

Towards Hybrid Link Traversal: Challenges and Research Directions for Heterogeneous Dataspaces

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Abstract

Decentralized dataspaces preserve data sovereignty by keeping data at its source, but querying a large number of autonomous sources with heterogeneous interfaces introduces significant challenges. Traditional federated engines fail to scale to this size, while existing Link Traversal-based Query Processing (LTQP) systems can only handle Linked Data documents. By ignoring the (query) capabilities of expressive interfaces, such as SPARQL endpoints, current LTQP engines suffer performance issues. To address this, we advocate for Hybrid LTQP, an execution strategy that combines the dynamic runtime discovery of link traversal with the performance benefits of delegating complex sub-queries to capable server-side interfaces. This paper reviews the state-of-the-art in hybrid traversal and identifies the critical challenges hindering its implementation: establishing exclusive groups without prior knowledge, integrating dynamic sub-queries into active query plans, enabling reliable service discovery, and deduplicating alternative architectural views. Solving the challenges identified in this paper is a strict prerequisite for making hybrid decentralized querying practically viable across heterogeneous dataspaces. Consequently, future research must prioritize these open problems to engineer the next generation of scalable, interface-aware LTQP query engines.

Keywords

Link Traversal-based Query Processing, Hybrid Link Traversal, SPARQL, Dataspaces, Query Optimization

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1. Introduction

While centralizing data into warehouses or lakes can benefit query engine performance, it is antithetical to the core principles of *dataspaces*. Dataspaces promote a decentralized ecosystem where data remains at the source, managed by independent participants under agreed-upon governance models. This architectural shift introduces significant technical challenges for query engines, which must now discover and query data across a vast network of sources, each enforcing its own usage policies.

Centralized querying requires collecting large volumes of proprietary or sensitive data in a single source, which conflicts with the requirements for fine-grained sovereignty and minimized data replication. Conversely, traditional *federated* approaches [1, 2, 3] support decentralization but typically assume a static federation of a small number (10–100) of uniform, expressive endpoints (e.g., SPARQL endpoints) [4]. Closely related are heterogeneous federation approaches [5, 6, 3], which operate similarly but allow different source interfaces, such as Triple Pattern Fragments (TPF) [7], Bindings-Restricted TPF (brTPF) [8], Star Pattern Fragments (SPF) [9], SaGe [10], smart-KG [11], and WiseKG [12]. Enforcing complex usage policies [13] on such heavyweight interfaces significantly impacts performance [14], and these engines struggle to scale to the thousands of sources characteristic of a dataspace.

In contrast, realistic dataspace environments are characterized by a massive scale of permissioned sources and high interface diversity [15, 16]. Beyond the structured fragments managed by traditional heterogeneous federated engines, dataspace participants may, for example, expose data through generic HTTP documents or derived views [17]. Because these participants are autonomous and decentralized, their availability and specific query capabilities are often unknown until the point of execution. Addressing this requires a *hybrid* federation approach: one that can dynamically discover and interoperate

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with highly heterogeneous sources at runtime, without the overhead of prior centralization or static configuration.

Link Traversal-based Query Processing (LTQP) is an approach that meets these needs by discovering data sources during execution via hypermedia links found in earlier results. Starting from seed references (e.g., a self-description or catalog entry), it follows links asynchronously. While LTQP is traditionally defined as an approach for querying over federated Linked Data documents by following links at runtime, data within a dataspace is often exposed through diverse access interfaces that may not support standard document-based traversal and querying.

We advocate for *Hybrid* LTQP, which we define below

Definition (Hybrid LTQP): We define *Hybrid Link Traversal Query Processing* as a query execution paradigm that advances beyond traditional Linked Data Document-centric traversal. It enables query engines to traverse, integrate, and actively leverage the computational capabilities of heterogeneous, expressive data interfaces alongside standard Linked Data Documents.

Examples of such expressive interfaces are SPARQL endpoints, materialized views, and TPF [7].

To leverage these varied levels of expressivity, a hybrid engine must offload computation to the server whenever possible. For instance, a SPARQL endpoint can execute complex sub-queries locally, while document-oriented interfaces can't. By leveraging internal indexing and local data knowledge, the endpoint could produce results far more efficiently than a client-side LTQP engine. Offloading these tasks to capable endpoints could significantly reduce overall query latency. However, building such a system introduces several unresolved challenges. In this paper, we review the state-of-the-art in hybrid querying, analyze related work, and outline open issues for future research.

2. Related Work

While Hybrid LTQP has previously been explored in literature, its exact definition varies across studies. Early works [18, 19, 20, 21] typically define a “hybrid” approach as the combination of live link traversal with traditional centralized or federated querying mechanisms. In contrast, our work employs a strictly live-traversal process, redefining the “hybrid” aspect as the traversal over heterogeneous data sources.

2.1. Hybrid Traversal - Local vs Remote

A branch of hybrid link traversal querying uses precomputed indexes, either located within the decentralized environment or computed locally to speed-up query execution or source discovery. The first of such approaches, uses precomputed indexes to rerank sources based on relevance to the query’s triple pattern and joins, and retrieves them to execute queries over. [22]. Other works [23] propose to cache information on sources to help in prioritization of sources.

Another approach is to store entire RDF graphs in local stores to quickly execute parts of the query. For this approach, non-blocking indexed-based operators [24] are used to leverage the locally computed indexes for faster query execution.

A major limitation is the freshness of the data in the store, which can quickly go stale for high velocity data sources. Various methods, like freshness (or coherence)-aware query processing exist [19]. An alternative approach, and the one closest to our definition, is a hybrid method that executes a federated query over known SPARQL endpoints concurrently with traversal-based query processing over URIs [20]. While this mitigates the issue of data freshness by querying sources at runtime rather than relying on static indexes, it requires the relevant SPARQL endpoints to be configured prior to query execution. Consequently, the system cannot utilize new SPARQL endpoints discovered during the traversal process.

2.2. Hybrid Traversal - Closed Systems

Other approaches integrate into the dataspace itself. ESPRESSO [25] is such an approach, it is a framework around the Solid dataspace specifically that defines several apps in the dataspace to perform certain *optimization* tasks. In the Solid environment, data is stored into personal data vaults. These vaults use the Linked Data Platform [26] to arrange the data into a folder-like structure. To avoid traversing the entire pod to query it, ESPRESSO uses the *Brewmaster* application to locally create indexes of the pod and uses the *CoffeeFilter* application to search its contents. In addition, ESPRESSO uses an overlay network to distribute end-user queries to relevant data resources across Solid servers, where each Solid server is mapped to a federated database node in the overlay network.

Unlike approaches based on link traversal or service discovery within Solid, ESPRESSO relies on an explicitly configured overlay network, with no mechanism for dynamically discovering participating servers or query endpoints. Furthermore, currently ESPRESSO only supports distributed keyword-based search, it does not support SPARQL queries over the sources.

While these approaches do hybrid link traversal, they significantly deviate from our interpretation of hybrid traversal. This in contrast to the setting described in this paper where the discovered sources themselves are heterogeneous, serendipitously discovered, and cannot be assumed to be present for every source within the dataspace.

2.3. Hybrid Link Traversal over Heterogeneous Sources

State-of-the-art LTQP engines, most notably Comunica-link-traversal [27, 28], demonstrate hybrid capabilities by dynamically federating over discovered SPARQL endpoints, TPF interfaces, and hypermedia documents. However, these systems are currently limited by a triple pattern level-only strategy: they fail to leverage the full expressivity of sophisticated interfaces. For instance, upon discovering a SPARQL endpoint, existing engines restrict interaction to simple *triple pattern* lookups and do not attempt sending sub-queries.

This often results in the client retrieving high volumes of intermediate results for performing local joins, and effectively negates the optimization and latency benefits of more expressive server-side execution.

3. Challenges with Hybrid Link Traversal

In this section, we will outline specific identified problems with hybrid link traversal making implementation difficult.

3.1. Establishing Exclusive Groups

In classical federated query processing, *exclusive groups* are sets of triple patterns within a query that can *only* be answered by a single source [1]. instead of retrieving results for individual triple patterns and joining them locally, the engine can delegate the entire group as a single sub-query (e.g., a Basic Graph Pattern) to the remote endpoint. This technique significantly reduces the number of remote requests and processed intermediate results [29, 1].

The identification of exclusive groups is a strict requirement for this sub-query delegation. If a query engine cannot verify that a group of patterns belongs *exclusively* to a single source, it cannot safely dispatch them as a conjunctive sub-query (BGP). Doing so risks *incomplete results* because the endpoint will execute a server-side join. If the endpoint lacks data for one of the patterns, or if valid join partners reside on a different server, the server-side join execution will filter out bindings that could have been successfully joined with data from another source. Thus, the exclusive grouping strategy fails to produce the valid distributed join results.RT: This last sentence seems incorrect.

Federation engines like *FedX* [1], *HiBISCuS* [30], Comunica [27], and *FedUp* [29] rely on either prior knowledge on the content of the federation members, or issue queries (such as ASK) to determine these groups.

For Hybrid Link Traversal to use SPARQL endpoints to compute joins locally, the engine should also identify exclusive groups. The problem is the unbounded nature of LTQP [31]. As during LTQP the query engine does not have access to all sources it will query over, exclusive groups can never be established.**RT:** Well, maybe they could be, but at least they can not be established in the same way as in federation in traditional (non-adaptive) query planning. Even if we move to a finite web [31], the engine would have to first traverse all reachable [31] sources before exclusive groups can be determined. This is antithetical to the streaming nature of LTQP engines [27, 32] and will delay time to first result. Thus, traditional approaches for determining exclusive groups are not applicable.

3.1.1. Research Avenues

RT: I'd repeat the main problem here first: what information could we use to safely determine exclusive groups?

Discoverable Indexes: Indexes can be effectively used to catalogue the data sources present in a dataspace. These indexes could be deployed as services, similar to approaches like ESPRESSO [25], pointing towards available source URIs and potentially exposing summary statistics. Using such indexes to quickly determine relevant sources, LTQP engines can apply traditional federation techniques to establish exclusive groups and perform source selection.**RT:** But it's not really link traversal anymore then, right? (which is fine) But maybe make it explicit, that the problem can then become a pure federation problem.

Caching: Similar to discoverable indexes, caches provide prior knowledge about the data sources present in a dataspace. Provided the cache maintains a degree of freshness and completeness, the query engine can perform a ‘simulated traversal’ offline over the cached data to identify relevant sources. Any gaps in this knowledge can then be supplemented by live traversal prior to or during execution. **RT:** There are some important issues here that are not mentioned yet. Caches can only be used for determining exclusive groups if we can guarantee that it is “complete” with respect to the current query. This is for example not always the case if we’re executing different queries. And if freshness can not be guaranteed, then we can not ensure correctness of the exclusive group.

Authoritativeness Assumptions: Establishing exclusive groups in a decentralized dataspace could be achieved through the notion of *source authority* [33]. Under standard Link Traversal, any reachable source can assert triples about any subject (e.g., a third-party source asserting $\langle P1 \rangle \text{ foaf:knows } \langle P2 \rangle$), creating a “wild west” of data where no single source can be deemed the exclusive authority for a topic. This lack of authority prevents the query engine from determining if a set of patterns can be safely delegated to a single dataspace or endpoint.

RT: I'd position this around our work on guided link traversal, where adding additional assumptions on the link structures can help us optimize more. A possible solution is restricting the scope of validity such that a data source is the *sole authority* for triples where the subject corresponds to the source’s own URI (or a URI within its controlled namespace). By enforcing this *Subject Authority Constraint*, the query engine can statically determine that all triples matching $\langle \text{SourceA} \rangle \text{ ?p ?o}$ **RT:** In a footnote, mention that also fragments on this URL should be considered reside exclusively at *Source A*. This allows the engine to safely treat such patterns as an exclusive group, enabling the delegation of complex sub-queries (e.g., BGPs) to hybrid sources like SPARQL endpoints without the risk of incomplete results.

3.2. Integrating Dynamically Delegated Sub-queries into the Query Plan

When exclusive groups are identified and subqueries are dynamically pushed to expressive server-side interfaces, the resulting intermediate results must fit into the existing query plan.**RT:** This sentence is a bit oddly phrased. I'd move it after the next sentence, and just make it say that exclusive groups can currently only be identified during query planning.

In the traditional optimize-then-execute approach, used by Comunica-link-traversal [27, 28] and SQUIN [32] (except in [34]), the query plan is constructed before execution and evaluated over dynamically discovered sources.

Thus, an exclusive group can only be integrated if the precomputed plan already produces an intermediate result that exactly matches the triple patterns in that group.**RT: I don't really understand this sentence.**

Example 3.1. As an example, consider a query plan with the join structure

$$(T_1 \bowtie T_2) \bowtie (T_3 \bowtie T_4).$$

If the engine discovers an exclusive group consisting of the triple patterns $\{T_2, T_3\}$, this group cannot be integrated into the plan. The plan never produces an intermediate result of the form $T_2 \bowtie T_3$, since these patterns occur in different subtrees. As a result, the exclusive group does not fit into the existing join structure and cannot be pushed as a single subquery.**RT: Ah. Now I understand what you mean. Still, I find it a weird way to explain it. I would suggest explaining it as the need for adaptively modifying the query plan as soon as an exclusive group has been identified, which may lead to branches of the plan being cut off, and joins branch being transformed. (similar to your example, maybe you could show a before/after plan once an exclusive group has been identified.)**

3.2.1. Research Avenues

A clear path forward is the use of adaptive query processing techniques. These techniques should enable low-cost plan switching without discarding intermediate results, as the large number of sources in a dataspace would otherwise lead to plan thrashing. Algorithms such as Eddies [35], SteMs [36] (as used in [34]), and STAIRs [37] are viable candidates due to their tuple-level adaptivity.

Despite their potential, the literature lacks a comprehensive analysis of adaptive frameworks within the context of LTQPRT: Well, there's Olaf's walking without a map paper, which surely must be mentioned here, and also Jonni's paper.. Specifically, there is little guidance on designing routing strategies, the logic determining how tuples are forwarded between operators, or the overhead costs associated with each.

Eddies offer a constrained form of adaptivity. In this model, the join operators are instantiated at the start of the query; the engine only adapts the *order* in which tuples visit these operators. Because these join operators (e.g., Symmetric Hash Joins) accumulate internal state as they process data, the engine is effectively locked into the existing operator instances.

Example 3.2. Consider a query with triple patterns $\{T_1, T_2, T_3, T_4\}$. If an Eddy starts by routing results from $T_1 \bowtie T_2$, the join state is physically stored within that specific operator's hash tables. If the engine later discovers that $T_2 \bowtie T_3$ is more selective due to newly traversed data, it can change the routing sequence, but it cannot easily dissolve the $T_1 \bowtie T_2$ state to form a $T_2 \bowtie T_3$ join without significant re-initialization overhead.

In contrast, **SteMs** (State Modules) decouple state from the join operation entirely. Since SteMs do not hold fixed join states, any tuple can be routed to any SteM in any order. However, this flexibility introduces a *recomputation tax*: if a SteM is probed before its join partners have arrived, the intermediate result is not captured and must be re-derived once more data is discovered via traversal.

Finally, **STAIRs** (Storage, Transformation and Access for Intermediate Results) attempt to bridge this gap through state migration. This allows a query engine to physically move join states between different operator instances. While this maximizes flexibility, state migration is a resource-intensive operation. Determining the optimal threshold for migration, and the number of required migrations to incorporate the diverse precomputed sub-queries discovered during traversal is an open challenge.

3.3. Service Discovery in Hybrid Query Environments

A fundamental challenge in Hybrid Link Traversal Querying is the gap between resource identifiers in Linked Data documents and the aggregated query services that host them. For standard LTQP, the reachability of data is determined by the recursive dereferencing of URIs, adhering to the “follow-your-nose” principle where a URI identifies a resource and provides its description. However, SPARQL endpoints function as aggregators: they contain data about resources but do not necessarily share the resource’s URI namespace, nor are they always explicitly linked from the resource descriptions themselves. Consequently, a query engine may possess a URI for a resource, but lack the URI pointing towards the SPARQL endpoint that contains that resource, rendering the data unreachable despite being publicly accessible.**RT: I don't understand this last sentence that well. Could you rephrase? Or even rephrase with an example?**

3.3.1. Research Avenues

Explicit Linkage Mechanisms An avenue for mitigating the service discovery gap is by using explicit links between resources and their hosting endpoints. This approach relies on extending the “follow-your-nose” principle to include service metadata within the resource description itself. Existing methods such as the Vocabulary of Interlinked Datasets (VoID) can embed pointers directly in the RDF data. For instance, a triple taking the form $\langle r, \text{void:sparqlEndpoint}, s \rangle$ allows a traversal engine to immediately identify that the resource r is queryable via the endpoint s . **RT: Does void:sparqlEndpoint allow regexes? In any case, we should mention something about generic patterns, as listing all resources may be unfeasible.** However, this relies on adoption of this approach while, for example, currently only 33% percent of endpoints expose VoID descriptions [38].

Service Registries and Indexing. Where direct linkage is absent, research directs towards the usage of external service registries and catalogs. Unlike linkage-based discovery, this approach treats the mapping between URI namespaces and SPARQL endpoints as a separate knowledge base. Within a dataspace, services such as ESPRESSO [25] can aggregate these connections, effectively bridging the gap between identifiers and query services. However, relying on purely centralized indexes is antithetical to the goals of a decentralized dataspace. Consequently, future directions should investigate distributed alternatives, such as Peer-to-Peer (P2P) discovery mechanisms, to prevent the reliance on single points of failure.

3.4. Alternative View Deduplication

In decentralized dataspaces, a single dataset can often be accessed through multiple architectural views. Consider a personal data store (PDS) following the Solid protocol [39]. Here, data is structured as a hierarchical resource tree where documents are interconnected via RDF predicates**RT: Mention LDP here**. A Link Traversal Query Processing (LTQP) engine navigates this store by dereferencing these URIs on-the-fly.

Data providers may also expose a SPARQL endpoint over the same PDS to improve query efficiency. Ideally, an LTQP engine should incorporate this endpoint into its query plan. However, because endpoint discovery is itself a byproduct of the traversal process, an engine may dereference several hypermedia documents before it encounters the metadata pointing to the endpoint.

This creates a synchronization problem: the engine ingests the same data from two different interfaces, leading to duplicated results and wasted computational resources. In the specific context of Solid, a client might assume that any (public) URI sharing the Pod’s root namespace is covered by the discovered endpoint. However, in general dataspace environments, this assumption is often unsubstantiated; alternative views may only cover specific subsets of data, such as historical archives versus high-frequency data.

RT: I like the direction of that last sentence! Let's include the nuances from our discussion on Tuesday here more explicitly (that views may not contain exactly the same data.)

3.4.1. Research Avenues

View Overlap Detection To determine an appropriate deduplication strategy, a query engine must first assess the data coverage of the discovered access interfaces. One approach from the literature involves issuing queries to both the local data store and the remote endpoint to calculate *coherence* [19], a metric indicating the extent to which the two sources agree. However, this method is restricted to queryable interfaces and incurs significant overhead, as each coherence check requires additional HTTP requests. An alternative is to rely on server-side metadata that explicitly denotes which underlying resources a given access interface aggregates. Unfortunately, this relies heavily on the widespread adoption of specific metadata vocabularies by data providers, which cannot easily be guaranteed in a decentralized, autonomous dataspace. **RT:** For this one (also also others for that matter, so might be better for conclusions), you could mention that this is best to take up within standardization efforts (e.g. dataspaces, LWS, ...)

Client-Side Deduplication Regardless of whether the data coverage of discovered access interfaces is known, the query engine must perform deduplication on the client side. **RT:** There's some nuance to explain here: 2 cases: if the overlap could be determined upfront, then no filtering afterwards is needed, as requests that would produce those duplicate results can be avoided. But if overlap if detected later, then we need such filtering indeed (either for everything, or only up to the point of where result-producing-requests can be avoided) This requires tracking the *provenance* (i.e., the source) of bindings **RT:** I would call it "intermediary results", as not everyone here may know this terminology during query processing to filter out duplicates based on the computed or assumed coverage of alternative interfaces. A naive approach is to maintain a "seen" set of previously produced bindings for each triple pattern, discarding any newly discovered matches that already exist in the set. However, retaining a complete history of bindings is highly memory-intensive and scales poorly for large queries. While probabilistic data structures like Bloom filters can drastically reduce this memory footprint, their inherent false-positive rates risk accidentally discarding unique, valid results.

If the engine knows which sources are covered by an alternative access interface, it can attempt further optimizations. **RT:** Aha, this is where you mention this second case. Might be good to announce the 2 cases in the beginning if this part. First, upon discovering the interface, any relevant links that have been discovered but not yet dereferenced can be pruned from the traversal queue. Second, the engine can better manage the lifecycle of the "seen" sets. Once the engine can guarantee that all data capable of producing duplicate results has been processed, the associated "seen" sets can be safely discarded to free memory. However, the mechanism for reliably detecting this point during execution remains an open research area.

4. Conclusion

Hybrid querying over heterogeneous sources has the potential to significantly reduce query latency and network overhead, by combining the dynamic discovery of link traversal with the computational efficiency of expressive interfaces like SPARQL endpoints,

However, as outlined in this paper, realizing a fully functional Hybrid LTQP engine introduces substantial technical hurdles. The unbounded nature of the web makes it difficult to establish exclusive groups for safe sub-query delegation without relying on prior knowledge, discoverable indexes, or strict authoritativeness assumptions. Furthermore, query engines must adopt highly adaptive processing techniques to seamlessly integrate dynamically discovered materialized sub-queries without thrashing the query plan or losing intermediate results. We also highlight the ongoing need for robust service discovery mechanisms to bridge the gap between resource identifiers and the endpoints aggregating them. Finally, we identify a challenge with alternative views and the inherent risk of result duplication due to them and identify two research directions for this challenge.

Ultimately, transitioning from a "lowest common denominator" traversal strategy **RT:** It's a bit vague what you mean by this. I assume it's about the fact that we only can send triple patterns to SPARQL

endpoints? to a fully hybrid approach is essential to leverage the diverse access interface expected in dataspaces. Addressing these open research avenues will be critical in developing query engines capable of scaling to the realistic demands of a decentralized data ecosystem.

RT: Could you include acks for my FWO funding?

Declaration on Generative AI

During the writing of this paper, the author(s) used Gemini in order to: Sentence Polishing, Rephrasing, and Text Creation. After using these tool(s)/service(s), the author(s) reviewed and edited the content as needed and take(s) full responsibility for the publication's content.

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