

# ST4SQL: A Spatio-Temporal Query Language Dealing with Granularities

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**ABSTRACT:** *In many different application fields the amount and importance of spatio-temporal data (i.e., temporally and/or spatially qualified data) is increasing in last years and users need new solutions for their management. In [4] a framework for modeling spatio-temporal data in relational databases has been proposed. The model is based on the notions of temporal, spatial, and spatio-temporal granularities. Once data have been modeled, they must be queried through query languages that are also able to exploit their spatio-temporal components. In this paper we propose a spatio-temporal query language, called **ST4SQL**. The proposed language extends the well-known **SQL** syntax and the **T4SQL** temporal query language [8]. The proposed query language deals with different temporal, spatial and spatio-temporal semantics. These semantics allow one to specify how the system must manage temporal and spatial dimensions for evaluating the queries. Moreover, the query language introduces new constructs for grouping data with respect to temporal and spatial dimensions. Both semantics and grouping constructs take into account and exploit data qualified with granularities.*

**Keywords:** Query language, Spatio-temporal semantics, Relational databases, Spatio-temporal data

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## 1. Introduction

Spatio-temporal databases are those databases providing support for representing and managing temporal, spatial and spatio-temporal data. They implement capabilities of both temporal and spatial databases and, further, they provide new constructs for representing spatio-temporal data, i.e., data with both a temporal and a spatial qualification.

Research on spatio-temporal databases began in the late 1990s with two different projects [16, 26]. Thenceforth, several papers [4, 5, 6, 15, 31] proposed to manage spatio-temporal data in databases by qualifying them with spatio-temporal granularities. In general, granularities represent partitions of a temporal and/or spatial domain in disjoint intervals or regions (called granules), e.g., municipalities, pollution areas. Spatio-temporal data can be temporally and/or spatially qualified associating them to the granule representing their temporal or spatial location. Granularities represent an abstraction of a domain where intervals and regions substitute points. By using granularities, temporal and spatial data can be analyzed with respect to this abstraction instead of points. In [4] authors proposed the definitions of spatial and spatio-temporal granularities and, based on them and on the notion of temporal granularity [5], they proposed a relational database for granularities. The database can be used to store information about temporal, spatial and spatio-temporal granularities. Authors proposed to use the database

for granularities to extend existing spatio-temporal databases in order to qualify spatio-temporal data with granularities.

Once spatio-temporal data have been modeled in relational databases, they should be queried. For this purpose we need a query language able to manage spatio-temporal data. *SQL* [12] includes the definition of temporal datatypes (e.g., date, timestamp, interval) and functions. On the other hand, the *ISO/IEC 13249-3 SQL/MM Part 3* [13] standard includes the definition of spatial datatypes for representing 2- and 3-dimensional geometries (e.g., points and regions) and functions for their management. However, in both cases, temporal and spatial information is considered as other classical data. The two standards do not provide to users neither any facility for simplifying the formulation of spatio-temporal queries nor constructs for the automatic management of temporal and spatial data and dimensions. A different approach has been discussed in several proposals about temporal query languages [8, 27]. In these proposals, temporal dimensions are in some sense considered “*metadata*” enriching classical information. In this way, particular meaning may be associated to temporal dimensions and this meaning can be used to better exploit temporal qualification of data. In particular, *T4SQL* [8] includes also the definition of four temporal semantics (i.e., atemporal, sequenced, current, and next) that can be specified by the user for the automatic management of temporal dimensions during query evaluation.

Based on temporal and spatial capabilities, some spatio-temporal query languages have been proposed [9, 10, 26, 30]. In general, these query languages define spatio-temporal datatypes and a set of functions on them. Thus, they lack of ad-hoc semantics for spatio-temporal dimensions that may help users to easily manage spatio-temporal data.

In this paper, we propose an extension of *T4SQL*, called *ST4SQL*, for supporting also spatial and spatio-temporal dimensions. *ST4SQL* introduces new constructs for querying data exploiting their spatio-temporal components. *ST4SQL* adds to *T4SQL* support to data qualified with granularities [4] and it introduces support for spatial and spatio-temporal dimensions. In particular, we extend the four temporal semantics proposed in *T4SQL* also to consider data qualified with temporal granularities. Moreover, we define four spatial semantics for dealing with data qualified with spatial granularities and we define spatio-temporal semantics. Semantics allow the user to specify how the system has to (automatically) manage temporal and spatial dimensions for the evaluation of the specified query. We propose also new constructs for grouping data (and applying aggregate functions) with respect to their spatial qualification based on granularities. *ST4SQL* is intended to ease the formulation of spatio-temporal queries.

We introduce the syntax of *ST4SQL* and we explain the usage of new constructs exemplifying them with queries on a real clinical database. Moreover, we briefly present its semantics showing how new proposed constructs may be translated into equivalent *SQL* queries.

The paper is organized as follow. In the next section we discuss main related work about temporal, spatial and spatio-temporal query languages and we introduce some background notions about granularities and the database model for representing them. In Section 3 we introduce the clinical spatio-temporal database we then use as an example. In Section 4 we describe the *ST4SQL* query language, exemplifying it with several queries. In Section 5 we conclude with final remarks and future work.

## 2. Background and Related Work

In this section we present main proposals in literature dealing with temporal, spatial, and spatio-temporal query languages for relational databases.

### 2.1 Temporal query languages

Several temporal query languages have been proposed [8, 24, 27] to overcome the limitations of *SQL* with respect to temporal databases.

In 1995, Snodgrass et al. proposed *TSQL2* [27], a temporal extension to the *SQL-92* standard query language. Despite many researches proved the usefulness of *TSQL2*, the project for incorporating some *TSQL2* capabilities into the *ISO SQL* standard has been canceled in 2001. However, some similar temporal features have been implemented in Oracle Database [19] and *IBM DB2* [11].

The draft proposed for adding temporal support in *SQL* standard, called *SQL/Temporal*, includes the support for two temporal dimensions and two semantics for temporal queries. Supported temporal dimensions are valid time, i.e., the time instants or intervals when an information is true in the modeled reality, and transaction time, i.e., the time interval during

which data are current and can be retrieved in the database [14]. Temporal semantics allow a user to specify how the DBMS has to (automatically) manage temporal information during the evaluation of a query. According to the sequenced semantics the query engine evaluates the query for each time instant in the time domain selecting only those tuples whose value for the considered temporal dimension includes the considered instant. Conversely, in the non-sequenced semantics the query is evaluated considering all tuples in the relations without regard to their value for the given temporal dimension.

In [8] Combi et al. proposed the *T4SQL* temporal query language. It extends *SQL/Temporal* adding support also for availability time (i.e., the time when the database system or user become aware of a fact), and event time (i.e., the time when a decision has been taken or an event happened determining the considered fact) [14]. Moreover, besides the mentioned sequenced and non-sequenced (called atemporal) semantics, *T4SQL* introduces other two semantics: (1) current (the DBMS evaluates the query only on those tuples where the value for the given temporal Dimension is equal or contains the current date) and (2) next (the DBMS considers only pairs of tuples related to the same entity and that are consecutive with respect to the temporal ordering).

Since in the following we will extend *T4SQL* adding spatial constructs, we remember here the main part of the syntax of *T4SQL*:

```
[SEMANTICS <sem> ON <dim> [TIMESLICE <ts_exp>]
      {,<sem> ON <dim> [TIMESLICE <ts_exp>]}]
SELECT <sel_element_list>
      [,TGROUING (<t_attr> [AS] <new_name>
                {,<t_attr> [AS] <new_name>})]
FROM <tables>
WHERE <conditions>
WHEN <t_conditions>
GROUP BY <group_element_list>
HAVING <g_cond>
```

where *<sem>* is the name of a semantics, *<dim>* is the name of a temporal dimension, *<ts\_exp>* is a temporal expression resulting in a time point or an interval, *<t\_attr>* is a temporal attribute, and *<t\_conditions>* is a set of selection conditions involving only temporal attributes.

For example, the following query retrieves, for each hospital patient, the symptoms whose valid time overlaps the valid time of the prescribed therapy according to the current state of the database.

```
SEMANTICS SEQUENCED ON VALID,
          CURRENT ON TRANSACTION
SELECT Symptom, PatId
FROM PatSymptom AS ps, PatTherapy AS pt
WHERE ps.PatId = pt.PatId
```

Both previous languages include support for temporal join and temporal grouping. The first one is an extension of join operator in which join conditions include one or more temporal dimensions. The latter allows to group tuples according to their value on a temporal dimension. On grouped tuples (temporal) aggregate functions can be applied.

### 2.2 Spatial query languages

Considering spatial databases, currently, many commercial DBMSs that implement spatial functionalities, offer spatial datatypes, relationships, and operations following the approach standardized in ISO/IEC 13249-3 SQL/MM Part 3 [13]. SQL/MM is a standard extending SQL with multimedia and application-specific features. In particular, its third part defines how to store, retrieve, and process spatial data in SQL. Spatial data types and functions for converting, comparing, and analyzing spatial data are defined. Some examples of DBMSs implementing this standard include PostGIS [22], Oracle Database [20], IBM’s DB2 [11], MySQL [18], Microsoft SQL Server [17].

SQL/MM includes functions for testing the validity of relationships between spatial data, and computing operations over

spatial objects. Spatial functions include, for example, *ST\_Area* that computes the area of the geometry on which it is applied and *ST\_Overlaps()* that tests whether the geometry on which it is applied overlaps the geometry provided as parameter. By using these functions spatially-enabled DBMSs allow one also to perform spatial selection (i.e., a selection operation based on a spatial predicate) and spatial join (i.e., a join operation which join condition includes a spatial predicate). For example, using the SQL/MM standard, the following query retrieves the identifiers for each two bank branches that have a non-empty overlap in the zones and also the overlapping area, encoded in a text representation. Note that the join condition contains a spatial predicate *ST\_Overlaps()* that tests whether the geometry on which it is applied (i.e., b1.area) overlaps or not with the geometry passed as parameter (i.e., b2.area). Moreover, the query returns, besides the ids of the retrieved branches, the geometry representing the intersection of the areas of the two joined branches. The intersection is computed by using the *ST\_Intersection* function and converted in a readable textual form by using *ST\_AsText()*.

```
SELECT b1.branch_id, b2.branch_id,
       b1.area.ST_Intersection(b2.area).ST_AsText()
FROM branches AS b1 JOIN branches AS b2
     ON (b1.area.ST_Overlaps (b2.area))
WHERE b1.branch_id < b2.branch_id
```

### 2.3 Spatio-temporal query languages

After TSQL2 and SQL/MM Part 3, despite several spatio-temporal query languages have been proposed [7, 9, 10, 26, 30], there have not been efforts for standardizing a spatio-temporal model and query language.

Erwig and Schneider [9] extend spatial functions adding a temporal components. In this way, spatial functions may be applied to moving points and regions. For example, the *trajectory()* function returns a polyline representing the trajectory of a moving point over time while *traversed* computes the region traversed in any time by a moving region.

A slightly different approach has been proposed by Sistla et al. [26]. Authors introduce a query language based on the FTL (Future Temporal Logic) [25]. FTL queries are specified in the *Retrieve-Where* form, where the *Retrieve* clause specifies which data have to be returned and the *Where* clause specifies the **FTL** formula representing the condition that the retrieved data must comply with. The *NextTime* and *Until* modal temporal operators may be used in the *Where* clause. Temporal conditions can use spatial relations in order to perform spatio-temporal selection.

For example, the following query retrieves the pairs of objects *o* and *n* such that the distance between them remains smaller than 5 miles until they both enter the polygon *P*.

```
RETRIEVE o, n
WHERE dist(o, n) ≤ 5
      UNTIL(inside^ (o, P) inside (n, P))
```

We note that, no commercial DBMS offers ad-hoc support for spatio-temporal querying. Thus spatio-temporal querying is based on the usage of temporal and spatial operators in the (usually SQL-based) query languages provided by the DBMS.

### 2.4 A framework for granularities

Considering granularities, in [4] a framework for modeling and qualifying spatio-temporal data in relational databases using spatio-temporal granularities has been proposed. Informally, a temporal granularity represents a partition of a time domain. Each element of a granularity is called *granule* and represents a set of time instants perceived and used as an indivisible entity. Granules must be disjoint (i.e., no point can belong to two different granules) and their order must coincide with time point order (i.e., for each pair of granules *i* and *j*, with *i* < *j*, all time points belonging to granules *i* must precede all time points in granule *j*) We can use these granules to provide data with a temporal qualification at the suitable granularity. In other words, a temporal granularity represents a temporal unit of measure. For example, we can associate a patient's hospitalization with the granule representing September 30, 2010, thus proving it with a temporal qualification.

Similarly, a spatial granularity represents a partition of a space domain. Granules must have disjoint interior and may have holes and may be composed by several disconnected regions. In [4] spatial granularities are defined by using a two-level model. The lower level represents the spatial domain, on which we can recognize geometrical information and in which vector data representing granules are defined. The higher level is a multidigraph structure used to access and manage granules. A multidigraph is a labeled directed graph with multiple labeled edges. In the multidigraph each node represents a spatial granule. On the other hand, edges represent relationships between granules (e.g., direction- and distance-based relations). Each edge is labeled with the name of the relationship it represents.

Based on these two frameworks, the notion of spatiotemporal granularity has been proposed in [4]. Spatio-temporal granularities represent the evolution over time of a spatial granularity. A spatio-temporal granularity has two components. The former is a temporal granularity that aggregates time points, while the latter is a mapping (called *spatial evolution*) that associates to each time point the spatial granularity valid on it.

In [4] a database for granularities has been designed. The database allows one to store all information describing temporal, spatial, and spatio-temporal granularities. The following is a part of the relational schema of the database for granularities.

```

GRANULE (id, index, start, end, granularity)
T-GRANULARITY (id, name, domain, extentS, extentE, anchor)
ST-GRANULARITY (id, name, tgranularity, evolution)
EVOLUTION (id, name)
VALIDTIMEHISTORY (evolution, sgranularity, since, to)
S-GRANULARITY (id, name, domain, extent, domain_rel)
NODE (id, sgranularity, geom, label)
NODE-LABEL (id, label)

```

where **T-GRANULARITY** and **GRANULE** represent temporal granularities and their granules, respectively. **S-GRANULARITY** contains information about spatial granularities, while **NODE** and **NODE-LABEL** represent spatial granules (the nodes of the multidigraph) and their labels, respectively. Finally, **ST-GRANULARITY**, **EVOLUTION**, and **S-GRANULARITY** represent spatio-temporal granularities and their spatial evolutions.

### 3. A Motivating Clinical Database

The designed database for granularities may be used to qualify and aggregate data, also during the querying phase, in an existing relational database. In [4] authors propose two ways to qualify and aggregate information by using granularities. The first way provides to add, in each table containing data we want to qualify, a reference (i.e., a foreign key) to the granule representing the temporal or spatial location. The second way allows one to qualify data with granularities directly during the querying phase. In this case, tuples we are interested in are associated to a temporal and/or a spatial location (e.g., a timestamp or a coordinate on the Earth surface). Later (e.g., in a query) it is possible to “*dynamically*” aggregate information with respect to the temporal or spatial locations we added by using suitable granularities. Let us consider for example the following database schema:

```

PATIENT (id, tax_code, surname, name, birth_date)
CONTACT (id, patient, contact_date, gaf, contact_location, duration)
PATIENT_DIAGNOSIS (id, patient, diagnosis, start_date, end_date)
DIAGNOSIS (id, category, icd10_code, description)
PATIENT_PROFESSION (id, patient, profession, start_date, end_date)
PROFESSION (id, description)
PATIENT_EMPLOYMENT (id, patient, employment, start_date, end_date)
EMPLOYMENT (id, description)

```

```

PATIENT_RESIDENCE(id,patient,address, subzone,start_date,end_date)
PATIENT_DOMICILE(id,patient,address,subzone,start_date, end_date,location)
CONTACT_LOCATION(id,description,location)
SUBZONE(id,description,quarter,municipality)
MUNICIPALITY(istat_code,name,inhabitants_2001,node)

```

It is a part of the schema of the Verona Psychiatric Case Register (PCR). Verona is a city in North-East of Italy. Its Health District includes a Community-based Psychiatric Service (CPS). CPS aims at providing responses to psychiatric patients' practical as well as psychological and social needs, while trying to alleviate and control their symptoms. CPS structures include a psychiatric ward, an outpatient department, a consultation liaison office, a 24-hour accident and emergency room, a night and week-end emergency room, a 24-hour staffed hostel, a group home, and two apartments. CPS provides also home visits in response to emergency calls and, for chronic patients, they may be planned in advance for offering regular, long-term support. PCR is the information system collecting information about patients' accesses to CPS since 1979 [3]. At the first contact with the psychiatric service (e.g., a visit), socio-demographic information, past psychiatric and medical history, and clinical data are routinely collected for patients aged 14 years and over. These data may be updated at successive contacts when required. Each patients' contact with CPS structures is recorded in PCR. Recorded contacts include: attendances at the out-patient clinic, domiciliary visits, telephone calls, day cares provided at the day hospital units, all admissions to the acute psychiatric ward, and private clinics. Data on about 28,700 patients and more than 1,500,000 psychiatric contacts have been recorded in PCR up to now.

To each patient a diagnosis is assigned according to ICD 10 [32] categories. Moreover, PCR stores also patients' employment and profession. All these data may be updated when required, and thus PCR records also their valid time.

PCR is used for administrative (e.g., for calculating direct costs for different groups of patients [1]), clinical (e.g., for organizing contacts with patients at regular intervals), and research purposes. In research activities, PCR is used for epidemiological studies about the utilization of services [2], studies about the correlation between geographical factors (e.g., position and distance of services) and service utilization [33], discovering service-related, area-based, and socio-demographic factors influencing accesses to health services [21, 28, 29], and comparisons with other case-register areas [23].

The relations of PCR we reported above exemplify the two ways for using granularities for aggregating and querying spatio-temporal data. The relation *Municipality* contains a reference (i.e., the node attribute) to the corresponding granule in the *Municipalities* spatial granularity. On the other hand, the *Patient\_domicile* relation has a location attribute representing the coordinates on the Earth surface where the domicile is located. Moreover, all temporal relations contain the *start\_date* and *end\_date* attributes representing the valid time of the tuples in the relations. Location and valid time attributes may be used for spatially and temporally aggregating data at the querying phase.

In the following we will exemplify the capabilities of our *ST4SQL* query language also presenting some queries on PCR. These queries refer to patients' personal information and contacts with CPS and they may be useful to get reports for administrative purposes.

#### 4. The ST4SQL Query Language

In this section, we present *ST4SQL*, our query language for dealing with spatio-temporal data and granularities. It is a spatial extension of *T4SQL*, the temporal query language proposed by Combi et al. [8] and discussed in Section 2. *ST4SQL* is an SQL-based query language. It extends the temporal constructs of *T4SQL* to deal with temporal granularities and it adds new spatial constructs, similar to the temporal ones, for dealing also with spatial dimensions and granularities.

*ST4SQL* queries are based on the usual *SELECT-FROM-WHERE* structure. *ST4SQL* adds new constructs for the "automatic" management of spatial dimensions. Supported dimensions include only valid space, that is the spatial location where an object instance is "true". Values for valid space can be modeled, for example, as points, lines, or regions.

As we already explained, one of the main contributions of *T4SQL* has been the introduction of four semantics for temporal queries. These semantics allows one to specify how the system has to manage temporal dimensions in order to evaluate queries. The proposed semantics for temporal querying are defined on time points. Thus, for example, the *SEQUENCED* semantics applied to valid time considers instant by instant the temporal domain and evaluates the query taking into account only those tuples whose valid time contains the considered time point. This approach allows one, for example, to query PCR for retrieving diagnosis of unemployed patients that are searching for a new job. Thus, the valid time of diagnoses must overlap the valid time of patients' employment status and this requirement is implicitly provided by the *SEQUENCED* semantics.

```

SEMANTICS SEQUENCED ON VALID
SELECT p.name,p.surname,d.description
FROM diagnosis AS d,patient_diagnosis AS pd,
      patient AS p,patient_employment AS pe,
      employment AS e
WHERE pd.diagnosis = d.id AND
      pd.patient = p.id AND
      pe.patient = p.id AND
      pe.employment = e.id AND
      e.description = 'searching for a new job'

```

However, we can consider semantics also on temporal granularities. In this way, the system evaluates queries considering granules instead of time points and, thus, constraining tuples with respect to their value on temporal dimensions according to temporal granules. To do that, we extend previous temporal semantics. Thus, each semantics requires the user to specify a time dimension (e.g., valid time) and a temporal granularity (e.g., months). Thus, semantics can be reformulated as follow:

- *SEQUENCED(td, tG)*: the DBMS evaluates the submitted query only on those tuples whose value for the temporal dimension *td* intersects the same temporal granule of the temporal granularity *tG*.
- *CURRENT(td, tG)*: the DBMS evaluates the query only on those tuples whose value over *td* intersects the granule of *tG* containing the current date.
- *NEXT(td, tG)*: this semantics considers only pairs of tuples belonging to the join of the tables specified in the FROM clause that are related to the same entity and that are in two consecutive granules of *tG*, with respect to the temporal ordering.
- *ATEMPORAL(td)*: it disables any support from the system, thus *td* is considered a classic attribute managed by the user and it does not require to *specify* a time granularity.

In other words, semantics based on granularities allow one to “*relax*” *T4SQL* queries by grouping tuples with respect to the granules of a temporal granularity instead of time points.

A spatial semantics allows one to specify how the query engine must manage spatial dimensions associated to tuples. We define four spatial semantics: *SPACELESS* (corresponding to the *ATEMPORAL* one), *SEQUENCED*, *CURRENT*, and *NEXT*. Semantics for spatial queries can be formulated as follow.

- *SEQUENCED(sd, sG)*: the DBMS evaluates the submitted query selecting only those tuples whose value for the spatial dimension *sd* intersects the same spatial granule.
- *CURRENT(sd, sG)*: the DBMS evaluates the query only on those tuples whose value over *sd* intersects the granule of the spatial granularity *sG* containing the current user's position (of course, it requires that the user's position is available or can be determined, for example by analyzing the user's IP address).
- *NEXT(sd, sG, R)*: this semantics considers only those tuples related to the same entity and that are in two consecutive granules of *sG* with respect to the ordering (eventually non-total) defined by the spatial relationship *R*.
- *SPACELESS(sd)*: it disables any support from the system, thus *sd* is considered a classic attribute managed by the user.

Note that we define only spatial semantics based on granularities, not on space points. This limitation is based on the consideration that joining only tuples spatially qualified exactly with the same point location is not meaningful, especially considering that many errors and uncertainties can affect spatial surveys.

The syntax of *ST4SQL* extends that of *T4SQL* (and that of *SQL*). Its (incomplete) *BNF* is as follows:

```
[SEMANTICS [<tsem> ON <tdim>
  [WITH TGRANULARITY <tG>] [THROUGH <attr>]
  [TIMESLICE <ts_exp>] [, ...] ,]
 [<ssem> ON <sdim> WITH SGRANULARITY <sG>
 [ORDERED BY <sr>] [THROUGH <attr>]
 [SPACESLICE <ss_exp>] [, ...] ,]
 [<tsem> ON <tdim> [TIMESLICE <ts_exp>] AND
 <ssem> ON <sdim> [SPACESLICE <ss_exp>]
 WITH STGRANULARITY <stG>
 [SIMPLIFY BY <s_aggr_func>]]]
SELECT <sel_element_list>
  [WITH <exp> [AS] <dim> [, ...]]
  [, TGROUP (<t_attr>) [AS] <new_name> [, ...]]
  [, SGROUP (<s_attr>) [AS] <new_name> [, ...]]
[FROM <tables>]
[WHERE<conditions>]
[WHEN <t_conditions>]
[WHEREABOUTS <s_conditions>]
[GROUP BY <group_element_list>]
[HAVING <g_cond>]
<tsem> ::= ATEMPORAL | CURRENT | SEQUENCED | NEXT
<ssem> ::= SPACELESS | CURRENT | SEQUENCED | NEXT
<dim> ::= <tdim> | <sdim>
<tdim> ::= AVAILABILITY | TRANSACTION |
          VALID TIME | INITIATING_ET | TERMINATING_ET
<sdim> ::= VALID SPACE
<group_element_list> ::= <group_element> [, ...]
<group_element> ::= <attribute> |
                   <temp_attr> ON <tG> | <space_attr> ON <sG>
```

where:

- **tG** and **sG** represent a temporal and a spatial granularity, respectively;
- **ts\_exp** and **ss\_exp** represent expressions corresponding to a time interval and a spatial region, respectively;
- **<sr>** is the name of a spatial relationship between granules whose definition must be present in the **SR** relation of the proposed database for granularities.

We introduce also some functions for accessing the value of tuples on temporal and spatial dimensions. As we will explain, these accessors can be used both for including the value of the tuples on a dimension in the output relation and for specifying any condition on the value of the tuples on a dimension. Given a tuple of a relation *R*, the value it takes over the temporal and the spatial dimensions can be recovered by using the following accessors:

- VALIDT (R)** returns the valid time of the R tuple;
- TRANSACTIONT (R)** returns the transaction time of the R tuple;
- AVAILABLET (R)** returns the availability time of the R tuple;
- INITIATING\_ET (R)** returns the initiating event time of the R tuple;
- TERMINATING\_ET (R)** returns the terminating event time of the R tuple;

**VALIDS (R)** returns the valid space of the R tuple;

As we already mentioned in previous section, Belussi et al. proposed two ways to qualify data with spatio-temporal granularities in two alternative ways: (1) indirect (tuples are associated to a temporal and/or spatial location that is qualified with granularities at query time) and (2) direct (tuples are associated directly to a granule in a temporal, spatial, or spatio-temporal granularity). In the first case accessors return the location associated to the tuples, while in the second case accessors return the granule associated to the tuples.

Since ST4SQL includes SQL and T4SQL, we now describe only new spatial constructs defined in ST4SQL.

#### 4.1 The clause **SEMANTICS**

The **SEMANTICS** clause allows one to specify the semantics to be applied for temporal, spatial, or spatio-temporal dimension. More semantics may be specified on different dimensions, while at most one semantics can be applied to a dimension, otherwise the query is considered to be not well-formed.

In particular, a semantics for a temporal dimension may be specified with the following syntax:

```
SEMANTICS <tsem> ON <tdim>
           [WITH TGRANULARITY <tG>] [TIMESLICE <ts_exp>]
```

while, with a similar syntax, the user may specify a spatial semantics:

```
SEMANTICS <ssem> ON <sdim>
           WITH SGRANULARITY <sG> [ORDERED BY <sr>]
           [THROUGH <attr>] [SPACESLICE <ss_exp>]
```

As we have already explained above, each semantics requires a dimension (<tdim> or <sdim>) on which it has to be applied. Moreover, optionally, a temporal semantics can be applied in relation to a temporal granularity <tG> that will be used to group and select the tuples on which the query has to be evaluated. Since spatial semantics are defined only on granularities, the **WITH SGRANULARITY <sG>** token is mandatory and <sG> is the name of the spatial granularity to be used.

The **TIMESLICE** and **SPACESLICE** tokens are optional and allow one to restrict the query evaluation only on those tuples whose value for the given temporal or spatial dimension intersects the value of <ts\_exp> and <ss\_exp> expressions, respectively. <ts\_exp> must represent a time interval, when the **WITH TGRANULARITY <tG>** option is not specified, or a pair (gstart,gend) representing a granule interval (where gstart and gend are the indexes of the first and the last granule in <tG> in the interval), when the **WITH TGRANULARITY <tG>** token is present. Similarly, <ss\_exp> is a list of labels of granules in the <sG> spatial granularity. The **TIMESLICE** and **SPACESLICE** tokens can be specified together with any semantics but the **CURRENT** one.

Note that, when no semantics is specified by the user on a temporal or spatial dimension, the **CURRENT** semantics is applied as default by the system on that dimension.

##### 4.1.1 **SPACELESS** spatial semantics

The **SPACELESS** spatial semantics corresponds to the **ATEMPORAL** one we described before. The **SPACELESS** semantics disables any support from the system and the user has, eventually, to manage explicitly the spatial dimension on which it is applied. No spatial granularity has to be specified together with the **SPACELESS** semantics unless the **SPACESLICE** option is specified: in this case the granularity is mandatory.

For example, the following query retrieves the patients living in a highly polluted area and contacting CPS structures located in little polluted areas.

Where **st\_contains** is the SQL/MM procedure that checks whether the object on which it is called contains the geometry provided as a parameter.

```

SEMANTICS SEQUENCED ON VALID TIME
      WITH TGRANULARITY Weeks,
      SPACELESS ON VALID SPACE
SELECT DISTINCT p.tax_code,p.name,p. surname
FROM patient AS p JOIN contact AS c
      ON p.id = c.patient
JOIN patient_domicile AS pd
      ON pd.patient = p.id
JOIN contact_location AS cl
      ON c.contact_location = cl.id,
s- granularity AS sg JOIN node AS n1
      ON n1.sgran = sg.id
JOIN nodelabel AS n11 ON n1.label = n11.id
JOIN node AS n2 ON n2. sgran = sg.id
JOIN nodelabel AS n12 ON n2.label = n12.id
WHERE sg.name = 'PM10' AND n11.label = 'High'
      AND n12.label = 'Low' AND
      n1.geom.st_contains(pd.location) AND
      n2.geom.st_contains(cl.location)

```

Using the *SPACELESS* semantics, the user disable the system support on the spatial dimension and she manage spatial data explicitly. For this reason, spaceless queries may be longer, more complex, and less readable than queries based on other semantics.

The output relation does not include the dimension on which the *SPACELESS* semantics has been applied.

Since the *SPACELESS* semantics disables any automatic behavior, it is possible to write a spaceless query in which spatial dimensions are explicitly managed in the way that it has the same result that could be obtained by applying any other spatial semantics.

#### 4.1.2 SEQUENCED spatial semantics

The *SEQUENCED* semantics evaluates the query granule by granule. For each granule, only those tuples whose value for the spatial dimension intersects the considered spatial granule are taken into account.

It allows us to retrieve, for example, the final report of contacts of patients living (when contacts occurred) in the same subzone (i.e., quarter) where contacts took place.

```

SEMANTICS SEQUENCED ON VALID TIME,
      SEQUENCED ON VALID SPACE
      WITH SGRANULARITY Subzones
SELECT c.patient,c.report
FROM patient_domicile AS pdom,contact AS c,
      contact_location AS cl
WHERE c.patient = pdom.patient AND
      c.contact_location = cl.id

```

The *SEQUENCED* temporal semantics restricts the query evaluation to the patients' domiciles valid when contacts occurred. On the other hand, the *SEQUENCED* spatial semantics provides that contacts occur in the same subzone where patient's domicile is located.

Similarly to what happens in the temporal case, the *SEQUENCED* semantics provides that the spatial dimension on which it is applied is included in the output relation. Thus, in the previous query both the time instant and the spatial granule are returned besides other required data.

Considering another example, the following query returns the patients whose residence is currently in a municipality that

intersects an area with a high level of acoustic noise.

```
SEMANTICS CURRENT ON VALID TIME,  
    SEQUENCED ON VALID SPACE  
    WITH SGRANULARITY DayAcousticNoise  
    SPACESLICE 'High'  
SELECT p.tax_code,p.name,p. surname  
FROM patient AS p,patient_residence AS pr,  
    subzone AS s, municipality AS m  
WHERE pr.patient = p.id AND  
    pr.subzone = s.id AND  
    s.municipality = m.istat_code
```

The *SEQUENCED* spatial semantics relates the patient's residence (through the subzone and municipality relations) with spatial granules representing the day acoustic noise levels. The **SPACESLICE** token has been used to restrict the query evaluation only to the spatial granule labeled with High, that we suppose to represent areas with a high level of noise pollution.

#### 4.1.3 CURRENT spatial semantics

The *CURRENT* spatial semantics can be used to restrict the query evaluation only on those tuples whose value over the specified spatial dimension intersects the spatial granule containing the current geographical position of the query submitter. This information can be locally saved, for example, in a system configuration file, or can be inferred from the DBMS server machine, for example analyzing the IP address from which the query has been received.

Considering the PCR database, the *CURRENT* semantics can be used, for example, by a physician for retrieving diagnoses only of patients currently living in the catchment area where she works.

```
SEMANTICS CURRENT ON VALID TIME,  
    CURRENT ON VALID SPACE  
    WITH SGRANULARITY CatchmentAreas  
SELECT DISTINCT p.name, p.surname, d.icd10_code  
FROM patient_domicile AS dom, patient AS p,  
    patient_diagnosis AS pdia, diagnosis AS d  
WHERE dom.patient = p.id AND p.id = pdia.patient  
    AND pdia.diagnosis = d.id
```

Similarly to the *CURRENT* temporal semantics, the spatial one cannot be paired with the **SPACESLICE** option and the spatial dimension is not included in the output relation.

#### 4.1.4 NEXT spatial semantics

Similarly to the temporal one, the *NEXT* spatial semantics allows one to relate tuples related to the same object in two consecutive granules. As in the temporal case, two tuples are considered to be related to the same object if they share the same value on a attribute provided by the user with the **THROUGH** token. On the other hand, conversely to the temporal case, a unique order is not defined between spatial points and granules. For this reason, the user must provide also the spatial relationship the system has to use for ordering spatial granules and thus to find consecutive granules. This can be done with the **ORDERED BY <sr>** token, where **<sr>** is the name of a spatial relationship defined in the **SR** table of the database for granularities.

For example, the following query relates the number of contacts for each patient in a quarter and in the quarters that are South of it.

The behavior of the *NEXT* spatial semantics is similar to the temporal one. The query is evaluated on pairs of tuples belonging to the relation resulting from the join between two instances of the relations in the **FROM** clause. These two instances are joined on the basis of the attribute specified in the **THROUGH** token. Among the obtained pairs of tuples the system selects only those whose values over the given spatial dimension intersect two consecutive granules according to the order defined by the relationship specified in the **ORDERED BY** token. In the **SELECT** and **WHERE** clauses, the **NEXT(<attr>)** function can

be used to refer to the value of the <attr> attribute in the tuple associated to the successor granule.

```

SEMANTICS NEXT ON VALID SPACE
      WITH SGRANULARITY Quarters
      ORDERED BY South THROUGH c.patient
SELECT c.patient, COUNT(c.id), COUNT (NEXT(c.id))
FROM contact AS c JOIN contact_location AS cl
      ON c.contact_location = cl.id
GROUP BY c.patient

```

By default, the spatial dimension on which the *NEXT* semantics is applied is not included in the output relation.

#### 4.1.5 Spatio-temporal semantics

In previous examples about the usage of spatial granularities, we always aggregated patients' domiciles using a unique *Municipality* granularity, supposing that it never changes over time. However, patients' domiciles are spatio-temporal data, thus they may be aggregated by using the spatial granularity that actually was valid at the time when domiciles were valid. Spatio-temporal granularities can be used in queries for this purpose. Spatio-temporal granularities temporally qualify spatial granularities, thus they allow one to spatially qualify data with respect to a spatial granularity whose definition changes over time. A spatio-temporal granularity groups spatial granularities with respect to granules of a temporal granularity. Each tuple is spatially qualified by using spatial granularities valid during the time granules intersecting the value of the tuple over the given temporal dimension. Since this value may be an interval and during this interval the spatial granularity may evolve, several spatial granularities may be used to qualify each tuple. For this reason the user may specify an aggregate function to apply to spatial granularities in order to obtain just one spatial granularity among all granularities valid during the interval. Aggregate functions include, for example, *first* (that returns the spatial granularity valid at the first instant in the interval), *last* (that returns the spatial granularity valid at the last instant in the interval), and operations for calculating union, intersection, and difference of spatial granularities [4]. If the aggregate function is not provided by the user, the system uses all spatial granularities valid in the interval, one by one, for querying the considered tuple, and all these combinations are used for obtaining the final resulting relation.

The user specifies a semantics on both the temporal and the spatial dimension making up the given spatio-temporal dimension. These temporal and spatial semantics are applied as we described above in this section, but, moreover, the temporal dimension is used also for selecting the spatial granularity valid at the same time of considered tuples.

We note that specifying the *SEQUENCED* semantics on the spatial dimension and the *CURRENT* semantics (or a timesliced *SEQUENCED* semantics) on the temporal one, a user can query data with respect to a spatially qualified timeslice. On the other hand, specifying the *SEQUENCED* semantics on the temporal dimension and the *CURRENT* semantics (or a spacesliced *SEQUENCED* semantics) on the spatial one, a user can query data with respect to the evolution over time of spatial granules. The *ATEMPORAL* and *SPACELESS* semantics cannot be used in spatio-temporal semantics.

In summary, a spatio-temporal semantics may be specified by using the following BNF syntax:

```

SEMANTICS <sem> ON <td> [TIMESLICE <ts_exp>] AND
      <sem> ON <sd> [SPACESLICE <ss_exp>]
      WITH STGRANULARITY <stG>
      [SIMPLIFY BY <s_aggr_funct>]

```

Where *td* and *sd* are a temporal and a spatial dimension, respectively, and *s\_aggr\_funct* is a spatial aggregate function.

For example, the following query retrieves information about patients' domicile valid in 2010 and located, during this interval, in an area with high or medium PM10 pollution level.

*PM10\_Weeks* is a spatio-temporal granularity representing daily PM<sub>10</sub> surveys associated to the *Weeks* temporal granularity. Thus, the query is evaluated on those tuples whose value over the valid space intersects a *PM10* granule valid in a week intersecting also the tuples value over the valid time.

The output relation includes both the temporal and spatial granules qualifying tuples.

```
SEMANTICS SEQUENCED ON VALID TIME
      TIMESLICE (2010 -01 -01 ,2010 -12 -31) AND
      SEQUENCED ON VALID SPACE SPACESLICE 'High','Medium'
      WITH STGRANULARITY PM10_Weeks
SELECT DISTINCT p.tax_code, p.name, p.surname
FROM patient_domicile AS pd, patient AS p
WHERE pd.patient = p.id
```

#### 4.2 The clause SELECT

After the **SEMANTICS** clause, ST4SQL queries continue with the usual **SELECT** and **FROM** clauses. As usual, the **SELECT** clause specifies what attributes (or expressions) have to be returned for each tuple in the resulting relation. The **FROM** clause contains the list of relations (that, eventually, may be the result of a subquery) required for processing the query.

As we explained before, only the **SEQUENCED** semantics provides to include the value for the dimension on which it is applied in the output relation. In order to change this behavior or include the value for the dimensions also when the other semantics are applied, the user can, besides classical attributes, use the **WITH** token in the **SELECT** clause.

Thus, for example, the **WITH** token can be added to the query we introduced above for retrieving diagnoses only of patients currently living in the catchment area where the submitter works. In this case also the geographical position of the domicile is included in the output relation.

```
SEMANTICS CURRENT ON VALID SPACE
      WITH SGRANULARITY CatchmentAreas
SELECT DISTINCT p.name, p.surname, d.icd10_code
      WITH VALIDS (dom) AS domicile
FROM patient_domicile AS dom, patient AS p,
      patient_diagnosis AS pdia, diagnosis AS d
WHERE dom.patient = p.id AND p.id = pdia.patient
      AND pdia.diagnosis = d.id
```

#### 4.3 The clause WHEREABOUTS

The **WHERE** clause contains the selection and join conditions to apply in order to select the set of tuples on which the query must be evaluated. In **ST4SQL**, the **WHERE** clause may include temporal and spatial conditions (i.e., conditions involving the value of temporal and spatial dimensions). In some cases, spatial conditions are complex and for the sake of clarity they could be kept divided from classical conditions. For this purpose, **ST4SQL** provides the **WHEREABOUTS** clause where spatial conditions can be specified. Let us consider for example the following query. It retrieves the duration of the contacts occurred at most one month after that the diagnosis of the corresponding patient has been established and at most at 1 km from the patient's domicile.

#### 4.4 The clause GROUP BY

The **GROUP BY** clause allows one to define groups of tuples on which aggregate functions may be applied. **ST4SQL** follows the user also to group tuples with respect to their value on a spatial dimension. To do that the user can use the spatio-temporal accessors we introduced above in the **GROUP BY** clause. For example, **GROUP BY VALIDS (T) ON <SG>** specifies that the query has to be evaluated on the groups of tuples which value for the valid space of the T relation are contained in the same granule of **<SG>**.

In order to include the value of groups on a dimension in the output relation, the user can include the **SGROUP** function in the **SELECT** clause. This function requires as parameter the accessor used in the **GROUP BY** clause and return the granule of **<SG>**.

```

SEMANTICS SEQUENCED ON VALID TIME,
          SPACELESS ON VALID SPACE
SELECT c.duration
FROM contact AS c JOIN patient_diagnosis AS pd
     ON c.patient = pd.patient
     JOIN contact_location AS cl
     ON c.location = cl.id
     JOIN patient_domicile AS pdom
     ON c.patient = pdom.patient
WHEN VALIDT(c)$start - VALIDT (pd)$start <INTERVAL '1' MONTH
WHEREABOUTS VALIDS (cl).st_distance (VALIDS (pdom))<1 KM

```

Thus, for example, the following query retrieves the average number of contacts per patient grouped with respect to the diagnosis category and the day acoustic noise level.

```

SELECT SGROUP (VALIDS (cl)), d. category,
        COUNT(c.id )/COUNT ( DISTINCT c.patient)
FROM contact AS c, patient_diagnosis AS pd,
     diagnosis AS d, contact_location AS cl
WHERE c.patient = pd.patient AND
     pd.diagnosis = d.id AND
     c.contact_location = cl.id
GROUP BY VALIDS (cl) ON DayAcousticNoise, d.category

```

The **HAVING** clause allows one to specify conditions on the groups. The **HAVING** clause provides conditions that have to be applied to each group of tuples, then only groups that meet such conditions are retrieved in the output relation. The **HAVING** clause may include aggregate functions, temporal and spatial conditions.

#### 4.5 On the semantics of ST4SQL

All *ST4SQL* queries can be translated into equivalent *SQL* queries by using standard constructs. Hence, it is possible to specify the semantics of *ST4SQL* with respect to the one of *SQL*. This can be done by showing how *ST4SQL* constructs can be translated into *SQL*. Here we show only a fragment of the *ST4SQL* semantics. All other cases are treated similarly to the showed one. In particular, we introduce only the translation of the **NEXT** spatial semantics.

The semantics of the accessors and other functions we introduced in the previous section about the syntax of *ST4SQL* has been already partially explained introducing them. More attention requires instead the **SEMANTICS** clause. As we have already explained, it allows one to specify how the temporal, spatial, and spatio-temporal dimensions have to be managed by the system for evaluating the query. In general, the clause is equivalent to some join and selection conditions that allow the system to select the right tuples on which the query have to be evaluated.

Let us briefly introduce the semantics of the **SEMANTICS** clause for some exemplifying cases. For example, let us consider the **ATEMPORAL** semantics. When it is specified on a temporal dimension (e.g., valid time), it disable any system support on the given dimension. Thus, the dimension associated to tuples is considered as usual atemporal attributes without any specific meaning. In this case the user has to manage these attributes when she needs. Of course, in this case the submitted query is translated into an equivalent *SQL* query without adding anything. Thus, given the following *ST4SQL* query:

```

SEMANTICS ATEMPORAL ON <tdim>
SELECT <sel_element_list>
FROM <tables>
WHERE <conditions>

```

its *SQL* equivalent query is:

```

SELECT <sel_element_list>
FROM <tables>
WHERE <conditions>

```

The **TIMESLICE** *<ts\_exp>* option can be specified together with the **ATEMPORAL** semantics. In this case *<ts\_exp>* must represent a time interval  $(t_1, t_2)$ . The meaning of **TIMESLICE** is to limit the query evaluation only to those tuples whose value for the given temporal dimension intersects the  $(t_1, t_2)$  interval. Since, no other relations between involved tables are required by the semantics, this constraint must be formulated separately for each table in the **FROM** clause. Thus, let

```

SEMANTICS ATEMPORAL ON <tdim> TIMESLICE <ts_exp>
SELECT <sel_element_list>
FROM T1, T2, ..., Tn
WHERE <conditions>

```

be the considered ST4SQL query. Then, it is equivalent to the following SQL query.

```

SELECT <sel_element_list>
FROM T1, T2, ..., Tn
WHERE <conditions> AND
intersect(VALIDT(T1), <ts_exp>) AND ...
AND intersect(VALIDT(Tn), <ts_exp>)

```

where we supposed that *<tdim>* is **VALID TIME**. Otherwise, the correspondent accessor must be used in place of the **VALIDT** function. **intersect**(*i1, i2*) is a function we defined for the sake of simplicity. It checks whether the given time intervals intersects each other. It can be easily implemented using usual SQL comparison operators and pair accessors.

Finally, also the **WITH TGRANULARITY** *<tG>* option can be specified in order to limit the query evaluation only to some granules of *<tG>*. In this case, *<ts\_exp>* represents a pair (*gstart, gend*) representing a granule interval where *gstart* and *gend* are the indexes of the first and the last granule in *<tG>* in the interval, respectively. The two elements of the pair are accessed by using *<ts\_exp>\$start* and *<ts\_exp>\$end*, respectively. In order to apply the specified timeslice, the **FROM** clause has to be extended with a subquery for calculating the time interval corresponding to the granule interval. Moreover, the **WHERE** clause must include, for each table  $T_i$  that includes the temporal dimension *tdim* (e.g., valid time), the conditions to constrain the query evaluation only to the calculated interval. Thus, let

```

SEMANTICS ATEMPORAL ON <tdim> TIMESLICE <ts_exp>
WITH TGRANULARITY <tG>
SELECT <sel_element_list>
FROM T1, T2, ..., Tn
WHERE <conditions>

```

be the considered ST4SQL query. It is equivalent to the following SQL query.

```

SELECT <sel_element_list>
FROM T1, T2, ..., Tn ,
      (SELECT MIN(g.start) AS start, MAX(g.end) AS end
       FROM t-granularity AS tg, granule AS g1, granule AS g2
       WHERE g1.TGRANULARITY = tg.id AND
             g2.TGRANULARITY = tg.id AND tg.name = <tG>
             AND g1.index = <ts_exp> $start
             AND g2.index = <ts_exp> $end) AS s1
WHERE <conditions> AND
      intersect(VALIDT(T1), (s1.start ,s1.end)) AND ...
      AND intersect(VALIDT(Tn), (s1.start ,s1.end))

```

Let us now consider a more complex semantics: the spatial *NEXT* semantics.

```

SEMANTICS NEXT ON <sdim> WITH SGRANULARITY <sg>
    ORDERED BY <sr> THROUGH <attr>
SELECT <attr_list>
FROM T1, ..., Tn
WHERE <conds > AND <sconds>

```

where *<attr\_list>* is a list of names of attributes in the  $T_1, \dots, T_n$  relations, while *<jconds>* and *<sconds>* are the sets of join and selection conditions. Note that, the distinction between join and selection conditions is not specified by the user, but can be easily computed by the system. Moreover, we remember that when the *NEXT* semantics is specified, also the **ORDERED BY** and **THROUGH** options are mandatory.

The *NEXT* semantics allows one to evaluate a query on the pairs of tuples representing two states of an object in two consecutive granules. To do that, the first thing the system has to do is to calculate the table containing the tuples representing the objects to pair. This table is the result of the join of all relations in the original **FROM** clause (i.e.,  $T_1, \dots, T_n$ ) according to the join conditions *<jconds>*. Thus, instance is calculated as follow:

```

(SELECT
FROM T1, ..., Tn
WHERE <jconds>) AS instance

```

where, as usual, when the user does not specify the join condition between all relations, the Cartesian product is applied. Then, in the **FROM** clause of the translated SQL query we have to:

- join two copies of the instance relation according to their values on the *<attr>* attribute, provided by the user as parameter of the **THROUGH** option;
- add the relations containing the information about the *<sg>* spatial granularity.

In the **WHERE** clause, we have to select only the pairs whose two constituent tuples intersect two consecutive spatial granules in *<sg>* with respect to the order defined by the *<sr>* spatial relation.

When the *NEXT* semantics is applied, the user can refer in the **SELECT** and **WHERE** clauses to the value of an attribute in the successor tuple by using the *NEXT(<attr>)* function. That is translated in the SQL query as *tnext.<attr>*, while all other attributes are referred as *t.<attr>*. Finally, the **SELECT** clause of the resulting SQL query contains all the attributes named by the user in the original ST4SQL query, where, eventually, the *NEXT(<attr>)* function is translated as explained.

Thus, the SQL query equivalent to the considered one is:

```

SELECT <attr_list>
FROM sgranularity AS sg JOIN node AS n
    ON (n.granularity = sg.id)
    JOIN node AS nnext
        ON (nnext.granularity = sg.id),
instance AS t JOIN instance AS tnext
    ON t.<attr> = tnext.<attr>
WHERE <sconds> AND sg.name = <sg> AND
    n.geom.st_intersects (VALIDS(t)) AND
    nnext.geom.st_intersects (VALIDS (tnext))
    AND NOT EXISTS(SELECT node .id FROM node
        WHERE node.granularity = sg.id AND
            evaluate(<sr>,n.geom, node.geom) AND
            evaluate(<sr>,node.geom, nnext.geom))

```

where `evaluate(r, g1, g2)` is a function we defined for evaluating the spatial relationship `r` between geometries `g1` and `g2` while `st_intersects` is the *SQL/MM* function testing whether two geometries intersect each other.

It is clear that SQL queries equivalent to *ST4SQL* ones are more complex, less readable, and they contain, in some cases, subqueries. Thus, computational complexity and performance issues arise. In order to keep query evaluation time reasonable, we need to take into account these problems and apply query optimization techniques and indexing structures when we will completely implement our language.

## 5. Conclusions

In this paper, we proposed a new query language, called *ST4SQL*, for dealing with temporal and spatial dimensions qualified with granularities. *ST4SQL* provides support for several temporal and spatial dimensions. *ST4SQL* introduces four temporal and spatial semantics. These semantics add a specific meaning to queries and allow the user to specify how the system has to (automatically) manage dimensions for evaluating the query. Moreover, *ST4SQL* allows the user to group data with respect to their spatio-temporal components. In *ST4SQL* all proposed constructs may use granularities for qualifying and querying spatio-temporal data. We introduced the syntax of *ST4SQL* and we exemplified it with several queries on a real psychiatric database. All *ST4SQL* queries can be translated into equivalent SQL queries, thus the semantics of *ST4SQL* can be specified with respect to the SQL one. The goal of *ST4SQL* is to provide constructs for querying spatio-temporal databases and exploit spatio-temporal components of data. New constructs may be translated into equivalent sets of SQL constructs. For this reason, the purpose of *ST4SQL* is not to improve the performance of spatio-temporal queries, but to facilitate their writing.

As for the ongoing work, we want to explore issues related to the spatio-temporal querying (e.g., spatio-temporal index structures) in order to extend *ST4SQL* and optimize the translation of *ST4SQL* toward SQL and the evaluation of obtained spatio-temporal queries.

## References

- [1] Amaddeo, F., Beecham, J., Bonizzato, P., Fenyo, A., Knapp, M., Tansella, M. (1997). The use of a case register to evaluate the costs of psychiatric care, *Acta Psychiatrica Scandinavica*, 91 (3) 189 -198.
- [2] Amaddeo, F., Bisoffi, G., Micciolo, R., Piccinelli, M., Tansella, M. (1997). Frequency of contact with community-based psychiatric services and the lunar cycle: a 10-year case-register study. *Social Psychiatry and Psychiatric Epidemiology*, 32, 323-326. 10.1007/BF00805436.
- [3] Amaddeo, F., Tansella, M. (2009). Information systems for mental health. *Epidemiologia e psichiatria sociale*, 18, 1 - 4.
- [4] Belussi, A., Combi, C., Pozzani, G. (2009). Formal and conceptual modeling of spatio-temporal granularities. In: *IDEAS*, 275 - 283, Cetraro, Calabria, Italy, Sept. ACM.
- [5] Bettini, C., Dyreson, C. E., Evans, W. S., Snodgrass, R. T., Wang, X. S. (1998). A glossary of time granularity concepts. *LNCIS*, 1399, 406 - 413.
- [6] Camossi, E., Bertolotto, M., Bertino, E., Guerrini, G. (2003). A multigranular spatiotemporal data model. In: *GIS-03*, 94 - 101. ACM Press, 7- 8.
- [7] Chen, C. X., Zaniolo, C. (2000). SQLST : A spatio-temporal data model and query language. In: *ER*, 96 -111.
- [8] Combi, C., Montanari, A., Pozzi, G. (2007). The T4SQL temporal query language. In: *CIKM*, 193 - 202, Lisbon, Portugal, ACM.
- [9] Erwig, M., Schneider, M. (1999). Developments in spatio-temporal query languages. In: *DEXA Workshop*, 441 - 449.
- [10] Güting, R. H., Böhlen, M. H., Erwig, M., Jensen, C. S., Lorentzos, N. A., Schneider, M., Vazirgiannis, M. (2000). A foundation for representing and querying moving objects. *ACM T. Database Syst.*, 25 (1) 1-42.
- [11] IBM. DB2. URL: <http://www.ibm.com/software/data/db2/>.
- [12] International Organization for Standardization. ISO/IEC 9075-14. (2008). Information technology *Database languages - SQL - Part 14: XML-Related Specifications (SQL/XML)*.

- [13] International Organization for Standardization. *ISO/IEC 13249-3 (2006). Information technology - Database languages - SQL Multimedia and Application Packages.*
- [14] Jensen, C. S., Dyreson, C. E., Böhlen, M. H., and et al. (1998). The consensus glossary of temporal database concepts - february version. *In: Temporal Databases*, 367 - 405.
- [15] Khatri, V., Ram, S., Snodgrass, R. T., O'Brien, G. M. (2002). Supporting user-defined granularities in a spatiotemporal conceptual model. *Ann. Math. Artif. Intell.*, 36 (1-2).
- [16] Koubarakis, M., Sellis, T. K., Frank, A. U., et al., editors. (2003). *Spatio-Temporal Databases: The CHOROCHRONOS Approach*, volume 2520 of LNCS. Springer.
- [17] Microsoft Corporation. SQL server. URL: <http://www.microsoft.com/sqlserver/2008/>.
- [18] Oracle. MySQL. URL: <http://www.mysql.com/>.
- [19] Oracle Corporation. Oracle database. URL: <http://www.oracle.com>.
- [20] Oracle Corporation. Oracle spatial. URL: <http://www.oracle.com>.
- [21] Pertile, R., Donisi, V., Grigoletti, L., Angelozzi, A., Zamengo, G., Zulian, G., Amaddeo, G., Drgs, F., other patient-, service-and area-level factors influencing length of stay in acute psychiatric wards: the Veneto region experience. *Social Psychiatry and Psychiatric Epidemiology*, 1-10.
- [22] Refrations Research. PostGIS. URL: <http://postgis.refrations.net>.
- [23] Rossi, A., Morgan, V., Amaddeo, F., Sandri, M., Tansella, M., Jablensky, A. (2005). Psychiatric out-patients seen once only in South Verona and Western Australia: a comparative case-register study. *Australian and New Zealand Journal of Psychiatry*, 39(5)414-422.
- [24] Sarda, N. L. (2005). Hsql: A historical query language. *In: Temporal Databases*, 110 - 140.
- [25] Sistla, A. P., Wolfson, O. (1995). Temporal triggers in active databases. *IEEE Trans. Knowl. Data Eng.*, 7 (3) 471 - 486.
- [26] Sistla, A. P., Wolfson, O., Chamberlain, S., Dao, S. (1997). Modeling and querying moving objects. *In: Proceedings of the Thirteenth International Conference on Data Engineering*, April 7-11, Birmingham U.K, pages 422-432. IEEE Computer Society.
- [27] Snodgrass, R. T., editor. (1995). *The TSQL2 Temporal Query Language*. Kluwer.
- [28] Tello, J., Mazzi, M., Tansella, M., Bonizzato, P., Jones, J., Amaddeo, F. (2005). Does socioeconomic status affect the use of community-based psychiatric services? A south Verona case register study, *Acta Psychiatrica Scandinavica*, 112 (3) 215-223.
- [29] Tello, J. E., Jones, J., Bonizzato, P., Mazzi, M., Amaddeo, F., Tansella, M. (2005). A census-based socio-economic status (ses) index as a tool to examine the relationship between mental health services use and deprivation. *Social Science & Medicine*, 61(10)2096-2105.
- [30] Viqueira, J. R. R., Lorentzos, N. A. (2007). SQL extension for spatio-temporal data, *VLDB J.*, 16 (2) 179 - 200.
- [31] Worboys, M. F. (1998). Imprecision in nite resolution spatial data. *GeoInformatica*, 2 (3) 257-279.
- [32] World Health Organization (WHO). *In: International classification of diseases (ICD)*. URL: <http://www.who.int/classifications/icd/en/>.
- [33] Zulian, G., Donisi, V., Secco, G., Pertile, R., Tansella, M., Amaddeo, F. (2010). How are caseload and service utilisation of psychiatric services influenced by distance? A geographical approach to the study of community-based mental health services. *Social Psychiatry and Psychiatric Epidemiology*, p. 1 - 11.