



Analytics for Massive Data Sets



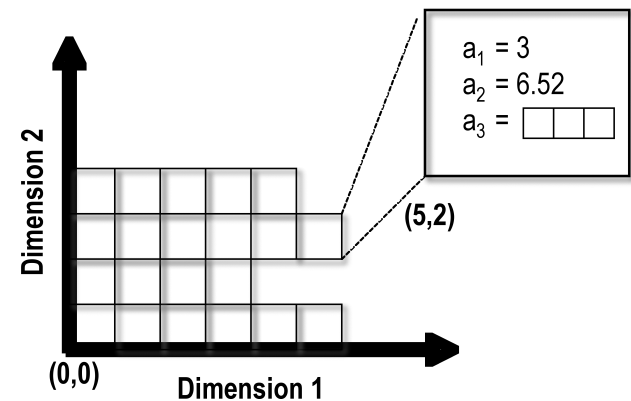
Big Data and Big Analytics

- Introducing SciDB
 - Open source, massively parallel DBMS and analytic platform
 - Array data model (rather than SQL, Unstructured, XML, or triple-store)
 - Extensible *micro-kernel* architecture
 - Reliable storage platform
- Analytic Example
 - Problem Description: Big Data, Big Analytics
 - Compare and Contrast SciDB with SQL and Map/Reduce
- Data, Queries and Scalability Graphs
 - What does SciDB look like from a user's perspective
 - Cloud deployment (EC2) up to 64 nodes

Introducing SciDB

- Open Source
 - Released under GPL-3
 - <http://www.scidb.org/>
 - Sponsored by Paradigm4 Inc.
 - Contributors from the US, Russia, India, and Europe

- **Array Data Model**





Why Arrays?

- Formal Mathematical Foundations
 - Closed algebra of Array → Array operations
 - Formal proofs of correctness for re-writes
 - Mathematical Model has broad applications
- Computer Science Pedigree
 - APL – *A Programming Language*
 - Lots of work on algorithms to compute Array → Array ops
 - Lots of work on efficient storage and data manipulation
- Ongoing Area of Research and Development
 - BLAS, LAPack, ScaLAPack – open-source tools
 - NAG – Numerical Algorithmics Group – proprietary platforms
 - R, SaS, etc - each have an efficient Linear Algebra Library inside

Introducing SciDB data definition

```
CREATE ARRAY Simple_Array <
  v1 : double,
  v2 : int64,
  v3 : string >
[ I = 0:*, 5, 0, J = 0:9, 5, 0 ];
```

Attributes
v1, v2, v3

Dimensions
I, J

Dimension size
* indicates unbounded

**Chunk
size**

**Chunk
overlap**

2 Programming Interfaces

**Array
Functional
Language
AFL**

```
aggregate(  
    filter ( Simple_Array,  
            v3 = 'Odd' ),  
    I,  
    avg ( Simple_Array.v1 )  
);
```

**Array
Query
Language
AQL**

```
SELECT avg ( S.v1 )  
    FROM Simple_Array S  
WHERE S.v3 = 'Odd'  
GROUP BY S.I;
```



SciDB: Analytic Features

- Versioned (no-overwrite) Storage System
 - UPDATE operations are append only
 - The state of the system can be reconstructed at a specified time
- Provenance
 - Log of all queries to reconstruct how results were derived
- Uncertainty and Statistical Reasoning
 - Types with error bars
 - Functions to perform statistical tests and analytics

Goal is to meet the requirements for scientific data management identified in Stonebraker. M, et al
“Requirements for Science Data Bases and SciDB”, CIDR 2009



SciDB vs. RDBMSs

on storage efficiency & complex computations

Relational Database

I	J	value
0	0	32.5
1	0	90.9
2	0	42.1
3	0	96.7
0	1	46.3
1	1	35.4
2	1	35.7
3	1	41.3
0	2	81.7
1	2	35.9
2	2	35.3
3	2	89.9
0	3	53.6
1	3	86.3
2	3	45.9
3	3	27.6

48 cells

SciDB

32.5	46.3	81.7	53.6
90.9	35.4	35.9	86.3
42.1	35.7	35.3	45.9
96.7	41.3	89.9	27.6

16 cells

- Speeds up data access in a distributed database
- Dramatic storage efficiencies as # of dimensions & attributes grows
- Facilitates drill-down & clustering by like groups
- Math functions like linear algebra run directly on native storage format



SQL vs. SciDB: Data Definition

```
CREATE TABLE USNOB (  
    Locn point                not null,  
    -- RA double, DECL double  
    ...  
    B_mag double precision    not null,  
    ...  
    R_mag double precision    not null  
    ...  
);
```

SQL

```
CREATE ARRAY USNOB <  
    B_Mag : double,  
    ...  
    R_Mag : double  
>  
[ RA(double)=*,72000,720, DECL(double)=*,36000,360 ];
```

SciDB



SQL vs. SciDB: Data Manipulation

```
SELECT U1.Locn, COUNT(*)
  FROM USNOB AS U1, USNOB AS U2
 WHERE box(U1.Locn, U1.Locn) &&
        box(point(U1.Locn[0] - 0.001,
                  U1.Locn[1] + 0.001),
            point(U1.Locn[0] + 0.001,
                  U1.Locn[1] - 0.001))
GROUP BY U1.Locn;
```

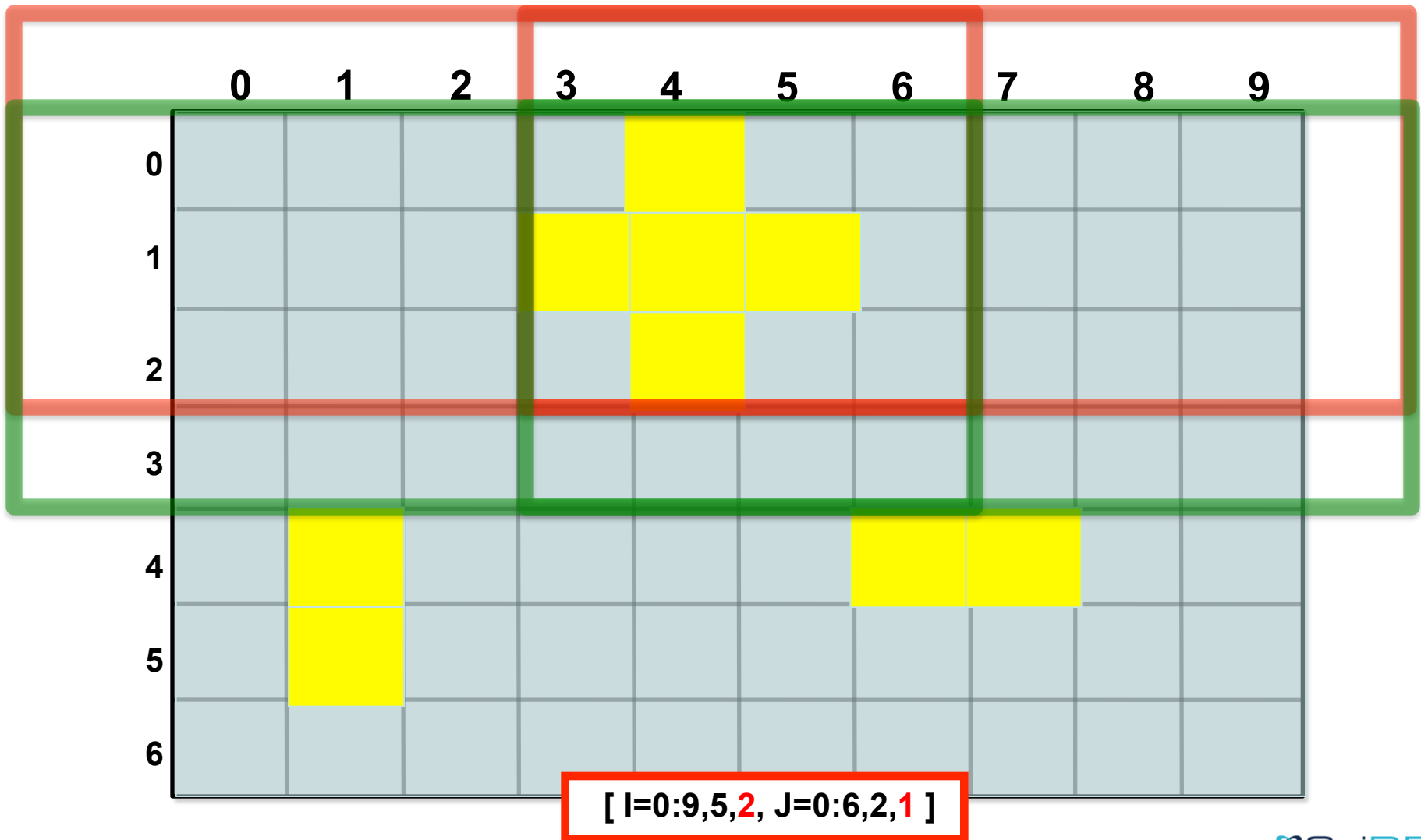
SQL

```
window (
    USNOB,
    0.001, 0.001,
    count(*)
);
```

SciDB

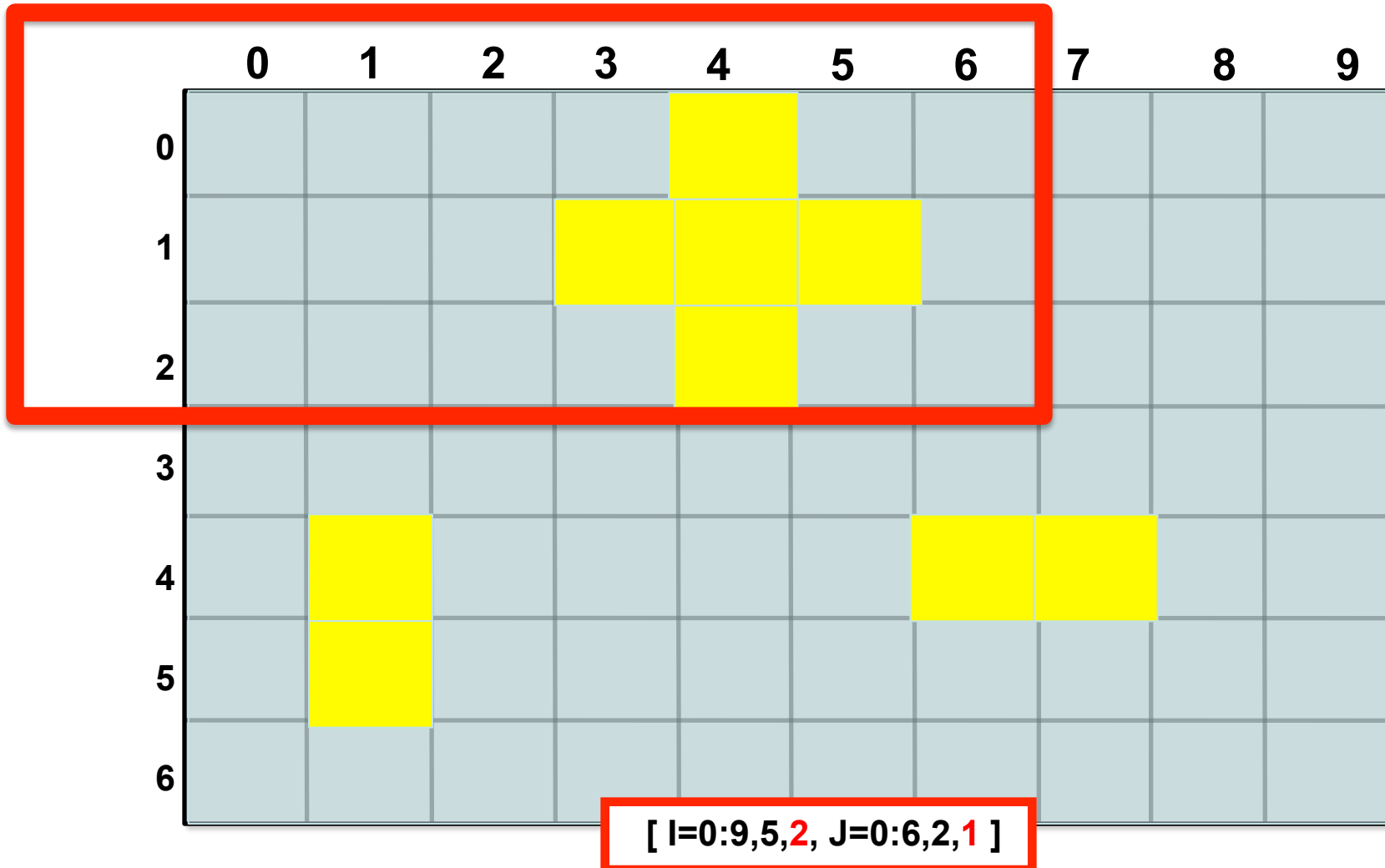


SciDB Storage: Arrays and Chunks



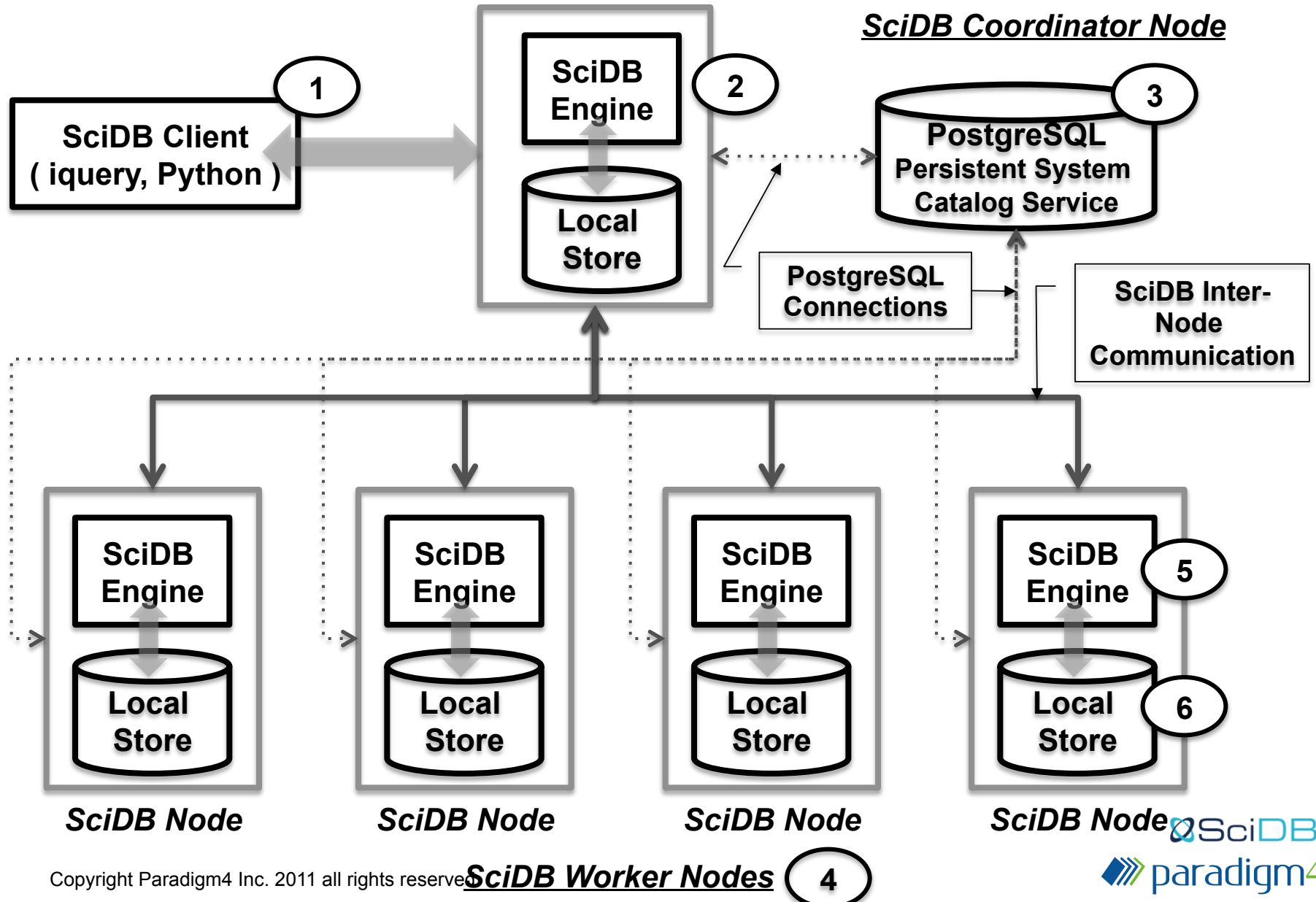


SciDB Storage: Arrays and Chunks



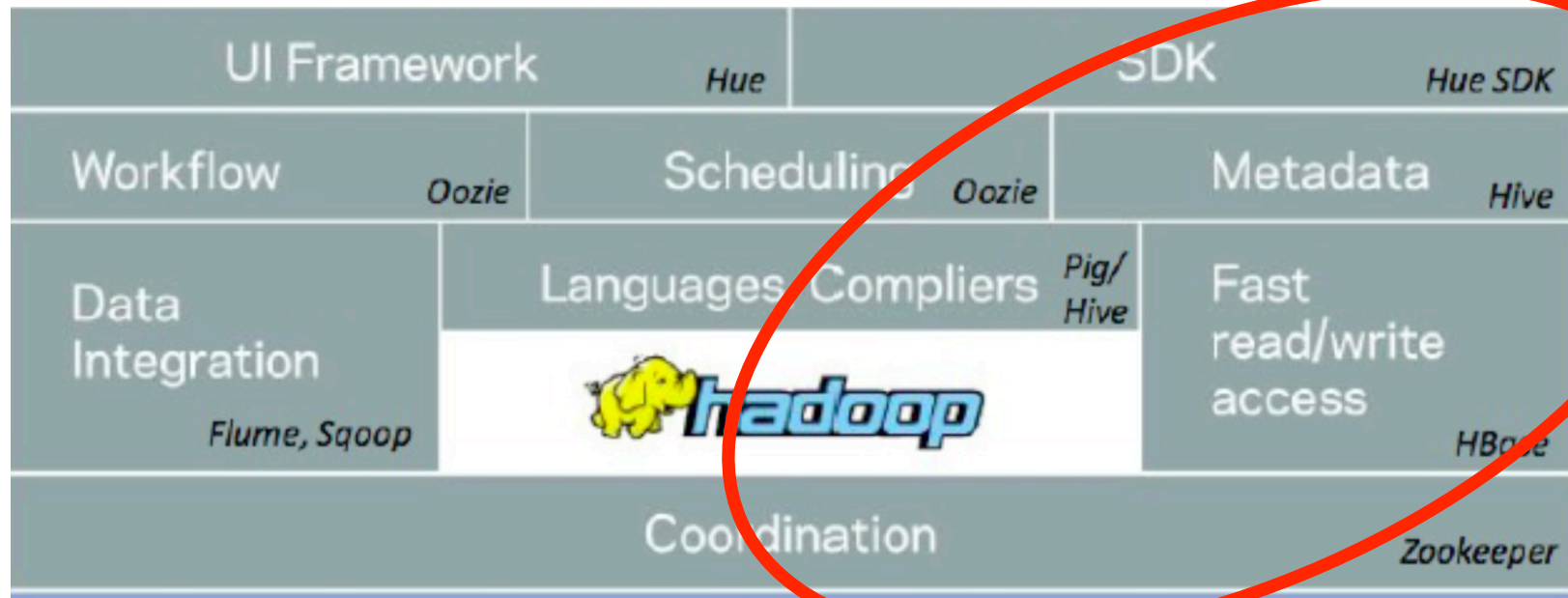


Shared Nothing MPP Architecture



SciDB is not Map/Reduce

- Build on the Best of Map/Reduce's many good ideas
 - Job scheduling over large numbers of nodes
 - Use k-copy replication of chunks to implement reliable storage
- SciDB's Specialized Storage Model Conveys Advantages
 - Block distribution and inter-node operations
 - Local parallelism for individual block to block operations





SciDB is not Map/Reduce

- Usage Model supports *ad hoc* queries
 - Use case lessons: Analysts like ‘R’ and dislike programmers
 - Interactive, conversational, explorative model
- Linear Algebra: hard to implement as a Map and a Reduce
 - Linear Algebraic operations are the foundations of statistical analysis
 - Linear Algebra operators are ‘not embarrassingly parallelizable’
 - multiply (transpose (A), A) is an $O(N^3)$ algorithm
 - inverse (A) is best implemented as an iterative algorithm
 - Cumulative sum, multi-dimensional WINDOW
- Deeply integrated data management & analytics has many advantages
 - Data is updatable

Fault-tolerance

by @jrecursive



Brief Performance Study

- Three Queries in our AFL Query Language

```
multiply ( transpose ( Simple_Array ), Simple_Array );
```

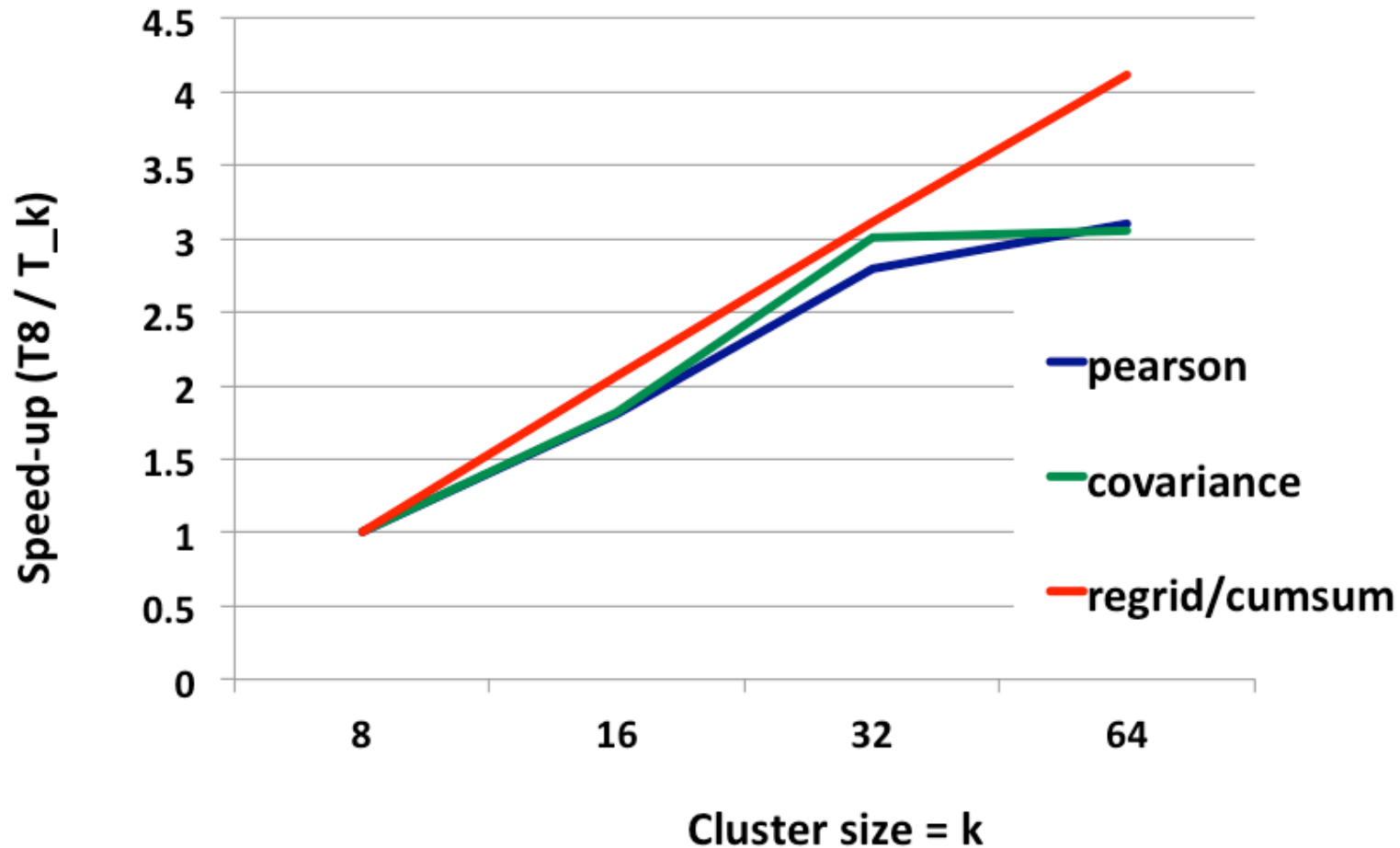
```
regrid( Simple_Array, 10, 10, avg (v2) );
```

```
cumsum (  
    filter ( Simple_Array, v1 = 'Odd' ),  
    I, v1 );
```

- Amazon EC2 Cluster of 8 Through 64 Nodes



Mathematical Operations



The “It’s Beer-o-Clock” Slide

- SciDB
 - Open-source, from the ground up array DBMS and analytic platform
 - <http://www.scidb.org>
- Paradigm4
 - Contributed all the open-source development to date
 - Will distribute and support enterprise-quality software along with P4 add-ons
 - <http://www.paradigm4.com>
- Why SciDB?
 - Query-centric usage model suits scientific / analytic users
 - Arrays support linear algebra and statistical processing
 - Shared-nothing architecture for scalability
 - Building on the shoulders of giants Map/Reduce, RDBMSs, ‘R’
 - Scalable data manipulation and linear algebra