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DESIGN OF A DECLARATIVE LANGUAGE FOR PERVASIVE SYSTEMS

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Abstract

This project aims at defining a completely declarative language to deal with wireless sensor networks and, more generally, with pervasive systems. In this work the main problems to be considered for the design of such a language are presented and discussed. Then a software architecture for pervasive systems is introduced and a language proposal is presented. The interaction between the declarative language and the existing architecture is described too. Finally some simple examples are reported to show how the language can be used by a final user, when the system will be implemented.

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Introduzione

Esistono molti contesti applicativi le cui fasi operative vengono monitorate tramite un elevato numero di dispositivi, muniti di appositi sensori, tra loro molto differenti per tecnologia e modalità di funzionamento. Si consideri, a titolo di esempio, il processo di produzione e trasporto dei vini di qualità che costituisce uno dei casi di studio di *ART DECO*, un importante progetto finanziato dal *Ministero dell'Università e della Ricerca italiano*.

Il processo deve essere monitorato in ogni sua fase, a partire dalla coltivazione dell'uva nel vigneto fino al trasporto su camion ed al mantenimento delle bottiglie nelle cantine. Ognuna di queste fasi richiede degli impianti di sensoristica molto diversi tra loro: le reti di sensori wireless possono essere la soluzione ideale per il monitoraggio di parametri quali la temperatura e l'umidità nel vigneto; l'applicazione di etichette *RFID* alle bottiglie può essere appropriata per la loro identificazione durante e dopo il trasporto; i dispositivi *GPS* permettono invece di conoscere l'attuale posizione dei camion che trasportano le bottiglie.

E' evidente l'interesse che può esistere nel considerare nel loro insieme tutti i dati che possono essere raccolti tramite questi impianti di monitoraggio. Ciò, però, risulta difficile in quanto le diverse tecnologie utilizzate non sono fra loro integrate e ciascuna di essere presenta all'utente delle interfacce di controllo ed interrogazione differenti.

Il lavoro presentato in questa tesi ha come obiettivo quello di analizzare il problema sopra descritto e porre le basi per una sua possibile risoluzione: l'idea principale è quella di definire un linguaggio di alto livello completamente dichiarativo, che permetta all'utente di interrogare un sistema pervasivo utilizzando una sintassi molto simile a quella dello *SQL* standard. Con il termine "sistema pervasivo" si intende una rete eterogenea formata da dispositivi molto differenti tra loro: reti di sensori, palmari, etichette *RFID*, ecc. Alla base della definizione del linguaggio dichiarativo di cui sopra vi è l'idea di astrarre il sistema pervasivo fino a considerarlo alla stregua di una base di dati, con l'obiettivo di mascherare la complessità della rete sottostante e mettere in risalto le informazioni che vengono raccolte, nonché le relazioni che sussistono tra di esse. In

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questo modo l'utente, non dovendosi preoccupare delle peculiarità dei singoli dispositivi, può interrogare l'intero sistema in maniera facile e veloce.

Lo scopo iniziale del progetto era quello di creare un linguaggio dichiarativo per l'interrogazione di reti costituite solamente da sensori tra loro omogenei. Si voleva riprendere l'idea di *TinyDB*, che storicamente è stato il primo progetto ad aver proposto la possibilità di astrarre una rete di sensori come una base di dati. Nonostante l'idea proposta fosse molto innovativa ed abbia suscitato molto interesse in ambito accademico, questo progetto è rimasto un prototipo senza mai essersi trasformato in un'applicazione commerciale.

Si è pensato quindi di utilizzare *TinyDB* come base di studio da cui partire per tentare di definire un sottoinsieme minimo di funzionalità necessarie in un qualsiasi sistema per l'interrogazione di reti di sensori. L'obiettivo originale era quello di implementare un sistema che supportasse tali funzionalità e che fosse facilmente portabile su diverse architetture fisiche.

In seguito, durante la fase di analisi dei requisiti, lo scopo del progetto si è modificato: l'idea di poter interrogare contemporaneamente diversi dispositivi, tra loro eterogenei sia nel tipo che nella tecnologia, ha spostato il focus sulla definizione di un linguaggio dichiarativo per sistemi pervasivi.

Il *Capitolo 1* presenta alcuni progetti simili, classificandoli secondo i diversi approcci utilizzati ed evidenziando le differenti peculiarità, potenzialità e limitazioni. A partire dall'analisi di questi progetti sono stati definiti gli aspetti fondamentali del linguaggio che viene proposto in questa tesi. Essi sono presentati nel *Capitolo 2* e sono classificabili nelle seguenti quattro categorie:

- Aspetti relativi alla rappresentazione e all'astrazione dei dati
- Aspetti relativi alla gestione dei dispositivi fisici
- Aspetti relativi alle caratteristiche funzionali del linguaggio
- Aspetti relativi alle caratteristiche non funzionali del linguaggio

L'eterogeneità dei dispositivi considerati ha portato alla necessità di suddividere il linguaggio in due parti:

- Un *Linguaggio di Basso Livello*, il cui scopo è quello di descrivere le operazioni di campionamento e le manipolazioni da effettuare sui dati raccolti dal singolo nodo (raggruppamenti, aggregazioni e filtraggi).
- Un *Linguaggio di Alto Livello*, che ha il compito di descrivere le manipolazioni da effettuare a partire dai dati prodotti dal basso livello.

Entrambi i linguaggi hanno una sintassi simile ad *SQL*, ma la semantica del linguaggio di basso livello si differenzia in buona misura da quella dello *SQL* standard, sia perché introduce clausole per la gestione del campionamento, sia perché è stato studiato per agevolare la generazione di dati aggregati a partire dai dati campionati. La semantica del linguaggio di alto livello è invece molto simile a quella dei linguaggi di interrogazione per database di streaming.

Il *Capitolo 3* presenta l'infrastruttura adottata per descrivere la semantica del linguaggio proposto. Si tratta di un'architettura per sistemi pervasivi creata dal *Politecnico di Milano* (area ingegneria del software), costituita da tre livelli che forniscono tre differenti gradi di astrazione dei dispositivi fisici:

- *Livello applicativo*: costituisce il *front-end* utilizzato dalle applicazioni per accedere ai dati provenienti dai dispositivi fisici;
- Livello degli oggetti logici: fornisce un'astrazione dei dispositivi fisici esponendo un'interfaccia che permette agli oggetti applicativi di interagire con l'hardware. Ciascun oggetto logico "wrappa" un singolo dispositivo o un gruppo omogeneo di essi, mascherandone la complessità di accesso;
- Livello di accesso fisico ai dispositivi: fornisce un'infrastruttura software che
 permette agli oggetti logici di accedere ai dispositivi fisici mascherando i
 problemi legati alla distribuzione.

A livello applicativo, un analizzatore di query manipola le interrogazioni sottomesse dall'utente e, utilizzando un componente dedicato (*registry*), ottiene informazioni circa

gli oggetti logici che compongono il sistema. Quindi, a partire da queste informazioni e da quanto richiesto dall'utente, seleziona gli oggetti logici che dovranno prender parte all'esecuzione della query.

Grazie all'architettura appena descritta, l'intero sistema pervasivo può essere astratto come un insieme di oggetti logici. In questo modo, ogni richiesta ad un nodo del sistema si traduce, di fatto, in una richiesta al corrispondente oggetto logico, tramite l'interfaccia da esso esposta. Proprio la definizione precisa di questa interfaccia permette di descrivere la semantica del linguaggio in termini di interazioni con gli oggetti logici. Ad esempio, una richiesta di campionamento ad un nodo che possiede un sensore di temperatura viene vista dal linguaggio come la lettura di un attributo dell'interfaccia esposta da quel nodo. Similmente, il passaggio di un tag *RFID* da un certo lettore viene visto dal linguaggio come un evento scatenato dalla stessa interfaccia.

Il Capitolo 4 descrive come le query sottoposte dall'utente possano essere rappresentate tramite grafi, i cui nodi definiscono le strutture dati e le componenti di alto e di basso livello delle query stesse. Viene in seguito presentato il meccanismo con cui tale grafo è utilizzato dal sistema per decomporre, distribuire ed eseguire la query sui vari nodi.

La definizione dei concetti appena descritti permette una migliore comprensione della semantica del linguaggio, che viene presentato in maniera formale nel *Capitolo 5* tramite una grammatica *EBNF*, scritta in modo tale da essere facilmente leggibile e comprensibile al lettore, ma non direttamente adatta come input per i software generatori di parser.

Il *Capitolo* 6 propone vari esempi di query per aiutare il lettore a comprendere meglio la semantica del linguaggio e le varie funzionalità offerte da quest'ultimo. Inoltre riprende il caso di studio relativo al monitoraggio dei vini e propone alcuni esempi realistici di utilizzo del linguaggio.

Il *Capitolo* 7 descrive i passi che dovranno essere seguiti per l'implementazione del sistema descritto in questa tesi e si focalizza in modo particolare sul parser, che è l'unico componente finora realizzato. In particolare viene presentata una trasformazione della grammatica, necessaria per rimuovere le ambiguità e permetterne l'utilizzo con un

software generatore di parser. Viene poi presentata la struttura a oggetti che costituisce l'interfaccia tra il parser ed il motore di esecuzione delle query.

Nel capitolo conclusivo il sistema proposto in questa tesi viene confrontato con il progetto *GSN*, sviluppato presso l'*EPLF di Losanna*, in quanto esso presenta diverse similitudini con questo lavoro. Il capitolo suggerisce inoltre alcuni possibili miglioramenti che potranno essere apportati al linguaggio.

L'Appendice A riporta la grammatica completa del linguaggio, in forma non ambigua, scritta utilizzando la sintassi del generatore di parser JavaCC.

1 Goals and phases of the project

The main idea of the project is to define a completely declarative language that allows the user to query a pervasive system in a similar way as SQL allows to query traditional databases. A pervasive system is a large heterogeneous network composed of many devices, belonging to different technologies like wireless sensors networks (WSN) [1] [2], RFID systems, GPS and other kinds of sensors.

We aim at providing a database like abstraction of the whole network in order to hide the high complexity of low level programming and allowing users to retrieve data from the system in a fast and easy way.

The original goal of our project was to define a *SQL* like declarative language only for *WSNs*. We aimed at defining at least a working subset of the functionalities supported in similar existing systems. Our idea was to develop an architecture deployable on different *WSN* technologies with little effort, but we were not initially interested in supporting query execution over different device technologies at run time. So, the goal was to develop a software executable on homogeneous sets of nodes, either in terms of technologies or of types of sensors.

The first idea changed many times during the initial phase of the project; in this chapter we describe the main steps that led us to the final goal of defining a declarative language for pervasive systems. Then we report a short survey on existing similar projects. Finally we briefly present *TinyDB* [3] [4] [5] that was the first attempt to abstract a sensor network as a database.

1.1 Projects goals evolution

Firstly, we considered the possibility of developing a system really supporting heterogeneity, also at runtime. Allowing the user to write queries, whose execution involves *WSN* nodes belonging to different technology classes, seemed very interesting.

To reach this goal we had to deal with two main issues: implementing communication between nodes implemented with different technologies and understanding the additional features required by the language to support runtime

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heterogeneity. Communication issues should be resolved by a middleware that provides a set of *API*s to manage the different hardware platforms in a uniform way. At this point, the language becomes a user friendly interface to that *API*. An implementation of the language is then a compiler or an interpreter that translates queries into a set of *API* calls.

The effort in language design does not change with the introduction of the full heterogeneity requirement. In fact the main part of the work is still the research of all the common features present in the different technologies.

The next extension to our project goals has been the introduction of new kinds of devices. We realized that our considerations about the language are not strictly limited to wireless sensor networks, and can be easily extended to any kind of mobile device equipped with sensors and communication tools (for example *PDAs*). In fact the only requirement for a device to be used in our architecture is the ability of sampling some physical magnitude and sending collected data to a base station. The only effect, rising from enlarging the set of supported devices, is the need of finding a more general method to manage the language non functional characteristics. In fact, while in *WSN*s the main issue to be handled is the power management, other aspects have to be considered in other types of devices.

From the point of view of the language functional characteristics, some problems came out when we tried to include also the possibility of managing *RFID* tags. There are many types of *RFID* technologies (*HF*, *UHF*, *UWB*, etc.), but in any case simpler tags are not equipped with sensors and cannot perform data manipulation or data transmission.

As we will better describe in the following, we found two possible solutions to this problem taking into account that the information of interest is the presence of a tag into the range area of an *RFID* reader.

The first one abstracts an *RFID* tag as it were a sensor, and the *ID* of the last reader which sensed the tag is considered as the sampled data. On the opposite, the second solution abstracts an *RFID* reader as it were a sensor, and the *ID* of the last tag detected by the reader is considered as the sampled data. In both cases, the communication with the network node that submits queries is done by readers, so it is clear that the first

solution provides a greater level of abstraction and then requires a greater amount of middleware to be implemented.

Moreover, with the introduction of *RFID* devices, we had to define also an event based semantics for data managing at the language level. In fact, differently from other kinds of nodes, the sampling operation on passive *RFID* is done when the tag is sensed by a reader and not at fixed instants. A time based semantics for *RFID* sampling can also be defined, but the event based one must be considered too.

The last step in the project goals definition has been the extension of the declarative language to generic pervasive systems. Although our first idea was to manage only networks composed of a limited number of nodes, we decided to define a language that can also be used on larger networks.

This choice was taken after a discussion with the participants to a project [6] about pervasive systems that is being developed at *Politecnico di Milano*. It defines an architecture for large heterogeneous networks that aims at providing a logical and distribution independent software platform through three different abstraction layers.

This architecture allowed us to focus only on the declarative language definition without taking into account low level programming issues. We only decided the positioning of the language on the existing architecture. Then we defined some constraints on the abstraction interfaces (logical objects) of the devices that will be part of queries. Finally we designed the language syntax and semantics, describing the queries behavior in terms of logical objects abstraction.

The introduction of the middleware for pervasive systems also allows a simpler high level testing of the language parser, without the need of a hardware simulator. In fact trivial implementations of logical objects can be used to perform a first test of the parser and the query execution engine.

1.2 Short survey of system features and relations with similar projects

It is not the first time that a database like abstraction approach is used in wireless sensor networks. In fact, in the last years, there was an increasing interest of the academic world on the problem of finding a high level model for programming this kind of systems.

Many different projects on WSNs have been developed and they introduced a lot of different network abstractions and high level programming models. Some of this projects are TinyDB [3] [4] [5], Cougar [7], Maté [8], Impala [9], Sina [10], DsWare [11], MaD-WiSe [12] [13], Kairos [14] and GSN [15]. There is also a certain number of surveys that try to classify each of the previous projects depending on the type of the provided abstraction [16] [17] [18] [19]. In particular S. Hadim and N. Mohamed [19] evaluated a set of representative middleware, analyzing the different approaches with their advantages and disadvantages. They also wrote out a clear summarizing table that is surely helpful to get the state of art on high level WSNs programming techniques.

For our purposes, we will mainly focus on two classification criteria. The first one is related to the architecture the project works on, while the second one is related to the kind of language provided for interacting with the system.

The architectural-based classification depends on:

- The type of supported nodes (sensors, RFID, PDA, etc.)
- The specific hardware model for each of the previous types (Crossbow MOTE [20], Intel MOTE [21], , ...; HF tags, UHF tags, UWB tags, ...)
- The supported heterogeneity level

The heterogeneity level can be evaluated on two different dimensions:

• **Deploy-time heterogeneity**: this dimension indicates the ability of the system to be deployed on different kinds of hardware platforms, so it can be considered as an index of the system hardware portability. For example, we will say that a system supports the deploy-time heterogeneity if it can be deployed on a **MICA**

MOTES WSN, but can also be re-compiled with little effort to be used on a different kind of network.

• *Run-time heterogeneity*: this dimension indicates the ability of the system to manage a network composed of different kinds of nodes (*RFID*, sensors, etc.).

We can then find out three macro groups of projects based on the supported architecture:

Architectural-based heterogeneity		Deploy-time heterogeneity		
		Little	Full	
-time geneity	Little	Homogenous system	Partially Heterogeneous system	
Runhetero	Full	-	Heterogeneous system	

Figure 1: Heterogeneity levels

Most of the existing *WSN* projects we have analyzed work on homogeneous systems. Sometimes they provide deploy-time heterogeneity through the underlying operative system portability (that is often *TinyOS* [22]).

The only project we found that is really intended to work simultaneously with different kinds of devices is *GSN* (*Global Sensor Network*), under development at the *EPFL of Lausanne*.

The second classification we are interested in is related to the type of language provided to the final user for interacting with the system. There are mainly two kinds of languages:

• **Declarative languages** focus on the results the user expects from the system rather than the way these results should be computed. The most famous and diffused declarative language to query databases is *SQL*. For this reason, many of the analyzed projects define *SQL* like languages to query the network. *TinyDB* was one of the first works that tried to follow this approach.

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• **Procedural languages** are imperative programming languages allowing the definition of the steps that should be executed by each network node to obtain the expected results. Often, systems using a procedural language provide a middleware with the *API*s that can be called by the language itself. While in the case of declarative languages the provided network abstraction is often a database, in the case of procedural languages many different abstractions have been adopted. For example, some projects are based on the concept of Virtual Machine (*MATE* and *MAGNET*) while others are based on the idea of event programming (*IMPALA*).

In some cases the system provides both a procedural language and a declarative language. An example is SINA that is composed of 4 layers. The first one is the SINA middleware. Then the SQTL (Sensor Query and Tasking Language) is a procedural language built over the middleware. The third level provides a built in declarative query language, defined in terms of SQTL scripts. Finally the application level can query the network using the declarative language, but also generating and directly executing SQTL scripts.

With respect to the above classifications, the final goal of our project aims at producing an entirely declarative *SQL* like language with full heterogeneity support and able to be executed on large heterogeneous pervasive systems.

1.3 Short TinyDB analysis

TinyDB is the most famous and cited work on declarative languages for sensor networks because it was the first attempt to design a processing system for extracting information from a *WSN*, abstracting it as a database. It was also the first project that provided an implementation of such a system.

In this paragraph the *TinyDB* language and architecture are briefly presented because it was the project that suggested us the first idea for our work. *GSN*, instead, is the most similar project w.r.t. our proposal, both in terms of final goals and adopted approach. For

this reason, it will be shortly presented in the conclusions of this thesis to allow a comparison between *GSN* and our language features.

We studied *TinyDB* papers [3] [4] [5] and tried to run *TinyDB* software on the *Tossim simulator* [23]. That work allowed us to appreciate the theoretical peculiarities of *TinyDB*, but also to find out some limitations in the implementation that makes *TinyDB* an interesting academic research, but not a very useful product [24].

TinyDB is written in nesC [25] over TinyOS operating system and runs on MICA Motes architecture. The innovative idea introduced by Madden was the possibility of thinking to a WSN as a database. Each node of the network can be actually thought as a source of tuples representing the results of sensor samplings.

Starting from this concept, Madden also introduced the idea of querying the network through a *SQL* like language. We now describe the mainly features of *TinyDB*.

1.3.1 Types of tables

In each *TinyDB* implementation there are:

- A sensor table (called "sensors")
- Some materialization tables

"Sensors" is a logical and distributed table that contains the sensor samplings coming from all nodes. Each record of the table is relative to a node of the network in a specified timestamp. The structure of the "sensor" table is then the following:

NodeID	Timestamp	Attribute 1	Attribute 2	 Attribute n

Figure 2: TinyDB sensors table

Each of attribute 1, attribute 2, ... attribute n fields refers to a sensor (e.g.: temperature, pressure, etc.) and contains the readings from that sensor. If a sensor is not present in a node, all the records that refer to that node will contain a null value in the

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field corresponding to the attribute. Note that if there are many different types of nodes in the network, many null values will be contained in the sensor table. The set of supported attributes and sensors is predefined and cannot be changed.

Materialized tables are physically stored and contain the results of a query on the "sensors" table. They are in fact buffers of fixed size, unlike the "sensors" table that is an unbounded logical stream. Materialization can be used to define a sliding window on the stream of sensors readings.

1.3.2 **Epoch**

TinyDB defines the concept of epoch as the time between the start of two consecutive sampling periods. Each epoch is identified by a number called timestamp.

Samplings are always done at the start of the epoch and, then, the remaining time of the interval is used to transmit collected data to the base station.

1.3.3 Types of queries

TinyDB supports two types of queries:

- Selection queries
- Materialization queries

The first type should be used to define the sampling rate of the network sensors and to collect data at the base station. The result of this kind of queries is a stream of records that comes to the base station with a rate equal to the specified sampling rate.

Every selection query is based on the "sensors" table, that must be present in the FROM clause. The simplest kind of selection query defines a view on "sensors", that is the only table contained in the FROM clause (Figure 3).

More complex queries can be written introducing aggregation operators and joins with materialization points.

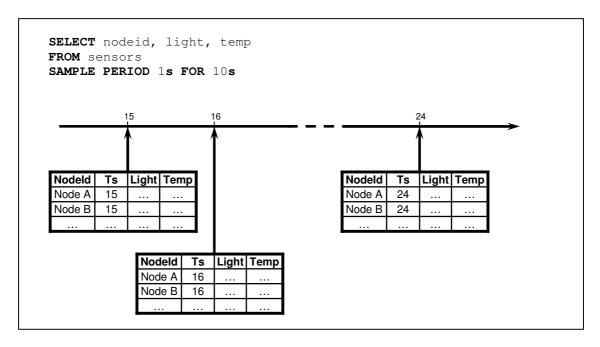


Figure 3: Simple TinyDB selection query

The latter points can be defined using materialization queries; they are composed of a selection query, with an additional "CREATE STORAGE POINT" clause to define the name, the structure and the size of the buffer.

An example of materialization query is reported in *Figure 4*. In that example an eight records buffer is reserved for the most recently readings. Each sample is taken every 10 seconds.

```
CREATE STORAGE POINT recentlight SIZE 8 AS

(
SELECT nodeid, light
FROM sensors
SAMPLE PERIOD 10s
)
```

Figure 4: Simple TinyDB materialization query

1.3.4 Termination of a query

In *TinyDB* there are three ways to stop a running query:

- Sending a statement "STOP QUERY {ID}" to the network. When a query is started the system assigns it an id, which can be afterwards used to stop the query.
- Specifying a FOR clause in the query. The FOR clause allows the definition of the running time. When this time expires the query is automatically stopped by the system. In the example reported in Figure 5 the execution will terminate after ten seconds.
- Specifying a termination event in the query. The syntax "STOP ON EVENT (param) WHERE cond (param)" can be used in a query to define the termination event.

```
SELECT nodeid, light, temp
FROM sensors
SAMPLE PERIOD 1s FOR 10s
```

Figure 5: TinyDB query with FOR clause

1.3.5 SQL Operators

The most important *SQL* operators provided by *TinyDB* are joins and aggregations. The join operator can be used in two different contexts. If the join is made between two or more materialized views a standard *SQL* join is obtained.

However, it is possible to make a join between the "sensors" table and a materialized view: in this case each record coming from the "sensors" is joined with the current content of the materialization view and the result is inserted in the output stream (Figure 6).

In the first case there aren't issues related to the semantics of the join, while in the second case the "sensors" table sampling rate can be different from the materialized view sampling rate. To manage this situation *TinyDB* provides two constructs: *COMBINE* and *LINEAR INTERPOLATE*.

```
SELECT COUNT(*)
FROM sensors AS s, recentlight AS r
WHERE s.nodeid = r.nodeid
   AND s.light < r.light
SAMPLE PERIOD 10s</pre>
```

Figure 6: TinyDB query with join

Another important *SQL* operation supported in *TinyDB* is the aggregation, that can be either temporal or non temporal.

In non temporal aggregations, each generated record refers to a single epoch and is just a classical aggregation of data produced in the current epoch. (*Figure 7*)

```
SELECT room, AVG(volume)
FROM sensors
WHERE floor = 6
GROUP BY room
HAVING AVG(volume) > [threshold]
SAMPLE PERIOD 30s
```

Figure 7: TinyDB query with non temporal aggregation

Classical aggregation operators (MIN, MAX, AVG, SUM and COUNT) are predefined in the system, but new operators can be defined by the user specifying the following three elements:

- f: merging function (that should be commutative and associative)
- *i*: initializer
- e: evaluator

Temporal aggregations allow to group data coming from different epochs. Consider the example in *Figure 8*. Suppose that only a node is executing the query. In this case the semantics is clear: every second the volume sensor of that node is sampled while every 5 seconds the average of the last 30 readings is calculated and the result is sent to the base station.

1. Goals and phases of the project

Now suppose that *n* nodes are executing the same query. Every second a sampling from each node is taken while every 5 seconds the average relative to the last 30 seconds data should be computed. Madden's article on *TinyDB* doesn't clarify the semantics of *WINAVG* in this case. In fact there are at least three different ways to compute the average:

- Average of all 30 * n values
- Temporal average on each node (30 values) and then average of previous results (*n* values)
- Average on each timestamp (n values) and then temporal average of previous results (30 values)

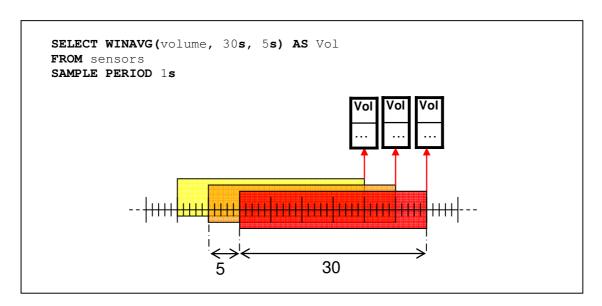


Figure 8: TinyDB query with temporal aggregation

1.3.6 Events

TinyDB provides also the concept of event to control the query start and termination and to react to external world changes. Events can be used by queries to decide:

- When the execution should start (*Figure 9*)
- When the execution should end (*Figure 10*)

Events can be generated both by the *OS* to signal some external events and by queries through the *OUTPUT ACTION SIGNAL* statement.

```
ON EVENT bird_detection(loc):
SELECT AVG(light), AVG(temp), event.loc
FROM sensors AS s
WHERE dist(s.loc, event.loc) < 10m
SAMPLE PERIOD 2s FOR 30s
```

Figure 9: TinyDB event used to start a query

```
...
STOP EVENT(param) WHERE COND(param)
```

Figure 10: TinyDB event used to terminate a query

1.3.7 Lifetime and power consumption

In *TinyDB* users can specify query duration through the *LIFETIME* clause. The system uses this clause to decide sensors sampling rate in order to grant appropriate power consumption.

This feature of the language shows that introducing non functional specifications in *SQL* like languages for *WSN*s can be useful.

1. Goals and	phases of	the	proj	ect
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2 Main issues in language design

In this chapter we will present the main issues we found out during the first phase of the project, when trying to define the features of the language. Those issues are the questions that have to be answered to clearly bound the goals and the characteristics of the language that is to be designed.

We classified the issues into four categories, identifying questions related to data representation and abstraction, questions dealing with physical devices and questions strictly related to the language expressiveness, both in terms of functional and non functional features.

2.1 Data representation issues

In this section we outline the ways in which data of interest will be presented to the user. We also describe abstractions that can be used to hide physical devices and to provide a database view of the whole network. The logical steps that allow the manipulations of samplings, coming from different sensors in order to build an output table, are presented too.

First of all we decided to use a database like abstraction, but there are some differences with the *TinyDB* abstraction. In fact we think that the concept of sampling cannot be completely hidden to the user. *TinyDB* abstracts the sampling process through the concept of "sensors" table, without allowing the user to modify any sampling parameter except for the sampling rate. This choice makes the system easily to be used and understood by the user, but it is only feasible when all the devices in the network are homogeneous and their features are already known at design time.

In our heterogeneous system the sampling operation should be managed in a more complex way, giving the possibility of defining some complex criteria to decide which nodes will perform the sampling operation. Moreover we would like to allow the user to define other sampling details, for example, in terms of sample rate and characteristics of the node performing the sampling.

2. Main issues in language design

Another problem to be managed is that different nodes can have different attributes and different limitations on the maximum achievable sampling rate. As we will better describe in the following, the above considerations led us to split the language in two parts: the lower level is intended to manage the sampling operations while the higher level has the role to manipulate the sampled data producing the query results. Due to the great differences between the two parts, two different languages should be designed. The low level one deals with sampling operations and can only perform some data manipulations on data sampled from a single sensor. So its semantics cannot be very similar to that of standard SQL; anyhow, we will try to define an SQL like syntax for it. On the contrary the high level language will be very similar to standard SQL because it must perform relational operations on the data streams produced by the low level part.

Note that the high level language has similar functions as the *TinyDB* language, because both are used to manipulate data streams coming from sensors. The low level language has not a counterpart in *TinyDB* and can be thought as a mechanism to generate data streams corresponding to the *TinyDB* "sensors" table (Figure 11).

The abstraction provided by the low level language does not only handle the parameters of the sampling operation, but also deals with the problem of deciding when sampling should be executed on different network nodes.

Users should be able to define the set of nodes that will take part in the query, specifying some conditions on the type (*e.g.*: *RFID*, *WSN*), the characteristics (*e.g.*: available sensors) and the current state (*e.g.*: battery level) of the nodes.

Thinking about the above problem and analyzing some case studies, we realized that in many real situations a sampling on a node should be started if and only if a certain value has been retrieved from a sampling done on another node. For example, suppose that a user requires a temperature monitoring in a certain zone. If temperature sensors are mounted over moving platforms, it is clear that the sampling operation on a node should be activated only when its platform enters the zone. Platforms locations are detected by position sensors that are nodes physically different from the temperature nodes and not directly connected to them.

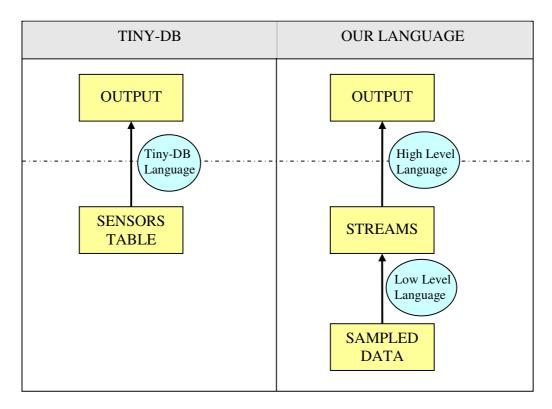


Figure 11: Comparison between TinyDB and our project

The above consideration suggested us that a specific operation should be supported to allow sampling activation depending on data sampled from other nodes. More specifically our idea is to force a sampling operation to start when a certain record is inserted into a stream by another query.

This operation can be logically thought as a join, followed by a filtering: in fact the same result of the above example can be obtained sampling continuously all the platforms positions and all the temperature nodes, then calculating the join between platforms location stream and temperature stream and finally eliminating all records relative to platforms currently out of the considered zone. This solution is not feasible in a real system because it requires the sampling of many sensors whose results are not used to compute the output stream. This is obviously too expensive from the point of view of power consumption.

The new operation, that we will call "*Pilot Join*" in the following, must be designed avoiding the above problem. To obtain this goal, latency between the event causing the sampling start and the first sample must be accepted.

Analyzing some real situations we find out that two kinds of *pilot join* are possible:

- Event based Pilot Join. When an event happens (i.e. a record is inserted into a stream) a certain set of nodes should start sampling (typically for a fixed period).
 - Suppose that, in the example of moving platforms, the temperature must be sensed once every time a platform crosses the zone borders. In this case an event based *pilot join* is required.
- Condition based Pilot Join. A continuous sampling (with frequency f) should be done on all the nodes that are connected to a base station that is in a list of base stations satisfying given criteria. This list is obtained extracting a fixed window from a stream and is updated with a frequency lower than f. Consider again the moving platforms example. Suppose that a running query is sensing (with low frequency) the position of all platforms and inserting them into a stream. Suppose also that the required behavior of the system is the continuous sampling of temperature sensors mounted over platforms whose last monitored position is in the zone. In this case a condition based pilot join is required.

The implementation of the condition based *pilot join* requires the definition of a data structure to store the list of records used as condition to activate and deactivate nodes sampling. This consideration led us to distinguish two table types that will be handled by the language.

- Stream Tables. Stream is the most important type of table and is an unbounded list of records produced by a query. Streams can be used to implement the event based join: the insertion of each new record in the stream is the event used to start the low level query in the nodes matching the record.
- Snapshot Tables. A snapshot table is a set of records produced by a query in a given period. During each period, all the records generated by the query are stored in a buffer; at the end of the period the snapshot table is cleared and filled with the records in the buffer. This type of table can be used to implement the condition based join: the snapshot table content is the list of nodes that are

currently satisfying the condition and that will be sampled during the next period.

2.2 Physical devices issues

In this paragraph we focus our attention on issues related to physical devices. We show the importance of giving a well defined meaning to the sampling concept. To reach this goal an abstraction of the sampling operation should be provided to hide the large hardware differences existing among different types of nodes.

Another problem that is introduced is the contrast between the ideal vision of hardware at the language level and the real hardware behavior (latency, failures, etc.). This will show the importance of formally defining a concept of timestamp in order to hide the above contrast.

The semantics of the sampling operation should be analyzed for all the device types that will be supported in our system in order to provide a good high level abstraction of the sampling concept. For network nodes having some sensors on board and able to probe some measures (temperature, pressure, speed, etc.) an intuitive meaning of sampling can be provided. When a sampling operation is invoked the required sensor is instantaneously probed and the read value is returned.

Consider now a node having on board also a storage device (*RAM*, *ROM*, *flash*, etc.). In this case the sampling operation can be extended to the reading of a section of that memory. Note that, from the user point of view, there are no differences between a sampled value obtained from probing a sensor and a value read from memory. In fact, in both cases, the user requests a sampling and the device returns a value.

So, a first abstraction of the sampling operation is a function-call with one parameter and a return value. The parameter is used to communicate to the device the name of the attribute to be sampled. It is not important if the attribute is then mapped to a real sensor or to a storage cell.

Particular attention is needed to define the sampling concept for *RFID* devices. First of all, it should be noted that both the tag and the reader can be seen as the device to be

sampled. If tags are considered as the devices, sampled data will be readers *ID*s. At the opposite, if readers are considered as the devices, sampled data will be tags *ID*s. Although we will support both the previous cases in our language, in the following we refer only to the first case because the software architecture for pervasive systems is thought to manage tags in this way. To better define an *RFID* sampling semantics it is important to remark that different kinds of *RFID* technologies exist.

HF-RFID (High Frequency RFID) is a technology working on short ranges: a tag is sensed when it passes near a reader (few centimeters). In this case, a first idea to define the sampling is to assume the ID of the tag currently passing near the reader as the sampled value. Obviously this is not a good idea because there is a high probability that, when a sampling is required, no tag is passing.

We found two possible solutions to this problem. The first one requires the introduction of a cached value: when a sampling is required the returned value is the cached one, i.e. the *ID* of the last reader that sensed the tag. Note that with this approach the sampling operation can be seen again as a function call.

On the contrary, the second solution requires a radical change in the abstraction: in fact the interaction model between the device and the upper software layer becomes synchronous. The idea is to abstract the sampling operation as an event: when the tag is sensed by a reader, an event flag is raised and a parameter of this event reports the *ID* of the reader. Note that, with the described abstraction, the sampling operation is not invoked by the user: there is not a concept of sampling rate and the execution of the next sampling operation is logically decided by the tag device.

In our language we will provide statements to support both sampling abstractions.

UHF-RFID (Ultra High Frequency RFID) is a technology working on long distances: the reader is able to detect the presence of a tag in a range of few meters. Since the reader to which a tag is connected changes slowly, in this case the sampled value can be abstracted by the current reader ID. Note that the sampling operation can return a null value if the tag is not in the range area of any reader in the system.

UWB-RFID (*Ultra Wide Band RFID*) is a fairly recent technology having a very long line-of-sight read range (dozens meters). These devices are active or semi-passive and the sampling operation semantics can be defined in a similar way to sensor nodes.

The last type of devices we consider in this summary are *GPS* (*General Positioning System*) sensors. They can be simply thought as a special kind of sensor node, having position as its unique attribute.

As we will better show in the next chapters, also a full wireless network can be abstracted as a single device. In this case the meaning of the sampling operation depends on the sampling semantics of the nodes composing the network.

Another issue related to physical devices deals with the contrast between the ideal vision of hardware at the language level and the real hardware behavior. In a WSNs query language time plays an important role: time values are used to specify sampling rates, to define queries execution length, to extract windows from streams, etc... Then, maintaining a timestamp shared by all nodes in the network is an important issue. This is not a trivial problem, but some algorithms can be found in the literature [26] to synchronize nodes' clock. Obviously, the result that can be obtained is not the sharing of an absolute time, but only a division of the time in fixed size intervals (called "epochs"). The shared timestamp is the number of the current epoch.

The computation performed by network nodes to manipulate sampled data requires a certain amount of time. In addition, the communication channels are not ideal and delays or packets loss can affect the transmission of data through the network. Therefore, latency can be found from the instant in which a record is generated by a query Q and the instant in which the same record is available for another query, using as input stream the output of Q.

The language should be able to hide this kind of problems as much as possible. To do this, a precise temporal semantics should be defined for all the operations provided by the language. As we will show in the following chapters, the concept of *native timestamp* can be introduced to formally specify a temporal semantics. We define the "native timestamp" of a sampled data the timestamp of the epoch in which the sampling was done. Then, we extend the concept to all records generated in the system, defining

2. Main issues in language design

some rules to assign a native timestamp to a record obtained from a manipulation of other records. The idea of native timestamps is especially useful to give an exact semantics to the *pilot join* operation.

2.3 Functional characteristics of the language

The functional characteristics of the language are the features that precisely define what the system can and what the system cannot do. They identify the set of operations and activities that can be required by user queries.

We identified a minimum set of functionalities, which must be implemented even in the first implementation of the language:

- Statements for the definition of sampling parameters (time, mode, etc.). These clauses are very important since they allow the extraction of raw data from network nodes. In fact obtaining high quality streams of raw data is essential because all other system functionalities work on them. The goal of these statements is to define when a node should or should not sample, the interval between two consecutive samplings and the condition under which sampled data must be held or deleted.
- Statements for the definition of operations to generate query output manipulating raw data. Also this set of clauses must be supported in a basic version of the language because, with the previous one, it forms the minimum kernel of instructions and clauses allowing queries definition. The goal of these statements is to provide classical SQL operators (filtering, aggregation, join, etc.) adapted for streaming systems.

We also identified a set of extra functionalities that are not strictly needed but, if implemented, can increase the system potential: statements for the definition of in-network data mining operations and statements for the definition of actuation commands. The latter functionalities are not implemented in the first release of the language and will be described in the last chapter of this thesis as future works.

2.4 Non functional characteristics of the language

The non functional characteristics of the language are the features that allow the definition of constraints on the offered services and of the system *QoS* (*Quality of Service*) [27] [28]. The main non functional requirements in *WSN*s are related to power management, but many other aspects can be considered, taking into account the wide heterogeneity of supported devices: different node types, different nodes latency and different available sensors.

Given the amount and the diversity of the possibly useful non functional constraints, it is not feasible to manage each possible constraint with a reserved clause. Thus, we think that a generalized abstract management could be a better solution: few clauses should be provided by the language to support all the non functional characteristics that can be considered now or discovered in the future.

The solution we present aims at providing a simple and general framework for policies management, allowing the user to express non functional requirements in a compact and abstract way.

Each physical device has a set of parameters that can be monitored and used to control query execution. For example, some sensor nodes can detect the power level reading the current battery voltage. The main idea in policy management is that this kind of sensed information can be used to take some decisions related to the quality of retrieved data rather than to the operations needed to produce query results. As an example, the power level can be used to decide which nodes of the network should execute the query.

Three main problems should be considered. The first one is that every device technology has a different set of parameters: for example the concept of power level cannot be applied to the *HF-RFID* technology because tags are passive devices powered by the *RFID* reader magnetic field.

The second problem is that the same parameter can be detected in different ways by different technologies: for example some devices can infer the power level starting from the number of executed operations, rather than measuring the battery voltage. Obviously

2. Main issues in language design

these details should be hidden to users: they only want to express conditions like "execute the query only on devices having a high power level".

The last problem is the different importance that a certain parameter can assume in different technologies. Consider still the battery level: it is very important in WSNs because charging or substituting the battery of a node is often difficult. The same information is still useful, but less critical, if related to a PDA, that can usually be easily recharged.

The abstract model we now propose as a solution will be clearer in the next chapter, when the architecture for pervasive systems will be presented. The idea is to abstract policies parameters as attributes of some logical object in the language: for a user there are no differences in requiring a power level or a temperature sample to a node. Every technology abstraction (logical object) should provide a mapping from its physical parameters to user-level policies. For example, the logical object of a sensor node will convert the battery voltage into a percentage indicating the current power state of the device. User queries will express conditions on that percentage.

Adopting the above solution, an important question has to be discussed: what should the query executor do if a policy-attribute that appears in a query is not managed by a technology abstraction? Note that this problem also exists when we consider standard attributes: for example, what should an executor do if a query requires a temperature sampling to a device that doesn't have any temperature sensor on board?

A possible, radical solution to that problem is excluding from the query execution all the devices that cannot provide a value for any of the attributes appearing in the query.

An alternative solution could be the definition of a well known set of policy attributes that should be implemented by all the logical objects. To maintain generality a default value for every well known attribute can be introduced so that, if an attribute is not supported by a device, the default value is used. For all other attributes (not present in the list) the default type value can be used when they are involved in selections, while devices should be excluded from the query execution when conditions refer to one or more non supported attributes.

A third solution, similar to the previous one, consists in using *NULL* values for all missing attributes, both in selections and in conditions, and then managing them with a *three valued logic* instead of a *Boolean logic*. This approach is the one used in standard *SQL* language [29] to deal with *NULL* values and can be summarized as follow:

- When an algebraic operator is applied to some operands, the produced result is *NULL* whenever at least one of the operands is *NULL*.
- *NULL* values are neither equal nor comparable with any other value (also if it is *NULL*, too). The comparison with a *NULL* always returns the *UNKNOWN* logical value.
- When a logical operator is applied to some operands, its logical value is computed using classical truth tables for three valued logic.
- The user can require that a condition must be (or not be) *TRUE*, *FALSE* or *UNKNOWN*. If this request is not explicitly specified, the default behavior is checking if the condition is *TRUE*.

We decided to adopt the latter solution because it is the more general and it maintains compatibility with the standard SQL approach. It is worth noticing that this feature of the SQL language is seldom used (and actually not known by common SQL users), but it is very useful in our language due to the above mentioned missing attributes issue.

Independently of the chosen way of implementing policy management, non functional clauses can be integrated in the declarative language to control some aspects of queries execution. For example, they can be used:

- To decide if a node should participate to a query
- To set the current sample rate
- To set the rate used for sending data out of the node
- To retrieve information about network nodes

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3 Architecture for pervasive systems

In this chapter the architecture for pervasive systems [6], over which we based our language, is briefly explained: the abstraction levels provided by the architecture are presented; then, three different ways of integrating our language with the existing platform are explored, with the motivation that led us to choose one of them; finally, we describe the interface that logical objects must implement to be used in our language.

3.1 Introduction to the architecture

Two are the main goals of the considered architecture: the first one is the definition of some abstractions for the different devices composing a pervasive system; the second one is the integration of data coming from peripheral devices in conventional information systems.

The architecture is intended to be used as a first step definition of the system during the design phase. It allows to design the system logical architecture and to provide a conceptual view of the pervasive system and of its logic, independently of its distribution.

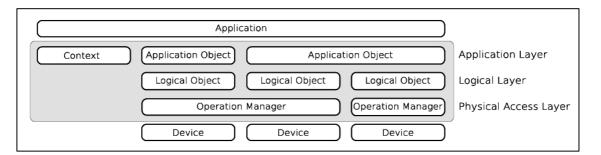


Figure 12: Layered architecture for devices abstraction

Figure 12 shows that three layers are introduced to describe peripheral devices at different abstraction levels:

• *The Physical Access Layer*, that is a representation of the underlying software infrastructure for the access to peripheral devices;

3. Architecture for pervasive systems

- The Logical Layer, that is built upon the device access layer and is defined to
 provide an abstraction of single or of homogeneous aggregates of physical
 devices:
- *The Application Layer*, that defines application level abstractions used by application programs to access data from peripheral devices.

The software components that populate each layer communicate with the higher one by sending events and with the lower one by invoking synchronous commands. Communication (both in terms of events and commands) can only occur between components defined in adjacent layers that are explicitly connected. The highest layer elements can also communicate among them.

Every layer is a container for a series of components. In the following we briefly present the three objects we are interested in, omitting all the other software components that are defined in the architecture.

3.1.1 Description of Operation Managers

The components of the *Device Access Layer* are the access point to devices. The goal of this layer is to allow modeling the access infrastructure to a specific device technology, abstracting from the required software distribution.

For example, in the case of the *RFID* systems, no software is executed on the tag and the interface provided by the *Operation Manager* is that of the reader or of a more complex off the shelf middleware that manages the reader (for example *EPC* [30] [31]). At the opposite, in the case of sensors, some software can be deployed on each sensor node.

At design time, only one *Operation Manager* exists for each software technology used to access peripheral devices, while at runtime some technologies allow the existence of different *Operation Manager* instances. In this case it is possible to have an instance for every device, or an instance for a group of devices (for example, if there are different sensor networks). Otherwise, in some cases (common with *RFID* systems) the *Operation Manager* can be a singleton.

3.1.2 Description of Logical Objects

A component in the logical layer (called *Logical Object*) provides a virtualization of a single device (or of a homogeneous group of devices) seen as a functional element. Every logical object offers some operations to the higher level and can raise events to that layer.

The *Logical Object* works dealing with the *Operation Manager*, hiding the interaction paradigm defined by the *Operation Manager* and hiding possible changes in this paradigm.

When a *Logical Object* deals with a single instance of an *Operation Manager*, it represents a virtualization of the device and, optionally, it can encapsulate the application logic executed on the device, thus hiding the communication trough the base station.

On the contrary, when *Logical Object* deals with aggregate devices, it represents a group of homogeneous peripheral devices viewed as a single functional element. In this case the *Logical Object* hides the presence of single nodes and aggregation is performed by the *Operation Manager* which provides aggregated results.

3.1.3 Description of Application Objects

This level contains components that perform abstraction of application layer concepts that are intended to be used by conventional applications to interact with the physical world. These abstractions get their data from *Logical Objects* and every abstraction can be bound to different *Logical Objects*.

The main abstraction on this level is called *Application Object* and defines a concept of the application domain that gets data from peripheral devices or that interacts with these kinds of objects. *Application Objects* define operations that can be externally invoked and events that can be raised.

3.2 Integration of the declarative language in the existing architecture

The relationship between the declarative language and the architecture must be analyzed from two different points of view.

3. Architecture for pervasive systems

Firstly, the set of objects able to receive a query from a user should be identified. The architecture layer these objects are placed into should be decided too. The objects that can accept queries should also be able to parse and distribute them to the devices involved in the execution.

Secondly, the mechanism for distributing and executing queries and the set of objects involved in this process must be identified.

In the following of this paragraph we introduce three possible solutions to manage the interaction between the user and the architecture:

- Queries handled by application objects (a). The user can submit queries to many application objects. Each of them can deal with the declarative language and distributes the received queries to the set of logical objects it is bound to. In this case, the same query submitted to different application objects produces different outputs because it is executed on different sets of logical objects.
- Queries handled by logical objects (b). The user can submit queries directly to logical objects, completely bypassing the application layer. This is a dummy solution because it requires that the user who submit a query knows the list of logical objects that will participate to the query. Moreover all the different results coming from logical objects have to be merged by the user.
- Queries handled by a query analyzer at application layer (c). The user can submit the query to a dedicated component, placed over the application layer. This component accepts a query in input and then parses it. If no syntactical error is found, the set of logical objects involved in the query execution is identified. To do this, a dedicated application object, that maintains a repository of logical objects, is used by the query parser. Finally, the query is divided into sub-queries that are sent to the executors. Results are then collected by the query analyzer and exposed to the user who submitted the query. The query analyzer, with the application object, forms the DBMS core seen by the user.

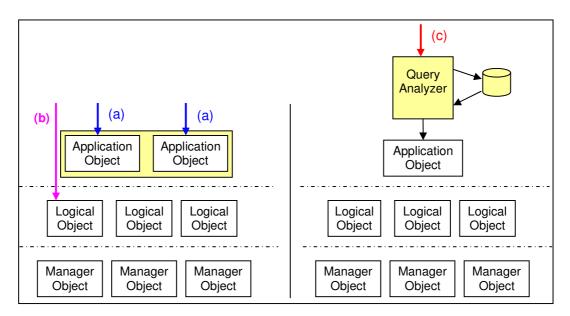


Figure 13: Interaction between the user and the architecture

We didn't consider solution (b) because we want a simple and single entry point for the user, who should not be concerned in the selection of logical objects involved in the query.

We chose to follow the third approach: every high level software that would interact with the network using the declarative language should use the query analyzer and inject queries through it.

3.3 Logical objects interface

Logical objects play an important role in the language-architecture interaction. In fact they are the abstractions of physical devices seen by the language. When a query expressed with the declarative language is parsed, a program for each logical object participating to the query is generated to define the behavior of the wrapped physical device. Independently of its nature, the program must deal with the physical device through the interface provided by the logical object.

In order to allow the query analyzer to build these programs, logical objects should expose a standard interface. We now list the functionalities that should be included in that interface to allow the language support all the requirements identified in *Chapter 2*:

- Retrieving the logical object identifier. Each device must univocally be identified in the system through an identifier and the interface should provide a way to retrieve it. It is not important the data type associated to the identifier: in the following we suppose that each device is associated to an *URI*.
- Retrieving the base station identifier. In many wireless technologies, mobile nodes are connected to a fixed base station through a wireless channel (in the case of RFID the base station is the RFID reader). Furthermore, a node can change the base station it is connected to. Although the connection to the base station can seem just a network issue, there are many situations in which the current base station identifier can be a data of interest for the user (as seen in the mobile platforms example in Chapter 2). The semantics of this field depends on the considered technology. For example in HF-RFID the base station is not the current reader, but the last cached one. If a technology doesn't support the concept of base station, a null value is returned.
- Retrieving attributes. All the other information that can be retrieved from a
 logical object is attributes. An attribute can be the abstraction of a sampled
 value, a cached one or a constant. Attributes can be classified in three
 categories:
 - *Static attributes:* represent constant values describing the characteristics of the node (*e.g.*: type of the node, maximum sampling rate, etc). These values don't change during the logical object lifecycle.
 - *Probing dynamic attributes*: when an object tries to retrieve this kind of fields, the logical object has to deal with the physical device to produce the value before returning it. Probing attributes can be mapped on a real sensor (*e.g.*: temperature, pressure, etc) or on a memory area (*e.g.*: information stored in a device *RAM*, *ROM*, *flash*, etc). Note that dynamic attributes also include policies whose value must be sampled whenever required (*e.g.*: battery level).
 - Non probing dynamic attributes: when an object tries to retrieve this kind
 of fields the logical object returns a local cached value without dealing

with the physical device. The cached value is periodically updated by the logical object.

- *Firing notification events*. As said before, some operations defined in the declarative language are activated after an event is raised. Thus, the logical object interface must be able to fire some events coming from physical devices.
- *Getting the list of supported attributes*. A function should be provided to allow the query analyzer discovering if a logical object supports a certain attribute. That function should also provide the type of a given attribute, both in terms of data type (integer, string, etc) and attribute type (static, probing or not probing).

In this paragraph we only define the required functionalities without giving a formal definition of the interface. In fact, an implementation is not needed to explain the semantics of our language. Note that many different implementations can be designed to support the described functionalities: for example a method for each attribute can be introduced or a single general method *getAttribute* (*attribute_name*) can be provided.

It is now possible to give a formal definition of the sampling concept: from the language point of view, a sample is the reading of a logical object attribute.

In the case of event based sampling, the sampling operation can be defined as the reading of a logical object non probing attribute after an event is detected.

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4 Language design

In this chapter we start presenting our language. More specifically we show how queries are seen by the user and how they are managed by the system. The language syntax with the semantics of each clause will be presented in the next chapter.

Firstly, we analyze how queries are structured from the user point of view and how they will be decomposed when executed. Then we introduce the idea of high and low level queries and of stream and snapshot tables and, using these concepts, we show how a user query can be seen as a graph. We also explain the meaning of abstraction level and we provide a formal classification of queries and sub-queries that can be conceptually found in the system. Finally, starting from a query graph example, we show how a user query is decomposed, sent to logical objects and executed.

The concepts presented in this chapter will be useful to better understand the language semantics.

4.1 Query analysis

Queries expressed with our language are quite complex, because they should describe all data manipulations to transform a set of sensor sampling operations to an output data stream. Thus, we start the presentation of language design showing how user queries can be decomposed in simpler components. In the following paragraphs we will show in detail what these components are and the possible relationships among them. Then, in the next chapter, we will present how each component can be specified using language statements.

The query decomposition can be considered from two different and orthogonal points of view: the user and the system one.

The user can perform a *syntactical decomposition*: the same query can be written as a single complex statement or as a set of simpler statements. For example, a simple query taking data from a sensor and inserting it in an output stream can be written either with a single statement or with a creation statement (to define the output stream) followed by an

4. Language design

insertion one (to control the sampling operation). Allowing the user to define simple queries and combine them to build the final *SQL* statement is similar to introduce the concept of function in a programming language syntax.

As we will see in the next chapter, not all the queries can be written as a single statement, but some choices in syntactical decomposition are left to users. Note that this decision is not very important at this point of the language definition, because it implies only modifications in the language syntax and in the parsing routine, but not in the query execution engine.

From the *system point of view* a user query can be decomposed in some blocks, each one describing a data structure (stream or snapshot) or the behavior of a sub-query (high or low level query). In the following of this chapter we present in details these blocks, showing their features and how they can be combined to build a valid user query.

The language syntax presented in this work is designed to allow user decomposition to be very similar to system decomposition: for example, if a user query contains a high level and a low level query, the user has to write two different statements. As said before, the language grammar can be modified to allow a more structured user decomposition, without changing the concepts presented in this chapter: in the previous example user can be allowed to write a single big statement containing both the high and low level subqueries.

4.2 Query decomposition (from the system point of view)

The building blocks composing a query can be classified in two classes: table definition and data management. The first type of block is used to define intermediate and final data structures that must be computed to produce query results. The second type of block is used to describe the set of operations needed to retrieve data from sensors or from existing data structures and to produce records for other data structures.

Two types of table are supported: *streams* and *snapshots*.

The *stream table* is the most common type of data structure and it is an unbounded table. Each record has a set of user-defined fields and a native timestamp field. The following operations are supported:

- *Insertion*. A new record can be inserted into the stream by a running query. The execution of this operation generates an insertion event that can be detected and used by other sub-queries.
- Reading. A window can be extracted from the stream by a running query. That
 window is defined by a timestamp value, that identifies the most recent record,
 and by a size, that can be expressed either in terms of a number of records or a
 time interval.

The *snapshot table* is a data structure needed by *pilot join* operations and it is a buffer characterized by a time value T that identifies the buffer size. The behavior of this data structure can be understood thinking that there are two memory areas: a local buffer and an output buffer. The time is divided into intervals of length T. All the new records inserted in the snapshot are stored in the local buffer. When the current interval changes, the content of the local buffer is copied to the output buffer and then the local buffer is cleared. When the snapshot is read, the current content of the output buffer (that is the set of records received during the previous interval) is returned.

Two types of query blocks relative to data management exist: *low level queries* and *high level queries*.

A *low level query* is used to define the behavior of a single device (or a group of devices abstracted by a single logical object). The main role of low level statements is to precisely define the sampling operations, but they also allow the application of some *SQL* operators on sampled data.

An object that is executing a low level query has to maintain a local buffer and to perform the following activities:

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- *Sampling data* by reading some logical object attributes and inserting the read values into the local buffer. This operation has to be executed when required by the query: periodically or when an event happens.
- *Performing SQL operations* (selection, aggregation, filtering, grouping, etc) on the current content of the local buffer and inserting obtained records into the stream or the snapshot that was indicated as query output. This activity should be executed when required by the query (if the output is a stream) or periodically with period *T* (if the output is a snapshot with size *T*).

The local buffer is conceptually unbounded and its size increases indefinitely. Practically, the executor should be able to delete old records that are no longer needed to perform required *SQL* operations.

It is worth noticing that *all the operations that can be executed at low level are* relative to data extracted from a single logical object. The goals of these operations should be discarding bad values and optionally aggregating a group of sampled values before sending them to high level queries. Two different low level queries can insert their results in the same stream or snapshot, performing in this way a SQL union operation.

The second type of data management block is composed of *high level queries*. They use one or more streams (generated by low level queries or other high level queries) as input, they perform *SQL* operations on windows extracted from input streams and insert the generated records in an output data structure. The activation of a high level query can be specified either in terms of a time period or in terms of an event (insertion of a record into a stream).

4.3 Query graph

In this paragraph we introduce a graphical notation to represent the concepts presented above. Then, using this notation, we show how a user query can be seen as a graph, whose nodes are streams, snapshots, low level and high level queries. Finally, we list all the constraints that a query graph should comply to represent a valid query.

As shown in *Figure 14*, we use a diamond to indicate a low level query (*LLQ*) and a circle to indicate a high level query (*HLQ*). Streams and snapshots are represented with rectangles having a single and a double border respectively. An exiting arrow can be drawn to indicate that a data structure (stream or snapshot) is an output of the user query. The star is used to indicate a generic set of logical objects that can be sampled by a low level query. Arrows are used to show a dataflow from a query to a data structure or vice versa. Then, a dashed oriented line is used to indicate a *pilot join* operation.

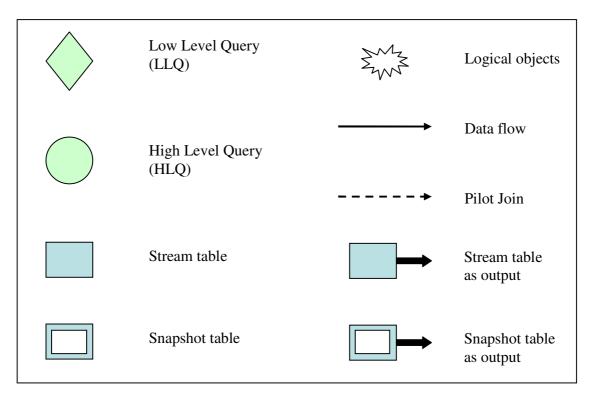


Figure 14: List of symbols used in query graphs

As an example, consider *Figure 15*. It is a graphical representation of a user query composed of two low level and two high level queries.

LLQ1 is used to sample a set of logical objects and to produce a stream. This stream is used as the source of both HLQ1 and HLQ2. HLQ1 generates a snapshot that is used to drive the *pilot join* of LLQ2: the list of logical objects that will execute LLQ2 will be periodically updated looking at the snapshot content. Finally, HLQ2 takes two streams in input and generates the output stream.

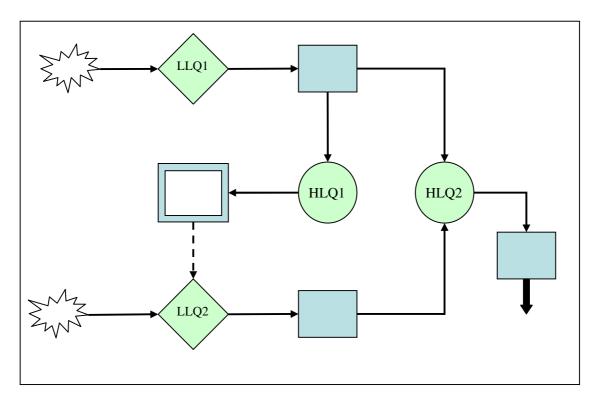


Figure 15: Example of query graph

Considering again the mobile platforms example introduced in the previous chapter, we now show that the graph in Figure 15 can represent the following query: "Execute a temperature sampling on the nodes mounted over the platform that is located nearest to a given point P. Then, return the sensed temperature and the location where the sample was taken."

The query LLQ1 is the query that periodically (with period T) samples platforms position. HLQ1 is activated every T instants to find the platform nearest to the point P. The ID of the base station mounted over that platform is then inserted in the snapshot. The query LLQ2 is activated on all the logical objects abstracting temperature sensors and currently connected to the base station with the ID contained in the snapshot. Finally, the query HLQ2 has the role of adding the platform location to the sampled temperatures using a join operation. Note that the sampling rate of the query LLQ2 should be greater than the rate used to update the snapshot.

In *Figure 16* the list of allowed connections in the query graph is shown. Low level queries can only have a set of logical objects as input and a single stream or snapshot as

output. They can contain a *pilot join* driven by one stream table (*event based pilot join*) or one or more snapshot tables (*condition based pilot join*). High level queries can have one or more streams as input, and a single stream or snapshot as output.

Even if not clear from the figure, two important rules must be respected. The graph, having data structures and queries as nodes and data flow arrows as edges, must be acyclic (*DAG*). Loops can be present in the graph only considering also *pilot join* arrows as edges: in this case each loop must contain at least a dashed edge.

The second rule forces each data structure to be the output of a single high level query or of a set of low level queries. A query is not valid if two different high level queries (or a low level and a high level queries) both insert their results in the same data structure.

The query graph expressivity can be extended adding some labels to the graph edges, as shown in *Figure 16*. These labels can be used to specify:

- The sampling type (event based or time based) of low level queries
- The query activation type (event based or time based) of low and high level queries

The semantics of the different sampling and activation types will be clearer when the complete language grammar will be presented and explained.

Now we consider again the issues related to timestamp management. We will use the graph representation of a query to describe how native timestamps are calculated and propagated. Low level queries are the first elements of the graph that generate native timestamp values. When activated, these queries compute some SQL operations on data contained in their local buffer and insert the obtained records in the output data structure. They set the native timestamp field of all the generated records equal to the current timestamp, i.e. the timestamp in which the query was activated.

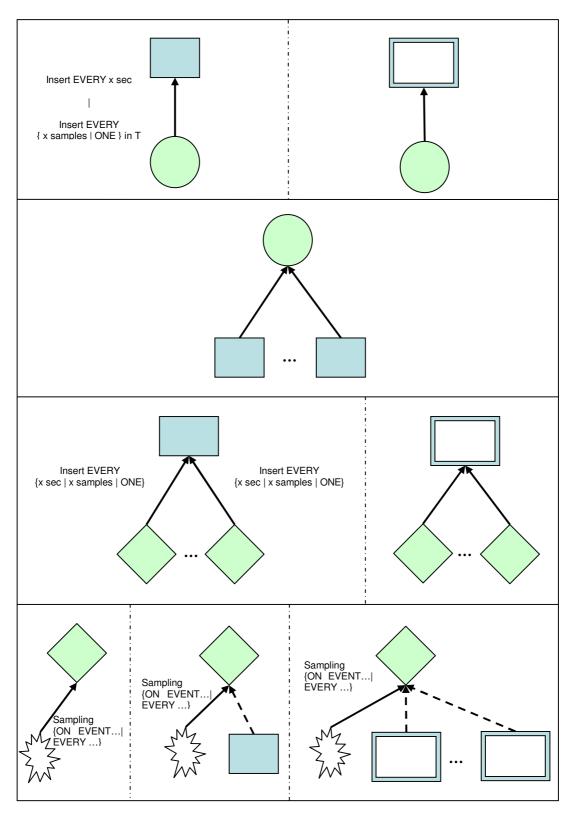


Figure 16: List of allowed connections in query graphs

Instead, high level queries use the following rules to attach a native timestamp value to the produced records. Each record is timestamped with:

- The native timestamp of the record that raised the event, if the query is event based
- The activation timestamp, if the query is activated periodically

To better explain how timestamps are managed by high level queries we now introduce two simple examples (*Figure 17*).

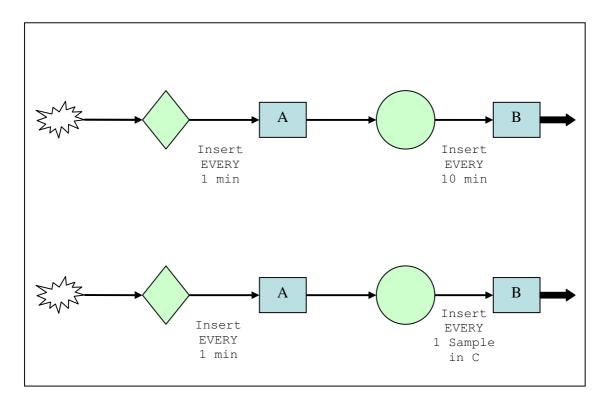


Figure 17: Query graphs and timestamps

The first query graph is composed of a low level query, activated every minute, and of a high level query, activated every 10 minutes. Suppose that the high level query takes the last ten values produced by the low level query to calculate an average. If this query is activated exactly after ten minutes, it should work on the records having a native timestamp in the interval (1 min, 10 min). But probably, due to network latencies, an offset exists such that, at the time 10 min, only records produced before (10 min - offset) are ready to be used.

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To give a clear semantics to the query execution, the system must estimate the maximum value of that offset. Then, the query is activated at the time (10 min + offset) and takes into account only the records having a native timestamp in the interval (1 min, 10 min). In this case, the native timestamp appended to the generated records is set equal to 10 min.

Note that if a record comes with a delay greater than the offset, the high level query is able to identify this anomalous latency and to immediately discard the record.

The second example shown in *Figure 17* is similar to the previous one, but now the high level query is activated by an event. Suppose that a record is inserted in the stream used as events source (*C*) and that it is timestamped with the value *15 min*. This event will be sensed by the high level query at a time greater than *15 min* due to network latencies. The idea is that this query activation must work on data having a native timestamp less than or equal to *15 min*. The generated records will be timestamped with the value *15 min*, too. Note that, to obtain this result, the real query execution can be delayed by an offset due to the same reasons described for the previous example.

To conclude, we defined some rules to assign a native timestamp to all records in the system. We also showed that a query must be executed after a certain delay from the ideal activation instant. This delay should be forecasted by the system and it is dependent on the position of the query in the graph.

This delay has also an impact on the *pilot join* operation semantics. In fact, it is clear that if a query inserts a record timestamped with T in a snapshot, the logical objects matching that record will start executing the query at a timestamp greater than T. This is the reason for which the query generating the snapshot should be very slow with respect to the query containing the pilot join operation.

4.4 Execution of a user query on the architecture for pervasive systems

In this section we show the matching between the existing architecture for pervasive systems and the idea of describing the network behavior via the query decomposition explained in the previous sections. Firstly, we show how a user query is decomposed. Then we analyze the architecture components that receive the obtained sub-queries. Finally we show how the results are generated and sent to the user.

Note that with the term *sub-query* we refer to a command or a set of commands that can be executed by an object of the architecture and that was obtained from the user submitted query. The way in which a language implementation represents these sub-queries is not important: they can be a set of statements expressed in a *SQL* like language or simply procedural programs generated by the query analyzer.

The query decomposition initially operated by the analyzer to extract the graph from the user query is not the only action that is executed. In fact the obtained low level queries can refer to some abstract attributes that should be still translated into concrete attributes, as described in *Chapter 2*.

So there are mainly three types of commands (or sub-queries) that can be found in the system:

- Low level, concrete sub-queries
- Low level, abstract sub-queries
- High level sub-queries

Note that the concept of abstract/concrete cannot be applied to high level sub-queries because they don't deal with sampling operations.

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We now consider a simple example to show the flow of commands and responses that is established during a query execution. The query graph used for the example is represented in *Figure 18*: there are two low level queries that insert data into a stream, used as input for a high level query.

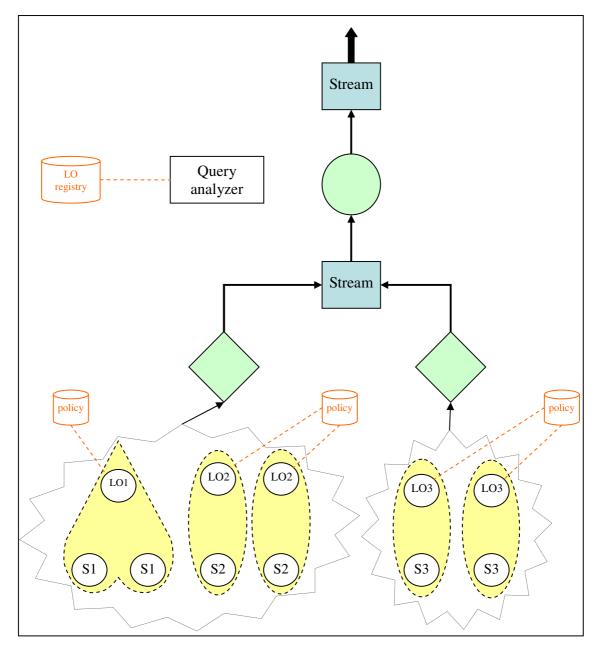


Figure 18: Example of query execution (1/3)

The figure also contains some other components that will be involved in the query execution:

- *The query analyzer*, that is a block representing all the software modules for the query parsing and distribution.
- *The logical object registry*, that represents all software modules and information needed to know the set of logical objects currently instantiated in the system.
- *Physical devices*, that are indicated with *S1*, *S2*, *S3* (note that same labels are used for homogeneous devices).
- Logical objects, that are indicated with LO1, LO2, LO3 (note that same labels are used to indicate different instances of the same logical object class). Logical object LO1 abstracts a group of homogeneous physical devices while LO2 and LO3 abstract a single device.
- *Policies*, that are an abstraction of the rules used by each logical object to perform the translation from abstract to concrete attributes.

The process of query decomposition and distribution is shown in *Figure 19*. Each thick arrow indicates that a set of commands is sent from a component to another one. The red arrow (QI) is the query submitted by the user. The parsing of QI produces:

- One "High level" sub-query (Q2) containing the statements for composing, filtering and aggregating data at the high level.
- Some "Low level, Abstract" sub-queries (Q3, Q4) containing the statements that are submitted to the logical objects and that define the behavior of sensor nodes.

It is worth noticing that the same low level query can be submitted to many different logical objects, also if they abstract different types of physical devices.

After the parser has analyzed the query Q1, buffers for the stream data structures are allocated. Then, the list of nodes that should take part in the query is generated using the registry. Finally the sub-queries Q3 and Q4 are sent to all the logical objects in the list.

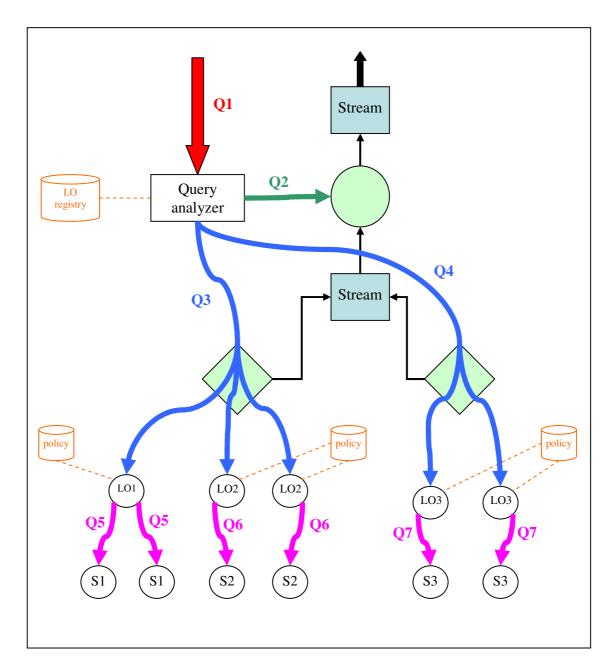


Figure 19: Example of query execution (2/3)

When a logical object receives an abstract command, it translates the command into a concrete one that can be understood by the object manager of the underlying physical device technology. The "Low level, Concrete" commands are indicated in the figure with the arrows labeled *Q5*, *Q6* and *Q7*.

It is to be noticed that the same abstract sub-query can be translated to different concrete sub-queries if it should be executed on different devices, recognizing different concrete policies constraints. For example, in the figure, the abstract query *Q3* has been translated to the concrete query *Q5* for a class of devices and to the concrete query *Q6* for another class of devices.

Figure 20 shows the data flow from the networks nodes to the output stream that is established after the query is initialized. Obviously, commands and data flows can change during the query execution if the list of logical objects involved in the query changes.

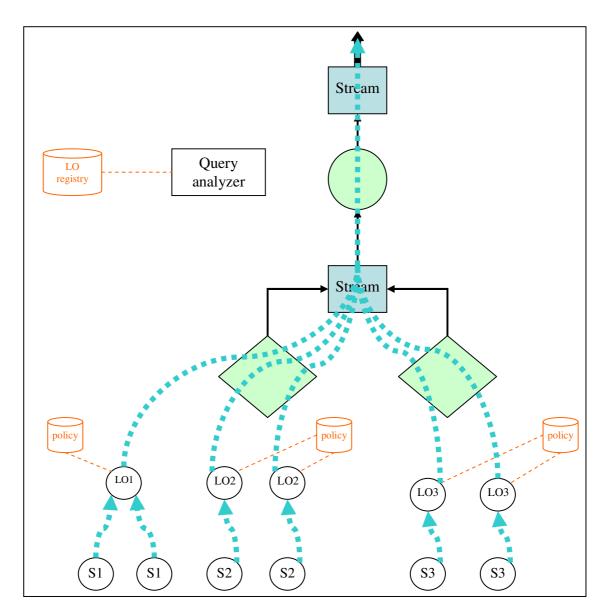


Figure 20: Example of query execution (3/3)

5 Language grammar and semantics

In this chapter we will present a formal definition of our language. The syntax is defined through an *EBNF* grammar and each production is explained and commented.

The grammar has been written to be clear to a human reader and not to be directly used with a parser generator software. In fact, we didn't take care of maintaining the grammar in the *LL* or *LR* classes because we decided to ensure the readability. Obviously, when the language will be implemented, a revision of the grammar is needed.

The language productions are grouped in small sets, each of them relative to a specific type of language statement. Firstly, we introduce the different types of language clauses; then we present the creation and the insertion statements that are needed to create and to populate data structures. We also describe low level and high level selections, needed to collect data from the pervasive system. Finally, we show the structure of the expressions allowed in the language.

In this thesis, grammar productions are reported into a single bordered box and numbered for an easier reference. Terminals are written in uppercase to make them more readily identifiable (*e.g.: SELECT*), while non terminals are put into angle brackets (*e.g.: <Group By Clause>*). Terminals that are composed of one or two characters are put into quotes, so that they cannot be confused with grammar metacharacters (*e.g.: 'l'*).

Furthermore, when two non terminals with the same name are defined for the high level part and the low level part of the language, a prefix is used to distinguish between them. The "*LL*" prefix is used to identify low level productions, while the "*HL*" prefix is used to identify high level productions.

5.1 Types of SQL Statements

All the supported query statements can be classified in two classes: low level and high level statements. The main difference between them is the data sources they work on. As said in previous chapters, low level queries take input data directly from the network nodes, communicating with them through the logical interfaces provided by the underlying architecture (logical objects). So, this kind of queries defines the behavior of

5. Language grammar and semantics

network nodes and specifies when sensors have to be sampled and when collected data should be sent out from the node.

Vice versa high level queries elaborate data coming from other (high or low level) queries. Their semantics is very similar to the semantics of a query language for streaming databases.

```
1. <Sql Statement> →

{ <LL Sql Statement> | <HL Sql Statement> }
```

All the statements, independently of their level, can be classified in two groups that correspond to the *CREATE TABLE* and the *INSERT* statements of standard *SQL*. In fact, each interaction with the system should produce some results that have to be inserted into a data structure. So, at least two kinds of statements are needed: one to define the table structure (creation statement) and the other to define how data should be queried to fill the table (insertion statement).

As said before, since there are two kinds of data structures in the system (stream and snapshot), both creation and insertion statements can be used to create and to populate the two types of tables. As we will show in the following, there are some differences between statements that refer to a stream table and statements which refer to a snapshot table: this is the reason that led to the definition of productions (3), (4), (6) and (7).

The productions introduced up to now identify eight different types of statement as shown in *Figure 21*.

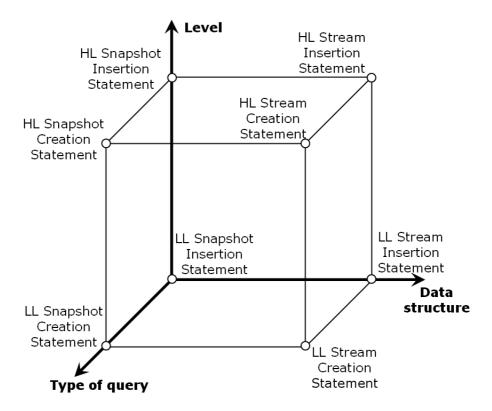


Figure 21: Language statements classification

Every creation statement is composed of a creation clause that defines the name and the fields of the data structure to be created; it allows also an optional part that will be discussed in the following. The creation clause is obviously the same for low and high level queries.

5. Language grammar and semantics

All the insertion statements are composed of an insertion clause followed by a select definition clause. The first one defines the data structure in which query results will be inserted and is the same at low and high level, while the second one is really the query part of the statement and is different for the low and the high level.

It is important to notice the differences between insertion statements that refer to streams and insertion statements that refer to snapshots. In fact, an *EVERY* clause is required when dealing with streams: it is used to define modality and frequency at which the query is computed and the results are inserted in the destination data structure. Otherwise, in the case of snapshots, that frequency is fixed equal to the size of the destination snapshot (for example, if the snapshot is a *30 seconds* buffer the query will be re-calculated every *30 seconds*).

The optional part of creation statements is a shortcut to define a creation query and a correspondent insertion query, writing only a single statement. Note that, when this optional block is specified, the creation query becomes just like an insertion query, except for the first part that requires a data structure creation.

It is not clear from the grammar how many insertion queries can be written to fill the same table. As explained in *Section 4.3*, there are no limits at the low level: a table can be populated by many low level statements at the same time and the union operator is applied among records coming from different queries. Vice versa, only one statement can be written at high level to fill a specified data structure.

The reason for this difference is that, at low level, the same information can be retrieved from very heterogeneous kinds of network node, on condition that different low level queries are used to appropriately set up involved technologies. For example, the monitoring of nodes position can be done both on devices having a *GPS* on board and devices equipped with an *RFID* tag, that indirectly calculate their position starting from the location of the nearest *RFID* reader. In this case, different queries should be written to manage these two kinds of nodes, but their results can be inserted in the same data structure.

```
8. <LL Stream Creation Statement> →
       <Create Stream Clause>
       [AS LOW `:' <LL Every Clause> <LL Select Definition>]
9. <LL Stream Insertion Statement> →
       <Insert Stream Clause>
       LOW ':' <LL Every Clause> <LL Select Definition>
10. <LL Snapshot Creation Statement> →
       <Create Snapshot Clause>
       [AS LOW ':' <LL Select Definition>]
11. <LL Snapshot Insertion Statement> →
       <Insert Snapshot Clause>
       LOW ':' <LL Select Definition>
12. <HL Stream Creation Statement> →
       <Create Stream Clause>
       [AS HIGH ':' <HL Every Clause> <HL Select Definition>]
13. <HL Stream Insertion Statement> →
       <Insert Stream Clause>
       HIGH ':' <HL Every Clause> <HL Select Definition>
14. <HL Snapshot Creation Statement> →
       <Create Snapshot Clause>
       [AS HIGH ':' <HL Select Definition>]
15. <HL Snapshot Insertion Statement> →
       <Insert Snapshot Clause>
       HIGH ':' <HL Select Definition>
```

Note that, in the previous productions, the *HIGH* and the *LOW* keywords have been introduced between the end of each level independent clause (creation, insertion) and the beginning of the next level dependent clause. The reason of this choice will be clearer when the parser implementation will be discussed.

5.2 Creation statements

We now consider creation statements that are very similar to the *CREATE TABLE* construct of standard *SQL*: they define the name of the data structure to be created and the list of fields. No other information must be provided for streams, while the size of the buffer must be specified for snapshots (we call this size "the duration" of the snapshot).

The name, the data type and optionally the default value have to be specified for each field. Besides general *SQL* types (*BOOLEAN*, *INTEGER*, *FLOAT* and *STRING*), *ID* and *TIMESTAMP* are supported because they have a particular meaning with respect to our architecture. *ID* is a special type that can be used to unambiguously identify specific nodes of the network. The real data type associated to *ID* can be a string, if logical objects are identified by an *URI*.

The *TIMESTAMP* data type is used for fields containing a time reference. It is the type of the "*Native Timestamp*" hidden field, that is present in each data structure, but it can also be used for other user-defined fields. The real type associated to timestamps is an integer value.

A particular remark is needed for the default value. We will see that an insertion clause can specify the value only for a subset of the data structure attributes: in this case the remaining fields of each inserted record are set to the default value of the attribute.

Note that the default value is optional in the creation statement: if not specified, a predefined default value is used, whose value depends on the field data type as shown in the following table:

Type	Default value
DOUBLE	0
INTEGER	0
BOOLEAN	FALSE
STRING	,
ID	/
TIMESTAMP	0

Figure 22: Data types and default values

The *OUTPUT* keyword can be used to specify that the created data structure is an output data structure. The system should know this information because the records

inserted in the output tables must be returned to the user who submitted the query. Vice versa, the content of the data structures that are not marked as output tables should only be transferred from the node that generated it to the nodes that will use it.

5.3 Insertion statements

Looking at productions (9), (11), (13) and (15), we can observe that each insertion statement is composed of an introductory part (the insertion clause) that defines the destination data structure, and a second part (the selection definition) that specifies the real query to be done on the sources.

The insertion clause is the same for low and high level queries and contains the name of the destination data structure and the subset of fields for which the selection statement will provide values. Obviously, all the elements in the field list must be attributes of the destination table. As said before, when the destination data structure is a stream, a third component appears in the insertion statement: it is the *EVERY* clause, used to specify how and when the selection part of the statement should be executed and its results appended in the destination table.

Production (22) is the *EVERY* clause for low level statements. If duration is specified, the selection part of the query is activated at regular intervals whose length is

the duration itself. Otherwise, a number of samples can be specified to obtain a completely different behavior: the selection is activated when the specified number of samples is generated by the sampling part of the query. While the first approach is *time based*, this second approach is *event based* and the activation instant of the selection part cannot be foreseen in advance. The *EVERY* clause allows also the *ONE* option that is just a useful shortcut for *1 SAMPLES*.

The semantics of the *EVERY* clause in the high level statements is similar to the one presented above. More specifically, it is exactly the same when duration is specified. However, when the event based approach is used, a table name must be provided as well the number of samples. In fact, high level statements can work on more than one input table, differently from the low level statements that work on data coming from a single logical object. So, in the high level case, the user must specify the name of the data structure that will be used as event source.

A last remark is the possibility to specify the keyword *SYNCHRONIZED* in the low level *EVERY* clause when the time based semantics is used: in this case the system will force the first execution of the selection part at a timestamp that is a multiple of the specified duration.

An example will clarify the concept. Suppose that a low level query, that inserts data into a table every 10 minutes, is started after the evaluation of the EXECUTE IF clause at the minute 27. If the SYNCHRONIZED keyword is not specified, the first activation of the query will be at the minute 37. In the opposite case, the start of the query will be postponed by 3 minutes, so that the first activation of the query will be at the minute 40.

The keyword *SYNCHRONIZED* is provided because there are some situations in which the user must exactly know when data will be inserted into a data structure (*e.g.*: when the content of that data structure is joined with another stream by a high level query). Note that the keyword is not allowed in the high level *EVERY* clause, because all the high level queries start together at the query startup.

```
20. <Insert Stream Clause> →

INSERT INTO STREAM <Data Structure Name> [<Field List>]
```

5.4 Low level SELECT definition

Now we consider the set of rules that define the low level selection. Production (25) shows that the statement is quite complex because it is composed of many clauses. However, some of them are optional and most of the common queries do not require all these clauses.

The selection statement has the role of specifying when data should be collected from the logical object acting as a query source and how it is manipulated to produce output records. A low level query can be decomposed in three main blocks:

- Sampling section specifies when the sensors must be sampled and the conditions to decide if a sampled record is valid, before storing it into a local buffer:
- **Data management section** is periodically activated and executes a selection on the local buffer to produce the results that will be inserted in the destination data structure.
- Execution conditions section defines the rules to establish if a certain logical object should execute the query (EXECUTE IF and PILOT JOIN clauses);

The simplest query a user can write contains only the *SELECT* and the *SAMPLING* clauses. All the network nodes will take part in such a query (this concept will be clearer when the *EXECUTE IF* clause will be presented).

5.4.1 Sampling section

The first clause we analyze is the *SAMPLING* one. First of all, we briefly recall what we mean with the term "sampling". Remember that a low level query is logically executed on a single network node from which data is collected, communicating through the interface provided by the logical object that wraps it. With the term "sampling" we mean that one or more attributes of the logical object are read and the obtained record is inserted into a local buffer, if it satisfies some conditions. Note that the reading of an attribute can imply a sensor sampling if the attribute is dynamic and probing.

As it can be seen in production (26) two types of sampling are supported: *event* based and time based. The first one is obtained writing the *ON EVENT* clause and forces a sampling whenever an event is raised by the logical object. The second is obtained writing the *SAMPLING IF_EVERY* clause and forces the sampling to be executed periodically with a certain frequency.

The *ON EVENT* clause requires the list of events that will be used to drive the sampling operation. The *IF_EVERY* clause is more complex because it allows a parametric definition of the sampling rate. In fact, a construct like an *if-elseif-else* is provided and the user can specify different sampling rates for each branch of the

construct. Moreover, the frequency can be expressed with a general expression, in order to calculate the optimal needed sampling rate (*e.g.*: the sampling rate can be defined as a function of the attribute giving the remaining power).

If at least one "if branch" is defined, the system evaluates the conditions in order of appearance until one of them is satisfied: at this point the sampling rate is set to the value indicated in the correspondent EVERY clause. If no condition is satisfied the sampling rate specified in the ELSE clause is used. Note that if a zero value or a negative value is obtained evaluating the sampling rate expression, the sampling operation will not be performed.

To clarify the distinction between event based and time based sampling, consider the following example. Suppose we want to retrieve the list of *RFID* readers that sense a specific *RFID* tag. The logical object that wraps the *RFID* tag probably has at least a static attribute (the tag *ID*), a dynamic non probing attribute (the last *RFID* reader who sensed the tag) and an event (that signal the sensing). In this situation, a reasonable sampling mode is the event based one: a record is appended to the local buffer whenever the event is raised.

The opposite situation is the case of a node with a temperature sensor on board, that should be sampled every *10 minutes*. In this case, the logical object that wraps the node has a dynamic probing attribute that returns the current temperature. To obtain the required sampling rate, a time based sampling mode must be used.

When the time based sampling is used, other two clauses can be specified in the statement. The first one allows to define the behavior of the node if the required sampling rate is too fast and then unsupported. The "DO NOT SAMPLE" option indicates that the query execution should be suspended until the required sampling rate become again acceptable. Otherwise the "SLOW DOWN" option allows keeping on the query execution, but with an automatic reduction of the sampling rate. If the clause is not specified, the option "DO NOT SAMPLE" is used as the default one.

The second clause is the *REFRESH* one and it allows the user to specify if the sample rate is fixed for the whole query life or if it should be revaluated periodically. If activated, the refresh mode can be event based or time based, exactly as the sampling

mode. If the time based mode is used, the duration is usually set to a very long interval with respect to the sampling frequency of the query. Note that the *REFRESH* clause can be useful also when no "*if branches*" are specified and the sampling rate is defined with a single expression: in this case the refresh forces a new evaluation of that expression.

The SAMPLING clause can be completed appending a WHERE clause, that allows the definition of a filtering condition. If specified, this condition is evaluated whenever a record is produced, before the insertion into the local buffer. If the condition is not satisfied, the record is immediately rejected without inserting it into the buffer. A discarded record is not considered anymore during the query execution. For example if the selection statement is activated every 3 records, the rejected one is not counted for the activation.

5.4.2 Data management section

With the *SAMPLING* clause we have defined the behavior of the low part of the query that reads logical objects attributes and inserts collected values into the local buffer. Now we present the *SELECT* clause that is the most important clause of the query.

The *SELECT* clause is evaluated with a frequency defined in the *EVERY* clause of the insertion query, as described before. The selection is executed on the local buffer that is ideally unbounded. In practice, all the fields specified in the *SELECT* clause are calculated on a bounded subset of the infinite buffer and then the system should be able to maintain only the needed portion of the buffer.

The idea behind the *SELECT* clause is the same as in standard *SQL*: they both have the role of specifying how data have to be collected from data sources (the unbounded buffer in our case; the tables in *SQL*) and manipulated to produce some records as result. However, there are important differences in the semantics, due to our decision of focusing especially on aggregates. In fact, we think that in many real situations the data of interest is directly an aggregation of some samplings rather than the list of all sampled records.

5. Language grammar and semantics

We now start describing the semantics of the data management section starting from the simpler case in which only the *SELECT* clause is specified. In this hypothesis, a single record is generated: each field of this record is an aggregation obtained from the current buffer content. From the grammar point of view, a field of the *SELECT* clause is an expression and can contain aggregation operators.

Differently from standard *SQL*, these operators have more than one parameter. The first one is the value on which the aggregation should be computed and can be an expression involving one or more logical object attributes. The second parameter is a duration or a number of records and represents the portion of the buffer that is used to calculate the aggregation. For example, the element *MIN* (temperature, 20 seconds) requires the minimum temperature value present in the buffer and relative to the last 20 seconds, while the element *MIN* (temperature, 20 SAMPLES) refers to the last 20 records in the buffer (independently of the epoch in which they were generated). Note that aggregations relative to different portions of the buffer can be used in different fields of the SELECT clause or, even, in the same expression. An example is shown in Figure 23.a: the SELECT clause extracts a percentage value indicating how many times a tag was seen by an RFID reader during the last ten seconds with respect to the last minute.

A third optional parameter can be introduced in each aggregate to define a filter. This condition is applied to the records that have to be aggregated, before the aggregation is done. For example the element COUNT (*, 60 seconds, temperature > 50) requires the number of temperature samplings that exceeded a threshold of 50 degrees during the last 60 seconds (Figure 23.b). Note that in the same figure the element COUNT (*, 60 seconds, temperature < 30) is computed starting from the same buffer portion, but applying a different filter condition.

An attribute can be present alone in the *SELECT* clause, out of any aggregation operator: it will be interpreted as a shortcut for *SUM* (attribute, 1 SAMPLES) and then the value of that attribute in the most recent record of the buffer will be returned (Figure 23.c).

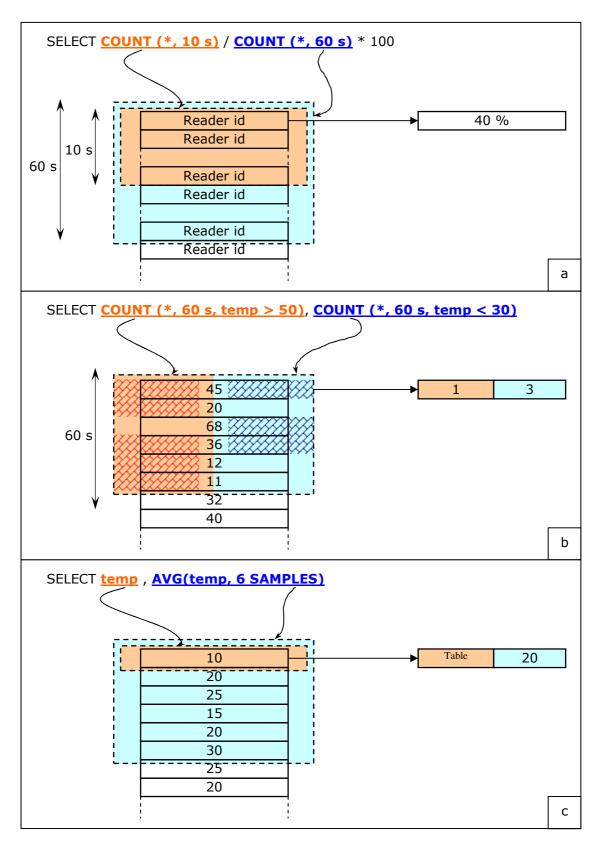


Figure 23: Low level buffer management and SELECT clause

In grammar production (37) each field of the *SELECT* clause is represented as an expression: *Section 5.6* will precisely clarify the set of attributes and operators that can be used in that expression. At the moment, the reader should only know that all the aggregation and standard operators, all the logical object fields and the *ID* and *GROUP_TS* special fields can be used. *ID* returns an unambiguous identifier of the logical object that is executing the query; while *GROUP_TS* returns a timestamp value and its meaning will be clear when the *GROUP BY* clause will be introduced.

It is worth noticing that *DISTINCT* and *ALL* keywords have the same semantics as in standard *SQL*, but they are not useful when the *SELECT* clause is used alone without other optional clauses: in fact, as seen before, in this case only a record is generated by the *SELECT* clause. A default value for each element of the *SELECT* clause can be specified and its meaning will be clear in the following.

We now extend the presented subset of the language with other clauses to add more functionality to the data management part of the statement. The *UP TO* clause allows to retrieve more than one record from the local buffer, giving to the *SELECT* clause a semantics more similar to the standard *SQL* one. A duration or a number of samples can be specified to identify the buffer window that will be returned. For example, if a duration of *1 minute* is specified in the *UP TO* clause, the selection returns all the records in the buffer that are relative to the last minute. This clause doesn't modify the semantics of the aggregations that can appear in the *SELECT* clause. Note that if the *UP TO* clause is not specified, an *UP TO* of *1 record* is taken as a default, falling again within the above explained situation of a simple *SELECT*. This justifies the semantics we gave to single attributes in the *SELECT* clause, when no optional clauses are present. *Figure 24* better clarifies the behavior of the *UP TO* clause.

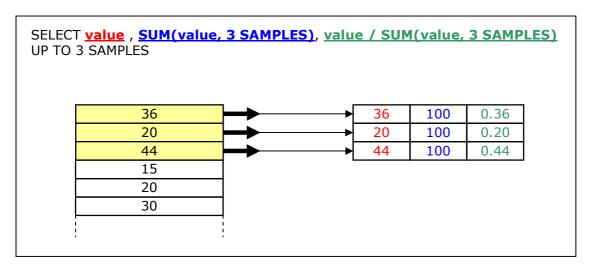


Figure 24: Low level buffer management and UP TO clause

```
42. <Up To Clause> →

UP TO { <Duration> | <Integer Constant> SAMPLES | ONE}
```

The *GROUP BY* clause allows a grouping of the records in the local buffer before the selection process described above is computed. Two kinds of grouping are supported: *attribute based* and *timestamp based*. If the grouping is done on an attribute, the (ideally infinite) buffer is splitted into many (ideally infinite) buffers. Each of them refers to a single value of the attribute. Then, the *SELECT* clause is evaluated on each buffer exactly as described above (*Figure 25.a*).

The second type of grouping is based on timestamps. A duration d and a number of groups n should be specified. In this case the local buffer is ideally replicated into n identical buffers. Then the records relative to the first d * i seconds are removed form the i-th copy of the buffer. Finally the SELECT clause is computed on each of the n buffers (Figure 25.b). The field $GROUP_TS$ retrieves the initial timestamp of the buffer replica from which the output record is generated.

The grouping can be made on more than one attribute simultaneously with an obvious semantics.

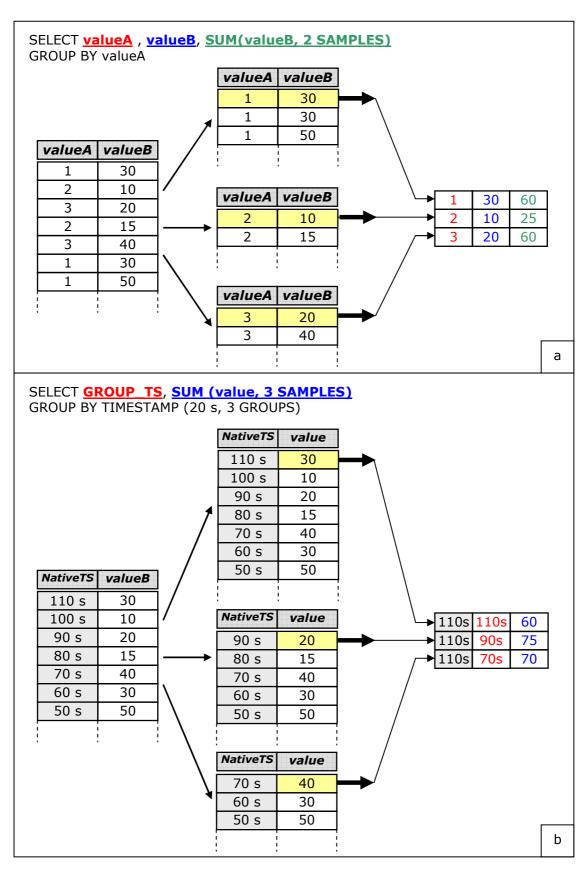


Figure 25: Low level buffer management and GROUP BY clause (1)

The *UP TO* can be used together with the *GROUP BY* clause: in this case the number of records specified by the *UP TO* clause indicates the number of records that will be considered by the selection process from each group that is generated by the *GROUP BY* clause (*Figure 26*).

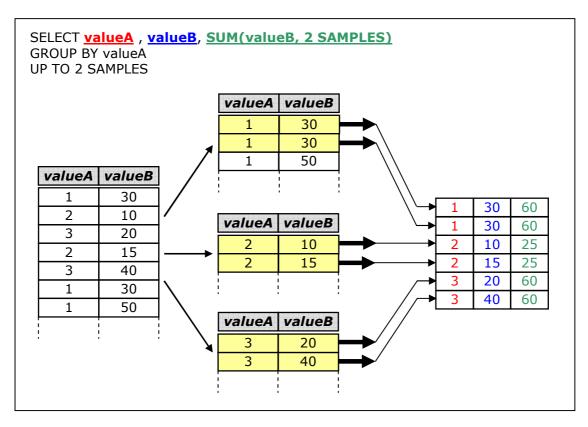
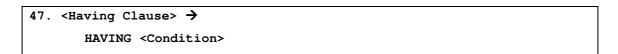


Figure 26: Low level buffer management and GROUP BY clause (2)

We said that when the selection process is activated, the first step that is executed is the application of the *GROUP BY* clause to split the content of the local buffer. Then, the *UP TO* clause is used to determine how many records from each portion of the buffer should be taken to produce output records. Finally, starting from each of these records, the expressions that appear in the *SELECT* clause are evaluated and the obtained records are inserted in the output data structure. The *HAVING* clause allows a filtering operation on the records identified by the *UP TO* clause, before output records are produced. In *Figure 27* an example that includes the *HAVING* clause is shown: records not satisfying the *HAVING* condition are indicated with a prohibition symbol.

Note that the *HAVING* clause covers both the *WHERE* and the *HAVING* clauses of traditional *SQL*. We decided to call this clause "*HAVING*", instead of "*WHERE*", because in many real queries it is used to filter aggregates, instead of single records.



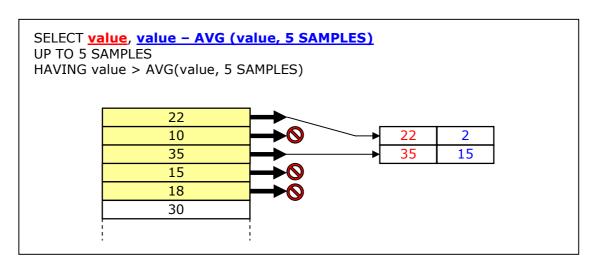


Figure 27: Low level buffer management and HAVING clause

The *EMPTY SELECTION* clause allows the user to specify the behavior of the query when the selection process doesn't produce any output record (this condition can happen, for example, when the *HAVING* clause filters all the records).

Two options are possible: *INSERT NOTHING* means that no records are inserted in the destination data structure when the buffer is empty, while *INSERT DEFAULT* means that, in the same situation, a default record is generated and inserted into the destination data structure. The fields of this record are set to the default values specified by the *SELECT* clause. (If these values are not specified, the default values of the attributes, defined with the data structure creation statement, are used).

If the *EMPTY SELECTION* clause is not specified, the *INSERT NOTHING* option is used as default.

```
48. <On Empty Selection Clause> > ON EMPTY SELECTION { INSERT NOTHING | INSERT DEFAULT }
```

Figure 28 shows a complete example of selection process in which all the clauses presented in this section are involved.

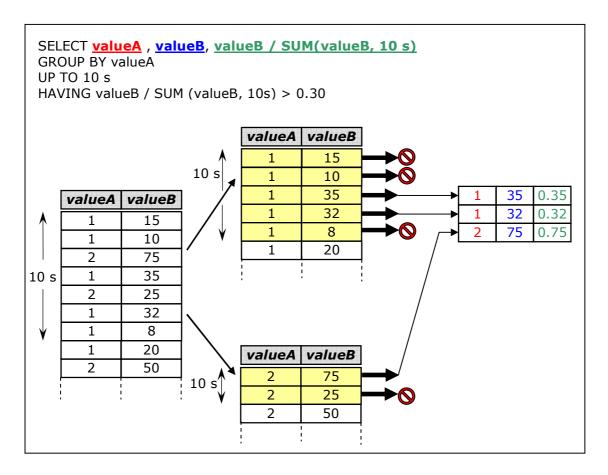


Figure 28: Complete example of low level buffer management

5.4.3 Execution conditions

To complete the analysis of the low level selection statement we now explain the clauses needed to define the set of logical objects that will be involved in the query execution: *PILOT JOIN* and *EXECUTE IF* clauses.

The first one is a special statement that allows a conditional execution of a query on a node. It is the implementation of the *pilot join* operation that was introduced in previous chapters and that was graphically represented as a dashed arrow in *Figure 15*. The condition, the query execution depends on, is based on the content of an existing data structure. If that data structure is a stream, the *pilot join* is event based and, whenever a

new record is appended to that table, the logical objects matching with that record start executing the query. On the contrary, if the data structure is a snapshot, the set of logical objects that execute the query are revaluated periodically with a period equal to the snapshot duration.

The grammar productions reported below show that a *pilot join* operation can be executed on more than one data structure simultaneously. As explained in *Chapter 4* there are some limitations, depending on the type of the used data structures: at most one *pilot join* can be defined with a stream, while an arbitrary number of *pilot joins* can be defined with snapshots. It is not possible to require simultaneously a *pilot join* with a stream and a *pilot join* with a snapshot.

The condition in production (51) must be a Boolean expression and can refer to logical object fields (including dynamic probing attributes) and to the correlated tables' fields. Aggregation operators cannot be present in the condition because it is evaluated before the query starts. Correlated table fields must be referred in the condition using the syntax: "data_structure_name.field_name".

Note that the most common use of the *pilot join* operation is to activate a certain low level query on the wireless sensor nodes that are currently connected to a base station present in a certain list: in this case, the *PILOT JOIN* clause has the following structure:

```
PILOT JOIN BaseStationList ON
    currentBaseStation = baseStationList.baseStationID
```

where "*currentBaseStation*" is a dynamic non probing attribute of the logical object that indicates the *ID* of the base station the node is currently wireless connected to.

5. Language grammar and semantics

The second clause that can be used to impose conditions on the execution of a query is the *EXECUTE IF* clause; it allows to define the set of logical objects that will participate to the query and it is essentially a condition that is evaluated before the query is distributed to network nodes. This condition must comply with the same constraints indicated for the *PILOT JOIN* clause. Moreover, it cannot contain references to correlated table fields because the *EXECUTE IF* condition is evaluated before the analysis of the *PILOT JOIN* clause. A useful Boolean function that can be used in the condition is *EXISTS* (attribute): it returns *TRUE* if the logical object supports the specified attribute, *FALSE* otherwise. If the keyword *ALL* is specified as parameter, the function returns *TRUE* if all the logical object attributes referred in the whole low level statement are supported. Note that this function is not bounded in the *EXECUTE IF* condition and can be used elsewhere in low level language expressions and conditions.

A *REFRESH* clause can be appended to the *EXECUTE IF* one in order to indicate if and when the system should revaluate the condition to update the list of logical objects executing the query. The syntax of the *REFRESH* clause has been explained before, when the *SAMPLING* clause was presented. If the *REFRESH* clause is not specified, the condition is evaluated only when the user submits the query: if satisfied, the logical object will be involved in query execution for the whole query life time; otherwise, if the condition is not satisfied, the logical object is excluded from the query once and for all.

52. <Execute If Clause> → EXECUTE IF <Condition> [<Refresh Clause>]

The last clause that can be specified in a low level selection statement is the *TERMINATE AFTER* one, that can be used to set the timeout after which the logical object will leave the query execution. This timeout can be specified both in terms of a time interval ("duration") and of a number of activations of the query data management section. The timeout starts after the *EXECUTE IF* condition is evaluated if the *PILOT JOIN* clause is not specified. Otherwise, it starts after the *pilot join* condition becomes satisfied.

The *TERMINATE AFTER* clause is optional and, when not specified, each logical object involved in the query continues executing it, until the user requires the termination of the whole submitted query.

```
53. <Terminate After Clause> >

TERMINATE AFTER

{<Duration> | <Integer Constant> SELECTIONS }
```

5.4.4 Low level statements UML

In this section an *UML* activity diagram is reported in order to summarize the concepts presented in previous sections. It aims at explaining the main steps that should be performed by a query engine to execute a low level statement on a logical object. Logical steps reported in the diagram have the role of clarifying the language semantics and of guiding the user that is writing queries. Diagram is intended neither to explain how a real language implementation can work, nor to explain which entities of the architecture should execute the different activities.

Figure 29 represents the logical steps executed by the query engine when all the execution conditions are satisfied: the activity flow on the left-hand side models the sampling section of the query, while the activity flow on the right-hand side models the data management section.

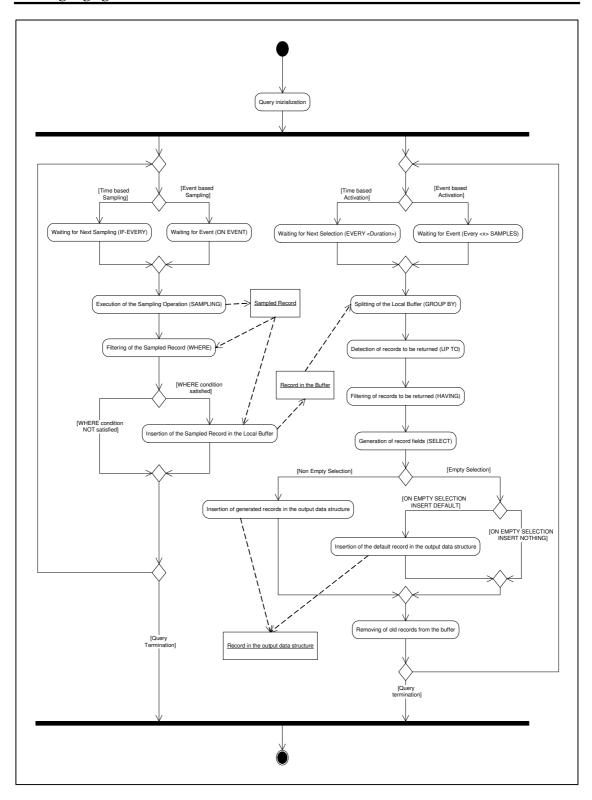


Figure 29: UML activity diagram of low level query buffer management

5.5 High level SELECT definition

The high level selection statement is syntactically and semantically similar to the standard SQL, but there are some important differences. The first one is related to the kind of data used as source. In fact, in standard SQL, each table listed in the FROM clause is a static table, while the sources of our high level queries are streams. We explain how these streams are managed when the FROM clause is presented.

The *UNION* keyword has exactly the same semantics as in standard *SQL* and allows the union of data coming from different selections. Obviously, even if not clear from the grammar, all the *SELECT* clauses must extract the same number and type of fields.

As described for the low level language, the *ON EMPTY SELECTION* clause has been introduced also at high level to ensure that, if needed, at least one record is produced every time the *SELECT* statement is evaluated. This clause can be useful to manage all the situations in which the user would like to know if a missing record is due to a communication error or to an empty selection result.

The overall structure of the *FROM* clause is the same as in standard *SQL*: a list of elements separated by a comma character must be provided; but each of these elements is not just the name of a database table: it defines a window on a stream. Production (58) shows that the window size can be defined either in terms of a time interval or a number of records.

The AS keyword allows the definition of an alias for the window, that have to be used elsewhere in the query to refer to that window. If an alias is not specified the name of the source stream is taken as default. Obviously, the alias definition is mandatory when the same stream is used twice in the same FROM clause.

The windows list in the *FROM* clause allows also the specification of static non streaming tables, whose content is fixed and known by all the sensors in the network. These tables are not obtained extracting a window from a stream, so the user has just to write their name. Static non streaming tables are useful to store parameters and information that were fixed during the setup and the deployment phase of the network. For example, physical connections between devices can be maintained.

Consider the case in which the position of some containers is traced with *GPS*. Suppose that a base station is placed in each container to monitor the temperature of the contained objects. In this situation the mapping between *GPS* nodes and base stations mounted over the same container can be stored in a static non streaming table. This information can be used to force a temperature sampling only in the containers located in a certain zone (using appropriately the *pilot join* feature).

Note that an alternative way to store static parameters, without using static tables, is to expose them through logical objects (in the example above the base station *ID* can be thought as an attribute of the logical objects wrapping the *GPS* devices).

Conditions and expressions used in *SELECT*, *WHERE* and *HAVING* clauses have the same syntax and semantics previously described for low level statements, but there are

some differences in attributes and aggregations; in fact, at the high level, attributes must be fields of a table contained in the *FROM* clause.

Aggregates are functions of one or two parameters. The first one is the name of the attribute the aggregation is relative to, while the second one is an optional *WHERE* clause that allows a filtering before the aggregation is computed. Note that the duration parameter that is supported at low level cannot be used at the high level. In fact, the aggregation is always calculated on the records obtained from the Cartesian product of the windows specified in the *FROM* clause.

So, the main difference between standard *SQL* aggregates and our high level aggregates is the possibility of defining the *WHERE* condition. This feature was introduced to allow users to write in a more compact way queries requiring nested statements when written in standard *SQL*. The above consideration allowed us to avoid supporting nested queries in our language.

Another difference is relative to the semantics of aggregates when the *GROUP BY* clause is not specified: in standard *SQL* all the source records are grouped together to produce a single output record, while in our language an output record is generated for each source record (*Figure 30*). To reproduce the standard *SQL* behavior we introduced the *ALL* keyword in the *GROUP BY* clause: if it is specified all source records are grouped together. Otherwise, whenever the *GROUP BY* clause specifies a list of fields, the semantics of the query is exactly the same as the standard *SQL* one.

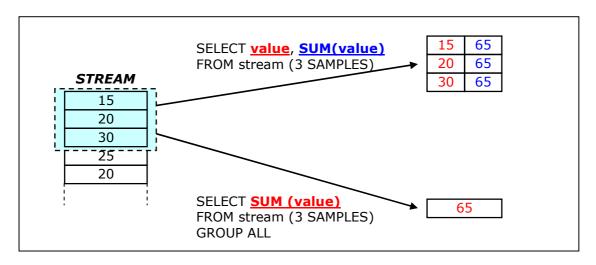


Figure 30: High level aggregations and GROUP BY clause

5.6 Expressions and conditions

Expressions and conditions are required in many language clauses. Although they have a common general structure and they support the same set of the operators, different attributes and aggregations can be used in different clauses. In fact, some clauses belong to high level statements, some others belong to low level statements; some clauses are evaluated before the queries are started, some others are evaluated during query execution. For these reasons, the set of available objects depends on the context in which an expression or a condition appears.

In this section we present the formal grammar of expressions, the complete set of basic blocks and the rules to know which blocks are available in a given context.

Although expressions and conditions are conceptually quite different for the user, they are both described with the same grammar productions. In fact, a condition is just an expression whose value is a Boolean data type. The grammar defines the set of operators that are allowed into expressions and fix their priorities, but doesn't perform type checking: a query that is valid from the grammar point of view can contain type errors.

As described in *Chapter 7*, these errors are handled by the parser, but not taking advantage of grammar features.

OPERATORS	DESCRIPTION					
OR	Boolean disjunction					
AND	Boolean conjunction					
XOR	Boolean exclusive					
NOT	Boolean negation					
IS NULL	Is null value					
BETWEEN	Between an inclusive range					
LIKE	Search for a pattern					
<	Less than					
>	Greater than					
<=	Less than or equal					
>=	Greater than or equal					
<>	Not equal					
!=	Not equal					
I	Bitwise disjunction					
&	Bitwise conjunction					
۸	Bitwise exclusive					
<<	Shift left					
>>	Shift right					
!	Bitwise negation					
+	Addition					
-	Subtraction					
*	Multiplication					
1	Division					
±	Sign					
()	Round brackets					

Figure 31: Set of supported operators

Priority

5. Language grammar and semantics

We now report the grammar of expressions, starting from the non terminal *Expression* and entering into the smallest details until reaching basic blocks. In this way, operators are introduced in increasing priority order (*Figure 31*).

The first introduced operators' class is that of Logical operators. Note that, although we use the word "*Boolean*" for non terminals, these operators are intended to work with a three valued logic.

```
65. <Expression> →
       <Expression Boolean Term> { OR <Expression Boolean Term> }*
66. <Expression Boolean Term> →
       <Expression Boolean Factor> { AND <Expression Boolean Factor> }*
67. <Expression Boolean Factor> →
       <Expression Boolean Test> { XOR <Expression Boolean Test> }*
68. <Expression Boolean Test> →
       { NOT }* <Expression Boolean Primary>
       [IS [NOT] { TRUE | FALSE | UNKNOWN }]
69. <Expression Boolean Primary> →
       <Expression Bit>
       [{ <Comparison Operator> <Expression Bit> |
       [IS [NOT] NULL] |
       BETWEEN <Expression Bit> AND <Expression Bit> |
       LIKE <String Constant> }]
70. <Comparison Operator> →
       { '<' | '>' | '=' | '>=' | '<=' | '<>' | '!='}
```

The following productions introduce bitwise operators.

```
71. <Expression Bit> →
{ <Expression Bit Term> { `|' <Expression Bit Term> }
```

The following productions introduce standard algebraic operators.

"Expression Value Primary" non terminal represents what we have previously called "basic block" of the expression. As said before, the grammar productions that defines this non terminal depends on the context in which the expression is used. In order to model this feature with the grammar, the whole set of productions describing an expression should be replicated once for each possible context, changing only the definition of the "Expression Value Primary" non terminal. To reduce grammar complexity and improve readability we decided to define that non terminal once, allowing every possible choice in all the contexts. In the following we will report a table that precisely shows which choices are valid in each context.

ID and GROUP_TS are special fields whose meaning was described above. A "Logical Object Field" is the name of a logical object attribute and, then, it is defined as an identifier. A "Pilot Join Field" is the name of a column of a data structure used to drive the pilot join operation. The user must specify such a field indicating both the data structure name and the field name, separated with a dot. A "Window field" is a column of a window extracted from a data structure: it can be referred both with an identifier and using the dot notation (in this case the identifier on the left of dot is the alias of the window).

```
85. <Window Alias> →

<Identifier>
```

"Constant" non terminal refers to numeric, logic, string or NULL values and it will be defined in the next section. It is worth noticing that numeric values are unsigned because their sign is already defined with the unary operators ("+" or "-") of production (78).

"Function Call" non terminal has been introduced to allow the user calling an external function from our language (e.g.: string functions: TRIM, LENGTH, etc). A function can have an arbitrary number of parameters and the language grammar allows passing arbitrary expressions to these parameters. Obviously, a real implementation of the language should define the supported set of external functions and provide their implementations.

"Exist Attribute" is a predefined function that allows to inspect the logical object structure. As said before, it takes the name of a field as the only parameter and returns a Boolean value indicating if the specified field exists or not in the logical object that is executing the query.

```
88. <Exists Attribute> ->

EXISTS `(' { <Logical Object Field> | ALL } `)'
```

Figure 32 reports a table that specifies which of the previous basic blocks can be used in each context: four different sets of basic blocks are identified. Expressions (and conditions) that appears as parameter of aggregates or functions are not reported in the table: the rule that has to be applied in this case is that these expressions should comply

5. Language grammar and semantics

with the same constraints of the expression in which the aggregate or the function is contained. For example, a low level aggregate cannot be part of an expression that is used as a parameter of a function contained in the *EXECUTE IF* clause, but the same aggregate can be used if the function is contained in the *SELECT* clause.

		Low Level SELECT	Dow Level HAVING	Duo Low Level WHERE Clause	Low Level IF-EVERY Clause	Low Level PILOT JOIN Clause	Clause	High Level SELECT	High Level WHERE	High Level HAVING
Logical Object Identifier	ID	Expr. Cond.		Cond. E/C Cond.			©	Expr. Cond. Cond.		
Low Level Group Timestamp	GROUP_TS									
Logical Object Attribute (Static – Dynamic Probing – Dynamic Non Probing)	LogicalObjectField	©		©			②			
Pilot Join Data Structure Field	PilotJoinField			②						
Window Field	WindowField						(②		
Constant Value	Constant			②			②	②		
External Function Call	FunctionCall	②		②		②				
Exists Function	Exists Attribute			©		②	(
Low Level Aggregation With / Without Condition	LLAggregate						0			
High Level Aggregation With / Without Condition	HLAggregate									

Figure 32: Table of allowed expressions in different contexts

5.7 Constants and identifiers

For completeness, in this section we report a set of productions representing simple grammar rules used to define the valid syntax of identifiers, numerical and literal values, etc.

Productions (89), (90), (91) and (92) define the sets of characters that can be used:

```
89. <Digit>  
{ '0' | '1' .. '9' }

90. <Literal>  
$\text{ 'A' .. 'Z' | 'a' .. 'z' }}

91. <Non Single Quote Char>  
$\text{ # all printable ASCII Chars, except single quote (') #}

92. <Non Double Quote Char>  
$\text{ # all printable ASCII Chars, except double quote (") #}
```

The following grammar rules define the syntax of identifiers in a classical way: an identifier starts with a literal or underscore and is composed only of literals, digits and underscores. Logical object fields, logical object events, data structure fields and data structure names are just identifiers. We introduced all of them to produce a clearer grammar.

Production (95) lists all the supported data types. Note that in this version of the language we have not introduced a *DataTime* type to limit the grammar complexity. In

5. Language grammar and semantics

fact, this kind of data type is not absolutely necessary due to the existence of timestamp values.

```
95. <Field Type> ->
{ ID | TIMESTAMP | BOOLEAN | INTEGER | FLOAT | STRING }
```

The following productions define the syntax of valid constant values for each data type. The "Signed Constant" non terminal represents the complete set of values that can be used in the language.

```
96. <Boolean Constant> →
       { TRUE | FALSE }
97. <Integer Constant> →
       { <Digit> } +
98. <Float Constant> →
       { <Integer Constant> [ `.' <Integer Constant> ] |
         `.' <Integer Constant> }['E' { '+' | '-' } <Integer Constant>]
99. <String Constant> →
       { <Single Quoted String Value> |
         <Double Quoted String Value> }
100. <Single Quoted String Value> →
       { ''' { <Non Single Quote Char> | '''' }* ''' }
101. <Double Quoted String Value> \rightarrow
       { `"' { <Non Double Quote Char> | `""' }* `"' }
102. <Null Constant> →
       NULL
```

In many parts of the language the concept of duration appears: it is a period of time that can be expressed with a floating point value followed by a time unit.

5.8 Summary of the grammar

A summary of the grammar presented in this chapter is reported in the following. The notation used for this summary is not as formal and precise as an *EBNF* grammar can be, but it is very intuitive and gives an overall view of the language.

Keywords are written with a bold font, while fields that should be completed by the user are enclosed within angle brackets. Square brackets indicate that a clause is optional, and a sequence of three dots indicates that a user can insert a list of elements. Underlined keywords are used to indicate the default option, when an optional clause is not specified.

5.8.1 Creation statements

```
CREATE [OUTPUT] STREAM <Data Structure Name>
   ( <Data Structure Field> <Field Type> [DEFAULT <Signed Constant>], ... )
```

```
CREATE [OUTPUT] SNAPSHOT <Data Structure Name>
  ( <Data Structure Field> <Field Type> [DEFAULT <Signed Constant>], ... )
WITH DURATION <Duration>
```

5.8.2 Low level creation statements

```
CREATE [OUTPUT] SNAPSHOT <Data Structure Name>
  ( <Data Structure Field> <Field Type> [DEFAULT <Signed Constant>], ... )
WITH DURATION <Duration>
AS LOW:
  #LLSelection
```

5.8.3 Low level insertion statements

```
INSERT INTO SNAPSHOT <Data Structure Name> [( <Data Structure Field>, ...)]
LOW:
#LLSelection
```

5.8.4 Low level selection statement

#LLSelection

```
SELECT [DISTINCT | ALL]
  Expression:
      ID | GROUP_TS |
      EXISTS ( { <Logical Object Field> | ALL } ) |
      <Logical Object Field> |
      <Pilot Join Field> |
      <Constant> |
      <Function> ( [<Expression>, ...] ) |
      <Aggregation operator> ( <Attribute>,
         { <Duration> | <Integer Value> SAMPLES | ONE } [, <Condition>] )
   [DEFAULT <Constant>], ...
[GROUP BY
      TIMESTAMP (<Duration> , <Integer Constant> GROUPS) |
      <Logical Object Field>
   [, <Logical Object Field> ...]
[HAVING <Condition>]
[UP TO
  {
      <Duration> |
      <Integer Constant> SAMPLES |
[ON EMPTY SELECTION { INSERT NOTHING | INSERT DEFAULT }]
SAMPLING
      ON EVENT <Logical Object Event>, ...
         IF <Condition> EVERY <Expression> <Time Unit>, ...
         ELSE EVERY <Expression> <Time Unit>
         EVERY <Expression> <Time Unit>
      [ \hbox{\tt ON UNSUPPORTED SAMPLE RATE } \{ \ \ \underline{\hbox{\tt DO NOT SAMPLE}} \ \ | \ \ \hbox{\tt SLOW DOWN} \ \ \} \, ]
      [REFRESH {ON EVENT <Logical Object Event>, ... | EVERY <Duration> | NEVER}]
   [WHERE <Condition>]
[PILOT JOIN <Data Structure Name> ON <Condition>,...]
[EXECUTE IF <Condition>
   [REFRESH {ON EVENT <Logical Object Event>, ... | EVERY <Duration> | NEVER}]
[TERMINATE AFTER { <Duration> | <Integer Constant> SELECTIONS }]
```

5.8.5 High level creation statements

```
CREATE [OUTPUT] SNAPSHOT <Data Structure Name>
  ( <FieldName> <FieldType> [DEFAULT <DefaultValue>], ... )
WITH DURATION <Duration>
AS HIGH:
  #HLSelection
```

5.8.6 High level insertion statements

```
INSERT INTO SNAPSHOT <Data Structure Name> [( <Data Structure Field>, ...)]
HIGH:
#HLSelection
```

5.8.7 High level selection statement

#HLSelection

```
#HLSelectionClause [UNION [ALL] #HLSelectionClause] ...

[ON EMPTY SELECTION { INSERT NOTHING | INSERT DEFAULT }]
```

#HLSelectionClause

5.	Language	grammar	and	semantics
•	Language	Sidillia	ullu	Schilaticics

In this chapter we present some queries written with the language proposed in this work. These queries are intended to explain all the features of the language.

For each example, we will report the request of the user, the interfaces of the involved logical objects, the code of the query and some comments to describe the behavior and to clarify possible doubts.

The first examples are very simple and are intended to exactly explain the sampling operation, highlighting the differences between event based and time based semantics. Query complexity grows in the next examples, where other features of the language are introduced. The last section of the chapter reports some examples related to a real case study.

In this chapter logical objects interfaces are described using a table, composed of four columns: the field name, the data type, the field type and a short description. Field types are indicated using the abbreviations reported in *Figure 33*.

Field Type	Abbreviation
ID	ID
Static attribute	S
Non probing dynamic attribute	NP
Probing dynamic attribute	P
Event	Е

Figure 33: Abbreviations for allowed field types of logical objects

In many examples the requirements refer to some parameters (e.g.: the ID of a logical object, a threshold, etc.). In the queries these parameters are enclosed within square brackets, to indicate that a real value should be provided before queries are submitted to the query analyzer.

6.1 Low Level queries with event based sampling

6.1.1 Example 1

Get the list of *RFID* readers that sense the passive tag with *ID* [tag], specifying also the transit timestamp.

RFID tag				
Logical object wrapping a single RFID tag				
Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier	
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag	
lastReaderChanged	-	Е	Notifies that the tag has been sensed by a reader	

Query

```
CREATE OUTPUT STREAM Table (readerID ID, ts TIMESTAMP) AS
LOW:

EVERY ONE

SELECT lastReaderID, GROUP_TS

SAMPLING ON EVENT lastReaderChanged

EXECUTE IF ID = [tag]
```

Alternative query

```
CREATE OUTPUT STREAM Table (readerID ID, ts TIMESTAMP)

INSERT INTO STREAM Table (readerID, ts)

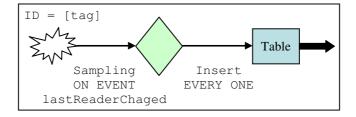
LOW:

EVERY ONE

SELECT lastReaderID, GROUP_TS

SAMPLING ON EVENT lastReaderChanged

EXECUTE IF ID = [tag]
```



The EXECUTE IF condition is specified in order to force the low level query to be executed only on the logical object wrapping the RFID tag having the specified ID. Whenever that tag is physically sensed by an RFID reader, the logical object updates the lastReaderId field and raises the lastReaderChanged event. The SAMPLING clause forces the query engine to be attached to that event and to perform a sampling operation whenever the event is raised.

Since a WHERE clause is not specified after the SAMPLING one, every sampling operation causes the insertion of a new record in the local buffer. The EVERY ONE clause means that the selection on the local buffer is performed once for each inserted record. Every time the SELECT clause is computed, the most recent record in the local buffer is inserted in the output stream: obviously, the lastReaderId field of that record contains the ID of the reader that sensed the tag.

Then, the behavior of the query is that a reader *ID* is immediately returned to the user every time the tag is sensed.

6.1.2 Example 2

Get the list of *RFID* tags that are sensed by the *RFID* reader with *ID* [reader], specifying also the transit timestamp.

RFID tag					
Logical object wrapping a single RFID tag					
Field Name	Data Type	Field type	Description		
ID	ID	ID	Logical object identifier		
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag		
lastReaderChanged	=	Е	Notifies that the tag has been sensed by a reader		
deviceType	STRING	S	Type of device		

Query

```
CREATE OUTPUT STREAM Table (tagID ID, ts TIMESTAMP) AS

LOW:

EVERY ONE

SELECT ID, GROUP_TS

SAMPLING ON EVENT lastReaderChanged

WHERE lastReaderID = [reader]

EXECUTE IF deviceType = "RRFID_TAG"
```

Alternative Query

```
CREATE OUTPUT STREAM Table (tagID ID, ts TIMESTAMP)

INSERT INTO STREAM Table (tagID, ts)

LOW:

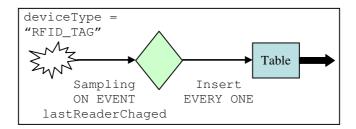
EVERY ONE

SELECT ID, GROUP_TS

SAMPLING ON EVENT lastReaderChanged

WHERE lastReaderID = [reader]

EXECUTE IF deviceType = "RFID_TAG"
```



This query requires to monitor an *RFID* reader and to report all the tags that are sensed by that reader. Suppose that each tag in the system is wrapped by a logical object and that no logical object wraps the *RFID* reader.

To reach the goal, the low level query must be executed on all logical objects wrapping a tag: whenever they are sensed by the reader, they send a record to the output stream.

The *EXECUTE IF* clause limits the set of logical objects that will take part to the query performing a filtering on the *deviceType* attribute: in this way a node that is not a tag (*e.g.*: a *WSN* node) is excluded.

The WHERE clause is needed because the *lastReaderChanged* event is raised whenever a tag is sensed by any of the RFID readers in the system, but we are interested only in a specific reader. Note that, when a different reader senses the tag, the WHERE clause is not satisfied, no new record is inserted in the local buffer and then the SELECT clause is not computed. On the opposite, when the WHERE condition is satisfied, a new record is inserted into the buffer and the behavior is the same as the previous query: the only difference is that, in this example, the ID of the logical object is returned instead of the last sensed reader ID.

6.1.3 **Example 3**

Get the list of *RFID* tags that are sensed by the *RFID* reader with *ID* [reader], only if they had been already viewed at least once in the last *30 seconds*.

RFID tag					
Logical object wrapping a single RFID tag					
Field Name	Data Type	Field type	Description		
ID	ID	ID	Logical object identifier		
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag		
lastReaderChanged	-	Е	Notifies that the tag has been sensed by a reader		
deviceType	STRING	S	Type of device		

Query

```
CREATE OUTPUT STREAM Table (tagID ID) AS

LOW:

EVERY ONE

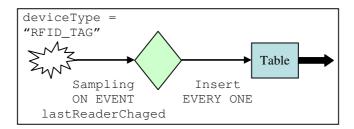
SELECT ID

HAVING COUNT(*, 30 s) >= 2

SAMPLING ON EVENT lastReaderChanged

WHERE lastReaderID = [reader]

EXECUTE IF deviceType = "RFID_TAG"
```



This query is very similar to the previous one: the only difference is the specification of the *HAVING* clause, needed to perform the required control on the last 30 seconds. When a certain tag "T" is sensed by the reader "reader" and the *WHERE* condition is satisfied, a record is appended to the local buffer and the selection is computed. Then, the *HAVING* clause is applied: the records produced during the last 30 seconds are counted. If the result of this operation is equal to one, no record is inserted in the output data structure because the tag was not sensed a second time during the last 30 seconds.

Note that a real implementation of the query executor should maintain a 30 seconds window of the ideally unbounded buffer.

The condition of the *WHERE* clause can be moved into the *HAVING* clause, without changing its semantics:

```
CREATE OUTPUT STREAM Table (tagID ID) AS
LOW:
    EVERY ONE
    SELECT ID
    HAVING lastReaderID = [reader]
        AND COUNT(*, 30 s, lastReaderId = [reader]) >= 2

SAMPLING ON EVENT lastReaderChanged
    EXECUTE IF deviceType = "RFID_TAG"
```

With this version of the query, whenever the tag "T" is sensed by any reader "R" in the system, a new record is appended to the local buffer. The EVERY ONE clause causes the activation of the selection: the record to be returned is filtered with the conditions specified in the HAVING clause. If "R" is not "reader" or "R" is "reader", but the count is less than two, the record is discarded and the set of records generated by the selection is empty. Since the clause ON EMPTY SELECTION is not specified, the default option SEND NOTHING is used and no record is inserted in the output data structure.

This approach is less human readable than the previous one and it requires the insertion of a record in the local buffer whenever the tag is sensed by a reader, even if this record is useless.

6.1.4 Example 4

Perform a 10 minutes monitoring of the tag with ID [tag]. At the end of this period, insert a single record in an output stream to report how many times the tag was sensed by the reader with ID [reader].

RFID tag				
Logical object wrapping a single RFID tag				
Field Name	Data Type	Field type	Description	
ID	ID	ID	Logical object identifier	
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag	
lastReaderChanged	-	Е	Notifies that the tag has been sensed by a reader	

Query

```
CREATE OUTPUT STREAM Table (counter INTEGER) AS

LOW:

EVERY 10 m

SELECT COUNT(*, 10 m) DEFAULT 0

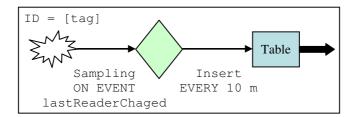
ON EMPTY SELECTION INSERT DEFAULT

SAMPLING ON EVENT lastReaderChanged

WHERE lastReaderID = [reader]

EXECUTE IF ID = [tag]

TERMINATE AFTER 1 SELECTIONS
```



This example is a one shot query: it requires the monitoring of a tag for a short interval and, then, it terminates. Note that all the queries in the previous examples do not specify the *TERMINATE* clause and they continuously run until the user explicitly stops them.

The execution conditions and the selection section of this query are the same as previous ones. So we focus our description only on the data management section.

Although the sampling operation is still event based, differently from previous examples, the selection process is time based and it is activated 10 minutes after the query starts. The local buffer records that are relative to the last 10 minutes are counted, and the obtained result is inserted in the output stream. At this point, the query is stopped because the *TERMINATE AFTER* clause requires the execution of the selection process only once.

A particular remark is needed to explain the query behavior when the tag is never sensed during the monitoring period. In this case, when the selection is computed, the local buffer is empty and no record can be returned although the *UP TO* clause is implicitly set to *ONE*. Since the *INSERT DEFAULT* option of the *ON EMPTY SELECTION* clause is specified, a default record is generated and appended to the output stream. The *counter* field of that record will be set to the value 0, as specified with the *DEFAULT* keyword in the *SELECT* clause.

6.1.5 **Example 5**

Every 10 minutes get a set of records, each of them related to one of the 10 minutes. All the tags that were sensed by the reader with ID [reader] all least three times during a minute must be returned, indicating also the final timestamp and how many times the tag was seen.

RFID tag					
Logical object wrapping a single RFID tag					
Field Name Data Type Field type Description					
ID	ID	ID	Logical object identifier		
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag		
lastReaderChanged	-	Е	Notifies that the tag has been sensed by a reader		
deviceType	STRING	S	Type of device		

Query

```
CREATE OUTPUT STREAM Table (tagID ID, ts TIMESTAMP, counter INTEGER) AS

LOW:

EVERY 10 m

SELECT ID, GROUP_TS, COUNT(*, 1 m)

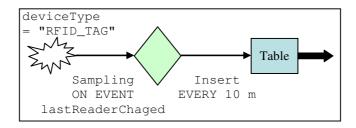
GROUP BY TIMESTAMP(1 m, 10 GROUPS)

HAVING COUNT(*, 1 m) >= 3

SAMPLING ON EVENT lastReaderChanged

WHERE lastReaderID = [reader]

EXECUTE IF deviceType = "RFID_TAG"
```



The only feature of the language used in this query and not yet explained in the previous examples is the *timestamp grouping*. Every ten minutes, when the selection is computed, the content of the local buffer is logically replicated in ten copies (10 *GROUPS*). The first replica contains all the records having a native timestamp less or equal to the current timestamp; the second replica contains all the records having a native timestamp relative to at least one minute before; and so on, until the tenth replica that contains data with a native timestamp relative to at least nine minutes before.

At this point the *HAVING* clause is applied to each replica of the buffer: records that are not discarded are appended to the *Table* stream.

6.2 Low level queries with time based sampling

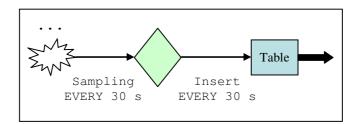
6.2.1 Example 6

Get the *ID* and the temperature of all temperature sensors placed in the area *Z*, sampling every *30 seconds* and assuming that all sensors are fixed.

WSN node						
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor					
Field Name	Field Name Data Type Field type Description					
ID	ID	ID	Logical object identifier			
temp	FLOAT	P	Temperature sensor			
locationX	FLOAT	S	Sensor location – X coordinate			
locationY	FLOAT	S	Sensor location – Y coordinate			
deviceType	STRING	S	Type of device			

Query

```
CREATE OUTPUT STREAM Table (sensorID ID, temperature FLOAT) AS
LOW:
    EVERY 30 s
    SELECT ID, temp
    SAMPLING
     EVERY 30 s
    EXECUTE IF deviceType = "SENSOR"
    AND is_in_zone_Z(locationX, locationY)
    AND EXISTS (temp)
```



This simple query introduces the time based sampling: the nodes involved in the query execution are sampled every 30 seconds (SAMPLING EVERY 30 s). The selection section of the query is activated with the same frequency, obtaining in this way a

semantics that is exactly equivalent to requiring the sampling operation whenever a record is inserted in the local buffer (*EVERY ONE*). Note that this particular circumstance, in which the time based activation is equivalent to the event based one, is not usually true. In fact, for example, the two semantics become different just appending a *WHERE* condition to the *SAMPLING* clause: when a sampled record is discarded, the selection is computed only if the time based activation is used.

Some considerations about the *EXECUTE IF* clause are needed. First of all, the query properly works only under the condition that temperature sensors are fixed or cannot enter or exit the area Z. In fact, if this condition is not satisfied, a node that is in the area Z while the *EXECUTE IF* clause is evaluated, keeps on sampling also after exiting the area. Vice versa, a node that is not in the area Z while the *EXECUTE IF* clause is evaluated, will not start sampling even if it will enter the area. The simplest way to handle moving nodes (*i.e.* nodes whose *locationX* and *locationY* logical object fields are dynamic) is to add a *REFRESH* clause to the *EXECUTE IF* condition.

The second consideration about the *EXECUTE IF* clause is relative to the evaluation of the condition when some logical objects do not expose the *locationX* or the *locationY* attributes: in this case *NULL* values are passed to the "*is_in_zone_Z*" function. If this external routine returns *FALSE* the *EXECUTE IF* condition is not satisfied, while if it returns *NULL* the condition becomes *UNKNOWN*, and the *EXECUTE IF* is still not satisfied.

6.2.2 **Example 7**

Every minute, sample the temperature and the pressure on the nodes having at least a temperature sensor on board. Then, after 10 minutes, return the record (or the records) having the maximum temperature value.

WSN node					
Logical	Logical object wrapping a single WSN node equipped with a temperature sensor				
Field Name Data Type Field type Description					
ID	ID	ID	Logical object identifier		
temp	FLOAT	P	Temperature sensor		
pressure	FLOAT	P	Pressure sensor		

Query

```
CREATE OUTPUT STREAM Table (sensorID ID, temperature FLOAT, pressure FLOAT) AS

LOW:

EVERY 10 m

SELECT ID, temp, pressure

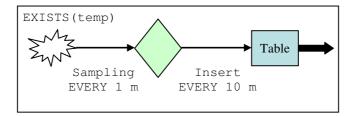
HAVING temp = MAX (temp, 10 m)

UP TO 10 m

SAMPLING

EVERY 1 m

EXECUTE IF EXISTS (temp)
```



The language feature introduced with this example is the *UP TO* clause, that is needed because the results that have to be returned are not aggregate values. The *HAVING* clause is used to filter the records considered by the *UP TO* clause, which do not have the maximum temperature value.

Note that the *EXECUTE IF* condition requires only the existence of the *temp* attribute. If the logical object provides this attribute, but not the *pressure* one, it will participate to the query and it will return a *NULL* value when a pressure sample is required.

6.3 Low level and High level queries

6.3.1 **Example 8**

Every 30 seconds get the average temperature of the sensors placed in the area Z sampling them depending on the following criteria:

- High power level: sample every 10 seconds and returns the average of 3 values
- Medium power level: sample every 30 seconds
- Low power level: don't sample (don't participate in the query)

Sampling rates should be updated once every hour.

WSN node						
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor					
Field Name	Field Name Data Type Field type Description					
ID	ID	ID	Logical object identifier			
temp	FLOAT	P	Sampled temperature			
locationX	FLOAT	S	Sensor location – X coordinate			
locationY	FLOAT	S	Sensor location – Y coordinate			
powerLevel	FLOAT	P	Current power levels of the devices expressed in a 1-100 scale			

Query

```
CREATE STREAM TempsInZ (temperature FLOAT) AS

LOW:

EVERY 30 s

SELECT AVG(temp, 30 s)

SAMPLING

IF powerLevel > 0.70 EVERY 10 s

ELSE EVERY 30 s

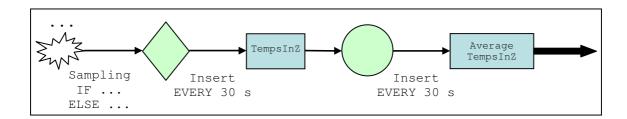
REFRESH EVERY 1 h

EXECUTE IF

EXISTS (ALL) AND is_in_zone_Z(locationX, locationY) AND powerLevel < 0.3

REFRESH EVERY 1 h
```

```
CREATE OUTPUT STREAM AverageTempsInZ (temperature FLOAT) AS
HIGH:
EVERY 30 s
SELECT AVG(temperature)
FROM TempsInZ (30 s)
```



This example is solved using a low level statement to specify the sampling parameters and, if needed, to perform the time average of data collected by a single node. The required spatial average is then computed by an apposite high level query.

It is worth noticing how the sampling frequency is specified in terms of the current power level. If it is low (< 30 %) the EXECUTE IF condition become FALSE and the

logical object do not participate to the query. Otherwise, the sampling condition is evaluated to establish the sampling rate: if the power level is high (> 70 %) the interval between two subsequent sampling operations is set to 10 seconds, else it is set to 30 seconds.

The correct behavior of the query can be obtained using the *REFRESH EVERY* clause, in order to force the execution of the above described process once every hour. Note that this clause must be specified both for the *EXECUTE IF* and the *SAMPLING* clause: otherwise, only a part of the whole process will be executed. Suppose, for example, to require only the refresh of the *EXECUTE IF* clause: a device having a high power level will not reduce its sample rate when the remaining energy becomes lower than the high threshold. Only when the power level reaches the low threshold the *EXECUTE IF* clause will be revaluated and the query will be stopped.

In the low level query the AVG aggregate, calculated on the last 30 seconds, performs the average of a different number of temperature samples depending on the current sampling rate: one record is used if the sampling time is equal to 30 seconds, while 3 records are used if the sampling is equal to 10 seconds.

The high level query is activated every 30 seconds and computes the spatial average of all the records arrived in the *TempsInZ* stream during the last 30 seconds: in this way the considered window certainly contains at most one record for each logical object that is executing the query.

6.3.2 Example 9

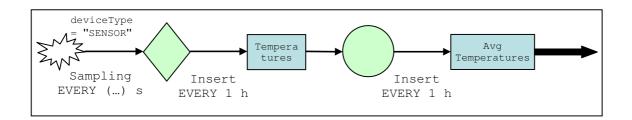
Every hour, get a record with the average temperature sensed at distances (0 m; 5 m), (5 m; 10 m), $(10 \text{m}; +\infty)$ from a fixed point P. Suppose that the temperature sensors are fixed, too. Use a greater sampling rate for sensors nearer to P.

WSN node					
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor				
Field Name	Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier		
temp	FLOAT	P	Sampled temperature		
locationX	FLOAT	S	Sensor location – X coordinate		
locationY	FLOAT	S	Sensor location – Y coordinate		
deviceType	STRING	S	Type of device		

Query

```
CREATE STREAM Temperatures
    (distanceFromP FLOAT, TempSum FLOAT, TempCounter INTEGER) AS
LOW:
    EVERY 1 h
    SELECT dist_from_P(locationX, locationY), SUM(temp, 1 h), COUNT(*, 1 h)
    SAMPLING EVERY (1000 / dist_from_P(locationX, locationY)) s
    EXECUTE IF deviceType = "SENSOR"
```

```
CREATE OUTPUT STREAM AvgTemperatures
    (temp0_5 FLOAT, temp5_10 FLOAT, temp10_inf FLOAT) AS
HIGH:
    EVERY 1 h
    SELECT
    SUM(TempSum, distanceFromP < 5) / SUM (TempCont, distanceFromP < 5),
    SUM (TempSum, distanceFromP > 5 AND distanceFromP < 10) /
    SUM (TempCont, distanceFromP > 5 AND distanceFromP < 10),
    SUM(TempSum, distanceFromP > 5 AND distanceFromP < 10),
    SUM(TempSum, distanceFromP > 10) / SUM (TempCont, distanceFromP > 10)
    FROM Temperature (1 h)
```



The only interesting point to be noted about this query is that the sampling rate is defined through an expression: in this way the interval between two sampling operations is a function of sensors location. No *REFRESH* clause is specified after the *SAMPLING* clause because the nodes are assumed to be fixed and, then, the value of the sampling rate expression is not time dependent.

The goal of the query is to compute an average of all the temperature samples taken during an hour by all the sensors placed at a certain distance from the point P. The

simplest way to reach this goal is appending all the sampled values to a stream and, then, computing their average using a high level query. The solution presented above uses a more efficient approach: a pre-aggregation of data collected by a single logical object is computed directly in the low level query. Thus, each logical object inserts only a record in the *Temperatures* stream every hour, independently of its distance from the point *P*. Note that this record does not contain directly an average, but a sum and a count, because the average operator is not distributive.

6.4 Queries with pilot join

6.4.1 Example 10

There is a set of tanks, containing some temperature sensors. A *HF-RFID* tag and a base station are mounted on each tank.

Whenever a tank is seen by the *RFID* reader with *ID* [reader], sample all the temperature sensors contained in that tank for a minute and return an average value for each sensor.

HF-RFID tag						
	Logical object wrapping a single HI-RFID tag					
Field Name	Field Name Data Type Field type Description					
ID	ID	ID	Logical object identifier			
lastReaderID	ID	NP	Cashed ID of the last reader that sensed the tag			
lastReaderChanged	-	Е	Notifies that the tag has been sensed by a reader			
linkedBaseStationID	ID	S	ID of the base station mounted over the same tank			
deviceType	STRING	S	Type of device			

WSN node				
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor			
Field Name	Field Name Data Type Field type Description			
ID	ID	ID ID Logical object identifier		
baseStationID	ID	NP	ID of the base station the WSN node is currently connected to	
temp	FLOAT	P	Sampled temperature	

Query

```
CREATE STREAM TanksStream (tagID ID, linkedBaseStationID ID) AS
LOW:

EVERY ONE
SELECT ID, linkedBaseStationID
SAMPLING ON EVENT lastReaderChanged
WHERE lastReaderID = [reader]
EXECUTE IF deviceType = "RFID_TAG"
```

```
CREATE OUTPUT STREAM AvgTemperatures (baseStationID ID, temp FLOAT) AS
LOW:

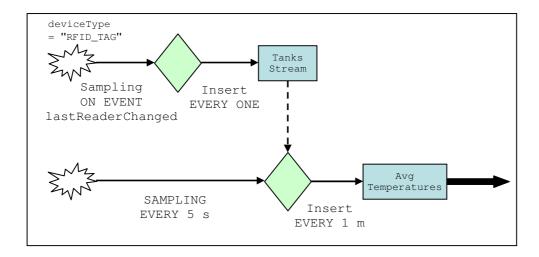
EVERY 1 m

SELECT TankStream.baseStationID, AVG(temp, 1 m)

SAMPLING EVERY 5 s

PILOT JOIN TankStream ON TankStream.linkedBaseStationID = baseStationID

TERMINATE AFTER 1 SELECTION
```



This example is intended to show the *pilot join* operation based on a stream. The first low level query is executed on each *RFID* tag and it inserts a record in *TankStream* whenever a tag is sensed by the reader with the *ID* [reader]. When this event happens, an instance of the second low level query is started on all the logical objects for which the *pilot join* condition is satisfied. This second query performs a sampling operation every 5 seconds, for a minute. Then, an average of the sampled temperatures is computed, a record is inserted in *AvgTemperatures* and the instance of the query is terminated.

6.4.2 Example 11

There is a set of tanks, containing some temperature sensors. A *UHF-RFID* tag and a base station are mounted on each tank.

The temperature sensors contained in the tanks that are in the range area of the *RFID* reader with *ID* [reader] must be sampled once every minute.

UHF-RFID tag				
	Logical object wrapping a single UHF-RFID tag			
Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier	
currentReaderID	ID	P	ID of the reader that is currently sensing the tag	
linkedBaseStationID	ID	S	ID of the base station mounted over the same tank	
deviceType	STRING	S	Type of device	

WSN node				
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor			
Field Name	Field Name Data Type Field type Description			
ID	ID	ID	Logical object identifier	
baseStationID	nID ID NP		ID of the base station the WSN node is currently connected to	
temp	FLOAT	P	Sampled temperature	

Query

```
CREATE SNAPSHOT TanksSnapshot (tagID ID, linkedBaseStationID ID)
WITH DURATION 1 h AS
LOW:
    SELECT ID, linkedBaseStationID
    SAMPLING EVERY 1 h
    WHERE currentReaderID = [reader]
    EXECUTE IF deviceType = "UHF_RFID_TAG"
```

```
CREATE OUTPUT STREAM Temperatures (sensorID ID, temp FLOAT) AS

LOW:

EVERY ONE

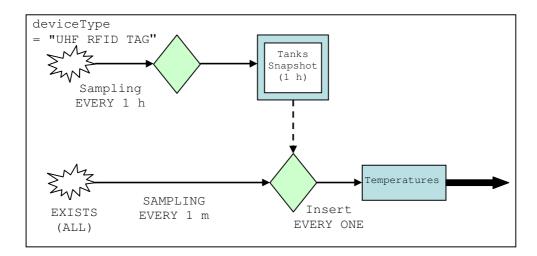
SELECT ID, temp

SAMPLING EVERY 1 m

PILOT JOIN TanksSnapshot

ON TanksSnapshot.linkedBaseStationID = baseStationID

EXECUTE IF EXISTS (ALL)
```



This example is intended to show the *pilot join* operation based on a snapshot. The first low level query fills a snapshot data structure having a duration of one hour. The query is executed on all the logical objects wrapping *UHF-TAGs* and each of them produces at most one record for every snapshot interval.

The second low level query uses the previous snapshot to drive the *pilot join* operation: the sensor nodes, currently connected to a base station that is present in the snapshot list, execute the query and sample their temperature sensor every minute. The *EXISTS (ALL)* is specified to guarantee that the nodes taking part to the query have a *baseStationID* attribute (*i.e.* they are *WSN* nodes) and a temperature attribute (*i.e.* they have a temperature sensor on board).

6.4.3 Example 12

There is a set of tanks, containing some temperature sensors. A *GPS* and a base station are mounted on each tank.

The temperature sensors contained in the tank nearest to point P must be sampled every 30 seconds. The distances of the tanks from the point P must be revaluated every hour.

GPS				
	Logical	object wrapping	g a GPS device	
Field Name	Field Name Data Type Field type Description			
ID	ID	ID	Logical object identifier	
linkedBaseStationID	ID	S	ID of the base station mounted over the same tank	
locationX	FLOAT	P	Sensor location – X coordinate	
locationY	FLOAT	P	Sensor location – Y coordinate	
deviceType	STRING	S	Type of device	

RFID tag				
	Logical object wrapping a single RFID tag			
Field Name Data Type Field type Description				
ID	ID ID		Logical object identifier	
baseStationID ID NP		ID of the base station the WSN node is currently connected to		
temp	FLOAT	P	Sampled temperature	

Query

```
CREATE STREAM TanksPositions (gpsID ID, linkedBaseStationID ID, distanceFromP
FLOAT) AS
LOW:
    EVERY ONE
    SELECT ID, linkedBaseStationID, dist_from_P(locationX, locationY)
    SAMPLING EVERY 1 h
    EXECUTE IF deviceType = "GPS"
```

```
CREATE SNAPSHOT NearestTank (gpsID ID, linkedBaseStationID ID)
WITH DURATION 1 h AS
HIGH:
    SELECT TanksPositions.gpsID, TanksPositions.linkedBaseStationID
    FROM TanksPositions (1 h)
    WHERE TanksPositions.distanceFromP = MIN(TanksPositions.distanceFromP)
```

```
CREATE OUTPUT STREAM Temperatures (sensorID ID, temp FLOAT) AS

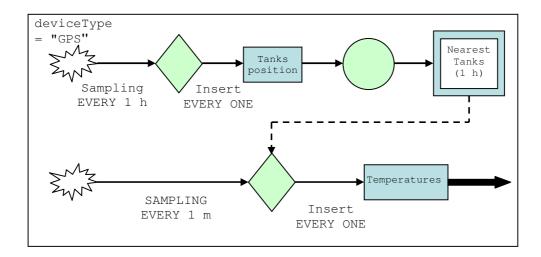
LOW:

EVERY ONE

SELECT ID, temp

SAMPLING EVERY 1 m

PILOT JOIN NearestTank ON NearestTank.linkedBaseStationID = baseStationID
```



This query is very similar to the one in the previous example, with the only difference that a snapshot data structure is populated by a high level query instead of a low level one.

6.5 Other queries

6.5.1 Example 13

Given two temperature sensors, produce a record once a minute indicating the node that sensed the highest average temperature.

RFID tag			
Logical object wrapping a single RFID tag			
Field Name Data Type Field type Description			
ID	ID	ID	Logical object identifier
temp	FLOAT	P	Sampled temperature

Query

```
CREATE STREAM TemperatureA (temp FLOAT) AS
LOW:

EVERY 1 m

SELECT AVG(temp, 1 m)

SAMPLING EVERY 5 s

EXECUTE IF ID = [IDSensA]
```

```
CREATE STREAM TemperatureB (temp FLOAT) AS

LOW:

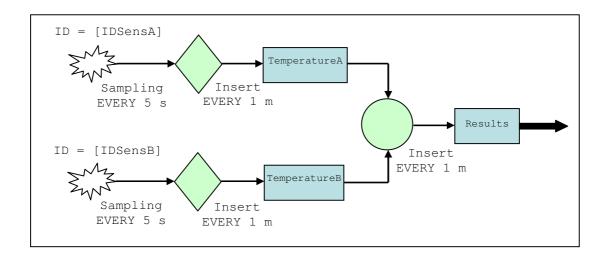
EVERY 1 m

SELECT AVG(temp, 1 m)

SAMPLING EVERY 5 s

EXECUTE IF ID = [IDSensB]
```

```
CREATE OUTPUT STREAM Results (sensor STRING) AS
HIGH:
    EVERY 1 m
    SELECT IIf (TemperatureA.temp > TemperatureB.temp, "A", "B")
    FROM TemperatureA (1 m), TemperatureB (1 m)
```



This example shows a high level query in which two windows are used in the *FROM* clause. The two low level queries produce a record every minute indicating the average temperature sensed during the last minute. The high level query is activated every minute and works on two windows with size of one minute: each of them contains exactly a record and then the Cartesian product generated by the *FROM* clause is composed of only a record.

Consider a variation of the query in which the *FROM* clause uses windows with a size greater than a minute. In this case each window will contain more than one record and their Cartesian product will contain many value pairs. If we are interested only in the comparison of temperatures sampled by the two nodes in the same minute, a *WHERE* condition on native timestamps can be added.

6.5.2 Example 14

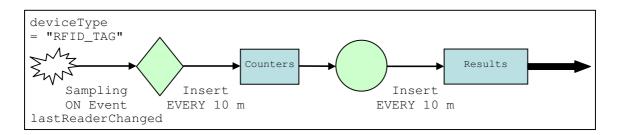
Every 10 minutes, get a record indicating the ID of the tag that was sensed by the RFID reader (with ID [reader]) the maximum number of times during the last 10 minutes.

RFID tag				
	Logical object wrapping a single RFID tag			
Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier	
lastReaderID ID NP Cashed ID of the last reader that sense				
lastReaderChanged - E Notifi		Notifies that the tag has been sensed by a reader		
deviceType	STRING	S	Type of device	

Query

```
CREATE STREAM Counters (tagID ID, counter INTEGER) AS
LOW:
    EVERY 10 m
    SELECT ID, COUNT(*, 10 m)
    SAMPLING ON EVENT lastReaderChanged
    WHERE lastReaderID = [reader]
    EXECUTE IF deviceType = "RFID_TAG"
```

```
CREATE OUTPUT STREAM Results (tagID ID) AS
HIGH:
    EVERY 10 m
    SELECT tagID
    FROM Counters (10 m)
    WHERE Counters.counter = MAX(Counters.counter)
```



No new feature of the language is introduced in this example. The low level query is used to count how many times each tag is sensed by the reader [reader] during the ten minutes period. The high level query is used to find the tag that reported the maximum count.

6.6 Case study queries

In this section we present some queries that can be used in a real context. As a case study, we consider a company which produces high quality wines and gives a great attention to the production process quality. On one side, the company wants to keep under strict control some important parameters of the production cycle to guarantee the quality and the integrity of the product itself; on the other side it also requires the complete traceability of all the operations performed on each of the produced wine lots until the final labeling phase. The availability through the network of information and controls could start as early as with the monitoring of grapes in the vineyard in order to timely identify sudden changes in the values of relevant parameters (*e.g.*: the soil temperature and humidity).

In the following examples we suppose that the devices used for supporting the different phases of the production and delivery processes are distributed as schematically shown in *Figure 34*.

WHERE	WHAT	HOW
vineyard	humidity, temperature, chemicals	sensors
cellar	humidity, temperature	sensors
bottle	tracking information	RFID tag
pallet	tracking information, temperature	RFID tag, sensors
truck	position information	GPS
workers	information system	PDA

Figure 34: Devices usage and types of the case study

6.6.1 Vineyard monitoring

The first query we consider has the role of monitoring environment parameters of the area in which vine is cultivated. More specifically we want to sample temperature and humidity every 30 minutes, returning these values, together with the location of the sensor, only if the sensed temperature is in a critical range. In order to provide more accuracy in the results, each node is required to sample its sensors every ten minutes and to consider the average of the three last values.

	WSN node			
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor			
Field Name	Field Name Data Type Field type Description			
ID	ID	ID	Logical object identifier	
temp	FLOAT	P	Sampled temperature	
humidity	FLOAT	P	Sampled humidity	
locationX	FLOAT	S	Sensor location – X coordinate	
locationY	FLOAT	S	Sensor location – Y coordinate	
deviceType	STRING	S	Type of device	
powerLevel	FLOAT	P	Current power levels of the devices expressed in a 1-100 scale	

PDA				
	Logical object wrapping a worker PDA			
Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier	
temp FLOAT P			Sampled temperature	
locationX	FLOAT	P	Sensor location – X coordinate	
locationY	FLOAT	P	Sensor location – Y coordinate	

In this situation, the user submitted query is only composed of a low level statement:

```
CREATE OUTPUT STREAM EnvironmentParameters
(sensorID ID, temp FLOAT, humidity FLOAT, locationX FLOAT, locationY FLOAT) AS
LOW:
    EVERY 30 m
    SELECT ID, AVG(temp, 30 m), AVG(humidity, 30 m), locationX, locationY
    SAMPLING
    EVERY 10 m
    EXECUTE IF EXISTS(temp) AND is_in_Vineyard(locationX, locationY)
    REFRESH EVERY 5 m
```

The *EXECUTE IF* clause requires the execution of the query on all the nodes currently placed in the yard and having on board a temperature sensor, but it does not restrict the query execution to wireless nodes only. Therefore, suppose that a worker is in the yard with a *PDA* which can sense the current temperature: this *PDA* will execute the query exactly as the wireless nodes (since the *PDA* has not a humidity sensor on board, *NULL* values will be produced).

In this example, it should be noticed that the *locationX* and *locationY* attributes are static for wireless nodes (and they are configured during the network deployment phase), while they are dynamic for *PDA*s: if a worker enters the yard, his *PDA* will start

executing the query, possibly recovering data which should be provided by a "dead" sensor.

The next query shows an important peculiarity of our language: the ability of querying the network state as "normal" data. Suppose that, in the previous example, we want to monitor the power state of the wireless nodes, in order to detect low powered devices and to allow the substitution of the batteries before nodes become inactive.

```
CREATE OUTPUT STREAM LowPoweredDevices (sensorID ID) AS
LOW:
    EVERY ONE
    SELECT ID
    SAMPLING EVERY 24 h
    WHERE powerLevel < 0.15
    EXECUTE IF deviceType = "WirelessNode"
```

If the user wants to know the number of low powered wireless nodes, a high level query performing the count aggregation can be written starting from the stream generated by the previous low level query:

```
CREATE OUTPUT STREAM NumberOfLowPoweredDevices (counter INTEGER) AS
HIGH:
EVERY 24 h
SELECT COUNT(*)
FROM LowPoweredDevices(24 h)
```

6.6.2 Transport monitoring

Now, we consider the monitoring of wine during the transport. In particular, suppose that every truck is equipped with a *GPS* and a base station. Suppose also that each pallet has a temperature sensor used to sense the temperature of the contained bottles. The query requires to produce the list of pallets whose temperature exceeded a certain threshold while the truck was travelling through a given zone, which is considered particularly critical.

GPS				
	Logical object wrapping a GPS device			
Field Name Data Type Field type Description				
ID	ID	ID	Logical object identifier	
linkedBaseStationID	ID	S	ID of the base station mounted over the truck	
locationX FLOAT P Sensor loc		Sensor location – X coordinate		
locationY	FLOAT	P	Sensor location – Y coordinate	
deviceType	STRING	S	Type of device	

WSN node				
Logical o	Logical object wrapping a single WSN node equipped with a temperature sensor			
Field Name	Data Type Field type Description			
ID	ID ID Logical object identifier		Logical object identifier	
baseStationID	ID NP 1 33 to		ID of the base station the WSN node is currently connected to	
temp	FLOAT	P	Sampled temperature	

```
CREATE SNAPSHOT TrucksPositions (linkedBaseStationID ID)
WITH DURATION 1 h AS
LOW:
    SELECT linkedBaseStationID
    SAMPLING
    EVERY 1 h
    WHERE is_in_CriticalZone(locationX, locationY)
    EXECUTE IF deviceType = "GPS"
```

```
CREATE OUTPUT STREAM OutOfTemperatureRangePallets (palletID ID) AS

LOW:

EVERY 10 m

SELECT ID

SAMPLING EVERY 10 m

WHERE temp > [threshold]

PILOT JOIN TrucksPositions

ON baseStationID = TrucksPositions.linkedBaseStationID
```

In this query the *pilot join* operation is used to activate the temperature sampling only on the pallets contained in the trucks that are driving into the critical zone. Note that the interface of logical objects wrapping *GPS* devices has an attribute *linkedBaseStationId* that retrieves the *ID* of the base station mounted on the same truck (this is a static attribute, whose value is defined during the network deployment time).

7 Prototype design

In this chapter the implementation components needed to build a prototype system, able to recognize the language presented in this thesis and to execute queries on a physical pervasive system, are described. Then, the intermediate goals to be achieved during the implementation phase are presented. Finally, the parser architecture is described in details, because at present it is the only implemented component and because it is strictly related to the language grammar presented in *Chapter 5*. A *UML* class diagram is provided to describe the internal representation of input queries, as a useful reference for developers who will implement the other prototype components.

7.1 Prototype components

To find out the required implementation steps, a review of the system architecture presented in *Chapter 3* and *Chapter 4* is needed. Looking at *Figure 12* and *Figure 19*, we can easily outline a set of components that must be implemented:

- *The infrastructure* for pervasive systems
- The operation managers, to provide access to technologies that should be supported
- The logical objects, to wrap physical devices that should be supported
- The query analyzer
- The logical objects registry
- The policy translator

There are also other components, to be designed and implemented, having an important role although they are not directly visible in the figures:

- An algorithm for time synchronization
- A query engine, to execute the queries parsed by the query analyzer
- A simulator of physical devices

7. Prototype design

In the following, a short description is provided for each of the previous components and the expected effort for implementing them is also indicated.

7.1.1 Infrastructure for pervasive systems

As described in *Chapter 3*, the infrastructure for pervasive systems is intended to help the design of a pervasive system from the conceptual and the logical points of view. At present, it is only a specification and then it should be implemented. For the purposes of this work, only a small subset of the architecture features is used and, therefore, only the implementation of this subset is really required.

7.1.2 Operation managers

The set of devices that can potentially be managed with our language is large. In the examples provided in this work, four technologies have been considered: passive *RFID*s, *WSN*s, *GPS* and *PDA*s. So, at least the operation managers that provide access to these technologies are required to be implemented. This is quite a complex task, requiring low level programming skills and hardware knowledge.

7.1.3 Logical objects

A logical object should be designed and developed for each type of physical device that should be supported. These components encapsulate the logic to manage probing and non probing dynamic attributes, as well as the logic to raise events to notify higher layers that something happened in the physical world. The effort in developing logical objects largely depends on the kind and the complexity of the physical device that is being wrapped.

Some special dummy logical objects should be provided to allow the testing of query analyzer and query engine, without using physical devices. These logical objects should not wrap real devices, but only implement an interface returning random data when its attributes are read and, possibly, fire some events at random instants.

7.1.4 Query analyzer

The query analyzer is the interface between the system and the final user, and it allows the submission of queries. It contains a parser that performs the syntactical analysis of queries: if they are error free, a data structure describing the query is generated to be used in further steps of the execution. The query analyzer contains also the logic needed to initialize and distribute the query to the involved logical objects.

The development of this component is quite simple, because a parser generator software can be used to automatically build the parser starting from the language grammar. The trickiest part of this component is the internal representation of queries, because the object framework should be designed to provide a clear view of the query structure and to simplify as much as possible the access to the query details.

7.1.5 Logical objects registry

This component is used by the query analyzer to evaluate the *EXECUTE IF* clause and to determine which objects should take part in the query execution. It maintains the list of the logical objects instantiated in the system, with cached values of their static attributes.

This component is only apparently simple because it should encapsulate all the logic to discover new devices in the pervasive system, to wrap them instantiating the appropriate logical objects and to efficiently maintain the list of instantiated components. The registry should also be able to discover that a device is leaving the system and to destroy the correspondent logical object.

So, although a simple limited implementation of the registry can be initially used, a big effort will be required to design and implement a complete version of it.

7.1.6 Policy Translator

This component deals with the translation of abstract policies into concrete ones: it should implement the rules to transform a concrete value, measured by a node, into an abstract value that can be understood by the user, and vice versa.

7. Prototype design

The policy translator will probably be embedded in logical objects implementation. For example, every logical object will expose a *powerLevel* attribute that is an abstract representation of the remaining energy amount: an internal voltage value (or a count of executed operations) is translated to a percentage value by the policy translator section of the logical object.

7.1.7 Algorithm for time synchronization

Many language clauses require the specification of a time interval: the period between two sampling operations, the period between two query activations, the duration of a snapshot data structure, the lifetime of a low level query are all examples of time intervals used in the language. The semantics defined in *Chapter 5* assumes that all logical objects share a common timestamp: so, they must have a clock and an apposite algorithm is needed to synchronize them.

7.1.8 Query engine

When a user submitted query is parsed, analyzed, decomposed and distributed, the query must be executed. A component for each involved logical object is needed to activate the sampling operations, to manage sampled data and to send the results to a data structure, as required in the user query. Another component is needed to execute high level statements, taking data from streams, performing *SQL* operations and inserting results in another data structure.

The query engine is then a relevant piece of software, composed of different parts, that should be carefully designed. The component that manages sampling operations will certainly be implemented from scratch, while more choices are available to build the data management section (both at low and high level).

Relying on an existing *DBMS* is the easiest solution: in this case only a wrapper is needed to translate the data management part of the query into a set of standard *SQL* statements. Steams and snapshots can be implemented using relational tables of the *DBMS* in a suitable way. Although this solution is the fastest to implement, it is not the best one due to its low performances.

Another approach to implement the query engine is to extend an existing open source *SQL* executor, introducing the required features.

The last solution is to implement the whole engine from scratch, using algorithms optimized for the execution of *SQL* operations on streams [32].

7.1.9 Simulator of physical devices

As said before, dummy logical objects can be used to test some components of the whole system without using physical devices. This approach doesn't allow all kinds of tests because real physical parameters (delays, latencies, packet losses, etc.) are not simulated. For this reason a simulator of the physical world and the physical devices can be useful.

7.2 Implementation steps

The whole project is a very complex and large system. For this reason, building a final and complete release is very hard without fixing some intermediate and incremental goals:

- Implementation of a simple parser, able to recognize the language and to notify syntactical errors.
- Extension of the previous parser to build some objects representing the parsed query.
- Local execution of low level queries on dummy logical objects or on the simulator.
- Local execution of low and high level queries on dummy logical objects or on the simulator.
- Execution of a low level query on a real *WSN* node directly connected to a *PC* through a serial cable.
- Execution of a low level query on a real *RFID* tag, with a reader directly connected to a *PC*.
- Execution of the full prototype on a real pervasive system.

7.3 Parser implementation

The parser is a component of the query analyzer and it has the role of analyzing language statements and representing them in a format that is optimized for further processing. It verifies if the text taken as input is a valid string of the language defined by the formal grammar presented in *Chapter 5*. If that is the case, it builds a data structure that describes all the clauses found in the input text; otherwise, an error is raised.

Many *compiler-compiler* software exist and they can be used to automatically build the parser. They take a grammar of the language as input (usually in *BNF* or *EBNF* form) and they produce the parser source code as output. Some pieces of code can usually be associated to each grammar rule and they will be executed when these rules are applied by the parser. In our project, these blocks are used to build the data structures that define the internal representation of queries.

Different parser generators can be classified on the basis of the programming language used to write the output code and on the basis of the grammar subset that is accepted as input. In fact, if only specific classes of grammars are allowed, high-performance algorithms can be used to parse the input string with a linear time complexity. The most common parser generators accept *LL*, *LR* or *LALR* grammars [33].

We decided to use the Java platform to implement the system components not directly interacting with the hardware. Among the Java based parser generators, we decided to use JavaCC (Java Compiler Compiler) that is an open source project able to generate top-down parsers. LL(k) is the accepted class of grammars, but semantic lookahead can be used to handle with other particular types of grammar [34].

As said before, the grammar explained in *Chapter 5* was written to be human readable and, then, it can contain some ambiguities. To use that grammar with JavaCC, some changes are needed. More specifically, all the ambiguities have to be removed as well as the left side recursions. The obtained grammar, written with the JavaCC syntax, is reported in Appendix A. Although the LL(k) class is supported, we tried to write a grammar that is LL(1) as much as possible. In fact, JavaCC allows to specify the lookahead to be adopted at each choice point. In this way, the obtained parser is mainly LL(1) with some LL(2) choice points. Note that, when the lookahead is not explicitly

specified, a default *lookahead* of one token is used by *JavaCC* and an automatic choice conflict control is provided.

Although in *Chapter 5* we have specified the whole language through an *EBNF* grammar, parser generators usually require to split the language definition in two blocks: tokens and grammar productions. Tokens are regular expressions and they specify the set of terminal symbols used in the productions. Thus, the software produced by the generator is composed of two components, called *lexical analyzer* and *syntactical analyzer*: the first one scans the input text and tokenizes it, while the second one checks the obtained tokens sequence to verify if it complies with grammar productions.

In Chapter 5 grammar, many non terminals symbols are introduced only to improve readability: they are all defined as an identifier or as two identifiers separated by a dot (e.g.: DataStructureName, DataFieldName, LogicalObjectField, PilotJoinField, etc.). In JavaCC grammar they have been substituted with only three non terminal symbols (Identifier, QualifiedIdentifier and OptionallyQualifiedIdentifier) to make easier the grammar rewriting w.r.t. LL(1) constrains. For clearness, whenever these three symbols are used, a comment is reported to specify the original non terminal they substitute.

There are few situations in which the specification of the *lookahead* clause was introduced. In the productions that allows the choice between a duration and a number of samples (or selections) a two tokens *lookahead* is needed. Consider for example the production that defines the *WindowSize* non terminal symbol:

```
void WindowSize() : {}
{ LOOKAHEAD(2) Duration() | SamplesNumber() }
```

If the window size is specified through a duration, a numeric constant (integer or float) followed by a unit of time is required. Otherwise, if the window size is defined in terms of a number of records, an integer constant followed by the *SAMPLE* keyword is required. If a float constant is found, it is certainly the beginning of a duration, but if an integer constant is found, decision cannot be taken until the next token is analyzed. So the choice point must be declared as *LOOKAHEAD* (2).

Another situation in which an explicit specification of *lookahead* is needed is the definition of the *Constant* symbol:

```
void Constant() : {}
{ ConstantNull() | ConstantBoolean() | ConstantString() |
  LOOKAHEAD(1) ConstantInteger() | ConstantFloat() }
```

To better understand this production, the explanation should start from the lexical analyzer. Two tokens are defined to recognize numeric values: CONSTANT_INTEGER and CONSTANT_FLOAT. The first one corresponds to integer numbers, while the second one corresponds only to the numeric values containing a dot. At the syntactical analyzer level, two productions are defined: ConstantInteger is exactly the CONSTANT_INTEGER token, while ConstantFloat is the union of CONSTANT_INTEGER and CONSTANT_FLOAT tokens. For these reasons, JavaCC signals a choice conflict in the production reported above. In fact, when a CONSTANT_INTEGER token is received from the scanner, both ConstantInteger and ConstantFloat productions can be matched. However, the first one is certainly the correct choice, without the needing of looking ahead. Since JavaCC chooses the firstly written production, specifying ConstantInteger before ConstantFloat (in the Constant production) is enough. To suppress the warning signaling, an explicit LOOKAHEAD (1) has been specified.

The last productions in which a numeric *lookahead* was used are relative to external function calls. The reasons that led us to using a *lookahead* of two tokens will be clearer in the following.

In *Chapter 5* the different kinds of allowed expressions were not described through the grammar, but rather with a table (*Figure 32*). In *JavaCC* these constraints have been specified directly in the grammar, using the *semantic lookahead* feature. It allows to specify any arbitrary Boolean expression, whose evaluation determines which option is taken at a choice point. We used this *JavaCC* feature as follows:

- We defined a Java enumeration (*ExpressionType*) to list the different kinds of expression.
- We added a parameter (*parExpressionType*) to the *Expression* production to indicate the type of the recognized expression.

- We propagated this information down to the production that defines the expression base blocks (*ExpressionValuePrimary*), trough parameter passing.
- We placed semantic lookahead clauses in the ExpressionValuePrimary production to define the set of allowed base blocks, depending on the value of parExpressionType.

Note that, in the current version of *JavaCC*, some constraints must be respected when *semantic lookahead* is used: in particular, if local variables (or parameters) are used in a *semantic lookahead* specification within the *EBNF* production for a certain non terminal, this non terminal mustn't be used in *syntactic lookahead*, or in a *lookahead* of more than 1 token; otherwise a compiler error will be present in the generated source code. This limitation forced us to split the definition of function calls in two cases: with and without parameters. Since these two alternatives start with the same token, a *lookahead* specification has been introduced.

A final consideration about the grammar reported in *Appendix A* is relative to a pair of features that are not present in the grammar of *Chapter 5*: the possibility of writing a list of statements separated by a ";" character and the possibility of introducing some comments in the query, using a *C* like notation.

Figure 35, Figure 36, Figure 37 and Figure 38 report a *UML* class diagram of the internal representation of queries, that is built by the parser.

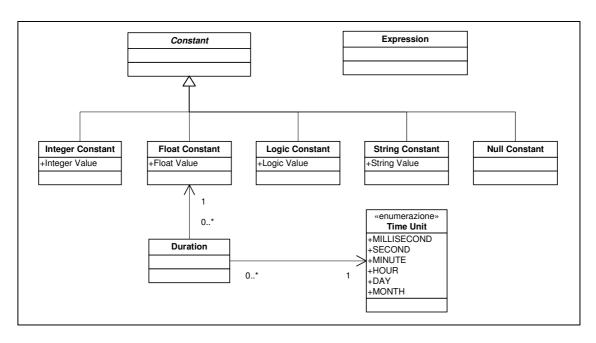


Figure 35: Internal representation of queries (1/4)

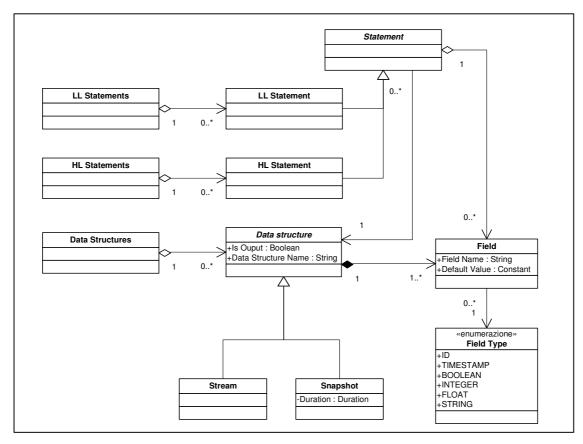


Figure 36: Internal representation of queries (2/4)

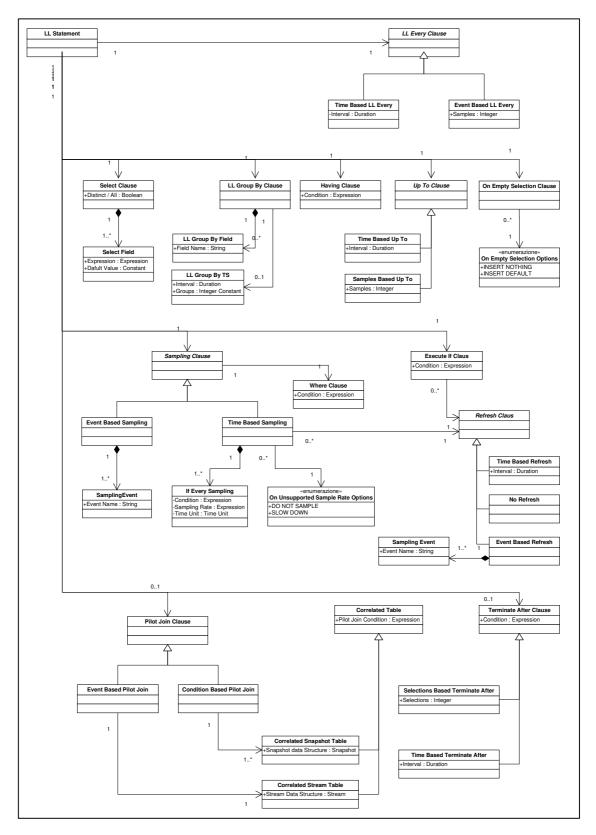


Figure 37: Internal representation of queries (3/4)

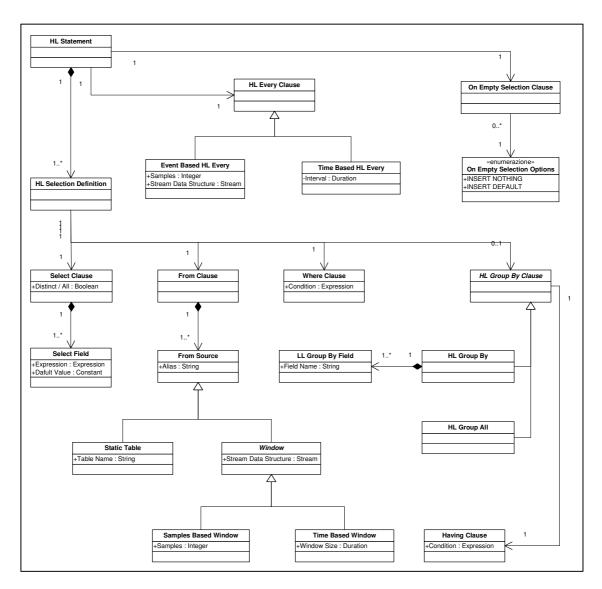


Figure 38: Internal representation of queries (4/4)

8 Conclusions

In this work we widely analyzed the main requirements a middleware should comply with in order to collect data from a pervasive system. We presented these features and discussed how they can be mapped on a fully declarative language. We also studied if and how similar projects handle the same issues. Then, as the result of this analysis phase, we defined a *SQL* like declarative language aiming at providing a friendly interface to the user, who wants to query a pervasive system without dealing with hardware details.

For each language clause we defined the syntax and explained the semantics, in order to allow the development of a prototype system. We also showed some query examples to better clarify the language semantics. These queries can be further used as preliminary tests of the prototype implementation.

It is important noticing that, although we tried to deal with every aspect of the language design and to define a clear semantics, the prototype development phase will help to discover possible mistakes and omissions and to find a solution for them.

In this chapter we briefly describe the state of art of the project, reporting the list of open issues and the possible extensions that are not handled in this thesis. We also present a concise comparison with the *GSN* project.

8.1 Concise comparison with GSN

The *Global Sensor Network Middleware* [15], designed at *EPLF of Lausanne*, manages the problem of querying a sensor network in a way similar to our approach. In that project two main concepts are introduced: *wrapper* and *virtual sensor*.

A wrapper is a component used to encapsulate the data received from data sources into the standard *GSN* data model. Two types of data sources are supported: event based and polling based. A virtual sensor is a specialized program that filters and processes data coming from wrappers or other virtual sensors.

A complex query is seen as a network of virtual sensors: this is not too different from our query graph idea. The *GSN* concept of wrapper is quite similar to our low level queries and that of virtual sensor is quite similar to our high level queries.

8. Conclusions

However, some differences between the two projects exist: we use only a *SQL* like declarative language while *GSN Middleware* uses *SQL* for data manipulation and *XML* for data structure definition. Moreover, *GSN* supports only event based activation of the query. The *pilot join* operation is the feature really improving our language potential with respect to *GSN*: it allows to control the sampling operation on some sensors depending on the results obtained from the manipulation of data coming from other sensors.

8.2 Open issues and future works

In this work, we explained how a query can be executed on the architecture for pervasive systems, however there are many issues that have still to be solved to allow a real integration between our language and the existing architecture. For example, some questions about the logical object registry behavior have to be answered:

- WHEN the registry instantiates logical objects?
- WHEN logical objects are destroyed?
- HOW new devices are detected by the system?

Probably, at the language level, a clause should be added to force new devices detection before the query initialization. A final decision about this question can be taken only when the registry component will be fully designed.

There is a set of useful language functionalities that are not implemented yet in the grammar presented in this work:

• Termination of a query. Although WSNs are often used to execute queries in a continuous way, we think that the language must however support a mechanism to stop a running query. At the moment we only introduced a clause in the low level language to allow specifying when an instance of a running low level query has to be stopped. This termination clause is related to a single logical object, but probably a clause should be added to require the termination of the whole user submitted query.

- Support for actuation queries. It could be interesting to design a set of clauses for allowing declarative queries to control the environment as well as monitoring it. It seems not so difficult extending our language to manage actuators and not only sensors. From an abstract, conceptual point of view, we think that sensors and actuators can be managed in a similar way.
- Extension of the EXECUTE IF clause. Some extensions to the EXECUTE IF clause can be explored to increase the language expressiveness. For example, an option to control the maximum number of logical objects that will take part in a query can be added.
- Definition of in-network data mining operations. Some data mining statements can be introduced in the language: they should be a set of clauses to require the execution of typical data mining primitives on intermediate streams, before they come to the node that submitted the query [35] [36]. The implementation of these statements is not strictly required to build a minimal query language for pervasive systems but it can be an additional useful service for the user.

Finally, an interesting extension is the definition of a procedural language that can be used to add new functionalities to the declarative one. More specifically, this language should provide the constructs to define new *aggregation operators* and new customized *grouping operators* (to be used in *GROUP BY* clauses). For example it can be useful to define a new grouping operator, able to split a set of records (containing a location field) depending on a squared grid.

Appendix A. Language EBNF

```
SEPARATORS
    Characters between tokens. They are not sent to the parser
*/
SKIP:{ " " | "\r" | "\t" | "\n" }
      COMMENTS
      C-style comments ( / * ... * / ): lexical states are used.
*/
MORE: { "/*" : COMMENT }
<COMMENT> MORE: { <~[]> }
<COMMENT> SKIP: { "*/" : DEFAULT }
/* Tokens of the *CREATE* clause */
TOKEN: {
     < KEYWORD CREATE: "CREATE" > | < KEYWORD OUTPUT: "OUTPUT" > |
     < KEYWORD_AS : "AS" >
/* Tokens of the *INSERT* clause */
TOKEN: { < KEYWORD_INSERT: "INSERT" > | < KEYWORD_INTO: "INTO" > }
/* Tokens of the *EVERY* clause */
TOKEN: {
     < KEYWORD_EVERY : "EVERY" > | < KEYWORD_IN: "IN" > |
     < KEYWORD_SYNCHRONIZED: "SYNCHRONIZED" >
/* Tokens of the *SELECT* clause */
TOKEN: {
     < KEYWORD_SELECT: "SELECT" > |
     < KEYWORD_DISTINCT: "DISTINCT" > |
     < KEYWORD_ALL : "ALL"
     < KEYWORD GROUPTS: "GROUP TS" >
```

```
/* Tokens of the *UNION* clause */
TOKEN: { < KEYWORD UNION: "UNION" > /* | < KEYWORD ALL: "ALL" > */ }
/* Tokens of the *WHERE* clause */
TOKEN: { < KEYWORD_WHERE: "WHERE" > }
/* Tokens of the *GROUP BY* clause */
TOKEN: {
     < KEYWORD_GROUP:
                         "GROUP" > | < KEYWORD_BY: "BY" >
     /* | < KEYWORD_ALL: "ALL" > */ |
     < KEYWORD_GROUPS:
                        "GROUPS" >
}
/* Tokens of the *HAVING* clause */
TOKEN: { < KEYWORD_HAVING: "HAVING" > }
/* Tokens of the *UP TO* clause */
TOKEN: { < KEYWORD UP:
                         "UP" >
                                      | < KEYWORD_TO: "TO" > }
/* Tokens of the *ON EMPTY SELECTION* clause */
TOKEN: {
     /* < KEYWORD ON: "ON" > | */
     < KEYWORD EMPTY: "EMPTY"> | < KEYWORD SELECTION: "SELECTION" > |
     /* < KEYWORD_INSERT: "INSERT" > | */
     < KEYWORD_NOTHING: "NOTHING" > | < KEYWORD_DEFAULT: "DEFAULT" >
/* Tokens of the *SAMPLING* clause */
TOKEN: {
     < KEYWORD_SAMPLING: "SAMPLING" > |
     /* < KEYWORD ON:
                         "ON"
                                   > | */
     < KEYWORD_EVENT:
                         "EVENT"
                                   > |
     /* < KEYWORD_IF:</pre>
                         "IF"
                                   > | */
     /* < KEYWORD_EVERY: "EVERY"
                                   > | */
     < KEYWORD ELSE:
                         "ELSE"
}
```

```
/* Tokens of the *ON UNSUPPORTED SAMPLING RATE* clause */
TOKEN: {
                          "ON"
     /* < KEYWORD ON:
                                       > | */
     < KEYWORD UNSUPPORTED: "UNSUPPORTED" > |
     < KEYWORD_SAMPLE: "SAMPLE" > | < KEYWORD_RATE: "RATE" > |
     < KEYWORD DO:
                     "DO"
                              > |
     /* < KEYWORD_NOT: "NOT" > | */
     < KEYWORD_SLOW: "SLOW" > | < KEYWORD_DOWN: "DOWN" >
/* Tokens of the *REFRESH* clause */
TOKEN: {
     < KEYWORD_REFRESH: "REFRESH" > |
     /* < KEYWORD_EVERY: "EVERY" > | */
     < KEYWORD_NEVER: "NEVER" >
}
/* Tokens of the *PILOT JOIN* clause */
TOKEN: {
     < KEYWORD_PILOT: "PILOT" > | < KEYWORD_JOIN: "JOIN" > |
     < KEYWORD_ON : "ON" >
}
/* Tokens of the *EXECUTE IF* clause */
TOKEN: {
     < KEYWORD_EXECUTE: "EXECUTE" > | < KEYWORD_IF: "IF" > |
     < KEYWORD_EXISTS : "EXISTS" >
/* Tokens of the *TERMINATE AFTER* clause */
TOKEN: {
     < KEYWORD_TERMINATE: "TERMINATE" > | < KEYWORD_AFTER: "AFTER" > |
     < KEYWORD_SELECTIONS: "SELECTIONS" >
}
/* Tokens of the *FROM* clause */
TOKEN: { < KEYWORD_FROM: "FROM" > /* | < KEYWORD_AS: "AS" > */ }
```

```
/* Tokens for DATA STRUCTURES */
TOKEN: {
     < KEYWORD_STREAM: "STREAM" > | < KEYWORD_SNAPSHOT: "SNAPSHOT" >
}
/* Others tokens */
TOKEN: {
     /* Token used to introduced the low level query */
     < KEYWORD_LOW:
                         "LOW"
                                    > |
     /* Token used to introduced the high level query */
     < KEYWORD_HIGH:
                         "HIGH"
                                    > |
     /* Token used, within DURATION, to introduce a TIME QUANTUM */
     < KEYWORD_WITH:
                         "WITH"
                                    > |
     /* Token used, within WITH, to introduce a TIME QUANTUM */
     < KEYWORD_DURATION: "DURATION" > |
     /* Token used to indicate sample number*/
     < KEYWORD_SAMPLES: "SAMPLES" > |
     /* Token used to indicate sample number equal to ONE*/
     < KEYWORD_ONE:
                         "ONE"
                                  > |
     /* Token used in Boolean expressions (in order to verify their
       * value: TRUE, FALSE or UNKNOWN) or to compare with a NULL value
      < KEYWORD IS : "IS" > | < KEYWORD BETWEEN: "BETWEEN" > |
     < KEYWORD_LIKE: "LIKE" >
/* ALGEBRAIC OPERATORS */
TOKEN: {
     < OPERATOR_MULTIPLY: "*" > | < OPERATOR_DIVIDE: "/" > |
     < OPERATOR_PLUS : "+" > | < OPERATOR_MINUS : "-" >
}
/* LOGIC OPERATORS */
TOKEN: {
     < OPERATOR_NOT: "NOT" > | < OPERATOR_XOR:
                                                    "XOR" > |
     < OPERATOR_AND: "AND" > | < OPERATOR_OR :</pre>
                                                      "OR" >
```

```
/* BITWISE OPERATORS */
TOKEN: {
     < OPERATOR_BITWISE_NOT: "!" > | < OPERATOR_BITWISE_XOR: "^" > |
     < OPERATOR BITWISE AND: "&" > | < OPERATOR BITWISE OR: "|" > |
     < OPERATOR_BITWISE_LSH: "<<"> | < OPERATOR_BITWISE_RSH: ">>" >
/* COMPARISON OPERATORS */
TOKEN: {
     < OPERATOR_GREATER: ">" > | < OPERATOR_LESS: "<" > |
     < OPERATOR_GREATER_EQUAL: ">=" > |
     < OPERATOR_LESS_EQUAL:
                             "<=" > |
     < OPERATOR_EQUAL:
                             "=" > |
     < OPERATOR_NOT_EQUAL: "<>" | !=" >
/* AGGREGATION FUNCTIONS */
TOKEN: {
     < FUNCTION_COUNT: "COUNT" > | < FUNCTION_AVG: "AVG" > |
     < FUNCTION_MAX : "MAX" > | < FUNCTION_MIN: "MIN" > |
     < FUNCTION SUM : "SUM" >
     /* | < FUNCTION_TIMESTAMP: "TIMESTAMP" > */
}
/* TIME UNITS */
TOKEN: {
     < TIMEUNIT_S: "seconds" | "s" > |
     < TIMEUNIT_M: "minutes"
                                 | "m" > |
     < TIMEUNIT_H: "hours"
                                 | "h" > |
     < TIMEUNIT_MS: "milliseconds" | "ms" > |
     < TIMEUNIT_D: "days"
                                 | "d" > |
     < TIMEUNIT_MT: "months"
                                | "mt" >
```

```
/* DATA TYPES */
TOKEN: {
     < TYPE_ID : "ID" > | < TYPE_TS : "TIMESTAMP" > |
     < TYPE_BOOLEAN: "BOOLEAN" > | < TYPE_INTEGER: "INTEGER" > |
     < TYPE_FLOAT : "FLOAT" > | < TYPE_STRING : "STRING"
/* CONSTANTS
* Note: DO NOT directly use in the grammar the tokens introduced
* in this section. Correspondent productions must be used.
*/
/* NULL constant */
          < CONSTANT_NULL: "NULL" > }
TOKEN: {
/* BOOLEAN constants:
* They define both the Boolean set of values (TRUE and FALSE)
* and the three valued logic set of values (TRUE, FALSE and UNKNOWN)
*/
TOKEN: {
     < CONSTANT_BOOLEAN_TRUE:
                                "TRUE"
     < CONSTANT BOOLEAN FALSE: "FALSE" > |
     < CONSTANT_BOOLEAN_UNKNOWN: "UNKNOWN" >
/* NUMERIC constants:
 They define integer and float values.
 Note: CONSTANT_FLOAT DO NOT corresponds to all numeric values,
  but only to the ones containing a dot ('.').
  e.g. '3.5' will be interpreted as a CONSTANT_FLOAT,
       while '3' will be interpreted as a CONSTANT_INTEGER.
*/
```

```
TOKEN: {
      <CONSTANT_INTEGER: ( <DIGIT> )+ > |
      <CONSTANT FLOAT:</pre>
         ( <CONSTANT_INTEGER> )? "." <CONSTANT_INTEGER>
         ( "E" ( <OPERATOR_PLUS> | <OPERATOR_MINUS> )?
         <CONSTANT_INTEGER> )?
         <CONSTANT_INTEGER> "E"
        ( <OPERATOR_PLUS> | <OPERATOR_MINUS> )? <CONSTANT_INTEGER>
      ) >
}
     STRING constants:
     They define the String type.
     - A String can be bounded both with single quotes
        (SINGLE_QUOTED_STRING) and with double quotes
        (DOUBLE_QUOTED_STRING) .
     - Quote characters contained in a String must be duplicated if
       the same type of quote is used to bound the string
       (e.g.: 'don''t').
     - Strings are defined using lexical states, in order to allow
       an easily detection of non properly closed strings.
*/
TOKEN: {
      < CONSTANT_SINGLE_QUOTED_STRING_START: "'" > :
                                               NON_SINGLE_QUOTED_STRING
       < CONSTANT_DOUBLE_QUOTED_STRING_START: "\"" > :
                                              NON_DOUBLE_QUOTED_STRING
<NON_SINGLE_QUOTED_STRING> TOKEN:{
<CONSTANT_SINGLE_QUOTED_STRING_VALUE: (~["'"] | "'" "'")* > :
                                     NON_SINGLE_QUOTED_STRING_END
}
```

```
<NON_DOUBLE_QUOTED_STRING> TOKEN:{
      < CONSTANT_DOUBLE_QUOTED_STRING_VALUE: (~["\""] | "\"" "\"")* >:
                                        NON DOUBLE QUOTED STRING END
<NON_SINGLE_QUOTED_STRING_END> TOKEN: {
      < CONSTANT_SINGLE_QUOTED_STRING_END: "'" > : DEFAULT
<NON_DOUBLE_QUOTED_STRING_END> TOKEN:{
      < CONSTANT_DOUBLE_QUOTED_STRING_END: "\"" > : DEFAULT
/* IDENTIFIERS */
TOKEN: {
                   ["0" - "9"]
      < #DIGIT:
      < #LITERAL:</pre>
                   ["a" - "z"] | ["A" - "Z"] > |
      < #UNDERSCORE: "_"</pre>
      < IDENTIFIER: ( <LITERAL> | <UNDERSCORE>)
                    ( <DIGIT> | <LITERAL> | <UNDERSCORE> ) * >
GENERIC PRODUCTIONS (Signs, Constants, Field names, Table names, etc.)
void StatementSequence():{}{ Statement()(";" Statement())* <EOF> }
void Sign():{}{ <OPERATOR_PLUS> | <OPERATOR_MINUS> }
void LogicValue():{}{
      <CONSTANT_BOOLEAN_TRUE> | <CONSTANT_BOOLEAN_FALSE> |
      <CONSTANT_BOOLEAN_UNKNOWN>
void ConstantNull():{}{ <CONSTANT_NULL> }
void ConstantBoolean():{} {
      <CONSTANT_BOOLEAN_TRUE> | <CONSTANT_BOOLEAN_FALSE>
```

```
void ConstantFloat():{}{ ( <CONSTANT_INTEGER> | <CONSTANT_FLOAT> ) }
void ConstantInteger():{}{ <CONSTANT_INTEGER> }
void ConstantString():{}{
           <CONSTANT_SINGLE_QUOTED_STRING_START>
            <CONSTANT_SINGLE_QUOTED_STRING_VALUE>
            <CONSTANT_SINGLE_QUOTED_STRING_END>
                                                     )
       Ι
       (
           <CONSTANT_DOUBLE_QUOTED_STRING_START>
            <CONSTANT_DOUBLE_QUOTED_STRING_VALUE>
            <CONSTANT_DOUBLE_QUOTED_STRING_END>
                                                     )
void Constant():{}{
       ConstantNull() | ConstantBoolean() | ConstantString() |
       LOOKAHEAD(1) ConstantInteger() |
       ConstantFloat()
}
void SignedConstant():{} { [ Sign() ] Constant() }
void ComparisonOperator(): {}{
       <OPERATOR_GREATER> | <OPERATOR_LESS>| <OPERATOR_GREATER_EQUAL> |
       <OPERATOR LESS EQUAL> | <OPERATOR EQUAL> | <OPERATOR NOT EQUAL>
void FieldType():{}{
       <TYPE_ID> | <TYPE_TS> | <TYPE_BOOLEAN> | <TYPE_INTEGER> |
       <TYPE_FLOAT> | <TYPE_STRING>
}
void TimeUnit():{}{
       <TIMEUNIT_S> | <TIMEUNIT_M> | <TIMEUNIT_H> | <TIMEUNIT_MS> |
       <TIMEUNIT_D> | <TIMEUNIT_MT>
}
```

```
void AggregationOperator():{}{
       <FUNCTION_AVG> | <FUNCTION_MIN> |
       <FUNCTION_MAX> | <FUNCTION_SUM>
void Duration():{}{ ConstantFloat() TimeUnit() }
void SamplesNumber():{}{
       ( ConstantInteger() <KEYWORD_SAMPLES> ) | ( <KEYWORD_ONE> )
void SelectionsNumber():{}{( ConstantInteger() <KEYWORD_SELECTIONS> ) }
void WindowSize():{}{ LOOKAHEAD(2) Duration() | SamplesNumber() }
void Identifier():{}{ <IDENTIFIER> }
void QualifiedIdentifier():{}{<IDENTIFIER> "." <IDENTIFIER> }
void OptionallyQualifiedIdentifier():{}{ <IDENTIFIER>["." IDENTIFIER>]}
void Aggregate(ExpressionType parExpressionType):{}{
    (
       ( <FUNCTION_COUNT> "(" "*" ) |
            ( AggregationOperator() "(" Expression(parExpressionType) )
      LOOKAHEAD( {parExpressionType == ExpressionType.LOW_LEVEL_ALL} )
       ( ", " WindowSize() [ ", " Expression(parExpressionType) ] ) |
       LOOKAHEAD( {parExpressionType == ExpressionType.HIGH_LEVEL_ALL})
       ( [ ", " Expression(parExpressionType) ] )
    ) ")"
```

```
void FunctionCall(ExpressionType parExpressionType):{}{
       <IDENTIFIER> (
      LOOKAHEAD (2) ( "(" ") " )
              "(" Expression(parExpressionType)
            ( ", " Expression(parExpressionType) )* ")" )
void ExistsAttribute():{}{
  <KEYWORD_EXISTS>
   "(" ( Identifier() /* LogicalObjectField */ | <KEYWORD_ALL> ) ")"
EXPRESSIONS PRODUCTIONS
**************************************
void Expression(ExpressionType parExpressionType):{}{
      ExpressionBooleanTerm(parExpressionType)
       ( <OPERATOR_OR> ExpressionBooleanTerm(parExpressionType) )*
void ExpressionBooleanTerm(ExpressionType parExpressionType):{}{
       ExpressionBooleanFactor(parExpressionType)
       ( <OPERATOR_AND> ExpressionBooleanFactor(parExpressionType) )*
void ExpressionBooleanFactor(ExpressionType parExpressionType):{}{
      ExpressionBooleanTest (parExpressionType)
       ( <OPERATOR_XOR> ExpressionBooleanTest(parExpressionType) )*
}
void ExpressionBooleanTest(ExpressionType parExpressionType):{}{
       ( <OPERATOR_NOT> ) * ExpressionBooleanPrimary(parExpressionType)
void LogicTest():{}{[ <KEYWORD_IS> [ <OPERATOR_NOT> ] LogicValue() ] }
```

```
void ExpressionBooleanPrimary(ExpressionType parExpressionType):{}{
     ExpressionBit (parExpressionType)
     (ComparisonOperator()ExpressionBit(parExpressionType)LogicTest())
     ( <KEYWORD_IS> [ <OPERATOR_NOT>]
     ( <CONSTANT_NULL> LogicTest() | LogicValue() ) )
     ( <KEYWORD_BETWEEN> ExpressionBit(parExpressionType) < OPERATOR_AND>
       ExpressionBit (parExpressionType) LogicTest() )
     ( <KEYWORD_LIKE> ConstantString() LogicTest())
     1
}
void ExpressionBit(ExpressionType parExpressionType):{}{
       ExpressionBitTerm(parExpressionType)
       ( <OPERATOR_BITWISE_OR> ExpressionBitTerm(parExpressionType) )*
}
void ExpressionBitTerm(ExpressionType parExpressionType):{}{
       ExpressionBitFactor(parExpressionType)
       (<OPERATOR_BITWISE_AND> ExpressionBitFactor(parExpressionType))*
void ExpressionBitFactor(ExpressionType parExpressionType):{}{
       ExpressionBitShift (parExpressionType)
       (<OPERATOR_BITWISE_XOR> ExpressionBitShift(parExpressionType) )*
void ExpressionBitShift(ExpressionType parExpressionType):{}{
 ExpressionBitTest (parExpressionType)
  [(<OPERATOR_BITWISE_LSH> | <OPERATOR_BITWISE_RSH>) ConstantInteger()]
```

```
void ExpressionBitTest(ExpressionType parExpressionType):{}{
       ( <OPERATOR_BITWISE_NOT> ) * ExpressionValue(parExpressionType)
}
void ExpressionValue(ExpressionType parExpressionType):{}{
       ExpressionValueTerm(parExpressionType)
       ( ( <OPERATOR_PLUS> | <OPERATOR_MINUS> )
            ExpressionValueTerm(parExpressionType) )*
void ExpressionValueTerm(ExpressionType parExpressionType):{}{
       ExpressionValueFactor(parExpressionType)
       ( ( <OPERATOR_MULTIPLY> | <OPERATOR_DIVIDE>
            ExpressionValueFactor(parExpressionType) )*
void ExpressionValueFactor(ExpressionType parExpressionType):{}{
       [ Sign() ] ExpressionValuePrimary(parExpressionType)
}
void ExpressionValuePrimary(ExpressionType parExpressionType):{}{
       Constant() | ExistsAttribute()
       LOOKAHEAD(2) FunctionCall(parExpressionType) |
       ( "(" Expression(parExpressionType) ")" )
       LOOKAHEAD({parExpressionType == ExpressionType.LOW_LEVEL_ALL } )
       ( <TYPE_ID> | <KEYWORD_GROUPTS>
         Aggregate(parExpressionType)
         OptionallyQualifiedIdentifier() /* PilotJoinField,
                                                  LogicalObjectField */
       )
       LOOKAHEAD({parExpressionType == ExpressionType.HIGH_LEVEL_ALL})
       ( Aggregate (parExpressionType)
            OptionallyQualifiedIdentifier() /* WindowField */ )
       LOOKAHEAD({parExpressionType == pressionType.LOW_LEVEL_NO_AGGR})
```

```
<TYPE_ID> |
          OptionallyQualifiedIdentifier()
          /* PilotJoinField, LogicalObjectField */ )
      LOOKAHEAD (
      {parExpressionType == ExpressionType.LOW_LEVEL_NO_AGGR_NO_PILOT}
      ( <TYPE_ID> | Identifier() /* LogicalObjectField */ )
     ELEMENTS LIST PRODUCTIONS
void FieldDefinitionList():{}{
      "(" FieldDefinition() ( ", " FieldDefinition() )* ")"
void FieldDefinition():{}{
      Identifier() /* DataStructureField */
      FieldType() [ <KEYWORD_DEFAULT> SignedConstant() ]
void FieldList():{}{
        "(" Identifier() /* DataStructureField */
         ( "," Identifier() /* DataStructureField */ )* ")"
void EventList():{}{
      Identifier() /* LogicalObjectEvent */
      ( "," Identifier() /* LogicalObjectEvent */ )*
}
void CorrelatedTableList():{}{CorrelatedTable()("," CorrelatedTable()*}
```

```
void CorrelatedTable():{}{
      Identifier() /* DataStructureName */
      <KEYWORD_ON> Expression(ExpressionType.LOW_LEVEL_NO_AGGR)
}
void WindowDefinitionList():{}{
      WindowDefinition() ( "," WindowDefinition() )*
void WindowDefinition():{}{
      Identifier() /* DataStructureName */
      [ "(" WindowSize() ")" ]
      [ <KEYWORD_AS> Identifier() /* DataStructureName */ ]
void FieldGroupingByList():{}{
 ( FieldGroupingBy() | FieldGroupingByTs() )( ", " FieldGroupingBy() )*
void FieldGroupingBy():{}{ Identifier() /* LogicalObjectField */ }
void FieldGroupingByTs():{}{
  <TYPE_TS> "(" Duration() "," ConstantInteger() <KEYWORD_GROUPS> ")"
}
void WindowFieldList():{}{
      OptionallyQualifiedIdentifier() /* WindowField */
      ( ", " OptionallyQualifiedIdentifier() /* WindowField */ )*
STATEMENTS PRODUCTIONS
void Statement():{}{CreationStatement() | InsertionStatement() }
```

```
void CreationStatement():{}{
       <KEYWORD_CREATE> [ <KEYWORD_OUTPUT> ]
       ( StreamCreationStatement() | SnapshotCreationStatement() )
}
void StreamCreationStatement():{}{
       CreateStreamClause() [ <KEYWORD_AS> StreamSelectionStatement() ]
void SnapshotCreationStatement():{}{
      CreateSnapshotClause()[<KEYWORD_AS> SnapshotSelectionStatement()]
void CreateStreamClause():{}{
       <KEYWORD_STREAM> Identifier() /* DataStructureName */
       FieldDefinitionList()
void CreateSnapshotClause():{}{
      <KEYWORD_SNAPSHOT> Identifier() /* DataStructureName */
      FieldDefinitionList() < KEYWORD_WITH> < KEYWORD_DURATION> Duration()
void InsertionStatement():{}{
      <KEYWORD_INSERT> <KEYWORD_INTO>
      ( StreamInsertionStatement() | SnapshotInsertionStatement() )
void StreamInsertionStatement():{}{
       InsertStreamClause() StreamSelectionStatement()
void SnapshotInsertionStatement():{}{
       InsertSnapshotClause() SnapshotSelectionStatement()
```

```
void InsertStreamClause():{}{
   <KEYWORD_STREAM> Identifier() /* DataStructureName */ [FieldList()]
}
void InsertSnapshotClause():{}{
   <KEYWORD_SNAPSHOT> Identifier() /* DataStructureName */[FieldList()]
void StreamSelectionStatement():{}{
       ( <KEYWORD_LOW> ":" LowEveryClause()
         LowSelectionStatement(ExpressionType.LOW_LEVEL_ALL) )
       ( <KEYWORD_HIGH> ":" HighEveryClause()
         HighSelectionStatement(ExpressionType.HIGH_LEVEL_ALL) )
void SnapshotSelectionStatement():{}{
       ( <KEYWORD_LOW> ":"
            LowSelectionStatement(ExpressionType.LOW_LEVEL_ALL) )
       ( <KEYWORD HIGH> ":"
            HighSelectionStatement (ExpressionType.HIGH_LEVEL_ALL) )
}
void LowEveryClause():{}{
       <KEYWORD_EVERY> (
       LOOKAHEAD (2) (
      Duration() [ <KEYWORD_SYNCHRONIZED> ] ) | ( SamplesNumber() ) )
void HighEveryClause():{}{
    <KEYWORD_EVERY> (
      LOOKAHEAD(2) ( Duration() ) |
      (SamplesNumber() <KEYWORD_IN> Identifier()/*DataStructureName*/))
}
```

```
void LowSelectionStatement(ExpressionType parExpressionType):{}{
       SelectClause(parExpressionType)
       [ GroupByClause()
                                          1
       [ UpToClause()
       [ HavingClause(parExpressionType) ]
       [ OnEmptySelectionClause()
       SamplingClause()
       [ PilotJoinClause()
                                         1
       [ ExecuteIfClause()
                                         ]
       [ TerminateAfterClause()
                                         1
void HighSelectionStatement(ExpressionType parExpressionType):{}{
       HighSelectionStatementDefinition(parExpressionType)
            <KEYWORD_UNION> [ <KEYWORD_ALL> ]
            HighSelectionStatementDefinition(parExpressionType) )*
       [ OnEmptySelectionClause() ]
void HighSelectionStatementDefinition(ExpressionType parExpressionType)
:{}{
       SelectClause(parExpressionType) FromClause()
       [ WhereClause(parExpressionType)
       [ GroupClause() [ HavingClause(parExpressionType) ] ]
void SelectClause(ExpressionType parExpressionType):{}{
       <KEYWORD_SELECT> [ <KEYWORD_DISTINCT> | <KEYWORD_ALL> ]
       FieldSelectionList(parExpressionType)
void FieldSelectionList(ExpressionType parExpressionType):{}{
       FieldSelection(parExpressionType)
       ( ", " FieldSelection(parExpressionType) )*
}
```

```
void FieldSelection(ExpressionType parExpressionType):{}{
   Expression(parExpressionType) [ <KEYWORD_DEFAULT> SignedConstant() ]
}
void WhereClause(ExpressionType parExpressionType):{}{
       <KEYWORD_WHERE> Expression(parExpressionType)
void GroupByClause():{}{
       <KEYWORD_GROUP> <KEYWORD_BY> FieldGroupingByList()
void GroupClause():{}{
    <KEYWORD_GROUP>((<KEYWORD_BY> WindowFieldList())|( <KEYWORD_ALL> ))
void HavingClause(ExpressionType parExpressionType):{}{
       <KEYWORD_HAVING> Expression(parExpressionType)
}
void UpToClause():{}{ <KEYWORD_UP> <KEYWORD_TO> WindowSize() }
void OnEmptySelectionClause():{}{
      <KEYWORD ON> <KEYWORD EMPTY> <KEYWORD SELECTION> <KEYWORD INSERT>
      ( <KEYWORD_NOTHING> | <KEYWORD_DEFAULT> )
void SamplingClause():{}{
    <KEYWORD_SAMPLING>
       OnEventClause()
    1
      SamplingIfEveryClause() [OnUnsupportedSRClause()] [RefreshClause()]
      [ WhereClause (ExpressionType.LOW_LEVEL_NO_AGGR) ]
}
```

```
void OnEventClause():{}{ <KEYWORD_ON> <KEYWORD_EVENT> EventList() }
void SamplingIfEveryClause():{}{
       (
            ( SamplingIfClause() SamplingEveryClause() )+
            [ <KEYWORD_ELSE> SamplingEveryClause() ]
       ( SamplingEveryClause() )
void SamplingIfClause():{}{
       <KEYWORD_IF> Expression(ExpressionType.LOW_LEVEL_NO_AGGR)
void SamplingEveryClause():{}{
       <KEYWORD_EVERY>
       Expression(ExpressionType.LOW_LEVEL_NO_AGGR) TimeUnit()
void OnUnsupportedSRClause():{}{
     <KEYWORD_ON> <KEYWORD_UNSUPPORTED> <KEYWORD_SAMPLE> <KEYWORD_RATE>
     ( <KEYWORD DO> <OPERATOR NOT> <KEYWORD SAMPLE> )
     ( <KEYWORD_SLOW> <KEYWORD_DOWN> )
void RefreshClause():{}{
   <KEYWORD_REFRESH> (
     OnEventClause() | ( <KEYWORD_EVERY> Duration() ) | ( <KEYWORD_NEVER> )
   )
void PilotJoinClause():{}{
       <KEYWORD_PILOT> <KEYWORD_JOIN> CorrelatedTableList()
}
```

```
void ExecuteIfClause():{}{
       <KEYWORD_EXECUTE> <KEYWORD_IF>
       {\tt Expression} \, ({\tt ExpressionType.LOW\_LEVEL\_NO\_AGGR\_NO\_PILOT})
        [ RefreshClause() ]
void TerminateAfterClause():{}{
       <KEYWORD_TERMINATE> <KEYWORD_AFTER>
        ( LOOKAHEAD(2) Duration() | SelectionsNumber() )
void FromClause():{}{
       <KEYWORD_FROM> WindowDefinitionList()
}
```

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