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To cite this version:
Benjamin Harbelot, Helbert Arenas, Christophe Cruz. Continuum: A spatiotemporal data model to represent and qualify filiation relationships. 4th ACM SIGSPATIAL International Workshop on GeoStreaming (IWGS) 2013, Oct 2013, Orlando, United States. hal-00869861

HAL Id: hal-00869861
https://hal.archives-ouvertes.fr/hal-00869861
Submitted on 4 Oct 2013

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Continuum: A spatiotemporal data model to represent and qualify filiation relationships

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ABSTRACT
This work introduces an ontology-based spatio-temporal data model to represent entities evolving in space and time. A dynamic phenomenon generates a complex relationship network between the entities involved in the process. At the abstract level, the relationships can be identity or topological filiations. The existence of an identity filiation depends on whether the object changes its identity or not. On the other hand, topological filiations are based exclusively on the spatial component, like in the case of growth, reduction, merging or splitting. When combining identity and topological filiations, six filiation relationships are obtained, forming a second abstract level. Upper-level filiation relationships provide better semantic vocabulary to describe the modeled phenomena, thus allowing the implementation of spatial, temporal and identity constraints. In this paper, we present a method based on identity and topological filiation relationships, to improve the capabilities of standard knowledge bases using Semantic Web technologies. Our method enables us to check the consistency of spatio-temporal and semantic data. An example is given in the field of urban growth to show the capabilities of the model.

Keywords
Spatio-temporal modeling, reasoning, semantic, filiation, spatio-temporal evolution, integrity constraints, spatial dynamics

Categories and Subject Descriptors
I.2.4 [ARTIFICIAL INTELLIGENCE]: Knowledge Representation Formalisms and Methods – Semantic networks.
I.2.3 [ARTIFICIAL INTELLIGENCE]: Deduction and Theorem Proving – Inference engines.
H.2.8 [DATABASE MANAGEMENT]: Database Applications – Spatial databases and GIS.

General Terms
Algorithms, Management, Verification.

1. INTRODUCTION
Several spatial and temporal approaches have been proposed to model urban and environmental processes. These approaches allow researchers to study the past and predict trends in the future by addressing the evolution of spatial entities or objects. Considering entities evolving in time, the notions of identity and change have led to the development of several conceptual models. Events and, processes generate complex relationships networks in space and time. There is a need to store a semantic representation of this evolution in the field of Geoinformation Science. A dynamic phenomenon can be defined as a set of entities evolving under the action of a process.

The modeling of entities varying in space and time is a challenge that has received much attention in the community of spatial information. Objects can move or change shape while maintaining their identity, for example, when a city changes its borders. Some objects may change giving rise to new objects, for instance when a land parcel is divided into several sub-parcels. A major concern is the identity of an object through time [1,2]. The problem is to determine the limits of the object identity through the time, in other words, answer the question: “To what extent can an entity vary before losing its identity?” [3,4,5].

During their existence, the objects maintain relations of various kinds between them. From a spatial perspective, models of topology as 9-intersection [6] or Region Connection Calculus [7] form the basics of what is known in the literature as qualitative spatial reasoning. By analyzing the topological spatial relationships between the spatial representation of objects co-existing at the same time or at different times, we can better understand the semantics of the relationships between objects. This analysis allows a better understanding of the evolution experienced by an object. An object can continue to exist, originate one or more new objects, and cease to exist. Qualitative spatial reasoning, based on topology, can help determine filiation relationships and help to identify the exact kind of evolution that is occurring.

Since the emergence of GIS, the need to represent real world dynamics, encouraged the scientific community to add a temporal dimension to GIS. Spatio-temporal dynamics generate a large amount of information that must be represented, stored and analyzed. Most of the models proposed in the literature of temporal GIS are based on relational databases [8,12,3,4,5,9]. While databases provide good capacity of storage and time query response through SQL language, their ability to handle the knowledge stored on the modeled objects is less obvious compared to Semantic Web technologies. The Semantic Web increases the capacity to represent knowledge through classes, relations and properties. In addition, the Semantic Web is widely recognized for its reasoning capabilities allowing to check the consistency of ontologies or to infer new knowledge from existing ones. Because of the previously described reasons the Semantic Web has gradually emerged in the field of GIS [10,11,12,13,14].

Many languages have been proposed to examine spatio-temporal semantic models. Faced with the heterogeneity of these languages, Open Geospatial Consortium (OGC) has
developed the standard GeoSPARQL to represent and query geospatial semantic web data. GeoSPARQL consists in a small ontology that adds the spatial dimension to spatial ontologies. It extends the SPARQL language by providing a set of additional function for handling geometries.

Even though temporal ontologies have been defined [15,16], there are still elements missing so as to allow an accurate depiction of evolving entities. To solve these limitations the concept of fluent has been introduced, allowing the change of the value of the object’s properties over time [17,18,19].

The work presented in this paper introduces a spatio-temporal ontology-based model called the Continuum Model. The Continuum Model tracks the evolution of entities through the time. It combines the spatial functions provided by GeoSPARQL with the temporal capabilities of the fluent ontology. The Continuum Model is additionally enhanced with the definition of filiation relationships that allow to qualify the evolution experimented by the entities. Based on filiation relationships defined in [1], the model exploits Semantic Web capabilities defining a filiation relationship hierarchy. By expressing integrity constraints using filiation relationships, it is possible to check consistency in the ontology with respect to the reality modeled. On the other hand, constraints allow inferring filiation relationships or better qualify them with regard to the defined hierarchy. Finally, based on semantic constraints provided in the ontology, the system is able to infer automatically some application dependent upper-level filiation relationships of the modeled field of study.

The paper is organized as follows: Section 2 introduces some basic concepts in order to develop our modeling approach. Section 3 presents spatial and temporal tools available in the Semantic Web literature. Section 4 presents and formalizes the Continuum Model. Section 5 presents integrity constraints for each filiation relationship and presents check-query and inference-query based on the constraints. Section 6 gives an example to switch from general filiation relationships to upper-level filiation relationships on the basis of semantic constraints. Finally, we present our conclusions in section 7.

2. FROM SPATIOTEMPORAL MODELING TO EVOLUTION

Some basic notions of spatio-temporal modeling have to be introduced in order to develop our approach.

2.1 Spatiotemporal entities

A representation of entities of the real world should comprise an identity that describes fixed semantics as well as a dynamic part that represents thematic and spatial properties that can change through time. Another special kind of evolution concerns the identity of the entity. In the literature, we denote two main kinds of spatio-temporal entities: 1) moving object, for example a taxi driving in a town, and 2) changing object, for example, an administrative unit whose borders change over time. In this paper our focus is on the latter.

2.2 Representing the time

The representation of temporal properties of information is a well-researched topic. In [20], the authors present an overview of the conceptual modeling of time. In our work, we consider a discrete, linearly-ordered time domain with a focus on absolute time. In our model, time is incorporated by associating intervals with class instances in the ontology. Intervals denote the time during which all the relations and properties of an instance are valid. Each interval is defined by start and end instants. The temporal granularity is application dependent, therefor the time units (e.g. Hours, Days, Weeks, Months, Seasons, Years) must be specified for each instant of time in the ontology.

2.3 Representing the topology

From the human point of view, when we observe a landscape, objects such as forests, buildings, rivers, etc. are “seen” or “perceived” in their context. The concept of neighborhood is implicit. For example, a river runs “through” a parcel, a building “is located in” agglomeration, an agglomeration is “juxtaposed” in a larger settlement. This kind of observation of geographic space has led to the study of topology. From the geographical point of view, the topology is the set of perceived relationships that allow us to situate an object in relation to others. In the literature, several models [6,7] have been proposed to study the topology of a geographical space. In all cases, we get an equivalent set of 8 spatial topological predicates (equals, disjoint, intersects, touches, crosses, overlaps, within, contains).

From a temporal point of view, topology also exists and has been introduced by [21] in order to represent actions and events occurring over time. Allen offers a representation based on temporal intervals. An interval is delimited by defining a start and a later end instant of time. Thus, Allen describes 13 mutually exclusive relationships between time intervals. The first six relations are before, meets, overlaps, during, starts and finishes have reverse, respectively, after, met-by, overlapped-by, contains, started by and finished-by. The last relationships is equals which does not have an inverse.

2.4 The nature of changing

Changes can occur on the position and shape of an object. In addition, knowledge about an object can evolve when changes occur on thematic attributes of objects. These space and semantic changes can occur simultaneously or independently from each other, giving rise to 8 possible different scenarios [22] as depicted in figure 1.

![Figure 1: Eights different scenarios of change](image)

The change can be continuous or discrete. We can classify models based on the type of changes they are able to support. In our model, we focus on shape and attribute changes. Moreover, we highlight a particular type of attribute change: identity modification.

2.5 Identity

The identity is a unique feature that distinguishes one object from another, which differentiates it from other properties, values or structures. Objects can change their position and shape while keeping their identity. In other cases new entities may be formed from ancient ones. A major issue then arises: “How far can an entity vary before losing its identity?”. Sometimes due to modeling constraints it may be more appropriate to destroy an object and to create a new one due to a large number of changes. The identity is a key point to prove the existence or non-existence of an object as well as to list the similarities or the differences between objects. The
existence refers to the physical presence of an object. For the conceptual objects, the existence is the perception of an object. There are generally three states of identity [4]: creation, continuation and elimination. There are previous research that identify more states of identity as in [5] which defines nine, but it is worth noting that in certain contexts, some of these states of identity seem to be impossible or contradictory to a domain ontology and therefore are not considered.

2.6 Filiation Relationships
Filiation relation defines the succession link that exists between different representations of the same object at different moments of time. This relationship is essential to maintain the identity of an entity that evolves. It is also necessary to identify children entities that might have been created due to some evolution. Filiation relation is intimately linked to the notion of identity. The establishment of a filiation relation induces a dependency in the identity. There are then two general types of filiation relations: continuation and derivation [1,2,3,4,5]. In the first case, identity is retained. The entity continues to exist but has experienced a change. While in the former, a new entity is created from another after being subjected to an evolution. Unlike the relations of continuation, derivation relations may concern several entities at the same time.

3. AVAILABLE TOOLS FOR SPATIO-TEMPORAL AND SEMANTIC DEFINITION
Several languages and ontologies have been proposed in order to enable the Geospatial Semantic Web [23]. Among these we can find GeoSPARQL, an OGC standard designed to provide the foundation for spatial reasoning and manage spatial data in RDF. In the time domain, there exists a proposed ontology to represent time, but the definition of time does not allow changing objects, which has led to the emergence of fluent.

3.1 Handling of spatial properties in the Semantic Web
GeoSPARQL is an emerging standard within the Open Geospatial Consortium (OGC). The main purposes of GeoSPARQL are: 1) provide a standard way to express and query spatial objects in RDF. 2) allow users to exchange data easily, and 3) provide a standard spatial indexing for triple stores.

3.1.1 Overview
Since it became an OGC standard, GeoSPARQL [24] has several arguments to be selected as a powerful language in geospatial application. Several spatially-enable SPARQL endpoints like Strabon¹, Parliament² or OpenSahara³ implements GeoSPARQL. Another strength of GeoSPARQL is to be built on existing standards. In fact, it is built on W3C Semantic Web standards like RDF, OWL and SPARQL but also on OGC standards like Simple Features and Spatial Relations. As a result, GeoSPARQL provides a standardized vocabulary for representing linked geodata and for writing SPARQL queries against geospatial RDF data. It also reuses common geometry serialization formats like GML, KML and WKT stored as strings and encoded as RDF Literals. Besides, it provides a structured vocabulary and semantics for geographic features and relationships. Finally, GeoSPARQL provides the ability to answer queries involving geographic features and relationships.

3.1.2 Enabling the spatial dimension
GeoSPARQL includes a small spatial ontology in RDFS/OWL for representation of spatial entities. There are three main classes in the GeoSPARQL ontology:

- geo:Feature : something which can have a spatial location such as a park or monument, etc.
- geo:Geometry : a representation of the location space, i.e., a set of coordinates
- geo:SpatialObject : the superclass of “Feature” and “Geometry”.

The geo:hasGeometry property links the “Features” (a thing) to their spatial representation. GeoSPARQL allows to associate multiple geometries to an object. The resource associated to the geometry then has an RDF literal representation, to which property is linked, named after the type of representation. For example, the geo:asWKT property connects geometry resource with a wktLiteral. The literal contains the geometry information in the specified format.

![Figure 2: the GeoSPARQL ontology](image)

GeoSPARQL implements a list of spatial concepts described in OGC/ISO Simple Features (ISO 19125), such as Point, Line, Polygon, among others, which could be specialized or generalized in a geometry concept hierarchies. It also provides metadata for each spatial object. Metadata concerns elements such as dimension, SRID (Spatial Reference system Identifier). Additionally GeoSPARQL supports topological relationships based on DE-9IM.

3.1.3 Query data with GeoSPARQL
GeoSPARQL provides a SPARQL query interface using a set of topological SPARQL extension functions for quantitative reasoning. GeoSPARQL covers different types, spatial properties, operations and relationships in order to express spatial queries. Thus, we can find descriptive datatype properties (e.g., dimension), topological relations (e.g., touches, intersects, contains) which can be computed with RCC8, 9-intersection, or Simple Feature. Furthermore, GeoSPARQL provides parameterized relations (e.g., within distance) and operations that produces new objects (e.g., buffer, union, intersect). All these types of spatial capabilities have been encoded into extensible filter functions.

3.2 The temporal component in the Semantic Web
One of the goals of the Semantic Web is to treat the temporal aspects of entities and perform reasoning with this information. However it is common to find ontologies representing the world as static, while in reality, the world is constantly evolving. In this section we describe previous research on the semantic representation of the temporal nature of objects.

3.2.1 Ontology of time
Several ontologies have been proposed in the literature in order to represent time. The most famous is OWL-Time. OWL-Time is an ontology of time designed for describing the temporal content of web pages and the temporal properties of

³ [https://dev.opensahara.com/](https://dev.opensahara.com/)
web services [15]. OWL-Time represents both instants and intervals of time. Properties are defined with specific time intervals. It supports the 13 Allen relationships to specify relations between intervals. Ontology OWL-Time also allows the use of time units such as seconds, minutes, hours, days, etc. Finally OWL-Time is the only known ontology to enable temporal aggregates, i.e. to represent expressions such as “every first Tuesday of the week”. OWL-Time consists of a main class named TemporalEntity, which has two subclasses Instant and Interval. Begins and ends properties have been defined to specify the beginning and end of a temporal entity. Other properties have been defined in order to enrich the relations between temporal entities: inside relation applies on instants and allows to define the latter as part of an interval. The class Interval has a subclass ProperInterval that corresponds to a range whose start and end times are not equal. Another ontology of time used to represent temporal information in OWL based on intervals of time is SWRL Temporal Ontology [16]. It represents time similar to OWL-Time, except that it does not allow to establish a topological relationship between instants or time intervals. However, SWRL Temporal Ontology offers some temporal built-ins for SWRL, a rule-based language. These built-ins allow checking of Allen relationships between different temporal entities when querying.

3.2.2 Ontology of Fluent
Ontologies of time are very similar in the way they represent time. But representing time is not sufficient to represent the evolution of an object. In order to do this we need upper-level temporal ontology based on ontologies of time. In [17], the authors introduce ontology of fluent to represent the fact that the properties of objects vary over time. The approach, called 4D-fluent, considers objects as having temporal parts, called timeslices, and the representation of change in the properties of an object is done through the use of fluent, which are properties that are valid only during a certain time interval. When the property of an object changes, a new timeslice is established, holding the new property of the object. The 4D-fluent approach was intended to model perdurant objects which are objects having several temporal parts representing them during a certain interval of time. An entity is defined by its temporal parts: timeslices. The properties that are valid only during a certain time interval are called fluent. In the 4D-Fluent approach, the used TimeInterval class corresponds to the Interval class defined in OWL-Time. Considering that the time interval associated with the timeslice represents the period during which the associated fluent is valid. If a timeslice is associated with many fluent then these fluent must be valid on the same interval of time.

A more recent development is TOWL (Time-determined Ontology Web Language) [18,19]. It is a language to extend the OWL language with a temporal dimension allowing the representation of time, change and state transitions. TOWL proposes a four layer architecture in order to expand the capabilities of OWL-DL as depicted in the figure 3.

The first layer of TOWL OWL-DL does not require further explanation. The second layer entitled Concrete Domains allows the representation of intervals and supports the 13 Allen relationships between intervals. The next layer is used to represent the time in a quantitative manner. Finally, the last layer uses the idea of perdurant objects using timeslices and fluent from the 4D-fluent approach.

However, some differences exist between 4D-fluent and TOWL. First, in TOWL, the timeslices are not necessarily associated with a time interval but can also be associated with an instant of time. When defining a fluent between two timeslices, the 4D-fluent approach requires the two timeslices to be associated with the same interval of time. With TOWL, the timeslices can have their own time interval, however, when a fluent is defined between two timeslices the equal relationship must be defined in order to ensure that these timeslices are valid on the same interval of time.

Another contribution of TOWL is to specify type of fluent. While with 4D-fluent, we had only a fluentProperty property, TOWL distinguishes between the fluentObjectProperty and the fluentDatatypeProperty. A sub-property of fluentObjectProperty connects two timeslices while a fluentDatatypeProperty property connects a timeslice to an object of type rdf:Literal and thus indicates changes in values.

3.3 Discussion
GeoSPARQL is an efficient standard that enables geospatial Semantic Web by providing a small reusable spatial ontology. GeoSPARQL is implemented in efficient triplestores allowing topology based calculations as well as other geospatial operations. However the temporal aspect is still excluded from this standard. To represent time some well-known ontologies such as OWL-Time and SWRL Temporal Ontology have been proposed in the literature. Representing time differs from representing evolution. That’s why upper-level ontology of time called ontology of fluent has recently emerged. But this ontology of fluent is rarely used in the field of spatio-temporal modeling. A notable exception is [25], however, this research does not implement filiation relationships in order to track entities over time and understand the kind of evolution that objects might have suffered. Additionally, the query language proposed is not a standard and suffers from lack of tools implementing it.

4. THE CONTINUUM MODEL
In this section, we present our spatio-temporal model inspired in GeoSPARQL and ontology of fluent in order to represent entities over time.

4.1 Overview
The Continuum Model extends GeoSPARQL allowing it to represent spatio-temporal dynamic objects. This extension is achieved by combining the GeoSPARQL Ontology with an ontology of fluent. Therefore, The feature class corresponding to entities with spatial component is new associated with timeslices. In this way we are able to create multiple representations of the object, each, corresponding to different instants or time intervals. Unlike GeoSPARQL, the hasGeometry relationship is no longer established between the feature and the geometry class but rather between a timeslice and a geometry allowing an object of the class feature to change its geometry (See Figure 4).

As in a fluent approach, the timeslice represents the state of an object during a certain interval or instant of time. When a property or relationship changes a new timeslice is created.

In the case of a relationship between two different objects, a relationship is valid only between timeslices of both objects.
coexisting at the same time. Therefore, this interval must be the same for each of the timeslices of objects affected by this relationship.

The design of our model has been influenced by [1,2]. In addition to hasGeometry and relations hasTime respectively for linking geometry and an interval of time to a timeslice, the model presented in this report implements the relationships 1) hasSpatialRelation, 2) hasFiliation. Figure 4 depicts the classes and relations implemented in our model.

![Image](image-url)

**Figure 4: The core architecture of our proposition**

The relationship hasSpatialRelation allows expressing the spatial topology between objects and must be established between two timeslices of different objects valid on the same interval or instant of time. Finally, our model focuses on the hasFiliation relationship. This relation is established between two timeslices, it allows to track and define the evolution of an object in the case of a continuation, and to track and define the transformation of one object to another in the case of a derivation. This relationship is designed to translate the semantic of the transition between two consecutive timeslices. This relationship can be specified more precisely combining identity-filiation and topological-filiation. Our model exploits the capabilities of specialization inherent in Semantic Web.

Figure 5 depicts the hierarchy of filiation-relation defined in the model. Details about each relationship are given in the section “Filiation relation integrity constraints”.

![Image](image-url)

**Figure 5: different level of filiation relationships**

### 4.2 OWL DL Continuum Model definition

To present our model, we will adhere to the Karlsruhe Ontology Model [26]. An OWL DL ontology is a structure \(O := (SO, \sigmaO, FO)\) consisting of:

- The underlying set \(SO\) containing:
  - Six disjoint sets \(sC\), \(sT, sR, sA, sI, sV\), \(sK_R\) and \(sK_A\) called concepts, datatypes, relations, attributes, instances, data values, relation characteristics (among Symmetric, Functional, Inverse Functional, Transitive) and attribute characteristics (Functional).
  - Four partial orders \(\leq_C\), \(\leq_T\), \(\leq_R\), \(\leq_A\) respectively on \(sC\) called concept hierarchy or taxonomy, on \(sT\) called type hierarchy, on \(sR\) called relation hierarchy and on \(sA\) called attribute hierarchy.
  - The signature \(\sigmaO\) containing:
    - Two functions \(\sigma_C : sR \rightarrow sC^2\) called relation signature and \(\sigma_A : sA \rightarrow sC \times sT\) called attribute signature, such that \(\sigmaO := \{\sigma_R, \sigma_A\}\).
    - The interpretation function \(FO\) containing the functions:
      - \(\iota_C : sC \rightarrow 2^{sC}\) called concept instantiation,
      - \(\iota_T : sT \rightarrow 2^{sT}\) called data type instantiation,
      - \(\iota_R : sR \rightarrow 2^{sR\times sR}\) called relation instantiation,
      - \(\iota_A : sA \rightarrow 2^{sA}\) called attribute instantiation,
      - \(\iota_K_R : sR \rightarrow 2^{sK_R}\) called relation characterization,
      - \(\iota_K_A : sR \rightarrow 2^{sK_A}\) called attribute characterization,
      - \(\iota_K_S : sR \rightarrow 2^{sK_S}\) called relation inverse specification.

- \(sC = \{\)TopConcept, SpatialObject, Feature, Geometry, TimeSlice, Time, TimeInterval, TimeInstant\},
- \(sT = \{\)TopType, dateType, String\},
- \(\leq_C = \{\)TopConcept, SpatialObject, Feature, (TopConcept, TimeSlice), (TopConcept, Geometry), (TopConcept, Time), (SpatialObject, Feature), (SpatialObject, Feature), (Time, TimeInterval), (Time, TimeInstant)\},
- \(\leq_T = \{\)TopType, Float\} \((\)TopType, String)\},
- \(sR = \{\)TopRelation, hasFiliation, hasSpatialRelation, hasTemporalRelation, isTimeSliceOf, hasGeometry, hasTime\}\(^4\), as WKT, hasDisjoint, hasOverlaps, hasTaches, hasContains, hasWithin, hasEqual, hasContinuation, hasDerivation, hasExpansion, hasContraction, hasSplits, hasFusion, hasDerivAnnexation, hasContiSeparation, hasDerivAnnexation, hasContiAnnexation before, meets, overlaps, during, starts, finishes, after, metBy, overlappedBy, contains, startedBy, finishedBy, Equal\},
- \(sA = \{\)TopAttribute, hasBeginInstant, hasEndInstant, hasTimeInstant\},
- \(\sigma_C = \{\)hasFiliation(TimeSlice, TimeSlice), hasSpatialRelation(TimeSlice, TimeSlice), isTimeSliceOf(TimeSlice, Feature), hasGeometry(TimeSlice, Geometry), hasTime(TimeSlice, Time), hasDisjoint(TimeSlice, TimeSlice), hasOverlaps(TimeSlice, TimeSlice), hasTouches(TimeSlice, TimeSlice), hasWithin(TimeSlice, TimeSlice), isWithin(TimeSlice, TimeSlice), hasDisjoint( Geometry, Geometry), hasOverlaps( Geometry, Geometry), hasTouches( Geometry, Geometry), (contains( Geometry, Geometry), (within( Geometry, Geometry), (equal( Geometry, Geometry)), (hasContinuation( TimeSlice, TimeSlice)), (hasDerivation( TimeSlice, TimeSlice)), (hasExpansion( TimeSlice, TimeSlice)), (hasContraction( TimeSlice, TimeSlice)), (hasSplits( TimeSlice, TimeSlice)), hasFusion( TimeSlice, TimeSlice), \)

\(^4\) Relations in bold represent the core of our model. More explanations are given in the next section.
\[ T^*_C \times \mathbb{P} \subseteq \Delta^1 \]

Time Intervals:

In the Continuum Model, \( t_0 \) is defined by the datatype property \( \text{hasBeginInstant} \) and \( t_f \) is defined by the datatype property \( \text{hasEndInstant} \). The spatial representation of the timeslice of an object is given through its geometry \( G \). The semantic component of the timeslice of an object is represented by S. It describes the nature of the entities and can be made up of one or more alphanumeric properties. Finally, a timeslice has an identity held by the property \( \text{isTimesliceOf} \) which connects it to an object \( O \). Each timeslice \( (TS) \) in the model has four components: 1) a time interval \( T^*_C \), 2) a geometry \( G \), 3) an identity \( O \) and 4) a semantic component \( S \) representing all other potential properties associated to a timeslice. We define all these properties using \( TS \) base symbol, as defined in [28], which stands for the qualities that distinguish a timeslice from another apart from its interval of existence, identity, and its geometry (note that \( \Delta S \equiv TS \) ).

In the Continuum Model a change on the spatial representation or on the semantic component generates a new timeslice which has a filiation relationship with the original timeslice, additionally we know that the time interval of the parent timeslice meets the time interval of the child timeslice. The filiation relationship between timeslice \( ts_1 \) and \( ts_2 \) is defined by the relationships between their spatial representations \( ts_{1g} \) and \( ts_{2g} \), their semantic definitions \( ts_{1s} \) and \( ts_{2s} \), their identity \( ts_{1i} \) and \( ts_{2i} \) and their time intervals \( t_{s1} \) and \( t_{s2} \). A filiation relationship is defined when a change occurs on the geometry, the semantic component or the identity.

5.1 Filiation relation integrity constraints

In this section we use a Tarski-style specification to describe the model filiation-relations. To represent time intervals we follow the semantics suggested in [27]. We can think of the temporal domain as a linear structure \( T \) composed by a set of temporal points \( P \). The components of \( P \) follow a strict order \( \prec \), which forces all points between two temporal points \( t_1 \) and \( t_2 \) to be ordered. By selecting a pair \( \{t_1; t_2\} \) we can limit a closed interval of ordered points. The set of interval structures in \( T \) is represented by \( T^*_C \).

Temporal Points:

\[ P \ni P' \subseteq \Delta^1 \]

We can further specialize the hasFiliation to define more complex relationships:

\[ \text{hasContinuation} : \text{hasFiliation} \]

In this case a change may occur only on the geometry or the semantic component but the identity is maintained.

\[ \text{hasContinuation(TS)} \ni \{ts_{1g} \in TS \mid \{ts_{1s} \in S, ts_{1i} \in O, ts_{i1} \subseteq ts_{1g} \} \}

\[ \text{hasDerivation} : \text{hasFiliation} \]

In this case the change may occur only on the geometry or the semantic component, while the identity must be different.
∀hasDerivation.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasDerivation → ts₂ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₁ ≠ ts₂) ∧ (meets(ts₁,ts₂))]  

hasExpansion: In this relationship the entity continues to exist but the geometry grows.

∀hasExpansion.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasExpansion → ts₂ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₁ ≠ ts₂) ∧ (meets(ts₁,ts₂) ∧ hasWithin(ts₂, ts₃))]  

hasContraction: In this case there is a reduction in the geometry size.

∀hasContraction.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasContraction → ts₂ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₁ ≠ ts₂) ∧ (meets(ts₁, ts₃) ∧ hasContains(ts₂, ts₄))]  

hasSplits: In this relationship, the parent entity ceases existing. While its geometry is divided, generating two new geometries corresponding each to a new entity. The union of the new geometry is equal to the former geometry.

∀hasSplits.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasSplits → ts₂ ∈ TS ∧ ts₁ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₂ ≠ ts₃) ∧ (ts₁ ≠ ts₃) ∧ (ts₂ ≠ ts₃) ∧ (ts₃ ≠ ts₄) ∧ (ts₂ ≠ ts₄) ∧ (ts₁ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (meets(ts₂, ts₃)) ∧ (meets(ts₃, ts₄)) ∧ (equal(ts₂, ts₃) ∧ hasEqual(ts₃, ts₄))]  

hasSeparation: In this case the parent entity continues existing, however its geometry originates a new geometry corresponding to a new entity. A “hasSeparation” relationship is similar to a “hasSplits” relationship with the difference that in hasSeparation the original entity remains.

∀hasSeparation.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasSeparation → ts₂ ∈ TS ∧ ts₁ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₂ ≠ ts₃) ∧ (ts₃ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (ts₂ ≠ ts₃) ∧ (ts₁ ≠ ts₃) ∧ (ts₂ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (ts₁ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (meets(ts₂, ts₃)) ∧ (meets(ts₃, ts₄)) ∧ (equal(ts₂, ts₃) ∧ hasEqual(ts₃, ts₄))]  

hasFusion: In this relationship the two parent entities merged and cease to exist to give rise to a new geometry corresponding to a new entity. Inverse to a “hasSplits” relationship. The resulting geometry is equal to the union of the former geometries.

∀hasFusion.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasFusion → ts₁ ∈ TS ∧ ts₂ ∈ TS ∧ (ts₁ ≠ ts₂) ∧ (ts₂ ≠ ts₃) ∧ (ts₁ ≠ ts₃) ∧ (ts₂ ≠ ts₄) ∧ (ts₁ ≠ ts₄) ∧ (ts₂ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (ts₂ ≠ ts₄) ∧ (ts₁ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ (meets(ts₁, ts₂) ∧ meets(ts₂, ts₃)) ∧ (equal(ts₁, ts₂) ∧ hasEqual(ts₂, ts₃))]  

hasAnnexation: In this case the two parent entities merge but the resulting entity keeps the identity of one of its parents.

∀hasAnnexation.TS [ ts₁ ∈ TS | ∀ts₂.(ts₁,ts₂) ∈ hasAnnexation → ts₂ ∈ TS ∧ ts₁ ∈ TS ∧ (ts₂ ≠ ts₁) ∧ (ts₂ ≠ ts₃) ∧ (ts₁ ≠ ts₃) ∧ (ts₁ ≠ ts₄) ∧ (ts₂ ≠ ts₄) ∧ (ts₁ ≠ ts₂) ∧ (ts₁ ≠ ts₃) ∧ (ts₁ ≠ ts₄) ∧ (ts₂ ≠ ts₃) ∧ (ts₂ ≠ ts₄) ∧ (ts₃ ≠ ts₄) ∧ meets(ts₁, ts₃) ∧ meets(ts₂, ts₄) ∧ (equal(ts₁, ts₃) ∧ hasEqual(ts₂, ts₄))]  

5 In our ontology, we define two separation relations which are hasComsSeparation and hasDervisSeparation in order to distinguish easily entities which continues existing and entities ceasing to exist when a separation occurs.

6 As in separation-filiation, we provide two annexation relations which are hasComAnnexation and hasDervisAnnexation.
The approach is naive because several check-queries have some common constraints. To check all filiation relationships, we run each of the check-queries. Thus, some constraints may be verified several times during this process. An optimized approach can be considered grouping some constraints as in [1].

5.3Inferring filiation relationship and knowledge

As we have seen, it is possible to check the consistency of the spatio-temporal ontology based on relations already defined. But sometimes, information may be incomplete and alter the semantic and understanding of the spatio-temporal phenomena modeled. In this section, on the basis of integrity constraints, we show how to infer filiation relation using SPARQL Update.

5.3.1Inferring basic filiation relationships

The transformation from a SPARQL check-query to SPARQL Update inference-query is trivial, so we show only one example of inference (see ).

Example 6: Inference-query for expansion relation

```
INSERT { ?ts1 ex:hasExpansion ?ts2 . }
WHERE { ?ts1 ex:hasExpansion ?ts2 . }  
```

All the filiation relationships discussed in the previous sections form the basis of spatio-temporal qualification of evolution. If these relations are standardized for all applications dealing with change on a geographical space, the semantic brought by these relations may be increased using contextual information defined in the ontology.

5.3.2 Increasing semantic of filiation relationships with contextual information

So far, filiation relationships of expansion, contraction, splits, separation, fusion and annexation were based on spatial, temporal and identity constraints. The Continuum Model uses ontology capabilities to define upper-level definitions based on filiation relationships. These upper-level relationships provide a better understanding of the evolution adding semantic constraints to previous filiation relationships. For example, the Dissolution of Czechoslovakia, which took effect on 1 January 1993, was an event that saw the self-determined division of the federal state of Czechoslovakia. The Czech Republic and Slovakia, entities which in 1960 within the framework of the Czechoslovak federalization, became immediate subjects of international law in 1993. Considering the representation of the phenomenon, we note that the state “Czechoslovakia” ceases to exist and its geometry splits into two new geometries, each corresponding to a new state: the “Czech Republic” and “Slovakia”. Finally, the union of the new geometry is equal to the former geometry. Thus, we notice that the phenomenon corresponds to a “split” filiation relationship. Consequently, it is possible to infer automatically that involving the “hasSplits” filiation relationships occurring between entities whose type is “State” is a “Dissolution” process.
6. EXAMPLE

For this example, we will use our model to study the urban evolution of the city of New Orleans, Louisiana. Figure 6 depicts the urban evolution of the entity “city of New Orleans” with timeslice, time interval and filiation relationships.

![Figure 6: Time frame of urban evolution](image)

Until 1870, Greenville, Jefferson, Lafayette and New Orleans cities grew, each having a hasExpansion relationship with its previous timeslice. In 1870, Greenville, Lafayette and Jefferson cities are absorbed by the city of New Orleans. This phenomenon is called a conurbation. We note that the parent entities merged, but one continues to exist meaning that the corresponding filiation relationship is hasAnnexation. Thus, we consider that an annexation occurring between cities is a conurbation. The Continuum model distinguishes the relation hasContiAnnexation meaning that the entity continues to exist after the process. The model also implements the relation hasDerivAnnexation implying that the entity ceases to exist after the annexation process. In this example, we will infer that a hasContiAnnexation relation between entities whose type is “city” leads to a isConurbation (Example 8: Inference-query for hasConurbation relation) upper-level filiation relationship. A hasDerivAnnexation relation between entities whose type is “city” leads to a hasConurbation (Example 8: Inference-query for hasConurbation relation) upper-level filiation relationships.

Example 7: Inference-query for isConurbation relation

```sun
INSERT {?ts1 ex:isConurbation ?ts2 .

?o1 rdf:type ex:City .
?o2 rdf:type ex:City .
}
```

Example 8: Inference-query for hasConurbation relation

```sun
INSERT {?ts1 ex:hasConurbation ?ts2 .

?ts1 ex:hasDerivAnnexation ?ts2 .
?o1 rdf:type ex:City .
?o2 rdf:type ex:City .
}
```

7. CONCLUSION

The research presented in this paper introduces the Continuum Model whose goal is to track the identity of spatio-temporal entities through time and capture the semantics of the modeled phenomena. The model is ontology-based and provides a solution that combines GeoSPARQL and ontology of fluent.

Additionally, the model focuses on filiation relationships. On the basis of filiation relationships presented in [1], we defined a hierarchy of these relations. Basic filiation relationships are subjected to integrity constraints affecting spatiality, temporality and identity.

The model makes it possible to check consistency in the ontology with respect to the modeled phenomena. It can happen that data sets are partially incomplete. Using the integrity constraints model, it is possible to infer filiation relationships or qualify them more specifically with respect to the filiation hierarchy defined in the Continuum Model.

Semantic depicted by basic filiation relationships may be increased to improve understanding of the phenomenon modeled. Therefore, we proposed adding semantic constraints which are application dependent to obtain upper-level filiation relationships. An example of semantics constraints is introduced in order to infer some upper-level filiation relationship on the basis of general ones.

8. ACKNOWLEDGMENTS

This research is supported by: 1) Conseil regional de Bourgogne. 2) Direction Générale de l’Armement, see: http://www.defense.gouv.fr/dga/. We would like to thank the cooperation of Perrine Pittet who helped us in the formalization of the Continuum Model.

9. REFERENCES


