

Describing a network of live datasets with the SDS vocabulary

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ABSTRACT

Multiple datasets can be derived from the same data, by transforming the objects on the incoming data stream. We want to give query processors transparency in how these datasets are related and what they contain. In this paper, we introduce the Smart Data Specification for Semantically Describing Streams (SDS) to annotate datasets with provenance information, describing the consumed stream and the applied transformations on that stream. We demo the execution of a pipeline that transforms a Linked Data Event Stream and publishes the data in a different structure as described in the SDS description. The SDS vocabulary and application profile bring together DCAT-AP, LDES and P-Plan. In future work, we will create a source selection strategy for federated query processors that considers this provenance information when selecting a dataset and interface to query the dataset.

CCS CONCEPTS

• **Computer systems organization** → **Embedded systems**; *Redundancy*; Robotics; • **Networks** → Network reliability;

KEYWORDS

pandoc, productivity, awesome

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1 INTRODUCTION

Data portals publish live datasets derived from streams. Streams are often related to other streams by transforming their data. This leads to problems when datasets are not accompanied by provenance information, the data portal has difficulty knowing whether or not this data is already published. Also, during dataset selection query agents are clueless as to what interfaces provide the same underlying

data and as a fallback, they query the same data multiple times.

For example in a GPS application that notifies the user about changes in routes due to construction sites, multiple (query) interfaces can be used. A time-based interface makes it easy to track the latest changes. On the other hand, a geospatial-based interface makes it easy to calculate whether or not a construction site will be encountered on a route. A SPARQL endpoint can be the solution if low availability and operation cost are of no concern[5]. Another possibility is to publish multiple Linked Data Event Streams (LDES)[12]. Each stream can be fragmented in a different way to accommodate the needs of the users.

Tailoring dataset interfaces to particular needs results in a plethora of interfaces which makes decent provenance a necessity. The provenance should cover the two steps a query agent takes to query an interface [4, 8]:

1. *Dataset discovery*: based on whether a dataset is going to contain statements that contribute to solving the query
2. *Interface discovery*: selecting the right interface that publishes the dataset

We set up a new interface for the Belgian street name registry and demonstrate that a query agent can find the optimal interface to execute particular queries.

2 RELATED WORK

DCTerms, DCAT and VoID: Exposing metadata about datasets is long established. Dublin Core Terms (DCTerms) can be used to provide basic information about resources, providing terms like *title*, *author*, *description* and *subject*[2]. Data Catalog Vocabulary (DCAT) is designed to facilitate interoperability between data catalogs published on the web[3]. DCAT also provides terms like *license*, which makes it possible to define a new license for an interface. The Vocabulary of Interlinked Datasets (VoID) focuses on explicitly linking datasets together on some predicate and defining subset datasets[1].

LDES: Linked Data Event Streams is a way of exposing an evergrowing set of immutable objects. These objects can be divided into fragments as HTTP resources that are linked together with the TREE specification. Fragmentations are used to provide semantic meaning to links between HTTP resources. Each HTTP resource can, for example, hold all items that start with a particular letter. A view description is used to define the meaning of the fragments and their links[12].

VoCaLS: Vocabulary of interoperable streams & On a Web of Data Streams (Dell Aglio): extends the ideas of DCAT with more information about streaming data[10]. The work defines a stream slightly differently than in this paper.

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VoCaLS focuses on streams that generate high throughput updates, this requires processors to use a windowing mechanism. In this paper, a stream is seen more broadly as a growing collection of objects, updates or otherwise.

P-Plan and PROV-O: The Ontology for Provenance and Plans (P-Plan) is an extension of the PROV-O ontology [7] created to represent the plans that guide the execution of scientific processes. P-Plan describes how plans are composed. This information can be used to keep track of provenance tree[6].

3 THE SMART DATA SPECIFICATION FOR SEMANTICALLY DESCRIBING STREAMS (SDS)

A stream in the context of the Smart Data Specification for Semantically Describing Streams (SDS) is a *physical* live channel that carries updates or items. A dataset can be derived from a stream as the collection of all updates or items. A *physical* channel can be any medium such as a Kafka stream, a WebSocket stream or even a file where updates are appended. A stream can carry any kind of data: CSV rows, mutable or immutable linked data objects, video stream bytes, etc.

A stream can be derived from a transformation applied to items of a different stream. This transformation is described with **p-plan** and resides in the SDS description of the resulting stream. The stream and the transformation correspond with **p-plan:Entity** and **p-plan:Activity** respectively. This is shown as the pink part of Figure 1. The transformation links to the previous stream with the **prov:used** predicate. With the power of the **p-plan**, query agents can understand how datasets are linked and what interface fits a specific query the best.

The SDS description also links to metadata about the resulting data with the **sds:dataset** predicate. This makes modifying the datasets' metadata possible after a transformation. This is represented as the green part in Figure 1.

Linking specific items to the correct stream is done with **sds:Record** objects. An **sds:Record** points to the data (**sds:payload**) and the corresponding stream (**sds:stream**). These small objects make it possible for multiple streams to use the same channel. Each transformation can thus push **sds:Record** objects and leave the original stream intact. A stream of immutable objects can still be transformed, for example, to calculate a hash or add a fragment id to the **sds:Record** object. The yellow part of Figure 1 gives a visual overview of **sds:Record**.

4 DEMO

Data published with Linked Data Event Streams can be partitioned or fragmented in a multitude of ways. This helps query agents resolve their queries with as few web requests as possible. A default fragmentation constitutes a timestamp fragmentation, this allows clients to replicate and synchronize the dataset efficiently. A substring fragmentation, on the other hand, makes autocompletion more efficient[11].

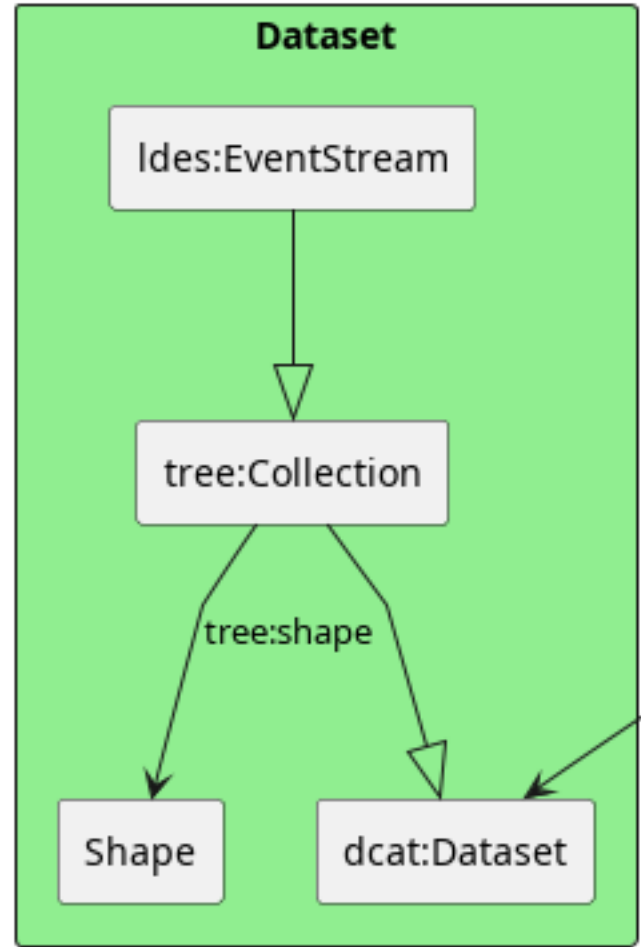


Figure 1: SDS Ontology

In this demo, we set up a pipeline starting from an existing LDES that exposes the registry of street names with a timestamp fragmentation. The pipeline calculates a substring fragmentation based on the name of the street and exposes a new LDES with the corresponding SDS Description and substring fragmentation.

When asking a query agent “What are the 10 latest updated street names?” starting from the newly created LDES, the query agent can derive from the SDS description that the current LDES is not suitable for this query. This query would require the query agent to request the entire LDES tree and manually find the 10 latest updates, whereas following the links from the SDS description back to the original LDES, this query would only require a few HTTP requests. One HTTP request gets the SDS description. The other requests get the latest updates due to the timestamp-based fragmentation.

To execute this pipeline we use a proof of concept pipeline runner called Nautirust[9]. This makes it easy to start the three required processes with the correct arguments. The three required steps are: read the original LDES with an LDES client, add buckets to the SDS Records and ingest the new SDS records in an LDES server.

5 CONCLUSION

With the introduction of the SDS ontology, it is possible to add a description to a stream and the resulting dataset, this adds provenance information. The provenance links streams together and transformations applied to the streams. The SDS ontology aligns well with long-established ontologies like DCAT and P-Plan to maximize interoperability.

With the SDS description, the query agent can now automatically select the right dataset and interface based on a given query.

Federated query processors, that utilize source selection based on this provenance information when selecting a dataset and interface to query the dataset, are still future work.

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