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LD Connector

References

Geographic Knowledge Graph GEOG 176C

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May 21st, 2018





 Facts, information, and skills acquired through experience or education; the theoretical or practical understanding of a subject. (Oxford Dictionary)

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Graph				
	simple graph	multigraph pseudograph		
	undirected graph	oriented graph directed graph	1 3 5 network	

 A graph is an ordered pair G = (V, E) ; where V is a set of vertices (nodes) and E is a set of edges (arcs) which are 2-element subsets of V (every edge is connected to two vertices)

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Knowledge Graph



- A KG is typically organized in the form of a graph, e.g., a directed multi-relational graph, such that the nodes represent (real-world) entities and edges represent their relations.
- A knowledge graph (KG) is a data repository that describes entities and their relationships across domains according to some schema, e.g., an ontology

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- A collection of triples/statement in the form of (*subject predicate object*)
 - (dbr:Santa_Barbara,_California , dbo:isPartOf , dbr:California)
- Example: Google Knowledge Graph, MicrosoftâĂŹs Satori, and Freebase to KGs based on W3C technologies such as DBpedia, YAGO, and Wikidata.

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Knowledge Graph



- These data repository can be linked with each other based on ontology alignment (owl:equivalentClass) and instance level alignment (owl:sameAs)
 - (dbr:Place, owl:equivalentClass, schma-org:Place)
 - (dbr:Santa_Barbara, California, owl:sameAs, freebase:Santa_Barbara, California)

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Knowledge Graph & Linked Data Cloud



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Google Assistant Google

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Google Duplex

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Santa Barbara

Apple Siri

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Applications of Knowledge Graph

• Information Retrieval, e.g. Google Knowledge Graph



18th-century mission & hundreds of animals Catholic church

Santa Barbara travel quide

Green space featuring a 1782 fortress







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References

Geographic Knowledge Graph

ADL Gazetteer Linked Data

adlg:university_of_california_at_santa_barbara_6486074	
rdf.type - adjoint-educational_facility adjoint.place	
adopathacAtemateNam - "University of California at Santa Barbara" "University of California Santa Barbara"	
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adgentheshtedteatere - 🏥 February 24, 2004	
adigont:partOf - Santa Barbara	
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• A geographic knowledge graphn is a KG which encodes geographic information.

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Geographic Knowledge Graph

- Specific problems for geographic knowledge graphs ¹:
 - How to meaningfully encode geographic information (complex geometries) in a KG? [7]
 - How to meaningfully visualize geographic knowledge graphs? [6]
 - How to enable GIS computations on geographic knowledge graphs on the fly?



• How to serialize complex geometries in geographic knowledge graphs?

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Multipart polygon representing the Lake of the Woods [7]

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GeoSPAR	QL			

@prefix rdfs: <http://www.w3.org/2000/01/rdf-schema#> .

```
@prefix geo: <http://www.opengis.net/ont/geosparql#> .
```

```
@prefix ex: <http://www.example.org/POI#> .
```

@prefix sf: <http://www.opengis.net/ont/sf#> .

ex:NationalMall a ex:Park;

rdfs:label "National_Mall";

geo:hasGeometry ex:NMPoly .

ex:NMPoly a sf:Polygon;

geo:asWKT "POLYGON((-77.050125_38.892086,_-77.039482_38.892036,_-77.039482_

↔ 38.895393,_-77.033669_38.895508,_-77.033585_38.892052,_-77.031906_

↔ 38.892086,_-77.031883_38.887474,_-77.050232_38.887142,_-77.050125_

→ 38.892086_))"^^**geo**:wktLiteral.

 Store the entire geometry and the coordinate system in a single RDF literal, thus eliminating any issues brought on by embedding complex structures as RDF

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GeoSPAR	QL			

• pros:

- storing serialized geometry data within RDF triples;
- supporting coordinate reference systems;
- maintaining the distinction between entities and their geometric representation;
- enabling geospatial queries on linked geographic data;
- o cons:
 - Challenges associated with the storage and transmission of large WKT strings;

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• Timely execution of SPARQL queries that make use of geospatial functions.

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GeoSPARQL

The Well-Known Text string for the geometry of Western Australia taken from Open Street Map

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base uri geometry ID http://ex.co/geometry/polygon/12345#113.05281,-38.11945/153.30671,-11.15957 geometry type WGS84 bounding box coordinates

Encoding scheme used to mint URIs representing a geometry [7]

- Using dereferencable Uniform Resource Identifiers (URIs) to represent the geometry as a resource;
- Including important information (geometry type; bounding box) in the URIs;

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• Can not get raw geometries information using SPARQL.

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Visualization of Geographic Knowledge Graph

DBpedia @ Browse using - | Formats -

C Faceted Browser C Sparal Endpoint

About: Santa Barbara, California

An Entity of Type : County seat, from Named Graph : http://dtbpedia.org, within Data Space : dtbpedia.org

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- Geographic Knowledge Graphs explicitly encode the semantic of geographic entities.
- Spatial information: can be easily visualized by maps.
- Semantic information: can not be easily visualized.

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Visualize Geographic KG in Multiple Ways

• Graph View



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Visualize Geographic KG in Multiple Ways

Map View



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Visualize Geographic KG in Multiple Ways

• Table View

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		hasAward	NSF 07-52366	
		thasChiefScientist	Johnson, Rodney	
		:hasCruiseType	http://voc.rvdata.us/cruise/op_science	
		:hasEndPortCall	Atlantic Explorer : 2009-03-20 : arrivePort	

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 A Semantically Enriched Visualization
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• Maps: extensively used to visualize GI and spatial relationships.



Difficult to directly express non-spatial relationships (semantic similarity) using such maps.

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References

A Semantically Enriched Visualization

A Semantically Enriched Visualization: An analogy of thematic maps to visualize the distribution of geographic features in a semantic space.

- \bullet Points: Geographic Coordinates \rightarrow Locations in the Semantic Space
- \bullet Polygons: Administration Regions/Continents \rightarrow Semantic Continents

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A Semantically Enriched Visualization

A Semantically Enriched Visualization:

- Semantically similar entities are clusters within the same region;
- The distance between geographic features represents how similar they are.

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A Semantically Enriched Visualization

- In this work, a **semantically enriched** geospatial data **visualization** and **searching** framework are presented.
- We evaluate it using a subset of places from DBpedia.



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- Multiple techniques:
 - Paragraph Vector
 - Spatial Clustering
 - Concave Hull Construction
 - Information Retrieval (IR) Model

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A Semantically Enriched Visualization



A semantically enriched visualization resembles cartographic layouts

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The Workflow



Paragraph Vectors Computing

- Paragraph Vector (or called Doc2Vec) is a representation learning method proposed by Natural Language Processing community.
- Idea: Give a collection of documents, Doc2Vec learns a high-dimensional continuous vector (embedding) for each document.
- The cosine similarity between the learned document vectors represents the semantic similarity between their corresponding documents.

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Paragraph Vectors Computing



The two-layer neural network architectural of Doc2Vec

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References

Paragraph Vectors Computing

Outputs of Doc2Vec:

- Embeddings of documents;
- Embeddings of word tokens in the document corpus.

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Paragraph Vectors Computing

- Data Source: All entities typed dbo:HistoricalPlace in DBpedia (21010 places)
 - Each historic place has an abstract, comments, images, and geographic coordinates.
- Method: Doc2Vec Model (PVDM [5])
 - **Textual data collection**: Treat each place as a document whose content is its abstract and comments
 - Textual data preprocessing: tokenization and lemmalization
 - Paragraph vector training: embedding dimension K = 300; window size N = 10; learning rate α = 0.025

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Information Retrieval Model

- Place Embeddings: the learned embedding of each historic place from Doc2Vec.
- Query Embeddings:
 - Utilize the *Doc2Vec.infer_vector()* function from gensim's Doc2Vec package
 - The TF-IDF score weighted embedding based on word embeddings of query word tokens
 - The simple average of the query tokens' embeddings after stop words removal
- Semantic Similarity Score Function: the cosine similarity between the query embedding and place embeddings
- An API ² is provided for the semantic searching functionality among *DBpedia* historic places.

http://stko-testing.geog.ucsb.edu:3050/semantic/search7searchText=grave%20yard. 🗆 🕨 4 🗐 🕨 4 🚊 🕨 4 🚊 🕨 4 🖉 - 🖉 4 🖉

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Semantic Similarity Map Construction

Spatialization: how to construct an overview of the semantic distribution of geographic entities such that it follows a cartographic tradition (*semantic continent*).



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Semantic Similarity Map Construction

K-means clustering: group these place embeddings into different clusters;

- Try #(*clusters*) from 2 to 30 and compute silhouette coefficient [8] of the clustering results;
- #(clusters) = 16 gives the highest silhouette coefficient;
- The descriptions of places in each cluster are combined as one document;
- Word clouds are produced from 10 word with highest TF-IDF score;
- Each cluster is named according to the its top 10 words.

Elementary Frame Maine Library Shaker Schoolhouse Oneroom

The word cloud for *Education* cluster

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References

Semantic Similarity Map Construction

Dimension reduction: to visualize the semantic distribution of geographic entities in a 2-dimensional space

- Different dimension deduction methods including PCA and t-SNE are experimented;
- t-SNE performs best and the clusters derived from k-means are still well separated.

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Semantic Similarity Map Construction

DBSCAN:

- Although t-SNE produces a good dimension reduction result, some points are far away from their cluster centroids and scattered in the 2D space.
- We apply DBSCAN [3] to each projected k-means cluster to extract the "core" parts of them.
- Visual interpretation are used to select the parameter combination for DBSCAN. (*Eps* = 1.1 and *MinPts* = 6)



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Semantic Similarity Map Construction

Concave Hull Construction: Chi-shape algorithm [2]

- It first constructs a Delaunay triangulation;
- It erodes the boundary by deleting boundary's edges until the longest edge less than a threshold.
- A normalized length parameter λ_p ∈ [1, 100] controls this threshold;
- To get optimal λ_p, a fitness score function [1] is used to balance the *complexity* and *emptyness* of the resulting concave hull.

$$\phi(P,D) = Emptiness(P,D) + C * Complexity(P)$$
(1)

P: the derived simple polygon; *D*: the Delaunay triangulation of the corresponding point cluster.



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Semantic Similarity Map Construction

Concave Hull Construction:

- We iterate λ_p from 1 to 100 and compute the average fitness score of all point clusters produced by DBSCAN;
- The optimal λ_p with the lowest average fitness score is 30.



The average fitness score for different $\lambda_{\rm p}$ among all DBSCAN clusters.

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Publishing the Semantic View Webmap in ArcGIS Online:



³ http://www.arcgis.com/home/item.html?id=7e15f98399ff4788a502fd04320bdafc < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > <

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Result

We have deployed a web-based user interface⁴ to showcase the functionality using the historical places dataset.



the search result of "grave yard" in the semantic space

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Result



the search result of "grave yard" in the geographic space

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References

Result



The pop-up window shows some basic information for

dbo:Istre_Cemetery_Grave_Houses.

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A Deep Integration of Geographic Linked Data with GIS

- From a GIS perspective, Linked Data seems almost like a one-way street.
- Considerations when integrate Linked Data with GIS:
 - How GIS and its users should interact with Linked Data?
 - How these key benefits of Linked Data can be maintained during conversion into GIS data formats and analysis without having to flatten the data back to a tabular format?
 - How to utilize the ontologies used to semantically lift Linked Data instead of merely relying on strings?

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ESRI Linked Data Connector

- Buffer Search on Geographic KG:
 - Find cities around the search center

Montara El Gran	Adda • • Foster Chy Mateo Belmont San Carlos Redwood City Fi	
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References

ESRI Linked Data Connector

- Property Enrichment for Geographic Entities (Utilize Semantic Web reasoning and ontologies to extract additional properties by using subsumption reasoning and (inverse) partonomical relations as examples):
 - People that died in San Francisco / Bay Area?

BJECTID	D6pediaIRI *	is_deathPlace_Of
1	http://dbpedia.org/resource/BrisbaneCalifornia	http://dbpedia.org/resource/Lafayette_Thomas
2	http://dbpedia.org/resource/San_Mateo,_California	http://dbpedia.org/resource/Robert_Cottle
3	http://dbpedia.org/resource/Pacifica_California	http://dbpedia.org/resource/Christian_Theodore_Pedersen
4	http://dbpedia.org/resource/Menio_Park_California	http://dbpedia.org/resource/Charles_N_Feiton
5	http://dbpedia.org/resource/Milbrae_California	http://dbpedia.org/resource/Albert_Johnson_(jockey)
6	http://dbpedia.org/resource/San_Mateo_California	http://dbpedia.org/resource/Greta_Johansson
7	http://dbpedia.org/resource/Redwood_CityCalifornia	http://dbpedia.org/resource/Walle_Herzer
8	http://dbpedia.org/resource/South_San_Francisco_California	http://dbpedia.org/resource/John_L_Wasserman
9	http://dbpedia.org/resource/Menio_Park_California	http://dbpedia.org/resource/Nan_Wood_Graham
10	http://dbpedia.org/resource/La_Honda_California	http://dbpedia.org/resource/Ben_Keith
11	http://dbpedia.org/resource/BurlingameCalifornia	http://dbpedia.org/resource/Michael_TGottleb
12	http://dbpedia.org/resource/Burlingame,_California	http://dbpedia.org/resource/Lurline_Matson_Roth
13	http://dbpedia.org/resource/San_Mateo_California	http://dbpedia.org/resource/Bil_Werle
14	http://dbpedia.org/resource/Redwood_CityCalifornia	http://dbpedia.org/resource/Con_Dempsey
15	http://dbpedia.org/resource/San_Mateo_California	http://dbpedia.org/resource/Morris_Kirksey
16	http://dbpedia.org/resource/Burlingame,_California	http://dbpedia.org/resource/Dick_Jones_(baseball)
17	http://dbpedia.org/resource/Daly_City,_California	http://dbpedia.org/resource/Babe_Pinelli
18	http://dbpedia.org/resource/Pacifica_California	http://dbpedia.org/resource/Mel_Stewart
19	http://dbpedia.org/resource/Menio_Park_California	http://dbpedia.org/resource/Cuckoo_Christensen
20	http://dbpedia.org/resource/Daly_City,_California	http://dbpedia.org/resource/Henry_Liu
21	http://dbpedia.org/resource/San_Mateo_California	http://dbpedia.org/resource/Wagner_Jorgensen
22	http://dbpedia.org/resource/San_Mateo_California	http://dbpedia.org/resource/Ray_Apolskis
23	http://dbpedia.org/resource/Redwood_City_California	http://dbpedia.org/resource/Bob_Garber
24	http://dbpedia.org/resource/Atherton_California	http://dbpedia.org/resource/John_R_Beckett
25	http://dbpedia.org/resource/Hillsborough_California	http://dbpedia.org/resource/Eugenio_Lopez_Jr.
26	http://dbpedia.org/resource/San Carlos, California	http://dbpedia.org/resource/Omar Ahmad (politician)
27	http://dbpedia.org/resource/Milbrae, California	http://dbpedia.org/resource/Ronnie_Montrose
28	http://dbpedia.org/resource/Menio Park, California	http://dbpedia.org/resource/Thomas A. Bailey
29	http://dbpedia.org/resource/Redwood_City_California	http://dbpedia.org/resource/Robert R. Barry
30	http://dbpedia.org/resource/Atherton_California	http://dbpedia.org/resource/Rajeev_Motwani
31	http://dboedia.org/resource/San Mateo, California	http://dbpedia.org/resource/Allan_Nevins
32	http://dboedia.org/resource/Redwood_City_California	http://dbpedia.org/resource/Jim Davenport
33	http://dboedia.org/resource/San Mateo, California	http://dbpedia.org/resource/Bert_Cole
34	http://dboedia.org/resource/Portola_Valley, California	http://dbpedia.org/resource/Albert H. Bowker
35	http://dbpedia.org/resource/Venio Park. California	http://dbpedia.org/resource/Abraham_Maslow
36	http://dboedia.org/resource/Pacifica, California	http://dbpedia.org/resource/Rapnar Hasselpren
37	http://dboedia.org/resource/Redwood_City_California	http://dbpedia.org/resource/Wally_Jay
38	http://dbpedia.org/resource/San Bruno, California	http://dbpedia.org/resource/Mosheim Feaster
39	http://dboedia.org/resource/San Carlos, California	http://dbpedia.org/resource/Tom Tennant
40	http://dbpedia.org/resource/Atherton, California	http://dbpedia.org/resource/Samuel M. Shortridge
41	http://dtpedia.org/resource/Redwood_City_California	http://dboedia.org/resource/Marioria_Deanne
42	http://dtpedia.org/resource/Redwood_City_California	http://dboedia.org/resource/Harry_Cheek
43	http://dbpedia.org/resource/San_Mateo_California	http://dboedia.org/resource/Ray_Medeiros
44	http://dtoedia.org/resource/Redwood_City_California	http://dbredia.org/resource/Boh Hoskins (American forthal
45	http://dtoedia.org/resource/Burlingame_California	http://dbredia.org/resource/lengison_Heaton
46	http://dtoedia.org/resource/Redwood_Chy_California	http://dbpedia.org/resource/Helen_Levitov_Sobell
47	http://dtoedia.org/resource/Redwood_Chy_California	http://dbpedia.org/resource/Chang-Lin_Tien
40	http://dhoadia.org/resource/San. Mateo. California	http://dbradia.org/rangurgaDavid_Breadan
40	http://dboedia.org/resource/Daty_Daty_California	http://dboadia.org/resource/Jeappe Manford
42	http://dboedia.org/carource/Dotola_Valley_California	http://dboadia.org/carourceEmile_Praining
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ESRI Linked Data Connector

- Relationship Finder between geospatial entities:
 - Which cities are 3-degree sister cities of Santa Barbara?



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