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A systematic overview of data federation systems — A simplified version

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This document is a simplified version of the paper published in the Semantic Web Journal and provided a systematic overview of 50 data federation systems. The motivation of this document is to provide a way of making users, such as data federation consumers and developers, to check or get knowledge of the characteristics of the considered data federation systems quickly rather than reading a paper with 60 pages.

Table 1 Summary of the selected data federation systems. Academic systems in italics

2	System	Provider	data federation systems. Academic systems in <i>italics</i> Description
3			
4	AllegroGraph [1]	Franz Inc.	Distributed graph & document DB supporting OWL, SPARQL, SHACL and federation
5	Amazon Athena [2] Amazon Neptune [4]	Amazon.com, Inc. Amazon.com, Inc.	Inter. cloud query service for Amazon S3 data, based on Presto [3] Fully-managed cloud graph DB (property graph, RDF), part of Amazon AWS
6	AnzoGraph DB [5]	Cambridge Semantics	Massively-parallel distributed graph DB (property graph, RDF) for large-scale analytics
7	Apache Drill [6, 7]	Apache Software Foundation	Distributed schema-free engine for interactive SQL queries on heterogeneous & nested data, inspired by Dremel [8]
8	Apache Jena [9]	Apache Software Foundation	SPARQL query engine of Jena framework and TDB triple store, supporting federation
9	Apache Spark [10, 11]	Apache Software Foundation	Multi-lang. (incl. SQL) distributed engine for large-scale data processing & analytics
10	BigDAWG [12, 13]	Intel Science & Technology Center for Big data	Polystore with heterogeneous storage engines for time series (SciDB), text (Accumulo) and relational data (PostgreSQL)
11	Blazegraph [14]	Systap	Triple store supporting SPARQL 1.1 federation and powering Wikidata (via a fork)
12	CloudMdsQL [15, 16]	Inria & LIRMM	Polystore integrating heterogeneous storage engines (incl. RDBMS, NoSQL, HDFS)
13	Comunica [17]	Univ. Ghent	Modular JS federated query engine for heterogeneous web sources, incl. SPARQL endpoints
14	CostFed [18]	Univ. Leipzig	Index-assisted, cost-based data federation system for SPARQL endpoints
15	DARQ[19]	Univ. HU Berlin	Earliest data federation system for SPARQL endpoints, cost-based
16	Data Virtuality [20]	Data Virtuality GmbH	Heterogeneous data integration solution combining data virtualization and ETL
	Denodo [21]	Denodo Technologies Inc.	Data virtualization solution for heterogeneous sources, also available as cloud service
17	Dremio [22] FEDRA [23]	Dremio Corporation Univ. Nantes (LINA lab.)	Data "lakehouse" (lake + warehouse) solution supporting heterogeneous data sources Data federation system for SPARQL endpoints exploiting data replication
18	FedX (RDF4J) [24, 25]	fluid Operations AG	On-demand (no statistics, query-time) data federation system for SPARQL endpoints
19	GraphDB [26]	Ontotext	Triple store featuring OWL reasoning, SPARQL federated queries & RDBMS access
20	HiBISCuS [27]	Univ. Leipzig	Source selection for SPARQL data federation (DARQ, FedX & SPLENDID extension)
21	IBM Cloud Pak for Data [28]	IBM	Data federation system with data discovery, governance, security and privacy solutions, also available as cloud service (formerly IBM Cloud Private for Data)
22	IBM Db2 Big SQL [29]	IBM	Massively-parallel Hadoop SQL engine for heterogeneous sources (formerly IBM SQL)
23	IBM InfoSphere Federation Server [30]	IBM	SQL-based data federation system for heterogeneous sources (formerly WebSphere Federation Server)
24 25	JBoss Data Virtualization [31]	Red Hat, Inc.	Data federation system based on Teiid and providing read/write access to heterogeneous sources, data security, and multiple user interfaces / APIs
26	Metaphactory [32, 33]	metaphacts GmbH	KG platform on top of SPARQL endpoints with two federation engines (Ephedra, FedX)
	Myria [34]	Univ. Washington	Cloud service for big data management/analytics with parallel & federated query engine
27	Neo4j (Fabric) [35]	Neo4j, Inc.	Federation solution of Neo4J graph DB (Cypher [36] queries on property graph model)
28 29	Obi-Wan [37, 38]	Inria & Polytechnic Institute of Paris	Ontology-Based Data Access (OBDA) [39] system on top of Tatooine [40] mediator for heterogeneous sources
30	Odyssey [41]	Univ. Aalborg & Univ. Nantes	Statistics & cost-based optimizer for SPARQL data federation (FedX extension)
31	Ontario [42]	L3S Research Center	Heuristics-based system using RDF Molecule Templates (introduced by its predecessor MULDER [43]) to describe/map source content as star-shaped RDF instance descriptions
32	Onto-KIT [44]	Univ. Toulouse	Data federation system focusing on Earth Observation data with hypergraph-based data model and query processing techniques
33	Oracle Big Data SQL [45]	Oracle Corporation	Data federation system for Oracle DB that accesses Hadoop storage & processing
34 35	Oracle DB (Spatial & Graph) [46, 47]	Oracle Corporation	Oracle DB component for semantic technologies with data federation capabilities (RDF views) over relational, graph, and RDF (SPARQL) sources
36	PolyWeb [48, 49]	Univ. NUI Galway	SPARQL-based data federation system for different sources on the Web (RDF & CSV data, RDBMS), focusing on source selection, query optimization & execution
37	Presto [3, 50]	Presto Foundation	SQL-based distributed query engine suitable to interactive (big) data analytics
38	Querona Data Virtualization [51]	YouNeedIT Sp. z o.o. Sp. k.	Data federation system for a variety of heterogeneous sources, based on Apache Spark and targeting big data analytics with the support of main BI tools
39	RDFLib [52]	RDFLib team	A pure Python package for working with RDF, supporting SPARQL 1.1 federation
40	SAFE [53]	Insight SFI Research Centre for Data Analytics	Data federation system for SPARQL endpoints exposing RDF data cubes with sensitive data, featuring access policy-aware data summaries, source selection & query execution
41	SAGE [54]	Univ. Nantes	SPARQL engine with "web preemption" (i.e., query suspend/resume) & federation capabilities
42	SAP HANA [55]	SAP SE	In-memory DB targeting with data federation capabilities, also available as cloud service
43	SAS Federation Server [56]	SAS Institute	Data federation system featuring data caches, masking, encryption & quality functions
44	SemaGrow [57]	IIT NCSR 'Demokritos'	Data federation system for SPARQL endpoints with statistics-based query optimization
	SPLENDID [58]	Univ. Koblenz-Landau	Data federation system for SPARQL endpoints that provide VOID [59] data statistics
45	SQL Server (PolyBase) [60]	Microsoft Corporation	SQL Server component for data federation supporting Hadoop and Azure storage
46	Squerall [61] Starburst [62]	Univ. Bonn Starburst Data, Inc.	Data federation system for heterogeneous sources built on Spark, Presto, and RML mappings Commercial distribution of Trino, extra security features, available on-premise/on-cloud
47	Stardog [63]	Stardog Union	KG platform including data federation of heterogeneous sources & query-time inference
48	Teiid [64]	Red Hat, Inc.	SQL-based engine for data federation of heterogeneous sources
49	TIBCO Data Virtualization [65]	TIBCO Software Inc.	Data federation system for heterogeneous sources, with data caching & security, massively parallel processing & GUI tools (formerly Composite, then Cisco Data Virtualization)
50	Trino [66]	Trino Software Foundation	SQL-based query distributed engine for interactive big data analytics, forked from Presto
51	Virtuoso [67–69]	OpenLink Software	Multi-model DB (object-relational, RDF, XML) with data federation facilities

Table 2

Evaluation of query language and data source sub-dimensions. Academic systems in italics. "-" denotes feature/information not found in the systems' official documentation, websites, or academic publications, to the best of our efforts

	Query language					Data			
System	SPARQL	SQL	Other	Relational	Graph- based	Aggregate- oriented	Structured Files	Web Service Paradigms	Other
AllegroGraph	√	_	Prolog	-	✓	-	_	-	_
Amazon Athena	-	✓	-	✓	✓	✓	✓	-	✓
Amazon Neptune	✓	_	-	-	✓	-	-	_	_
AnzoGraph DB	✓	_	Cypher	✓	✓	-	✓	✓	_
Apache Drill	_	✓	-	✓	-	✓	✓	✓	✓
Apache Jena	✓	_	_	_	✓	_	_	_	_
Apache Spark	_	✓	_	✓	_	_	✓	-	_
BigDAWG	_	_	BigDAWG Query	✓	_	/	_	-	/
Blazegraph	/	_		_	✓	_	_	_	_
CloudMdsQL	_	_	CloudMdsQL	✓	/	/	_	_	_
Comunica	✓	_	GraphQL	_	√	_	✓	_	
CostFed	/	_	-	_	✓	_	_	_	_
DARQ	√	_	_	_	~	_			
· -	_	- ✓	-		√	- ✓	<i>-</i> ✓	_ ✓	_
Data Virtuality		√		V ✓		∨ ✓	∨ ✓	∨ ✓	· /
Denodo	-	√	GraphQL	V ✓	-	∨ ✓	∨ ✓		V
Dremio		-	-		-		•	-	_
FEDRA	√	-	-	-	✓	-	-	-	-
FedX (RDF4J)	√	-	-	_	√	-	-	-	-
GraphDB	√	✓	Cypher	✓	✓,	-	-	-	-
HiBISCuS	✓	-	-		✓	-,	-	-,	-
IBM Cloud Pak for Data	-	√	-	√	-	✓	✓,	✓	√
IBM Db2 Big SQL	-	√	-	√	-	✓	✓	-,	✓
IBM InfoSphere Federation Server	-	✓	-	√	-		✓.	✓.	✓.
JBoss Data Virtualization	-	\checkmark	-	√	-	✓	\checkmark	\checkmark	\checkmark
Metaphactory	✓	-	-	✓	✓	✓	-	\checkmark	-
Myria	_	\checkmark	MyriaL	_	✓	✓	✓	-	\checkmark
Neo4j (Fabric)	_	-	Cypher	-	✓	-	-	-	-
Obi-Wan	✓	-	-	✓	✓	\checkmark	_	-	-
Odyssey	✓	-	-	-	✓	-	-	-	-
Ontario	✓	_	_	✓	✓	✓	✓	_	_
Onto-KIT	✓	_	_	_	_	_	✓	_	_
Oracle Big Data SQL	_	✓	_	✓	_	✓	✓	_	✓
Oracle DB (Spatial & Graph)	✓	_	_	✓	✓	_	-	_	_
PolyWeb	✓	_	_	✓	✓	_	✓	_	_
Presto	_	/	_	√	_	✓	_	_	✓
Querona Data Virtualization	_	/	_	/	_	√	✓	_	/
RDFLib	✓	_	_	_	✓	_		_	_
SAFE	· /	_	_	_	· /	_	_	_	_
SAGE	· /	_	_	_	√	_	_	_	_
SAP HANA	_	/	_	✓	_	_	_	_	/
SAS Federation Server	_	_	FedSQL	√	_			_	/
		_	reasQL	_	- ✓	_	_	-	V
SemaGrow	· ,	_	_	_	,	_	_	_	_
SPLENDID	√	- ✓	_	_ ✓	✓	_ ✓	<i>-</i> ✓	-	_
SQL Server (PolyBase)		V	-	√	-	✓ ✓	✓ ✓	-	_
Squerall		-/	-		-			-	-/
Starburst	-	✓	-	√	-	√	√	-	\
Stardog	✓	-	-	√	\checkmark	✓	✓	-	√
Teiid	-	✓_	-	✓	-	✓	✓	✓	√
TIBCO Data Virtualization	-	✓_	-	√	-	✓	✓	\checkmark	√
Trino		√	-	✓,	-	✓	-	-	✓
Virtuoso	√	✓	_	√	✓				
Number	27	22	10	32	30	24	24	10	20

Table 3

Evaluation of the data security dimension. Academic systems in italics. "-" denotes feature/information not found in the systems' official documentation, websites, or academic publications, to the best of our efforts. Subscript ng denotes the use of named graph-based solutions to hide (mask) sensitive information in selected graphs to certain users, and possibly (for AnzoGraph DB) expose sanitized named graph views

Data security

System	Data security									
System	Authentication	Authorization	Auditing	Encryption	Data masking					
AllegroGraph	√	√	_	_	-					
Amazon Athena	✓	✓	✓	✓	_					
Amazon Neptune	✓	✓	✓	✓	_					
AnzoGraph DB	· /	· ✓	_	_	\checkmark_{ng}					
Apache Drill	✓	√	_	✓	- ng					
Apache Jena	· /	_	_	_	_					
Apache Spark	✓	✓	_	✓	_					
BigDAWG	_	_	-	_	_					
Blazegraph	_	_	_	_	_					
CloudMdsQL	_	_	_	_	_					
Comunica	_	_	_	_	_					
CostFed	_	_	_	_	_					
DARQ	_	_	_	_	_					
Data Virtuality	✓	✓	_	_	_					
Denodo	√	·	✓	✓	✓					
Dremio	· /	· /	_	· /	<i>'</i>					
FEDRA		_		-	<u> </u>					
FedX (RDF4J)		_	_	-	_					
GraphDB	√		_ 	_ 	_					
HiBISCuS	v	v	v -	v	_					
IBM Cloud Pak for Data			_ ✓	_ ✓	_ ✓					
	V	√	· · · · · · · · · · · · · · · · · · ·	· /	v					
IBM Db2 Big SQL	V	∨ ✓	V	√	_					
IBM InfoSphere Federation Server	· .				-					
JBoss Data Virtualization	√	√	-	✓	-					
Metaphactory	-	✓	-	-	-					
Myria	_	-	-	-	-					
Neo4j (Fabric)	√	✓	-	-	-					
Obi-Wan	_	-	-	-	-					
Odyssey	-	-	-	-	-					
Ontario	_	-	-	-	-					
Onto-KIT	_		-	-	-					
Oracle Big Data SQL	√		-	-	-					
Oracle DB (Spatial & Graph)	✓	✓	-	✓	✓					
PolyWeb	-	-	-	-	-					
Presto	√	√	✓	-	-					
Querona Data Virtualization	✓	✓	-	✓	✓					
RDFLib	_	-	-	-	-					
SAFE	✓	✓	-	-	-					
SAGE	_	-	-	-	-					
SAP HANA	√	-,	-	-,	-,					
SAS Federation Server	✓	✓	-	✓	✓					
SemaGrow	_	-	-	-	-					
SPLENDID	<u> </u>	-	-	-	-					
SQL Server (PolyBase)	✓	\checkmark	✓	✓	-					
Squerall	-	-	-	-	-					
Starburst	✓.	✓	✓	✓	-					
Stardog	✓	✓	-	-	\checkmark_{ng}					
Teiid	✓	✓	-	✓	-					
TIBCO Data Virtualization	✓	✓	-	✓	-					
Trino	✓	✓	-	✓	_					
Virtuoso	✓	✓	_	✓	-					
Number	32	29	10	20	8					

Table 4 Evaluation of the interface dimension. Academic systems in italics. "-" denotes feature/information not found in the systems' official documentation, websites, or academic publications, to the best of our efforts

		G 1	Application programming interface						
System	Graphical interface	Command line interface	JDBC Driver	ODBC Driver	Web API	ADO.NET	SPARQL HTTP API		
AllegroGraph	√	✓	-	-	✓	_	✓		
Amazon Athena	 	✓	\checkmark	✓	-	_	-		
Amazon Neptune	✓	✓	✓	-	\checkmark	-	-		
AnzoGraph DB	✓	✓	_	-	✓	-	✓		
Apache Drill	✓	✓	✓	✓	✓	_	-		
Apache Jena	_	✓	✓	-	-	_	✓		
Apache Spark	✓	✓	✓	✓	_	_	_		
BigDAWG	_	✓	_	_	✓	_	_		
Blazegraph	√	✓	_	_	✓	_	_		
CloudMdsQL	_	✓	✓	_	_	_	_		
Comunica	✓	✓	_	_	_	_	✓		
CostFed	/	_	_	-	_	_	_		
DARQ	-	✓	_	_	_	_	_		
Data Virtuality	/	/	/	✓	✓	_	_		
Denodo	/	<i>'</i>	· /	√	<i>'</i>	✓	_		
Dremio	/	/	· /	√	· /	_	_		
FEDRA	_		_	_	_	_	_		
FedX (RDF4J)		<i>_</i>	_	_	- ✓	_	- ✓		
GraphDB	V	√	_	_	√	_	∨ ✓		
HiBISCuS		_	_	-	v	_	_		
IBM Cloud Pak for Data		- ✓		_	_ ✓				
		√				-	-		
IBM Db2 Big SQL		∨ ✓	∨ ✓	V	- ✓	_	-		
IBM InfoSphere Federation Server	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	√	V		√	-	-		
JBoss Data Virtualization	· .	· .				-	_		
Metaphactory	/	√	-	-	√	-	✓		
Myria		√	_	-	√	-	-		
Neo4j (Fabric)	√	✓	✓	-	✓	-	-		
Obi-Wan	-	√	-	-	-	-	✓		
Odyssey	-	✓	-	-	-	-	-		
Ontario	_	✓	-	-	-	-	-		
Onto-KIT	/	-	-	-	-	-	-		
Oracle Big Data SQL	✓	✓.	-	-	-,	-	-,		
Oracle DB (Spatial & Graph)	√	✓	-	-	✓	-	✓		
PolyWeb		-,	-,	-,	-,	-	-		
Presto	√	✓.	√	✓.	✓		-		
Querona Data Virtualization	√	√	✓	✓	-	✓	-		
RDFLib	-	✓	-	-	-	-	-		
SAFE	-	-,	-	-	-	-	-		
SAGE	√	✓	-	-,	-,	-	-		
SAP HANA	√	✓	✓.	✓.	✓.	✓	_		
SAS Federation Server	✓	✓	✓	✓	\checkmark	-	-		
SemaGrow	✓	✓	-	-	_	-	_		
SPLENDID	-	✓	-	-	-	-	-		
SQL Server (PolyBase)	√	✓	✓	✓	-	✓	_		
Squerall	✓	✓	-	-	-	-	-		
Starburst	✓	✓	✓	✓	✓	-	-		
Stardog	✓	✓	-	-	✓	-	✓		
Teiid	✓	✓	✓	✓	✓	✓	-		
TIBCO Data Virtualization	✓	✓	✓	✓	✓	✓	-		
Trino	/	✓	✓	✓	✓	_	_		
Virtuoso	✓	√	√	✓	✓	✓	✓		
Number	38	45	24	18	27	7	11		

Table 5

Evaluation of development dimension. Academic systems in italics. "F." and "L." denote "First" and "Latest" respectively. Subscript letters further qualify available deployment options: n = native; c = containerized; a = Amazon AWS; m = Microsoft Azure; g = Google Cloud Platform."-" denotes feature/information not found in the systems' official documentation, websites, or academic publications, to the best of our efforts

	Main dev	velopmer	t language	D	eployment		Comm.	Open		Rele	ease	
System	C/C++	Java	Others		IaaS/PaaS	SaaS	support	source	F. Year	F. Version		L. Version
AllegroGraph	_	√	Lisp	\checkmark_{nc}	\checkmark_{am}	_	√	_	2004	6.4.0	2021	7.2.0
Amazon Athena	-	\checkmark	_	-	_	\checkmark_a	✓	-	2017	_	2021	_
Amazon Neptune	_	✓	_	_	_	\checkmark_a	✓	_	2018	1.0.1.0	2021	1.0.5.1
AnzoGraph DB	✓	-	_	\checkmark_{nc}	\checkmark_a	-	✓	_	-	2.0	2021	2.3
Apache Drill	_	✓	_	\checkmark_{nc}	-	_	_	✓	2012	M1	2021	1.19
Apache Jena	_	✓	_	\checkmark_n	_	_	_	✓	2012	2.7.0	2021	4.2.0
Apache Spark	_	_	Scala	\checkmark_{nc}	_	-	_	✓	2014	1.0	2021	3.2.1
BigDAWG	_	✓	_	\checkmark_n	_	_	_	✓	2015	_	2017	0.0.5
Blazegraph	_	✓	_	\checkmark_n	_	_	_	✓	2019	2.1.5	2020	2.1.6rc
CloudMdsQL	_	✓	_	\checkmark_n	_	_	_	✓	2017	_	2017	_
Comunica	_	_	JavaScript	\checkmark_{nc}	_	_	_	✓	2018	1.0.0	2021	1.22.3
CostFed	_	✓	_	\sqrt{nc}	_	_	_	✓	2016	_	2018	_
DARQ	_	✓	_	\checkmark_n	_	_	_	✓	2006	_	2008	_
Data Virtuality	_	_	_	\checkmark_{nc}	_	_	✓	_	_	_	2021	2.4
Denodo	_	_	_		/	_	√	_	2002	1.0	2020	8.0
Dremio	_	/	_	\sqrt{nc}	\sqrt{amg}	_	· /	✓	2017	1.1	2021	19.0
FEDRA	_	~	_	√ nc	\checkmark_{am}	_	_	✓	2017	_	2015	-
FedX (RDF4J)	_	/	_	\checkmark_n	_	_	<i>-</i>	✓	2013	_	2013	3.7.4
GraphDB	_	√	_	\checkmark_n	_	_	√	_	2011	6.2	2021	9.10
HiBISCuS	_	√	_	\checkmark_{nc}	_	_	_	<i>-</i> ✓	2013	1	2014	1
				\checkmark_n	,							
IBM Cloud Pak for Data	-	<i>-</i> ✓	-	√ _c	\checkmark_{amg}	√		-	2018	2.1.0	2021	4.0
IBM Db2 Big SQL	-	V	-	\checkmark_n	-	√	√	-	2017	-	2020	7.1.0
IBM InfoSphere Federation Server	-	-	-	\checkmark_n	-	-	√	-	-	-	2019	10.5.0
JBoss Data Virtualization	-	✓	-	\checkmark_{nc}	-	-	✓	✓	2014	6.0.0	2018	6.4.0
Metaphactory	-	-,	-	✓	\checkmark_a	-	-		2015	-	2021	4.3.0
Myria	-	✓	-	\checkmark_n	-	-	-	✓	2014	1	2017	1
Neo4j (Fabric)	-	✓	-	\checkmark_{nc}	\checkmark_{amg}	✓	√	✓	2020	4.0.11	2021	4.3.7
Obi-Wan	-	✓	-	\checkmark_n	-	-	-	✓.	2020	-	2020	-
Odyssey	-	✓	-	\checkmark_n	-	-	-	✓	2016	-	2019	-
Ontario	-	-	Python	\checkmark_n	-	-	-	✓	2018	-	2021	-
Onto-KIT	-	\checkmark	-	\checkmark_n	-	-	-	✓	2020	-	2020	-
Oracle Big Data SQL	-	-	-	\checkmark_n	_	-	-	-	-	3.0.1	2021	4.1.1
Oracle DB (Spatial & Graph)	-	-	-	\checkmark_{nc}	-	-	✓	-	2016	-	2021	21c
PolyWeb	_	✓	_	\checkmark_n	_	-	-	✓	2017	_	2017	_
Presto	-	✓	_	\checkmark_{nc}	-	-	✓	✓	2013	0.54	2021	0.265.1
Querona Data Virtualization	_	_	_	\checkmark_n	_	_	✓	_	2015	_	2020	_
RDFLib	-	-	Python	\checkmark_n	-	-	-	✓	2002	1.1.1	2021	6.1.1
SAFE	-	✓	_	\checkmark_n	_	-	-	✓	2017	_	2017	_
SAGE	_	_	Python	\checkmark_{nc}	_	_	_	✓	2019	1.1	2021	2.3
SAP HANA	✓	_	_	\checkmark_{nc}	\checkmark_{ag}	_	✓	_	2018	1.0.SPS12	2020	2.0.SPS05
SAS Federation Server	√	_	_	\checkmark_n	- ag	_	✓	_	2013	3.2	2021	4.4
SemaGrow	_	/	_	\checkmark_{nc}	_	_	_	/	2014		2021	2.2.1
SPLENDID	_	✓	_	\checkmark_n	_	-	_	√	2011	_		_
SQL Server (PolyBase)	/	_	_	\checkmark_{nc}	_	_	✓	_	2016	2016	2019	2019
Squerall	_	_	Python	\checkmark_{nc}	_	-	_	✓	2018	0.1	2019	0.2
Starburst	_	/	-	\checkmark_{nc}	\checkmark_{amg}	-	/	_	2019	0.188-e	2021	364-e LTS
Stardog	_	/	_	\checkmark_{nc}	√ _{amg}	√	/	_	2011	0.7.3	2021	7.7.3
Teiid	_	√	_	\checkmark_{nc}	• a	_	/	√	2009	6.0.0	2021	16.0.0
TIBCO Data Virtualization	_		_		./		V	-	2007	7.0.5	2020	8.4.0
Trino	_	- ✓	_	\sqrt{nc}	\checkmark_{am}	_	V /	<i>-</i> ✓	2007	0.54	2021	364
Virtuoso	_ ✓	-	-	\checkmark_{nc}	\checkmark_{am}	_	√	-	_	-	2021	8.3
Number	5	31	7	49	12	6	26	30	_	_		_

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The following Table 6 lists the specific sources supported by each investigated data federation system, obtained from available systems' documentation and publications. Sources are classified on a *local*, *per-system* basis, along the source types defined in Section 6 of the original paper, with additional source information — such as the specific kind(s) of relational, graph-based or aggregate-oriented system — reported next to the source name via subscript letters (see table caption for legend). We remark the following:

- Some sources correspond to data access interfaces that can be configured to connect additional systems beyond the ones explicitly listed in the table. In particular, companies such as CData¹ and Progress² commercialize connectors for the relational SQL-based JDBC, ODBC, ADO.NET and OLE DB interfaces that can be used to access a myriad of heterogeneous data sources, possibly different from the ones listed in Table 6 (e.g., GraphQL sources via specific connectors³) and possibly using a different data model that is transparently adapted to the relational one by the connector (e.g., via flattening of nested data). In Table 6, besides the supported data access interfaces, we explicitly list only the sources that are directly and natively supported by a system without relying on such third party connectors / adapters.
- Structured files are distinguished from other source types with the same data model (e.g., relational sources for CSV files, aggregate-oriented specifically, document-based for JSON files) by virtue of direct access to raw file contents by the data federation system. In some cases, however, access to stored structured files may require metadata services external to the filesystem (e.g., Hive Metadata Store) for locating and interpreting file contents, or may leverage processing services (e.g., from Hadoop) co-located with the nodes storing the file in a distributed filesystem (e.g., HDFS), for instance to push down data access operations and computations (e.g., filtering, sorting) close to where raw file data reside, this way reducing communication costs.
- Some of the data federation systems investigated in this survey are also listed as supported sources (marked with * subscript) of other systems in Table 6, reflecting the fact that the virtual data sources obtained through data federation can be used themselves in downstream federations. As a limit case (e.g., AllegroGraph), a system may list only itself as a supported data source, which occurs when the system offers both storage and data federation capabilities, and the latter are restricted to instances of the same system.
- Test sources (e.g., emulating /dev/null) and system-specific connectors used to access configuration, performance or log data of the system itself are omitted in Table 6, for simplicity.

Table 6

Supported data sources. Academic systems in *italics*. Additional source information in subscript position: * = investigated system; a = specialized web API; r = RDF triple store; g = property graph store; k = key-value store; w = wide-column store; d = document store; g = search engine; g = hardware + software appliance; g = MDX (MultiDimensional expressions) support. SPARQLp denotes the SPARQL protocol

System	Relational	Graph- based	Aggregate-oriented	Structured Files	Web Service Paradigms	Other
AllegroGraph		Allegro- Graph _{r*}				
Amazon Athena	Amazon Redshift, MySQL, PostgreSQL, Vertica	Amazon Neptune _{rg*}	Amazon Document DB_d , Amazon Dynamo DB_d , Amazon OpenSearch $_s$, HBase $_w$, Redis $_k$	Common Log Format, CSV, JSON, ORC, Parquet		Amazon AWS System Manager Inventory _a , Amazon CloudWatch _a , Amazon Timestream
Amazon Neptune		SPARQLp				
AnzoGraph DB	Derby, Google BigQuery, Hive, HSQLDB, IBM DB2, Impala, JDBC, MariaDB, MS SQL Server _* , MySQL, PostgreSQL, SAP ASE	SPARQLp		CSV, JSON, Parquet, SAS7BDAT, SAS XPT, XML	HTTP / REST	

¹https://www.cdata.com/drivers/

²https://www.progress.com/connectors

³https://www.cdata.com/drivers/graphql/

1	System	Relational	Graph- based	Aggregate-oriented	Structured Files	Web Service Paradigms	Other
2 3 4 5 6	Apache Drill	Derby, Druid, Hive, H2, MS SQL Server*, MySQL, Oracle DB*, PostgreSQL		Cassandra _w , Elasticsearch _s , HBase _w , MapR-DB _w , MongoDB _d , Splunk _s	Avro, Common Log Format, CSV, Excel, JSON, Parquet, SequenceFile, XML	HTTP / REST	Kafka, OpenTSDB
7	Apache Jena		Jena API, SPARQLp				
8 9 10 11	Apache Spark	Hive, JDBC			any file (content field + metadata), Avro, CSV, JSON, ORC, Parquet		
12	BigDAWG	PostgreSQL		Accumulo _w	•		SciDB
13	Blazegraph		SPARQLp				
	CloudMdsQL	Derby	Sparksee _g	$MongoDB_d$			
14 15	Comunica		SPARQLp, TPF		RDF		
	CostFed		SPARQLp				
16	DARQ		SPARQLp				
17 18 19 20 21	Data Virtuality	Amazon Redshift, ClickHouse, Data Virtuality*, Derby, Exasol, Google BigQuery, Greenplum, Hive, HSQLDB, H2, IBM DB2, IBM Informix, IBM Netezzah, Ingres, JDBC, MDX**, MetaMatrix**, MS SQL Server**, MySQL, Oracle DB**, PostgreSQL, SAP ASE, SingleStore, Snowflake, Teradata	Neo4j _{g*}	$MongoDB_d$, $Redis_k$	CSV, Excel, JSON, XML	HTTP / REST	DHL Track & Trace _a , Google Ads _a , Google Analytics _a , InterSystems Caché, Kdb+, LDAP, ModeShape, Salesforce _a
22 23 24 25 26 27 28	Denodo	Amazon Athena*, Amazon Redshift, Databricks, Denodo*, Derby, Google BigQuery, Greenplum, Hive, IBM DB2, IBM Informix, IBM Netezzah, Impala, JDBC, MS Analysis Service*, MS Azure SQL Database, MS SQL Server*, MS Azure SQL Database, MS SQL Server*, MS Azure Synapse Analytics, Mondrian*, MySQL, Oracle DB*, Oracle Essbase*, Oracle TimesTen, PostgreSQL, Presto*, SAP ASE, SAP Business Warehouse*, SAP HANA*, Snowflake, Teradata, Trino*, Vertica, Yellowbrick*		Amazon OpenSearch _s , Cassandra _w , Elasticsearch _s , MongoDB _d	CSV, Excel, JSON, XML	SOAP / WSDL	ITPilot (website wrapper generator), LDAP, Salesforce _a , SAP Business _a
29 30	Dremio	Amazon Redshift, Hive, MS SQL Server*, MySQL, Oracle DB*, PostgreSQL, Teradata		Amazon OpenSearch _s , Elasticsearch _s , HBase _w , MongoDB _d	CSV, Excel, JSON, Parquet		
31	FEDRA		SPARQLp				
32 33	FedX (RDF4J)		RDF4J API, SPARQLp				
34	GraphDB	IBM DB2, MS SQL Server*, MySQL, Oracle DB*, PostgreSQL	GraphDB $_r$, SPARQLp				
35	HiBISCuS	*/ ************************************	SPARQLp				
36 37 38 39	IBM Cloud Pak for Data	Amazon Redshift, Derby, Google BigQuery, Greenplum, Hive, IBM DB2, IBM Db2 Big SQL*, IBM Db2 Warehouse, IBM DVM, IBM Informix, IBM Netezzah, Impala, MariaDB, MS SQL Server*, MySQL, Oracle DB*, PostgreSQL, SAP ASE, SAP HANA*, Snowflake, Teradata	ν. 1	${\bf MongoDB}_d$	CSV, Excel	OData	IBM Db2 Event Store, Salesforce _a , SAP Gateway OData _a
40 41 42 43 44 45	IBM Db2 Big SQL	Amazon Athena*, Amazon Redshift, Derby, Google BigQuery, Greenplum, Hive, IBM DB2, IBM Db2 Big SQL*, IBM Db2 Warehouse, IBM DVM, IBM Informix, IBM Integrated Analytics System\$\mu\$, IBM Netezza\$\mu\$, IBM PureData\$\mu\$, Impala, MariaDB, MS Azure SQL Database, MS SQL Server*, MySQL, Oracle DB*, PostgreSQL, SAP ASE, SAP HANA*, Teradata		Amazon OpenSearch s , CouchDB $_d$, MongoDB $_d$	Parquet		IBM MQ, Salesforce _a
46 47 48	IBM InfoSphere Federation Server	IBM DB2, IBM Informix, MS SQL Server*, Oracle DB*, SAP ASE, Datacom/DB, Teradata, IBM Netezzah			Excel, XML	SOAP / WSDL	BioRS, IBM MQ, IDMS, IMS

System	Relational	Graph- based	Aggregate-oriented	Structured Files	Web Service Paradigms	Other
JBoss Data Virtualization	Actian Vector, Amazon Redshift, Exasol, Greenplum, Hive, Hive, IBM DB2, IBM Informix, IBM Netezza _h , Impala, Ingres, JBoss Data Virtualization _* , MariaDB, MetaMatrix _* , MS Access, MS SQL Server _* , Mondrian _m , MySQL, Oracle DB _* , PostgreSQL, Presto _* , SAP ASE, SAP HANA _* , SAP IQ, Teradata, Vertica		Accumulo _w , Amazon OpenSearch _s , Cassandra _w , Couchbase _d , HBase _w , MongoDB _d , Red Hat Data Grid _k , Solr _s	CSV, Excel, XML	HTTP / REST, OData, SOAP / WSDL	Google Sheets _a , LDAP, ModeShape OSIsoft PI, Red Ha Directory Server, Salesforce _a , SAP Gateway OData _a
Metaphactory	JDBC	Amazon Neptune _{rg*} , GraphDB _r , SPARQLp, Stardog _{r*} , Virtuoso _{r*}	Elasticsearch _s		HTTP / REST	
Myria		SPARQLp	Amazon OpenSearchs	CSV		SciDB
Neo4j (Fabric)		Neo4j _{g*}				
Obi-Wan	PostgreSQL	Jena TDB _r	$MongoDB_d$, $Redis_k$			
Odyssey	TostglesQE	SPARQLp	Mongo Da, Redisk			
Ontario	MySQL	Neo4j _{g*} ,	$MongoDB_d$	CSV, XML		
Ontario	MySQL	SPARQLp	WollgoDbd			
Onto-KIT				CSV, ENVI, JSON		
Oracle Big Data SQL	Hive		$HBase_w$, $Oracle$ $NoSQL_k$	Avro, CSV, JSON, ORC, Parquet, XML		Kafka
Oracle DB (Spatial & Graph)	Oracle DB _*	SPARQLp				
PolyWeb	MySQL	SPARQLp		CSV		
Presto	Amazon Redshift, Druid, Google BigQuery, Hive, Iceberg, Kudu, MS SQL Server _* , MySQL, Oracle DB _* , Pinot, PostgreSQL		Accumulo _w , Cassandra _w , Elasticsearch _s , MongoDB _d , Redis _k			Kafka, Prometheu
Querona Data Virtualization	Actian Matrix, Actian Vector, ADO.NET, Alibaba AnalyticDB for MySQL, Alibaba Data Lake Analytics, Amazon Athena*, Amazon Aurora, Amazon Redshift, ClickHouse, Databricks, dBASE, Denodo*, Drill*, Exasol, Google BigQuery, IBM DB2, JDBC, MariaDB, MS Access, MS SQL Server*, MS Azure Synapse Analytics, MySQL, ODBC, OLE DB, Oracle DB*, PostgreSQL, SAP HANA*, SAS Scalable Performance Data Server, Spark*, Teradata, Teradata Aster, Vertica		Amazon OpenSearch $_s$, DataStax $_w$	CSV, Excel, MSG/EML (email), PDF (metadata)		Kafka
RDFLib		SPARQLp				
SAFE		SPARQLp				
SAGE SAP HANA	Amazon Athena*, Google BigQuery, IBM DB2, IBM Netezzah, MS SQL Server*, Oracle DB*, SAP ASE, SAP HANA*, SAP IQ, SAP MaxDB, Teradata	SPARQLp				SAP HANA Streaming Analytic
SAS Federation Server	dBASE, Greenplum, Hive, IBM DB2, IBM Informix, IBM Netezza _h , Impala, MS Access, MS SQL Server*, MySQL, Oracle DB*, Paradox, PostgreSQL, Progress OpenEdge RDBMS, SAP ASE, SAP HANA*, SAS Federation Server*, SAS Scalable Performance Data Server, Teradata					Btrieve, Salesforce SAP RFC _a
SemaGrow		SPARQLp				
SPLENDID		SPARQLp		0011 10011		
SQL Server (PolyBase)	MS SQL Server*, ODBC, Oracle DB*, Teradata		$MongoDB_d$	CSV, JSON, ORC, Parquet, RCFile		
Squerall	MySQL		Cassandra _w , Couchbase _d , Elasticsearch _s , MongoDB _d	CSV, Parquet		

Aggregate-oriented

Accumulow, Amazon

HBase_w, MongoDB_d, Redis_k, Splunk_s

DataStax_w,
Elasticsearch_s, MS
Azure Cosmos DB_d,

MongoDB_d, Splunk_s

Accumulo_w, Amazon OpenSearch_s,

Cassandra_w,
Couchbase_d, HBase_w,

Infinispan_k, MongoDB_d, Solr_s

DynamoDB $_d$,

Elasticsearch_s

OpenSearch_s, Cassandra_w,

Cassandra_w

Amazon

Amazon

Amazon

Amazon

DynamoDB_d,

OpenSearch_s,

Cassandra_w,

Couchbase a

Accumulo_w,

Cassandra_w

Redisa

Elasticsearch,

 HBase_w , $\mathsf{MongoDB}_d$,

Elasticsearch_s HBase_w, MarkLogic_d, MS Azure Cosmos

 DB_d , $\mathrm{MongoDB}_d$, Splunk_s

SimpleDB_{kw},

Graph-

SPARQLp,

Stardog_{r*}

System

Starburst

Stardog

Teiid

TIBCO Data

Virtualization

Relational

Amazon Redshift, ClickHouse, Druid, Google BigQuery, Greenplum, Hive, IBM DB2, IBM

Server*, MS Azure Synapse Analytics, MySQL, Oracle DB*, Pinot, PostgreSQL, SAP HANA*, SingleStore, Snowflake,

Amazon Athena, Amazon Aurora, Amazon

Redshift, Derby, Exasol, Google BigQuery, Hive, H2, IBM DB2, Impala, MariaDB, MS

Actian Vector, Amazon Athena*, Amazon Redshift, Derby, Exasol, Greenplum, Hive, HSQLDB, H2, IBM DB2, IBM Informix, IBM

HSQLDB, H.Z, IBM DB2, IBM INOTINIX, IB Netezzah, Impala, Ingres, JDBC, MariaDB, MDX_m, MetaMatrix_{*}, MS Access, MS SQL Server_{*}, Mondrian_m, MySQL, Oracle DB_{*}, PostgreSQL, Presto_{*}, SAP ASE, SAP HANA_{*}, SAP IQ, Teiid_{*}, Teradata, Vertica

Amazon Redshift, Drill_{*}, Google BigQuery, Greenplum, Hive, HP Neoview_h, HSQLDB, IBM DB2, IBM Informix, IBM Netezza_h, MS

Access, MS SQL Server_{*}, MySQL, Oracle DB_{*}, PostgreSQL, SAP ASE, SAP Business Warehouse_m, SAP Business Warehouse_m, SAP HANA_{*}, Snowflake, Teradata, Tibco

Amazon Redshift, ClickHouse, Druid, Google BigQuery, Hive, Iceberg, Kudu, MS SQL Server*, MySQL, Oracle DB*, Pinot, PostgreSQL, SingleStore

Firebird, IBM DB2, IBM Informix, Ingres,

MS SQL Server*, MySQL, Oracle DB*, PostgreSQL, Progress OpenEdge RDBMS,

ComputeDB, TibcoDataVirtualization_{*}, Vertica

SQL Server_{*}, MySQL, Oracle DB_{*}, PostgreSQL, SAP ASE, SAP HANA_{*},

Starburst*, Teradata, Vertica

Snowflake, Teradata

Iceberg, JDBC, Kudu, MS SQL

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Virtuoso

SAP ASE

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SPARQLp

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Web Service Paradigms

HTTP/

REST,

OData,

WSDL

HTTP /

OData

SOAP / WSDL

OpenAPI, SOAP /

Other

Amazon Kinesis.

Google Sheetsa,

Salesforce_a

Kafka, Prometheus,

Google Sheetsa, Jiraa,

LDAP, Salesforce_a

Google Sheets_a, InterSystems Caché, JPA/JPQL sources, LDAP, MS Active

ModeShape, OSIsoft PI, Red Hat Directory

Server, Salesforce_a, SAP Gateway OData_a

Eloqua_a, Facebook_a,

Google Adsa, Google

Calendar_a, Google
Calendar_a, Google
Calendar_a, Google
Contacts_a, Google
Sheets_a, HubSpot_a,
IMAP, Marketo_a, MS
Sharepoint, MS

Sharepoint Excel Services_a, NetSuite_a,

RSS, Salesforce_a, SAP RFC_a, Twitter_a

Amazon Kinesis,

Kafka, Prometheus

Google Sheets.

Directory,

Structured Files

Avro, CSV, JSON, ORC,

RCFile, SequenceFile

CSV ISON

CSV, Excel, JSON, XML

CSV, Excel,

JSON, XML

Parquet,

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