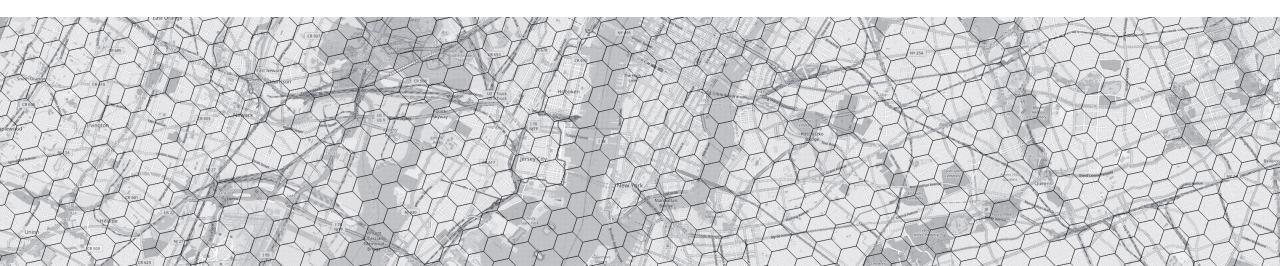
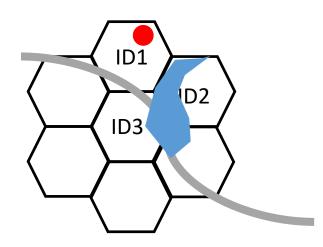


A Planet Scale Spatial-Temporal Knowledge Graph Based On OpenStreetMap And H3 Grid

Martin Böckling, Heiko Paulheim & Sarah Detzler ESWC 2024, 26.05.2024



Background



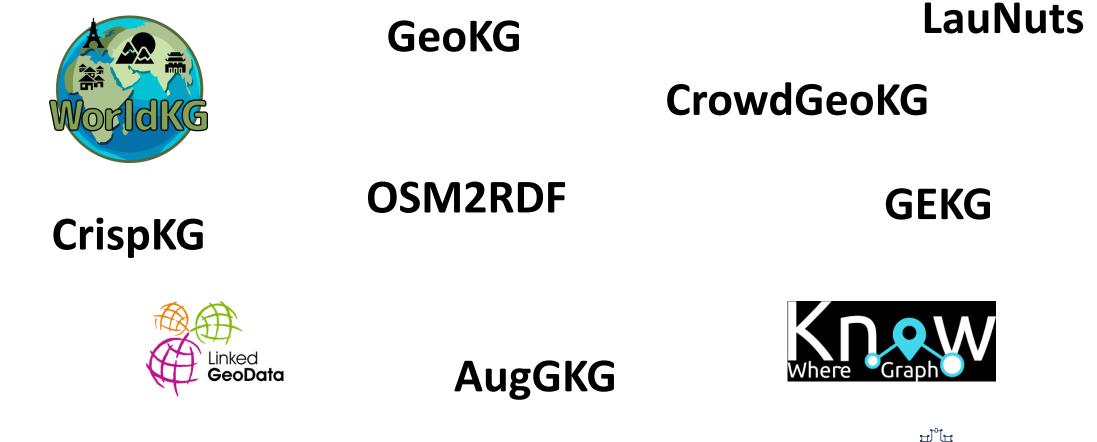
Grid Cell ID	hasRiver	hasCampfire	Wildfire
ID1	True	True	False
ID2	False	False	True
ID3	True	False	True

- Use Grid cells as single elements within datasets
- No inclusion of surrounding elements
- Limited data base



State of the art

Spatio-(Temporal) Knowledge Graphs





State of the art

Spatio-(Temporal) Knowledge Graphs - WorldKG

WorldKG constructs a Spatial Knowledge Graph using OpenStreetMap transposing the features from OpenStreetMap

- WorldKG contains the data for the complete planet for one time snapshot
- Contains 828 million triples and 113 million entities
- Each geometry associated to a OpenStreetMap feature is translated into a Point WKT format





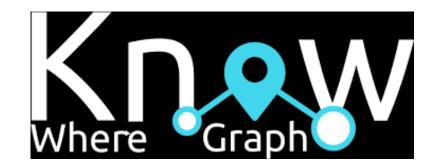
State of the art

Spatio-(Temporal) Knowledge Graphs - KnowWhereGraph

KnowWhereGraph contains a Spatio-Temporal Knowledge Graph modeling multiple datasets into the Knowledge Graph

- KnowWhereGraph models thematic datasets (Wildfires, Earthquakes, ...) and Place Centric datasets (S2 Grid, ZIP codes, ...) over time span
- Contains 13+ billion triples
- Models place centric datasets and thematic datasets based on spatial predicates

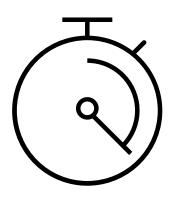
Janowicz, K., et al. (2022). Know, Know Where, KnowWhereGraph: A densely connected, cross-domain knowledge graph and geo-enrichment service stack for applications in environmental intelligence. *AI Magazine*, *43*(1), 30-39.

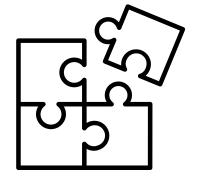




Motivation & Research Goals







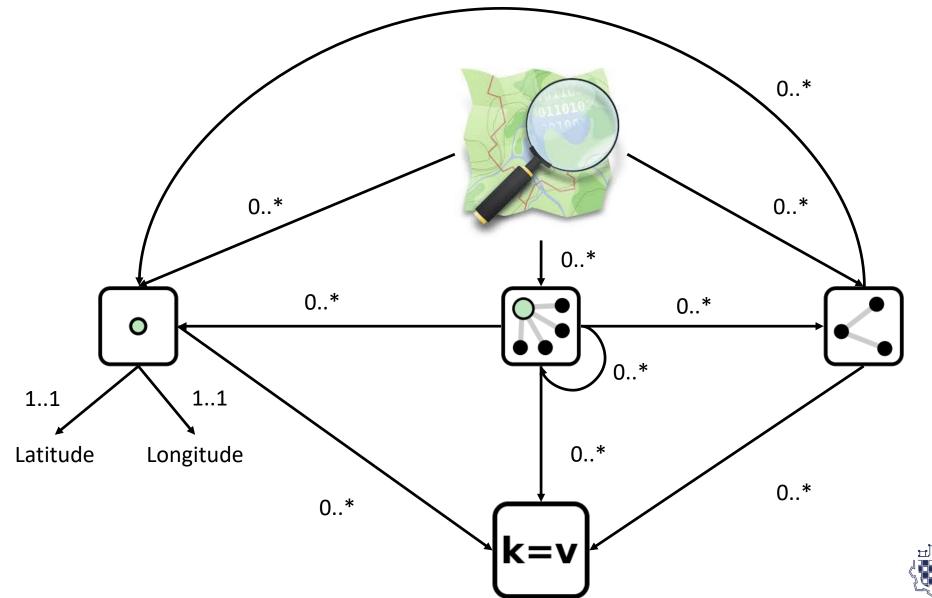
Cover spatial data over global extent

Provide temporal dimension to KG

Provide extensibility to KG

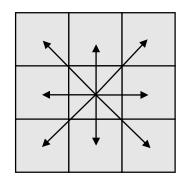


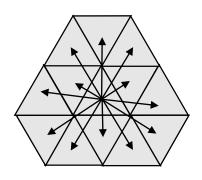
OpenStreetMap data structure

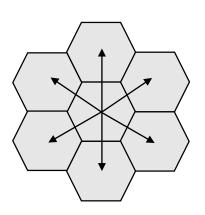




Extensibility through spatial grid

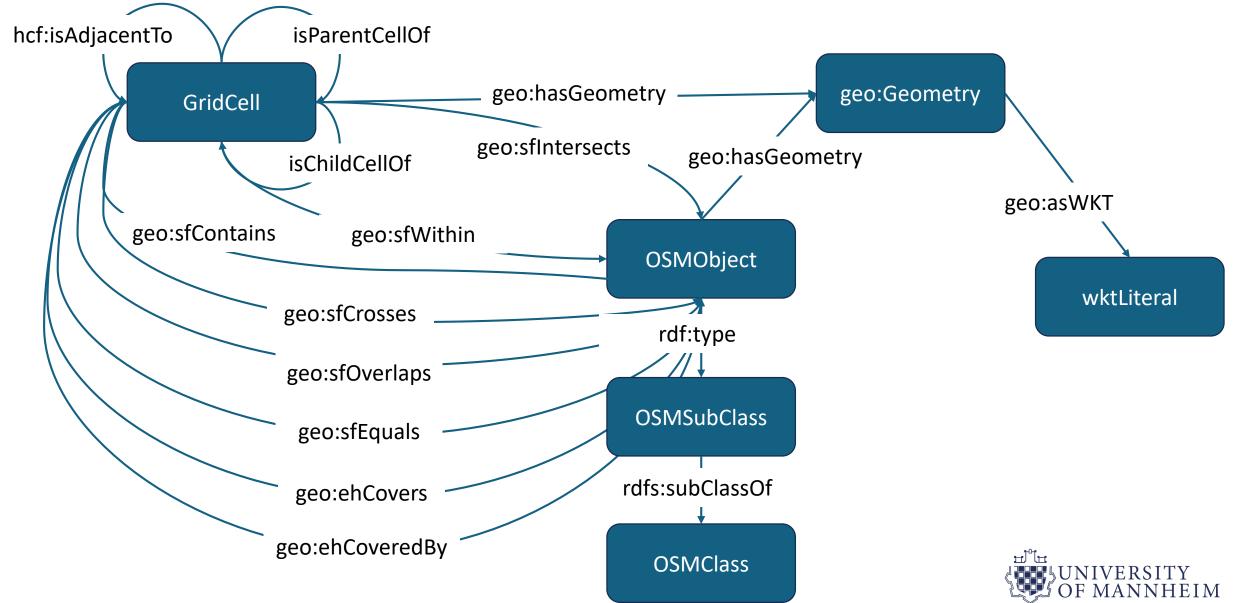




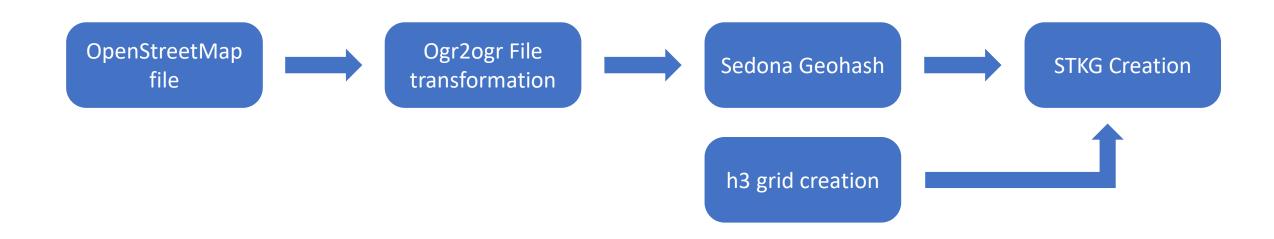


- Selection for regular grids between triangular, square and hexagonal grids
- Usage of hierarchical spatial grid to make sure global representation on different resolutions
- Selection to use hexagonal grid system
 - More accurate representation during tesselation of space
 - Modeling of neighbor information has same distance to all neighboring cells
- Selection to use h3 grid system for a hierarchical discrecte global grid
- Grid aggregation possibility based on population density

Knowledge Graph ontology

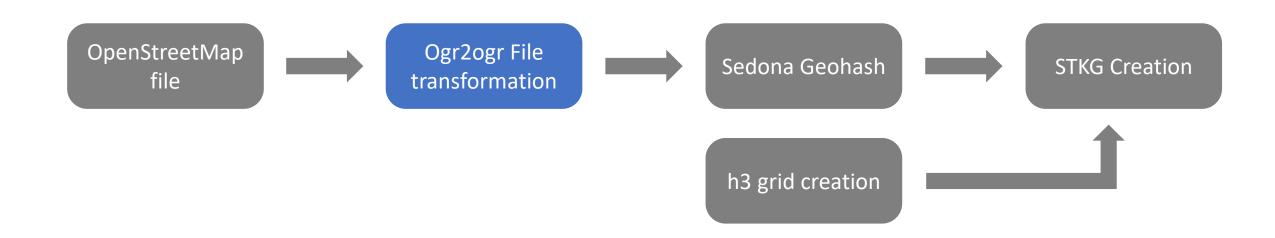


Spatio-Temporal Knowledge Graph creation





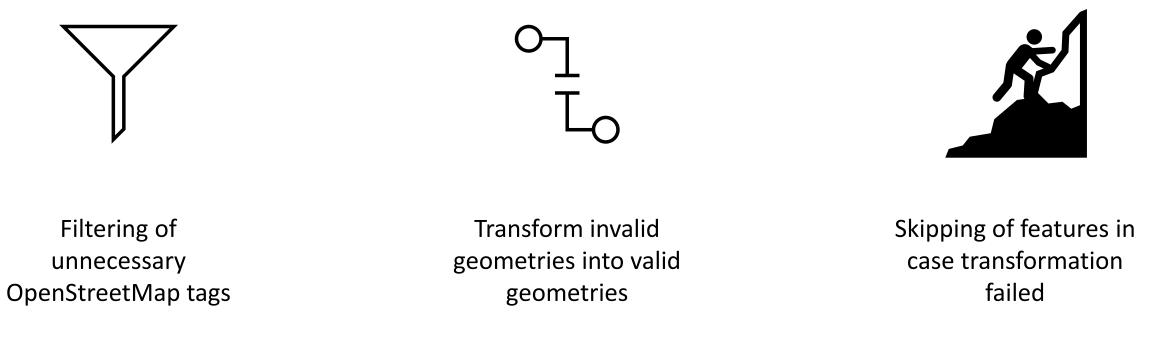
Spatio-Temporal Knowledge Graph creation





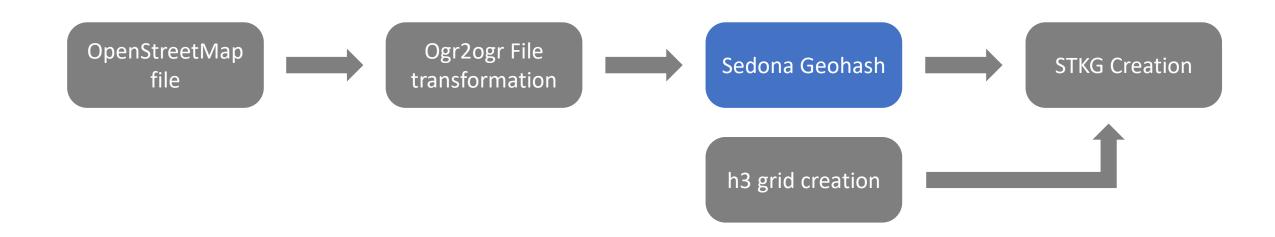
OpenStreetMap to GeoParquet

Transform the pbf file into a GeoParquet file to support processing of Apache Sedona for Spatial Temporal Knowledge Graph creation





Spatio-Temporal Knowledge Graph creation





Geohash writer with Apache Sedona

Possibility to push queries down to GeoParquet based on bbox property in GeoParquet metadata

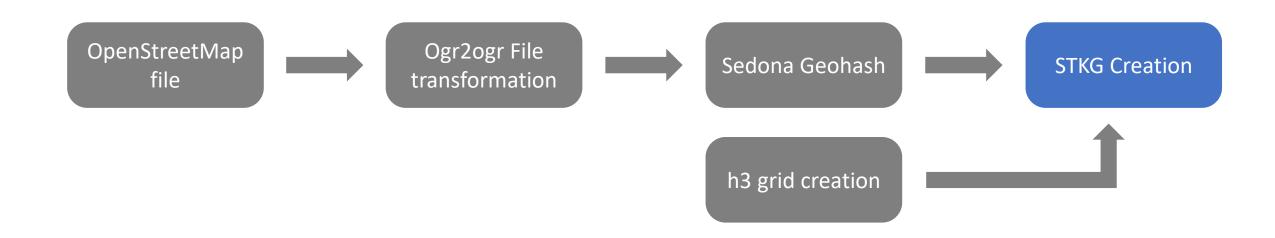
100



Increased query performance with sorted geohashes Query window constructed based on join predicate and geometry location

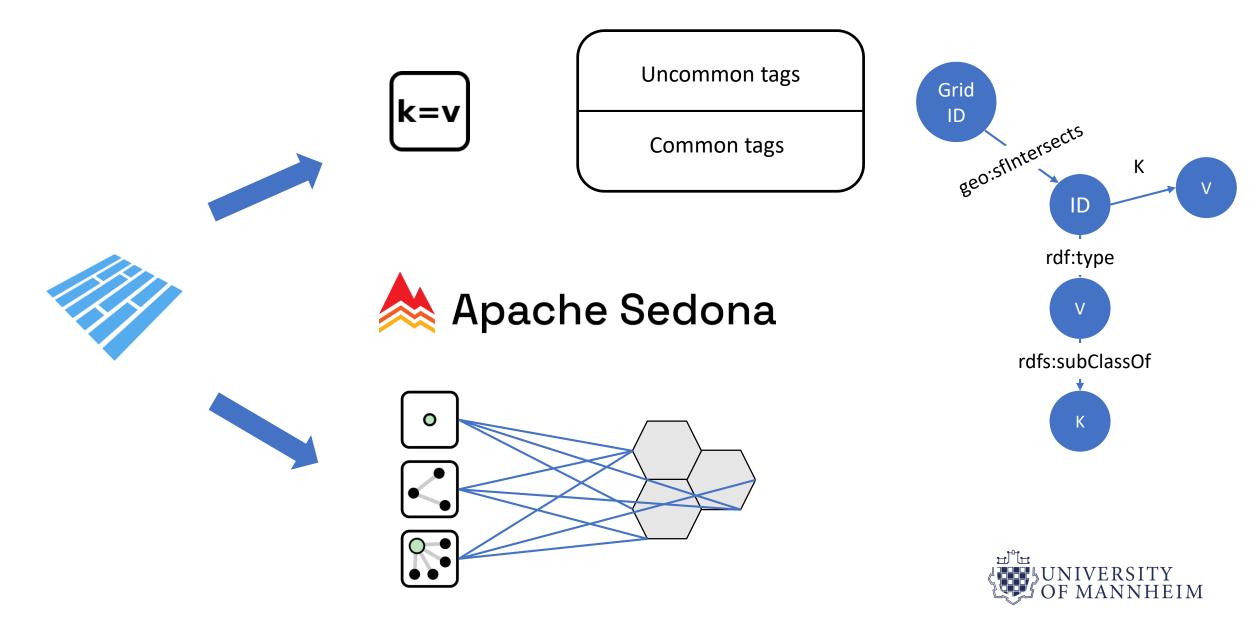


Spatio-Temporal Knowledge Graph creation





Spatio-Temporal Knowledge Graph creation



Instance of Spatio Temporal Knowledge Graph

Way: Universität Mannheim B6 (240974013)

Version #12

addr:street -> place (bzw. entfernt)

Edited <u>11 months ago</u> by Ropino Changeset #138367415

Tags

addr:city	Mannheim	
addr:place	B6	
addr:postcode	68159	
amenity	university	
name	Universität Mannheir B6	
operator	Universität Mannheir	
operator:type	university	

Nodes

14 nodes

Download XML

« Version #1 · View History · Version #12 »

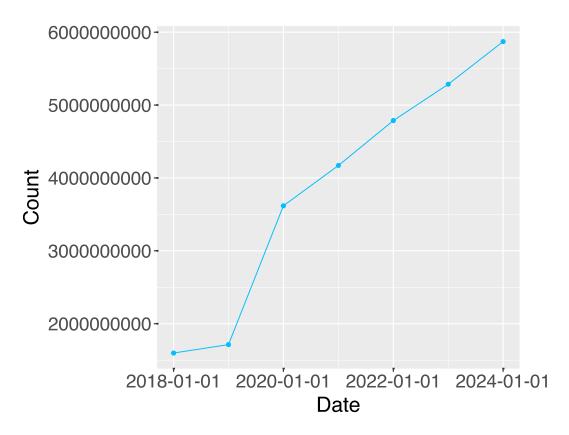


Instance of Spatio Temporal Knowledge Graph

Subject	Predicate	Object	Date
240974013	rdf:type	university	2024-01-01
university	rdfs:subClassOf	amenity	2024-01-01
240974013	addr:city	Mannheim	2024-01-01
240974013	addr:postcode	68159	2024-01-01
240974013	name	Universität Mannheim B6	2024-01-01
240974013	operator	Universität Mannheim	2024-01-01
240974013	operator:type	university	2024-01-01
240974013	geo:hasGeometry	geo240974013	2024-01-01
geo240974013	geo:asWKT	POLYGON()	2024-01-01
881fae61b9fffff	geo:sfContains	240974013	2024-01-01
881fae61b9fffff	geo:ehCovers	240974013	2024-01-01
881fae61b9fffff	geo:sfIntersects	240974013	2024-01-01
881fae61b9fffff	hcf:isAdjacentTo	881fae6183fffff	2024-01-01

Spatio Temporal Knowledge Graph characteristics

Characteristic	Count	
Triple	27,042,753,856	
Distinct entities	1,841,912,579	
Distinct predicates	98,955	



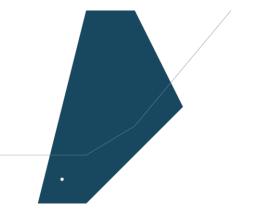


Conclusion and outlook

In total our Knowledge Graph provides over 27 billion entities over a temporal period of 2018 to 2024 (yearly snapshots).







Usage of scalable framework allows large scale processing

No direct possibility to use SPARQL on Knowledge Graph Extension to model OSM geometry relation between each other

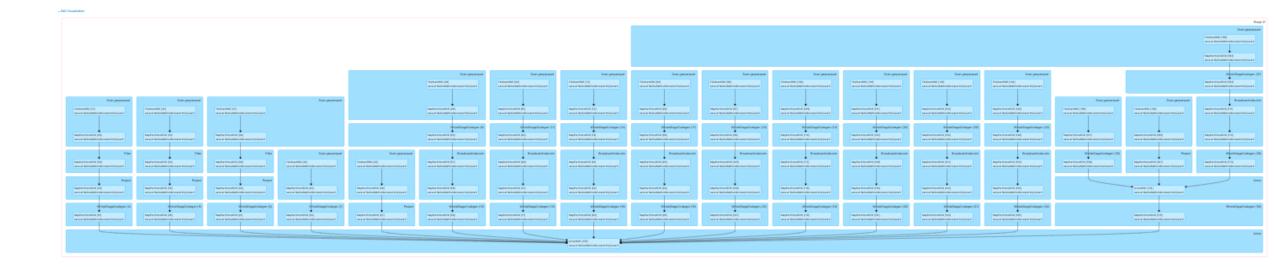






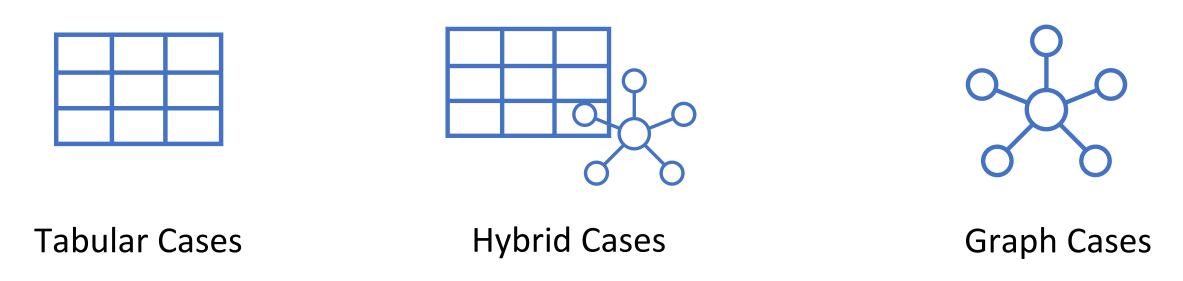
Appendix

Spark DAG



Evaluation Plan

Different spatial datasets are compared by dividing the dataset into three different dataset constellations and performing machine learning tasks



Research questions are evaluated based on ablation studies in the different dataset constellations



Preliminary results

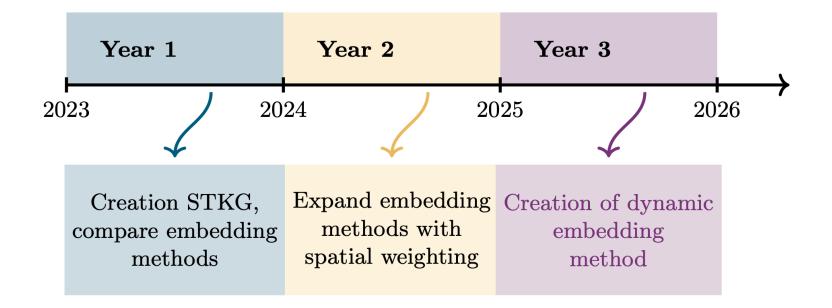
Dataset containing wildfires over time span of 2010-2021 in California containing weather, landscape and modeled on hexagonal grid cells

Dataset	F1	AUC
Tabular Case	0.3478	0.6816
Hybrid Case	0.3803	0.8748
Network Case	0.0107	0.5341

Based on results derived from Böckling, M., Paulheim, H., & Detzler, S. (2023). Wildfire Prediction Using Spatio-Temporal Knowledge Graphs



Research Methodology and Approach





Data Preparation - Weather

- Pointwise measurements of weather variables
- Need to interpolate data over created spatial grid
- Used interpolation technique:

Kriging:

$$\hat{Z}(s_0) = \sum_{i=1}^N \lambda_i * Z(s_i)$$

Weight λ_i is determined by a semivariogram

Semivariogram determines spatial autocorrelation and fits function to data

• Each constructed grid cell has now interpolated values for weather variables





- λ_i : Weight at i
- $Z(s_i)$: Value at point s_i
- $\hat{Z}(s_0)$: Prediction at point s_0

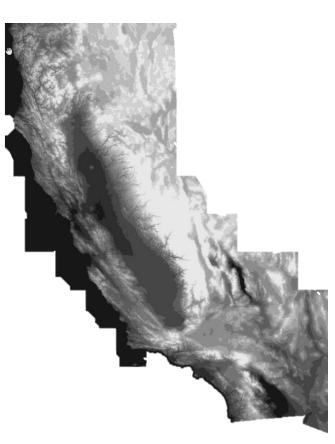
Data Preparation – Elevation & Landcover data



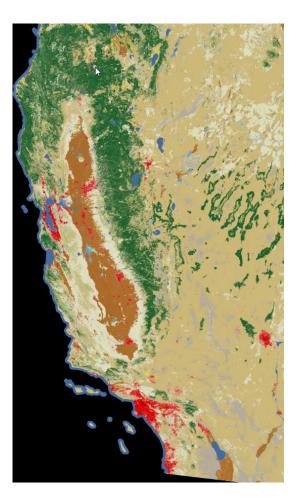




Elevation



Land cover



Data Preparation – Elevation & Landcover data



- Both datasets fine granular
 - Elevation 60m*60m tiles
 - Landcover 90m*90m tiles
- Elevation numeric dataset
- Landcover categorical dataset
- Elevation dataset gets aggregated with weighted mean to single grid cell



 Landcover dataset gets aggregated with weighted majority vote to single grid cell

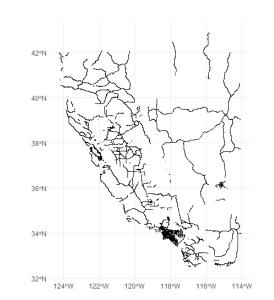


Data Preparation – Openstreetmap & Wildfire data



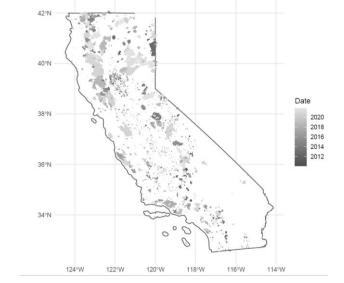






Openstreetmap

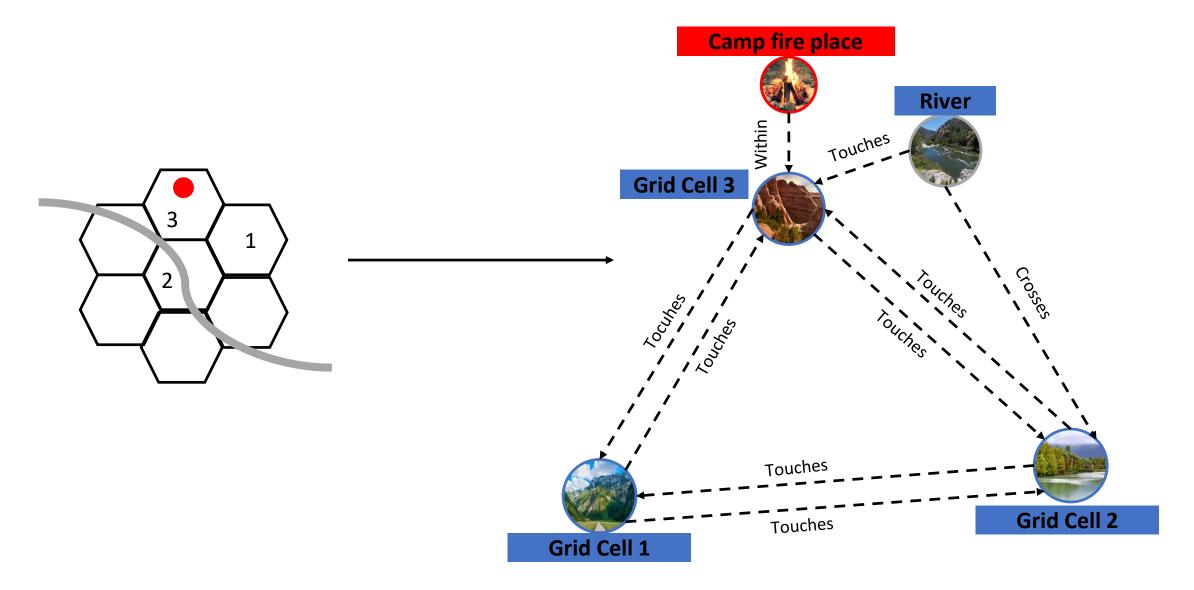
- Extract columns related to potential wildfires
- Extract necessary geometry types
- Join Openstreetmap to Grid Cell based on overlap



Wildfire

- Transform year and days to date
- Join wildfire to Grid Cell based on relation overlap

Transform data to spatial knowledge graph



Build up spatial knowledge graph – DE-9IM

- DE-9IM is topological model to build relationships between geometric objects
- DE-9IM model is based upon a 3x3 intersection matrix
- Measures the dimension dim between two geometries a and b
- Differentiates between Interior I, Boundary B and Exterior E aspects of geometries

 $DE9IM(a,b) = \begin{bmatrix} \dim(I(a) \cap I(b)) & \dim(I(a) \cap B(b)) & \dim(I(a) \cap E(b)) \\ \dim(B(a) \cap I(b)) & \dim(B(a) \cap B(b)) & \dim(B(a) \cap E(b)) \\ \dim(E(a) \cap I(b)) & \dim(E(a) \cap B(b)) & \dim(E(a) \cap E(b)) \end{bmatrix}$

- Value range for dim: {Ø (empty set), 0 (point), 1(lines), 2 (areas)}
- For Knowledge graph creation all geometry objects are related with DE-9IM to created spatial grid
- The pattern of result can be transformed into spatial predicates like the following:
 - Overlap, Touches, Within