled
in by using UML, e.g., Sheng and Benatallah (2005).
Another modelling approach includes an extension to
the Object-Role Modelling (ORM) by context
information presented in Hendricksen et al. (2003).

- **Object oriented models.** Modelling context by using
object-oriented techniques offers to use the full power
of object orientation (e.g., encapsulation, reusability,
inheritance). Existing approaches use various objects to
represent different context types (such as temperature,
location, etc.), and encapsulate the details of context
processing and representation. Access the context and
the context processing logic is provided by well-defined
interfaces. Hydrogen (Hofer et al., 2002) uses such an
object-oriented example. We explain the system in
more detail in Section 3.

- **Logic based models.** Logic-based models have a high
degree of formality. Typically, facts, expressions and
rules are used to define a context model. A logic based
system is then used to manage the aforementioned
terms and allows to add, update or remove new facts.
The inference (also called reasoning) process can be
used to derive new facts based on existing rules in the
systems. The contextual information needs to be
represented in a formal way as facts. One of the first
approaches was published by McCarthy and Buvac
(1997).

- **Ontology based models.** Ontologies represent a
description of the concepts and relationships.
Therefore, ontologies are a very promising instrument
for modelling contextual information due to their high
and formal expressiveness and the possibilities for
applying ontology reasoning techniques. Various
context-aware frameworks use ontologies as underlying
context models. We describe some of them in
Section 3.

The conclusion of the evaluation presented in Strang and
Linnhoff-Popien (2004), based on six requirements, show
that ontologies are the most expressive models and fulfil
most of their requirements. Korpipää et al. (2003) present
some requirements and goals having designed a context
ontology:

- **simplicity:** the used expressions and relations should be
as simple as possible to simplify the work of
applications developers

- **flexibility and extensibility:** the ontology should support
the simple addition of new context elements and
relations

- **genericity:** the ontology should not be limited to special
kind of context atoms but rather support different types
of context

- **expressiveness:** the ontology should allow to describe
as much context states as possible in arbitrary detail.
Numerous tools are available to define declarative representations and to publish and share ontologies developed by the World Wide Web Consortium, e.g., the Resource Description Language (RDF) (W3C, 2000) and the Web Ontology Language (OWL) (W3C, 2004b). A single context atom can be described with a couple of attributes. The two most obvious are:

- **Context type.** The context type refers to the category of context such temperature, time, speed, etc. This type information may be used as a parameter for a context query or a subscription, e.g., `subscribeToChanges` ('temperature'). It is important to use meaningful type names, hence, as the system grows, some names might not be unique anymore. For example the type position may belong to a mobile device or a user. One solution for creating well-structured type names is the use of cascaded names (Korpipää and Mäntyjärvi, 2003) as shown in Table 2.

- **Context value.** Context value means the raw data gathered by a sensor. The unit depends on the context type and the applied sensor, e.g., degree Celsius, miles per hour, etc.

In most cases, context type and context value are not enough information to build a working context-aware system. Additional attributes that might be useful include:

- **Time stamp.** This attribute contains a date/time-value describing when the context was sensed. It is needed e.g., to create a context history and deal with sensing conflicts.

- **Source.** A field containing information how the information was gathered. In case of a hardware sensor it might hold the ID of the sensor and allow an application to prefer data from this sensor.

- **Confidence.** The confidence attribute describes the uncertainty of this context type. Not every data source delivers accurate information, e.g., location data suffers inaccuracy depending on the used tracking tool.

### Table 2  Example context vocabulary

<table>
<thead>
<tr>
<th>Context type</th>
<th>Context</th>
</tr>
</thead>
<tbody>
<tr>
<td>Environment:Temperature</td>
<td>Cold</td>
</tr>
<tr>
<td>Environment:Temperature</td>
<td>Normal</td>
</tr>
<tr>
<td>Environment:Temperature</td>
<td>Hot</td>
</tr>
<tr>
<td>Environment:Light:Source</td>
<td>50 Hz</td>
</tr>
<tr>
<td>Environment:Light:Source</td>
<td>60 Hz</td>
</tr>
<tr>
<td>Environment:Light:Source</td>
<td>NotAvailable</td>
</tr>
<tr>
<td>Device:Activity:Placement</td>
<td>AtHand</td>
</tr>
<tr>
<td>Device:Activity:Placement</td>
<td>NotAtHand</td>
</tr>
</tbody>
</table>

Part of a flexible context model is an extendable context vocabulary to deal with abstract descriptions rather than technical data. It simplifies the description of various context atoms and context instances. Table 2 shows a small part of an example vocabulary from Korpipää and Mäntyjärvi (2003). Notice that not all contexts have to be available at a time. In contrast to temperature, a light source is not always measurable.

### 3 Existent systems and frameworks

In this section, we discuss three types of existing context-aware systems. Firstly, we briefly describe a special case of context-aware systems, the so-called location-aware systems. Secondly, we describe an existing context-aware system assisting hospital information management. Thirdly, we draw our main focus on context-aware frameworks, as basic building block for context-aware systems and applications.

#### 3.1 Location-aware systems

Context-aware systems dealing with location information are widespread and the demand for them is growing due to the increasing spread of mobile devices. Examples for location-aware systems are various tourist guide projects where information dependent to the current location is displayed. Other examples can be found in Espinoza et al. (2001), Priyantha et al. (2000), Burrell and Gay (2002) and Kerer et al. (2004). A couple of different location aware infrastructures are available in order to collect position data. These include GPS satellites, mobile phone towers, badge proximity detectors, cameras, magnetic card readers, barcode readers, etc. These sensors can provide either position or proximity information. In addition, they differ in price and accuracy. Some need a clear line of sight, other signals can travel through walls, etc. As a detailed example, we introduce an indoor location sensing system from Harter et al. (2002), who describes a location-aware system using ultrasonic technique. To each entity (person or equipment), which should be detectable, a small sending unit called *bat* is attached. These bats have globally unique identifiers and contain *ultrasonic transducers*. To monitor the signals sent by the bats, receivers are installed at the rooms ceilings and connected by a wired network. The third hardware type needed is a *base station*. It periodically sends radio messages with specific bat ids and resets the receivers. A corresponding bat reacts by emitting an ultrasonic impulse which is caught by the receivers. By recording the time of arrival of signals the distance between the bat and the receiver can be calculated. The bat’s exact position is then determined by using multilateration (an extension of trilateration). A challenge the authors where confronted with due to the use of ultrasonic technique was the incorrect measurement because of unwanted reflections of the signals. The problem was solved by using a statistical outlier rejection algorithm to improve the accuracy of the calculated positions.
3.2 Context-aware systems

The systems named in the prior chapter use only one aspect of context, namely location information. The use of different types of context atoms such as noise, light and location allows the combination to high-level context objects. These elements are necessary to build more adaptive, useful and user-friendly systems. As an example for this kind of context-aware infrastructures serves the system presented by Munõz et al. (2003), which extends the instant messaging paradigm by adding context-awareness to support information management within a hospital setting. All users (in this case physicians, nurses etc.) are equipped with mobile devices to write messages that are sent when a specified set of circumstances is satisfied. For example, a user can formulate a message that should be delivered to the first doctor that enters room number 108 after 8 am. The contextual elements this system is aware of include location, time, roles and device state. Its context functionality is moved to agents which include three modules (layers). The perception module gathers raw context information from data sources (sensors, users, other agents, the server). The perception module gathers raw context information from data sources (sensors, users, other agents, the server). The reasoning module governs the agents’ actions and finally the action module triggers a user-specified event. All messages between agents are XML encoded.

3.3 Context-aware frameworks

Context-aware systems capable of dealing with special types of context are well-suited for specific conditions, e.g., in hospital scenarios. These systems can be optimised for the situations they are used in. They do not have to be flexible and extensible. To actually simplify the development of context-aware applications an abstract framework is needed. Such a generic infrastructure not only provides a client with access to retrieve context data, it also permits the simple registration of new distributed heterogeneous data sources. In this section different context-aware frameworks are introduced and compared based on various design criteria (architecture, resource discovery, sensing, context model, context processing historical context data, security and privacy).

3.3.1 Architectures

The most common design approach for distributed context-aware frameworks is a classical hierarchical infrastructure with one or many centralised components using a layered architecture as presented in Section 2. This approach is useful to overcome memory and processor constraints of small mobile devices but provides one single point of failure and thereby lacks robustness. The architecture of the Context Managing Framework presented by Korpipää et al. (2003) is depicted in Figure 2. Four main functional entities comprise this context framework: the context manager, the resource servers, the context recognition services and the application.

Whereas the resource servers and the context recognition services are distributed components, the so-called context manager represents a centralised server managing a blackboard. It stores context data and provides this information to the client applications.

The Service-Oriented Context-Aware Middleware (SOCAM) project introduced by Gu et al. (2004a, 2004b) is another architecture for the building and the rapid prototyping of context-aware mobile services. It uses a central server as well, here called context interpreter, which gains context data through distributed context providers and offers it in mostly processed form to the clients. The context-aware mobile services are located on top of the architecture, thus, they make use of the different levels of context and adapt their behaviour according to the current context.

One further extensible centralised middleware approach designed for context-aware mobile applications is a project called Context-Awareness Sub-Structure (CASS) presented in Fahy and Clarke (2004). In Figure 3, the system architecture of CASS is presented. The middleware contains an Interpreter, a ContextRetriever, a Rule Engine and a SensorListener. The SensorListener listens for updates from sensors, which are located on distributed computers called sensor nodes. Then the gathered information is stored in the database by the SensorListener. The ContextRetriever is responsible for retrieving stored context data. Both of these classes may use the services of an interpreter. The ChangeListener is a component with communication capabilities that allows a mobile computer to listen for notification of context change events. Sensor and LocationFinder classes also have built-in communications capabilities. Mobile clients connect to the server over wireless networks. To reduce the impact of intermittent connections local caching is supported on the client side.
Context Broker Architecture (CoBrA) (Chen et al., 2003) is an agent based architecture for supporting context-aware computing in so-called intelligent spaces. Intelligent spaces are physical spaces (e.g., living rooms, vehicles, corporate offices and meeting rooms) that are populated with intelligent systems that provide pervasive computing services to users. Central to CoBrA is the presence of an intelligent context broker that maintains and manages a shared contextual model on the behalf of a community of agents. These agents can be applications hosted by mobile devices that a user carries or wears (e.g., cell phones, PDAs and headphones), services that are provided by devices in a room (e.g., projector service, light controller and room temperature controller) and web services that provide a web presence for people, places and things in the physical world (e.g., services keeping track of peoples and object whereabouts). The context broker consists of four functional main components: the Context Knowledge Base, the Context Inference Engine, the Context Acquisition Module and the Privacy Management Module. To avoid the bottleneck problem CoBrA offers the possibility of creating broker federations.

The Context Toolkit (Salber et al., 1999; Dey and Abowd, 2000a; Dey, 2001), another context-aware framework, takes a step towards a peer-to-peer architecture but it still needs a centralised discoverer where distributed sensor units (called widgets), interpreters and aggregators are registered in order to be found by client applications. The toolkits object-oriented API provides a superclass called BaseObject which offers generic communication abilities to ease the creation of own components.

Another framework based on a layered architecture is built in the Hydrogen project (Hofer et al., 2002). Its context acquisition approach is specialised for mobile devices. While the availability of a centralised component is essential in the majority of existent distributed content-aware systems, the Hydrogen system tries to avoid this dependency. It distinguishes between a remote and a local context. The remote context is information another device knows about, the local context is knowledge our own device is aware of. When the devices are in physical proximity they are able to exchange these contexts in a peer-to-peer manner via WLAN, Bluetooth, etc. This exchange of context information among client devices is called context sharing. Figure 4 shows the management of a device’s context which consists of its own local context and a set of remote contexts gathered from other devices. Both local and remote context are made up of context objects. The superclass ContextObject is extended by different context types, e.g., LocationContext, DeviceContext, etc. This approach allows the simple addition of new context types by specialising ContextObject. A context type has to implement the methods toXML() and fromXML() from the ContextObject class in order to convert the data from and to a XML stream.

The architecture consists of three layers which are all located on the same device (Figure 5). The Adaptor layer is responsible for retrieving raw context data by querying sensors. This layer permits a sensor’s concurrent use by different applications. The second layer, the Management layer, makes use of the Adaptor layer to gain sensor data and is responsible for providing and retrieving contexts. The so-called Context server offers the stored information via synchronous and asynchronous methods to the client applications. On top of the architecture is the Application layer, where the appliance code is implemented to react on specific context changes reported by the context manager. Due to platform and language independency, all inter-layer communication is based on a XML-protocol.
The **CORTEX** system is an example for a context-aware middleware approach. The architecture is based on the **Sentient Object Model** (Biegel and Cahill, 2004) which was designed for the development of context-aware applications in an ad-hoc mobile environment. The model’s special suitability for mobile applications depends on the use of STEAM, a location-aware event-based middleware service designed specifically for ad-hoc wireless networking environments.

A sentient object is an encapsulated entity consisting of three main parts, as depicted in Figure 6, *Sensory capture, Context hierarchy and Inference engine*. Via interfaces a sentient object communicates with sensors which produce software events and actuators which consume software events. As Figure 6 shows, sentient objects can be both producer and consumer of another sentient object. Own sensors and actuators are programmed using STEAM. For building sentient objects, a graphical development tool is available which allows developers to specify relevant sensors and actuators, define fusion networks, specify context hierarchies and production rules, without the need to write any code.

The **Gaia** project (Roman et al., 2002; Gaia Project, 2005), another middleware infrastructure, extends typical operating system concepts to include context-awareness. It aims at supporting the development and execution of portable applications for active spaces. **Gaia** exports services to query and utilise existing resources, to access and use current context, and it provides a framework to develop user-centric, resource-aware, multi-device, context-sensitive and mobile applications. The current system consists of the **Gaia** kernel and the application framework as depicted in Figure 7.

**Figure 6** The sentient object model

**Figure 7** Architecture of the Gaia system

In this paper, we focus on **Gaia’s** parts concerning context-awareness, namely the **Event Manager**, the **Context Service** and the **Context File System**. The Event Manager service is responsible for event distribution in the active space and implements a decoupled communication model based on suppliers, consumers, and channels. Each channel has one or more suppliers that provide information to the channel and one or more consumers that receive the information. The reliability is increased as suppliers are exchangeable. With the help of the **Context Service**, applications may query and register for particular context information and higher level context objects. Finally the **Context File System** makes personal storage automatically available in the users present location. It constructs a virtual directory hierarchy to represent context as directories, where path components represent context types and values. For example, to determine which files have the context of location == RM2401 && situation == meeting associated with them, one may enter the /location:/RM2401/situation:/meeting directory.

### 3.3.2 Resource discovery

As sensors in a distributed network may fail or new ones may be added, a discovery mechanism to search for and find appropriate sensors at runtime is important. For these purposes the **Context Toolkit** offers the already mentioned discoverer. The discoverer works as registry component which interpreters, aggregators and widgets have to notify about their presence and their contact possibilities. After registration the components are pinged to ensure that they are operating. If a component does not respond to a specified number of consecutive pings, the discoverer determines that the component is unavailable and removes it from its registry list. Customers may find appropriate components querying the discoverer either via a white page lookup (a search for the components name) or a yellow page lookup (a search for specific attributes). In case the lookup was successful the discoverer returns a handle to contact the context component.

**SOCAM** offers a discovery mechanism as well called **Service Locating Service**. In **Gaia** different context providers are stored in a registry component. A pure peer-to-peer context-aware system such as **Hydrogen** only uses local built-in sensors and does not connect to distributed sensors, therefore, no discovery mechanism is involved.

### 3.3.3 Sensing

The **Context Toolkit** authors presented a new approach to handle different data sources. Derived from the use of widgets in GUI development, they introduced so-called **context widgets** to separate applications from context acquisition concerns. In these widgets the complexity of sensing is hidden. Furthermore, they abstract the gained
context information (e.g., the accurate position of a person might not be of value but the application should be notified when this person enters another room) and as widgets are encapsulated software components, they are reusable. Each widget owns some attributes that can be queried by applications, e.g., the IdentityPresence widget, implemented by the authors, offers attributes such as its location, the last time a presence was detected and the identity of the last user detected. Beside the polling mechanism an asynchronous way of data retrieval is possible too: if an application subscribes to a widget, it is notified when the widgets context changes. The IdentityPresence provides the callbacks PersonArrives(location, identity, timestamp) and PersonLeaves (location, identity, timestamp) which are triggered when a person either arrives or leaves a room. The separation of acquisition and use of context permits a simple exchange of widgets since e.g., identity may be sensed in various ways like Active Badges, video recognition, etc. This manner of building reusable sensor units that make the action of sensing transparent to the customer (whether it is a centralised server or a distributed client component), became widely accepted in distributed context-aware systems: CASS applies sensor nodes, SOCAM uses context providers, the Context Managing Framework refers to resource servers and CoBrA makes use of context acquisition components.

3.3.4 Context model

An efficient model for handling, sharing and storing context data is essential for a working context-aware system. The Context Toolkit handles context in simple attribute-value-tuples, which are encoded using XML for transmission. As already described above Hydrogen uses an object-oriented context model approach with a superclass called ContextObject which offers abstract methods to convert data streams from XML representations to context objects and vice versa. More advanced ways of dealing with context data based on ontologies are found in SOCAM, CoBrA and the Context Managing Framework. The SOCAM authors divide a pervasive computing domain into several sub-domains, e.g., home domain, office domain etc., and define individual low-level ontologies in each sub-domain to reduce the complexity of context processing. Each of these ontologies implemented in OWL provides a special vocabulary used for representing and sharing context knowledge. CoBrA also uses an own OWL-based ontology approach, namely COBRA-Ont (Chen et al., 2003, 2004). Listing 1 shows parts of an COBRA-Ont example. The ontology structure and vocabulary applied in the Context Managing Toolkit are described in RDF. Parts of its vocabulary are used as an example in Table 2 in Section 2.

Listing 1  CORBA-ONT example

```xml
<loc:LocationContext>
  <rdf:type rdf:resource="#tme:InstantThing"/>
  <loc:locationContextOf>
    <per:Person>
      <per:name rdf:datatype="$xsd:string">Harry Chen</per:name>
    </per:Person>
  </loc:locationContextOf>
  <loc:locationWithin rdf:resource="#webgeo:Japan"/>
  <tme:at rdf:datatype="#xsd:dateTime">2004-02-23T11:23:00</tme:at>
</loc:LocationContext>
```

In Gaia context is represented in a special manner, namely through 4-ary predicates in the way Context(<ContextType>, <Subject>, <Relater>, <Object>) written in DAML+OIL. The <ContextType> refers to the type of context the predicate is describing, the <Subject> is the person, place or thing with which the context is concerned, and the <Object> is a value associated with the <Subject>. The <Relater> relates the <Subject> and the <Object> such as a comparison operator (=, >, or <), a verb, or preposition. An example for a context instance is Context(temperature, room 3231, is, 98 F). This syntax is used for both, representing context and for forming inference rules.

3.3.5 Context processing

As soon as the raw context data is sensed by a data source, it has to be processed as its customers mostly are rather interested in already interpreted and aggregated information than in raw, fine-grained data. Whereas context aggregation refers to the composition of context atoms either to collect all context data concerning a specific entity or to build higher-level context objects, context interpretation refers to the transformation of context data including special knowledge. These forms of context data abstraction ease the application designer’s work tremendously. The Context Toolkit offers facilities for both context aggregation and context interpretation. The context aggregators (former called context servers) are responsible for composing context of particular entities by subscribing to relevant widgets, context interpreters provide the possibility of transforming context, e.g., in a simple case returning the corresponding e-mail address to a passed name. Like widgets aggregators and interpreters inherit communication methods from the superclass BaseObject and have to be registered at the discoverer in order to be found. The Context Managing Framework presented by Korpip et al. (Figure 2) offers various processing facilities as well. The resource servers’ tasks are complex. First they gather raw context information by connecting to various data sources. After the preprocessing and feature abstraction crisp limits and fuzzy sets are used for quantisation. But now
the data are delivered by posting it to the context manager’s blackboard. The context recognition services are used by the context manager to create higher-level context objects out of context atoms. In this vein new recognition services are easy to add.

In SOCAM, the Context Reasoning Engine reasons over the knowledge base, its tasks include inferring deduced contexts, resolving context conflicts and maintaining the consistency of the context knowledge base. Different inference rules used by the reasoning engine can be specified. The interpreter is implemented by using Jena2 (Jena, 2005), a semantic web toolkit.

In CoBrA the so-called Inference Engine processes context data. The engine contains the Context Reasoning Module responsible for aggregating context information. It reasons over the Context Knowledge Base and deduces additional knowledge from information acquired from external sources.

In CASS deriving of high-level context is also based on an inference engine and a knowledge base. The knowledge base contains rules queried by the inference engine to find goals using the so-called forward chaining technique. As these rules are stored in a database separated from the interpreter neither recompiling nor restarting of components is necessary when rules change. Table 3 shows an example rule database entry containing criteria to display rather indoor than outdoor activities in a CASS based tour-guide application.

<table>
<thead>
<tr>
<th>Rain</th>
<th>Brightness</th>
<th>Temperature</th>
<th>Goal</th>
</tr>
</thead>
<tbody>
<tr>
<td>Wet</td>
<td>Dull</td>
<td>Cold</td>
<td>Indoor</td>
</tr>
</tbody>
</table>

In CORTEX, the whole context processing is encapsulated in Sentient Objects. The sensory capture unit performs sensor fusion to manage uncertainty of sensor data (sensing conflicts) and to build higher-level context objects. Different contexts are represented in a so-called context hierarchy together with specific actions to be undertaken in each context. Since only one context is active at any point in time (concept of the active context) the number of rules that have to be evaluated are limited. Thus efficiency of the inference process is increased. The inference engine component is based on C Language Integrated Production System (CLIPS). It is responsible for changing application behaviour according to the current context by using conditional rules. Gaia’s context processing is hidden in the Context Service Module allowing the creation of high-level context objects by performing first order logic operations such as quantification, implication, conjunction, disjunction, and negation of context predicates. One example of a rule is Context(Number of people, Room 2401, >, 4) AND Context(Application, Powerpoint, is, Running) ⇒ Context(Social Activity, Room 2401, Is, Presentation). Almost all current context-aware frameworks permit the aggregation and interpretation of raw context data. Only in a few cases the higher-level abstractions are handled by the application layer, such as in Hydrogen.

### 3.3.6 Historical context data

Sometimes it might be necessary to have access to historical context data. Such context histories may be used to establish trends and predict future context values. As most data sources constantly provide context data, the maintenance of a context history is mainly a memory concern, so a centralised high-resource storage component is needed. Since in a server-based architecture the context data provided by sensors has anyway to be stored at the server-side to offer it to customers, the majority of these systems has the facility to query historical context data. The Context Toolkit, CoBrA, CASS, SOCAM and CORTEX save sensed context data persistently in a database. A further advantage of using a database is the use of the Structured Query Language (SQL) which enables to read and to manipulate operations at a high abstraction level. In the CoBrA and the CASS architecture the persistent storage is called Context Knowledge Base. Additionally a set of APIs is offered to assert, delete, modify and query the stored knowledge. CASS uses its database not only to save context data but also to store domain knowledge and inference rules needed for creating high-level context. Due to limited memory resources a peer-to-peer network of mobile devices like Hydrogen is not able to offer persistent storage possibilities.

### 3.3.7 Security and privacy

As context may include sensitive information on people, e.g., their location and their activity, it is necessary to have the opportunity to protect privacy. For these purposes the Context Toolkit introduces the concept of context ownership. Users are assigned to sensed context data as their respective owners. They are allowed to control the other users’ access. New components involved in this access control are the Mediated Widgets, Owner Permissions, a modified BaseObject and Authenticators. The MediatedWidget class is an extension of a basic widget which contains a so-called widget developer specifying who owns the data being sensed. The Owner Permission is the component that receives permission queries and determines to grant or to deny access based on stored situations. These situations include authorised users, time of access etc. The modified BaseObject contains all the original methods augmented with identification mechanisms. Now applications and components have to provide their identity along with the usual request for information. Finally the authenticator is responsible for proofing the identity by using a public-key infrastructure. CoBrA includes an own flexible policy language to control context access, called Rei (Kagal et al., 2003). This policy language is modelled on deontic concepts of rights, prohibitions, obligations and dispensions and controls data access through dynamically modifiable domain dependent policy rules.
4 Discussion of approaches

In Table 4 we have summarised the main aspects of the discussed approaches. The architectural style of a context-aware system is mainly driven by the context acquisition method. The main criteria for a reasonable architectural approach is the separation of concerns between the context acquisition and the user components as proposed by Dey (2000). All the frameworks presented in this paper support this separation of concerns.

The sensing technology is implemented differently in every framework. It is important, that the concrete sensing mechanism is encapsulated in separate components, to support the aforementioned separation of concerns. Furthermore, it encapsulates sensing and allows to access the contextual data via defined interfaces. Currently, there is no standard description language or ontology for sensing contextual information from various sources to enable reuse across various middleware systems and frameworks. Therefore, proprietary solutions as used by the different frameworks, have emerged. SOCAM use the most sophisticated approach for sensing context information. External virtual sensors are consumed via web services (by using SOAP). Internal providers for querying sensors are consumed by using context events represented in OWL based on a predefined ontology.

The context model and the context processing logic supported by different frameworks is a major criteria for providing intelligent and adaptable context-aware services or applications. As mentioned before, ontologies provide a rich formalism for specifying contextual information. Based on such ontological models, highly sophisticated ontology reasoning engines can derive new concepts to adapt the service behaviour accordingly. The major drawback of the Context Toolkit is, therefore, its context model, a set of attribute-value tuples. The development of intelligent context processing and aggregation is limited due to the fact that such attributes do not have a meaning. Furthermore, using non-ontology based models requires a lot of programming effort and tightly couples the context model to the rest of the system. Moreover, the lack of declarative semantics of programs does not allow reasoning and knowledge sharing amount systems. For example, SOCAM uses a general upper ontology to specify basic common contextual properties and to refine this general ontology. Domain-specific ontologies can be defined to provide very fine grained possibilities for specifying and formalising context.

<table>
<thead>
<tr>
<th>Architecture</th>
<th>Sensing</th>
<th>Context model</th>
<th>Context processing</th>
<th>Resource discovery</th>
<th>Historical context data</th>
<th>Security and privacy</th>
</tr>
</thead>
<tbody>
<tr>
<td>CASS</td>
<td>Centralised middleware</td>
<td>Sensor nodes</td>
<td>Relational data model</td>
<td>Inference engine and knowledge base</td>
<td>n.a.</td>
<td>Available</td>
</tr>
<tr>
<td>CoBra</td>
<td>Agent based</td>
<td>Context acquisition module</td>
<td>Ontologies (OWL)</td>
<td>Inference engine and knowledge base</td>
<td>n.a.</td>
<td>Available</td>
</tr>
<tr>
<td>Context Toolkit</td>
<td>Widget based</td>
<td>Context widgets</td>
<td>Attribute-value tuples</td>
<td>Context interpretation and aggregation</td>
<td>Discoverer component</td>
<td>Available</td>
</tr>
<tr>
<td>CORTEX</td>
<td>Sentient object model</td>
<td>Context component framework</td>
<td>Relational data model</td>
<td>Service discovery framework</td>
<td>Resource management component framework</td>
<td>Available</td>
</tr>
<tr>
<td>Gaia</td>
<td>MVC (extended)</td>
<td>Context providers</td>
<td>4-ary predicates (DAML + OIL)</td>
<td>Context-service module (first-order logic)</td>
<td>Discovery service</td>
<td>Available</td>
</tr>
<tr>
<td>Hydrogen</td>
<td>Three layered architecture</td>
<td>Adapters for various context types</td>
<td>Object-oriented</td>
<td>Interpretation and aggregation of raw data only</td>
<td>n.a.</td>
<td>n.a.</td>
</tr>
<tr>
<td>SOCAM</td>
<td>Distributed with centralised server</td>
<td>Context providers</td>
<td>Ontologies (OWL)</td>
<td>Context reasoning engine</td>
<td>Service locating service</td>
<td>Available</td>
</tr>
</tbody>
</table>
Resource discovery mechanisms are currently rarely used in the presented frameworks. Such dynamic mechanisms are important, especially in a pervasive environment, where available sensors, the context sources, change rapidly (new ones are added or removed). SOCAM, for example, which is the only system that is based on a service-oriented architecture, offers a service locating service, which dynamically binds to available context providers. These providers can also be changed at runtime. The lack of resource discovery support is a major drawback of many frameworks, because it implies that the used context sources are stable and permanently available, which is not always the case in real-world applications. Erroneous behaviour of one or more context sources may lead to a decreased availability of the context-aware service or application.

Managing historical context data provides the ability to implement intelligent learning algorithms which allow to provide highly adaptable context-aware services. Furthermore, based on learning algorithms, contextual information can be predicted to proactively provide a certain set of services to the user. Many of the systems store contextual information but none of them do not use learning techniques to provide context-aware service proactively.

Another important aspect is security and privacy. Contextual information mostly considers user profile information and other sensitive information. Therefore, concepts are needed to express policies and to define ownership of context information. CoBrA uses the Rei policy language, to express (security) policies about contextual information. Gaia uses several mechanisms to define privacy restrictions and secure communication for tracking locations of people. The Context Toolkit implements the concept of context ownership, which is used to allow access to the context to the owner only. The other frameworks do not use security concepts, which is one of their major drawbacks. When dealing with sensitive data, secure connections for context acquisition as well as privacy of different user specific contextual information is very important.

5 Conclusions and future work

In this survey paper we described different design principles and context models for context-aware systems and presented various existent middleware and server-based approaches to ease the development of context-aware applications. The direct comparison of the named systems and frameworks shows their similarity concerning the layered structure. Especially remarkable is the strict division of the context data acquisition and use. Thus context sources become reusable and are able to serve a multitude of context clients. Although most authors refer to abstract context sources, the currently mainly used and tested sources are physical sensors. Virtual and logical sensors are capable of providing useful context data as well and should be more incorporated in ongoing research.

Other often disregarded aspects are security and privacy issues. These facets belong to the most important components of a context-aware system as the protection of sensitive context data must be guaranteed. Many systems totally lack security modules, others provide basic security mechanisms and only a few systems offer advanced and sufficient security options. Probably the main problem in the presented approaches is the variety of used context encodings and ways to find and access context sources. Every system and framework uses its own format to describe context and its own communications mechanisms. We believe that standardised formats and protocols are important for the enhancements of context-aware systems to make the development of context services the focus rather than the communication between context sources and users. In our opinion web service technologies seem to be an appropriate solution to achieve that aim as they provide standardised methods for service description and access.

Our future work in this area will investigate the use of service-oriented and autonomic computing concepts for building context-aware service frameworks. We believe that standardised technologies and protocols, such as WSDL and SOAP, could help to build more interoperable context aware services. Furthermore, the use of ontologies for building a context model is an important approach (as can be seen from existing approaches) to build more sophisticated algorithms, which derive new contextual knowledge or patterns to proactively aggregate new context-aware services. Autonomously orchestrating atomic context-aware services into higher level services based on context information and available QoS parameters provides high potential for offering more accurate services to the user.

References


A survey on context-aware systems


Context-aware applications (or the system infrastructure to support them), however, have never been widely available to everyday users. In this survey of research on context-aware systems and applications, we looked in depth at the types of context used and models of context information, at systems that support collecting and disseminating context, and at applications that adapt to the changing context. Through this survey, it is clear that context-aware research is an old but rich area for research. The difficulties and possible solutions we outline serve as guidance for researchers hoping to