

# Pre-culling geometric linked building data for lightweight viewers

Linked Data

Master's dissertation submitted in order to obtain the academic degree of

Master of Science in de ingenieurswetenschappen: architectuur

Supervisor: Prof. ir.-arch. Paulus Present

Counselors: Ir.-arch. Jeroen Werbrouck  
Prof. dr. ir. arch. Ruben Verstraeten

Philippe Soubrier 01702837 philippe.soubrier@ugent.be  
Academic year: 2022–2023

# Contents

1	Introduction	6
1.1	Proposal . . . . .	6
1.2	Research questions . . . . .	8
1.2.1	Can LDBIM be culled? . . . . .	9
1.2.2	Can existing semantics be used? . . . . .	9
2	Linked Data	10
2.1	RDF and triples . . . . .	10
2.2	Ontologies and reasoning . . . . .	12
2.3	Triplestores and SPARQL . . . . .	12
2.4	Complexity of the data graph . . . . .	13
2.4.1	BIM geometry . . . . .	13
2.4.2	LDBIM geometry . . . . .	14
3	State of the art	15
3.1	Existing BIM viewers and ontologies . . . . .	15
3.1.1	Qoniq and LOD Streaming for BIM . . . . .	15
3.1.2	ld-bim.web.app . . . . .	16
3.1.3	AEC related ontologies . . . . .	16
3.1.4	GIS related ontologies . . . . .	19
3.2	On the market viewers comparison . . . . .	20
3.2.1	General Features . . . . .	20
3.3	Culling algorithms . . . . .	21
4	Dynamic Queries	22
4.1	Requirements . . . . .	22
4.2	Beyond geometry . . . . .	23
4.2.1	BCF integration . . . . .	23
4.2.2	Visualising semantic . . . . .	24
4.3	In situ WKT location . . . . .	24
4.4	In viewer "bot:Space" identification . . . . .	27
4.5	In query OBJ geometry filtering . . . . .	29

5	Modular Approach	31
5.1	Data fetching . . . . .	31
5.1.1	SPARQL fetcher . . . . .	31
5.1.2	Database fetcher . . . . .	31
5.2	Cache manager . . . . .	31
5.2.1	LRU algorithm . . . . .	31
5.3	Query processing . . . . .	31
5.3.1	Query builder . . . . .	31
5.3.2	Query composer . . . . .	31
5.4	Interactions . . . . .	31
5.5	Sequences . . . . .	31
6	Prototype	32
	References	35
	Referenced webistes	37

# List of Figures

1.1	Sequence diagram - basic concept . . . . .	8
1.2	Illustration of culling principle . . . . .	8
2.1	Triple structure . . . . .	10
2.2	Evolution of LOD during the life-cycle of a building . . . . .	13
3.1	Illustration of the BOT ontology . . . . .	17
3.2	Sequence diagram with external database - concept . . . . .	18
3.3	Illustration of the FOG and OMG ontologies . . . . .	19
3.4	Performance viewers . . . . .	21
4.1	Serialization, OMG level 1 . . . . .	25
4.2	In viewer "bot:Space" identification, with raytracing . . . . .	27

# List of Tables

1.1	Size of test-models in Johansson et al., 2015 . . . . .	7
3.1	Acceleration techniques used by tested viewers . . . . .	20
4.1	FOG ontology geometry formats . . . . .	23

# Short abstract

This is my short abstract.

# Abstract

This is my abstract

# Chapter 1

## Introduction

From 2D, to 3D and now to [BIM](#). The evolution of the Architecture, Engineering and Construction ([AEC](#)) industry has been a long and complex one. The introduction of 3D modeling was the first major step in the industry's evolution, as it allowed for more accurate representations of buildings. No longer solely relying on 2D drawings, a 3D model of a building can be used to create various representations, from a simple 2D floor plan to a full 3D model. Following the adoption of 3D modeling, the implementation of Building Information Modelling ([BIM](#)) emerged as another significant milestone. [BIM](#) adds an extra layer of information on top of the 3D model. As the digital representation of a building's physical and functional characteristics, BIM serves as a repository for semantics originating from various applications throughout the design and construction processes, including cost estimation, energy analysis, and production planning.

However, as mentioned in Werbrouck, [2018](#), the next challenge for the [AEC](#) industry is related to the domain-specific nature of current [BIM](#) softwares, which remains closed off to other disciplines. This data management challenge is currently being addressed by the Linked Building Data Community Group ([LBD-CG](#)) and other research entities, such as the University of Ghent, through the use of Web of Data technologies<sup>1</sup>. This emerging milestone will be discussed in this thesis under the term Linked Data [BIM](#) ([LDBIM](#)).

### 1.1 Proposal

Each of these evolutions has brought, and will continue to bring, a significant amount of data together. This volume is expected to grow exponentially in the future as the industry shifts towards a more digital approach and opens up to other stakeholders. The data graphs will not only expand in terms of semantics but also in geometry. This

---

<sup>1</sup>W3C, [2023](#).

makes visual querying, or simply put, 3D exploration of models, an increasingly difficult task. Especially when looking at newer devices used in the industry such as mobile phones, and tablets, which are becoming more and more powerful, but still have limited computational resources in comparison to office computers.

To bring this volume of geometric data in perspective, Table 1.1 shows the size of the test-models used in Johansson et al., 2015, a study from 2015 on the performance of BIM viewers for large models with the following description:

“ Although the Hotel model contains some structural elements they are primarily architectural models. As such, no Mechanical, Electrical or Plumbing (MEP) data is present. However, all models except the Hospital contain furniture and other interior equipment. ” (Johansson et al., 2015)

Model	# of triangles	# of objects	# of geometry batches
Library	3 685 748	7318	11 195
Student House	11 737 251	17 674	33 455
Hospital	2 344 968	18627	22 265
Hotel	7 200 901	41 893	62 624

Table 1.1: Size of test-models in Johansson et al., 2015

These models demonstrate how basic BIM models can already contain a significant amount of data. LDBIM will not only bring together new stakeholders but also be able to keep track of multiple geometry versions for each object, should they occur. Therefore, this thesis proposes a new approach to the visual querying of LDBIM models, wherein viewers will not have to load the entire model into memory. Instead, after filtering at the source, only the geometry needed for the visual tasks at hand will be loaded, while maintaining the original link to each resource for further processing and use cases. This filtering step is commonly referred to as culling in the computer graphics industry and is illustrated in Figures 1.1 and 1.2.

Figure 1.1 illustrates the basic idea of this thesis, presenting an extra step in the communication between a user, represented here by a Hand Held Device (HHD), and a database storing the model. An HHD has been chosen to exemplify a low-powered device used in the field, which requires a lightweight 3D viewer to visualize and explore the digital twin of the building. The HHD is assumed to have no knowledge of the LDBIM model and only receives the geometry that needs to be displayed from the database. On the other hand, the database is assumed to possess, or have access to, all the knowledge of the model and the necessary semantics to perform the culling. In addition to the culling, Web of Data technologies behind LDBIM allow for a more flexible approach to data management. The expressive capabilities of SPARQL Protocol and RDF Query Language (SPARQL) enable complex and fine-grained queries, in contrast to current BIM approaches, for data retrieval. This offers a broad range of



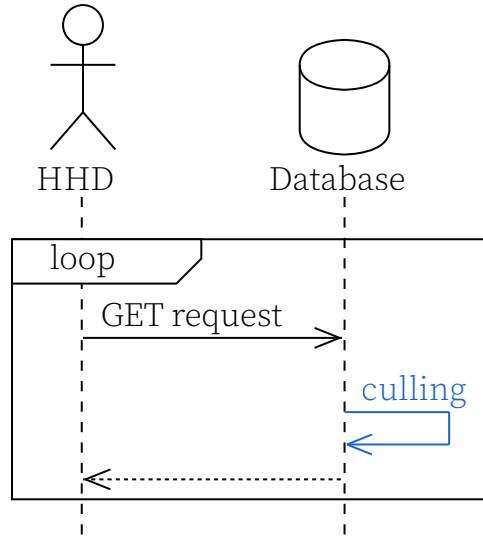


Figure 1.1: Sequence diagram - basic concept

end-use cases tailored to multiple stakeholders. In this thesis, this translates to user or application-specific query adaptation capabilities.

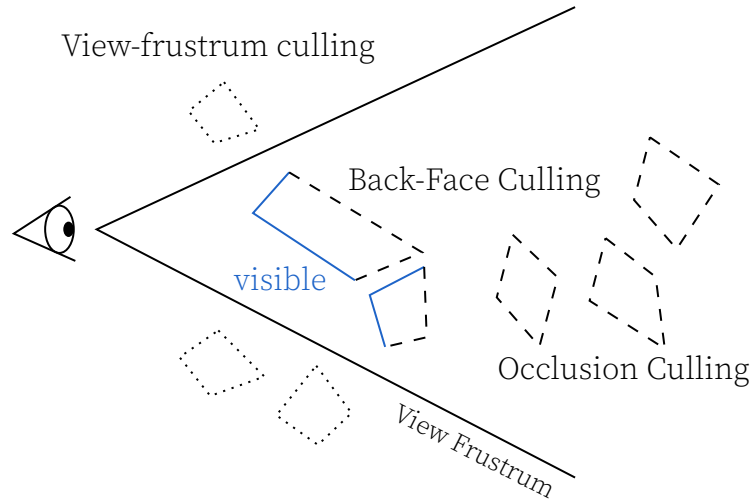


Figure 1.2: Illustration of culling principle, based on Cohen-Or et al., 2003

Figure 1.2 showcases multiple culling techniques to showcase some culling principles. The first technique, *frustum culling*, is used to determine which objects are visible to the user. The second technique, *occlusion culling*, is used to determine which objects are occluded by or behind other objects. And lastly, *back-face culling*, is used to determine which faces, and not whole objects, are facing away from the user.

## 1.2 Research questions

This thesis proposes the introduction of culling algorithm technology within the context of LDBIM to address the previously mentioned issue of the scene's size, by culling

the scene at its source prior to sending it to the viewer. As culling algorithms have been extensively researched and continue to evolve, as described in section 3.3, the research questions in this thesis concentrate on assessing the feasibility of introducing such algorithms in LDBIM. It aims to propose a set of possible solutions tailored to this specific problem, while highlighting possibilities for future research and specific use cases.

### 1.2.1 To what extent can LDBIM geometry be culled to be streamed to lightweight viewers?

This thesis focuses on computing with data snippets or triples inside a LDBIM model, rather than within. This means that the smallest unit of data that can be culled is the one described in a single triple, which is, in the most likely scenario, a single Level of Detail (LOD) of a single element. It implies that geometry is defined and separated at the object level. Furthermore, this thesis will not address culling techniques such as back-face culling, as these will be left to the viewer itself, not the database.

Determining the necessary data snippets for the viewer is a key question. The fundamental requirements for the viewer include, first, the geometry itself, which involves selecting the appropriate geometry format for the application as well as the additional visual information such as color, texture, etc. Second, the identifier of each element is of crucial importance in order to maintain the link to other semantic resources in the graph. This enables the viewer to retrieve those resources for a multitude of use cases, transforming it into a user-friendly visual query tool.

### 1.2.2 Can existing semantics and ontologies be used to feed possible culling algorithms?

Unlike the computer graphics industry, this interconnected context already contains both explicit and implicit relationships within the graph, the latter being derived through inferencing. This is similar to Johansson and Roupé, 2009 and their paper where they utilized the semantics of a BIM model in Industry Foundation Classes (IFC) format to develop culling techniques. However, this thesis will concentrate on the use of Semantic Web resources. As such, it will examine both AEC-specific and AEC-related ontologies, such as those related to Geographic Information System (GIS), to determine if they can be employed to feed culling algorithms.

# Chapter 2

## Linked Data

As mentioned in the [Introduction](#), the evolution from [BIM](#) to [LDBIM](#) is an evolution of the data *management* layer. “Linked Data”, as stated by the World Wide Web Consortium ([W3C](#)), is a collection of interrelated datasets on the Web, formatted in a standard way that is accessible and manageable by Semantic Web tools. The same applies to the relationships among them.<sup>2</sup> The following collection of Semantic Web technologies explores the required environment to achieve this goal.

### 2.1 [RDF](#) and triples

At the core of the Semantic Web is the Resource Description Framework ([RDF](#)), a data model for describing resources on the Web. RDF is a graph data model that consists of *triples*, which are statements about resources. A triple consists of a subject, a predicate, and an object. The subject is the resource that is being described, the predicate is the property of the subject, and the object is the value of the property. Both the predicate and the object can, in turn, become the subjects of other triples. Listing [2.1](#) shows an example of an [RDF](#) database described in the Turtle format.

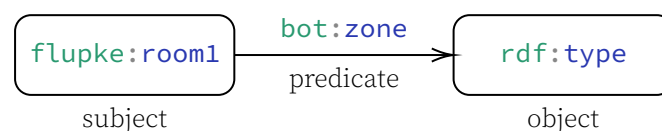


Figure 2.1: Triple structure

The basic, yet versatile, structure of a triple is illustrated in Figure [2.1](#). Both the subject and object are considered as nodes in the data graph, and they are linked by the predicate, which is referred to as an edge. Multiple triples can thus create and link

---

<sup>2</sup>W3C, [2015b](#).

multiple nodes or enrich a connection between two nodes by creating new edges between them. Each element contains a single resource that can be one of the three types: a [URI](#), a literal, or a blank node. A Uniform Resource Identifier ([URI](#)) identifies the name and/or location of a resource on the web and, as its name states, is unique and unambiguous, thus enabling queries and reasoning of the same nature. A literal is a value, and a blank node is an anonymous resource, sometimes used as a placeholder when the exact resource is not known or not necessary to specify. Due to their nature, a subject must be either a [URI](#) or a blank node, a predicate exclusively a [URI](#), and the object may be any of the three types. As [URI](#) descriptions can be very long, a prefix can be used to shorten them. This is illustrated in Listing 2.1 with the `@prefix bot: <https://w3id.org/bot#>`, which declares that `bot:Zone` refers, in its full length, to the address `<https://w3id.org/bot#Zone>`.

```
@prefix fog: <https://w3id.org/fog#> .
@prefix omg: <https://w3id.org/omg#> .
@prefix bot: <https://w3id.org/bot#> .
@prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix rdfs: <http://www.w3.org/2000/01/rdf-schema#> .
@prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix rdfs: <http://www.w3.org/2000/01/rdf-schema#> .
@prefix xsd: <http://www.w3.org/2001/XMLSchema#> .
@prefix flupke: <http://flupke.archi#> .

flupke:room1 rdf:type bot:Zone ;
  bot:containsElement flupke:coneOBJ ;
  bot:containsElement flupke:cubeGLTF .

flupke:coneOBJ omg:hasGeometry flupke:coneOBJ_geometry-1 ;
  rdf:type bot:Element .

flupke:cubeGLTF omg:hasGeometry flupke:cubeGLTF_geometry-1 ;
  rdf:type bot:Element .

flupke:coneOBJ_geometry-1 rdf:type omg:Geometry ;
  fog:asObj_v3.0-obj
  ↪ "https://raw.githubusercontent.com/flol3622/AR-Linked-BIM-viewer/_
  ↪ main/public/assets/database_1/coneOBJ.obj"^^xsd:anyURI
  ↪ .

flupke:cubeGLTF_geometry-1 rdf:type omg:Geometry ;
  fog:asGltf_v1.0-gltf
  ↪ "https://raw.githubusercontent.com/flol3622/AR-Linked-BIM-viewer/_
  ↪ main/public/assets/database_1/cubeGLTF.gltf"^^xsd:anyURI
  ↪ .
```

Listing 2.1: Example of an [RDF](#) database in turtle format

This basic concept can be extrapolated to describe and store any kind of data. The advantage for the [AEC](#) industry would be to allow any stakeholders to describe and enrich the knowledge base of a building.

## 2.2 Ontologies and reasoning

When looking at Listing 2.1, a distinction can be made between two types of statements: some refer to classes or properties, such as `bot:Zone` or `bot:containsElement`, while others refer to instances such as `flupke:room1`. The former is referred to as the TBox for “terminology”, and the latter is referred to as the ABox for “assertions”. The TBox, the ontology layer, is used to describe instances in the ABox and their relationships.

By developing an ontology, the domain of interest and the relationships between the classes and properties can be described. This is achieved by defining the classes and properties of the domain and their relationships. The ontology is then used to reason about the domain, inferring new facts based on the ontology and the existing facts within the domain. This is done by a reasoner, which is software capable of performing the reasoning itself on the ontology and associated data. As mentioned, the reasoner can be used to infer new facts, check if created facts are consistent with the ontology, and check if the ontology itself is consistent.<sup>3</sup> It is often integrated with [RDF](#) databases, also known as triplestores or graph databases.

Classes, properties, and their relationships can be defined using Resource Description Framework Schema ([RDFS](#)), which is a vocabulary for describing [RDF](#) schemas using a basic set of constructs. As an extension of [RDFS](#), Web Ontology Language ([OWL](#)) is a vocabulary for describing ontologies using a more expressive set of constructs tailored to the needs of ontologies. Both [RDFS](#) and [OWL](#) are considered to be formal ontologies themselves, as they describe the classes and properties of the domain of [RDF](#).

## 2.3 Triplestores and [SPARQL](#)

As briefly discussed in 2.2, triplestores are [RDF](#) databases that store data in the form of a graph. They are used to store and query Linked Data and are often integrated with a reasoner. The data itself is retrieved and modified using the [SPARQL](#).<sup>4</sup> In contrast to Structured Query Language ([SQL](#)), [SPARQL](#) queries are able to work across multiple triplestores, called [SPARQL](#) endpoints. These are known as federated queries, and their results are combined into a single result set. This is useful when the data is distributed across multiple triplestores in a decentralized manner.<sup>5</sup> For example, multiple stakeholders participating in a project, each with their own database.

---

<sup>3</sup>W3C, 2015a.

<sup>4</sup>W3C, 2015c.

<sup>5</sup>Ontotext, 2022.

## 2.4 Complexity of the data graph

The complexity of the data graph is a major concern when working with [LDBIM](#). This section discusses the origins of the different sources of geometric data that enrich it.

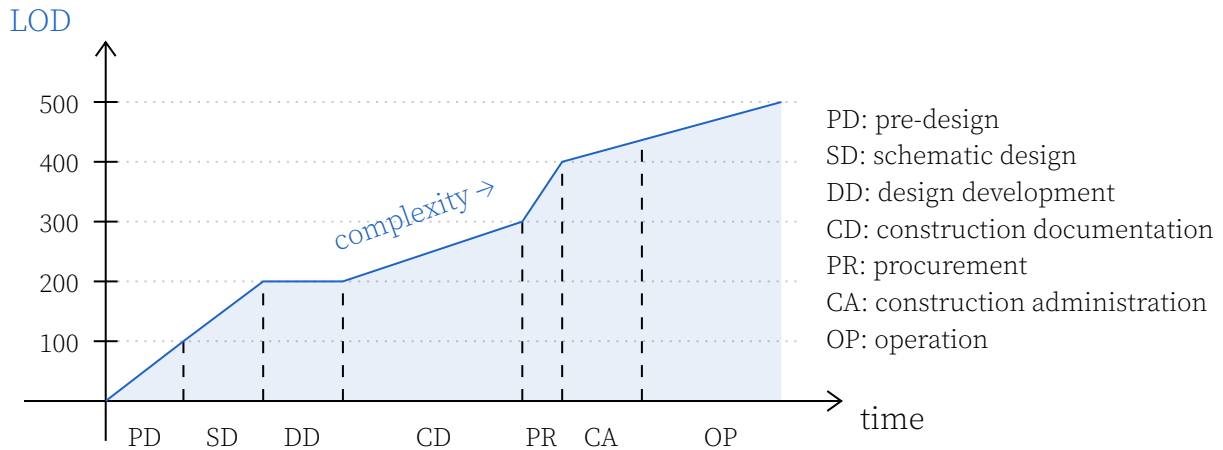


Figure 2.2: Evolution of [LOD](#) during the life-cycle of a building.

Based upon the Macleamy Curve (Ilozor & Kelly, [2012](#))

random values at the moment

### 2.4.1 [BIM](#) geometry

The 3D model of a building consists of a multitude of sub-models, describing objects for all the different stakeholders participating in the project. Some describe very large objects, and some very small parts. Both can be defined in their most simple and abstract form or have an intricate and complex geometry. For instance, a door can simply be defined as a box, or up to the level of the screw-thread for the hinge system. The level of abstraction is here described as the [LOD](#), which is most of the time pre-selected for the needs of a [BIM](#) model, and is applied throughout a single model.

As shown in Figure [2.2](#), a standard BIM workflow goes through multiple phases, each with their associated model and [LOD](#). This makes it an important concept in the [AEC](#) industry, as it allows for a very efficient workflow. The modeling step is approached from a top-down perspective, starting with rougher geometries describing the broader ideas of a concept model and evolving to a more refined model for the construction documentation phase. As the last and longest-standing model, a higher [LOD](#) can be used to describe subtle changes in the evolution of a building during the operation phase.

### 2.4.2 LDBIM geometry

The interconnectivity of semantics can also be applied to geometry descriptions. This could allow the co-existence of multiple LODs in a single model database. Besides storing the evolution of a single element's geometry, it enables the linking of the different LODs, described in 2.4.1, to each other. Not only that, but extending onto the size of the models described in Table 1.1, already existing Mechanical, Electrical and Plumbing (MEP), structural, alongside many other stakeholders' geometry can be added.

# Chapter 3

## State of the art

As mentioned in Johansson et al., 2015, existing research on the performance of currently used BIM viewers is quite limited. This state-of-the-art research will, therefore, focus on the overall features of some promising newer viewers and the ontologies that will be used in this thesis.

### 3.1 Existing BIM viewers and ontologies

#### 3.1.1 Qoniq and LOD Streaming for BIM

Qonic focuses on developing an open platform BIM viewer. With the use of Unity to enable cross-platform compatibility, they focused on two main aspects: performance and aesthetics. The latter refers to the visual quality of the viewer, offering both a seamless experience for the viewer as well as a pleasant one, with, for example, the implementation of ambient lighting and shadow castings. The first and most researched aspect of their viewer, the performance, is mainly focusing on a LOD culling algorithm.

(T. Strobbe, personal communication, November 25, 2022)

##### 3.1.1.1 Qoniq's approach to LOD streaming

Their core research is developing a dynamic LOD streaming model. Starting from the geometry and semantics of an IFC file, they compute an LOD hierarchy tree of the model. Through multiple mesh decimation algorithms, they reduce the number of triangles of each object's mesh, regardless of the semantics associated with that object. On top of that, a filtering algorithm is implemented in the streaming model to filter out objects, regarding their semantics, that are not relevant to the current camera position. In doing so, they both reduce the size of models far from the viewpoint and evaluate the need to show certain objects based on their nature, extracted from semantics in the IFC file, and their distance to the camera. The resulting dynamic LOD streaming



model is reevaluated at each camera move in Unity.  
(T. Strobbe, personal communication, November 25, 2022)

Unity was chosen as it allows for writing once and deploying everywhere. This means that the viewer can be used on any platform, including mobile devices and browsers. The performance results are thus related to the hardware capabilities of each device, with the exception of the browser, where the performance of Unity’s WebGL build is limited to a scene size of 2Gb.<sup>6</sup>

### 3.1.1.2 Advantages and trade-offs

Being able to run on many platforms, offering a smooth viewer experience and a pleasing aesthetic makes it an ideal candidate for lightweight viewers on the job site. However, the LOD library has to be computed on every model update. The decimation algorithms are furthermore computational results that are not humanly reviewed. This means that the quality of the resulting meshes is not guaranteed for the lower LODs, which are, as illustrated in Figure 2.2, already modeled in previous design phases. LDBIM could, by interconnection, recall previous LODs in the viewer’s scene. Without the need for computational remodeling. Nevertheless, Qonic serves as this thesis’s goal, outside the LDBIM context.

### 3.1.2 ld-bim.web.app

“The purpose of the app is to showcase our LBD toolset and to demonstrate the capabilities of Linked Building Data to newcomers.”<sup>7</sup>

<https://ld-bim.web.app/> demonstrates a viewer built around an RDF database. It separates the data from an IFC file into semantics, stored in the previously mentioned graph, and a glTF model, together with a JavaScript Object Notation (JSON) file containing a reference table. Extra local or remote graphs can be added to the User Interface (UI). As it contains a SPARQL engine to query and visualize, in the form of highlighting, the results of the query in a 3D viewer. The viewer is based on the ifc.js project, which is itself based on the three.js 3D JavaScript library.

### 3.1.3 AEC related ontologies

As mentioned in the second research question 1.2.2, this section will discuss AEC-related technologies that are actively researched by the LBD-CG<sup>8</sup>.

---

<sup>6</sup>Unity, 2023.

<sup>7</sup>Rasmussen and Schlachter, n.d.

<sup>8</sup>LBD-CG, 2022.

### 3.1.3.1 BOT

The Building Topology Ontology (BOT) proposes a set of classes and properties, “which provides a high-level description of the topology of buildings including storeys and spaces, the building elements they may contain, and the 3D mesh geometry of these spaces and elements.” (Rasmussen et al., 2020), as illustrated in Figure 3.1. This high-level description could be fed to portal-culling algorithms in a situation where the visibility is contained within one `bot:Space` or `bot:Storey`, or it could extend the scope to `bot:adjacentZone`<sup>9</sup>. Additionally, it could play a part in the construction of the Bounding Volume Hierarchy (BVH) needed for other occlusion culling algorithms, such as the Coherent Hierarchical Culling algorithm (CHC)<sup>++</sup> (Johansson et al., 2015).

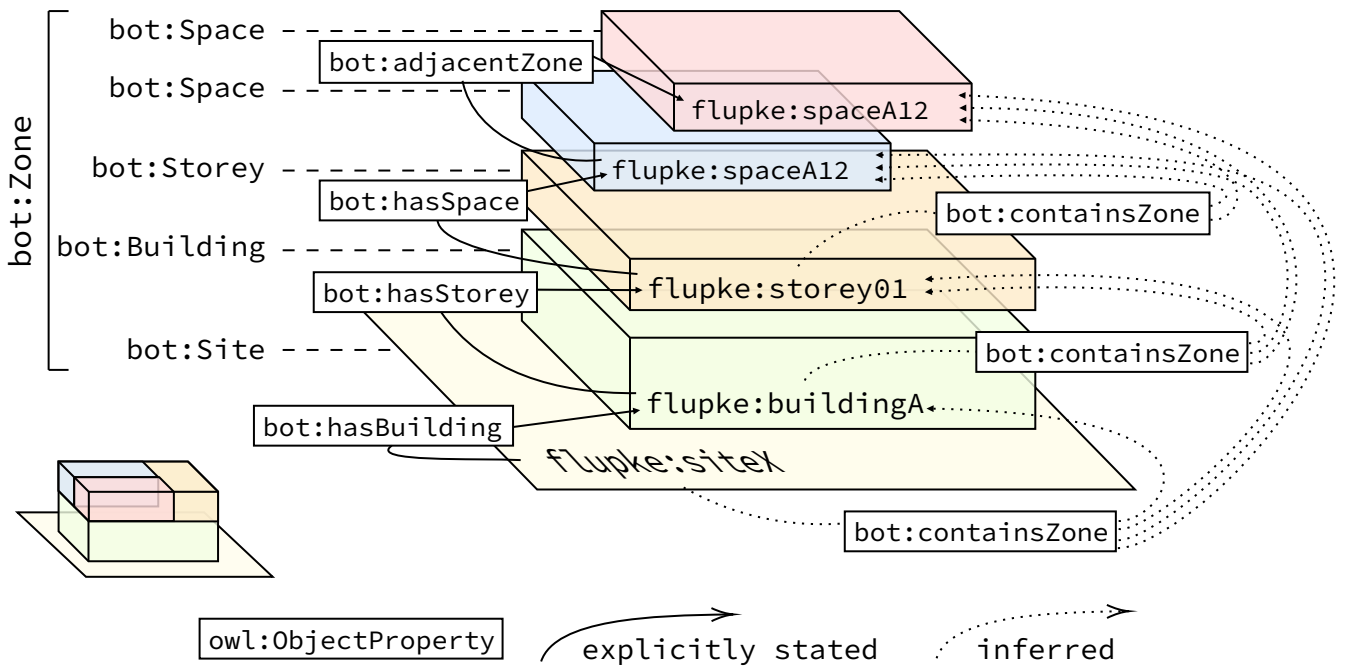


Figure 3.1: Illustration of the BOT ontology, based on Rasmussen et al., 2020.

### 3.1.3.2 FOG and OMG

With the help of File Ontology for Geometry formats (FOG) and Ontology for Managing Geometry (OMG), geometry descriptions can be linked in the data graph. The innovation lies in the choice to store it either inside or outside the graph, by means of one triple referring to a literal or an URI. Listing 2.1 showcases multiple examples of objects assigned with a geometry description using an URI (Bonduel et al., 2019).

<sup>9</sup>Linietsky et al., 2023.

```
flupke:coneOBJ_geometry-1 fog:asObj_v3.0-obj "https://..."^^xsd:anyURI .
```

Listing 3.1: Example of FOG usage

Listing 3.1 describes a subject of datatype `xsd:anyURI` from the Extensible Markup Language (XML) Schema Definition (XSD)<sup>10</sup>. The versatile approach of Bonduel et al., 2019 also proposes the following datatypes: `xsd:string` for American Standard Code for Information Interchange (ASCII)-based geometry descriptions or `xsd:base64Binary` for binary geometry descriptions.

The format of the geometry is assigned directly by the predicate in Listing 3.1, which is `fog:asObj_v3.0-obj`. This further infers the statements in Listing 3.2.<sup>11</sup>

```
flupke:coneOBJ_geometry-1 fog:asObj "https://..."^^xsd:anyURI ;
ex:LOD "100"^^xsd:integer .
```

Listing 3.2: FOG inference examples

The possibility of introducing an external geometry location adds a new participant to Figure 1.1, the external database, from which some files are expected to be fetched. This is illustrated in Figure 3.2.

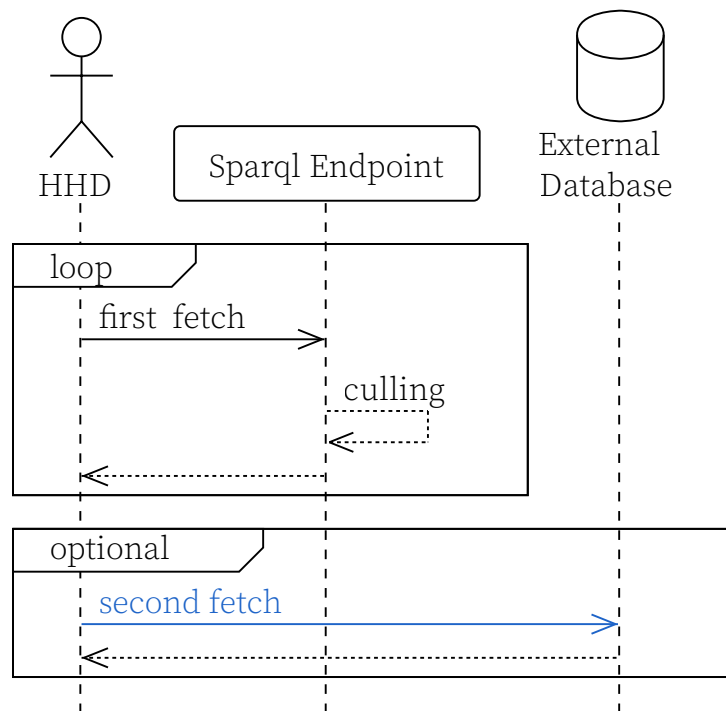


Figure 3.2: Sequence diagram with external database - concept

<sup>10</sup>Carrol and Pan, 2006.

<sup>11</sup>Bonduel et al., 2020.

Bonduel et al., 2019 refers to the proposal of the **LBD-CG** stated in Holten Rasmussen et al., 2018 to write Linked Data patterns on three possible levels, “each having a different degree of complexity”. The first and second levels are illustrated in Figure 3.3. Level 2 allows assigning multiple geometry descriptions to a single object, each with, for example, a different **LOD**.

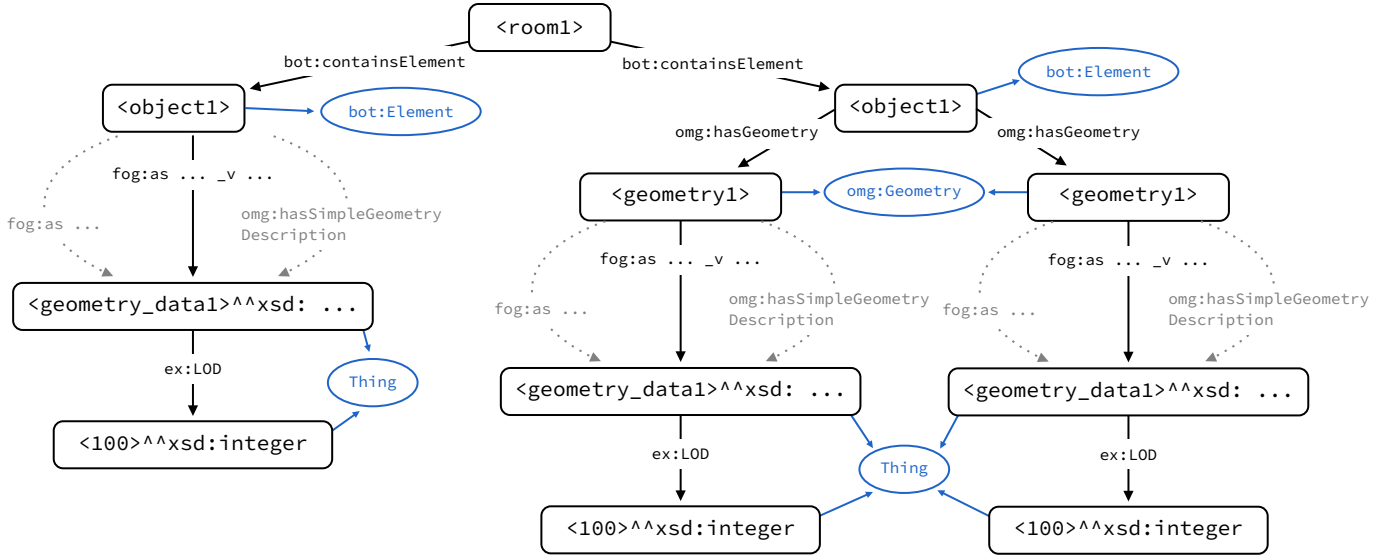


Figure 3.3: Illustration of Level 1 (left) and Level 2 (right) of the **FOG** and **OMG** ontologies, based on Bonduel et al., 2019. **LOD can't be assigned to literal, needs to be changed**

### 3.1.4 GIS related ontologies

The technological field of study, Geographic Information System (**GIS**), is closely related to the **BIM** domain. The central standards organization, Open Geospatial Consortium (**OGC**), which actively maintains the **GIS** standards, is also prominent in the Semantic Web scene. They recognize widely adopted standards such as GeoSPARQL.<sup>12</sup>

#### 3.1.4.1 GeoSPARQL

“The OGC GeoSPARQL standard supports representing and querying geospatial data on the Semantic Web. GeoSPARQL defines a vocabulary for representing geospatial data in RDF, and it defines an extension to the SPARQL query language for processing geospatial data. In addition, GeoSPARQL is designed to accommodate systems based on qualitative spatial reasoning and systems based on quantitative spatial computations.”<sup>13</sup>

As multiple triplestores and **SPARQL** endpoints support the GeoSPARQL extension, it is a viable candidate for spatial and **LOD** culling algorithms. Such algorithms require spa-

<sup>12</sup>OGC, 2023a.

<sup>13</sup>OGC, 2023b.

tial data, such as the distance from the viewpoint to the object. Spatial query functions proposed in this extension are needed for this purpose. The functions can compute on nodes of geospatial geometry as if they are expressed using Well-Known Text (WKT) or the Geography Markup Language (GML). These expressions can be assigned by using the predicates `geo:asWKT` or `geo:asGML`. However, GeoSPARQL comes with some limitations that are less prevalent in the GIS domain, which mostly requires 2D data (Perry & Herring, 2012), in contrast to BIM where 3D distance functions would be needed. Despite such limitations, GeoSPARQL remains a viable solution for spatial querying, and workarounds could be employed to address them.

## 3.2 On the market viewers comparison

Johansson et al., 2015 mentioned in their paper the lack of research about objective BIM viewers comparison and made one as a result. The size of the model they tested can be found in Table 1.1. They evaluated the following viewers:

- DDS CAD Viewer
- Tekla BIMsight
- Autodesk Navisworks
- Solibri Model Viewer

### 3.2.1 General Features

Their study had two main goals. Firstly, evaluating existing viewers and their capabilities, they identified the acceleration techniques used, which are presented in Table 3.1.

BIM viewer	Acceleration technique
Solibri 9.0	VFC   DC (optional)   HAGI (optional)
Naviswork 2015	VFC   DC (optional)   CPU OC (optional) GPU OC (optional)
BIMsight 1.9.1	VFC
DDS 8.0	VFC   DC (optional)
DDS 10.0	VFC   DC

Table 3.1: Acceleration techniques used by tested viewers from Johansson et al., 2015. (View Frustum Culling (VFC), Drop Culling (DC), Hardware Accelerated Geometry Instancing (HAGI), Central Processing Unit (CPU), Occlusion Culling (OC))

Secondly, they implemented modern culling algorithms and strategies such as CHC++.

The worst-case scenarios are shown in Figure 3.4 against the Solibri viewer. The results are quite promising, but as concluded by the authors, the gains are limited to the capacities of the Graphics Processing Unit (GPU), Video Random Access Memory (VRAM), and Random Access Memory (RAM), as discussed in 1.2.

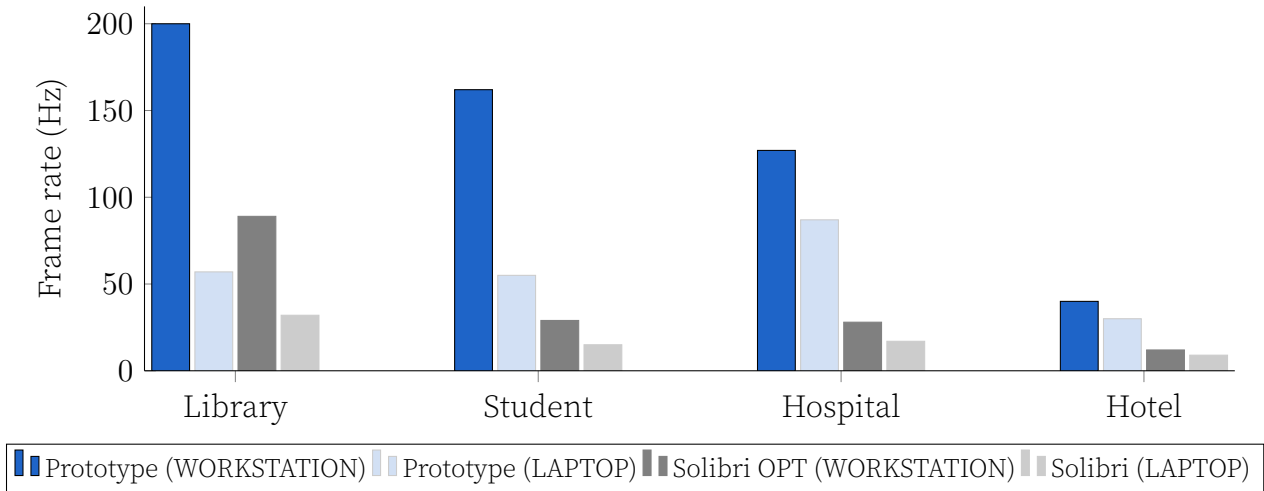


Figure 3.4: Comparison in rendering performance.  
from Johansson et al., 2015

### 3.3 Culling algorithms

Johansson et al., 2015 presented in their paper a new BIM viewer equipped with the powerful CHC++. This is a third-generation occlusion culling algorithm developed by Mattausch et al., 2008a, the first being the CHC (Bittner et al., 2004), followed by the Near Optimal Hierarchical Culling (NOHC) (Mattausch et al., 2008b). Their conclusion stated that although occlusion culling is very efficient, it is still bound to the scene size, which is limited by hardware capabilities. More precisely, the GPU, VRAM, and RAM capacities.

# Chapter 4

## Dynamic Queries

This chapter introduces the concept of dynamic querying. In this thesis, it refers to the automatic generation of queries responsible for obtaining the data needed to visualize building elements from a [BIM](#) model within an [RDF](#) graph. The examples are presented as static SPARQL queries, since the automation itself depends on the implementation or framework used. A link to the appendix is provided, where the implementation within the prototype of this thesis is explained.

The structure is as follows: first, the requirements for the viewer are researched, as its functioning will dictate the output of the queries. Second, the capabilities of the viewer are explored in relation to the visualization of semantic data, thus emphasizing the added value of working with Linked Data. Third, three types of dynamic queries for culling are presented, each with its own advantages and disadvantages.

### 4.1 Requirements

A viewer designed to visualize data stored in an [RDF](#) graph is required to understand the data stored within it. Therefore, the requirements for the viewer align with those of its source, the [RDF](#) graph. Section 3.1.3.2, which discusses the use of both the [FOG](#) and [OMG](#) ontologies, offers an overview of the available options in terms of file format and file source. The [FOG](#) ontology supports the description of a wide range of geometry formats, as illustrated in Table 4.1. In conjunction with the [OMG](#) ontology, which allows for the description of the file source using the datatype of the literal, it can be concluded that the viewer should be able to handle a broad spectrum of file formats, preferably described in the [FOG](#) ontology, and accept both remote files and literal values.

COLLADA	Compressed LAS	Compressed Nexus	DWG
E57	GeoJSON	Well Known Text SFA	GML
IFC	IGES	LAS Point Cloud	Nexus
OBJ	PCD Point Cloud	Uncompressed LAS	Revit
Rhino	Shapefile	Simple Feature Access	SketchUp
SPFF	STEP SPFF	Uncompressed Nexus	SVG
PLY	STL	Well Known Binary SFA	X3D
glTF			

Table 4.1: List of geometry formats that can be assigned with the [FOG](#) ontology.<sup>14</sup>

## 4.2 Beyond geometry

This section highlights the advantages a viewer based on Linked Data has over a viewer based on traditional file-based systems, by extending the thought of a 3D viewer to its ability to visualise non geometric data from its source. [LDBIM](#) links geometrical entities to their corresponding semantic data, which can be visualised in the viewer. This allows for the visualisation of data that is not directly related to the geometry of the building elements, such as the physical properties of a wall or the cost of a door. The possibilities are endless, as long as the data is available in the [RDF](#) graph. The following sections will discuss different possible implementations.

### 4.2.1 BCF integration

As a first possible implementation of non-geometric data, this section examines the [BIM](#) Collaboration Format ([BCF](#)) buildingSMART standard. [BCF](#) is an open file format that enables the creation and communication of issues about [BIM](#) models<sup>15</sup>. Both it and its translation in the Semantic Web as [BIM](#) Collaboration Format Ontology ([bcfOWL](#)) (Schulz et al., 2021) link a screenshot, a camera angle, and a list of concerned entities to form a specific issue<sup>16</sup>.

This type of semantic offers two types of implementations. The first is the positioning of a screenshot, together with its camera position and orientation, within the 3D scene. This allows for the visualization of issues, which can be linked to the screenshot, in the viewer, offering communication integration. The second implementation involves the metadata surrounding issues that can be used as visual properties to feed specific queries. This type of implementation is discussed in the next section, [Visualising semantic](#).

<sup>14</sup>Bonduel et al., 2020

<sup>15</sup>"BIM Collaboration Format (BCF) - buildingSMART Technical", [n.d.](#)

<sup>16</sup>"What is BCF - BIMcollab", [n.d.](#)



## 4.2.2 Visualising semantic

When examining semantics such as physical properties of entities, free from geometric and spatial data (see [BCF integration](#)), a visual interpretation superimposed on the viewer offers powerful rendering possibilities. By coloring elements based on their properties, both physical and non-physical, the viewer can be used to detect anomalies or insights in the model, in a feature-rich output medium.

Physical properties such as thermal, acoustic, structural, and others, or non-physical properties like cost, time, and pathologies, can all be described and linked in the [RDF](#) graph. The expressive capabilities of [SPARQL](#) enable complex and fine-grained queries, offering application-specific query creation about these properties. By selecting a specific subset or combining them, a user or developer can transform the viewer into a powerful tool.

As such, the viewer is expected to comply with a multitude of requirements, which are not all covered in this thesis. Chapter [Modular Approach](#) thus proposes a modular approach, allowing for a step-by-step implementation of the viewer, starting with the most basic requirements, which are adopted in Chapter [Prototype](#), while leaving room for future extensions.

## 4.3 In situ WKT location

The first type of dynamic query identifies the room in which the observer/camera is located using the [WKT](#) serialization of `bot:Space` entities. It proposes to base its culling algorithm on super-elements such as rooms, thus grouping the scene into a limited number of elements within meaningful boundaries. As entities are primarily viewed within their allocated rooms, this approach takes advantage of the spatial organization of buildings using the [BOT](#) ontology. Moreover, not all building elements have a meaningful [WKT](#) serialization (e.g., a door), which makes the use of super-elements necessary.

This approach is limited to 2D, as the widely adopted GeoSPARQL functions used to achieve this are constrained to 2D. Nonetheless, the approach can be extended to 3D if the GeoSPARQL functions are extended to 3D in the future or the needed introduction of a 3D [SPARQL](#) is adopted.

Listing [4.1](#) proposes a static query in which a location is first assigned to the variable `?location` to create an easily reusable query. This location is the [WKT](#) serialization of a point, the coordinates of which can be updated at each move of the camera in the viewer. It therefore represents the location of the observer both on-site, “in situ” for augmented reality use cases, and in the 3D scene.

The query then combines two sets of so-called “entities”, represented in the graph as `bot:Element` or `bot:Space`, both of which are linked in the graph to their correspond-

ing [WKT](#) and geometric serialization using an [OMG](#) level 1 pattern explained in Figure 3.3. This pattern level links the literal directly to the entity, without the need for an intermediate node, but without the possibility to assign multiple serializations of the same type. The geometry literal is assigned using the [FOG](#) ontology, while the [WKT](#) literal is assigned using the GeoSPARQL ontology, as illustrated in Figure 4.1. As this last serialization requires a [WKT](#) format, a literal assigned with `geo:asWKT` is required, and queried for in the query.

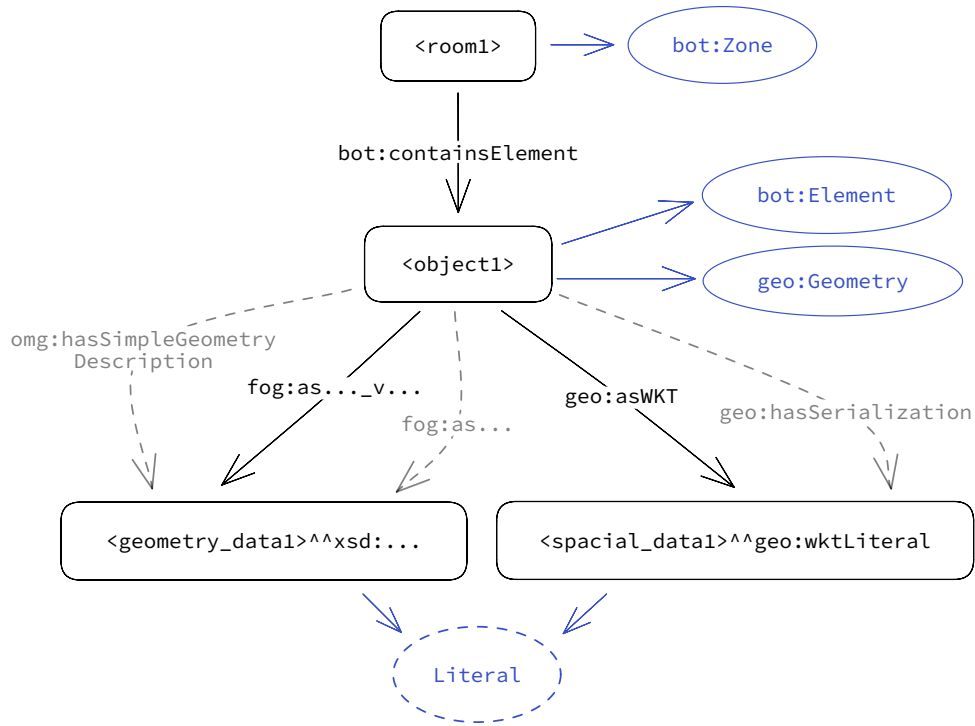


Figure 4.1: Serialization of a geometry using the [OMG](#) level 1 pattern.

The separation into two sets allows for querying entities that have a [WKT](#) serialization but are not located in a `bot:Space` (such as a floor, which would not be linked to a room) and entities that are located in a `bot:Space` that has a [WKT](#) serialization. The latter variant uses the `bot:containsElement` or `bot:adjacentElement` properties to select entities within a room.

After filtering out spaces, the geometry of which is not needed by the viewer, it filters the entities' geometry based on a list of implemented/accepted formats by the receiving viewer. These last consist of a list of [FOG](#) super-properties inferred by their assignment. In other words: a triple in the [RDF](#) graph of the form `?entity fog:asObj_v3.0 ?geometry_data` infers `?entity fog:asObj ?geometry_data` (as shown in Figure 4.1). This allows for the use of a single property to query related formats while still permitting the declaration of more specific formats if needed.

In the end, the query returns a list of entities that are located in the same room as the

observer, or are not located in a room but have a [WKT](#) serialization that is validated by the GeoSPARQL filter function. Together with the needed metadata, which is further explained in [Section 5.4](#).

```
PREFIX bot: <https://w3id.org/bot#>
PREFIX fog: <https://w3id.org/fog#>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
PREFIX geof: <http://www.opengis.net/def/function/geosparql/>
PREFIX geo: <http://www.opengis.net/ont/geosparql#>
PREFIX xsd: <http://www.w3.org/2001/XMLSchema#>
PREFIX inst: <https://172.16.10.122:8080/projects/1001/>

SELECT DISTINCT ?entity ?fog_geometry ?datatype
WHERE {
  BIND(STRDT("POINT(5000 -5000)", geo:wktLiteral) AS ?location)

  {
    ?entity geo:asWKT ?entityWKT .
    FILTER(geof:sfWithin(?location, ?entityWKT))
  }
  UNION
  {
    ?space rdf:type bot:Space .
    ?space geo:asWKT ?spaceWKT .
    FILTER(geof:sfWithin(?location, ?spaceWKT))
    ?space bot:containsElement bot:adjacentElement ?entity .
  }

  FILTER NOT EXISTS { ?entity rdf:type bot:Space }
  ?entity ?fog_geometry ?geometryData .
  FILTER(?fog_geometry IN (fog:asObj, fog:asStl, fog:asGltf, FOG:asIfc))
  BIND(DATATYPE(?geometryData) AS ?datatype)
  FILTER(?datatype = xsd:anyURI)
}
```

Listing 4.1: Dynamic culling query using GeoSPARQL

In the case of [Listing 4.1](#), the GeoSPARQL `FILTER` uses the function `geof:sfWithin` to evaluate if the given point coordinates are within the [WKT](#) serialization, which in most cases represents a [WKT](#) polygon. For example:

```
"POLYGON ((40 10, 30 10, 30 40, 40 40, 40 10))"^^geo:wktLiteral
```

The use of GeoSPARQL results in a very efficient query, as the filtering is done directly in the query and does not require any post-processing. Although the lack of 3D capabilities is a limitation, it allows, in 2D to easily implement more complex algorithms leveraging the use of different functions. With which, together with a level 2 [OMG](#) pattern, [LOD](#) culling based on the distance to the observer can be implemented.

Selecting rooms and entities outside the room in which the observer is present is also conceivable when leveraging the use of the `bot:adjacentZone` property. Or using the GeoSPARQL distance to a room.

## 4.4 In viewer "bot:Space" identification

In this culling approach, ray tracing is employed to assess the room where the observer is situated. As illustrated in Figure 3.2, out of the three participants, only one utilizes a 3D engine—a component capable of handling geometric modeling operations, such as ray tracing—referred to as the 3D viewer. In essence, both the SPARQL endpoint and the external database lack the ability to perform 3D operations or evaluations. Consequently, this approach capitalizes on the 3D viewer to pinpoint the room in which the observer is positioned.

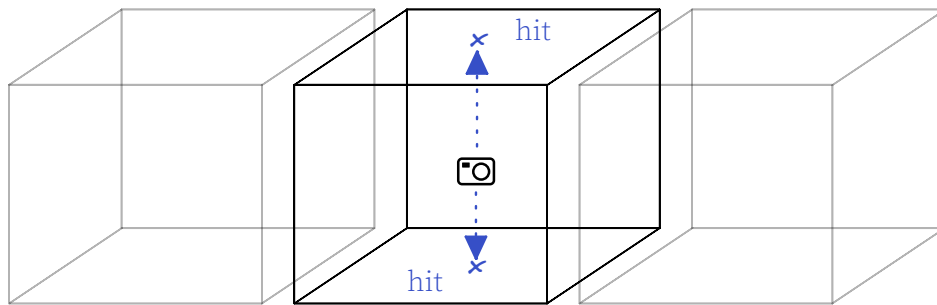


Figure 4.2: In viewer "bot:Space" identification, with raytracing

This process involves casting a ray from the observer's position both upwards and downwards, and evaluating the first intersection with a `bot:Space` entity. Unlike the previous approach, the geometry of the rooms themselves is required by the viewer, necessitating an initial query to load the rooms' geometry. By utilizing multiple layers in the viewer, the rooms' geometry can be concealed from the user while still being employed for ray tracing. Listing 4.2 demonstrates the function used in this thesis prototype to assess the location. The scope of the picking is limited to the rooms using the cache's metadata, in this case `lru: RefLRU`. Subsequently, two picking operations are carried out and compared, as depicted in Figure 4.2; if they match, the intersected room's URI is returned.

Once the room is identified, the query in Listing 4.3 is executed to retrieve the entities located within the room. This is achieved by querying the `bot:containsElement` property of the room. Similar to the previous method, it necessitates the combination of two sets of entities in this case: those located in the room and in the adjacent rooms themselves. This enables the optimization of the viewer by evaluating only the adjacent rooms, rather than all the rooms in the database.

```

function findRoom(viewer: RefViewer, lru: RefLRU, position: number[]):
↪ string | undefined {
  const down = [0, -1, 0];
  const up = [0, 1, 0];
  const rooms: string[] = [];

  lru.current?.forEach((value, key) => {
    if (value.botType === "Space") {
      rooms.push(key);
    }
  });

  function pick(direction: number[]) {
    return viewer.current?.scene.pick({
      origin: position,
      direction: direction,
      pickSurface: true,
      includeEntities: rooms,
    });
  }

  const resultDown = pick(down);
  const resultUp = pick(up);

  if (resultDown?.entity === resultUp?.entity) {
    return resultDown?.entity?.id;
  }
}

```

Listing 4.2: Typescript code for raytracing in viewer

The remainder of the query closely mirrors the previous one (Listing 4.1), with the distinction that it necessitates the additional storage of the `?botType` property to keep track of `BOT` classes and to identify `bot:Space` elements locally.

Due to the absence of `bot:adjacentZone` relations in the database used for the prototype, the query is unable to directly query the adjacent rooms. Instead, the property `bot:adjacentElement` is employed to query related room's adjacent walls and, consequently, the corresponding rooms. The lack of `BOT` relations in the utilized database was identified as a recurring issue in this thesis. Chapter ?? delves into this problem in greater detail.

This approach takes advantage of the 3D viewer and its engine to contribute to the culling process, although the first step requires downloading all the rooms' geometry. However, this step can be replaced by an initial manual localization by the user. By utilizing the adjacent rooms during subsequent queries, the impact on performance and

resource usage remains minimal throughout the operation. In contrast to the [In situ WKT location](#), this method is not limited to 2D but can be employed in 3D. However, it requires the geometry format of the room to be compatible with the viewer.

```
PREFIX bot: <https://w3id.org/bot#>
PREFIX fog: <https://w3id.org/fog#>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
PREFIX geof: <http://www.opengis.net/def/function/geosparql/>
PREFIX geo: <http://www.opengis.net/ont/geosparql#>
PREFIX xsd: <http://www.w3.org/2001/XMLSchema#>
PREFIX inst: <https://172.16.10.122:8080/projects/1001/>

SELECT DISTINCT ?entity ?fog_geometry ?datatype ?botType
WHERE {
    BIND (inst:room_1xS3Bck291UvhgP2dvNvkw AS ?room) .

    {
        ?room bot:adjacentElement ?adjacentWall .
        ?adjacentWall rdf:type bot:Element .
        ?entity bot:adjacentElement ?adjacentWall .
        ?entity rdf:type bot:Space .
    } UNION {
        ?room bot:containsElement ?bot:adjacentElement ?entity .
    }
    FILTER (?botType != bot:Space)

    ?entity ?fog_geometry ?geometryData .
    ?entity rdf:type ?botType .
    FILTER(STRSTARTS(str(?botType), "https://w3id.org/bot#"))
    FILTER(?fog_geometry IN (fog:asObj, fog:asStl, fog:asGltf, FOG:asIfc))
    BIND(DATATYPE(?geometryData) AS ?datatype)
    FILTER(?datatype = xsd:anyURI)
}
```

Listing 4.3: Querying in viewer "bot:Space" identification

## 4.5 In query OBJ geometry filtering

```

PREFIX bot: <https://w3id.org/bot#>
PREFIX fog: <https://w3id.org/fog#>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
PREFIX geof: <http://www.opengis.net/def/function/geosparql/>
PREFIX geo: <http://www.opengis.net/ont/geosparql#>
PREFIX xsd: <http://www.w3.org/2001/XMLSchema#>
PREFIX inst: <https://172.16.10.122:8080/projects/1001/>
PREFIX jsfn: <http://www.ontotext.com/js#>

SELECT ?entity ?fog_geometry ?datatype ?botType
WHERE {
    BIND("5000,-15000,2000" as ?position)

    # select the room
    ?room a bot:Space .
    ?room fog:asObj ?obj .
    BIND(DATATYPE(?obj) AS ?type)
    FILTER(?type = xsd:string)
    FILTER(jsfn:insideObjAABBox(?obj, ?position))

    # select the entities in the room
    ?room bot:containsElement bot:adjacentElement ?entity .
    ?entity ?fog_geometry ?geometryData .
    FILTER(?fog_geometry IN (fog:asObj, fog:asStl, fog:asGltf, FOG:asIfc))
    BIND(DATATYPE(?geometryData) AS ?datatype)
    FILTER(?datatype = xsd:anyURI)
    ?entity rdf:type ?botType .
    FILTER(STRSTARTS(str(?botType), "https://w3id.org/bot#"))
}

```

Listing 4.4: Querying in query OBJ geometry filtering

```

function insideObjAABBox(objString, point) {
    var lines = objString.split("\n");
    var vertices = [];
    for (var i = 0; i < lines.length; i++) {
        if (lines[i].trim() !== "") {
            var coords = lines[i].split(" ");
            var vertex = [
                parseFloat(coords[0]),
                parseFloat(coords[1]),
                parseFloat(coords[2]),
            ];
            vertices.push(vertex);
        }
    }
    var verticesT = transpose(vertices);
    var AABOX = [];
    for (var i = 0; i < 3; i++) {
        var minVal = Math.min.apply(Math, verticesT[i]);
        var maxVal = Math.max.apply(Math, verticesT[i]);
        AABOX.push([minVal, maxVal]);
    }
}

```

# Chapter 5

## Modular Approach

The theoretical model proposed in this thesis is informed by the practical experiences and observations gained through the development of the prototype.

### 5.1 Data fetching

#### 5.1.1 [SPARQL](#) fetcher

#### 5.1.2 Database fetcher

### 5.2 Cache manager

#### 5.2.1 [LRU](#) algorithm

### 5.3 Query processing

#### 5.3.1 Query builder

#### 5.3.2 Query composer

### 5.4 Interactions

### 5.5 Sequences



# Chapter 6

## Prototype

# List of Acronyms

AEC	Architecture, Engineering and Construction . . . . .	6
ASCII	American Standard Code for Information Interchange . . . . .	18
BCF	BIM Collaboration Format . . . . .	23
bcfOWL	BIM Collaboration Format Ontology . . . . .	23
BIM	Building Information Modelling . . . . .	6
BOT	Building Topology Ontology . . . . .	17
BVH	Bounding Volume Hierarchy . . . . .	17
CHC	Coherent Hierarchical Culling algorithm . . . . .	17
CPU	Central Processing Unit . . . . .	20
DC	Drop Culling . . . . .	20
FOG	File Ontology for Geometry formats . . . . .	17
GIS	Geographic Information System . . . . .	9
GML	Geography Markup Language . . . . .	20
GPU	Graphics Processing Unit . . . . .	21
HAGI	Hardware Accelerated Geometry Instancing . . . . .	20
HHD	Hand Held Device . . . . .	7
IFC	Industry Foundation Classes . . . . .	9
JSON	JavaScript Object Notation . . . . .	16
LBD-CG	Linked Building Data Community Group . . . . .	6
LDBIM	Linked Data BIM . . . . .	6
LOD	Level of Detail . . . . .	9
LRU	Least Recently Used	
MEP	Mechanical, Electrical and Plumbing . . . . .	14
NOHC	Near Optimal Hierarchical Culling . . . . .	21
OC	Occlusion Culling . . . . .	20
OGC	Open Geospatial Consortium . . . . .	19
OMG	Ontology for Managing Geometry . . . . .	17
OWL	Web Ontology Language . . . . .	12
RAM	Random Access Memory . . . . .	21

RDF	Resource Description Framework . . . . .	10
RDFS	Resource Description Framework Schema . . . . .	12
SPARQL	SPARQL Protocol and RDF Query Language . . . . .	7
SQL	Structured Query Language . . . . .	12
UI	User Interface . . . . .	16
URI	Uniform Resource Identifier . . . . .	11
VFC	View Frustum Culling . . . . .	20
VRAM	Video Random Access Memory . . . . .	21
W3C	World Wide Web Consortium . . . . .	10
WKT	Well-Known Text . . . . .	20
XML	Extensible Markup Language . . . . .	18
XSD	XML Schema Definition . . . . .	18

# References

- Bittner, J., Wimmer, M., Piringer, H., & Purgathofer, W. (2004). Coherent hierarchical culling: Hardware occlusion queries made useful. *Computer Graphics Forum*, 23, 615–624. <https://doi.org/10.1111/J.1467-86>
- Bonduel, M., Wagner, A., Pauwels, P., Vergauwen, M., & Klein, R. (2019). Including widespread geometry formats in semantic graphs using rdf literals. <http://lib.ugent.be/catalog/pug01:8633665>
- Cohen-Or, D., Chrysanthou, Y., Silva, C., & Durand, F. (2003). A survey of visibility for walkthrough applications. *IEEE Transactions on Visualization and Computer Graphics*, 9, 412–431. <https://doi.org/10.1109/TVCG.2003.1207447>
- Holten Rasmussen, M., Lefrançois, M., Bonduel, M., Anker Hviid, C., & Karlshøj, J. (2018). Opm: An ontology for describing properties that evolve over time. *CEUR Workshop Proceedings*, 2159, 24–33.
- Ilozor, B. D., & Kelly, D. J. (2012). Building information modeling and integrated project delivery in the commercial construction industry: A conceptual study. *Journal of Engineering, Project, and Production Management*, 2, 23–36. <https://doi.org/10.32738/JEPPM.201201.0004>
- Johansson, M., & Roupé, M. (2009). Efficient real-time rendering of building information models. <https://www.researchgate.net/publication/220758081>
- Johansson, M., Roupé, M., & Bosch-Sijtsema, P. (2015). Real-time visualization of building information models (bim). *Automation in Construction*, 54, 69–82. <https://doi.org/10.1016/j.autcon.2015.03.018>
- Mattausch, O., Bittner, J., & Wimmer, M. (2008a). Chc++: Coherent hierarchical culling revisited. *Computer Graphics Forum*, 27, 221–230. <https://www.academia.edu/14388994>
- Mattausch, O., Bittner, J., & Wimmer, M. (2008b). Chc++: Coherent hierarchical culling revisited. *Computer Graphics Forum*, 27, 221–230. <https://www.academia.edu/14388994>
- Perry, M., & Herring, J. (2012). *Ogc geosparql - a geographic query language for rdf data* (OGC Implementation Standard 11-052r4). Open Geospatial Consortium. <http://www.opengis.net/doc/IS/geosparql/1.0>

- Rasmussen, M. H., Lefrançois, M., Schneider, G., & Pauwels, P. (2020). Bot: The building topology ontology of the w3c linked building data group. *Semantic Web*, 11, 1–20. <https://www.researchgate.net/publication/342802332>
- Schulz, O., Oraskari, J., & Beetz, J. (2021). Bcfowl: A bim collaboration ontology.
- Werbrouck, J. (2018). *Linking data : Semantic enrichment of the existing building geometry*. <http://lib.ugent.be/catalog/rug01:002494740>

# Referenced webistes

- Bim collaboration format (bcf) - buildingsmart technical.* (n.d.). <https://technical.buildingsmart.org/standards/bcf/>
- Bonduel, M., Wagner, A., & Pauwels, P. (2020). *Fog: File ontology for geometry formats.* <https://mathib.github.io/fog-ontology/>
- Carrol, J. J., & Pan, J. Z. (2006). *Xml schema datatypes in rdf and owl.* <https://www.w3.org/TR/swbp-xsch-datatypes/#sec-xm1s-dt>
- LBD-CG. (2022). *A list of ontologies related to linked building data.* <https://github.com/w3c-lbd-cg/ontologies>
- Linietzky, J., Manzur, A., & the Godot community. (2023). *Portal — documentation de godot engine (stable) en français.* [https://docs.godotengine.org/fr/stable/classes/class\\_portal.html](https://docs.godotengine.org/fr/stable/classes/class_portal.html)
- OGC. (2023a). *Geosparql - a geographic query language for rdf data.* <https://www.ogc.org/standard/geosparql/>
- OGC. (2023b). *Open geospatial consortium.* <https://www.ogc.org/>
- Ontotext. (2022). *What is sparql?* <https://www.ontotext.com/knowledgehub/fundamentals/what-is-sparql/>
- Rasmussen, M. H., & Schlachter, A. (n.d.). *Ld-bim - bim meets linked data.* <https://ld-bim.web.app/>
- Unity. (2023). *Manual: Memory in unity webgl.* <https://docs.unity3d.com/2023.2/Documentation/Manual/webgl-memory.html>
- W3C. (2015a). *Semantic web: Inference.* <https://www.w3.org/standards/semanticweb/inference>
- W3C. (2015b). *Semantic web: Linked data.* <https://www.w3.org/standards/semanticweb/data>
- W3C. (2015c). *Semantic web: Query.* <https://www.w3.org/standards/semanticweb/query>
- W3C. (2023). *Linked building data community group.* <https://www.w3.org/community/lbd/>
- What is bcf - bimcollab.* (n.d.). <https://www.bimcollab.com/en/resources/openbim/about-bcf/>