Qualitatively Correct Bintrees With Roles

Leif Harald Karlsen\textsuperscript{1} and Martin Giese\textsuperscript{2}

\textsuperscript{1} Department for Informatics, University of Oslo, Norway
leifhka@ifi.uio.no
\textsuperscript{2} Department for Informatics, University of Oslo, Norway
martingi@ifi.uio.no

Abstract
In this paper we extend previous work on using bintrees as an efficient representation for qualitative information about spatial objects. Our approach represents each spatial object as a bintree satisfying the exact same qualitative relationships to other bintree representations as the corresponding spatial objects. We prove that such correct bintrees always exist and that they can be constructed as a sum of local representations, allowing a practically efficient construction. Our representation is both efficient, w.r.t. storage space and query time, and can represent many well-known qualitative relations, such as the relations in the Region Connection Calculus and Allen’s Interval Algebra.

1998 ACM Subject Classification I.2.4 Knowledge Representation Formalisms and Methods

Keywords and phrases Qualitative spatial data, Bintree, Data structure

Digital Object Identifier 10.4230/LIPIcs...

1 Introduction

Spatial and temporal data types are ubiquitous in today’s software, with a growing number of spatially aware devices gathering and publishing data. Spatial and temporal data are used in a great number of highly valuable applications, such as route planning, automatic navigation, and modeling of physical processes. However, temporal and especially geospatial data are normally represented as complex numerical objects. The relationships between objects are implicit in these numerical representations, and advanced algorithms (e.g. from computational geometry) are needed to determine them, and indexing these objects for efficient query answering is complex. During the last decades, several geospatial and temporal database systems have been developed, featuring advanced indexing mechanisms and efficient numerical algorithms for answering queries over these data types (see e.g. [10, 16, 11]). Despite these advances, geospatial and temporal data are still significantly more difficult to handle than more traditional types of data. Spatial data types also often lag behind when new knowledge representations are introduced and in many cases need special treatment.

The present work stems from the following observations:

(i) Many applications of geospatial data are mostly concerned with qualitative relations like overlaps or containment of geometries, rather than quantitative properties like distance, area, etc.

(ii) For such qualitative applications, resorting to expensive computations on the numerical representations of geometries for each query seems wasteful. It would be sufficient to store a (pre-computed) database table for overlap, containment, or any other relations of interest, treating these relations like any other in a RDBMS. But this, on the other hand also seems wasteful, in terms of space, since such tables could be quadratic in the number of geometries (for binary relations, cubic for ternary and so on), despite obvious
redundancies, like e.g. the transitivity of the containment relation or symmetry of the overlaps relation.

(iii) Numerical representations of spatial objects are subject to precision errors. E.g. even though two objects are touching in the real world, their numerical representations might not, due to insufficient precision in their numerical representation. These errors are difficult or sometimes even impossible to fix numerically without introducing other errors[2]. However, if we construct a new qualitative representation of the objects, we can fix such errors during the translation by using domain knowledge about the objects. For instance, we might know that every country $c$ touches all countries $c'$ whenever their numerical representations have a smallest distance of 1 kilometer, or that the spatial extent of any capital of a country is contained in the extent of that country.

(iv) Most approaches to qualitative spatial representation are either targeted at complex reasoning tasks and therefore do not scale to large real-world datasets, or focus on a particular set of relations for a particular type of spatial data.

Our approach is to construct a linear bintree-representation for each spatial object that are qualitatively correct w.r.t. any given set of relations definable from a given first-order language. This representation scales to real-world datasets without limiting the approach to any fixed set of relations.

The linear bintree [19, 20] consists of a set of bit-strings, each representing a small chunk of space obtained by recursively dividing space. Bintrees have the convenient property that they can be stored as a regular database relation. Furthermore, the bintrees can themselves be indexed by normal database index structures, like B-trees, since they only consist of sets of bit-strings where each bit-string can be represented by one integer. Another desirable feature of bintrees is that they allow variable resolution, so we can have low resolution (few and short bit-strings) for homogeneous areas and high resolution (many and long bit-strings) for heterogeneous areas where more detail is necessary. The bintree has previously been used as an indexing structure for geometries and as an efficient representation for images. Bintrees are now considered obsolete as index structures for geometries, as R-trees[6] and their variants (see e.g. [14] for an overview) have better performance. However, for our purpose of representing qualitative information, we will see that the bintree is a good fit.

The concrete problem this paper addresses is the following: Given a set of objects with a spatial interpretation and a set of qualitative relations, construct a bintree representation that returns the same answers to queries with the given relations over the spatial objects. We have previously constructed both theory [8] and an implementation [9] for constructing such qualitatively correct bintrees, with promising results. However, our previous work has been restricted to the construction of bintree-representations that are correct only w.r.t. part-of and overlaps relationships. In this paper we will extend the theoretical foundation to allow for representations that are correct with respect to a more expressive set of relations.

The paper is outlined as follows: In Section 2 we introduce the spatial objects we work with and the key notions and results needed for expressing and constructing correct bintrees; in Section 3 we explain how to construct correct bintrees and why this is a good representation for qualitative information; in the Sections 4 and 5 we extend the expressiveness of the relational language our bintrees are correct w.r.t. in two directions; in Section 6 we show several examples of common qualitative relations that our bintrees can represent; Section 7 discusses related work and Section 8 concludes the paper.
2 Spatial Objects and Correct Bintrees

We will start by introducing the central concept of spaces, the elements which we aim to represent correctly.

Definition 1. A space lattice \( S = (S, \prec_S, \top_S, \bot_S) \) is a bounded, distributive lattice with top element \( \top_S \) and bottom element \( \bot_S \). We will let \( \otimes_S \) and \( \oplus_S \) be the induced meet and join operators respectively, and call the elements of \( S \) spaces. We will let \( S^+ := S \setminus \{ \bot_S \} \).

Typical examples of such space lattices is the lattice of geometrical objects (polygons, lines points) where \( \prec_S \) is geometric containment, the lattice of temporal intervals where \( \prec_S \) is temporal containment, the lattice of sets where \( \prec_S \) is the subset-relation, and so on. Thus, the goal of this paper is to construct efficient representations of the qualitative relationships between such structures. In order to do this, we need to compute the relationships between the spatial objects, however checking all possible relationships between all possible spaces would be very complex, as this has a complexity of \( O(n^k) \) (for \( n \) elements with relations of arity \( k \)). A property of qualitative relationships like overlaps and contains is that they are local, that is, they depend only on the spatial parts inside the elements, and nothing more. Thus, we want to exploit this locality in a similar fashion as the bucket sort-algorithm does, where the elements to sort are first distributed into a set of of buckets/intervals partitioning the universe. The buckets are sorted individually, before being gathered into a sorted list.

In a similar fashion, we will construct a set of chunks of space partitioning the space-lattice’s universe \( \top_S \), and construct locally correct representations in each chunk. We will call such a chunk a block. The blocks are most naturally construed in a recursive fashion where we start with \( \top_S \) and recursively split blocks into two smaller blocks, until we reach some desired property (e.g. the desired resolution or the desired number of spatial objects overlapping each block.) This splitting forms a binary tree, so each block can be represented as the path from the root (\( \top_S \)) down to that block. Furthermore, such a path can be compactly represented as a bit-string (a 0-bit and 1-bit denotes a left-edge and a right-edge resp.) Note that every bit-string denotes a chunk of space, and that if \( s \) is a bit-string which is a prefix of \( s' \), then the block denoted by \( s \) spatially contains the block denoted by \( s' \). If we let a set of bit-strings denote their union, we can represent more complex spaces that can spatially overlap and contain other spaces.

Our representation should allow efficient updates, and since relationships are locally determined, inserting a new object into our representation should only affect the representation of the blocks overlapping the object to insert. However, for such a local insert to be possible we need to know which block each representation was constructed in.

Therefore, it would seem natural to let each element’s local block-representation be a set of bit-strings, each contained in that block, which satisfies the same qualitative relationships as the spatial objects they represent. A set of such bit-strings is in fact a linear bintree. The bintree is thus a binary trie data structure, similar to the quadtree and octree. For a discussion and comparison of these three structures, see e.g. [19]. Below follows the formal definition of both bit-strings and bintrees.

Definition 2. Let \( \mathbb{B} \) be the set of bit-strings with \( \varepsilon \) being the empty bit-string and \( b \circ b' \) to be the concatenation of the bit-strings \( b \) and \( b' \). Let \( b_1 \circ b_2 \Leftrightarrow \exists b \in \mathbb{B}(b_2 \circ b = b_1) \) and \( b_1 \sim b_2 \Leftrightarrow \exists b \in \mathbb{B}(b_1 = b \circ 0 \land b_2 = b \circ 1) \). Define a block-set \( B \) to be a non-empty, finite set of bit-strings such that if \( b \in B \) then \( B \) also contains all \( b' \in \mathbb{B} \) such that either \( b \sim b' \) or \( b \circ b' \).

Definition 3. Define the \( T \)-lattice \( T := (T, \prec_T, \top_T, \bot_T) \) where \( T = \mathcal{P}_{fin}(\mathbb{B}) \) is the set of bintrees, (where \( \mathcal{P}_{fin} \) is the finite powerset) such that \( t \in T \) contains no two distinct
Figure 1 The left figure shows the bit-string representation of some blocks and the right figure the spatial extent (in gray) of the bintree \{110,0110,00110\} in 2D.

![Bit-string and Spatial Extent](image)

Elements \(b_1, b_2\) where either \(b_1 \prec b_2\) or \(b_1 \sim b_2\). Furthermore, let \(\top_T = \{\epsilon\}\), \(\bot_T = \emptyset\), and \(t \prec t' \iff \forall b \in t \exists b' \in (t \prec b')\)

It should be easy to see that the \(T\)-lattice is a space lattice. Thus, bintrees behave similarly to spaces, which allows them to be used as representations for spaces. In Figure 1 we can see an example of both blocks and a bintree, and their spatial extent (assuming regular splitting in each space division). Note that we put no restriction on the number of dimensions our spaces has, and the same holds for our bintrees. In the same way we alternate between splitting along the x- and y-axis in the 2D case, we would cycle through all \(k\) dimensions in a \(k\)-dimensional space. We will now introduce our models, which will allow us to precisely define correctness of bintree-representations.

**Definition 4.** Given a space lattice \(S = (S, \prec_S, \top_S, \bot_S)\), a finite set of constants \(C\) and a block-set \(B\), an \(S\)-model \(M\) is a first order model over the similarity type \((\prec; C \cup B)\) with universe \(S\), \(\prec\) is an atomic spatial sentence such that \(\prec\) is a well-formed formula (BNF) defined by the BNF \(\varphi := \psi | \neg \varphi | \varphi_1 \land \varphi_2\), where \(\psi\) is an atomic spatial formula.

Given a set of constants \(C\) and a block-set \(B\), an (atomic) spatial sentence \(\varphi(\vec{c})\) is a first order sentence such that \(\varphi(\vec{x})\) is an (atomic) spatial formula and \(\vec{c} \in (B \cup C)^{|\vec{c}|}\).

**Definition 5.** Given a set of spatial formulae \(\Gamma\), we will say that a \(T\)-model \(M_T\) is \(\Gamma\)-correct w.r.t. an \(S\)-model \(M_S\) if \(M_T \models \varphi(\vec{c}) \iff M_S \models \varphi(\vec{c})\) where \(\vec{c} \in (B \cup C)^{|\vec{c}|}\) and \(\varphi(\vec{c}) \in \Gamma\).

**Example 7.** Let \(\Gamma := \{x \prec y, \exists^+ z (z \prec x_1 \land z \prec x_2)\}\). In Figure 2 we can see an example of a \(\Gamma\)-incorrect and a \(\Gamma\)-correct bintree model for the constants \(A, B, C\) \(\cup \{b \in B \mid |b| \leq 4\}\) (where \(|b|\) is the length of the bit-string \(b\), w.r.t. a geometric model \(M_G\). \(M_T\) is just an approximation from above, which is how bintrees are normally used as index structures. We
We will now formalize the fact that our qualitative relations are locally determined, that is, for any spatial formula \( \varphi(\bar{c}) \in \Gamma \) and any \( \bar{c} \in (C \cup B)^{|\bar{c}|} \), but it is not sound, i.e. the converse implication does not necessarily hold. For instance, \( M_G \models \varphi \) but \( M_G \not\models \varphi \). As stated above, for efficiency reasons we will construct our bintrees locally. Thus, we need a proof.

Let \( M_T \) be the set of all constants \( c \) with \( \bar{c} \in S \) such that \( c \in C \cup B \). By definition the \( M_T \)-interpretation of the elements of \( \beta \) forms a partition on \( T_S \), so \( T_S = \bigoplus_{b \in \beta} b^{M_S} \) and \( b_1^{M_S} \otimes_S b_2^{M_S} = \bot_S \). This, together with distributivity, we know that \( c_1^{M_S} \prec_S c_2^{M_S} \) is equivalent to \( \forall b \in \beta (b^{M_S} \otimes_S c_1^{M_S}) \prec_S (b^{M_S} \otimes_S c_2^{M_S}) \) for any \( c_1, c_2 \in C \cup B \). By similarly, we have that \( \exists z \in S^+ (\land_{1 \leq i \leq k} z \prec_S c_i^{M_S}) \) is equivalent to the local \( \exists z \in S^+ (\land_{1 \leq i \leq k} z \prec_S c_i^{M_S}) \) for any \( c_1, \ldots, c_k \in C \cup B \). ▶

**Theorem 10.** For any \( S \)-model \( M_S \) there exists a locally \( \Gamma \)-correct bintree-model \( M_T \).

**Proof.** Proof done by model construction: For each \( b \in \beta \), construct the set of all locally true atomic spatial sentences occurring (either positively or negatively) in some \( \varphi \in \Gamma \):

\[
T_b := \left\{ \psi(\bar{c}) \mid M_S \models_b \psi(\bar{c}), \ \land_{1 \leq i \leq |\bar{c}|} \left( c_i \in C \cup B \land c_i^{M_S} \otimes b^{M_S} \neq \bot_S \right) \right\}
\]

Any \( T \)-model of all \( T_b \)s is locally \( \Gamma \)-correct. Then, let \( T_b' \) be the skolemization of \( T_b \), and \( T_b'' \) be the set of all constants occurring in any sentence in \( T_b' \). Define \( K_b'' \) to be the set of all constants
occurring in $T^w_{\beta}$, and $K^b_{\beta} := (C \cup B) \setminus K^b_{\beta}$. So $\{K^b_{\beta}, K^+_{\beta}\}$ partitions $C \cup B$ and $K^b_{\beta}$ is the set of constants that should have a non-empty interpretation, locally in $b$.

We will now construct the $\mathcal{T}$-model. First, for each $b \in \beta$, generate a set $W_b \subseteq B$ of size $|K^b_{\beta}|$ of pairwise $\prec$-unrelated bit-strings $b'$ such that $b' \prec b$. Then, let $w_b : K^b_{\beta} \to W_b \cup \{b\}$ be a bijective function on $K^b_{\beta} \setminus B$ and $w_b(b') = b$ for $b' \in K^b_{\beta} \cap B$. Then, define $I_b(c) := \bigoplus_{\mathcal{T}} \{w_b(c') \mid (c' \prec c) \in T^w_{\beta}, c' \in K^b_{\beta}\}$ for each $c \in K^b_{\beta}$, and $I_b(c) := \bot_{\mathcal{T}}$ for $c \in K^b_{\beta}$. So $I_b$ is the locally correct interpretation of the constants in $C \cup B$, and it should be clear that $(c^{M_{\mathcal{S}}} \otimes_S b^{M_{\mathcal{S}}}) \prec_S (c'^{M_{\mathcal{S}}} \otimes_S b^{M_{\mathcal{S}}})$ if and only if $I_b(c_1) \prec_{\mathcal{T}} I_b(c_2)$ and $\exists z \in S^+ \left(\bigwedge_{1 \leq k} z \prec_S (b^{M_{\mathcal{S}}} \otimes_S c^{M_{\mathcal{S}}})\right)$ if and only if $\exists z \in T^+ \left(\bigwedge_{1 \leq k} z \prec_{\mathcal{T}} I_b(c_i)\right)$ for any $c_1, \ldots, c_k \in C \cup B$ and any $b \in \beta$. Finally, let $c^{M_{\mathcal{T}}} := \bigoplus_{\mathcal{T}} \{I_b(c) \mid b \in \beta\}$ for each $c \in C \cup B$. $\mathcal{M}_{\mathcal{T}}$ is now a $\Gamma$-correct $\mathcal{T}$-model w.r.t $\mathcal{M}_{\mathcal{S}}$. □

3 How To and Why Construct Correct Bintrees

The proof of Theorem 10 illustrates how one could design an algorithm for construction of correct bintree-models. We can write an almost direct translation of the steps in the proof to an algorithm. That is, for each $b \in \beta$ do the following: Find all $c \in C \cup B$ overlapping $b$ and compute their $\Gamma$-relationships, $T^w_b$; skolemize and extract the atomic sentences, $T^w_{\beta}$; generate a set of blocks and assign each non-empty element a block, and propagate according to the $\prec$-relationships in $T^w_{\beta}$; finally, sum up the local representations to form the model.

The algorithmic complexity of such a model construction is $|\beta|$ times the complexities of first constructing $T^w_{\beta}$, and then generating and distributing the elements of $W_b$. It should be easy to see that the latter has complexity $O(|T^w_{\beta}|)$. Note that constructing $T^w_b$ from $T_{\beta}$ and $T^w_{\beta}$ from $T^w_{\beta}$ are both linear in the size of $T_b$. Lastly, we have that constructing $T_b$ requires computing whether $\mathcal{M}_{\mathcal{S}} \models \varphi(c)$ holds for each atomic spatial sentence generated from the atomic spatial formulas of $\Gamma$ and the constants $B \cup C$. This gives us a total complexity of $O(|\beta| \cdot o^k)$, where $o = \max_{c \in \beta} |\{c \in C \cup B \mid \mathcal{M}_{\mathcal{S}} \models \varphi(c, b)\}|$, that is, the largest number of elements from $C \cup B$ that overlaps any $b \in \beta$, and $k$ is the largest number of free variables occurring in any atomic spatial formula occurring in any $\varphi \in \Gamma$. This means that we in practice can construct correct bintree-models for any $\mathcal{S}$-model, however, why still remains to be answered. Below we discuss the main properties of the representation making it suitable for representing qualitative information.

The bintrees can be stored and queried in a relational database as a binary relation $(id, block)$, where we encode the bit-strings as integers and where both the IDs and the bit-string integers can be indexed by a normal B-tree. This allows for highly efficient query answering. In [9] we discuss this representation in more detail and present a benchmark that show that overlaps and containment queries are on average 2.7 times faster over our correct bintrees than over the corresponding geometries. The comparison was done with Postgis[16], a state of the art geospatial database, over real-world datasets where the largest sets has over a million geometries.

Note that our constructions also allow a more efficient insertion than reconstructing the entire model upon each insert: Assume we already have constructed a $\Gamma$-correct model $\mathcal{M}_{\mathcal{T}}$ for the $\mathcal{S}$-model $\mathcal{M}_{\mathcal{S}}$ and constants $C \cup B$, but now want to construct a $\Gamma$-correct model for the extended model $\mathcal{M}_{\mathcal{S}}'_{\mathcal{T}}$ for $C \cup C' \cup B$. Since we only need local $\Gamma$-correctness, we only need to update $I_b$ for each $b \in \beta$ where $\mathcal{M}_{\mathcal{S}}_{\mathcal{T}} \models \varphi(c, b)$ for any $c \in C'$. Thus, a larger $B$ gives a more efficient insert-operation as we have a higher resolution. Observe also that the only requirements we put on the interpretations of the elements of $B$, is that $\{b \circ 0, b \circ 1\}$ partitions $b$. Thus, we are free to interpret $b \circ 0$ and $b \circ 1$ in such a way that there is approximately the
We have now seen that we can construct a correct bintree-representation for any space lattice. As we will see shortly, we only need roles that consist of a single name to express the relations without holes. Observe also that we always have $\pi$ where $(\ldots)$ will now see that a small extension to our bintree representations allows us to accommodate $\delta$ and $t$ for interior and $\triangledown$ as an intersection of all the $\pi$ roles can relate the different types of parts $i$ of RCC8, namely $\mathcal{R}$, which allows us to talk about the different parts of a space.

### Definition 11.
A role is a set of names. A role-set is a set of roles containing $\emptyset$.

As we will see shortly, we only need roles that consist of a single name to express the relations of RCC8, namely $i$ for interior and $b$ for boundary. However, we will also see examples where using multiple names to denote a part is useful.

### Definition 12.
Given a role-set $\mathcal{R}$, an $R$-roled space lattice $\mathcal{S}$ is a tuple $(\mathcal{S}, \prec, \top, \bot, \pi_{\mathcal{S}})$ where $(\mathcal{S}, \prec, \top, \bot, \pi_{\mathcal{S}})$ is a space lattice and $\pi_{\mathcal{S}} : \mathcal{R} \times \mathcal{S} \to \mathcal{S}$ is a function where $\pi_{\mathcal{S}}(\emptyset, s) = s$ and $\pi_{\mathcal{S}}(r \cup u, s) = \pi_{\mathcal{S}}(r, s) \oplus s \pi_{\mathcal{S}}(u, s)$ for any $(r, u) \in \mathcal{R}^2$ such that $r \cup u \in \mathcal{R}$ and any $s \in \mathcal{S}$.

The reader can read $\pi_{\mathcal{S}}(r, u)$ as “$u$’s $r$-part”. Intuitively one can think of a $\{n_1, \ldots, n_k\}$-part as an intersection of all the $\{n_i\}$-parts. For instance the role $\{i, h\}$, where $h$ is short for hole, denotes holes in an interior, whereas $\{i\}$ denotes all of the interior, both with and without holes. Observe also that we always have $\pi_{\mathcal{S}}(r, s) \prec_{\mathcal{S}} s$. We will now introduce the corresponding bintrees.

### Definition 13.
Given a role-set $\mathcal{R}$, an $R$-roled block is a pair $(r, b)$ such that $b \in \mathcal{B}$ and $r \in \mathcal{R}$. Let $\mathcal{B}_\mathcal{R}$ be the set of $R$-roled blocks. Also let $\delta(r, t) := \{b \mid (r, b) \in t\}$ for any $r \in \mathcal{R}$ and $t \subseteq \mathcal{B}_\mathcal{R}$ and let $\Sigma_{\mathcal{R}}(t) := \bigoplus_{r \in \mathcal{R}} \delta(r, t)$.

An $R$-roled bintree $t$ is an element of $\mathcal{P}_{\mathcal{B}_\mathcal{R}}(\mathcal{B}_\mathcal{R})$, such that for any role $r \in \mathcal{R}$ we have that $\delta(r, t)$ is a bintree, and $\delta(r, t) \odot \tau \delta(u, t) = \bot \tau$ for any $r \neq u$. Let $T_{\mathcal{R}}$ be the set of $R$-roled
bintrees. Furthermore, let $\top_n := \{(\emptyset, \varepsilon)\}$, $\bot_n := \emptyset$, $\pi_n(t, r) := \{(u, b) \in t \mid r \subseteq u\}$, $t \prec_n t' \iff \Sigma_n(t) \prec_n \Sigma_n(t')$ and $\mathcal{T}_n := (\mathcal{T}_R, \prec_n, \top_n, \bot_n, \pi_n)$.

While the different roles for the parts are implicitly defined for spaces like geometries, (such as being the interior of a polygon), we explicitly state the roles each block should have in the bintree. So the boundary of a bintree $t$, $\pi_t(\{b\}, t)$, is the set of blocks having a role $r$ such that $b \in r$. We can then define the touching relations as $\mathrm{ov}(\pi_t(x), \pi_t(y))$. So even though two bintrees seem to touch geometrically (e.g. if one has a block $b$ and the other a block $b'$ and $b \sim b'$) they will not necessarily touch according to our definition. This makes it easier for us to construct correct bintree-models, as we still only have to care about overlaps and part-of relationships. Note also that it is possible to construct bintree-models that satisfy sentences that are unsatisfiable by any $\mathcal{S}$-model for a particular space lattice $\mathcal{S}$. For instance, it is easy to make a bintree model with two objects that have a partially overlapping interior, but that have disjoint boundaries, which is impossible for any geometrical model. Thus, we cannot use our representation for reasoning (that is, make a representation for a set of sentences and then query for all entailments). However, as our bintrees only function as a representation of the relationships of a given $\mathcal{S}$-model and is constructed to satisfy exactly these, this is not a problem.

Definition 14. Given an $R$-roled space lattice $\mathcal{S}$, a set of constants $c$, and a block-set $B$, an $R$-roled $\mathcal{S}$-model $\mathcal{M}$ is a first order model over the similarity type $\langle \prec; \pi; C \cup B \rangle$, where $\pi$ is a family of unary function symbols $\pi_r$ for each $r \in R$, that is an $\mathcal{S}$-model over $\langle \prec; C \cup B \rangle$ and where $\pi_r(c)_{\mathcal{M}} := \pi_S(r, c_{\mathcal{M}})$ for any $r \in R$ and $c \in C \cup B$.

Definition 15. Given a role-set $R$, an $R$-roled atomic spatial formula is a first order formula on one of the forms $\pi_{r_1}(x) \prec \pi_{r_2}(y)$ or $\exists \bar{z} \left( \bigwedge_{1 \leq k \leq n} \pi_{r_k}(x_k) \right)$ for some $r_1, \ldots, r_n \in R$. Let $R$-roled formulae and $R$-roled (atomic) spatial sentences be defined analogously as in Definition 5, but where $\psi$ is an $R$-roled (atomic) spatial formula.

Note that $(\pi_g(s))_{\mathcal{M}_g} = s$ for any $R$-roled $\mathcal{S}$-model $\mathcal{M}_g$, so we sometimes write $x$ instead of $\pi_g(x)$ in the definitions of spatial formulae. To save ink, let $\bar{r} = \{r\}$ for any role-name $r$.

Example 16 (RCC8). Assume we have the names $b$ for “boundary”, and $i$ for “interior”, where $\pi_b(x)$ denotes $x$’s boundary and $\pi_i(x)$ denotes $x$’s interior, we can now express the RCC8-relations with $\Gamma$ equal to the set of formulae:

\[
\begin{align*}
DJ(x, y) := & \neg \mathrm{ov}(x, y) & EC(x, y) := & \mathrm{ov}(x, y) \land \neg \mathrm{ov}(\pi_i(x), \pi_i(y)) \\
PO(x, y) := & \mathrm{ov}(\pi_b(x), \pi_i(y)) \land (x \not\prec y) \land (y \not\prec x) & EQ(x, y) := & x \prec y \land y \prec x \\
TPP(x, y) := & x \prec y \land \mathrm{ov}(\pi_b(x), \pi_b(y)) \land (y \not\prec x) & NTPP(x, y) := & x \prec y \land \neg \mathrm{ov}(\pi_b(x), \pi_b(y))
\end{align*}
\]

Let $G$ be set of two-dimensional geometries (i.e. (multi)polygons, (multi)lines, (multi)points) contained in some universe $\top_G$ and with $\prec_G$ being geometric containment, then $G$ is a space
lattice. So a Γ-correct TR-model w.r.t. G-models M will correctly represent all RCC8-relations between the elements of C as interpreted by M. In Figure 3 we see an example of a correct bintree-model with respect to the RCC8-relations.

**Theorem 17.** Any R-roled TR-model M_T is Γ-correct if and only if it is locally Γ-correct, w.r.t. an R-roled S-model M_S. Furthermore, for any R-roled S-model M_S there exists a locally Γ-correct R-roled TR-model M_T.

**Proof.** The arguments for the first part are analogous to the proof of Theorem 9, just substitute c_i with π_r(c_i).

The second part is done by a similar model construction as for the proof of Theorem 10. So, construct T'''_b in the same way, but note that now the elements of T'''_b are on the forms π_t(c_1) ∼ π_u(c_2) and v ∼ π_r(c). Now, define K_b^+ to be the set of expressions (v and π_r(c) where v is a skolem-constant and c ∈ C ∪ B) occurring in T'''_b, and K_b^- as before. Let W_b be a set of size |K_b^+| of pairwise ⊼ unrelated bit-strings b' where b' ~ b. Then, let w_b : K^+_b → W_b ∪ {b} be a function such that assigns a unique element from W_b to each c ∈ K_b^+ \ \{π_b(b') | b' ∈ B\}, and w_b(e) := b for each c ∈ K_b^+ ∩ \{π_b(b') | b' ∈ B\}. We then define I_b(π_r(c)) := ⊕_T_b \{(r, w_b(e)) | (e ∼ π_r(c)) ∈ T'''_b, e ∈ K_b^+\} for each π_r(c) ∈ K_b^+ and I_b(e) := ⊗_T_b for e ∈ K_b^-.

It should now be clear that (b^M_S ⊗_S (π_r(c_1))^M_S) ⊼_S (b^M_S ⊗_S (π_u(c_2))^M_S) if and only if I_b(π_r(c_1)) ⊼_T I_b(π_u(c_2)) and ∃z ∈ S^+ (A_{i≤k} z ∼_S (b^M_S ⊗_S (π_r(c_i))^M_S)) if and only if ∃z ∈ T'''_b (A_{i≤k} z ∼_T I_b(π_r(c_i))) for any c_1, . . . , c_k ∈ C ∪ B and any b ∈ β. Finally, we let c^M_T := b ∈ B ⊕_r∈R I_b(π_r(c)) for each c ∈ C ∪ B. M_T is now an R-roled TR-model that satisfies exactly the true spatial sentences of M_S generated from Γ and C ∪ B.

From the above proof, we can see that the construction of correct roled bintrees is done in a similar fashion as the normal bintrees, and we only need a minor update of any algorithm used for constructing normal correct bintrees.

Observe also that we can compress our roled bintrees in the following manner: Assume that in the set of skolemized atoms T'''_b we have a π_r(c) such that (π_r(c) ∼ e_1) ∈ T'''_b ⇔ (π_u(c) ∼ e_1) ∈ T'''_b and (e_2 ∼ π_r(c)) ∈ T'''_b ⇔ (e_2 ∼ π_u(c)) ∈ T'''_b for any expressions e_1, e_2 ∈ K_b^+. If then there is no formula φ(⃗x) ∈ Γ such that eq(π_r(c), π_u(c)) ⇔ φ(⃗c), we can let w_b(r, c) = w_b(u, c), thus reducing the size of our bintree-representation. This can for instance be done for the RCC8-relations (letting eq(π_r(c), π_u(c)) cannot introduce a new relationship, if π_r(c) and π_u(c) has the exact same relationships to other elements).

Note also that any role can be represented as a fixed length bit-string by enumerating all role-names occurring in Γ and represent each role r as the bit-string having 1s at the bit-positions corresponding to the numbers given to the role-names in r, and 0 everywhere else. We can then represent our roled bintrees as ternary relations (id, block, role) where each column can be indexed by a normal B-tree. Thus, querying roled bintrees is almost as efficient as querying our normal bintrees.

### 5 Extension: Order

Introducing roles allows us to construct much richer bintree-models. However, having only the part-of relations allows only relations based on sharing of different types of parts, we are still unable to describe many interesting qualitative relationships, such as temporal relationships, relative size and relative direction. In this section we will extend our language to also include a different type of partial order which will enable us to express these relationships.
Definition 18. An ordered $R$-roled space lattice $S$ is a tuple $(S, <_S, ⊑_S, ⊢_S, \pi_S)$ where $(S, <_S, ⊑_S, ⊢_S)$ is an $R$-roled space lattice and $<_S$ is a strict partial order such that if $a <_S b$ then $a \otimes_S b = ⊢_S$ and for any pair $c, d \in S$ we have $c <_S a \land d <_S b \rightarrow c <_S d$.

The reader can read the statement $x < y$ as “$x$ is before $y$”. The rest of the definitions are analogous to before:

Definition 19. Let $t <_T t' \Leftrightarrow \forall b \in t \forall b' \in t'(b <_B b')$ where $b <_B b'$ for bit-strings $b, b'$ iff there exists some $b''$ such that $b <_B b'' \circ 0$ and $b' <_B b'' \circ 1$. Then let $t <_T t' \Leftrightarrow \Sigma_R(t) <_S \Sigma_R(t')$ and $T^R_R := (T_R, <_T, ⊑_T, ⊢_T, \pi_T)$.

Definition 20. Given an ordered $R$-roled space lattice $S$, a set of constants $C$, and a block-set $B$, an ordered $R$-roled $S$-model $M$ is a first order model over the similarity type $(\prec, <_S; C \cup B)$ that is an $R$-roled $S$-model over $(\prec; \pi; C \cup B)$, and where $<_S$ and $(b \circ 0)^{M_S} <_S (b \circ 1)^{M_S}$ for any $(b \circ 0), (b \circ 1) \in B$.

Definition 21. Let an atomic ordered $R$-roled spatial formula be a first order formula that is either an atomic $R$-roled spatial formula or a formula on the form $x < y$. Let ordered $R$-roled spatial formulae and (atomic) ordered $R$-roled spatial sentences be defined analogously as $R$-roled spatial formulae and (atomic) $R$-roled spatial sentences, but where each $\psi$ is an atomic ordered $R$-roled spatial formula.

Example 22 (Allen’s Interval Algebra). Assume we have the role-names $i$ for interior, $f$ for first, $l$ for last, and the role-set $R := \{\emptyset, i, f, l\}$. Let $\pi_j(x)$ denote the interval consisting of only $x$’s first point, and $\pi_l(x)$ denote the interval consisting of only $x$’s last point, and $\pi_i(x)$ is the interior of $x$’s interval. We can then express the relations of Allen’s Interval Algebra:\[
\begin{align*}
\text{before}(x,y) & := \pi_i(x) < \pi_j(y) & \text{meets}(x,y) & := \pi_l(x), \pi_j(y) \\
\text{overlaps}(x,y) & := \pi_l(x), \pi_j(y) \land (x \neq y) \land (y \neq x) & \text{equal}(x,y) & := \pi_l(x), \pi_l(y) \\
\text{starts}(x,y) & := \pi_l(x), \pi_j(y) \land \pi_l(x) < \pi_j(y) & \text{during}(x,y) & := \pi_l(x), \pi_j(y) \land \pi_l(x) < \pi_j(y) \land \pi_l(y) < \pi_j(y) \\
\text{ends}(x,y) & := \pi_l(x), \pi_j(y) \land \pi_l(x) < \pi_j(y) & \text{after}(x,y) & := \pi_l(y) < \pi_l(x) \land \pi_l(y) < \pi_j(x)
\end{align*}
\]

Given the set $I$ of time intervals contained in some universe $\top_S$, with $\prec_S$ being temporal containment, and $x <_I y$ is the temporal before, it should be obvious that this forms an ordered $R$-roled space lattice. Thus, any correct $T^<_R$-model w.r.t. such an $I$-model $M$ will correctly represent all Allen’s Interval-relations between the elements of $C \cup B$ as $M$.

Theorem 23. Any ordered $R$-roled $T^<_R$-model $M_T$ is $\Gamma$-correct if and only if it is locally $\Gamma$-correct, w.r.t. an ordered $R$-roled $S$-model $M_S$. Furthermore, for any ordered $R$-roled $S$-model $M_S$ there exists a locally $\Gamma$-correct ordered $R$-roled $T^<_R$-model $M_T$.

Proof. For the first part, note that, since $\beta$ is a partition of $\top_S$ and $(b \circ 0) <_S (b \circ 1)$ for any $(b \circ 0), (b \circ 1) \in B$, we have that $<_S$ is a total order on $\beta$. This implies that $e_1^{M_S} <_{<_S} e_2^{M_S}$ if and only if $\forall b \in \beta \left( (e_1^{M_S} \otimes_S b^{M_S}) <_S (e_2^{M_S} \otimes_S b^{M_S}) \right)$. The rest of the proof is analogous to the proof of Theorem 9.

For the second part, we again have to construct a locally correct model. So, construct $T^{w_0} b$ in the same way as before for each $b \in \beta$, but this time the elements of $T^{w_0} b$ can also be on the form $e_1 < e_2$. We define $w_0$ in the same way, but we assign each expression an element of $W_0$ such that $(e_1 < e_2) \in T^{w_0} b$ implies $w_0(e_1) <_{_B} w_0(e_2)$. We then define both $I_S(e)$ and $M_T$ in the same way as before. Now, $M_T$ is an ordered $R$-roled model that satisfies exactly the same ordered $R$-roled spatial sentences generated from $\Gamma$ as $M_S$. ▶

Again we see that the construction of correct bitrees with order requires only a small extension to the previous algorithm. Furthermore, a nice feature of encoding bit-strings as
integers as described in Section 3 is that the \(<\)-ordering of the blocks corresponds to the normal \(<\)-ordering on their integer representations, thus we can reuse the B-tree index over the blocks to efficiently answer \(<\)-queries as well.

6. Expressiveness and More Examples

- **Example 24 (Holes).** To both \(G\) and \(I\) we can add an additional role-name, \(h\), for “hole”, that can be combined with e.g. \(i\) to represent holes in the interior of a polygon or interval, or with \(b\) to represent geometries that have an open boundary. We can now express:

\[
\text{surroundedBy}(x, y) := x < \pi_{(h, i)}(y) \quad \text{hasHoles}(x) := \exists^+ z(z < \pi_b(x))
\]
\[
\text{hasOpenBoundary}(x) := \pi_{(i, h)}(x) < \pi_{(h, i)}(x) \quad \text{hasHole}(x, y) := eq(\pi_{(i, h)}(x), y)
\]

- **Example 25 (Relative size and direction).** One dimensional attributes like size, length, projection down to the north-south and east-west axis can easily be represented by introducing an appropriate role-name, e.g. \(d\), and let \(\pi_S(d, x) < \pi_S(d, y)\) hold if \(x\) has a smaller value than \(y\) on the \(d\)-axis. If we then also let for each \(b \in \beta\), \(\pi_S(d, b)\) be an interval along this axis such that \(\beta\) contains both the smallest and largest values, our constructing algorithm will be a normal bucket-sort with \(\beta\) being the set of buckets.

If we introduce the role-names \(n\) for the projection along the north-south and \(e\) for the projection down to the east-west, we can express the following relations from the Cardinal Direction Calculus[12], e.g.:

\[
\text{northOf}(x, y) := \pi_{(h, n)}(y) < \pi_{(h, n)}(x) \wedge ov(\pi_{(h, e)}(x), \pi_{(h, e)}(y))
\]
\[
\text{northEastOf}(x, y) := \pi_{(h, n)}(y) < \pi_{(h, n)}(x) \wedge ov(\pi_{(h, e)}(x), \pi_{(h, e)}(y)) \wedge \pi_{(h, e)}(x) < \pi_{(h, e)}(y)
\]

and the rest of the directional-relations are defined similarly. Note that \(\pi_S(\bar{n}, x)\) and \(\pi_S(\bar{e}, x)\) is the projection of a two-dimensional object down to each dimension. We can of course also do this for three-dimensional (or higher) objects and introduce a role-name, \(u\) for the up-down axis, and relations such as \(\text{above}(x, y)\) and \(\text{between}(x, y, z)\). If we combine the directional roles with the interior-role, e.g. \(\{i, u\}\), we can express

\[
\text{onTopOf}(x, y) := ov(\pi_{(i, n)}(x), \pi_{(i, n)}(y)) \wedge ov(\pi_{(i, e)}(x), \pi_{(i, e)}(y)) \wedge
\]
\[
\pi_{(i, e)}(x) < \pi_{(i, e)}(y) \wedge \pi_{(i, n)}(x) < \pi_{(i, n)}(y)
\]

that is, \(x\) and \(y\) overlap in the two-dimensional plane, but \(x\) and \(y\) are touching along the up-down axis, yet \(x\)‘s interior is above \(y\)’s.

- **Example 26 (Orientation).** If we have the directional roles \(\{n, e\}\) as described above, we can introduce two more role-names \(f\) for front and \(b\) for back, and then introduce orientational relations, e.g. \(\text{northOriented}(x) := \pi_{(n, f)}(x) < \pi_{(n, f)}(x) \wedge ov(\pi_{(e, b)}(x), \pi_{(e, b)}(x))\) and similarly for the rest of the directions. If we allow unions of relations in our query language (this is trivial in SQL), we can express relative orientation, that is, \(\text{orientedTowards}(x, y)\) as the union of the 8 relations on the form \(\text{northOf}(x, y) \land \text{southOriented}(x)\).

- **Example 27 (Egg-Yolk).** If we have a space-lattice \(S\) with indeterminate boundaries (that is, an inner and outer boundary where the real boundary is somewhere in between) we can introduce two new role-names \(y\), for yolk, and \(w\) for white, and let \(\pi_S(y, s)\) be the region within the inner boundary and \(\pi_S(w, s)\) be the region within the outer boundary. We can then introduce all the 46 relations from the Egg-Yolk RCC5 calculus[5], e.g.:

\[
R_2(x, y) := PO'(\pi_{(y)}(x), \pi_{(y)}(y)) \land \neg ov(\pi_{(y)}(x), y) \land \neg ov(\pi_{(y)}(y), x)
\]
\[
R_{11}(x, y) := PO'(\pi_{(y)}(x), \pi_{(y)}(y)) \land PO'(\pi_{(y)}(x), \pi_{(y)}(y)) \land (\pi_{(y)}(x) < \pi_{(y)}(y)) \land \neg ov(\pi_{(y)}(x), \pi_{(y)}(y))
\]
where $PO'(x, y) := ov(x, y) \land (x \neq y) \land (y \neq x)$. $R_2(x, y)$ states that the white of the two partially overlap whereas the yolks are disjoint from each other’s eggs, and $R_{11}(x, y)$ that $x$’s white partially overlap both $y$’s white and yolk, and $x$’s yolk is contained in $y$’s white.

It is also possible to combine any of the above relation-sets whenever the underlying space-lattice has a natural interpretations for each relation-set’s roles. For instance, for spatio-temporal objects one could combine Allen’s Interval Algebra and RCC8.

There is, of course, qualitative information that cannot be represented by our bintrees, e.g. unknown data such as $EC(a, b) \lor PO(a, b)$ but where we do not know which, since our representation is a concrete model; unions such as $a \prec b \otimes c \land a \neq b \land a \neq c$, we can only state that $ov(a, b) \land ov(a, c)$; space-lattices that require infinite sets of roles, such as fuzzy sets with membership-roles in $[0, 1]$; formulae with role-variables, such as $R(z, x, y) := \pi_z(x) \prec y$; or shape-relations, we have not found a way to express formulae that can state e.g. concavity.

7 Related Work

There has been done much work on efficient representations of transitive relations and structures for reachability queries in directed graphs (see e.g. [15, 22, 7]) which can be used to represent our containment relationships. However, these representations do not facilitate efficient construction or update of these structures from a set of spatial objects. They are also less expressive, as they do not have any concept similar to our roles or the $<$-ordering.

In [17] the authors developed a qualitative representation of spatial data based on arrays of representative points. However, this representation has the same drawbacks as above.

There has also been done a lot of work on representing qualitative spatial information as a set of assertions in some spatial logic, whereby the main information extraction method is logical reasoning based on either logical calculi or constraint solving (see e.g. [21, 3] for an overview). These representations are more focused on the complex reasoning problems rather than efficient query answering, and therefore do not scale to large datasets.

In [13] the authors construct a compact representation for the RCC8 and CDC (Cardinal Direction Calculus) relations over polygons. They construct a minimum bounding rectangle (MBR) for each polygon and store the correct relations in normal database tables between polygons that have overlapping MBRs. They then show that if two polygons have disjoint MBRs, the RCC and CDC relations can be answered using only the MBRs, and given overlapping MBRs they check the explicit relation tables. However, their approach is limited to RCC and CDC relations over 2D polygons, whereas our approach can handle a more expressive set of relations over elements from any space lattice.

8 Conclusion and Future Work

We have seen that we always can construct a bintree representation for any space-lattice that is correct w.r.t. any predefined set of qualitative relations expressible in our formula language. This formula language is expressive enough to express most of the common qualitative spatial relations. Our bintree representations are compact, can be stored naturally in any tuple-based representation (RDBMS, RDF, etc.) and allow highly efficient query answering as they can be stored in a relational database and indexed by B-trees.

In the future we want to extend our implementation to also handle the role and order extensions and test these against real-world datasets with expressive relation-sets. It would also be interesting to try to extend the language of our relations, to for instance allow
intersections, unions, or some restricted form of universal quantification in our formulae
without effecting the computational properties of the algorithms.

References


