



Building **Linked** Spatio-Temporal **Data** from Vectorized Historical Maps

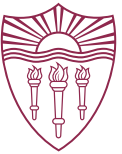
April 21st, 2020

Basel Shbita¹, Craig A. Knoblock¹, Weiwei Duan²,
Yao-Yi Chiang², Johannes H. Uhl³, and Stefan Leyk³

¹ Information Sciences Institute and Department of Computer Science, USC

² Spatial Sciences Institute and Department of Computer Science, USC

³ Department of Geography, University of Colorado Boulder

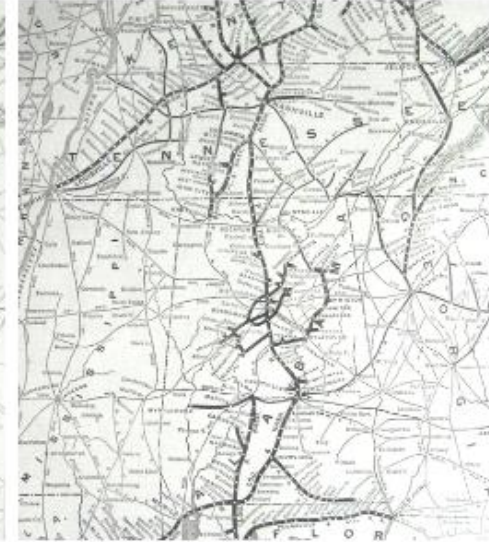
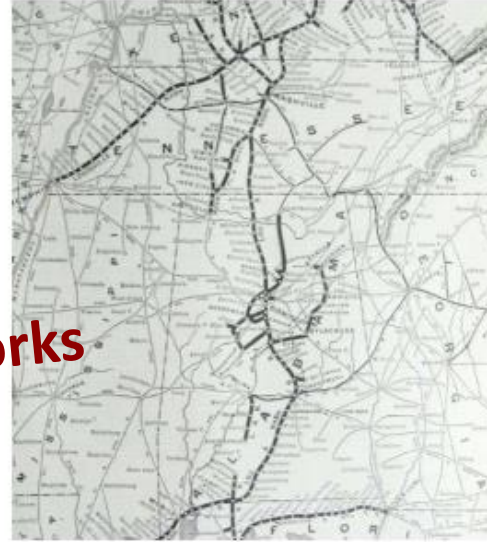


Problem

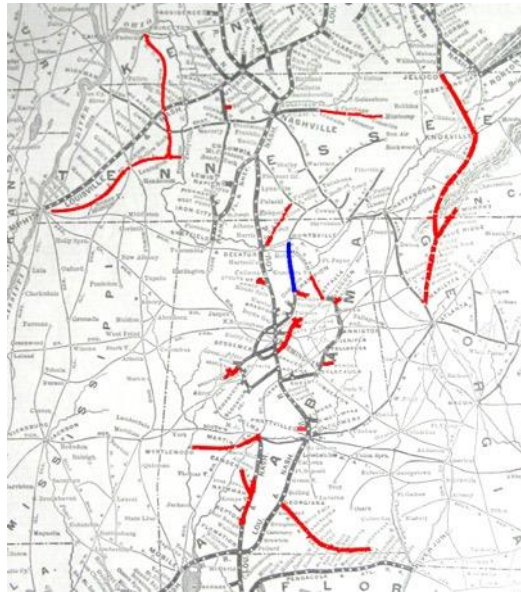
Digitized Historical Maps
=
Rich sources of *information*

1886

1904

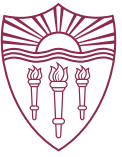


Natural- & Human-made features
Wetlands **Railroad Networks**



Labor-intensive to analyze
across time & space

Additional *discovery*



Idea

Decompose
to building blocks

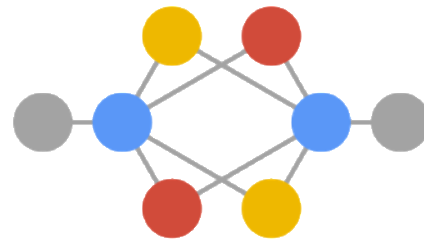


OpenStreetMap

then use

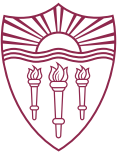


Linked Data & the Semantic Web



to build a
Knowledge Graph

Why Linked Data?



Break down
data barriers

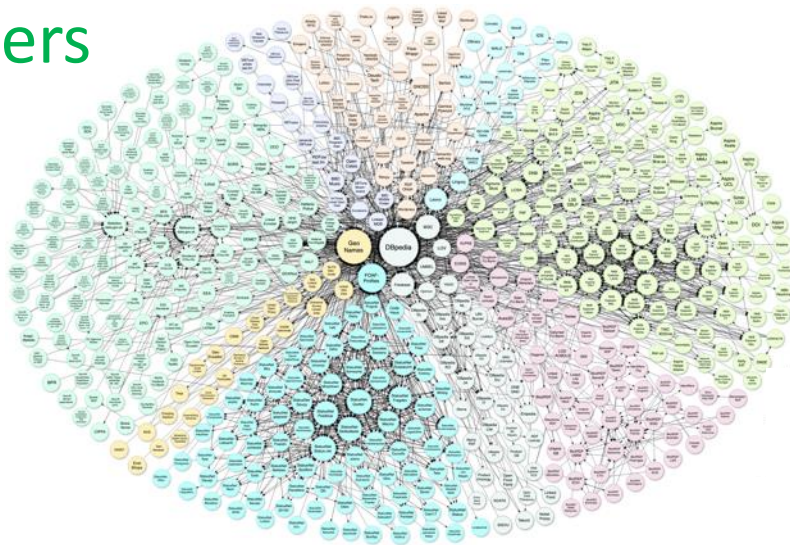
Across-domains

Structured

interpreted by **humans & machines**

Semantic

relationships, properties, metadata

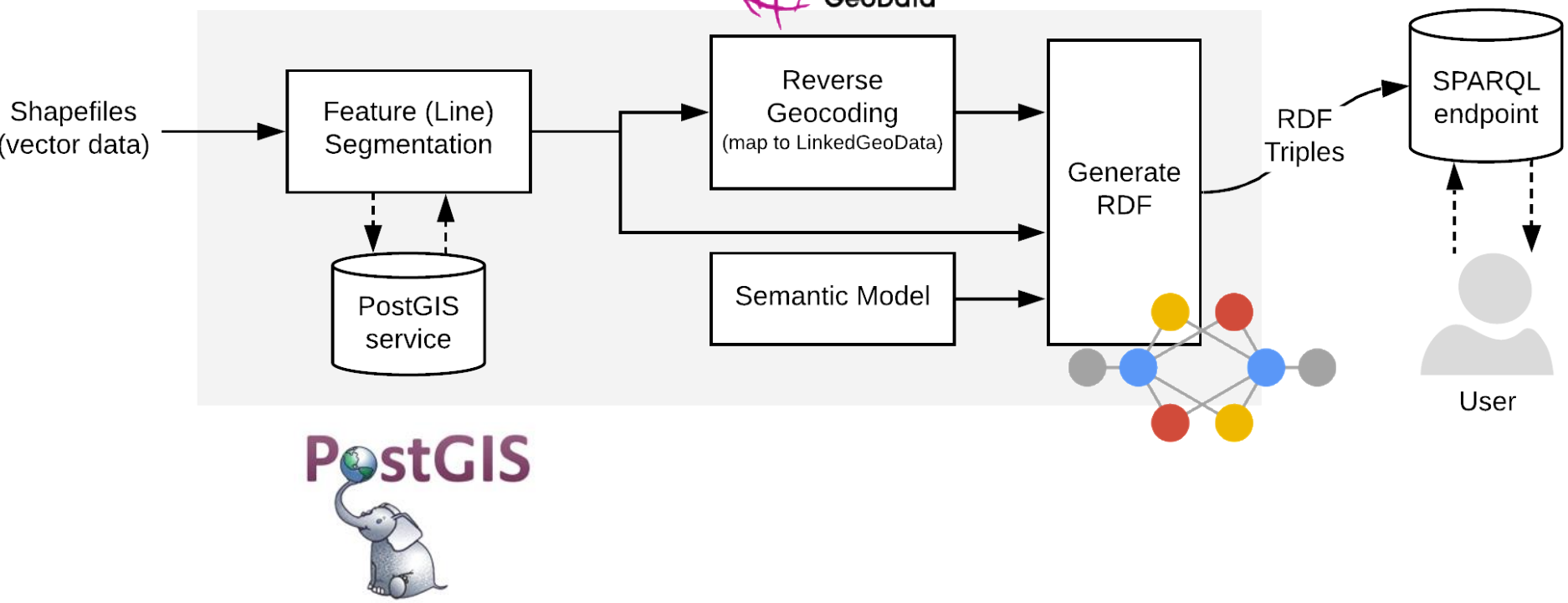
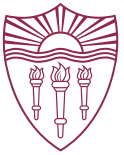


Make data
widely available

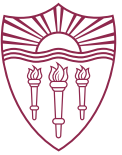
Fuel Discovery

Query & visualize

Our Approach

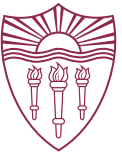


Automatic Feature Segmentation



- Goal: create **partitions** of geo features (segments)
 - Entity matching/linking & entity “partitioning” task
 - “**Building Blocks**”
- Use a spatially-enabled database service (PostGIS)
 - PostgreSQL extension
 - **Manipulate & transform** spatial data
- Allow **incremental** additions over time





Feature Segmentation – cont'd

segments of different
map edition

foreach $i \in \mathcal{M}$ **do**

current “building blocks”

foreach $k \in \mathcal{L}$ **do**

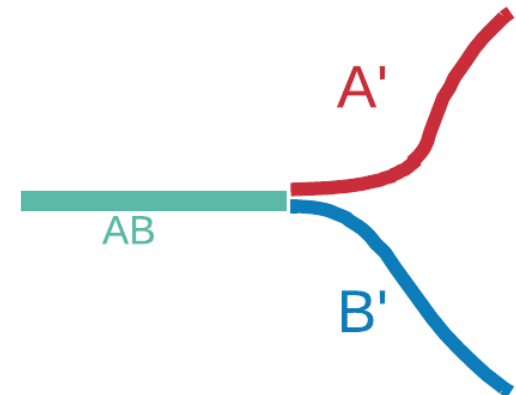
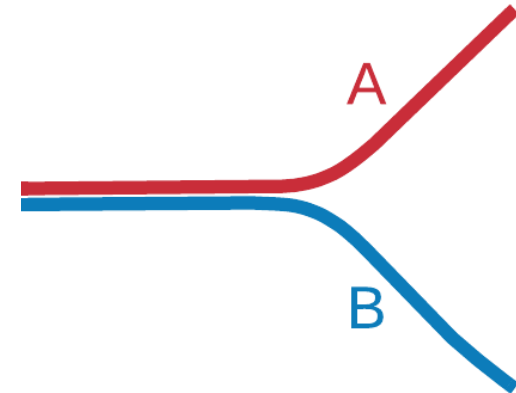
$$\mathcal{F}_\alpha = \mathcal{F}_i \cap \mathcal{F}_k;$$

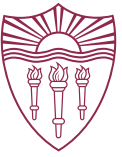
$$\mathcal{F}_\gamma = \mathcal{F}_k \setminus \mathcal{F}_\alpha;$$

end

$$\mathcal{F}_\delta = \mathcal{F}_i \setminus \left(\bigcup_{j \in \mathcal{L}} \mathcal{F}_j \right);$$

end





Feature Segmentation – cont'd

segments of different
map edition

foreach $i \in \mathcal{M}$ **do**

current "building blocks"

foreach $k \in \mathcal{L}$ **do**

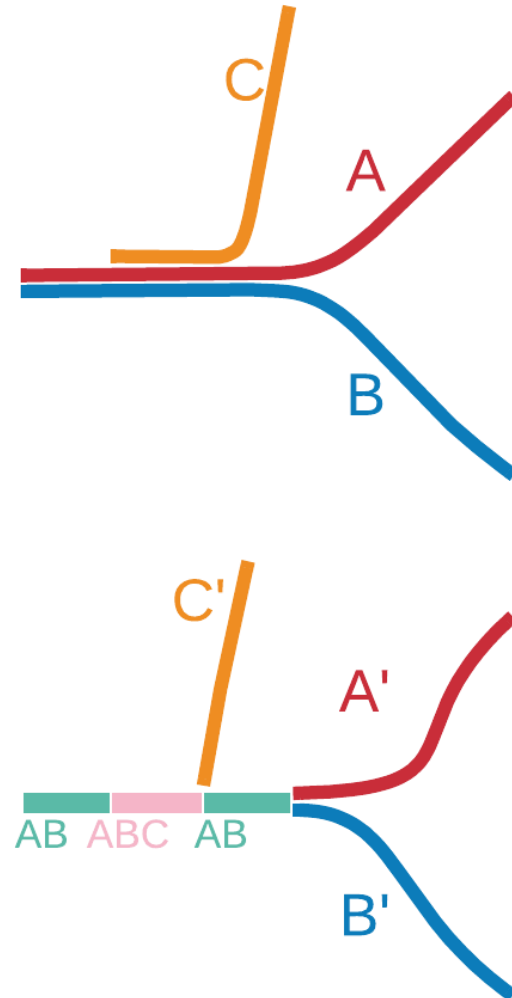
$$\mathcal{F}_\alpha = \mathcal{F}_i \cap \mathcal{F}_k;$$

$$\mathcal{F}_\gamma = \mathcal{F}_k \setminus \mathcal{F}_\alpha;$$

end

$$\mathcal{F}_\delta = \mathcal{F}_i \setminus (\bigcup_{j \in \mathcal{L}} \mathcal{F}_j);$$

end



Geo-linking

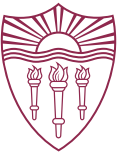


- Goal: map segments to **Linked Open Vocabularies**
 - Entity matching
 - Enrich data to fuel **discovery**
- Use a **reverse geocoding** service (OpenStreetMap)
 - **LinkedGeoData** instances



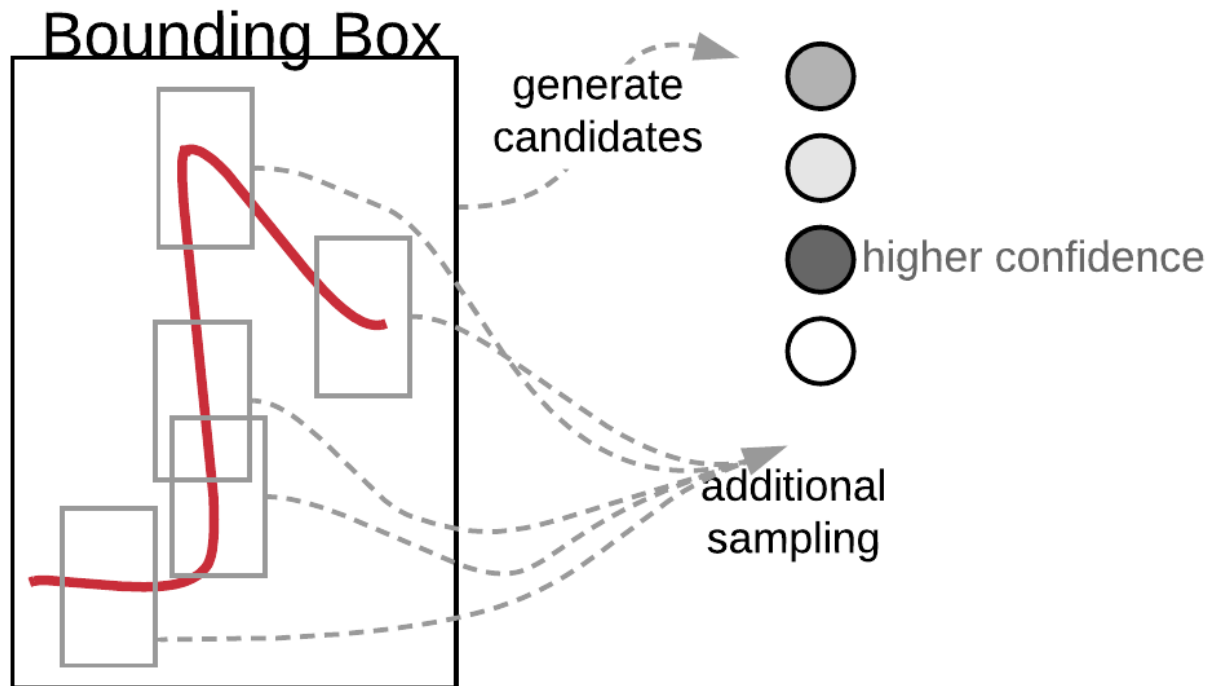
OpenStreetMap

Geo-linking – cont'd

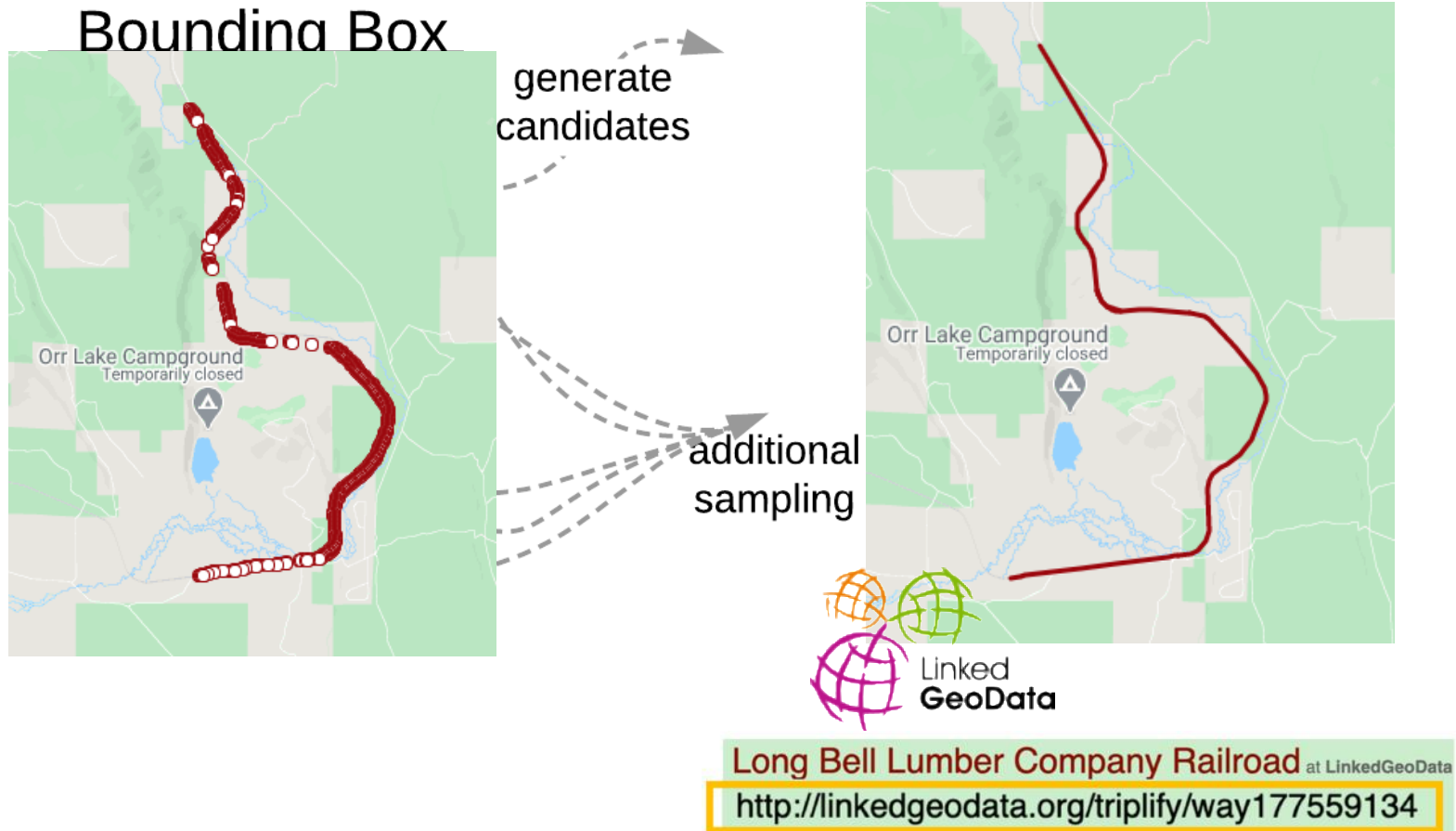
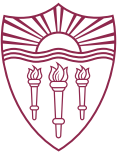


```
 $B_s$  = bounding box wrapping  $s$ ;  
 $\mathcal{L}$  = reverse-geocoding( $B_s, T$ );  
  for 1... $N$  do  
    |  $e$  = randomly sample a Point in segment  $s$ ;  
    |  $E$  = reverse-geocoding( $e, T$ );  
    |  $\mathcal{L}$ .add( $E$ );  
  end  
filter out instances with a single appearance in  $\mathcal{L}$ ;  
return  $\mathcal{L}$ ;
```

Geo-linking – cont'd



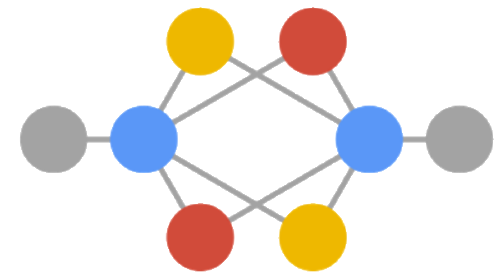
Geo-linking – cont'd



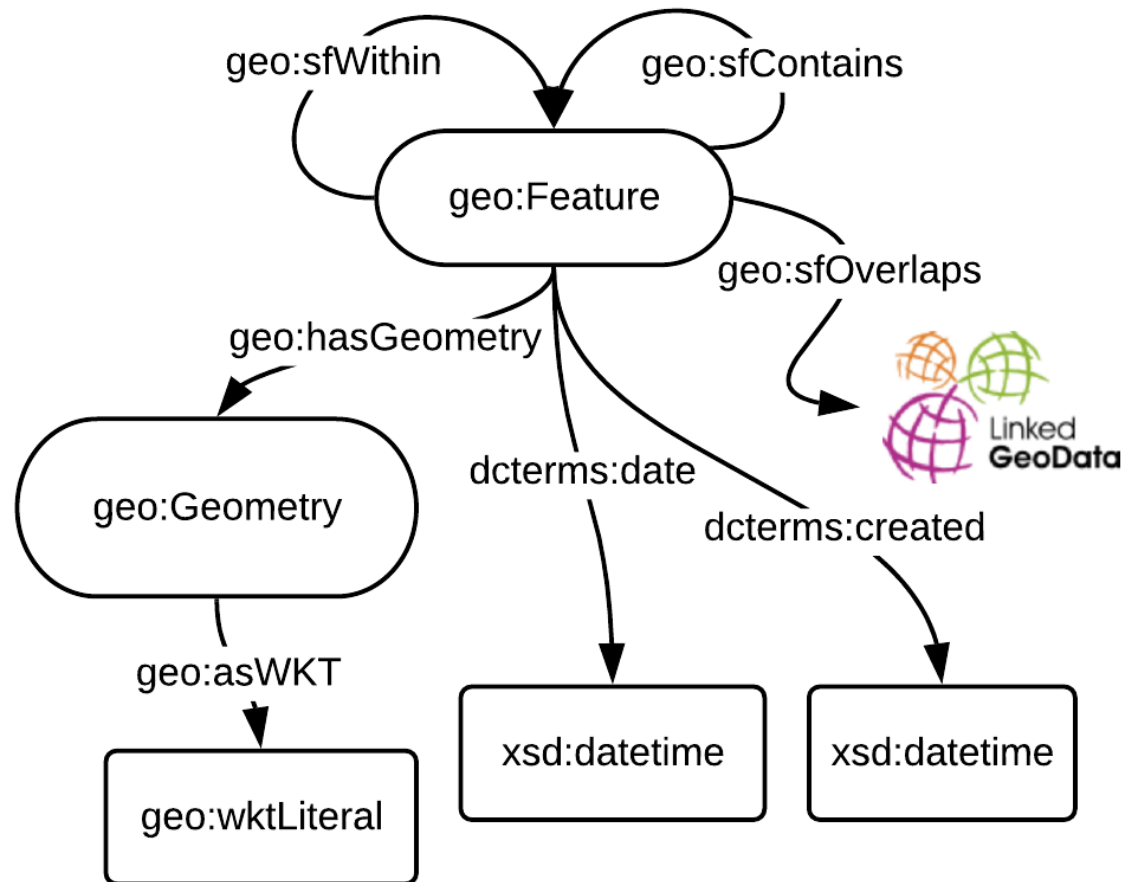
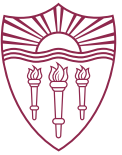
Generate RDF



- Goal: construct a **KG** from the data we collected
 - Useful **semantic representation**
 - Support downstream **spatial reasoners**
- Construct a meaningful **semantic model**
 - OGC GeoSPARQL **standard**
 - **Universal** conventions
 - Easily queried



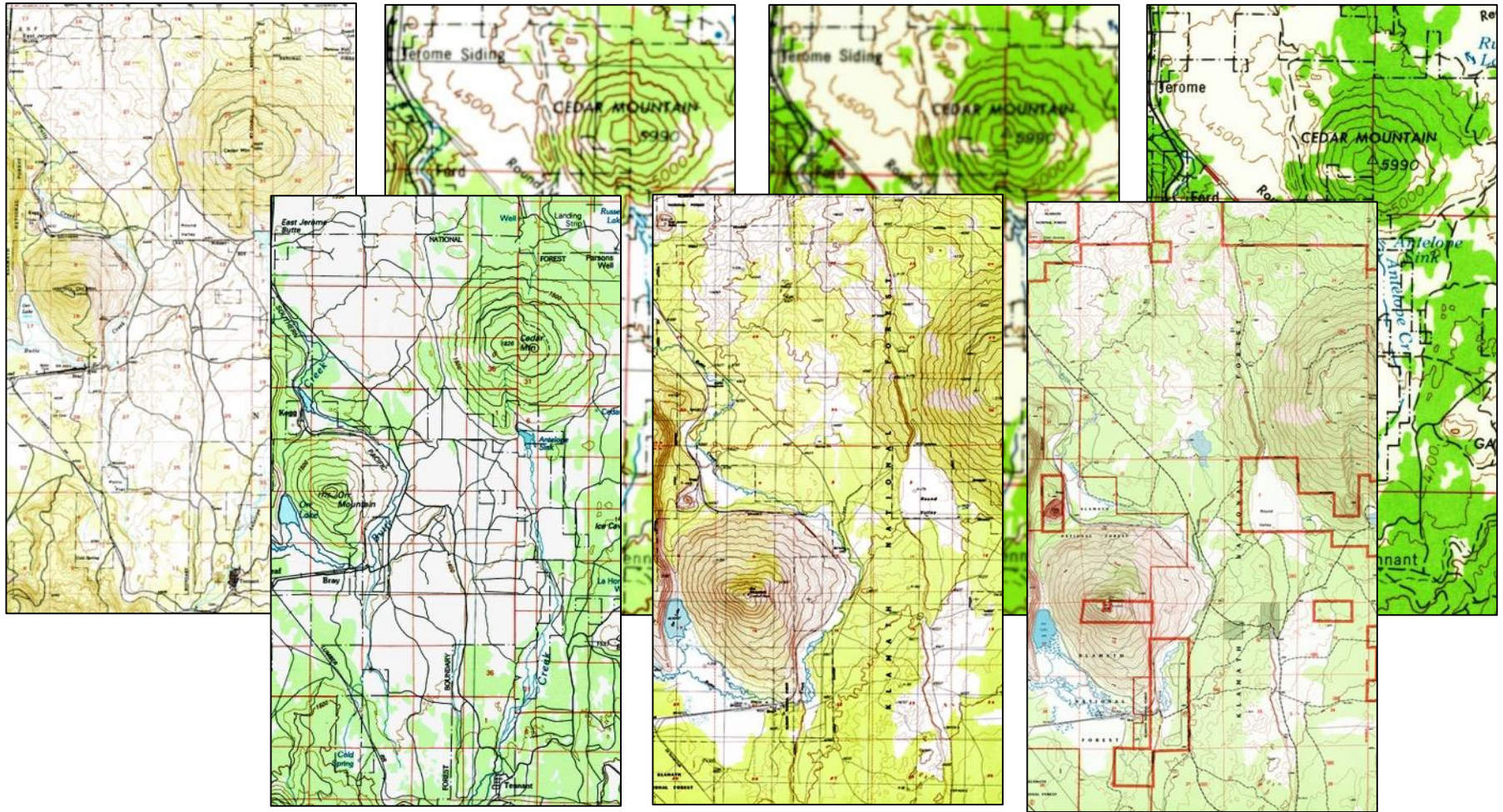
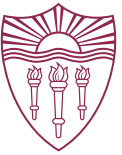
Semantic Model

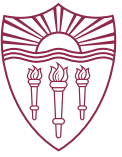




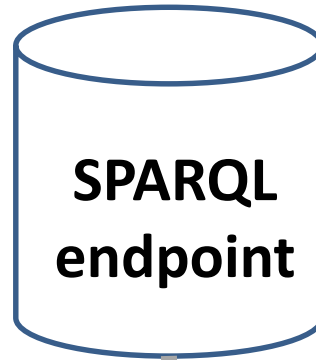
We constructed a KG
Now what?

Use Case

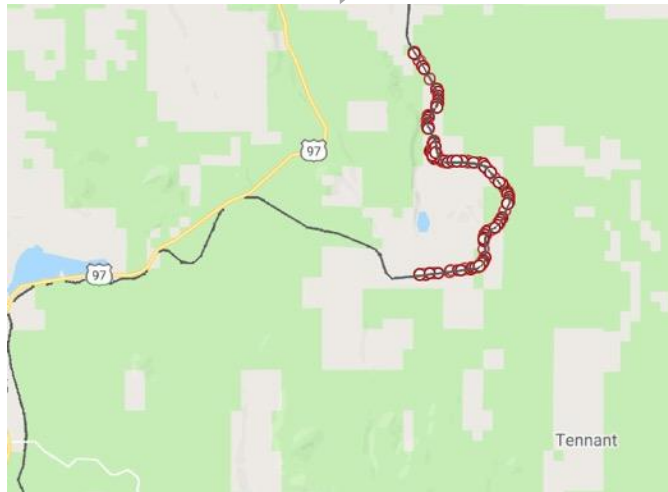




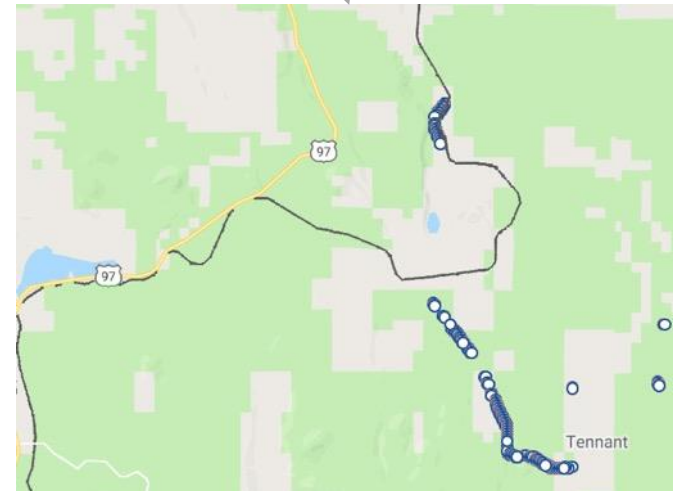
Use Case

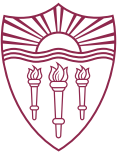


Railroad segments that are **similar** in 1962 and 2001



Railroad segments that are **present** in 1962 but are **not present** in 2001

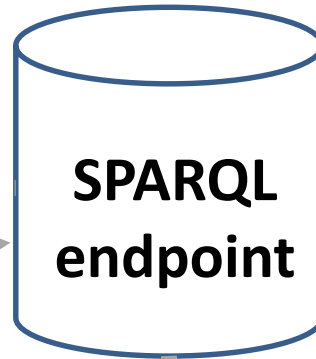




Use Case



Linked
GeoData



SPARQL
endpoint

Can you show me a
subset of what's
abandoned?

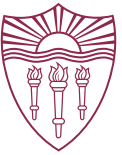
Long Bell Lumber Company Railroad at LinkedGeoData

<http://linkedgeo.org/triplify/way177559134>

Property	Value
lgdo:changeset	12836533 (xsd:int)
dcterms:contributor	lgdo:user194231
geom:geometry	lgdg:way177559134
rdfs:isDefinedBy	lgd:meta/way177559134
rdfs:label	Long Bell Lumber Company Railroad
dcterms:modified	2012-08-23T21:09:32 (xsd:dateTime)
lgdo:tiger%3Acfcc	B11
lgdo:tiger%3Acounty	Siskiyou, CA
lgdo:tiger%3Aname_base	Long Bell Lumber Company RR
lgdo:tiger%3Areviewed	no
lgdo:tiger%3Asource	tiger_import_dch_v0.6_20070809
lgdo:tiger%3Atlid	113280414:113280416:113280418:113280420:113280421
rdf:type	<ul style="list-style-type: none"> spatial:Feature lgdm:Way lgdo:AbandonedRailway lgdo:Railway,Thin
lgdo:version	1 (xsd:int)



Evaluation

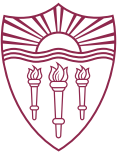


Railroad data from a collection of **historical maps**:

- Bray, California (7)
- Louisville, Colorado (4)

- **Segmentation**
 - Runtime
 - Number of nodes
- **Geo-linking**
 - Runtime
 - Correctness (Precision, Recall & F1)
- **RDF**
 - Query time
 - Query complexity
 - Query robustness

Results



- Segmentation

Table 1. Segmentation Statistics for Bray

Year	# vecs	Runtime (s)	# nodes
1954	2382	<1	1
1962	2322	36	5
1988	11134	1047	11
1984	11868	581	24
1950	11076	1332	43
2001	497	145	57
1958	1860	222	85

Table 2. Segmentation Statistics for Louisville

Year	# vecs	Runtime (s)	# nodes
1965	838	<1	1
1950	418	8	5
1942	513	5	8
1957	353	4	10

Results – cont'd

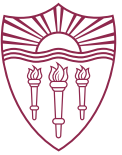


- Geo-linking

Table 3. “Geo-linking” Results

	Precision	Recall	F1
BRA-baseline	0.193	1.000	0.323
BRA	0.800	0.750	0.774
LOU-baseline	0.455	1.000	0.625
LOU	0.833	1.000	0.909

Results – cont'd



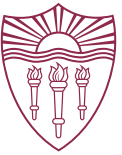
- RDF

```
SELECT ?f ?wkt WHERE {  
  ?f a geo:Feature ;  
    geo:hasGeometry [ geo:asWKT ?wkt ] ;  
    dcterms:date "1962-01-01T00:00:00"^^xsd:dateTime ;  
    dcterms:date "2001-01-01T00:00:00"^^xsd:dateTime .  
  FILTER NOT EXISTS { ?f geo:sfContains _:_ } }
```

Table 4. Query Time Statistics (in milliseconds)

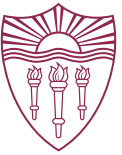
	avg	min	max
SIM-BRA	12	10	18
SIM-LOU	11	9	20
DIFF-BRA	10	8	20
DIFF-LOU	10	9	14
UNIQ-BRA	14	8	28
UNIQ-LOU	15	9	17

Discussion



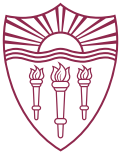
- Complexity of **changes** in **original** topographic maps
- Quality & **level of detail**
- **Crowdsourcing**
 - LinkedGeoData
- How can we do better?
 - Segmentation:
 - Optimize **buffer size** hyperparameter (heuristics/learning)
 - Normalize & **denoise** the original data
 - **Parallel** processing
 - Geo-linking:
 - Expand to **additional KBs** (Wikidata)

Related Work

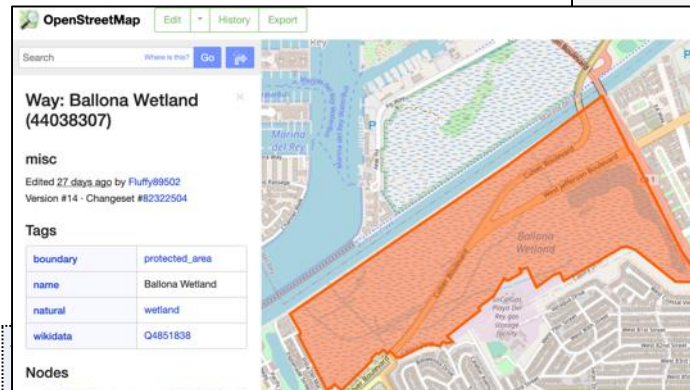


- Transforming geospatial vector data into **RDF**
 - Kyzirakos et al. [1], Usery et al. [2]
 - Do not address:
 - Geospatial entity **intra-linking** or **distant linking**
 - **Semantics**
- Geographical data **conflation**
 - Li et al. [3], Ruiz et al. [4]
 - Do not address:
 - **Linked Data** or **Semantics**
- Geospatial **data integration** in the web
 - Prudhomme et al. [5]
 - Do not address:
 - Geospatial entity **intra-linking** or **distant linking**

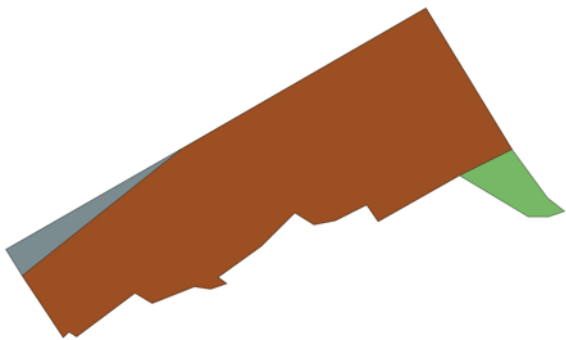
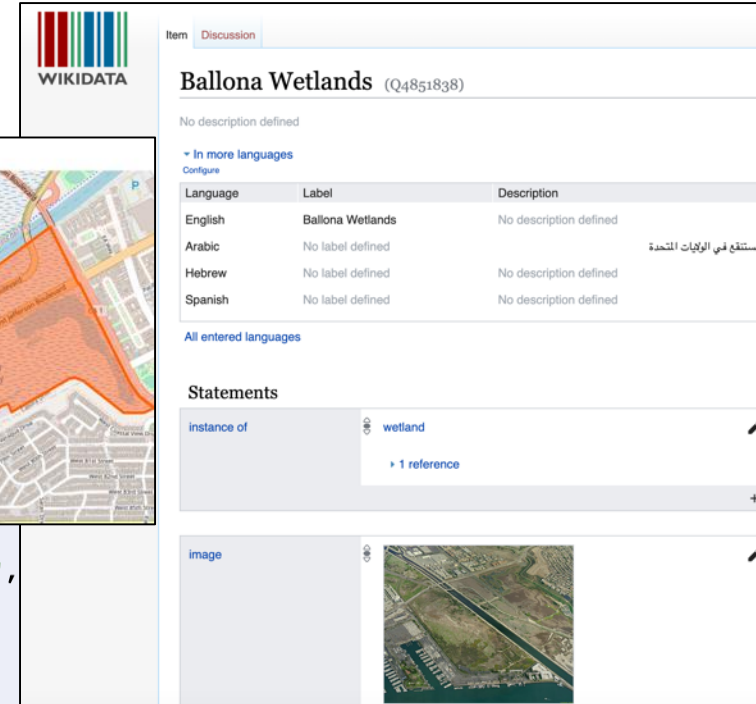
Future (present) Work



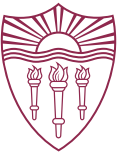
- Extend
 - Wetlands
 - Forests
 - Highways



```
id: 44038307,  
timestamp: "2020-03-18T01:26:12Z",  
version: 14,  
changeset: 82322504,  
user: "Fluffy89502",  
uid: 10165657,  
- nodes: [  
  ],  
- tags: {  
  boundary: "protected_area",  
  name: "Ballona Wetland",  
  natural: "wetland",  
  wikidata: "Q4851838"  
},  
},
```



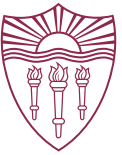
Conclusions



- Unsupervised approach to **integrate, relate, & interlink** geospatial data from digitized resources
- Publishable structured semantic-rich linked **spatio-temporal data**
- Enables users to easily **understand & analyze** geographic information across **time & space**
- Fuel **discovery**

- Source code available at:
<https://github.com/usc-isi-i2/linked-maps>

References



- [1] Kyzirakos, K., Vlachopoulos, I., Savva, D., Manegold, S., Koubarakis, M.: *Geotriples: a tool for publishing geospatial data as rdf graphs using r2rml map-pings* (2014)
- [2] Usery, E.L., Varanka, D.: *Design and development of linked data from the national map* (2012)
- [3] Li, L., Goodchild, M.F.: *An optimisation model for linear feature matching in geographical data conflation* (2011)
- [4] Ruiz, J.J., Ariza, F.J., Urena, M.A., Blázquez, E.B.: *Digital map conflation: a review of the process and a proposal for classification* (2011)
- [5] Prudhomme, C., Homburg, T., Ponciano, J.J., Boochs, F., Cruz, C., Roxin, A.M.: *Interpretation and automatic integration of geospatial data into the semantic web* (2019)