



The Urban Toolkit: A Grammar-based Framework for Urban Visual Analytics

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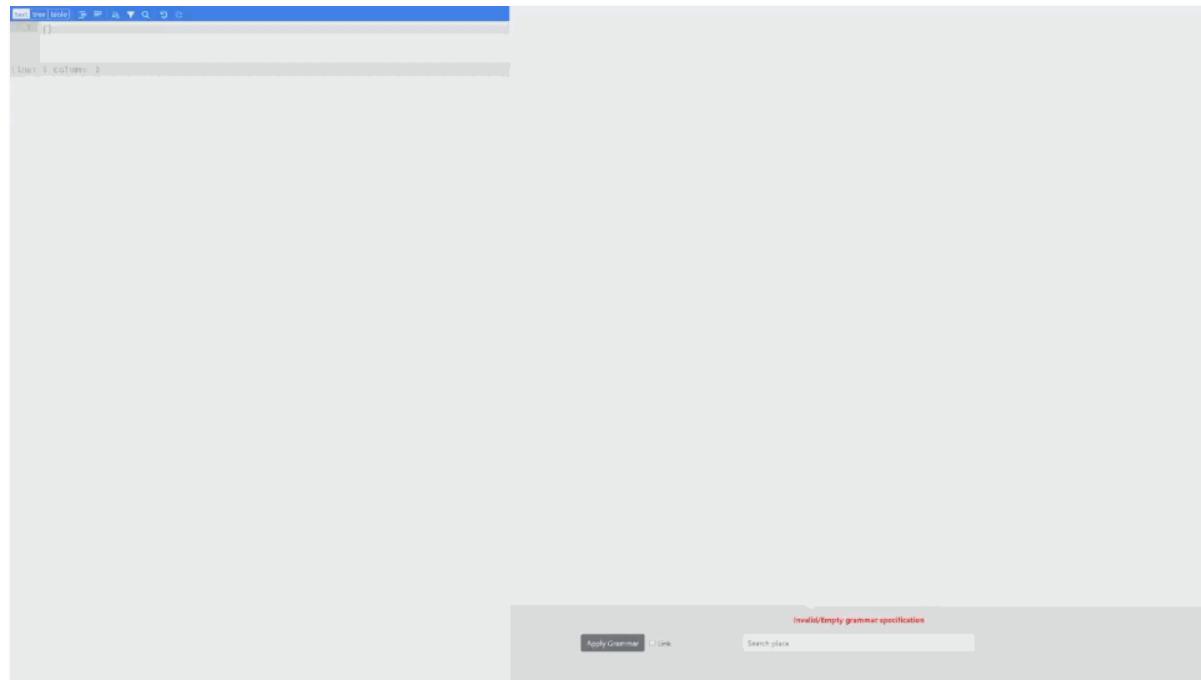
Fabio Miranda, University of Illinois Chicago

IEEE VIS 2023





The Urban Toolkit





Diverse data





Diverse data



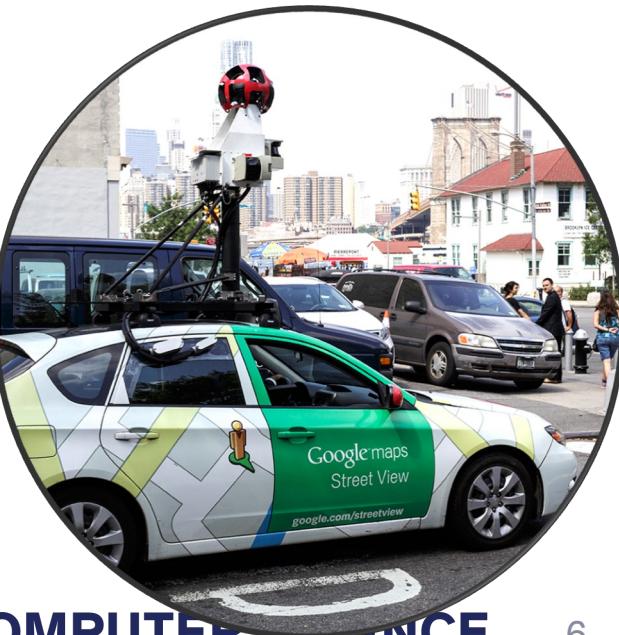


Diverse data





Diverse data



Occupational therapists

Urban noise experts

Transportation
engineers

Public health experts

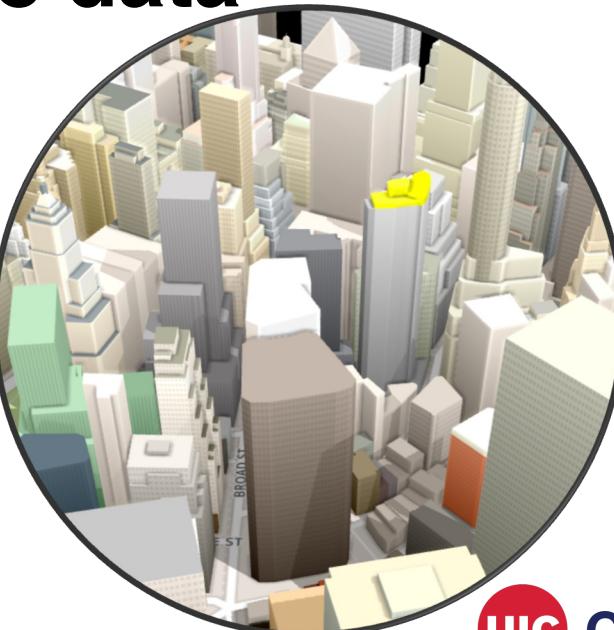
Environmental
scientists

Architects

Urban planners

Diverse users

Diverse data





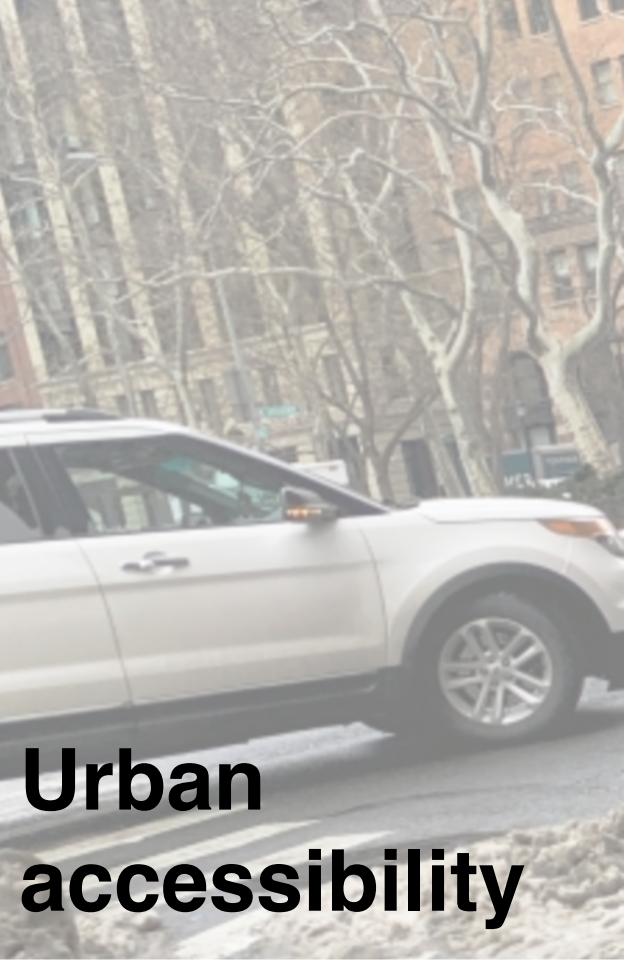
Diverse users

Diverse data

Diverse problems



UIC COMPUTER SCIENCE



Urban accessibility



Air and noise pollution



Sunlight access



Impact of climate change

Diverse problems

Diverse users

Diverse data





Urban visual analytics surveys

Visual Analytics in Urban Computing: An Overview

Yixian Zheng, Wenchao Wu, Yuanzhe Chen, Huamin Qu, *Member, IEEE*, and Lionel M. Ni, *Fellow, IEEE*

Abstract—Nowadays, various data collected in urban context provide unprecedented opportunities for building a smarter city through urban computing. However, due to heterogeneity, high complexity and large volumes of these urban data, analyzing them is not an easy task, which often requires integrating human perception in analytical process, triggering a broad use of visualization. In this survey, we first summarize frequently used data types in urban visual analytics, and then elaborate on existing visualization techniques for time, locations and other properties of urban data. Furthermore, we discuss how visualization can be combined with automated analytical approaches. Existing work on urban visual analytics is categorized into two classes based on different outputs of such combinations: 1) For *data exploration and pattern interpretation*, we describe representative visual analytics tools designed for better insights of different types of urban data. 2) For *visual learning*, we discuss how visualization can help in three major steps of automated analytical approaches (i.e., cohort construction; feature selection & model construction; result evaluation & tuning) for a more effective machine learning or data mining process, leading to sort of artificial intelligence, such as a classifier, a predictor or a regression model. Finally, we outlook the future of urban visual analytics, and conclude the survey with potential research directions.

Index Terms—Urban computing, visual analytics, visualization, visual learning, spatio-temporal, multivariate

1 INTRODUCTION

WITH the development of science and technology, urbanization process has been accelerating worldwide, which on one hand improves people's life quality, on the other hand gives rise to serious problems, such as environmental pollution, traffic congestion and ever-increasing

quite a few issues which have not been addressed satisfactorily. Recently, Zheng et al. [3] presented a survey on urban computing, which introduced general framework, key research problems, methodologies, and applications mainly based on automated data mining approaches. However, as

> 150 papers (Zheng et al., 2016)

A survey of urban visual analytics: Advances and future directions

Zikun Deng¹, Di Weng² (✉), Shuhan Liu¹, Yuan Tian¹, Mingliang Xu^{3,4}, and Yingcai Wu¹ (✉)

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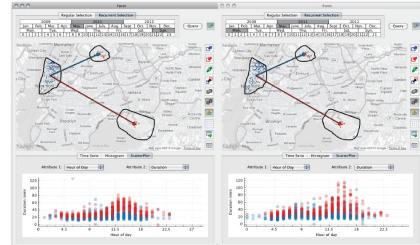
Abstract Developing effective visual analytics systems demands care in characterization of domain problems and integration of visualization techniques and computational models. Urban visual analytics has already achieved remarkable success in tackling urban problems and providing fundamental services for smart cities. To promote further academic research and assist the development of industrial urban analytics systems, we comprehensively review urban visual analytics studies from four perspectives. In particular, we identify 8 urban domains and 22 types of popular visualization, analyze 7 types of computational method, and categorize existing systems into 4 types based

knowledge and expertise into the analysis loop. Thus, urban visual analytics [7] is used to empower urban experts using a combination of intuitive data visualization and fast computational methods, enabling experts to visually and interactively perceive, explore, manipulate, and reason about urban data [8].

When developing an urban visual analytics approach, practitioners like urban analysts and researchers may have the following four questions:

1. Which urban *domain problems* have been solved or remain unsolved by visual analytics?
2. What *visualization* techniques have been applied to visually interpret urban data?

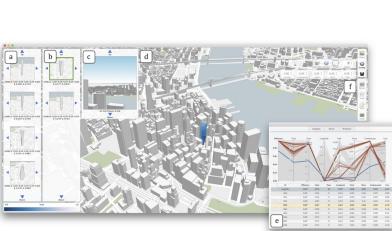
> 200 papers (Deng et al., 2022)



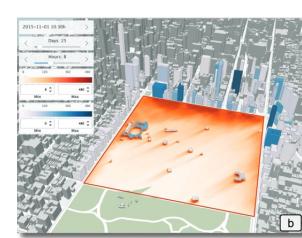
TaxiVis
(Ferreira et al., 2013)



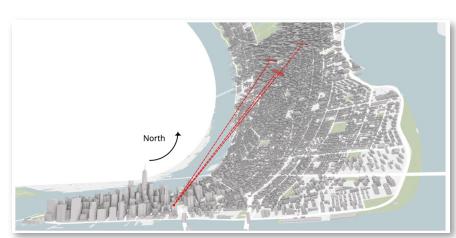
Urbane
(Ferreira et al., 2015)



Catalogue
(Doraiswamy et al., 2015)



Shadow Profiler
(Miranda et al., 2019)



UrbanRama
(Chen et al., 2020)



UTK
(Moreira et al., 2023)

2014

2016

2018

2020

2022

2023

2013

2015

2017

2019

2021

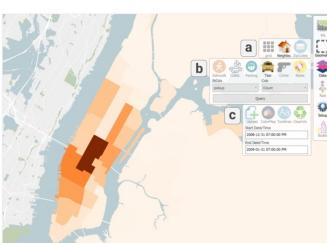
Taxi Patterns
(Doraiswamy et al., 2016)



Urban Pulse
(Miranda et al., 2016)



Raster-Join
(Doraiswamy et al., 2018)

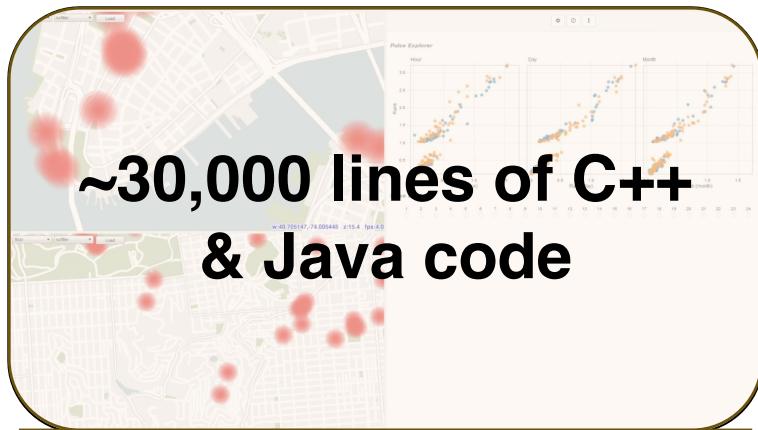


Urban Mosaic
(Miranda et al., 2020)

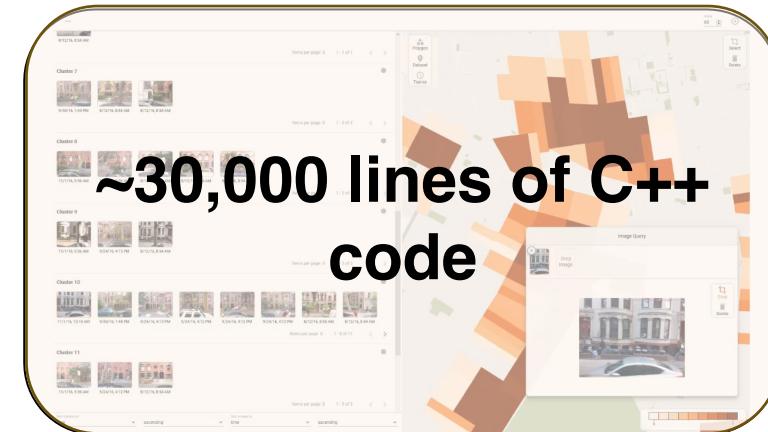


Urban Rhapsody
(Rulff et al., 2022)

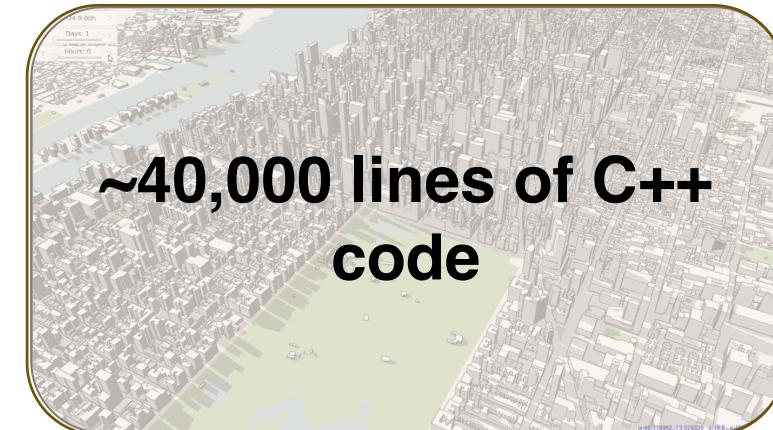




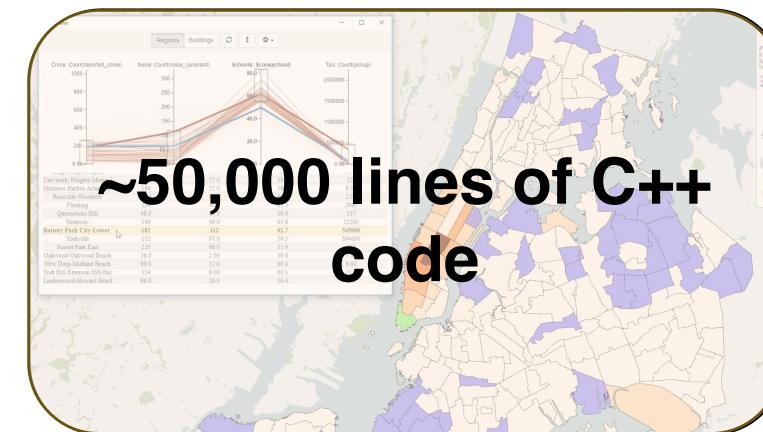
Urban Pulse:
Large-scale data mining of social media data



Urban Mosaic:
Interactive exploration of large imagery data



Shadow Profiler:
City-scale assessment of sunlight access



Urbane:
Interactive exploration of large data





Current state of urban tools

Lack of flexibility:

Tools and techniques are not translatable to other domains or regions
(Acuto et al., 2018)

Lack of extensibility:

Hard to add new functionalities needed for specific workflows
(Lobo et al., 2020)

Lack of reproducibility:

Results are rarely reproducible
(Ziegler and Chasins, 2023)

Lack of scalability:

Analyses are rarely interactive, limiting the number of hypothesis from the exploratory process
(Ziegler and Chasins, 2023)

Lack of accessibility:

Tools and techniques are often difficult to use, limiting stakeholders' ability to conduct large-scale analysis
(Kontokosta, 2021)





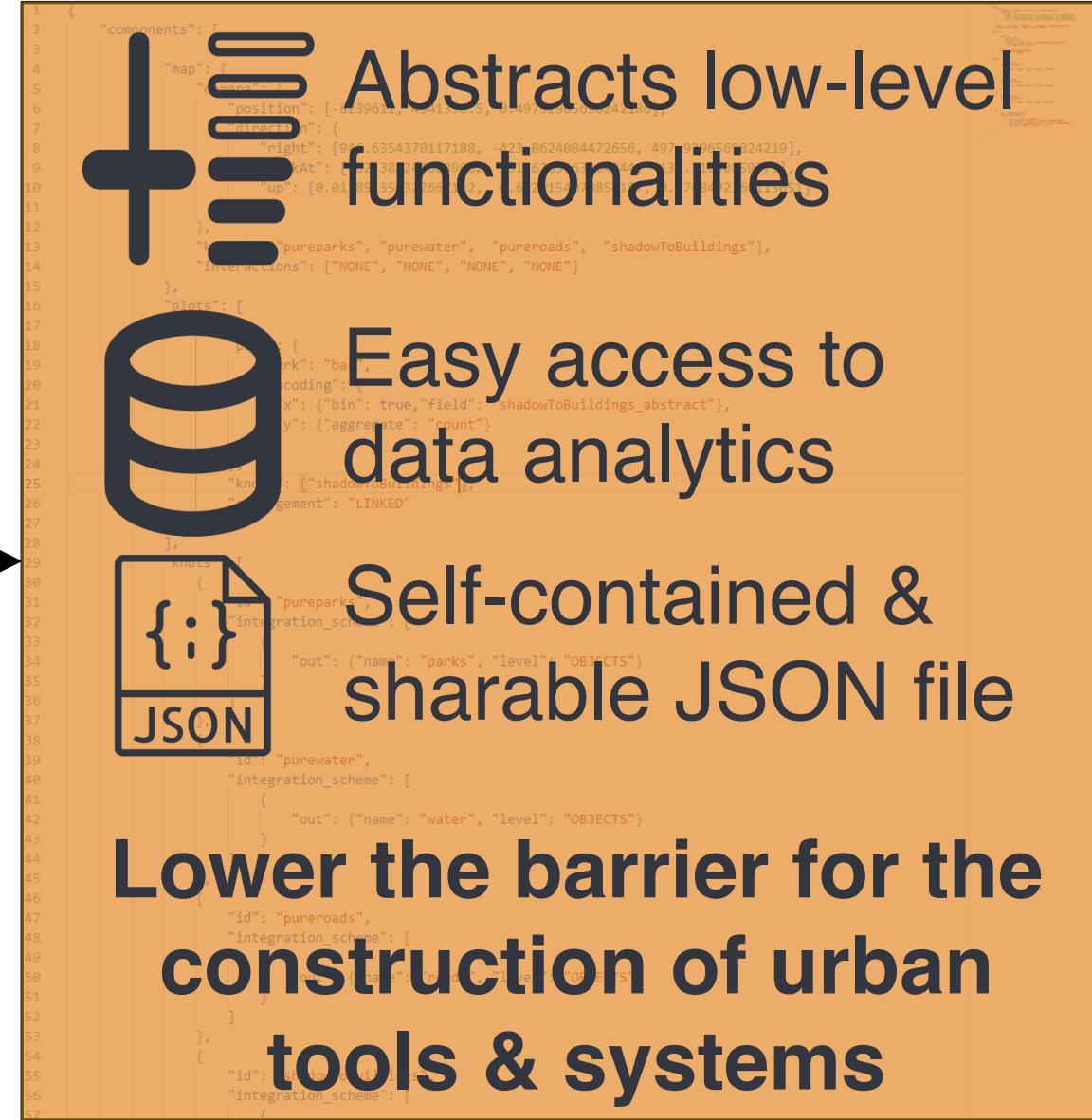
How to support (1) accessible,
(2) flexible, (3) extensible, (4)
reproducible and (5) scalable
tools across domains and users?



```

1 #include "UrbaneMapView.hpp"
2
3 #include <QApplication>
4 #include "../MapView/BuildingRenderingLayer.hpp"
5
6 #include "../MassingGeneration/massinggeneration.h"
7 #include "../Util/ColorMapDivergent.hpp"
8 #include "UrbaneManager.hpp"
9
10 #include <QElapsedTimer>
11 #include <QThread>
12 #include <QDir>
13
14 #include <vector>
15
16 UrbaneMapView::UrbaneMapView(const QString &filename, const QRectF &vp, QWidget *parent)
17 | : MapView(filename, vp, parent), graphLayer(NULL)
18 {
19     initialized = false;
20     skyExposureData = false;
21     this->centerIndex = GridIndex(1024, 1024);
22     this->currentLayer = NULL;
23     this->lotUpdate = true;
24 }
25
26 UrbaneMapView::~UrbaneMapView() {}
27
28 void UrbaneMapView::initializeGL()
29 {
30     if(!initialized) {
31         MapView::initializeGL();
32         this->buildingScore.initComputeShader();
33         this->skyScore.initComputeShader();
34     }
35     initialized = true;
36 }
37
38 void UrbaneMapView::paintGL()
39 {
40     this->showOsd(false);
41
42     // Lot data initialization in manager
43     // TODO Don't know of a better place to do this
44     if(lotUpdate && this->parcelLayer->isDataReady()) {
45         updateLotDataDB();
46         lotUpdate = false;
47     }
48
49     UrbaneManager *manager = UrbaneManager::getInstance();
50     QPair<RenderingOperation, UIOperation> state = manager->getState();
51
52     RenderingOperation operation = state.first;
53     UIOperation what = state.second;
54     switch(operation) {
55     case RenderingOperation::UpdateVis:
56     {
57         bool updateFunction = false;

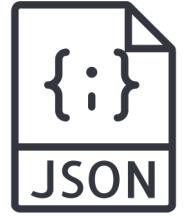
```





Views

```
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        "camera": {  
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        },  
        ...  
      },  
      "plots": [...],  
      "knots": [...]  
    }  
  ]  
}
```

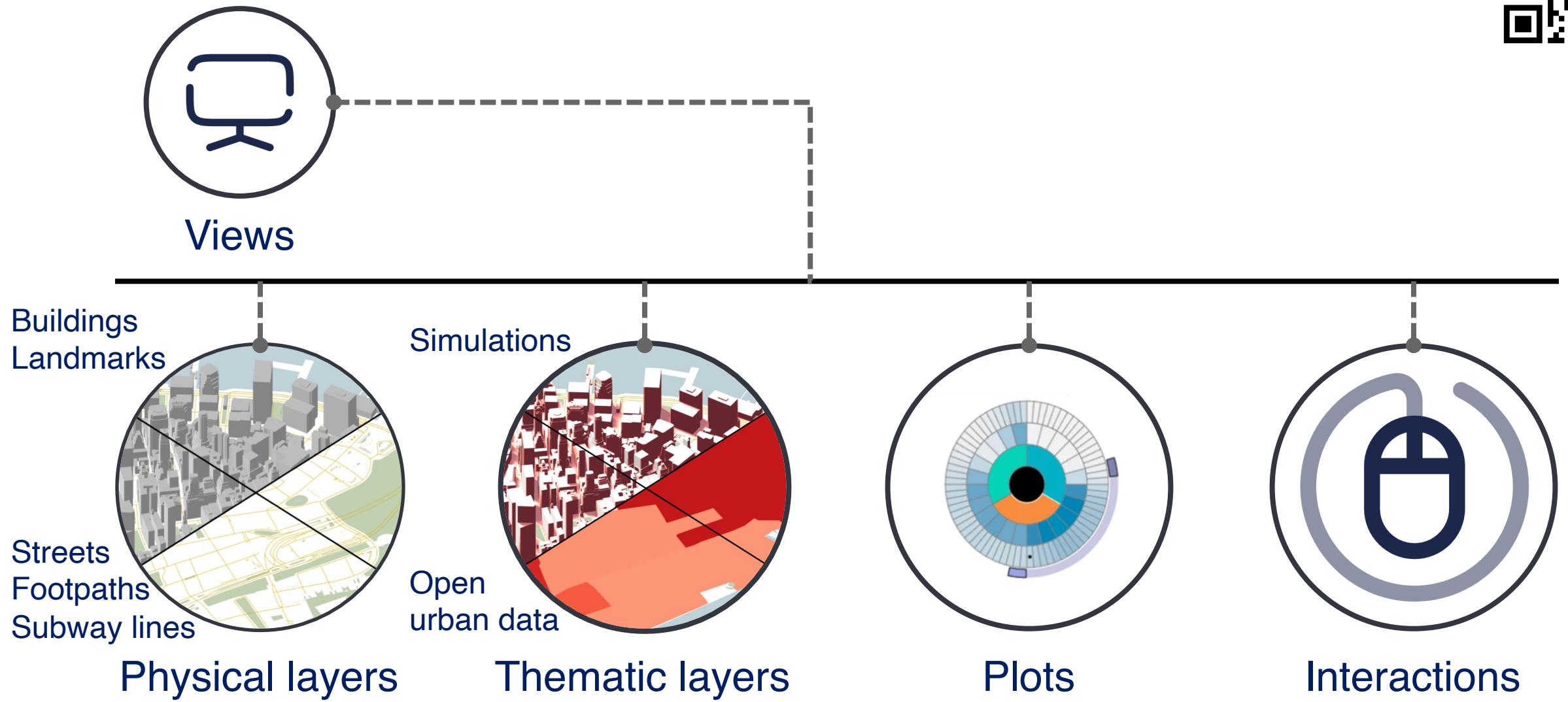


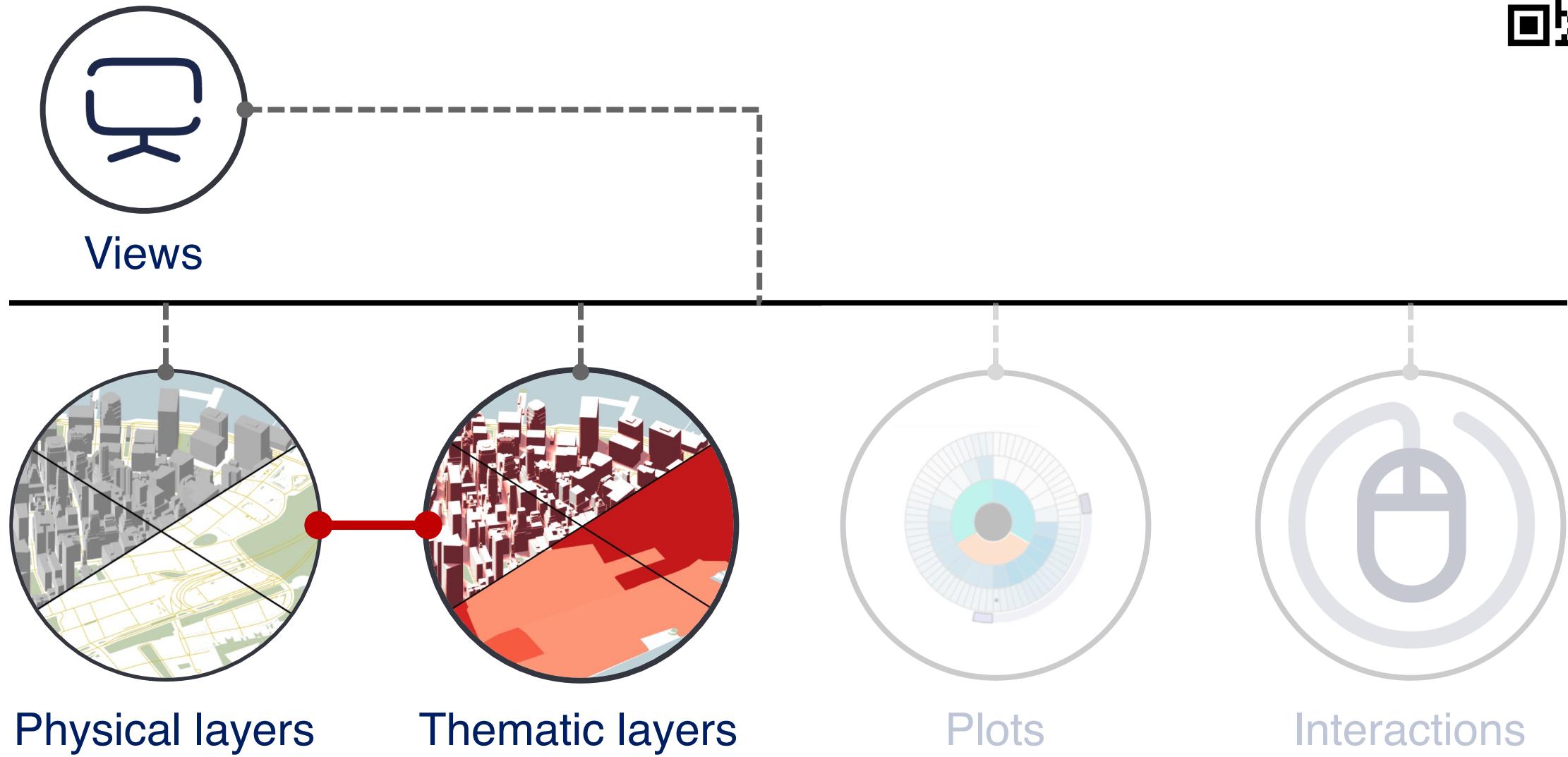
{;}

JSON



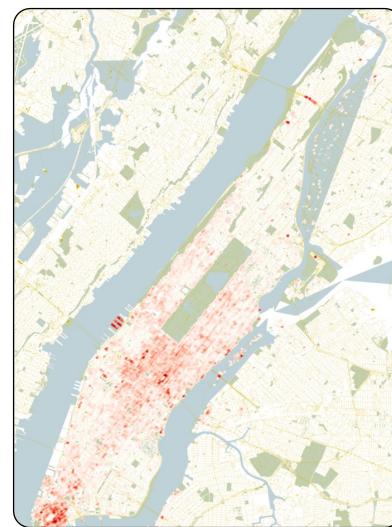
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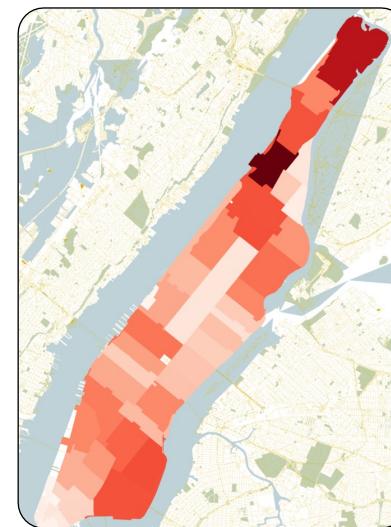




Views



Grid



Areas



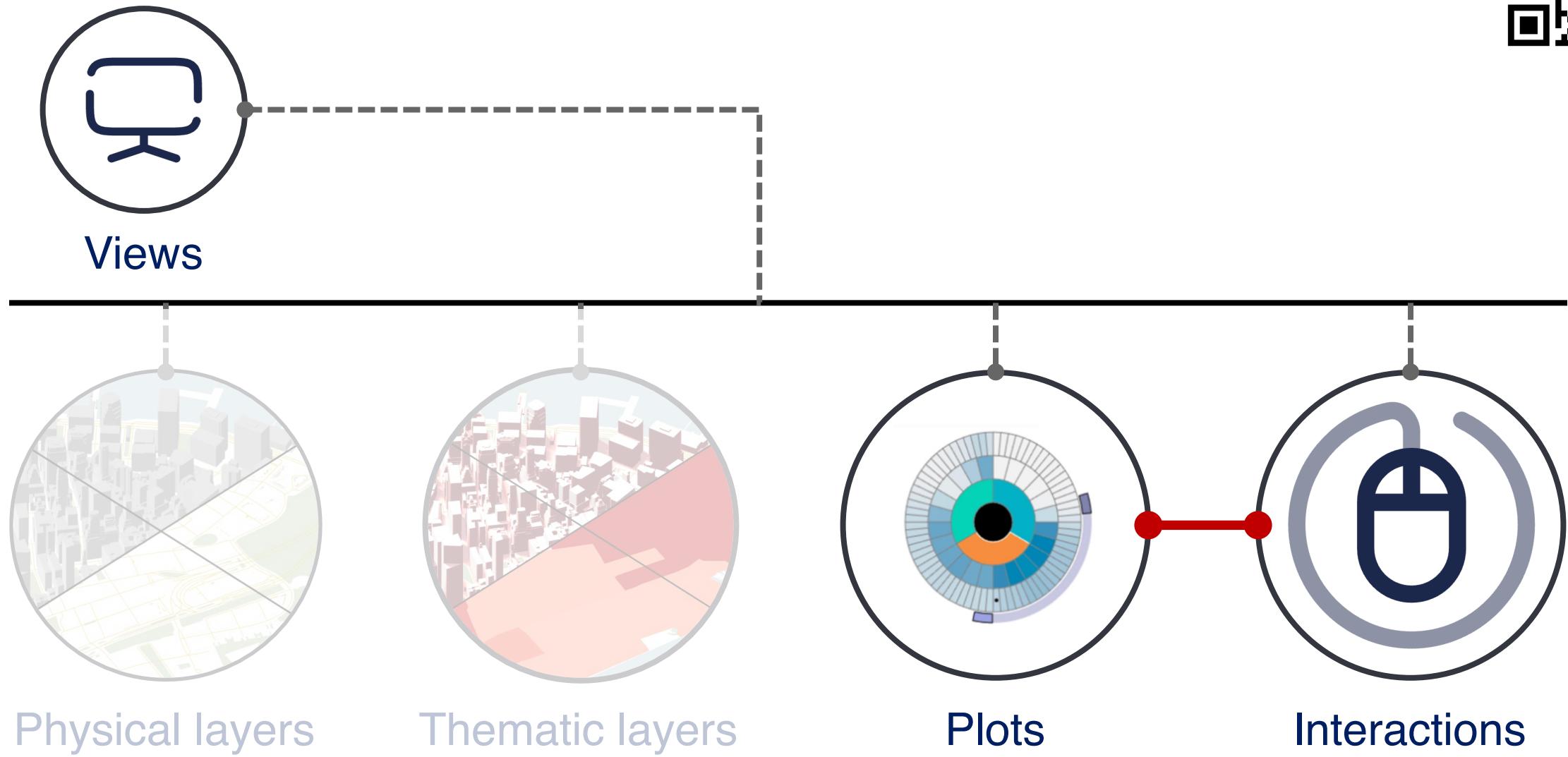
Buildings



Networks

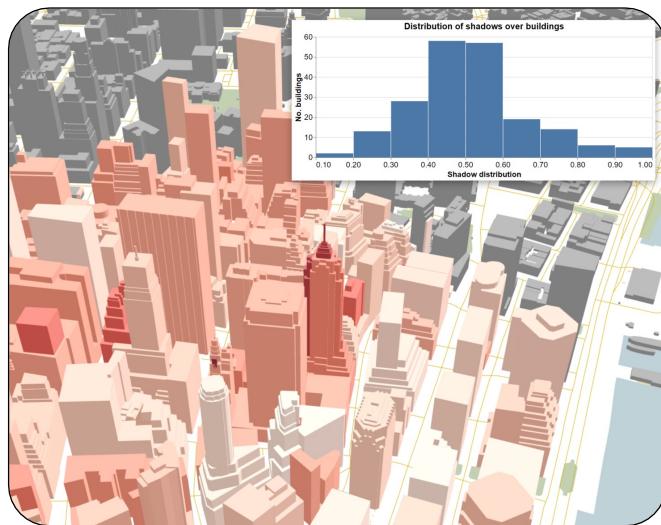


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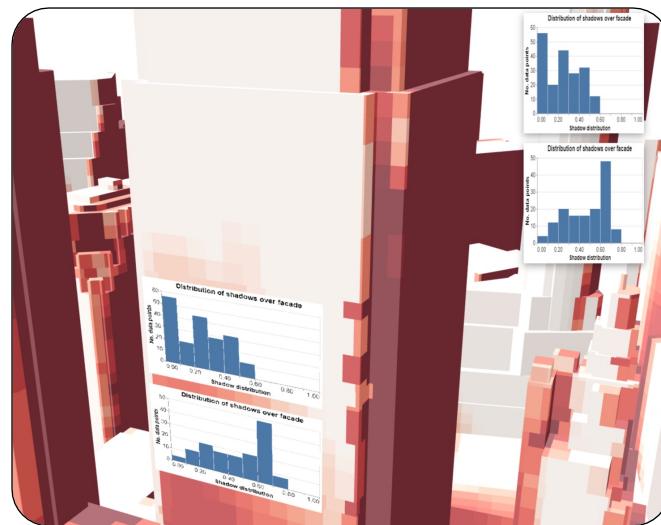


Views

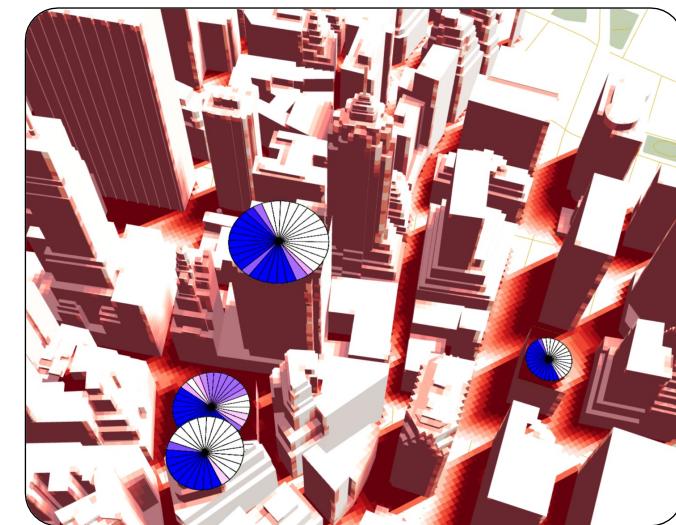


Juxtaposed

Taxonomy by Mota et al. (VIS 2022)



Embedded (surface)

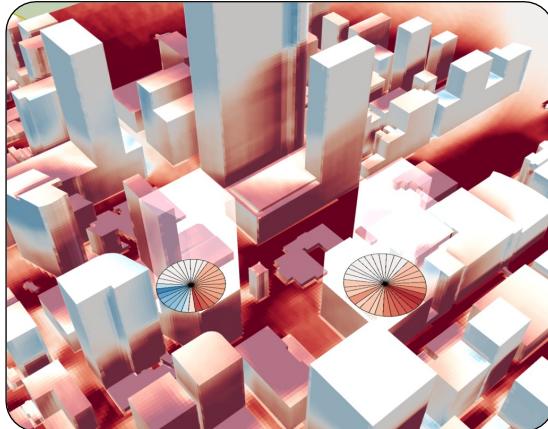
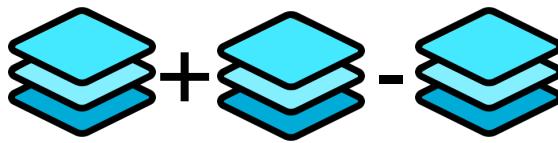


Embedded (crosscut)

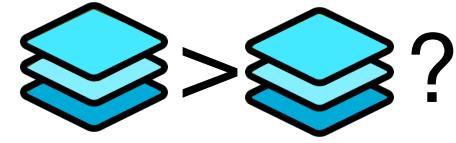


Operating with layers

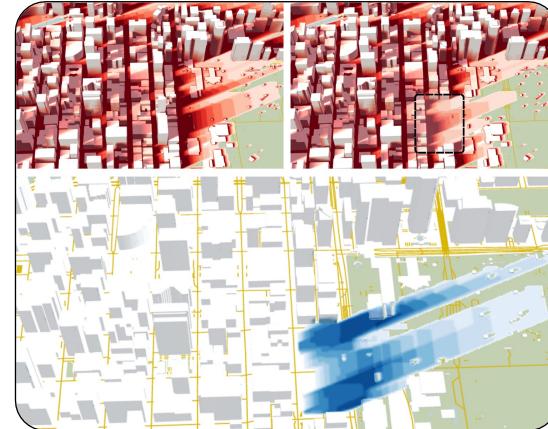
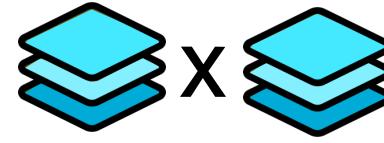
Arithmetic operations



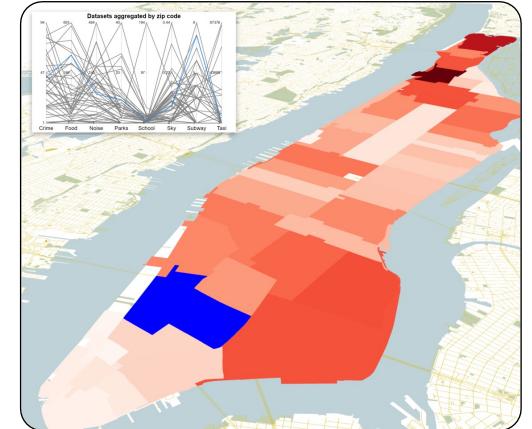
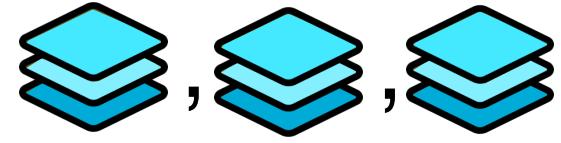
Logical operations



What-if analyses



Data exploration



Editing the
JSON
specification

Visualization

Invalid/Empty grammar specification

Apply Grammar



Search place



Data functionalities

- Converting data: Pandas' DataFrame, GeoJSON, shapefile, CSV and Protocolbuffer Binary Format (PBF).
- Loading data: OpenStreetMap (OSM).
- Generating data: Sunlight access simulation (Miranda et al., TVCG 2018).

```
import utk
uc = utk.OSM.load('Chicago,USA', layers=['buildings'])
uc.save('chicago')
uc.view()
```



Evaluation

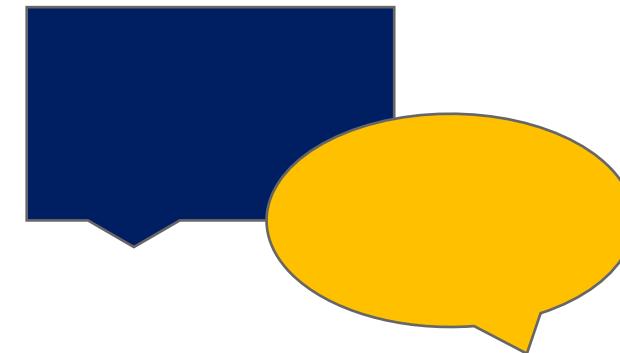
Case studies

- Motivated by real-world problems
- Inspired by previous collaborations



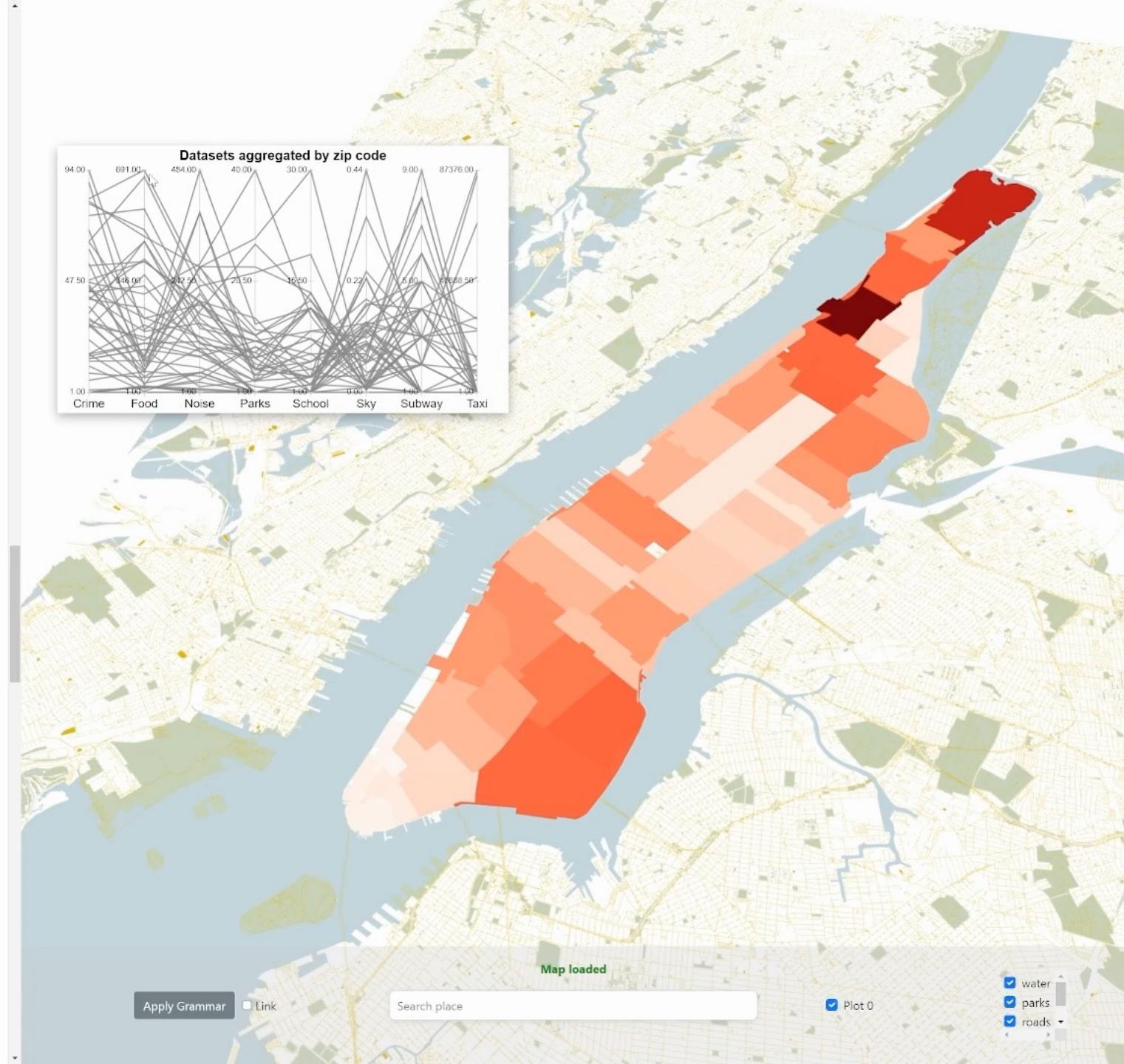
Experts' feedback

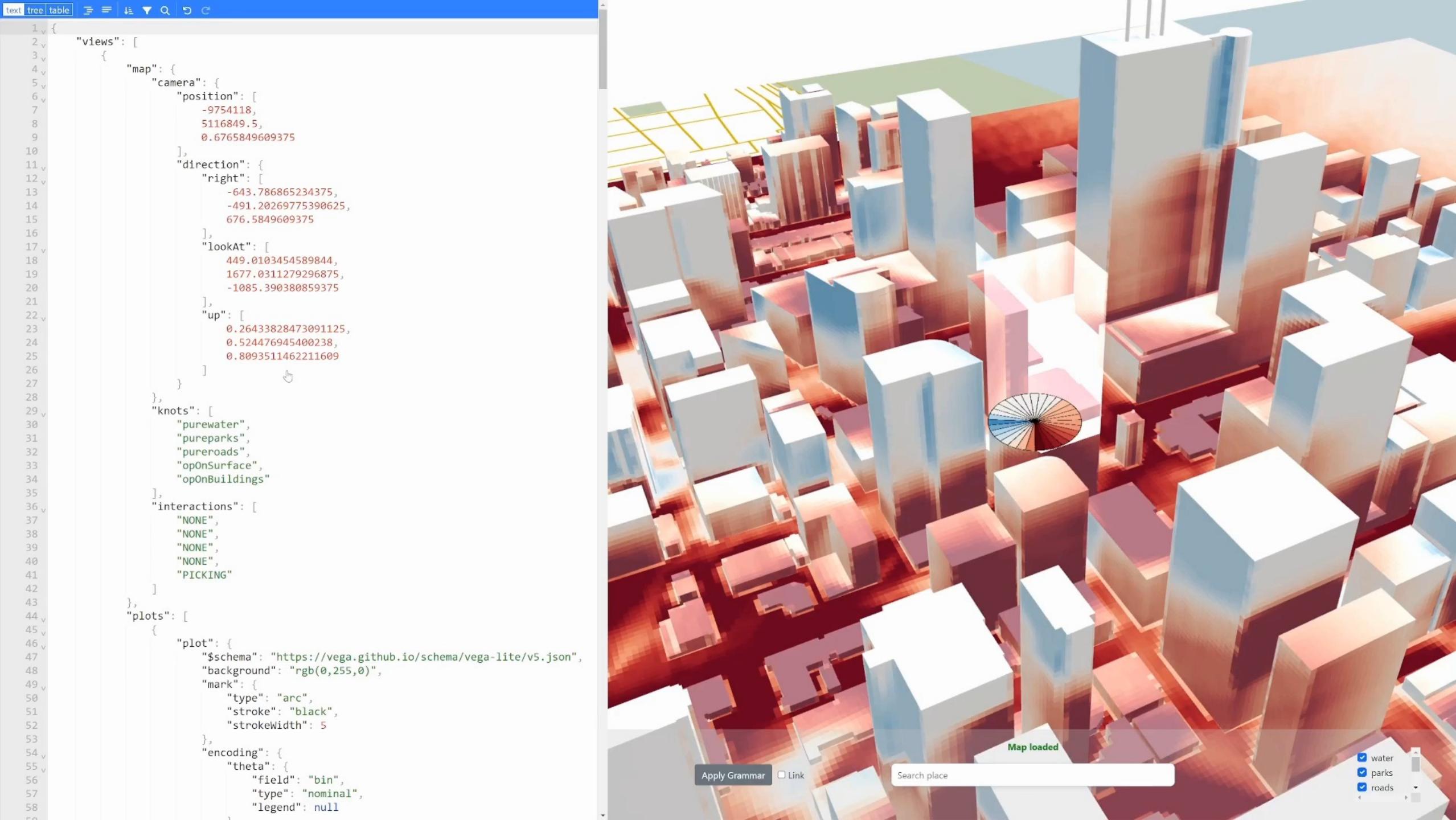
- One-hour semi-structured interviews
- Their perspectives on UTK's usability, limitations and needed features



```
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241 v  
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291 v  
292 v  
293
```

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 }
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 "restaurantsToZip",
 "subwayToZip",
 "schoolToZip",
 "skyToZip",
 "parksToZip"
],
"arrangement": "LINKED",
"interaction": "HOVER"
},
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 "linkingScheme": [
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 "order": 1
 },
 {
 "value": "noiseToZip",
 "order": 2
 }
]
 }
]







Experts' feedback

“... facilitates engagement not only across disciplines, but also across urban communities” (Weather scientist)

“... researchers could easily share their visualizations instead of cumbersome GIS files” (Urban planner)

“3D makes it more attractive to users and a great tool for communication” (Urban planner)

“It would require training to educate people on the grammar, with examples to showcase its use” (Urban planner)



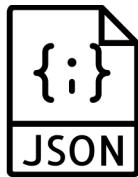
Future opportunities

- Incorporate other functionalities that have been highlighted in previous urban visual analytics works:
 - Computational topology
 - Wavelet
 - Techniques for model inspection



Transportation experts

- What-if scenarios
- Model inspection
- ...



Weather experts

- What-if scenarios
- Model inspection
- Data wrangling
- ...



The Urban Toolkit

- Open urban data
- Modeling & simulation results
- Crowdsourced data
- Urban sensing data



Communities

- Engagement
- ...



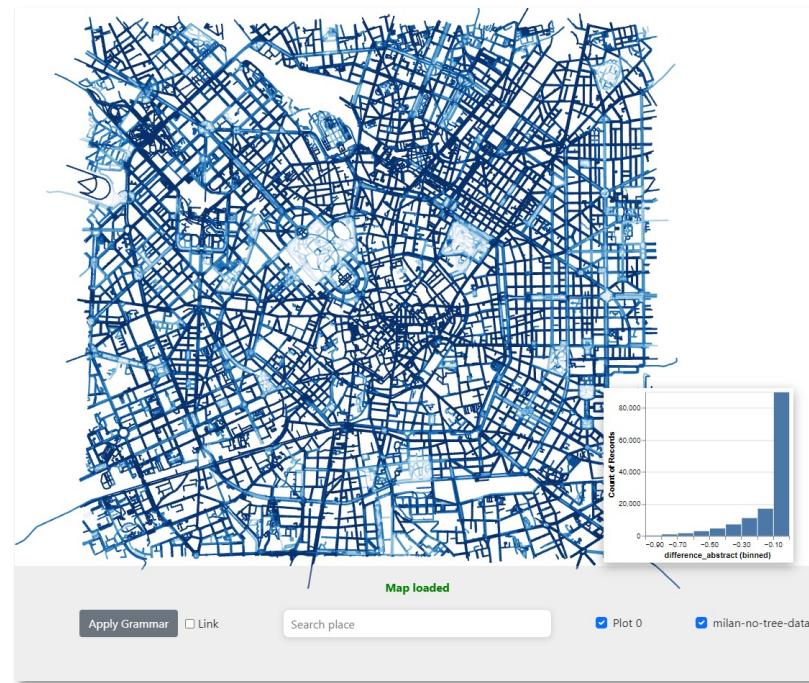
Policy makers

- What-if scenarios
- Engagement
- ...

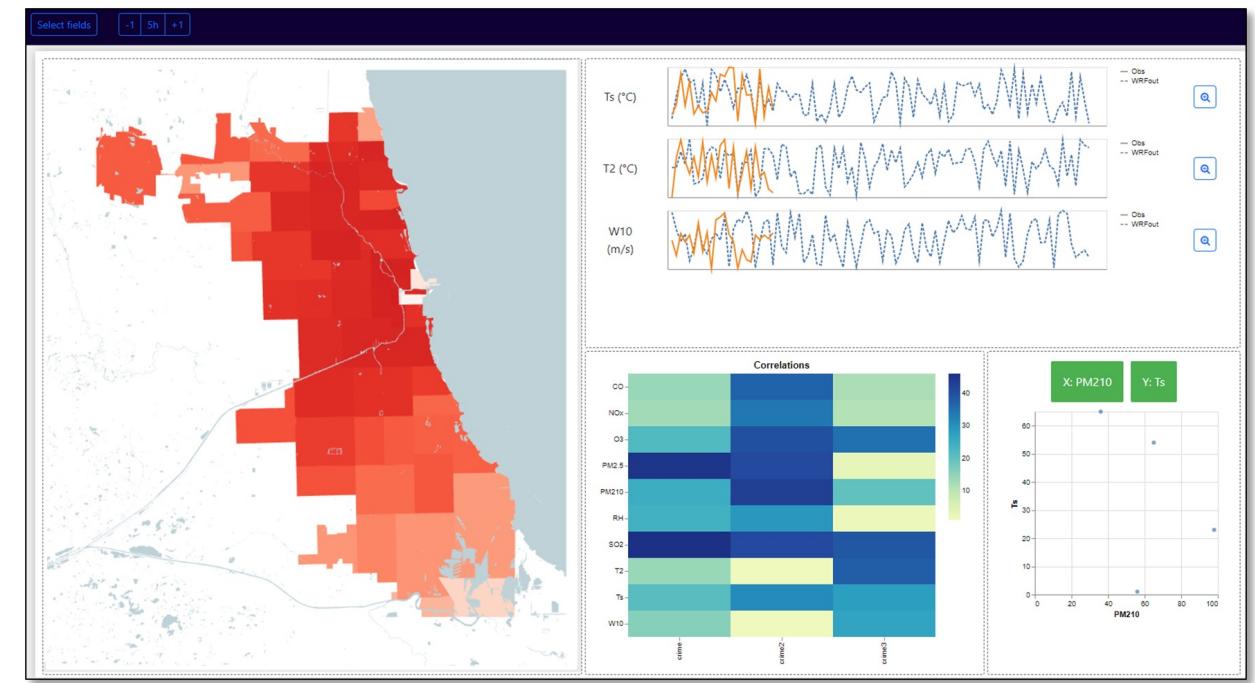




Ongoing works extending UTK

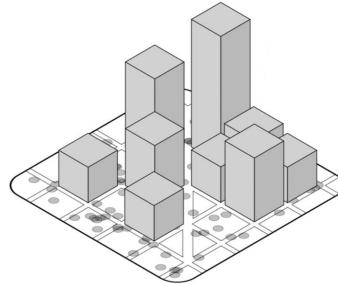


Urban accessibility



Environmental justice





UrbanTK

.....

Code & tutorials:
urbantk.org



IIT UFF



Gustavo Moreira, University of Illinois Chicago
Maryam Hosseini, Massachusetts Institute of Technology
Md Nafiul Alam Nipu, University of Illinois Chicago
Marcos Lage, Universidade Federal Fluminense
Nivan Ferreira, Universidade Federal de Pernambuco
Fabio Miranda, University of Illinois Chicago

The Urban Toolkit
A Grammar-based Framework for Urban Visual Analytics

Getting Started GitHub Tutorials

While cities around the world are looking for smart ways to channel new advances in data collection, management, and analysis to address their day-to-day problems, the complex nature of urban issues and the overwhelming amount of available structured and unstructured data have posed significant challenges in translating these efforts into actionable insights. In the past few years, urban visual analytics tools have significantly helped tackle these challenges. With this in mind, we present the Urban Toolkit, a flexible and extensible visualization framework that enables the easy authoring of web-based visualizations through a new high-level grammar specifically built with common urban use cases in mind.

The toolkit is described in the [paper](#):
The Urban Toolkit: A Grammar-based Framework for Urban Visual Analytics
Gustavo Moreira, Maryam Hosseini, Md Nafiul Alam Nipu, Marcos Lage, Nivan Ferreira and Fabio Miranda
IEEE Transactions on Visualization and Computer Graphics (Accepted at IEEE VIS 2023, to appear)