



Data Management and Modeling in Support to Scientific applications


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<http://dexl.lncc.br>

Outline

- Introduction
- Linked Science and Scientific Hypotheses
- Athletes Trajectory DW
- Astronomy integration as linked science
- QEF – Query Engine for data Intensive Applications
- Final remarks

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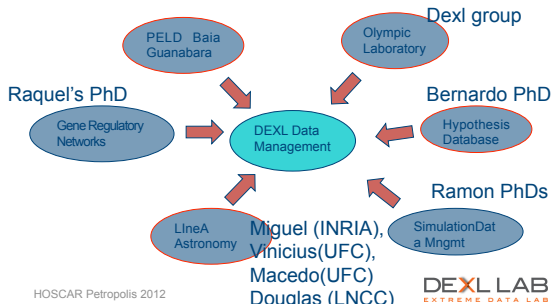
The DEXL Lab Mission

- To support in-silico science with data management techniques;
 - To develop interdisciplinary research with contributions on data modelling, design and management;
 - To develop tools and systems in support to in-silico science;


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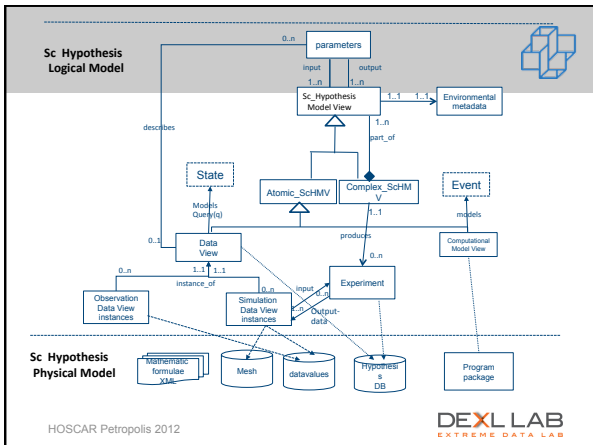
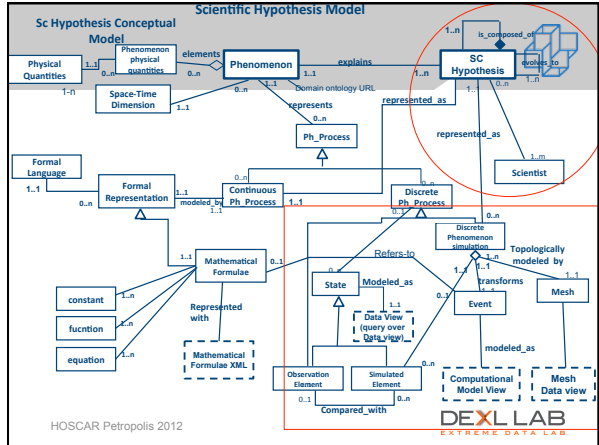
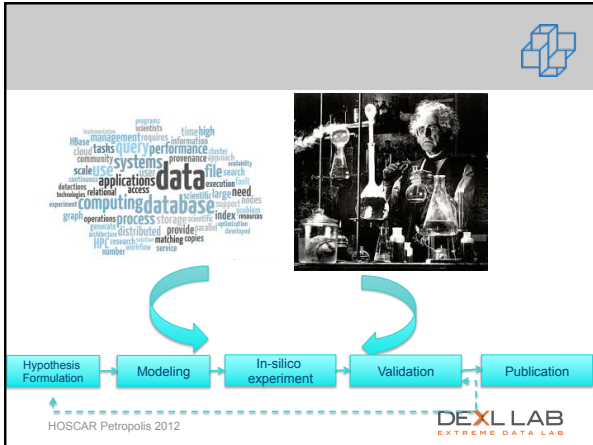


Current projects in the Lab



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Points of Investigation


- Modelling and Management of Hypothesis (Hypothesis DB)
- Processing of Data Intensive Scientific Workflows
- Storage and Management of Meshes

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Linked Science (or Linked Open Science)

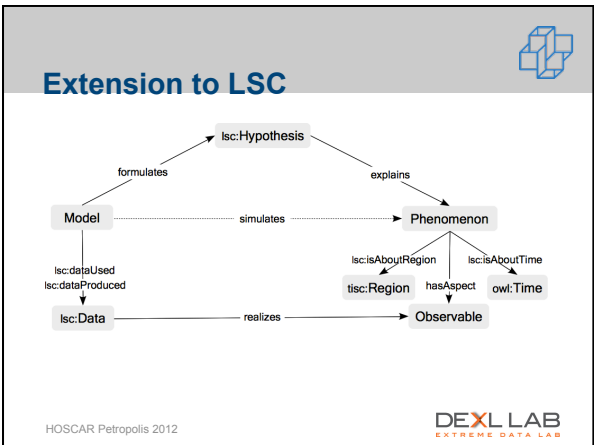
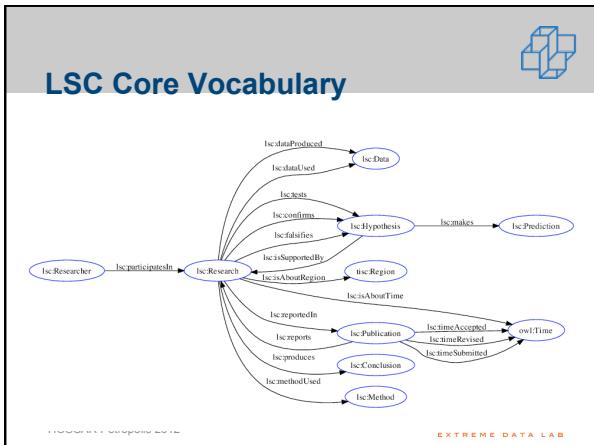
- Is an initiative to interconnect all scientific assets;
- It is a combination of:
 - Linked data, semantic web
 - Open source;
 - Scientific workflows and provenance (OPM);
 - Scientific models;
 - Cloud computing;
 - ...

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Linked Science Core Vocabulary (LSC)

- Defines a vocabulary (LSC) with “basic” terms for science;
 - More specific terminology shall be added by individual communities (minimal ontological commitment)

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Published Research as Linked Data (1)³

rdfs:Class	rdf:Resource	rdf:Literal
lsc:Researcher	authors1 <small>rdf:value</small>	"P.J. Blanco, M.R. Pivello, S.A. Urquiza, and
lsc:Research	research1 <small>dc:description</small>	"Simulation of hemodynamic conditions artery."
lsc:Publication	pub1 <small>dc:title</small>	"On the potentialities of 3D-1D coupled mi-
lsc>Data	dataset1 <small>dc:description</small>	"Flow rate of 5.0 l/min as an inflow bounda-
lsc>Data	dataset2 <small>dc:description</small>	"1D mechanical and geometric data from A
lsc>Data	dataset3 <small>dc:description</small>	"MRI images processed for reconstructing 1
Phenomenon	p17 <small>dc:description</small>	"Blood flow in the carotid artery."
lsc:Region	region1 <small>dc:description</small>	"The carotid artery, a part of the human CV
owl:IntervalEvent	beat1 <small>dc:description</small>	"A heart beat with period $T = 0.8$ s."
Observable	ob1 <small>dc:description</small>	"Blood flow rate."
Observable	ob2 <small>dc:description</small>	"Blood pressure."
lsc:Hypothesis	h17 <small>rdfs:label</small>	"h17(h13, h15, h16)"
Model	m17 <small>dc:description</small>	"3D-1D coupled model with lumped windke-

³Blanco et al.'s published research as an LSC instantiation.

Semantic engineering of hypotheses

rdfs:Class	rdf:Resource	rdf:Literal
lsc:Data	dataset4 <small>dc:description</small>	"Plots of hemodynamic observables in the left f
lsc:Data	dataset5 <small>dc:description</small>	"Plots of hemodynamic observables in the caro
lsc:Data	dataset6 <small>dc:description</small>	"Scientific visualization of hemodynamic obser
lsc:Data	dataset7 <small>dc:description</small>	"Scientific visualization of hemodynamic obser
lsc:Prediction	predict1 <small>rdf:value</small>	"Sensitivity of local blood flow in the carotid arte
lsc:Prediction	predict2 <small>rdf:value</small>	"Sensitivity of the cardiac pulse to the pre
lsc:Conclusion	conclusion1 <small>rdf:value</small>	"3D-1D coupled models allow to perform qua-

⁴Blanco et al.'s published research as an LSC instantiation.


Q1.] Find in Blanco et al.'s microtheory a hypothesis (if any) explaining phenomena of blood flow in microvascular vessels and show which model formulates it.

```

PREFIX rdfs: <http://www.w3.org/2000/01/rdf-schema#>
PREFIX dc: <http://purl.org/dc/elements/1.1/>
PREFIX lsc: <http://linkedsience.org/lsc/ns#>
SELECT ?hypothesis_name ?model_name
WHERE {
?h rdfs:label ?hypothesis_name .
?m rdfs:label ?model_name .
?h a lsc:Hypothesis .
?p a lsc:Phenomenon .
?m a lsc:Model .
?h lsc:explains ?p .
?m lsc:formulates ?h .
?p dc:description ?d .
FILTER regex(?d, "blood flow", "i") . FILTER regex(?d, "microvascular",
"i")
}
    
```



Final remarks – Hypothesis DB

- An opportunity to publish the artifacts produced during the in-silico scientific life-cycle;
- Project is still in its infancy. We intend to develop an application to support hypothesis management
- Bernardo Gonçalves PhD work



PROCESSING SCIENTIFIC VISUALIZATION DATA


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Introduction

- A query processing-based technique to compute the pre-processing stage of scientific visualization of blood flow in an artery.
- Use the QEF engine to model and evaluate the workflow

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QEF – Query Engine for Data Intensive Applications




Adaptive and Extensible Query Engine

- Extensible to data types
- Extensible to application algebra
- Extensible to execution model
- Schedule operations in grid nodes
- Adaptive execution model

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Objective




- Offer a query processing framework that can be extended to adapt to data intensive application needs;
- Offer transparency in using resources to answer queries;
 - Query optimization transparently introduced
 - Standardize remote communication using web services even when dealing with large amount of unstructured data
 - Run-time performance monitoring and decision

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The problem


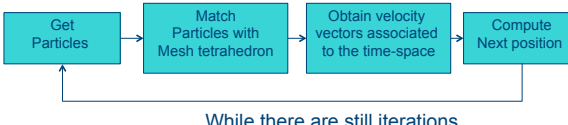


- Data sets
 - Mesh – tetrahedrons in 3D
 - Dataset of velocity and time
 - Virtual particles in an initial position
- Trajectories without collision
- A number of iterations through time-space

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A scientific workflow





```


graph LR
  A[Get Particles] --> B[Match Particles with Mesh tetrahedron]
  B --> C[Obtain velocity vectors associated to the time-space]
  C --> D[Compute Next position]
  D --> A
  
```

While there are still iterations

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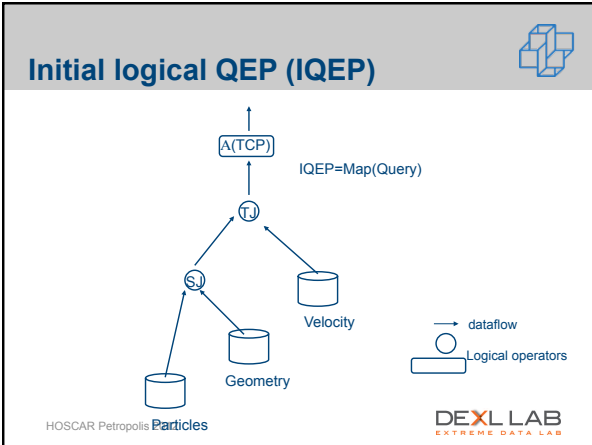
Modelling in QEF



- Each data set is a relation
 - Geometry (id, <3DPoint>)
 - Velocity (id, tetrahedronId,time,<velocity>)
 - Particle (id, iteratio, 3DPoint)
- Each activity of the workflow is an operator
 - Spatial-temporal join
 - Map (trajectory computing program)
- Add control operators
 - Orbit – to control iteration
 - Split/merge
 - Fold / unfold

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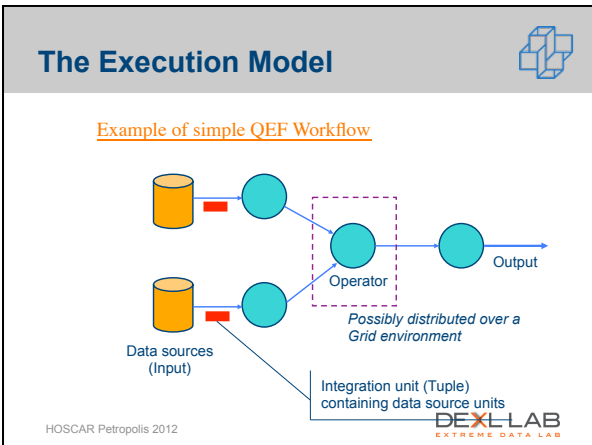
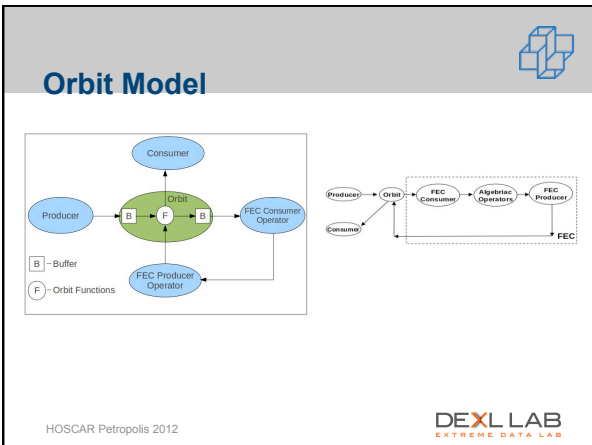


Control Operators

- Add data-flow and transformation operators
- Isolate application oriented operators from execution model data-flow concerns
- parallel grid based execution model:
 - **Split/Merge** - controls the routing of tuples to parallel nodes and the corresponding unification of multiple routes to a single flow
 - **Send/Receive** - marshalling/ unmarshalling of tuples and interface with communication mechanisms
 - **B2I/I2B** - blocks and unblocks tuples
 - **Orbit** - implements loop in a data-flow
 - **Fold/Unfold** - logical serialization of complex structures (e.g. PointList to Points)

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Distribution and Parallelization

Operator distribution

A Query Optimizer selects a set of operators in the QEP to execute over a distributed environment.

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General Parallel Execution Model

Remote QEP

In order to parallelize an execution, the initial QEP is modified and sent to remote nodes to handle the distributed execution.

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Modifying IQEP to adapt to execution model

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Grid node allocation algorithm (G2N)

Grid Greedy Node scheduling algorithm (G2N)

Introduction

- Offers maximum usage of scheduled resources during query evaluation.

Principles

- Basic idea : "an optimal parallel allocation strategy for an independent query operator ... is the one in which the computed elapsed-time of its execution is as close as possible to the maximum sequential time in each node evaluating an instance of the operator".


Architecture

Implem.

Conclusion

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Implementation

- Core development in Java 1.6.
- Globus toolkit 4.
- Derby DBMS (catalog).
- Tomcat, AJAX and Google Web Toolkit for user interface.
- Runs on Windows, Unix and Linux.
- source code, demo, user guide available at:
<http://dexl.incc.br>



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Final remarks



- QEF is a complete engine for processing data intensive applications;
- Is extensible for:
 - data types, data sources
 - User operations
 - Data management operators
- Current applications
 - Open linked Data Processing (PELD integration)
 - SkyMap workflow
 - Data Replication

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PROCESSING ASTRONOMY DATA – LINEA LABORATORY


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Context

- Analytical Workflows process a large part of Catalog data
 - Catalogs are supported by few indexes, thus most queries scan tens-to-hundreds of millions of tuples
- Parallelization comes as a rescue to reduce analyses elapsed-time, but
 - Compromise between:
 - Data partitioning and degree of parallelization;
 - Current solutions consider:
 - Centralized files to be distributed through nodes (MapReduce)
 - Distributed databases (Qserv) to serve Workflow engines
 - Centralized databases to serve Workflow Engine (Orchestration LineA)
 - Partitioned database to serve distributed queries (HadoopDB)

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Processing Scientific workflows on Database data

Users
- Ad-hoc queries
- downloads

Scientific workflows
-- Analyses

DB

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Traditional WF – Database decoupled architecture

Workflow engine

act1 → Act 2 → act3

Database

DB₁ DB₂ DB₃

Data is consolidated as input
The workflow

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Orchestration Layer at LineA Portal

Catalog DB

Spatial partitioning

Etapa 1: Data Retriever

Etapa 2: Data Organizer, Skymap

Etapa 3: Skyadd, Hystogram.PNG

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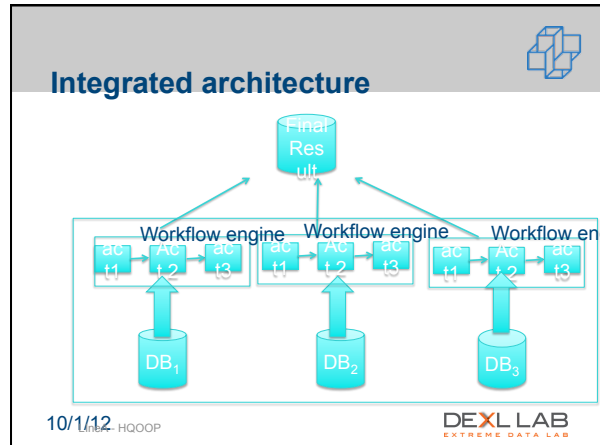
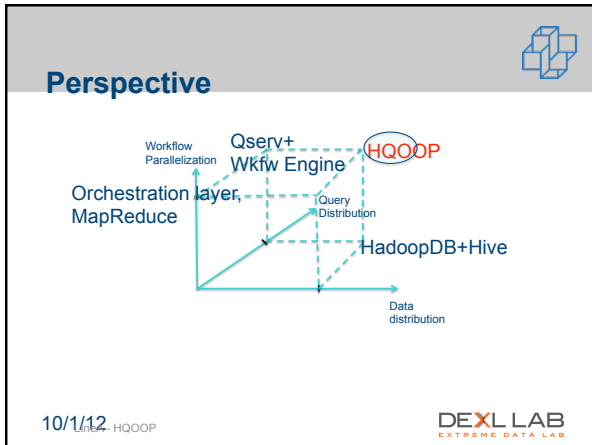
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HQOOP -Parallelizing Pushed-down Scientific Workflows

- Partition of data across cluster nodes
 - Partitioning criteria
 - Spatial (currently used and necessary for some applications)
 - Random (possible in SkyMap)
 - Based on query workload (Miguel Liroz-Gestau's Work)
- Process the workflow close to data location
 - Reduce data transfer
- Use Apache/Hadoop Implementation to manage parallel execution
 - Widely used in Big Data processing;
 - Implements Map-Reduce programming paradigm;
 - Fault Tolerance of failed Map processes;
- Use QEF as workflow Engine
 - Implements Mapper interface
 - Run workflows in Hadoop seamlessly;

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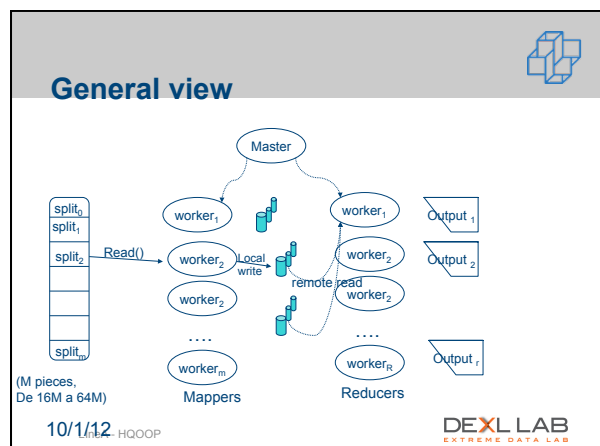



Map and Reduce

- Interface:
 - Map(key1,value1) -> list (key2,value2)
 - Reduce(key2,list<value2>)-> list(value2)
- Map and reduce are functions written by the user according to application;
- Map: takes a <key,value> pair; key and value are of any datatype, and produces a list of intermediate key,value pairs;
- The framework groups the output of Map by the value of key2, producing a list of associated value2;
- The reduce function takes the pair <key2, list<value2>> and produces its output;

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Process

- Initially, the MR framework splits the input into M partitions of fixed size;
- It initializes the Master node;
- The master node creates M+R workers and assign them maps and reduce functions, accordingly;
- Each Map reads its partition of the input and generates in memory its output;
- Periodically a process reads the buffer and groups the output values by key. It then writes the output to one of the R partitions, informing the master about its complete status and the partition addresses;
- Finally, the reduce reads each of its partitions and iterates over the keys, producing the results that are written to the output file.



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Fault Tolerance

- Master keeps record of worker status (idle, in-progress, completed)
- It pings workers periodically
- If worker ping times-out, it is considered as dead and all completed work is re-scheduled to another node;



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Task Granularity

- M and R much larger than the number of workers machines;
- Google defines M in terms of the size of the input partition (between 16M and 64M), and R a small multiple of the number of workers machine;
- Usual numbers:
 - Worker machines: 2000
 - Mappers: 200.000
 - Reducers: 5.000
- Reducers are in small number as they produce each an output file


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Partitioning of Intermediate results

- Intermediate results produced by map are re-partitioned into "R" fragments;
- Default partitioning function is:
 - Hash(key) mod R;
 - More semantically meaningful partition desired, if possible

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Summing-up

- MapReduce proposes a simple interfaces with robust framework to support parallelization of applications with huge number of data entities;
- Processes an iteration over keys;
- The framework has been implemented by Goggle, apache(Hadoop)
- The main exported elements are pairs of Key, value
- Deals with fault tolerance of workers but not that of master
- No application based optimization is possible due to lack of function implementation semantics;
- File based

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Hadoop(Google MapReduce) - Weaknesses

- No expressive query language
 - Expressive query language allow developers to formulate high-level tasks
- No optimization based on function semantics
- No semantic-based partitioning strategy
 - Semantic-based partitioning foster parallelization and data access according to application characteristics

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HadoopDB - a step in between [Abouzeid09]

- Offers parallelism and fault tolerance as Hadoop, with SQL queries pushed-down to postgresQL DBMS;
- Pushed-down queries are implemented as Map-reduce functions;
- Data are partitioned through nodes.
 - Partitioning information stored in the catalog
 - Distributed through the N nodes

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HadoopDB architecture

```

    graph TD
        SQL[SQL query] --> SMS[SMS Planner]
        SMS --> MR[MapReduce Framework]
        SMS --> Catalog[Catalog]
        MR --> Node1[Node 1]
        MR --> Node2[Node 2]
        MR --> NodeN[Node n]
        subgraph Node1
            TT1[Task Tracker]
            DB1[(Database)]
            DN1[DataNode]
        end
        subgraph Node2
            TT2[Task Tracker]
            DB2[(Database)]
            DN2[DataNode]
        end
        subgraph NodeN
            TTn[Task Tracker]
            DBn[(Database)]
            DNn[DataNode]
        end
    
```

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Example

```

Select year(SalesDate), sum(revenue)
From Sales
Group by year(SalesDate)
    
```

a) Table partitioned by year(SalesDate) b) no partitioning by year(SalesDate)

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Summing-up

- HadoopDB extends Hadoop with expressive query language, supported by DBMSs
- Keeps Hadoop MapReduce framework
- Queries are mapped to MapReduce tasks
- For scientific applications is a question to be answered whether or not scientists will enjoy writing SQL queries
- Algebraic like languages may seem more natural (eg. Pig Latin)

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Experiment Set-up

- Cluster SGI
 - Configurations: 1, 4 and 95 nodes;
 - Each node:
 - 2 proc. Intel Zeon – X5650, 6 cores, 2.67 GHz
 - 24 GB RAM
 - 500 GB HD
- Data
 - Catalog DC6B

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Preliminary Results



- Preliminary results are encouraging:
 - Baseline Orchestration layer (1 node DB + 46 proc. nodes) – approx. 46 min
 - 1 node DB + 94 nodes HQOOP – approx. 12.3 min
 - 95 nodes; 1 Master + (94 DB part. + HQOOP) – approx. 2.10 min
 - 95 nodes (1 Master + 94 DB part. Hadoop +Python) – approx. 2.4 min

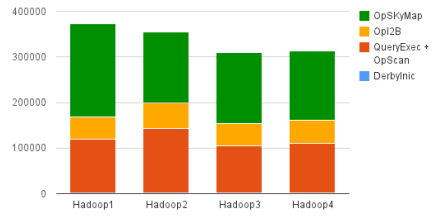
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Execution with 4 nodes



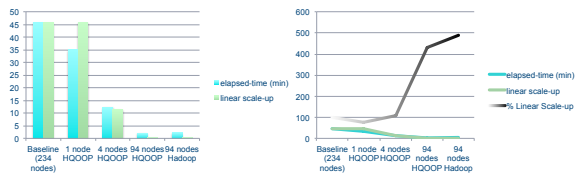
Elapsed-time total: 11.27 min



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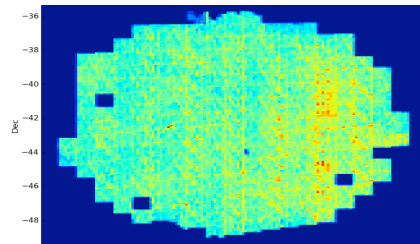
Overall performance



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Resulting Image



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TLPP Processing

- Transfer Data
 - From telescope to data storage sites
- Load
 - Data Ingestion Procedures
 - Data Management
 - Data Replication
 - Data Model
 - Query Processing
- ★ Process
 - Scientific Workflows
 - Data locality
 - Provenance
 - Store of workflow results in DB
- Publish
 - Linked-Data


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Final Remarks


- HQOOP shows interesting initial results
 - Evaluate with other pipelines
 - Enhancing data partitioning and load procedure
 - Allow workflow to be passed as parameter
 - Deal with fault tolerance
 - Evaluate other possible configurations

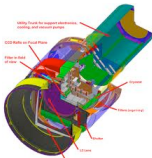
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Dark Energy Survey and LSST – Large Synoptic Survey Telescope



Cerro Pachón – Future site of the LSST






- 800 images p/ night during 10 anos !!
- Map 3D of the Universe
- 30 TeraBytes per night
- 30 PetaBytes in 10 years


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LSST – simulated image from sky




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Context




- **Dark Energy Survey**
 THE DARK ENERGY SURVEY
 - Astronomic project to explain:
 - Acceleration of the universe
 - Nature of dark energy
 - Data production
 - DECam takes images of 1GB (400/night)
 - Images are analyzed; galaxies and stars are identified and catalogued
 - Catalogs are stored in relational databases

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Context




- **Database features**
 - Single relation (the catalog)
 - Initially: 1 billion tuples x 1000 attributes (300GB)
 - The size of db is increasing each day
- **Many astronomical surveys gathering data from the "same" sky:**
 - Sloan Digital Sky Survey (SDSS III)
 - DES
 - LSST
 - Gemini,

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Problem




- Different catalogs holding information from the "same" sky object;
- Integrating these catalogs provide a more comprehensive view of the sky
- How to build a linked data view of huge (Billion of objects) databases?

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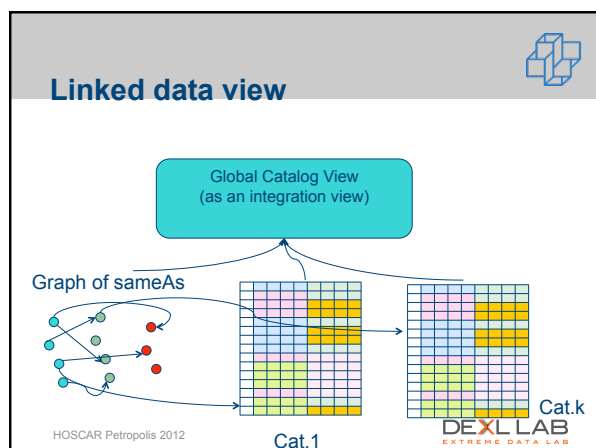
Initial Proposal



- Build an integrated view using linked data views;
- Materialize "sameAs" relationships among objects, according to matching algorithms;
 - Combine:
 - graph representation
 - sameAs
 - Linked data view of catalogs
 - Relational database

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Final remarks

- Managing scientific Big data is a hot topic
- Linked Data may contribute on publishing scientific results in the context of linked science
- Many challenges with respect to providing linked data in the context of Big Data
- Lots of fun ahead !!!

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Obrigado !

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