

# Towards Multi-Facet Snippets for Dataset Search

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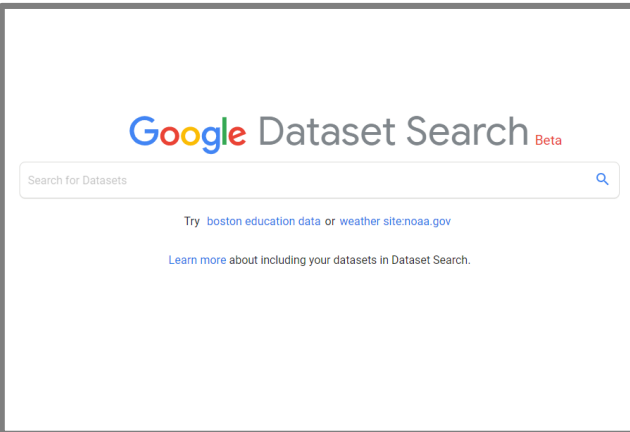
<sup>2</sup> Bosch Center for Artificial Intelligence, Renningen, Germany



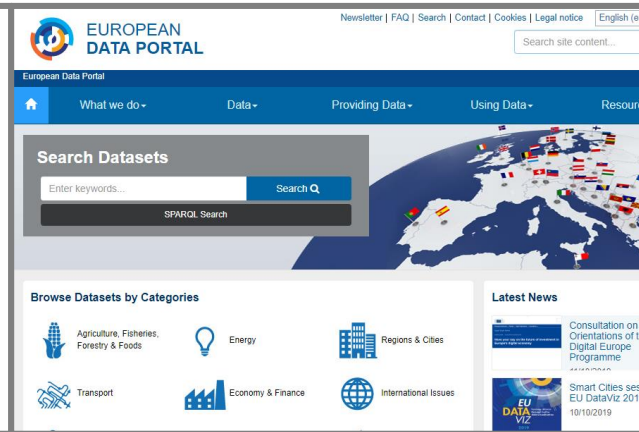
**BOSCH**  
Invented for life



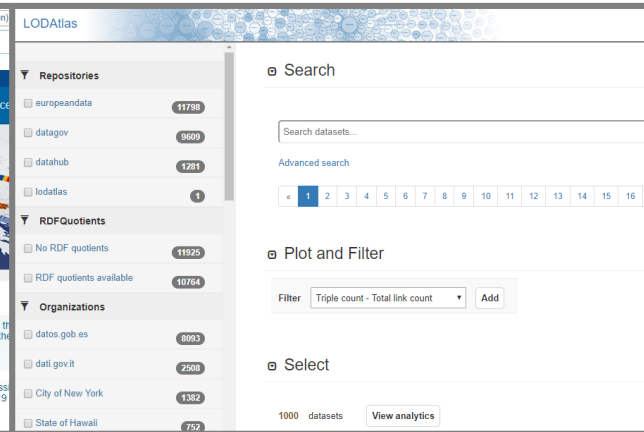
# Dataset search systems: Conveniently find relevant datasets.



*Google Dataset Search*

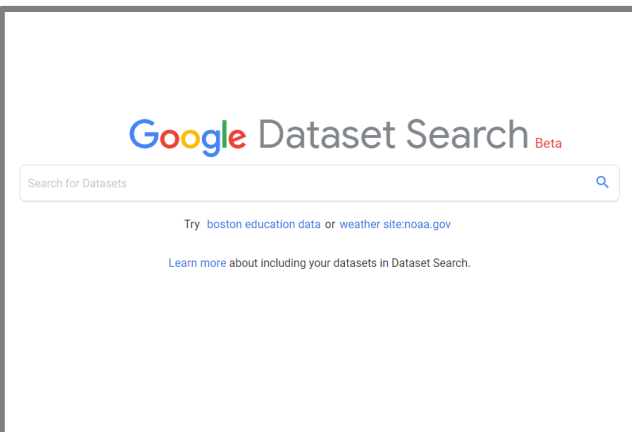


*European Data portal*

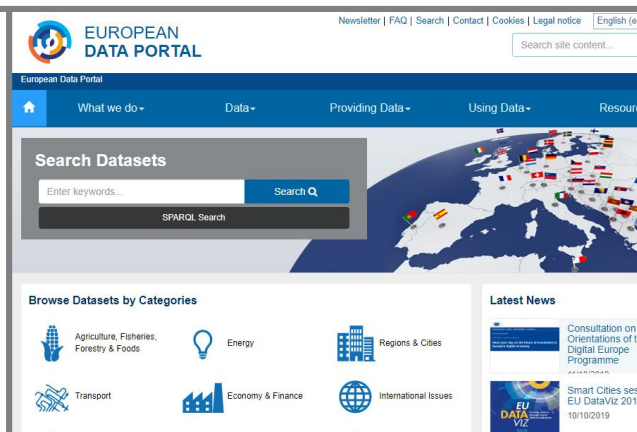


*LODATlas*

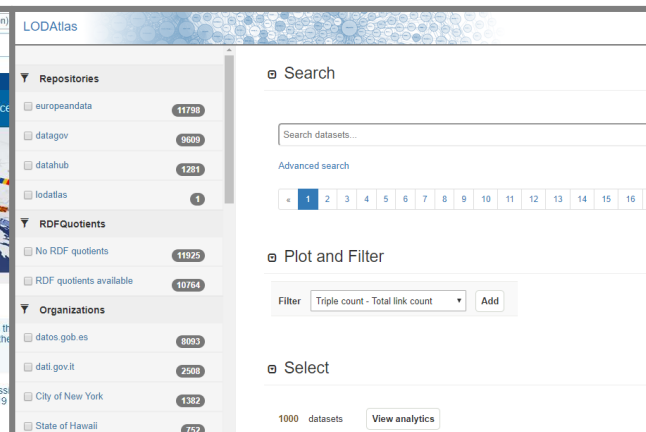
# Dataset search systems: Conveniently find relevant datasets.



Google Dataset Search

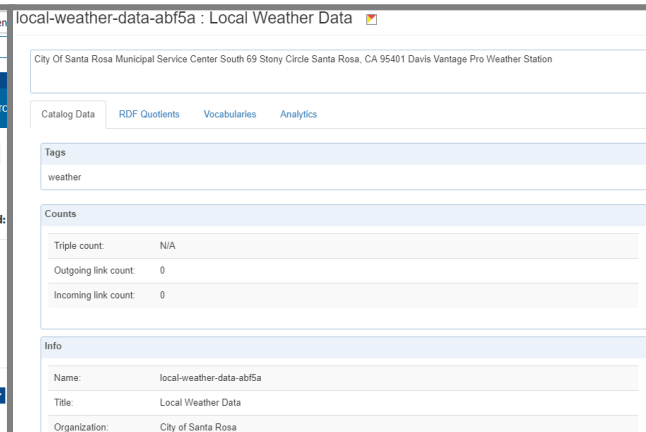
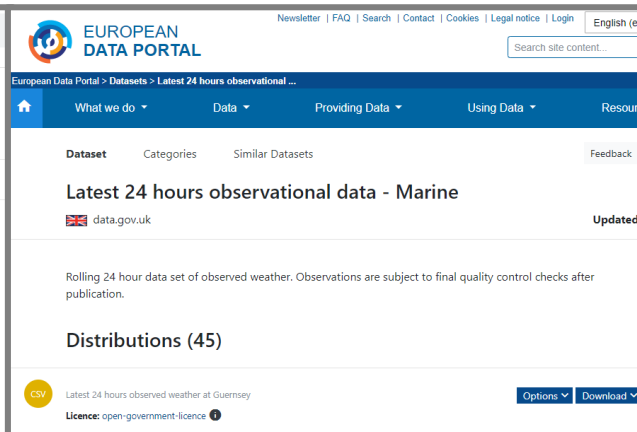
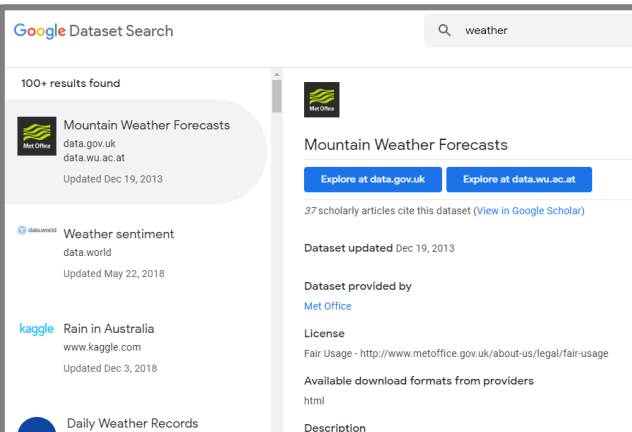


European Data portal



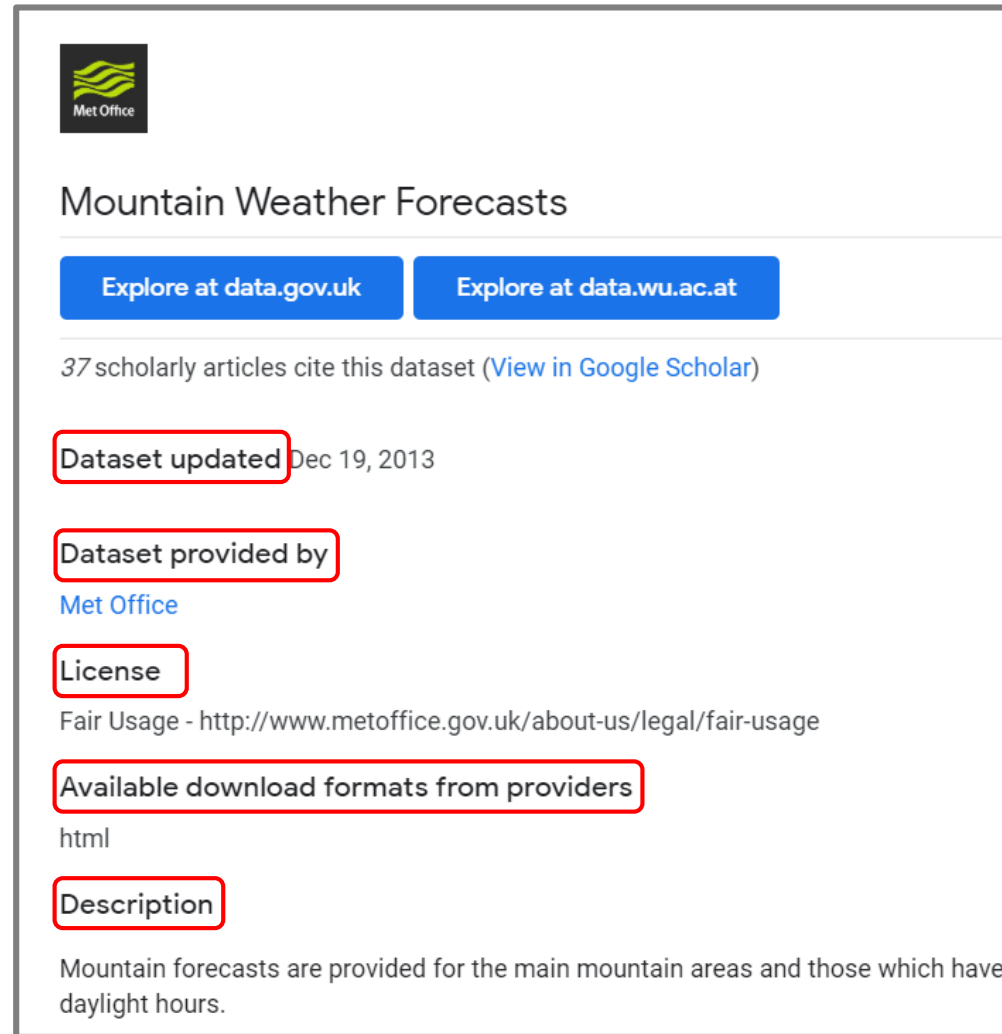
LODAtlas


Existing systems provide **only metadata** for users.



## Metadata:

- No detailed information of dataset content
- **Limited utility** for users to judge the relevance.





### Mountain Weather Forecasts

[Explore at data.gov.uk](#) [Explore at data.wu.ac.at](#)

37 scholarly articles cite this dataset ([View in Google Scholar](#))

**Dataset updated** Dec 19, 2013

**Dataset provided by**  
[Met Office](#)

**License**  
Fair Usage - <http://www.metoffice.gov.uk/about-us/legal/fair-usage>

**Available download formats from providers**  
html

**Description**  
Mountain forecasts are provided for the main mountain areas and those which have daylight hours.

*Google Dataset Search system*

## Dataset Snippet: Complementary to metadata

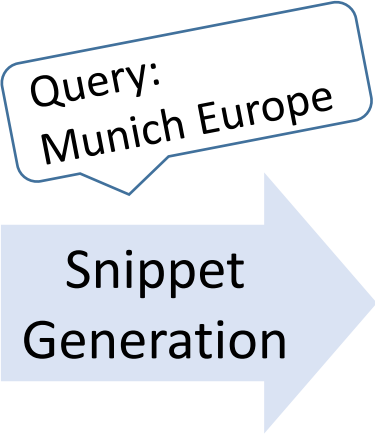
- a **size-limited subset** of triples
- **exemplify** dataset content
- **illustrate** the relevance to the query

# Dataset Snippet: Complementary to metadata

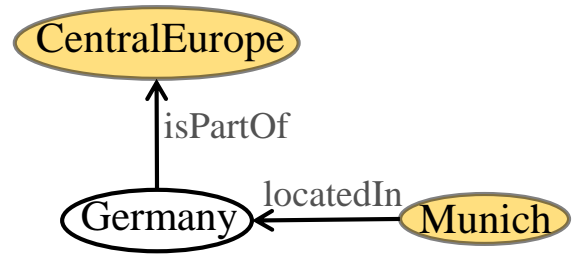
- a **size-limited subset** of triples
- **exemplify** dataset content
- **illustrate** the relevance to the query

⟨Augsburg-TYPE-City⟩  
⟨Berlin-capitalOf-Germany⟩  
⟨Berlin-locatedIn-Germany⟩  
⟨Berlin-neighboringCity-Dresden⟩  
⟨Berlin-TYPE-Capital⟩  
⟨Berlin-TYPE-City⟩  
⟨Germany-isPartOf-CentralEurope⟩  
⟨Germany-TYPE-Country⟩  
⟨Munich-locatedIn-Germany⟩  
⟨Munich-TYPE-City⟩  
⟨Munich-neighboringCity-Augsburg⟩

*An RDF dataset*



⟨Germany-isPartOf-CentralEurope⟩  
⟨Munich-locatedIn-Germany⟩



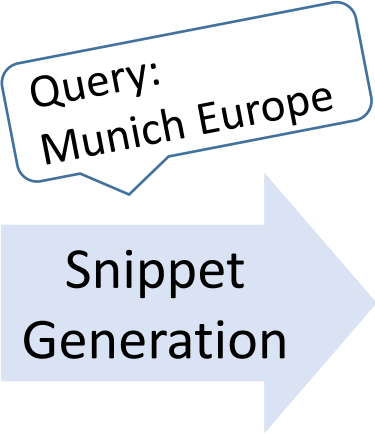
*An example snippet*

# Dataset Snippet: Complementary to metadata

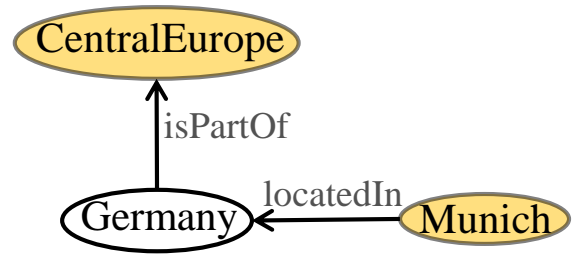
- a **size-limited subset** of triples — Size limit  $k$
- **exemplify** dataset content — Contain central elements of Schema and Instance
- **illustrate** the relevance to the query — Contain query keywords

⟨Augsburg-TYPE-City⟩  
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*An RDF dataset*



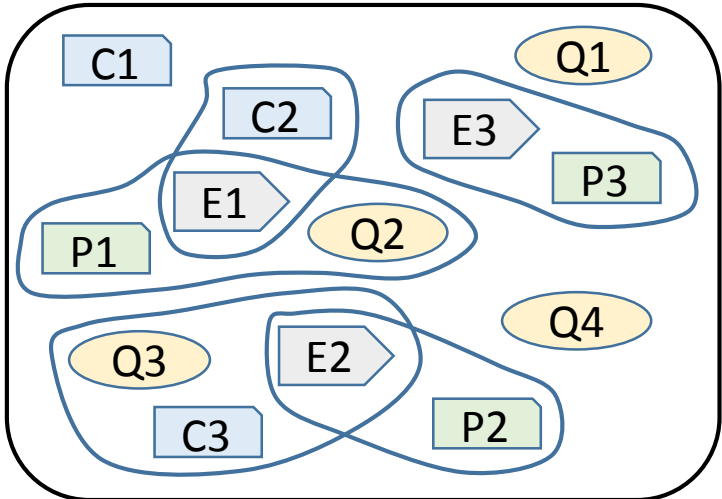
⟨Germany-isPartOf-CentralEurope⟩  
⟨Munich-locatedIn-Germany⟩



*An example snippet*

# Problem Formulation: A **Weighted Maximum Coverage** Problem

- Input: a collection of sets
- Select at most  $k$  sets to maximize the total weight of covered elements
- Consider *keywords, classes, properties, entities* as elements



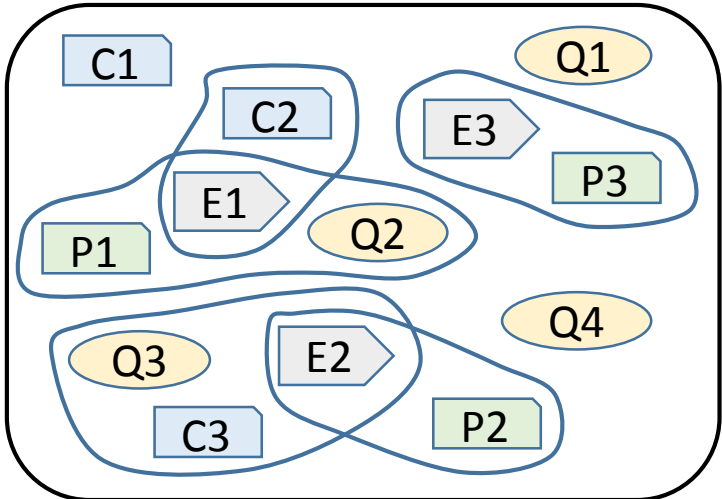
$$U = Q \cup \text{Cls}(T) \cup \text{Prp}(T) \cup \text{Ent}(T)$$

| Snippet                          |   | WMC Problem       |
|----------------------------------|---|-------------------|
| Each triple                      | ↔ | A set             |
| keyword, class, property, entity | ↔ | Weighted elements |
| Size limit                       | ↔ | At most $k$ sets  |
| Relevant to keywords and content | ↔ | Maximum weight    |



# Problem Formulation: A **Weighted Maximum Coverage** Problem

Optimize the **coverage of Keywords, Classes, Properties, and Entities.**



$$U = Q \cup \text{Cls}(T) \cup \text{Prp}(T) \cup \text{Ent}(T)$$

| Snippet                          |   | WMC Problem       |
|----------------------------------|---|-------------------|
| Each triple                      | ↔ | A set             |
| keyword                          | ↔ | Weighted elements |
| class                            |   |                   |
| property                         | ↔ |                   |
| entity                           |   |                   |
| Size limit                       | ↔ | At most $k$ sets  |
| Relevant to keywords and content | ↔ | Maximum weight    |

# Experiment

Datasets: 311 real datasets from *Datahub*

Queries:

- Real queries from *data.gov.uk*
- Artificial queries selected from *DMOZ*

Evaluation Metrics:

- Coverage of keywords
- Coverage of Paths between keywords
- Coverage of Dataset Schema
- Coverage of Data instance

Details about our **Evaluation Metrics & Baselines**  
**Tomorrow 12:00**, Session 1B, **FPAA** level 0.

# Result

- Our approach achieved a **Balance** among four evaluation metrics

|              | coKyw    | coCnx    | coSkm         | coDat         | Average       |
|--------------|----------|----------|---------------|---------------|---------------|
| IlluSnip     | 0.1000   | 0.0540   | 0.6820        | 0.3850        | 0.3053        |
| TA+C         | 0.9590   | 0.4703   | 0.0425        | 0.0915        | 0.3908        |
| PrunedDP++   | <b>1</b> | <b>1</b> | 0.0898        | 0.2133        | 0.5758        |
| CES          | 0.9006   | 0.3926   | 0.3668        | 0.2684        | 0.4821        |
| <b>coKSD</b> | 0.8352   | 0.3595   | <b>0.8651</b> | <b>0.4247</b> | <b>0.6211</b> |

- Our approach got **consistent** scores on different query groups

|             | coKyw  | coCnx  | coSkm  | coDat  | Average |
|-------------|--------|--------|--------|--------|---------|
| data.gov.uk | 0.7643 | 0.2882 | 0.8249 | 0.3870 | 0.5661  |
| DMOZ-1      | 0.8977 | 0.7955 | 0.8873 | 0.4726 | 0.7633  |
| DMOZ-2      | 0.8433 | 0.2444 | 0.8710 | 0.4569 | 0.6039  |
| DMOZ-3      | 0.8395 | 0.2337 | 0.8693 | 0.4145 | 0.5893  |
| DMOZ-4      | 0.7936 | 0.1877 | 0.8521 | 0.3731 | 0.5516  |

## Conclusion

- Our approach (coKSD) achieved a **Balance** between evaluation metrics, can be used as a better dataset snippet than existing baselines.

## Future Work

- **Better** snippet for dataset search
- **Faster** generation process

Thanks for your time!

Q&A

Contact: [xxwang@smail.nju.edu.cn](mailto:xxwang@smail.nju.edu.cn)

- Keywords

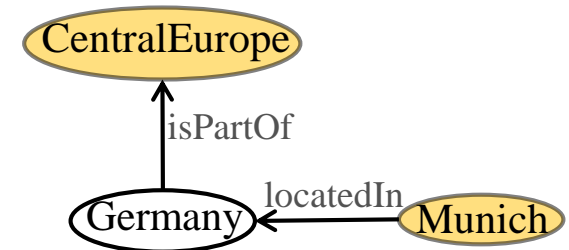
To match user's data needs as much as possible

$$\text{maximizes } q(S) = \sum_{x \in U_{t_i \in S} \text{cov}(t_i)} w(x), \quad \text{subject to } |S| \leq k$$

$$w(x) = \begin{cases} \alpha \cdot \frac{1}{|Q|}, & x \in Q \\ \beta \cdot \text{frqCls}(x), & x \in \text{Cls}(T) \\ \beta \cdot \text{frqPrp}(x), & x \in \text{Prp}(T) \\ \gamma \cdot \left( \frac{\log(d^+(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^+(e) + 1)} + \frac{\log(d^-(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^-(e) + 1)} \right), & x \in \text{Ent}(T) \end{cases}$$

Query:  
Munich Europe

<Augsburg-TYPE-City>  
 <Berlin-capitalOf-Germany>  
 <Berlin-locatedIn-Germany>  
 <Berlin-neighboringCity-Dresden>  
 <Berlin-TYPE-Capital>  
 <Berlin-TYPE-City>  
 <Germany-isPartOf-CentralEurope>  
 <Germany-TYPE-Country>  
 <Munich-locatedIn-Germany>  
 <Munich-TYPE-City>  
 <Munich-neighboringCity-Augsburg>



- **Classes and properties**

To exemplify central schema elements

$$\text{maximizes } q(S) = \sum_{x \in \bigcup_{t_i \in S} \text{cov}(t_i)} w(x), \quad \text{subject to } |S| \leq k$$

$$w(x) = \begin{cases} \alpha \cdot \frac{1}{|Q|}, & x \in Q \\ \beta \cdot \text{frqCls}(x), & x \in \text{Cls}(T) \\ \beta \cdot \text{frqPrp}(x), & x \in \text{Prp}(T) \\ \gamma \cdot \left( \frac{\log(d^+(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^+(e) + 1)} + \frac{\log(d^-(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^-(e) + 1)} \right), & x \in \text{Ent}(T) \end{cases}$$

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Central schema elements:

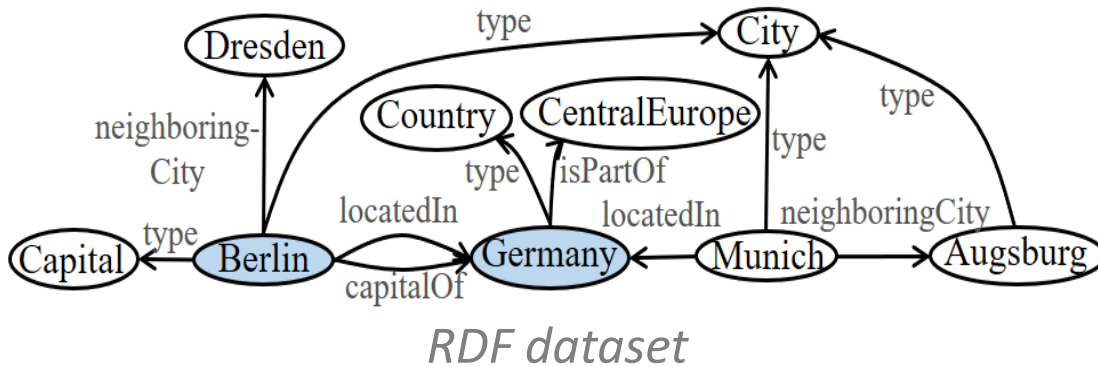
- Frequent **classes & properties**
  - e.g., **City**: 3 times
  - locatedIn**: 2 times
  - neighboringCity**: 2 times

- **Entities**

To show central elements at data level

$$\text{maximizes } q(S) = \sum_{x \in U_{t_i \in S} \text{cov}(t_i)} w(x), \quad \text{subject to } |S| \leq k$$

$$w(x) = \begin{cases} \alpha \cdot \frac{1}{|Q|}, & x \in Q \\ \beta \cdot \text{frqCls}(x), & x \in \text{Cls}(T) \\ \beta \cdot \text{frqPrp}(x), & x \in \text{Prp}(T) \\ \gamma \cdot \left( \frac{\log(d^+(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^+(e) + 1)} + \frac{\log(d^-(x) + 1)}{\sum_{e \in \text{Ent}(T)} \log(d^-(e) + 1)} \right), & x \in \text{Ent}(T) \end{cases}$$



Central entity:

- High **in-degree** and **out-degree**



# Approach

/\*Greedy Algorithm: **coKSD**\*/

Input: A dataset  $T$ , a keyword query  $Q$ , a size bound  $k$

Output: An optimum dataset snippet  $S \subseteq T$

1.  $S \leftarrow \emptyset$ ;

2. while  $|S| < k$  do

3.  $t^* \leftarrow \operatorname{argmax}_{t \in (T \setminus S)} (q(S \cup \{t\}) - q(S))$ ;

4.  $S \leftarrow S \cup \{t^*\}$ ; *//select an optimal triple in each step*

5. end while

6. return  $S$ ;

Approximation ratio:  $1 - \frac{1}{e}$