

Modeling Inhabited Smart Spaces to Support Interoperable IoT-Based Applications

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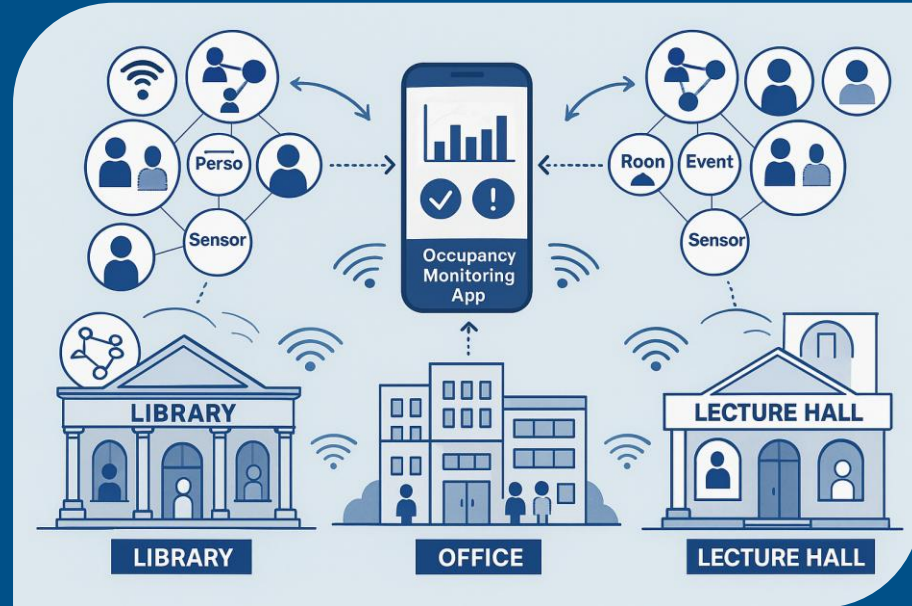
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MDM 2025



** Most images have been generated using ChatGPT*

Why Smart Spaces Matter

Smart spaces improve comfort, efficiency, and safety — by making environments aware of and responsive to people.



Energy-efficient workplaces



Responsive learning environments



Real-time healthcare support



Comfortable, adaptive living



Data-driven retail experiences

Typical IoT App Development

1. Start from a Specific Deployment

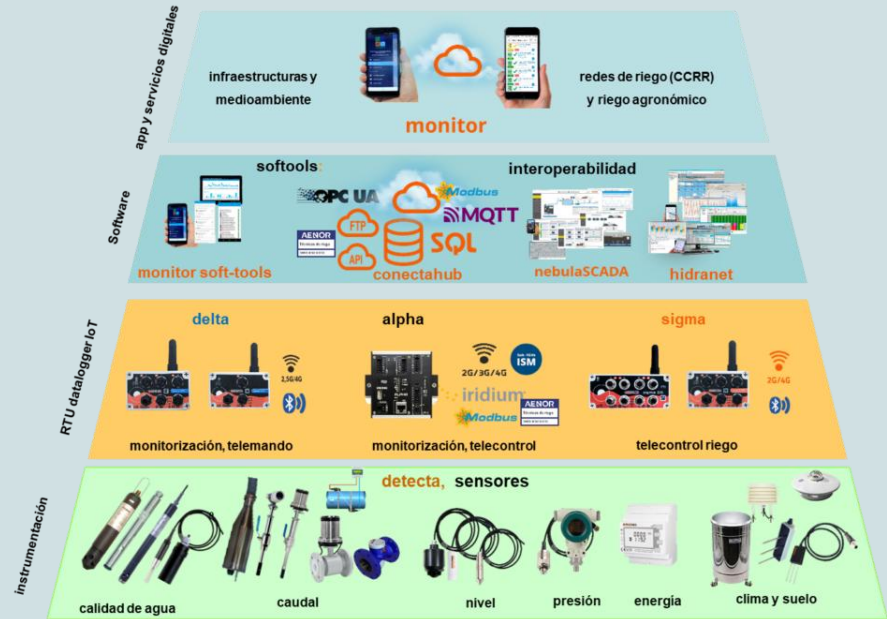
- E.g., “Developing an occupancy app for Building A (e.g., an office)”
- Uses particular sensors (e.g., WiFi APs, motion detectors)

1. Hardcoded Logic and Data Access

- Floor plans and room names are manually embedded in code.
- Sensor IDs / mappings are hardwired.
- Activity definitions (e.g., meeting = conference room + schedule) are custom.

1. App Works Only in That Space

- Cannot reuse in a different building (e.g., a classroom or shopping mall).
- Logic /infrastructure assumptions tied to one space.



Typical IoT App Development

1. Start from a Specific Use Case

- E.g., “Develop an app for Building A (e.g., a classroom) that uses particular sensors and detectors)”

1. Hardcoded Logic

- Floor plans are embedded in the app
- Sensor IDs / IP addresses are hardcoded
- Activity definitions are hardcoded (e.g., room + schedule)

1. App Works Only in One Place

- Cannot reuse the app in another classroom or building
- Logic /infrastructure is tied to the specific space.



Limited Reusability



High Redundancy



Poor Maintainability



Lack of Interoperability



Difficult to Scale

redes de riego (CCRR)
y riego agronómico

operabilidad



bulaSCADA

hidranet

sigma



telecontrol riego



presión

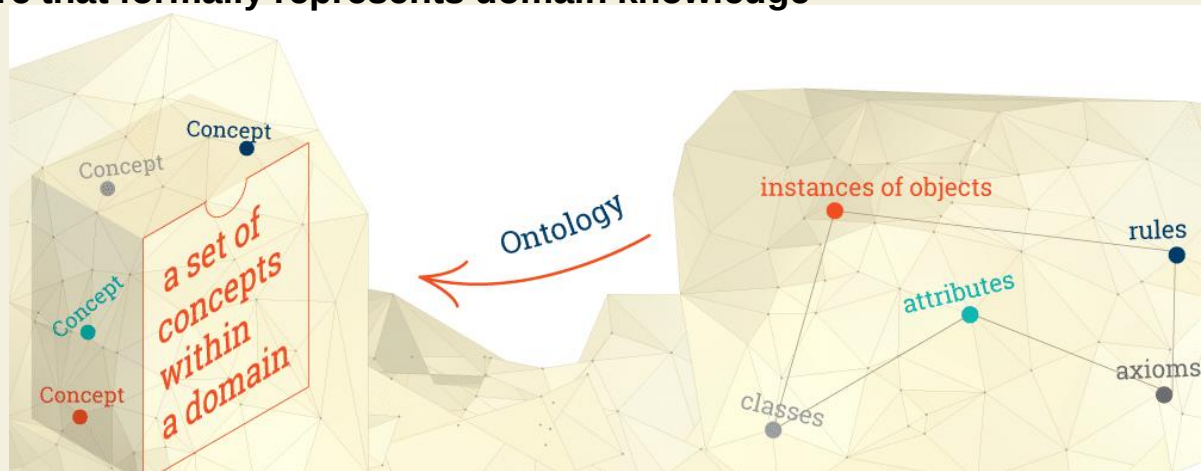
energía

clima y suelo

Ontologies for Modeling Semantic Knowledge

- **Ontology:** Knowledge structure that formally represents domain knowledge

- Classes/Concepts
- Instances (Objects)
- Properties/Attributes
(Relationships)



Related Works and Challenges

- **Existing Ontologies in Smart Spaces**

- **SSN & SOSA**
 - Focus on sensor, observation, and actuator descriptions
- **SAREF & SAREF4Building**
 - Smart device roles and commercial building topologies.
- **Brick Schema**
 - Sensor infrastructures in buildings (HVAC, lighting).
- **BOT**
 - Spatial relationships (containment, adjacency) in lightweight fashion.

- **Gaps and Limitations**

- **Existing vocabularies are:**
 - Primarily **system- and device-centric**, not **inhabitant-centric**.
 - **Lack integration of spatial-temporal uncertainty** (e.g., probabilistic locations).
 - **Provide limited support for recurring activities and spatial hierarchies across domains.**
- **Challenge: Semantic Interoperability Across Spaces**

Why Model Inhabited Smart Spaces?

- **Smart Spaces Are People-Centric**
 - IoT-enabled environments (e.g., campuses, offices, homes) aim to enhance human experience.
 - Applications include personalized HVAC control, occupancy-based regulation compliance, and activity tracking.
 - Human presence and behavior must be central to smart space modeling



Our goal:

- Develop a schema that enables semantic modeling of inhabited smart spaces (i.e., people, activities, spatial hierarchies, and uncertain sensor data).
- Support the development of interoperable, space-agnostic IoT-based applications.

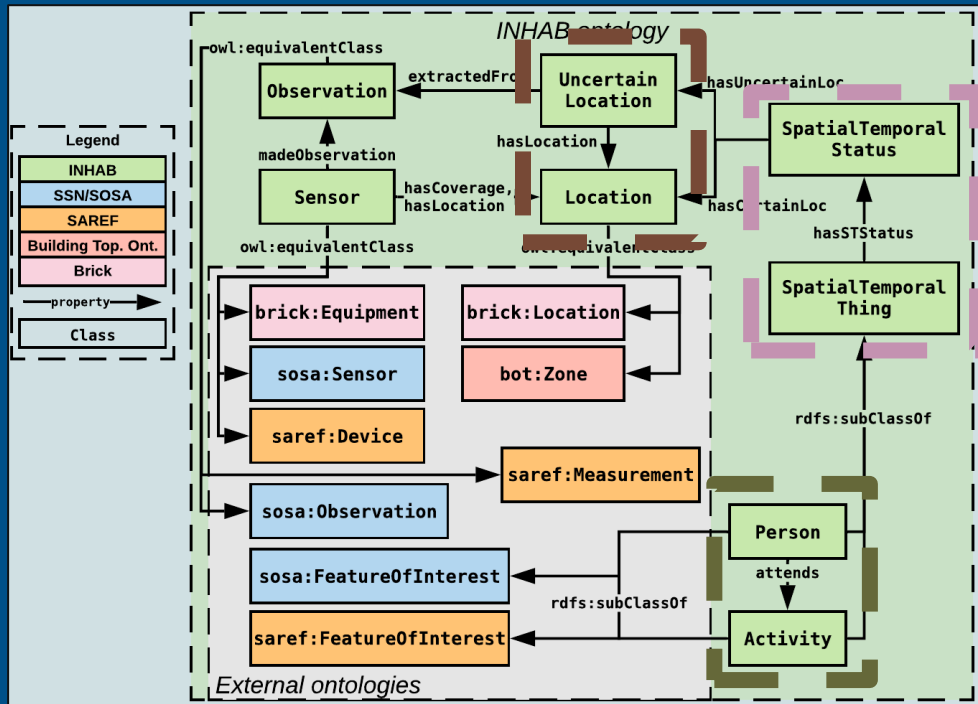
The INHAB Schema Overview

Key Features

- Integrates existing ontologies (e.g., SSN/SOSA, SAREF, Brick, BOT).
- Models people, activities, spatial hierarchies, and uncertainty in sensor data.

Three Core Modules

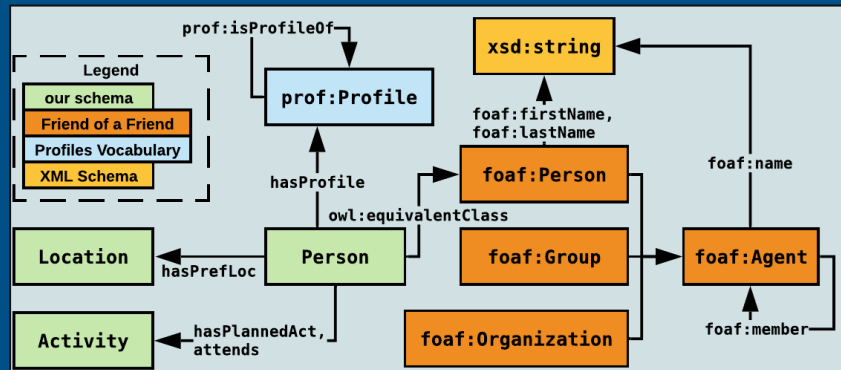
- People & Activities**
- Location & Topology**
- Spatial-Temporal Representation**



People & Activities Module

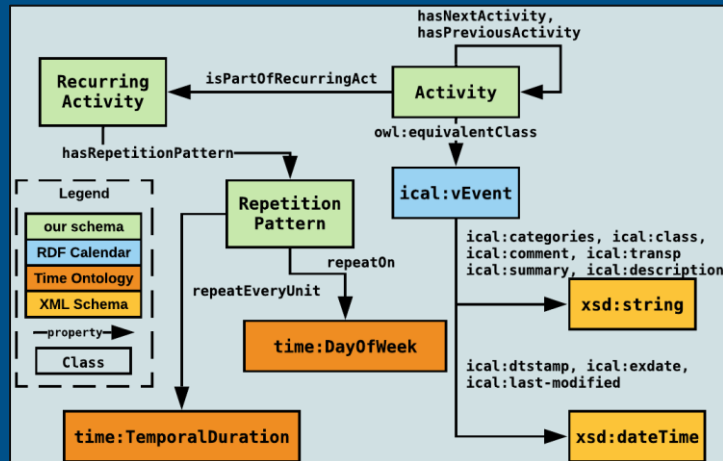
• People Representation

- Uses FOAF, DX-Prof for profiles (e.g., faculty, staff).
- Links to locations (e.g., office) and scheduled activities.



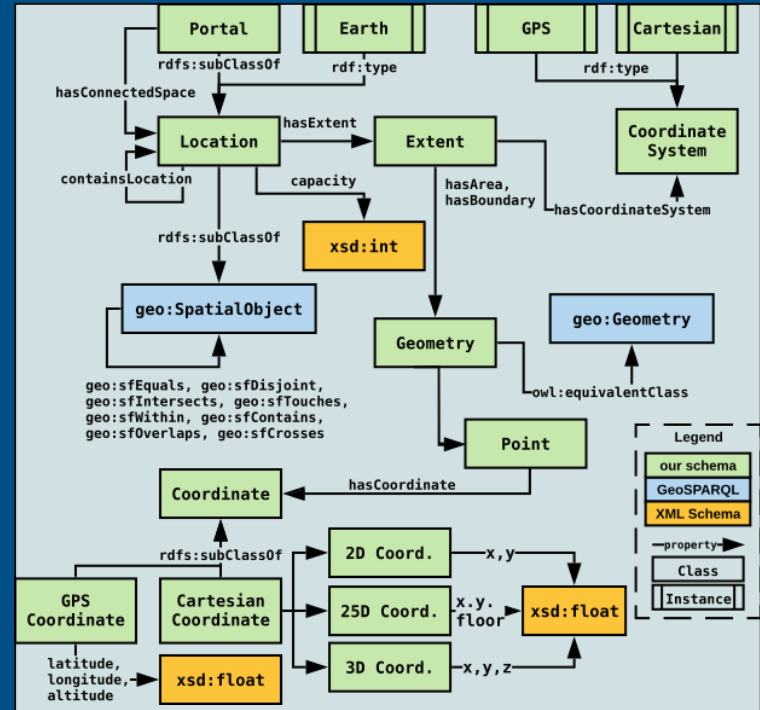
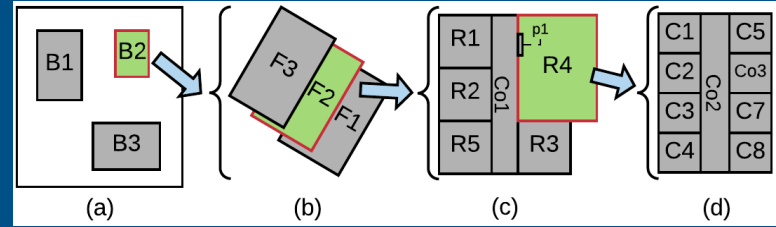
• Activity Representation

- Uses RDFcal and OWL-time for metadata (name, time, recurrence).
- Supports recurring activities for optimizing building operations.



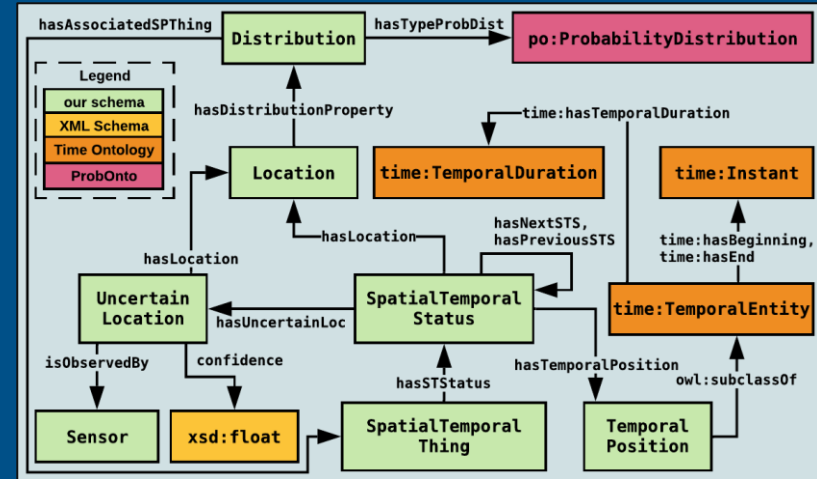
Location Modeling

- **Spatial Hierarchy**
 - Buildings → Floors → Rooms → Subspaces (e.g., cubicles).
- **Topological Relationships**
 - Explicit definitions: containment, adjacency, intersection.
- **Coordinate Systems**
 - Supports both GPS and Cartesian systems.
 - Allows definition of areas and boundaries using extensible geometry.



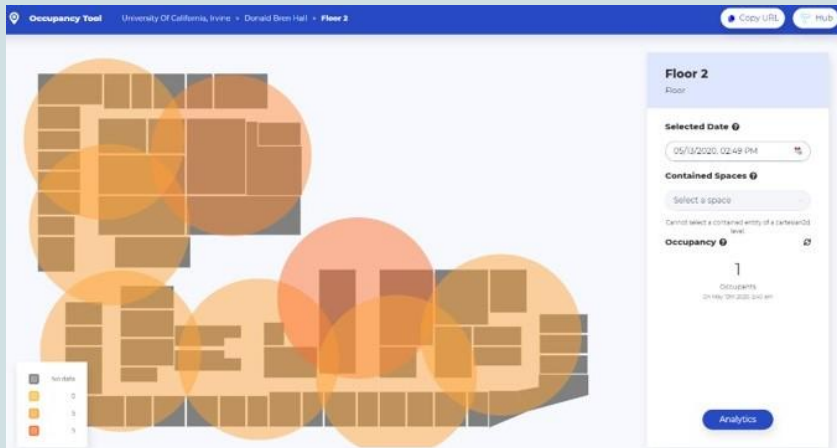
Spatial-Temporal Representation

- **Temporal Localization**
 - People and activities linked to time-bound locations.
- **Handling Uncertainty**
 - inhab:UncertainLocation class includes confidence levels.
 - Allows probabilistic modeling via distributions (e.g., uniform).
- **Coordinate Systems**
 - Supports both GPS and Cartesian systems.
 - Allows definition of areas and boundaries using extensible geometry.



Usage via SPARQL Queries

- Occupancy monitoring



Count number of people per building, floor, region, etc.

Retrieve people and their spatial-temporal location

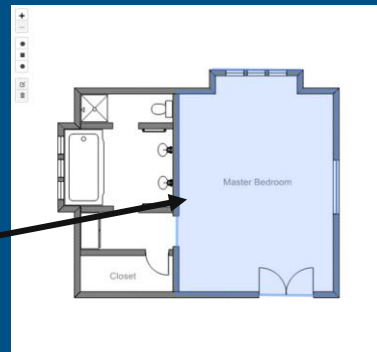
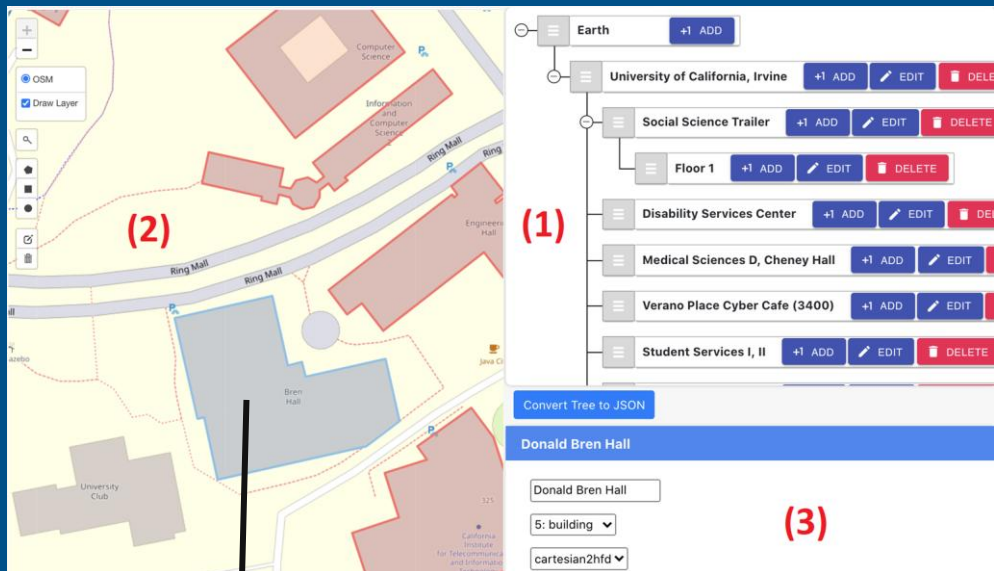
For those on the smart space (last 10 minutes)

Count them, taking into account an “uncertain location” (i.e., they are on campus, they are in the building, they *might* be in a floor or another)

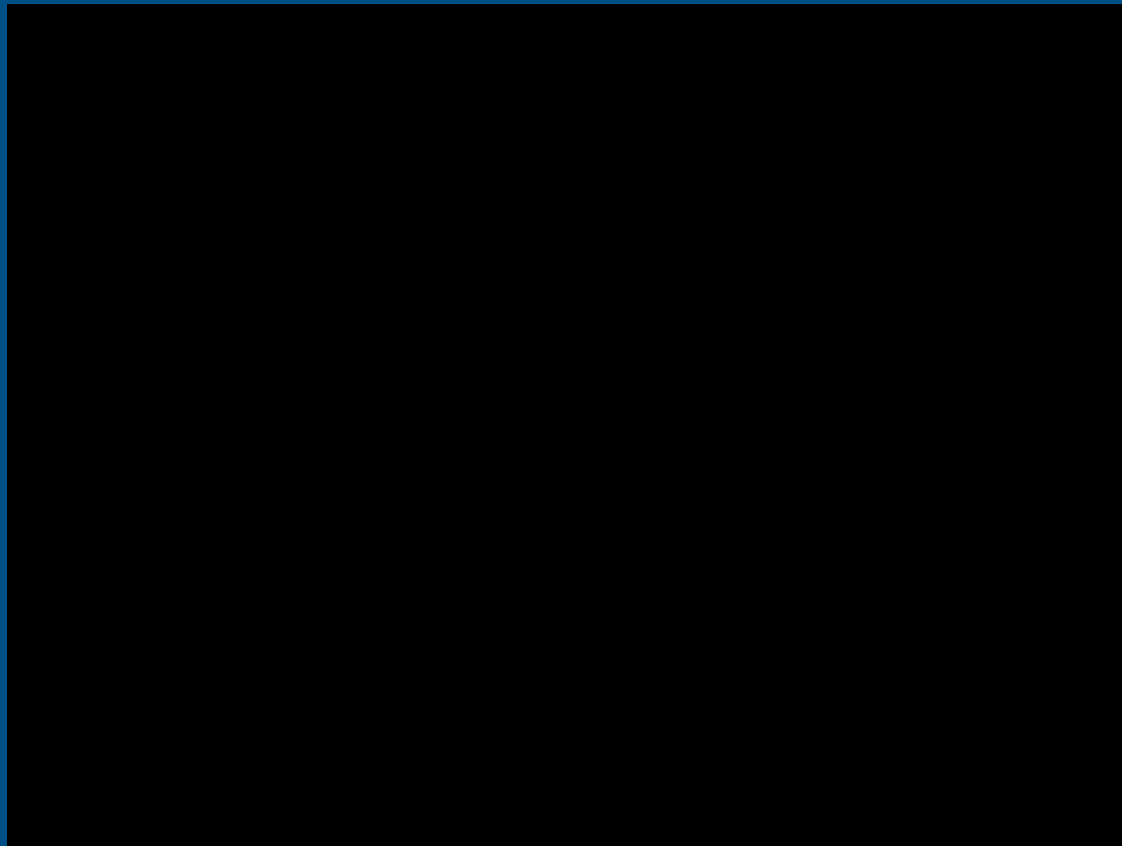
Count them, taking into account a “certain location” (i.e., they are on campus, they are in the building, they *are in this floor X*)

Annotation Tool

- **Purpose**
 - GUI for modeling spaces using INHAB schema.
- **Components**
 - Hierarchy Definition
 - Extent Drawing (map or floor plan)
 - Metadata Entry
- **Advantages**
 - Reduces annotation errors and complexity
 - Helps visualize spatial extent



Annotation Tool: Demo



Developing Inhab-Enabled Systems

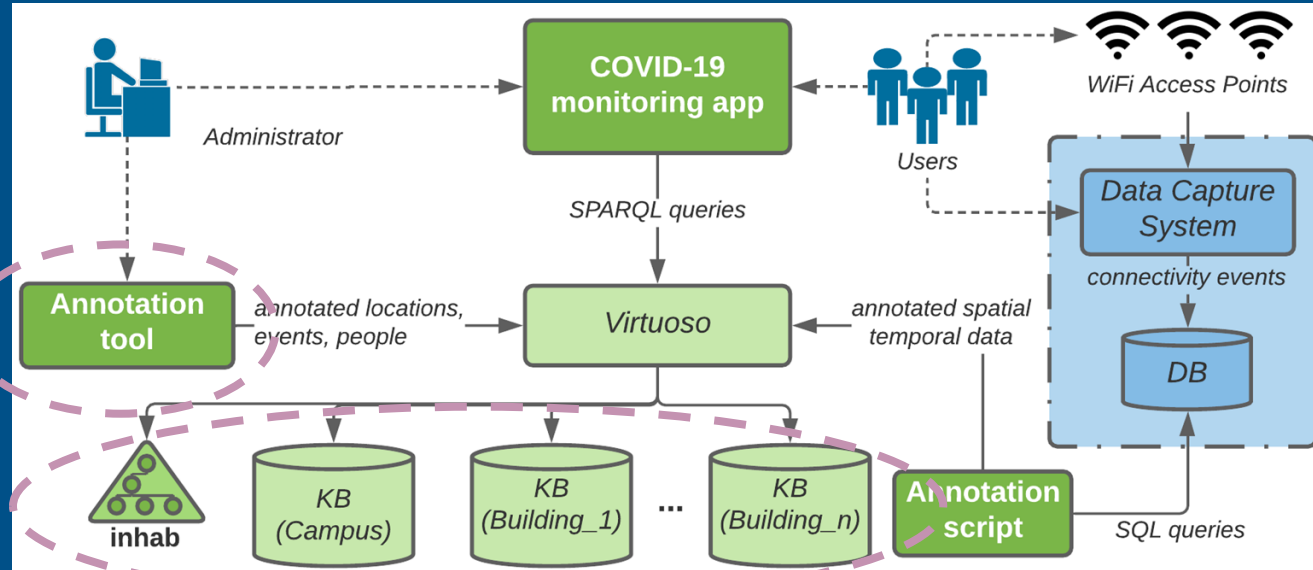
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UC Irvine campus

154 annotated buildings

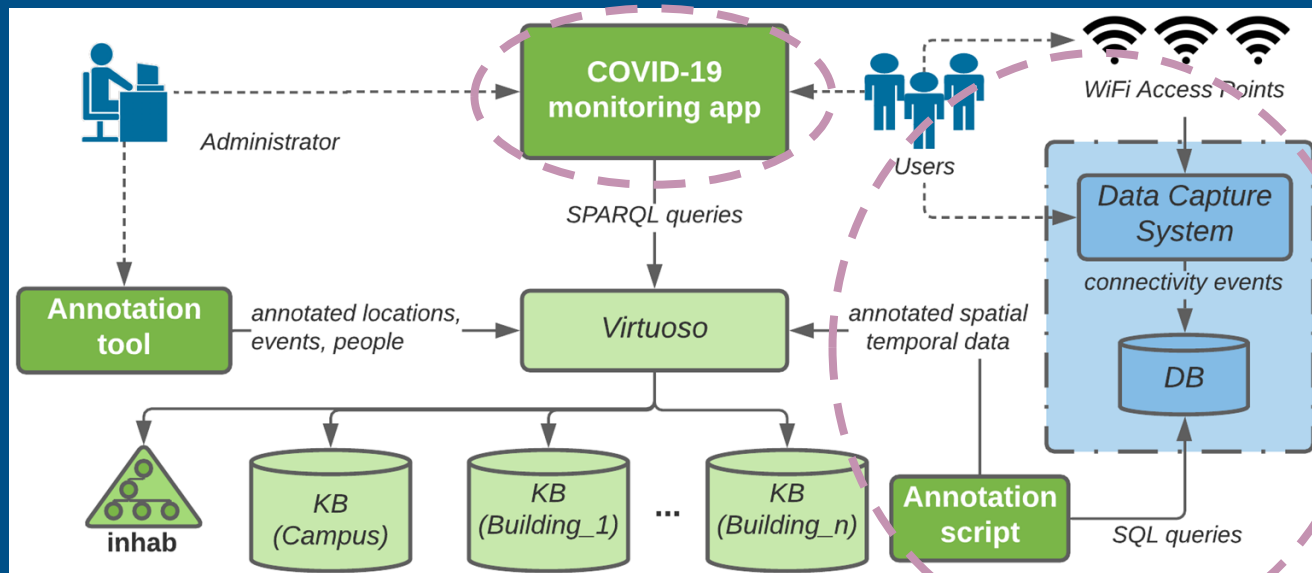
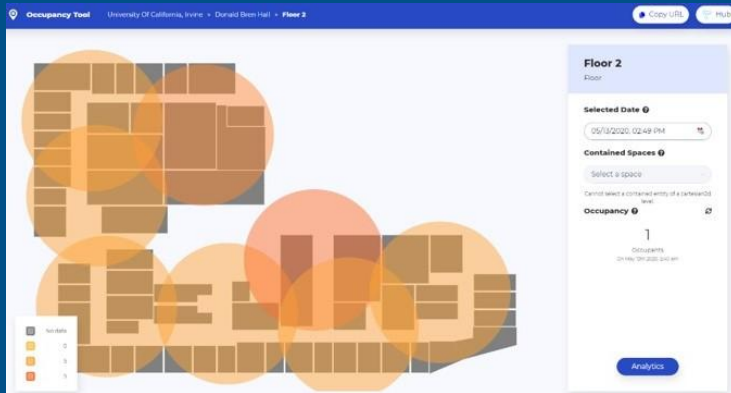
Detailed spatial hierarchy for selected buildings (e.g., DBH → 360 rooms / 6 floors)

Containment relationships for over 350 WiFi AP coverage areas.



UCI-KB → includes people and high-level spatial structure.

DBH-KB, FRH-KB, etc. → detailed space layouts and temporal event data per building



Converts WiFi connectivity logs into RDF triples describing spatial-temporal events

Syslog records (device MAC address, timestamp, and WiFi AP ID).

Models uncertain location using `inhab:UncertainLocation` with confidence values based on AP coverage.

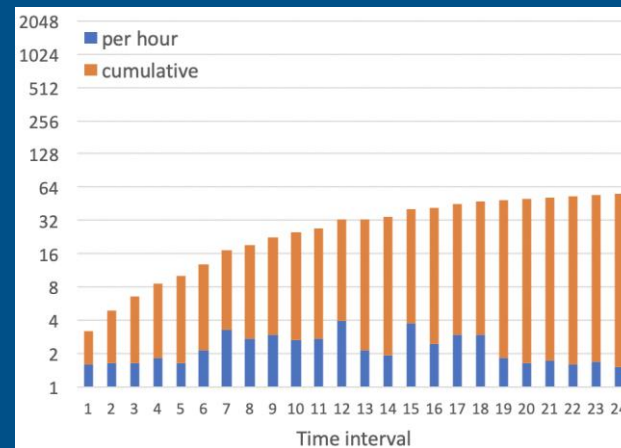
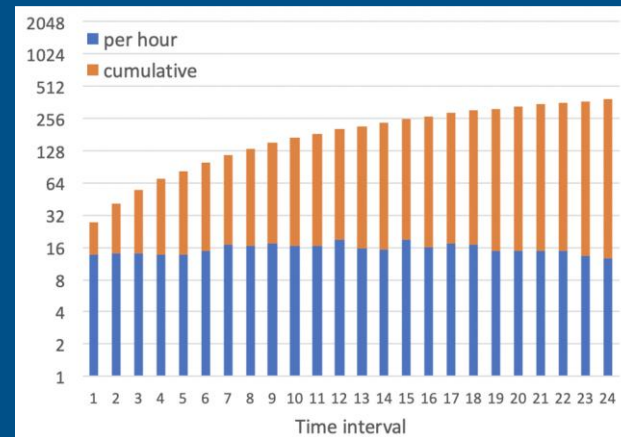
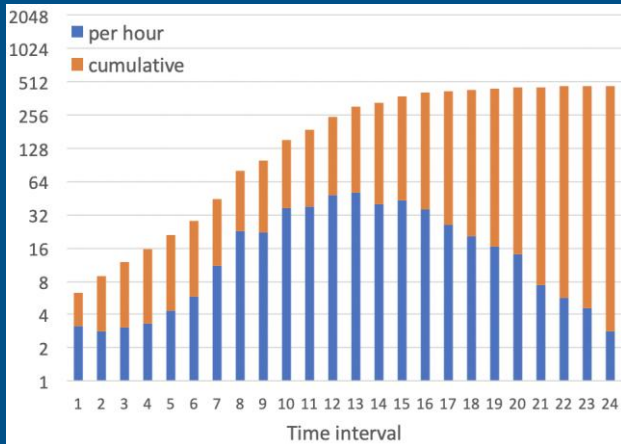
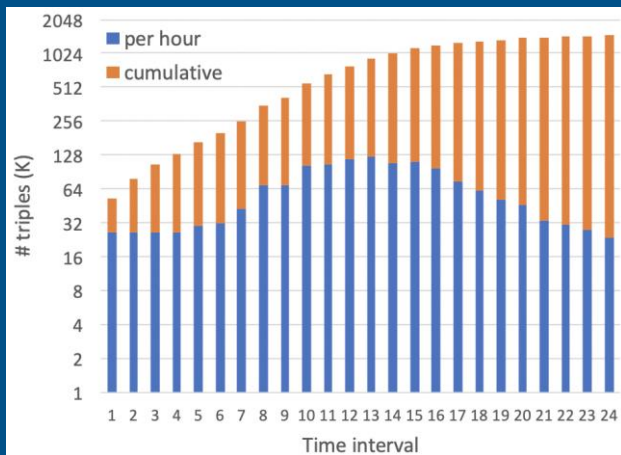
Size of KB

- Two one-week data snapshots—pre and post campus partial closure due to COVID-19 (early and late March 2020, respectively).
- ≈ 800 triples to annotate a building
- $\approx 2\text{M}$ pre-closure and $\approx 600\text{K}$ during closure triples required to represent “people”
- spatial-temporal triples per building proportional to its size and inhabitants (e.g., $\approx 5\text{M}$ for DBH)

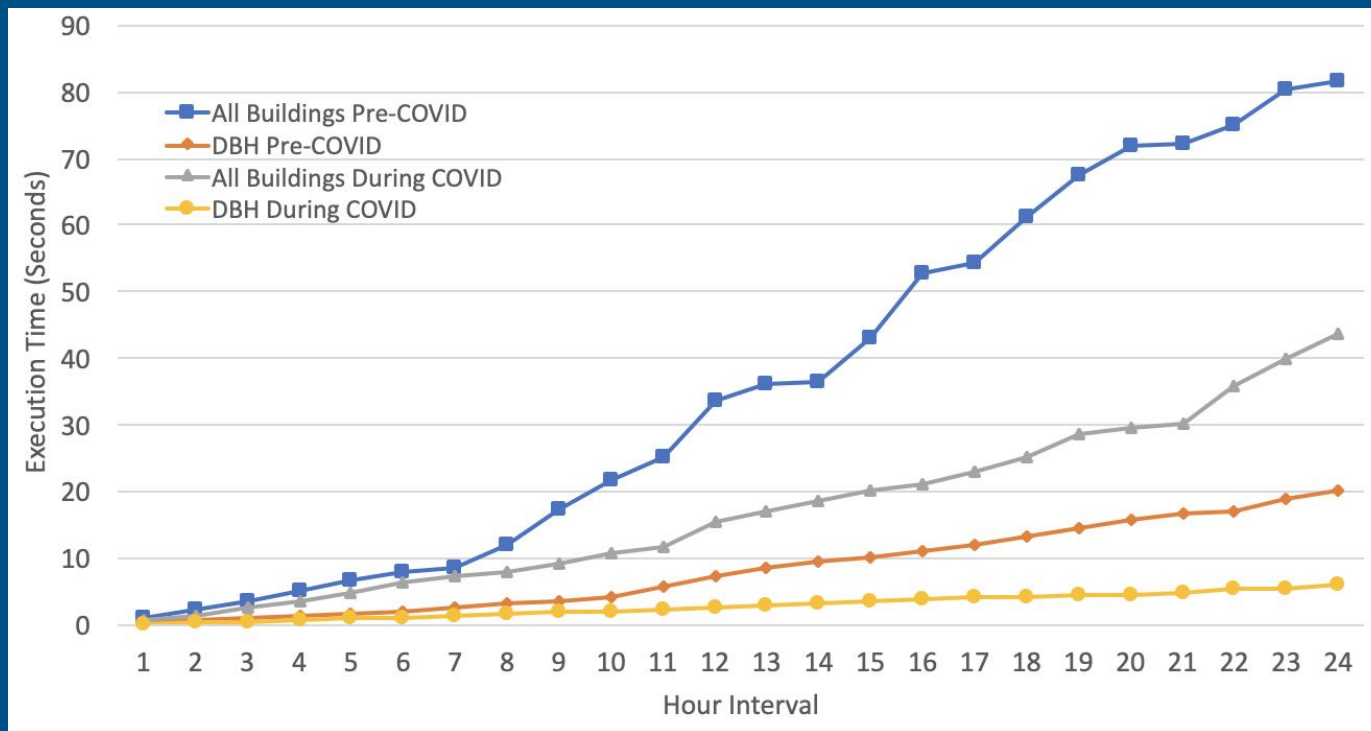
#triples	UCI-KB	DBH-KB	ICS-KB	ICS2-KB	FRH-KB
space	379	8,731	124	193	289
activity	-	21	0	0	105
people [†]	2,255,778	-	-	-	-
people [§]	584,052	-	-	-	-
spat-temp [†]	-	5,192,778	1,013,339	713,543	2,796,318
spat-temp [§]	-	2,487,957	283,925	69,809	949,746
total	135,609	7,689,466	1,297,388	783,545	3,746,353

TABLE I: Triples per KB. pre-campus closure[†], during closure[§].

Size vs. Time



Query Performance



Discussion and Future Direction

- **INHAB** schema enables **interoperable, inhabitant-aware** applications.
- Aligns with existing ontologies, supports **uncertainty** and **spatial-temporal reasoning**.
- **Limitation:** More fine-grained representation, more size, slower queries
 - Common tradeoff in KGs/ontologies to achieve interoperability
 - Mitigations: Storing only current triples (e.g., last day) in the operational KG, use efficient RDF/KG DBs
- **Future Work:** Enhancing annotation tool and optimizing knowledge graph storage.

Thank you!
Questions?