

## Building Database Infrastructure for Managing Semantic Data





- Semantics support in the database
- Our model
  - Storage
  - Query
  - Inference
- Use cases: Enhancing database queries with semantics

#### **Semantic Technology**

- Facts are represented as <u>triples</u>
- Triple is the basic building block in the semantic representation of data
- Triples together form a graph, connecting pieces of data
- New triples can be inferred from existing triples
- RDF and OWL are W3C standards for representing such data



# Using a Database for Semantic Applications

- Database queries can be enhanced using semantics
  - Syntactic comparisons can be enhanced with semantic comparisons
- All database characteristics become available for semantic applications
  - Scalability: Database type scale backed by decades of work difficult to match by specialized stores
  - Security, transaction control, availability, backup and recovery, lifecycle management, etc.

# Using a Database for Semantic Applications (contd.)

- SQL (an open standard) interface is familiar to a large community of developers
  - Also SQL constructs can be used for operating on semantic data
  - Existing database users interested in exploring semantics to enhance their applications
- Databases are part of infrastructure in several categories of applications that use semantics
  - Biosurveillance, Social Networks, Telcos, Utilities, Text, Life Sciences, GeoSpatial

# **Our Approach**

- Provide support for managing RDF data in the database for semantic applications
  - Storage Model
  - SQL-based RDF query interface
  - Query interface that enables combining with SQL queries on relational data
  - Inferencing in the database (based on RDFS and user-defined rules)
  - Support for large graphs (billion+ triples)

#### **Technical Overview**



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### Semantic Technology Stack



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### Semantic Technology Storage



#### **Storage: Schema Objects**



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### **Model Storage**

Optional columns for related enterprise data

Application table 1

![](_page_10_Figure_3.jpeg)

### **Internal Semantic Store**

![](_page_11_Figure_1.jpeg)

# **Storage: Highlights**

- Generates <u>hash-based IDs</u> for values (handles <u>collisions</u>)
- Does <u>canonicalization</u> to handle <u>multiple lexical</u> <u>forms</u> of same value point
  - Ex: "0010"^^xsd:decimal and "010"^^xsd:decimal
- Maintains <u>fidelity</u> (user-specified lexical form)
- Allows long literal values (using CLOBs)
- Handles <u>duplicate</u> triples
- No limits on amount of data that can be stored

![](_page_13_Picture_0.jpeg)

### Semantic Technology Query

![](_page_13_Picture_2.jpeg)

# **RDF Querying Problem**

#### Given

- RDF graphs: the data set to be searched
- Graph Pattern: containing a set of variables
- Find
  - Matching Subgraphs
- Return
  - Sets of variable bindings: where each set corresponds to a Matching Subgraph

## **Query Example: Family Data**

Data: :Tom :hasParent :Matt :Matt :hasFather :John :Matt :hasMother :Janice :Jack :hasParent :Suzie :Suzie :hasFather :John :Suzie :hasMother :Janice :John :hasName "JohnD"

(?x :hasFather ?y) (?y :name ?name)', Variable bindings: x = :Matt y = :John name = "John D" Matching subgraph: '(:Tom :hasParent :Matt) (:Matt :hasFather :John)

(:John :name "John D")',

![](_page_15_Figure_4.jpeg)

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## **RDF Query Approaches**

- General Approach
  - Create a new (declarative, SQL-like) query language
  - e.g.: RQL, SeRQL, TRIPLE, N3, Versa, SPARQL, RDQL, RDFQL, SquishQL, RSQL, etc.
- Our SQL-based Approach
  - **Embedding** a graph query in a SQL query
  - SPARQL-like graph pattern embedded in SQL query
- Benefits of SQL-based Approach
  - Leverages all the powerful constructs in SQL (e.g., SELECT / FROM / WHERE, ORDER BY, GROUP BY, aggregates, Join) to process graph query results
  - RDF queries can easily be combined with conventional queries on database tables thereby avoiding staging

# **SDO\_RDF\_MATCH** Table Function

#### Input Parameters

SDO\_RDF\_MATCH (

Query,	← SPARQL-like graph-pattern (with vars)
Models,	← set of RDF/OWL models
Rulebases,	← set of rulebases (e.g., RDFS)
Aliases,	$\leftarrow$ aliases for namespaces
Filter	$\leftarrow$ additional selection criteria

- Return type in definition is <u>AnyDataSet</u>
- Actual return type is determined at compile time based on the graph-pattern argument

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## **Query Example: SQL-based interface**

select x, y, name from TABLE(SDO\_RDF\_MATCH( '(:Tom :hasParent ?x) (?x :hasFather ?y) (?y :name ?name)', SDO\_RDF\_Models('family'), .., .., .));

Returns the name of Tom's grandfather

Х	Y	NAME
Matt	John	"John D"

![](_page_18_Figure_4.jpeg)

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# Combining RDF Queries with Relational Queries

- Find salary and hiredate of Tom's grandfather(s)
- SELECT emp.name, emp.salary, emp.hiredate FROM emp, TABLE(SDO\_RDF\_MATCH( '(:Tom :hasParent ?y) (?y :hasFather ?x) (?x :name ?name)', SDO\_RDF\_Models('family'), ...)) t

WHERE emp.name=t.name;

# **RDF\_MATCH Query Processing**

- Subsititute aliases with namespaces in search pattern
- Convert URIs and literals to internal IDs
- Generate Query
  - Generate self-join query based on matching variables
  - Generate SQL subqueries for rulebases component (if any)
  - Generate the join result by joining internal IDs with UriMap table
  - Use model IDs to restrict IdTriples table
- Compile and Execute the generated query

# Table Columns returned by SDO\_RDF\_MATCH

Each returned row contains one (or more) of the following cols (of type VARCHAR2) for each variable ?x in graph-pattern:

Column Name	Description
X	Value matched with ?x
x\$rdf <u>VTYP</u>	Value TYPe: URI, Literal, or Blank Node
x\$rdf <u>LTYP</u>	Literal TYPe: e.g., xsd:integer
x\$rdf <u>CLOB</u>	CLOB value matched with ?x
x\$rdf <u>LANG</u>	LANGuage tag: e.g., "en-us"

Projection Optimization: Only the columns referred to by the containing query are returned.

# **Optimization: Table Function Rewrite**

- TableRewriteSQL()
  - Takes RDF Query (specified via arguments) as input
  - generates a SQL string
- Substitute the table function call with the generated SQL string
- Reparse and execute the resulting query
- Advantages
  - Avoid execution-time overhead (linear in number of result rows) associated with table function infrastructure
  - Leverage SQL optimizer capabilities to optimize the resulting query (including filter condition pushdown)

![](_page_23_Picture_0.jpeg)

#### Semantic Technology Inference

![](_page_23_Picture_2.jpeg)

### **Inference: Overview**

- Native inferencing in the database for
  - RDF, RDFS
  - User-defined rules
- Rules are stored in rulebases in the database
- RDF graph is entailed (new triples are inferred) by applying rules in rulebase/s to model/s
- Inferencing is based on forward chaining: new triples are inferred and stored ahead of query time
  - Minimizes on-the-fly computation and results in fast query times

## Inferencing

- RDFS Example:
  - A rdf:type B, B rdfs:subClassOf C
  - => A rdf:type C
  - Ex: Matt rdf:type Father, Father rdfs:subClassOf Parent
  - => Matt rdf:type Parent
- User-defined Rules Example:
  - A :hasParent B, B :hasParent C
  - => A :hasGrandParent C
  - Ex: Tom :hasParent Matt, Matt :hasParent John
  - => Tom :hasGrandParent John

# Creating a rulebase and rules index (SQL based)

- Creating a rule base
  - create\_rulebase('family\_rb');
  - insert into mdsys.RDFR\_family\_rb values( 'grandParent\_rule', '(?x :hasParent ?y) (?y :hasParent ?z)', NULL, '(?x :hasGrandParent ?z)', .....);
- Creating a rules index
  - create\_rules\_index('family\_idx',sdo\_rdf\_models('family'),sdo\_ rdf\_rulebases('rdfs','family\_rb)

## **Query Example: Family Data**

![](_page_27_Figure_1.jpeg)

Y	NAME
John	'John D'

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:Tom

:Jack

![](_page_28_Picture_0.jpeg)

## Semantic Technology Enhancing Database Queries with Semantics

![](_page_28_Picture_2.jpeg)

#### Semantics Enhanced Search Medical Information Repositories

- Multiple users might use multiple sets of terms to annotate medical images
  - Difficult to search across multiple medical image repositories

![](_page_29_Figure_3.jpeg)

#### Semantics Enhanced Search Geo-Semantics

- Enhance geo-spatial search with semantics
  - Create an ontology using business categorizations (from the NAICS taxonomy) and use that to enhance yellow pages type search

![](_page_30_Figure_3.jpeg)

![](_page_31_Figure_0.jpeg)

😇 X:-122.03617238233494 / Y:37.771975033167

![](_page_32_Figure_0.jpeg)

E X:-122.23058808973013 / Y:37.909627030373784

![](_page_33_Picture_0.jpeg)

## Biosurveillance

- Biosurveillance application: Track patterns in health data
- Data from 8 emergency rooms in Houston at 10 minute intervals
- Data converted into RDF/OWL and loaded into the database
- 8 months data is 600M+ triples
- Automated analysis of data to track patterns:
  - Spike in flu-like symptoms (RDF/OWL inferencing to identify a flu-like symptom)
  - Spike in children under age 5 coming in

## **Data Integration in the Life Sciences**

"Find all pieces of information associated with a specific target"

- Data integration of multiple datasets
  - Across multiple representation formats, granularity of representation, and access mechanisms
  - Across In-house and public sets (Gene Ontology, UniProt, NCI thesaurus, etc.).
- Standardized and machine-understandable data format with an open data access model is necessary to enable integration
  - Data-warehousing approach represents all data to be integrated in RDF/OWL
  - Semantic metadata layer approach links metadata from various sources and maps data access tool to relevant source
- Ability to combine RDF/OWL queries with relational queries is a big benefit
- Lilly and Pfizer are using semantic technology to solve data integration problems

#### **Use Case: SenseLab Overview**

![](_page_36_Figure_1.jpeg)

#### Login

The SenseLab Project is a long term effort to build integrated, multidisciplinary models of neurons and neural systems, using the olfactory pathway as a model. This is one of a number of projects funded as part of the <u>Human Brain Project</u> whose aim is to develop neuroinformatics tools in support of neuroscience research. The project involves novel informatics approaches to constructing databases and database tools for collecting and analyzing neuroscience information, and providing for efficient interoperability with other neuroscience databases.

- Overview
- Membrane Properties Resource

![](_page_36_Figure_6.jpeg)

#### Brain Database Research

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Part of this work published in the Workshop on Semantic e-Science

#### Courtesy: SenseLab, Yale University

#### **Relational to Ontological Mapping**

![](_page_37_Figure_1.jpeg)

#### Courtesy: SenseLab, Yale University

Detail

#### **Use Case: Integrated Bioinformatics Data**

![](_page_38_Figure_1.jpeg)

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Part of this work published in Journal of Web Semantics

Source: Siderean Software

## **Use Case: Knowledge Mining Solutions**

![](_page_39_Figure_1.jpeg)

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#### **Semantic Operators in SQL**

- Two new first class SQL operators to semantically query relational data by consulting an ontology
  - SEM\_RELATED (<col>,<pred>, <ontologyTerm>, <ontologyName> [,<invoc\_id>])
  - SEM\_DISTANCE (<invoc\_id>) ← <u>Ancillary Oper.</u>
  - Can be used in any SQL construct (ORDER BY, GROUP BY, SUM, etc.)
- Semantic indextype
  - An index of type semantic indextype introduced for efficient execution of queries using the semantic operators

#### **Ontology-assisted Query**

![](_page_42_Figure_1.jpeg)

#### **Ontology-assisted Query**

![](_page_43_Figure_1.jpeg)

#### **Ontology-assisted Query**

![](_page_44_Figure_1.jpeg)

### Summary

- Semantic Technology support in the database
  - Store RDF/OWL data and ontologies
  - Infer new RDF/OWL triples via native inferencing
  - Query RDF/OWL data and ontologies
  - Ontology-Assisted Query of relational data

• More information at:

http://www.oracle.com/technology/tech/semantic\_technologies/index.html