

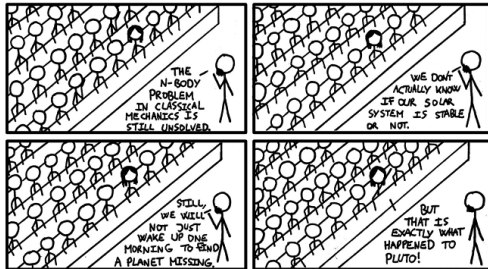
Analyzing Massive Astrophysical Datasets

Can Pig/Hadoop or a relational DBMS help?

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Outline

Introduction

Astrophysics Simulation

Hadoop

Showdown

N-body Problem

N-body problem is the problem of predicting a future state of a gravitational system from initial conditions.

N-Body Problem does not have an analytic solution for $N > 3$:

- ▶ Computer science: NP-complete
- ▶ Math: Galois Theorem
- ▶ Astrophysics: Numerical Solution

N-body Problem Cont.

N-body simulation

parameters:

- ▶ N : number of bodies
- ▶ Δt : simulation timestep
- ▶ T_{final} : simulation length
- ▶ $T_{checkpoint}$: checkpoint frequency

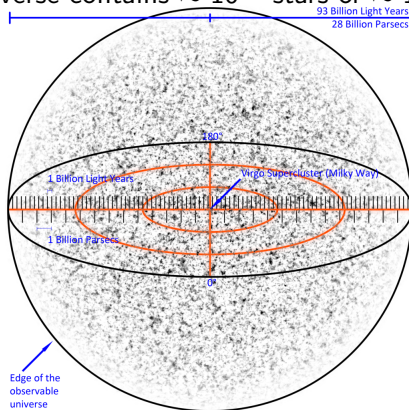
N-body simulation workflow:

```
1:  $t \leftarrow 0$ 
2: while  $t < T_{final}$  do
3:   for particle  $p$  in  $N$  do
4:     Compute all forces on  $p$ 
5:     Compute position at  $t + \Delta t$ 
6:   end for
7:    $t = t + \Delta t$ 
8:   if  $(t - t_{last}) > T_{checkpoint}$  then
9:     Create checkpoint for  $t$ 
10:     $t_{last} \leftarrow t$ 
11:   end if
12: end while
```

Astrophysics Simulation Scale

Name	# of Particles	Snapshot Size
dbtest128g	$4.2 * 10^6$	169 MB
cosmo50	$3.36 * 10^7$	1.4GB
como25	$9.168 * 10^8$	3.6GB

Visible universe contains $\approx 10^{22}$ stars or $\approx 10^{80}$ atoms.



Simulation Analysis

Strategy:

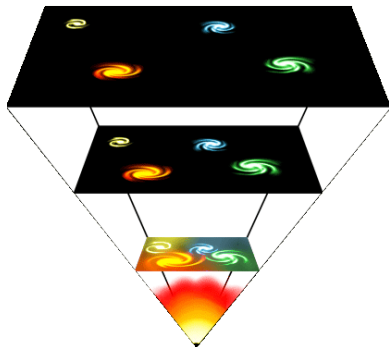
Analyze checkpoint files.

Tools:

- ▶ IDL
- ▶ Python/Perl
- ▶ C++ libraries

Trend:

- ▶ Constraints on CPU IO
RAM
- ▶ Distributed memory
model?



Could we use a Parallel DBMS or MapReduce?

If the problem can be partitioned correctly MapReduce takes care of the parallelism for you.

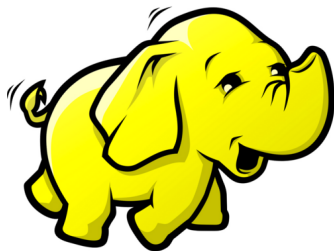
VS

Parallel DBMS tackle similar problems in data-mining. These systems could be adapted to the Scientific workload.

Hadoop

Hadoop:

- ▶ An open source MapReduce implementation
- ▶ API for Java, C++, python ...
- ▶ Pig Programing language
- ▶ HDFS distributed filesystem



Queries:

Q1: Return all particles whose property X is above a given threshold at step S_1 .

$$\pi(iOrder)\sigma_{(x>T)}(S_1)$$

Q2: Return all particles of type T within distance R of point P .

$$\pi(iOrder)\sigma_{p.x^2+p.y^2+p.z^2>R^2}(S_1)$$

Q3: Return all particles of type T within distance R of point P whose property X is above a threshold computed at timestep S_1 .

$$\pi(iOrder)\sigma_{(p.X>T)\&(p.x^2+p.y^2+p.z^2>R^2)}(S_1)$$

Q4: Return gas particles destroyed between step S_1 and S_2 .

$$\pi(iOrder)\sigma_{p_1.iOrder\notin S_2}(S_1 \times S_2)$$

Q5: Return all particles whose property X changes from S_1 to S_2 .

$$\pi(iOrder)\sigma_{(p_1.iOrder=p_2.iOrder)\&(p_1.X\neq p_2.X)}(S_1 \times S_2)$$

Single Node Performance:

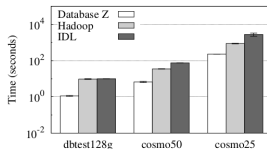
Specs:

- ▶ IDL: 16CPU, 128 GB RAM
- ▶ DBMS/Hadoop: 8CPU, 16GB RAM.

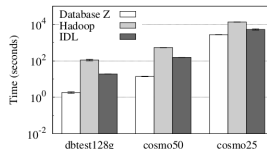
Notes:

Hadoop and IDL had to read in the snapshot file every time.

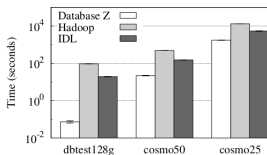
DBMS had indexed data available to it.



Query: Q1,Q2,Q3

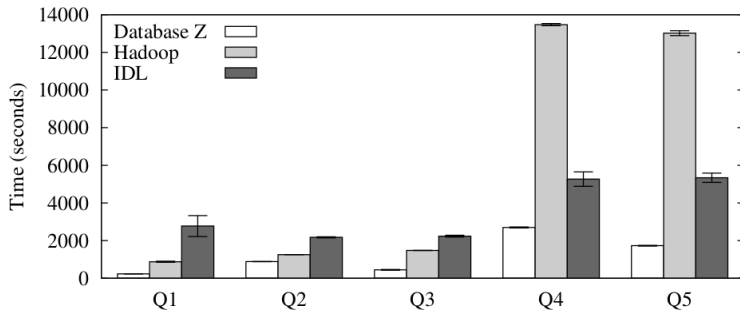


Query: Q4



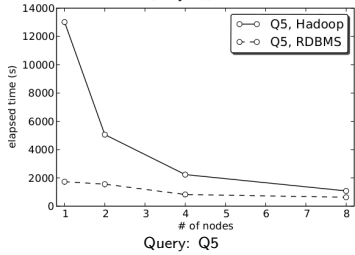
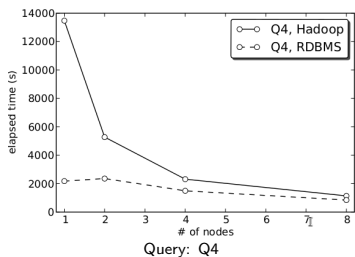
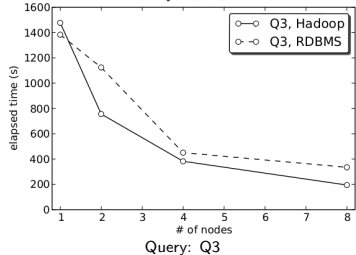
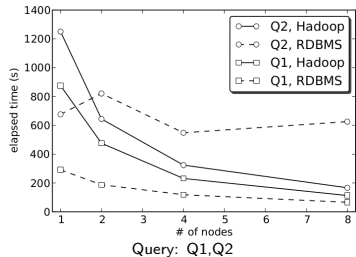
Query: Q5

Single Node Large Dataset:



Query Performance on 32GB dataset.

Multi-Node Performance:



Questions?