SPOOL:
A SPARQL-based ETL Framework for OLAP over Linked Data

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Background: Proliferation of Linked Data

- Linked Data (LD) has been used in various domains.
  - Publishing and connecting data on the Web.
  - Datahub[1] contains more than 10,000 datasets.
  - Linked Open Data Cloud[2] reports more than 1,000 domains.
- Many LD datasets hold useful numerical values.
  - Population
  - Food consumption
  - Money usage
  - etc.

[1] http://datahub.io/
[2]http://lod-cloud.net/

Motivation: Analytical processing

- Analyzing numerical data on LD datasets can reveal important facts.
  - LD datasets have more complicated structures (i.e. graph structure).
  - Various ontologies are used for each dataset.
- Preparation for analysis is laborious.
  - Understanding structures of target datasets.
  - Extracting necessary information for analyses.
  - Developing analytical processors.

Research purpose 1: reduce this effort

Motivation: Large LD datasets

- Previous work [3] tried to achieve the purpose.
  - ETL framework for OLAP over LD datasets.
  - Processing LD datasets in local servers.
  - The algorithm requires to read all data.
- Problems
  - Downloading large datasets takes long time.
  - Reading all data is inefficient.

Research purpose 2: efficient processing

[3] Inoue et al., An ETL Framework for Online Analytical Processing of Linked Open Data, WAIM 2013

Objective

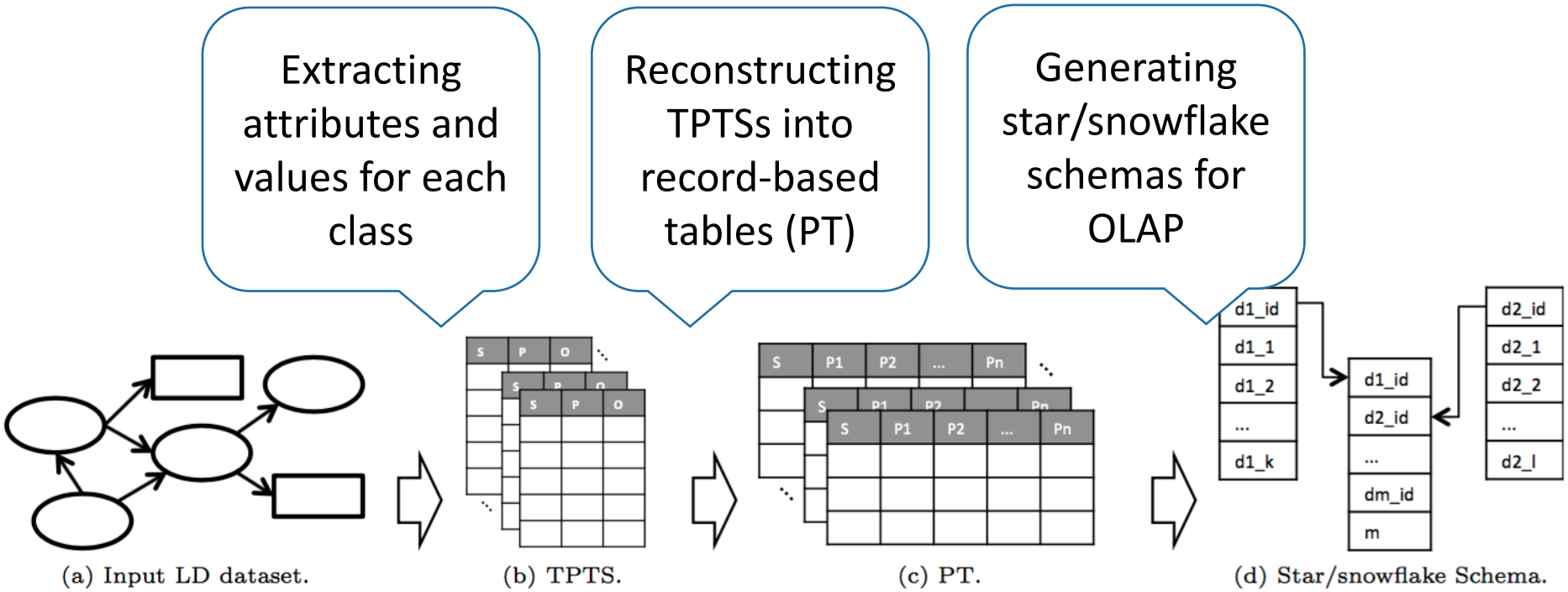
- Achieve the purposes
  - Reducing efforts for preparation of analyses.
  - Efficiently processing large LD datasets.
- Overcome the previous work [3]
  - Aiming at enabling OLAP for LD datasets
    - OLAP is typical and powerful analytical processing paradigm.
  - Processing datasets w/o downloading whole datasets.
  - Formally defining ETL process for LD datasets (cf. paper).

[3] Inoue et al., An ETL Framework for Online Analytical Processing of Linked Open Data, WAIM 2013

SPOOL: proposed framework

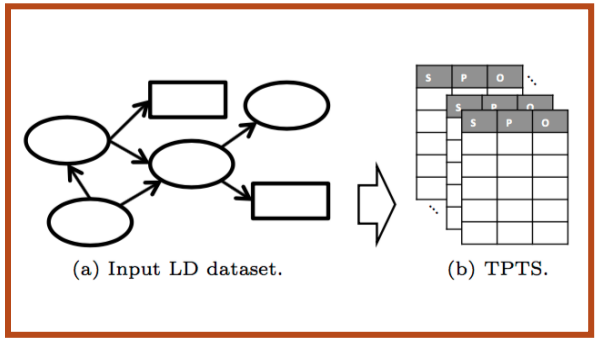
- Idea
  - Extracting only necessary information through SPARQL endpoints of LD datasets.
  - Utilizing search engine optimization on SPARQL endpoints.
- Components
  - SPARQL-based Type-partitioned Triple Store (TPTS)
    - TPTS: extracting OLAP-related information for LD datasets
      - This process is originally offline process.
    - A series of SPARQL queries to construct TPTS.

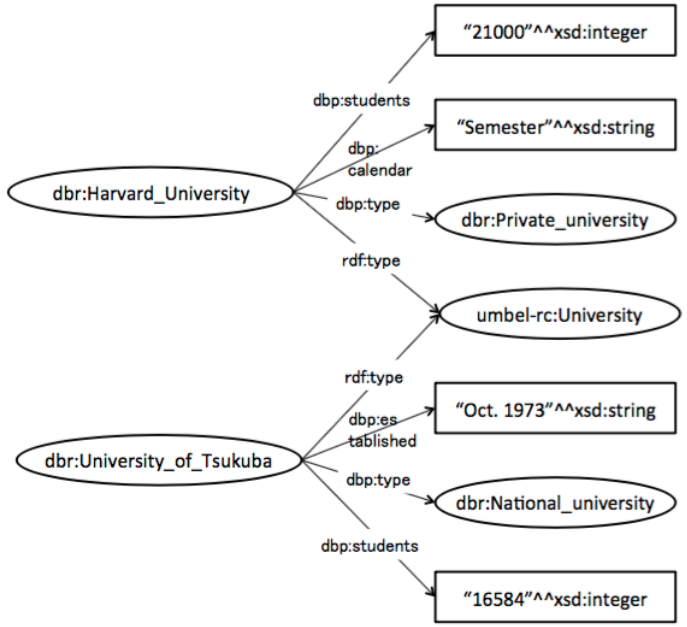
TPTS approach[3]: overview



[3] Inoue et al., An ETL Framework for Online Analytical Processing of Linked Open Data, WAIM 2013

TPTS approach: TPTS extraction

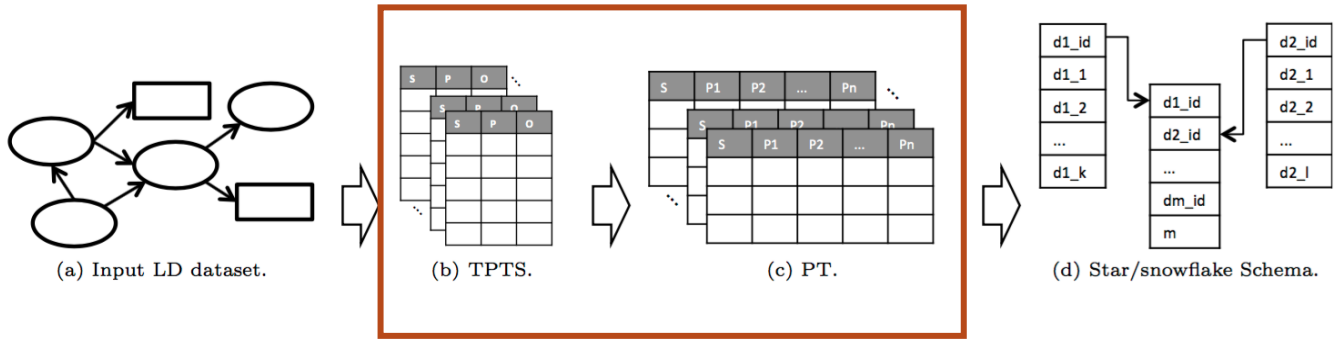




- Extract triples which subject is of a class (identifying rdf:type)
  - e.g., umbel-rc:University

subject	predicate	object	object_type
dbr:Harvard_University	dbp:students	21000	xsd:integer
dbr:Harvard_University	dbp:calendar	Semester	xsd:string
dbr:Harvard_University	dbp:type	dbr:Private_university	resource
dbr:University_of_Tsukuba	dbp:established	Oct. 1973	xsd:string
dbr:University_of_Tsukuba	dbp:type	dbr:National_university	resource
dbr:University_of_Tsukuba	dbp:students	16584	xsd:integer

TPTS approach: PT extraction



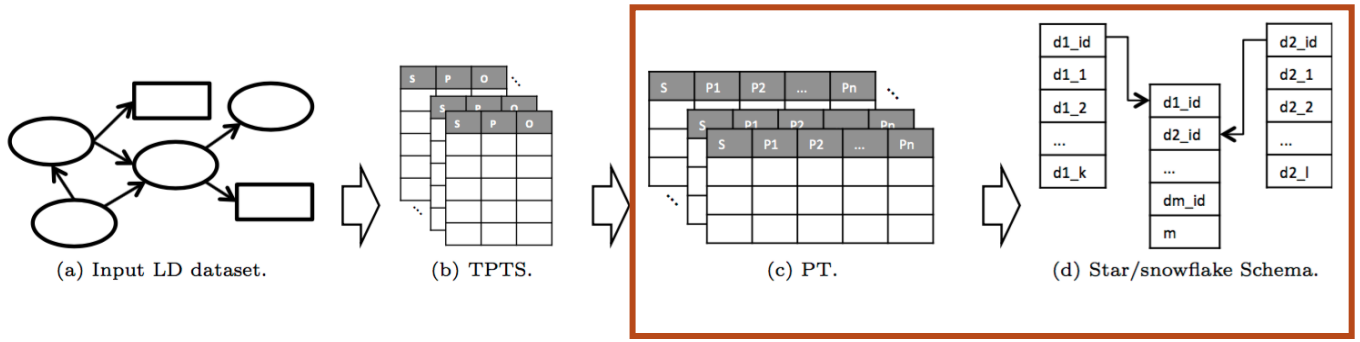
(a) Input LD dataset. (b) TPTS. (c) PT. (d) Star/snowflake Schema.

subject	predicate	object	object_type
dbr:Harvard_University	dbp:students	21000	xsd:integer
dbr:Harvard_University	dbp:calendar	Semester	xsd:string
dbr:Harvard_University	dbp:type	dbr:Private_university	resource
dbr:University_of_Tsukuba	dbp:established	Oct. 1973	xsd:string
dbr:University_of_Tsukuba	dbp:type	dbr:National_university	resource
dbr:University_of_Tsukuba	dbp:students	16584	xsd:integer

- Reconstruct tables in TPTS into record-level tables (or PTs) using distinct predicates

subject	dbp:students (xsd:integer)	dbp:calendar (xsd:string)	dbp:type (resource)	dbp:established (xsd:string)
dbr:Harvard_University	21000	Semester	dbr:Private_university	*null*
dbr:University_of_Tsukuba	16584	*null*	dbr:National_university	Oct. 1973

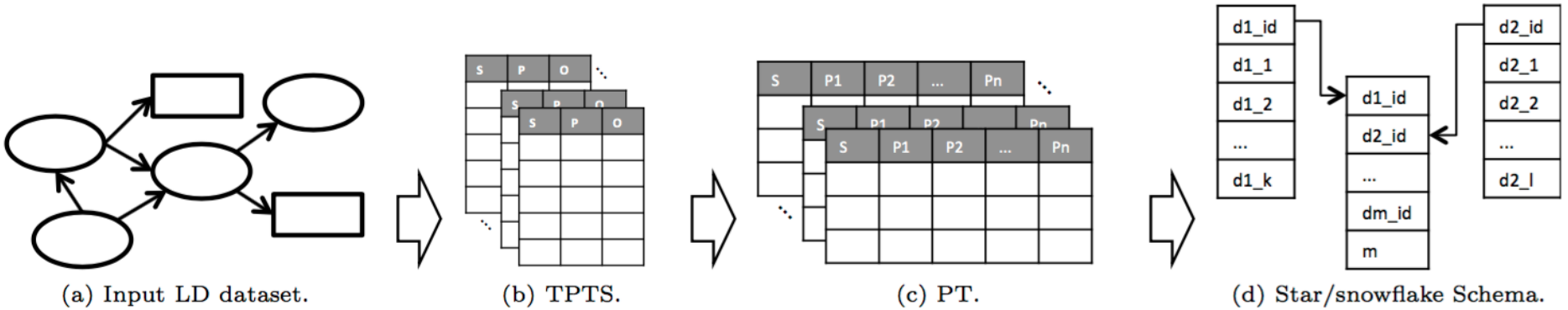
TPTS approach: Schema generation



(a) Input LD dataset. (b) TPTS. (c) PT. (d) Star/snowflake Schema.

- Generating star/snowflake schema from PTs.
  - From PT, attributes for each class are obtained.
  - An attribute is specified as measure for OLAP.
  - Other attributes in the same table as measure are considered as dimensions.

SPOOL framework: idea



(a) Input LD dataset. (b) TPTS. (c) PT. (d) Star/snowflake Schema.

- Classes and attributes are required for determining PT structures.
- Instances of classes are required during PT construction.

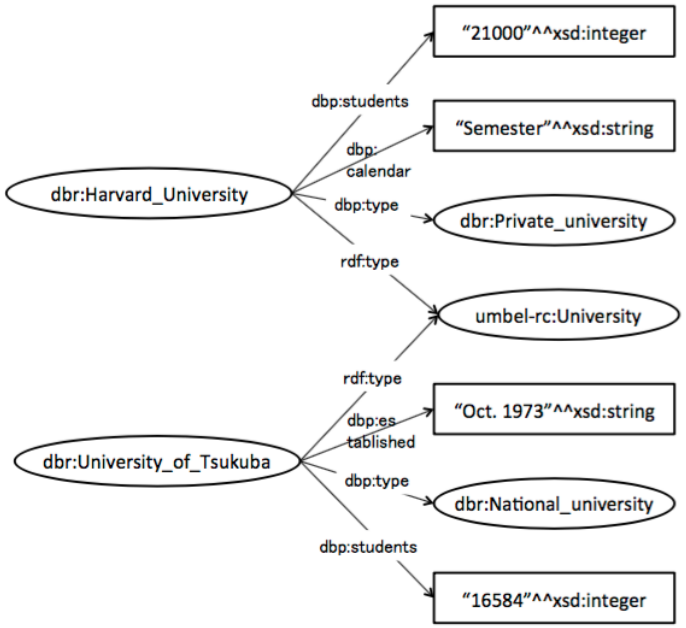
SPOOL framework: class and attributes

- Obtaining classes

```
SELECT distinct ?o
WHERE { ?s rdf:type ?o. }
```

- Obtaining attributes of a class x

```
SELECT distinct ?p datatype(?o)
WHERE { ?s rdf:type <x>; ?p ?o. }
```



subject	dbp:students (xsd:integer)	dbp:calendar (xsd:string)	dbp:type (resource)	dbp:established (xsd:string)

SPOOL framework: materialization

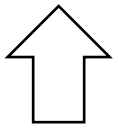
- In this step, values of attributes for instances of classes are extracted.

Algorithm 1 A SPARQL query generation algorithm for materializing a joined property Table

Input: Joined property table J_x
Output: Query q

1: $S \leftarrow \{\text{“?}s\text{”}\}$, $W \leftarrow \{\text{“?}s$ `rdf:type` $\text{”} + x\}$
2: **for** $i = 0$ **to** $|J.\mathcal{P}|$ **do**
3: $S \leftarrow S \cup \{\text{“?}v\text{”} + i\}$
4: $W \leftarrow W \cup \{\text{“?}s\text{ ”} + J.\mathcal{P}[i] + \text{“ ?}v\text{”} + i\}$
5: **end for**
6: $q \leftarrow$ “`SELECT` ” $+ implode(\text{“ ”}, S)$
$+$“ `WHERE` {” $+ implode(\text{“. ”}, W) + \text{“}\}\text{”}$

```
SELECT ?s ?v0 ?v1 ?v2 ?v3
WHERE {
        ?s rdf:type umbel-rc:University.
        ?s dbp:students ?v0.
        ?s dbp:calendar ?v1.
        ?s dbp:type ?v2.
        ?s dbp:established ?v3.
}
```



subject	dbp:students (xsd:integer)	dbp:calendar (xsd:string)	dbp:type (resource)	dbp:established (xsd:string)

Empirical study

- Purpose: Check applicability of SPOOL framework.
- Datasets
  - CIA World Factbook [4]
  - DBpedia [5]
- Methodology
  - Apply SPOOL to these datasets and observe the outputs.
- Results (cf. paper) indicate
  - Applicability of SPOOL framework for LD datasets.

[4] http://wifo5- 03.informatik.uni- mannheim.de/factbook/snorql/
[5] http://dbpedia.org/sparql

Related work: OLAP for LD

- Dedicated ontology based approaches (e.g., [Kaempgen et al. 2011])
  - Dedicated vocabularies indicate which parts of LD datasets form OLAP cubes.
  - Vocabularies are RDF Data Cube vocabulary, Open Cube vocabulary, and their extensions.
- Human-supported ETL (e.g., [Niinimaki et al. 2009])
  - This kind of approach determines mapping from an LD dataset to OLAP schema with help of users.

Conclusion

- SPOOL framework
  - OLAP schema and instances extraction from SPARQL endpoints of LD datasets.
  - Advantages
    - No need to download whole datasets in advance.
    - Utilizing search engine performance on SPARQL endpoints.
    - Small amount of human efforts is required.
- Future work
  - Enrichment of dimension hierarchy using external vocabularies.
  - Update mechanism.
  - Missing **rdf:type** situations.