16 MNISIKLIS: Indoor Location Based Services for All

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Abstract

MNISIKLIS is an integrated system aiming to provide universal, indoor location-based services focusing on navigation. This paper presents the overall MNISIKLIS architecture and certain implementation details. In the context of the Design for All approach, the system targets to the support of several types of users, including persons with disabilities as well as elderly, by exploiting multimodal interaction. Moreover, the system implements efficient path finding algorithms and provides advanced user experience through highly personalized services. MNISIKLIS adopts Semantic Web technologies (e.g., ontologies and reasoning methods) for representing and managing application models. Furthermore, MNISIKLIS exploits modern positioning techniques in order to achieve high quality positioning. The paper discusses the algorithms and the models that accommodate the services provided by the system. Additionally, an analysis of the positioning subsystem, the user interaction subsystem and the peripheral infrastructure is given. Hence, a new paradigm in the area of location-based systems is presented.

Keywords: indoor navigation, ontology, RFID, dead reckoning, multimodal interaction

16.1 Introduction

Nowadays, the increasing demand for advanced, personalized, context-aware, intelligent and “always available” applications has led to the convergence of information technology and telecommunication domains. As a result, concepts like pervasive
computing, context-awareness and artificial intelligence started to play an essential role to the application development community and researchers, thus, leading to the so-called ambient intelligence paradigm (Aarts et al. 2001).

Location-based services (LBS) constitute a popular domain of context-aware applications. Indoor pedestrian way-finding, in particular, is a very challenging area, mainly due to the unsuitability of the mature and widely established outdoor positioning technologies for use inside buildings. The Global Positioning System (GPS) is an excellent technology that can be used for the determination of absolute location in outdoor environments, but is almost useless indoors.

Furthermore, the vision of Design for All (Stephanidis 2001), which targets at enabling people of all ages and abilities to access services and environments in order to improve their quality of life, becomes more and more popular. The adoption of multimodal interfaces is central to this vision, since Ambient Intelligence aims to enhance human-machine interaction by placing the user at the centre of the computing environment.

This paper presents the MNISIKLIS system that provides real-time, indoor LBS to a wide range of users. The main novelties of the implemented system are listed below:
1. For the first time, to our knowledge, passive UHF Radio Frequency Identification (RFID) technology is used for proximity sensing.
2. Positioning is based on a multi-sensor fusion process, involving Wi-Fi positioning and the Dead Reckoning technique.
3. The system provides a multimodal user interface, thus implementing the Design for All paradigm.
4. The implemented services heavily rely on semantic models and knowledge reasoning techniques. Hence, the overall service logic is highly human-centered.

The rest of this paper is organized as follows. Section 16.2 discusses related work in the area of indoor location-based systems. A generic architecture of the system and the description of the services supported are provided in Section 16.3. Sections 16.4–16.6 present the main subsystems of MNISIKLIS and give some implementation details. Finally, some concluding remarks are provided in the last section of the paper.

16.2 Related Work

In this section, we present some indicative systems for location-aware applications. Most of them mainly focus on navigation services for either indoor or outdoor environments and, thus, do not support the full spectrum of LBS services as MNISIKLIS
does. For example, most of them either just implement a navigation service without providing multimodal interaction or do not exploit efficient positioning methods.

iNAV (Kargl et al. 2007) is a navigation framework aiming to providing guidance in indoor environments. It exploits the COMPASS (Kargl & Bernauer 2005) middleware in order to achieve localization and facilitate service discovery. Nevertheless, iNAV mainly targets at typical users, since it does not provide any advanced user interaction features. CoINS (Lyardet et al. 2006) is a context-aware indoor navigation system that involves a complex mechanism for spatial modeling and room detection. With regard to the route selection process, the system exploits an optimized version of the Dijkstra algorithm. However, CoINS does not currently support any multimodal interfaces to support diverse user classes. Another pedestrian navigation and exploration system is presented in (Wasinger et al. 2003). The system exploits GPS, infrared beacons and a magnetic compass as positioning technologies and emphasizes on supporting different modalities. Although, the system involves multimodal interaction, it does not investigate the concept of interaction with multiple devices and it implements only a core navigation service.

IRREAL (Baus et al. 2002) is another indoor navigation system, based on infrared beacons, that adapts the presentation of route directions to the specific device capabilities. The application does not fully support interaction with disabled users. A pedestrian navigation system that investigates complex aspects like multi-criteria path selection and integrated positioning for both indoor and outdoor environments is described in (Gartner et al. 2004). Although the system supports audio guidance, it is not targeting to disabled users. In (Bikakis et al. 2006), the authors exploit Semantic Web technologies in order to develop a context ontology for supporting indoor navigation services. However, this approach does not examine in detail the efficiency of positioning techniques and the presentation of path instructions to the user.

16.3 System Architecture and Implemented Services

The MNISIKLIS platform includes three main subsystems (as shown in Figure 16.1):

- **Positioning Subsystem.** It comprises the overall equipment and the algorithms used to estimate the user’s position. Specifically, it consists of sensors and positioning techniques, the location fusion component and the interfaces between them.
- **Middleware.** The middleware consists of the services and the navigation algorithms developed as well as the application models. It is also responsible for gluing together the other subsystems.
- **User Interaction Subsystem.** The user interaction subsystem involves the user device (hardware and software), the input/output interfaces and the content selection and representation algorithms.
Apart from these core platform ingredients, a peripheral infrastructure for LBS content provisioning and management has been developed. Such infrastructure includes a GIS system and a Semantic Content Management System (SCMS). In the following sections, we elaborate on the main components of each subsystem.

In the context of the MNISIKLIS project, the following services have been implemented:

- **Static Navigation.** The user asks the system to determine a “suitable” route to a certain destination. The service takes into consideration the application models (e.g., user profile) in order to compute the “best” path and guide the user with the most suitable landmarks.

- **Dynamic Navigation.** An extension of static navigation that periodically traces the user position. In case it detects a significant deviation of the user from the predetermined path, it helps her to find her way by providing more detailed information.

- **Where-Am-I.** The user asks for her current position inside a building. The system responds by providing details about the last known user position. The information about a specific location is organized and presented in different levels of detail.

![MNISIKLIS Architecture](image)
• Exploration. While the user is moving inside the building, the system provides (“pushes”) information about the nearest locations that she may be interested in. Such Points Of Interest (POIs) may have been explicitly stated by the user or not (e.g., significant exhibits in a museum).

• Nearest POIs. The system computes the POIs that are closer to the user. The main difference from the exploration service is the push-based nature of the latter. Hence, the system may always return points that are not located close to the user.

### 16.4 Positioning Subsystem

#### 16.4.1 Sensing Technologies

Accurate indoor positioning can be achieved through two main technology approaches. The first approach is based on radio technology (WLAN, Bluetooth) as well as infrared and ultrasound beacons (Schiller & Voisard 2004). The second approach exploits inertial sensors (accelerometers and gyroscopes) and non-inertial sensors (magnetic compasses) along with appropriate dead reckoning algorithms. Proximity sensing, based on RFID technology has been recently introduced (Koide & Kato 2006) as an alternative. In the MNISIKLIS platform we have adopted three technologies: UHF RFIDs, Dead Reckoning (DR) for pedestrian users (Fang et al. 2005, Dippold 2006) and WLAN Received Signal Strength Indicator (RSSI). This section describes the main sensing technologies that compose the proposed positioning subsystem.

#### 16.4.1.1 RFID Technology

UHF RFIDs (868 MHz) used in the system provide longer ranges (in the order of 1m) with only 50 mW of RF power compared with other approved RFID frequency bands. The RFID reader is the nano-UHF from TAGSENSE, while the RFID tags are manufactured by Texas Instruments and support the EPC1 GEN2 standard for read-write capabilities. In the current implementation an identity has been stored in the tag’s EEPROM memory and has been associated with a specific building location in the location server database. The navigated person carries a mobile RFID reader which continuously scans for tags and transfers the specific identity of the proximal RFID to the PDA.

#### 16.4.1.2 Dead Reckoning

In the MNISIKLIS platform a prototype dead reckoning system has been developed, based on a commercially available high performance 3-axis electronic compass
(Honeywell HM3300) and a 3-axis accelerometer (Hitachi H48C). The 3-axis magnetometer is used for heading information independently from sensor orientation. For the estimation of distance traveled from a person, various algorithms have been proposed for step detection and step length estimation based on 3-axis accelerometers (Weinberg 2008, Ladetto et al. 2000). The sensor measurements are collected at the sensor unit which is attached to the user’s belt (see Figure 16.2) and are, subsequently, transferred via a Bluetooth link to a PDA for processing by the dead-reckoning algorithm.

As a first preprocessing step, the raw accelerometer data are low-pass filtered with a moving average filter. The acceleration measurements on the Z (gravitational) axes or from any of the horizontal acceleration axis can be used in the DR algorithm.

The step length, for walking on a flat path, is influenced by the walking frequency and the variance of the accelerometer during one step (Ladetto et al. 2000). Specifically, the predicted step length is computed using the following step model:

\[
\text{step length of } k \text{ step} = A + B \cdot f_k + C \cdot \text{var}_k
\]  

(16.1)

where

- \(A, B, C\) are coefficients estimated through linear regression,
- \(f_k\) is the walking frequency at time \(t_k\) obtained by the equation \(f_k = \frac{1}{t_k - t_{k-1}}\), with \(t_i\) being the detecting time of the \(i^{th}\) step, and
- \(\text{var}_k\) is the variance of the acceleration in the direction of movement, during the \(k^{th}\) step: \(\text{var}_k = \frac{\sum_{t=k-1}^{k} a_t - \text{mean}(a_k)}{N}\), where \(a_t\) is the acceleration at time \(t\), \(\text{mean}(a_k)\) and \(N\) is the mean value of the accelerations data and the number of the acceleration samples during the \(k^{th}\) step, respectively.
Finally, the walking distance of m steps on a straight flat path is obtained by the equation:

$$\text{walking distance} = \sum_{i=1}^{m} (A + B \cdot f_i + C \cdot \text{var}_i)$$

(16.2)

The position estimations of the dead reckoning algorithm are transferred from the PDA using a WLAN to the location server for further processing (fusion).

### 16.4.2 Location Server

The location server is the core component of the positioning subsystem. It processes the data received from the mobile device and generates the final estimation of the user’s current position. Hence, it is equipped with suitable software which handles the communication with the mobile device, the analysis of collected data and the implementation of inference algorithms (Figure 16.3).

The server takes advantage of a database that stores the spatial data and the history of estimations. Below, we provide a more detailed presentation of each component.

#### 16.4.2.1 Communication Component (CC)

The role of this component is the communication with the mobile device. It receives the requests with the collected data and checks their validity (e.g., if the measured value from a WLAN access point is in the range of predefined min-max values).

![Fig. 16.3. Location Server Architecture.](image_url)
The form of the data vector is as follows:

\[
<userId, IE \_ Id_1 = value_1, \ldots, IE \_ Id_N = value_N, \ X = x_i, Y = y_i, Z = z_i, Orientation >
\]

where
- \( userId \) is the unique identifier of the user,
- \( IE \_ Id_i = value_i \) is the pair (Infrastructure Element Identifier, value),
- \( X, Y, Z \) are the coordinates of the estimated location as they have been computed by the DR algorithms (Z denotes the floor level),
- \( Orientation \) is the compass’ measurement for the orientation of user (deviation from the North).

After that, CC proceeds to the quantization of the WLAN RSSI values at N discrete levels: \( S_1, S_2 \ldots S_N \). If, for example, the value from the access point with \( IE \_ Id Id_1 \) is between -30 dBm and -47.5 dBm, the min-max values are -30dBm and -100dBm respectively and \( N = 4 \), the value \( S_1 \) is assigned as the observed value of this access point (assumption of uniformly distributed values).

### 16.4.2.2 First Level Fusion Engine

This engine uses multiple readings from WLAN access points and RFID tags and it is based on a Dynamic Bayesian Network (DBN), which is used for the location inference (Sekkas et al. 2006). The structure and the required probability distributions for the DBN are loaded from an XMLBIF file (XMLBIF 1998) which is the result of a training phase. The output of this engine consists of a vector of probabilities \( \{P_1, P_2, \ldots, P_N\} \) for each symbolic location \( L_1, L_2, \ldots, L_N \) at time \( t \). The probability \( P_{bi} \) represents the probability of the user being at location \( L_i \) given her previous location \( L_j \) at time \( t-1 \) and given the values/observations, \( O^{(t)} \) of the aforementioned elements/sensors associated with this user. Such probability is calculated as follows:

\[
P_{bi} = P(L_i^{(t)} | L_j^{(t-1)}, O^{(t)}) = \frac{P(L_i^{(t)} \& L_j^{(t-1)} \& O^{(t)})}{P(L_j^{(t-1)} \& O^{(t)})} \quad (16.3)
\]

### 16.4.2.3 DR Data Converter

The DR data converter takes as input the coordinates \( (x, y) \) of the estimated location by the DR algorithms executed at the mobile device. Data processing comprises the phase of computation of distances and the phase of homogenization.

In the first phase, for each location \( L_i \) with coordinates \( (x_i, y_i) \) we compute the 2D Euclidean distance from the estimated location, as the DR component cannot locate the user between successive floors.
In the homogenization phase the converter applies a simple transformation rule in order to calculate the probability \( P_{ci} \) of the user being at location \( L_i \). The distance of the location \( L_i \) from the estimated location is not taken into account for the calculation of probabilities when it exceeds a predetermined threshold \( d_i \). For locations of distance \( d_i \) greater than \( d_i \) or locations found in a different floor than the estimated \((x, y)\) coordinates, we assign zero probability. For the remaining locations we compute the probability \( P_{ci}' \) according to the equation:

\[
P_{ci}' = \frac{1}{d_i^2} , \; i = 1,2,...,N
\]  

(16.4)

which indicates that this probability is reversely proportional to the square distance. The final value of the probability \( P_{ci} \) is calculated after the normalization procedure so that the sum of probabilities equals unity:

\[
P_{ci} = \frac{P_{ci}'}{\sum_{i=1}^{N} P_{ci}'}, \; i = 1,2,...,N
\]

(16.5)

Thus, the converter produces as output the vector of probabilities \(<P_{c1}, P_{c2},...,P_{cN}>\) which feed the 2nd level fusion engine.

**16.4.2.4 Second Level Fusion Engine**

The second level fusion engine takes as input the probabilities calculated for each symbolic location by the first level fusion engine and the DR converter and produces the final estimation of the current position of the user. Specifically, it uses the probability vectors \(<P_{b1}, P_{b2},...,P_{bN}>\) and \(<P_{c1}, P_{c2},...,P_{cN}>\) and combines them appropriately so that it calculates the final probability \( P_i \) for location \( L_i \). The contribution of probabilities \( P_{bi} \) and \( P_{ci} \) is determined by the corresponding weights \( w_b \) and \( w_c \) \((w_b + w_c = 1)\) according to the following combination rule

\[
P_i = w_b * P_{bi} + w_c * P_{ci} , \; i = 1,2,...,N
\]

(16.6)

The final estimated location is calculated as the location with the highest probability and is stored in the database. Moreover, it is sent to the mobile device as feedback for the DR algorithms and updates the DBN in the first level fusion engine.

**16.5 Middleware**

**16.5.1 Application Models**

Four ontologies are the basis for MNISIKLIS: i) the spatial ontology (Indoor Navigation Ontology – INO), ii) the User Navigation Ontology (UNO), iii) the
Device Ontology (DO), and iv) the Content Ontology (CO). The instances of the aforementioned ontologies are connected through semantic relationships in order to provide more intelligent location services.

- **Indoor Navigation Ontology (INO):** The spatial ontology is an extended version of the INO (Tsetsos et al. 2006; INO 2008), based on the OWL-DL language (OWL 2004). Specifically, it describes concepts and relationships that correspond to every basic spatial element typically found in indoor environments.

- **User Navigation Ontology (UNO):** UNO (Kikiras et al. 2006) is an ontology that contains the necessary concepts and relations to define the main characteristics and abilities of users, facilitating the provision of highly personalized services. Additionally, UNO design is based on the international standard defined by the World Health Organization (WHO) (ICF 2001).

- **Device Ontology (DO):** Our approach adopts a device ontology in order to represent basic features and the functionality supported by various user devices (e.g., mobile phones, PDAs, headphones). The knowledge captured by the ontology refers to hardware capabilities (e.g., display size, resolution) as well as device supported modalities (e.g., input/output modes).

- **Content Ontology (CO):** Content Ontology describes general categories of content with their properties and relations (Figure 16.4). CO includes two main categories of concepts. *Digital Content* is the high level concept including properties related to the general characteristics of the described content. Low level concepts describe each specific content type (e.g., Text, Image, Video).

Fig. 16.4. Content Ontology concepts and their properties.
16.5.2 LBS Algorithms

In the context of MNISIKLIS, a hybrid rule-based navigation algorithm has been designed and implemented. The core navigation algorithm (see code below) involves two main steps to compute the best traversable path for each user:

**Creation of user-compatible building graph.** In this step, a number of path elements incompatible with the user profile (capabilities) are excluded from the building graph. Specifically, several disability rules are applied to the INO instances with respect to the user profile, eliminating the path elements that cannot be traversed by a certain user. These rules are expressed through the SWRL language (Horrocks et al. 2004). For example, all stairway elements should be excluded in case of a wheel-chaired user. The following rule captures that knowledge:

\[
\text{uno:WheelchairedUser}(u) \land \text{ino:Stairway}(s) \rightarrow \text{ino:isExcludedFor}(s,u)
\]

Thus, the overall graph of the building is reduced to a user compatible form.

**Path computation.** This step takes into consideration several metrics in order to compute the best traversable path for each user. Since, in personalized human navigation systems, we expect to have many different constraints (per user) during route computation, we decided not to adopt such a “monolithic” approach. As a result, the algorithm assigns bonus/penalty points to the route elements according to several parameters. The main measures that affect the route computation are:

- **The route complexity.** The complexity of the route instructions is a very important factor in human navigation, since people usually do not follow the shortest but the simplest path. The simplest-path algorithm proposed in (Duckham & Kulik 2003) computes the “easiest-to-describe” path in a graph and has similar computational complexity with a shortest path algorithm.

- **The Euclidean route distance.** Since the distance of a route is usually the main criterion in human navigation, we also take into account the Euclidean distance of a path.

- **The user profile.** User capabilities and preferences play significant role during the path computation process. For example, in the case of a user that prefers to use stairs, the system would penalize paths including the elevator.
Foreach Path Element PE
    Assign bonus/penalty value to PE, according to perceptual rules and user preferences
Endfor
Compute the total path length
TotalPathRank = f(path length, bonus vector, penalty vector)
Endfor
Return the path with the maximum TotalPathRank
End

The aforementioned core navigation algorithm was developed for the static navigation service. The rest of the services (see Section 16.3) were implemented as extensions/modifications of this core algorithm. In this section, we describe in detail the most challenging implementation parts of these services.

Dynamic navigation was the most computationally-hard service since it should spontaneously provide a new route description to user in case the system detects her “far enough” from the initially computed path. It also imposes strict deadlines, since it has to provide useful information to the user at minimal response times. Specifically, the extended algorithm of dynamic navigation does not guide the user back to the initial path, but it computes a totally new path description given the current user trace.

Exploration was also a time critical service, since it had to inform user about nearby location points or regions before she passes them. To achieve acceptable response times, the implemented algorithm takes advantage of the graph structure of the building by examining only the locations that are close to the user position.

Finally, the “K-Nearest POIs” was a memory consuming service, as it computes and manages paths for all POIs of the user. Once again, the graph structure of the building facilitated the service implementation.

16.5.3 Peripheral Systems

16.5.3.1 GIS-Based Ontology Population

In our approach, we use a GIS system in order to define points on the map of a specific building. We categorize points according to the concepts defined in INO. A layered architecture is adopted (Figure 16.5). The blueprints of each floor constitute the lower layer and are used as a reference for the rest of the layers. Based on each floor’s map, corridors and other spatial elements are defined. In our approach, 16 layers are used for each floor. There are layers related to navigational or transition points as well as layers devoted to facilitate users’ guidance. Moreover, auxiliary
points are defined close to transition points (i.e., room entrances) for orientation purposes. Further details on the overall procedure are provided in (Kolomvatsos et al. 2007).

Once the GIS data are in place, one can proceed with the actual ontology population. Each of the aforementioned layers is exported as a shape file (i.e., a popular GIS data storage format) which is subsequently imported in a table of a spatial database. Subsequently, a series of algorithms are used to create the instances based on INO concepts.

16.5.3.2 Semantic Content Management System (SCMS)

Since, the main goal of MNISIKLIS platform is to provide to users services and content according to their location, a Content Management System (CMS) was developed for managing content. We should remind that CMSs are systems used for the creation, organization and the manipulation of digital content (Boiko 2005). Semantic Web technologies can provide efficiency in content retrieval and interoperability. Hence, an ontology used for the annotation of content gives a common view and semantic meaning in order to have effective content retrieval, enabling the exploitation of content in knowledge-based systems and processes (i.e., rule-based path computation). The MNISIKLIS SCMS uses the CO (Figure 16.4) in order to semantically annotate content items.

The architecture of our SCMS is comprised by three layers: Data Layer, Logic Layer and Presentation Layer (see Figure 16.6).
Data Layer: This layer is responsible for the content storage and retrieval in/from the physical media.

Logic Layer: It deals with the actual administration of the content. Logic Layer processes the retrieved data in order to pass them in the appropriate form to the presentation layer.

Presentation Layer: This layer is used by users in order to retrieve information and by the administrators of the SCMS in order to insert, retrieve and update content and the associated metadata. Content entities can be linked with INO elements through a graphical user interface.

16.6 User Interaction Subsystem

MNISIKLIS, in contrast to other indoor location based systems, targets a wide range of user population. The main user groups include:

1. Non-disabled users. This group is comprised of the typical users that do not have any impairment.

2. Elderly people (older than 65 years old). Some special features of this group are that they are not fully acquainted with modern technologies. They may also present some degree of disability in perception, memory, vision, hearing and physical movement. Hence, the User Interface (UI) should be as simple and tangible as possible.
3. *People with partial or total vision loss.* If the UI is visual, then it should be as distinguishable as possible (colors, fonts etc). In the case of audio or tactile UI there are no constraints except for the quantity of information given.

4. *People with locomotive disabilities.* Users that use a wheelchair or have very limited ability to walk.

### 16.6.1 Building an Accessible to All System

One of the main tasks of MNISIKLIS was to create a system Accessible to All. In order to achieve this goal, the design of the UI was based on Design for All principles (Stephanidis 2001). Hence, an analysis of the user groups and their needs was performed from the early stages of the project implementation, mainly based on the analysis of the existing scientific literature. To better define those groups, the International Classification of Functioning, Disability and Health (ICF 2001) was used.

### 16.6.2 MNISIKLIS Devices

In the context of the MNISIKLIS system, four types of user terminal devices are used by the target groups for accessing the LBS: Tablet PC, Smart Phone, PDA (Personal Digital Assistant) and Mobile Phone. The basic criteria for the selection of those devices were ergonomy and economy. Of high importance was also their compatibility to some assistive technology and their characteristics, such as processor speed, screen size and networking. Depending on the user profile, some input/output peripherals may be connected wirelessly (e.g., Bluetooth) or through cables to some devices, such as earphones, head-mounted screens and Braille displays. *Table 16.1* presents the possible combinations of peripheral devices with the user device.

<table>
<thead>
<tr>
<th></th>
<th>Head-mounted LCD screen</th>
<th>Braille Display</th>
<th>Bluetooth Earphones</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tablet PC</td>
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</tr>
<tr>
<td>Smart Phone</td>
<td></td>
<td></td>
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<tr>
<td>PDA</td>
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<td>√</td>
<td>√</td>
</tr>
<tr>
<td>Mobile Phone</td>
<td></td>
<td></td>
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</tr>
</tbody>
</table>

### 16.6.3 Multimodal User Interfaces

The UI of MNISIKLIS is multimodal. It adopts three modalities for input/output: visual, audio and haptic. In the haptic mode the user can interact using the keyboard, special buttons, the touch-screen, or the Braille display. The user can also interact
orally, through a speech-based dialogue subsystem (Fellbaum & Kouroupetroglou 2008) using a Mobile Phone. In the visual mode, the output is presented through a common graphical UI (GUI). Combinations of the above described modalities are also supported.

**Visual Modality:** In the case of a GUI, the output data may be in the form of text, image and map. Specifically, the user has at her disposal a textual menu consisting of buttons and selection lists to make a service choice. When the user has requested a specific navigational service, she receives an output consisting of textual instructions, images of the nearby POIs, landmarks and corridors and a map depicting the current floor (see Figure 16.7). The user is navigated on a turn-by-turn basis using textual instructions that are given dynamically according to her position and notifications, whenever necessary. The map, in this case, depicts the segment of the path/itinerary that is already covered by the user and the remaining part left to travel. When the user asks for an informational service such as Nearest POIs, Where-Am-I or Exploration, then she obtains three layers of proliferating information in the textual instructions. She also obtains the current map depicting her position and the position of the Nearest POIs.

The user can choose among five different text font sizes, two font colors (black or white) and five colors for the drawing on the maps. The maps are represented in

![Screenshot of the graphical UI.](image-url)
Scalable Vector Graphics (SVG) format and can be transposed and zoomed-in and -out by the user without any loss of detail.

Audio Modality: A person without hearing loss is able to obtain audio output in the form of synthetic speech or in combination with the visual output. The design of the GUI ensures support of traditional Interactive Voice Response (IVR) mode. For the case of the blind users full spoken-based dialogue interaction (Speech-only User Interface) is supported (Freitas & Kouroupetroglou 2008).

Haptic Modality: In the case of haptic interaction, the user makes use of her hands to provide input (touch screen, buttons) or/and obtain output (Braille display). The output on the Braille display is just the same as if the user was using audio output.

16.6.4 Architecture of User Interaction Subsystem

The User Interaction subsystem of MNISIKLIS adopts a client/server architecture. The client part is installed on the users’ terminal. It constitutes a cross-platform software component compliant with a variety of devices. Specifically, the client was installed on a Tablet PC, a PDA, a Smart Phone and a speech interaction server. The server side of the subsystem, on the other hand, receives the requests of the client and, in collaboration with the subsystems of Positioning Services, Content Management and GIS, composes and returns the output to the user, regardless of the device used. The architecture of the User Interaction subsystem is depicted in Figure 16.8.

![User Interaction Subsystem Architecture](image-url)
The client is developed using Java and other open source technologies. Java was selected because of its interoperability and platform neutrality. Moreover, there are various tools available to support the technologies integrated in MNISIKLIS.

Maps are visualized through SVG and are displayed using the Batik open-source framework (from Apache Foundation). This caters for efficient map handling (zoom, rotate, drawing of POIs and paths etc). On the PDA, the Tinyline SVG toolkit is used to perform the same tasks.

16.6.5 Evaluation and Field Trials

Currently, a number of trials with real users are underway. Users belong to all target groups addressed by MNISIKLIS. The field tests take place in the building of Department of Informatics and Telecommunications – National and Kapodistrian University of Athens and a university museum. Users are asked to select/execute a certain service (for example, follow the instructions of navigation). All the users’ choices, significant intermediate results, outputs and positions are logged for post-processing and analysis. Furthermore the users are asked to change their profile, preferences and selected POIs, resulting in changes to UNO and INO ontologies. Finally, after having used the system, the users are asked to fill in some questionnaires. The results of these tests are expected to bring forward directions for further research and improvement of MNISIKLIS.

16.7 Conclusions and Future Work

In this paper, we presented MNISIKLIS, an integrated framework that provides indoor location-based services for all. The system exploits semantic technologies in order to represent the application models and the estimation of user position is achieved through advanced positioning techniques. Furthermore, it implements multimodal interfaces for supporting both able-bodied and disabled users. As already mentioned, currently, we are working on the validation of MNISIKLIS through a set of trials.

However, a number of issues remain open for further research in the area of indoor location-based services. The path generation methodology could be enhanced by taking advantage of path prediction techniques and historical information. Moreover, the incorporation of landmarks during the navigation process could substantially facilitate comprehension of route directions by the user. Another issue that we are working on is the incorporation of other technologies and elements (infrared beacons, ultrasound sensors) in the positioning system, thus enhancing its accuracy and robustness. Furthermore, different methods and techniques (e.g.,
Evidential Reasoning) will be examined in order to improve the results of the second level fusion. Additionally, Kalman and particle filtering will be utilized for improving the dead-reckoning and the sensor fusion position estimations (Evennou & Marx 2006), as well.

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References


