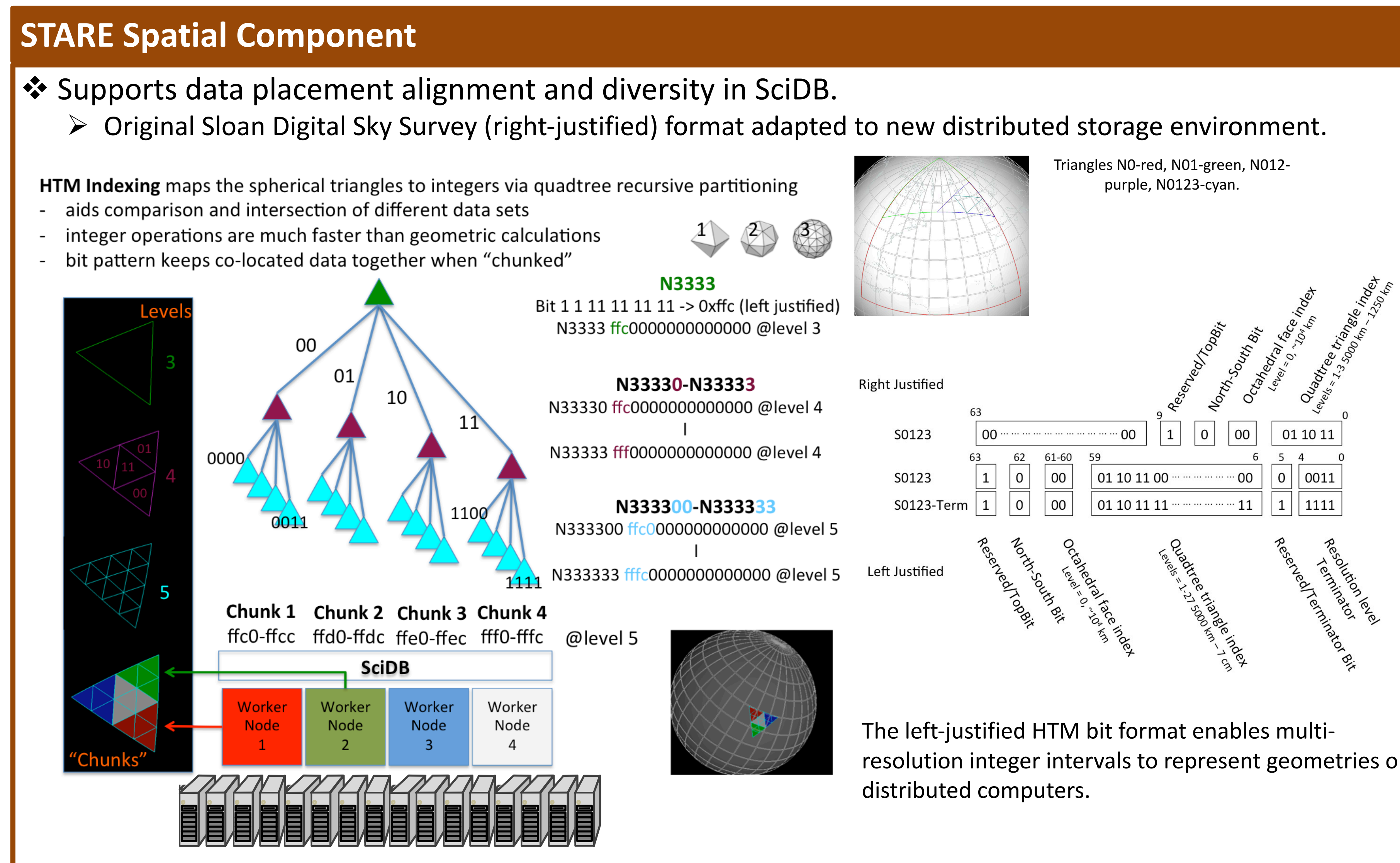


# An Innovative Infrastructure with a Universal Geo-spatiotemporal Data Representation Supporting Cost-effective Integration of Diverse Earth Science Data

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### Abstract

- The SpatioTemporal Adaptive Resolution Encoding (STARE) is a **unifying scheme encoding geospatial and temporal information** for organizing data on scalable computing/storage resources, minimizing expensive data transfers.
- STARE provides a compact representation that **turns set-logic functions into integer operations**, e.g. conditional subsetting, taking into account representative spatiotemporal resolutions of the data in the datasets.
- STARE **geo-spatiotemporally aligns data placements of diverse data** on massive parallel resources to maximize performance.
- Automating important scientific functions (e.g. regridding) and computational functions (e.g. data placement)** allows scientists to focus on domain specific questions instead of expending their efforts and expertise on data processing.
- With STARE-enabled automation, SciDB+STARE provides a database interface, reducing costly data preparation, increasing the volume and variety of interoperable data, and easing result sharing.
- Using SciDB+STARE as part of an integrated analysis infrastructure dramatically eases combining diametrically different datasets.



### STARE Temporal Component

- Hierarchical like HTM
- Bit ranges capture important time resolutions
- Implementation flexible, supporting multiple schema for bit ranges
- Not (necessarily) a tree, but
- Supports intervals like HTM

Meanings of bit ranges (from least to most significant) *prototype*

Range	Starting Bit	Ending Bit	No. Bits	Denoting
0	0	2	3	Resolution
1	3	12	10	millisecond
2	13	24	12	Second
3	25	29	5	Hour
4	30	32	3	Day of week
5	33	34	2	Week
6	35	38	4	Month
7	39	48	10	Year
8	49	58	10	Kilo-annum
9	59	62	4	Mega-annum
10	63	63	1	Before/After

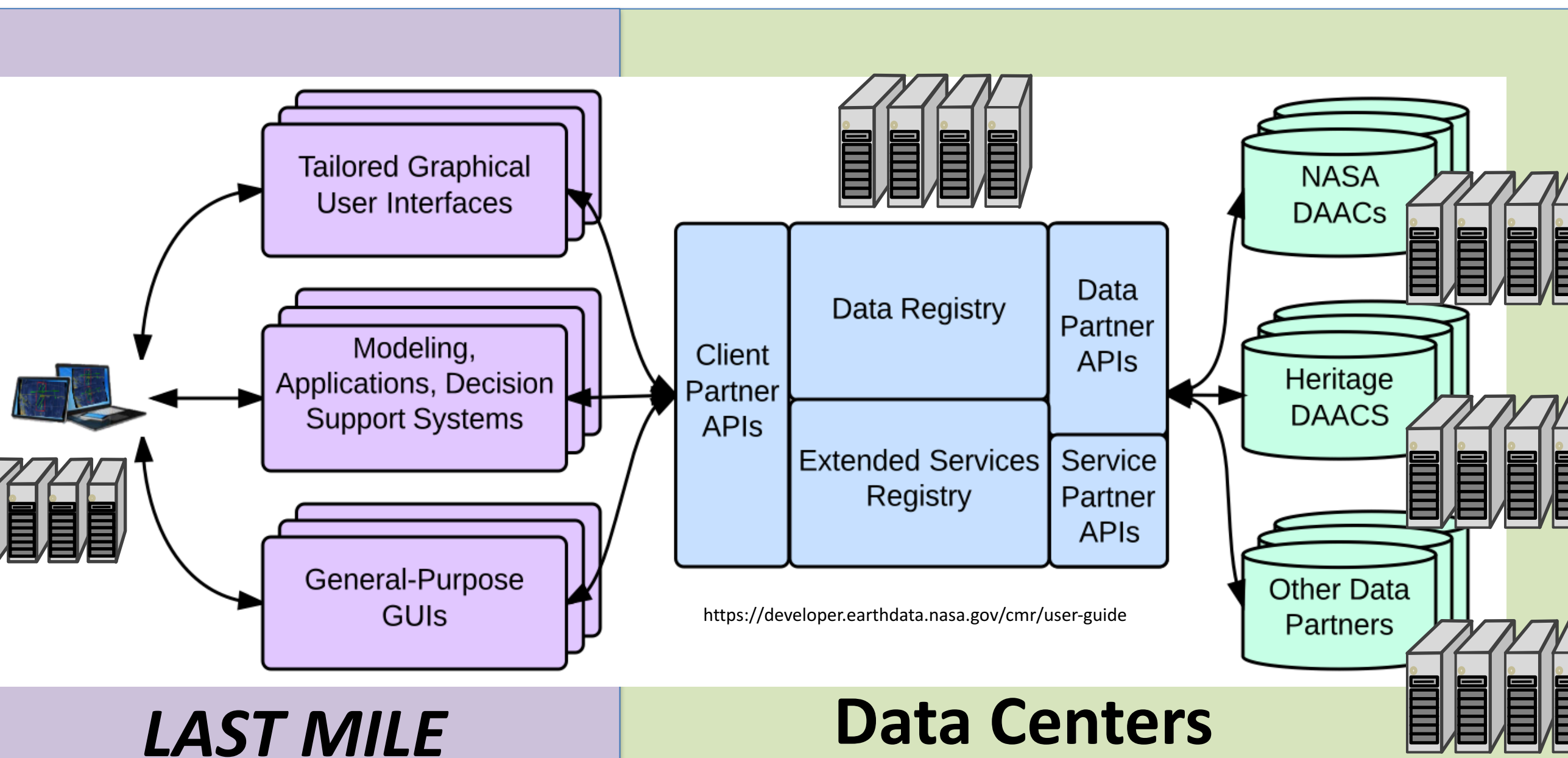
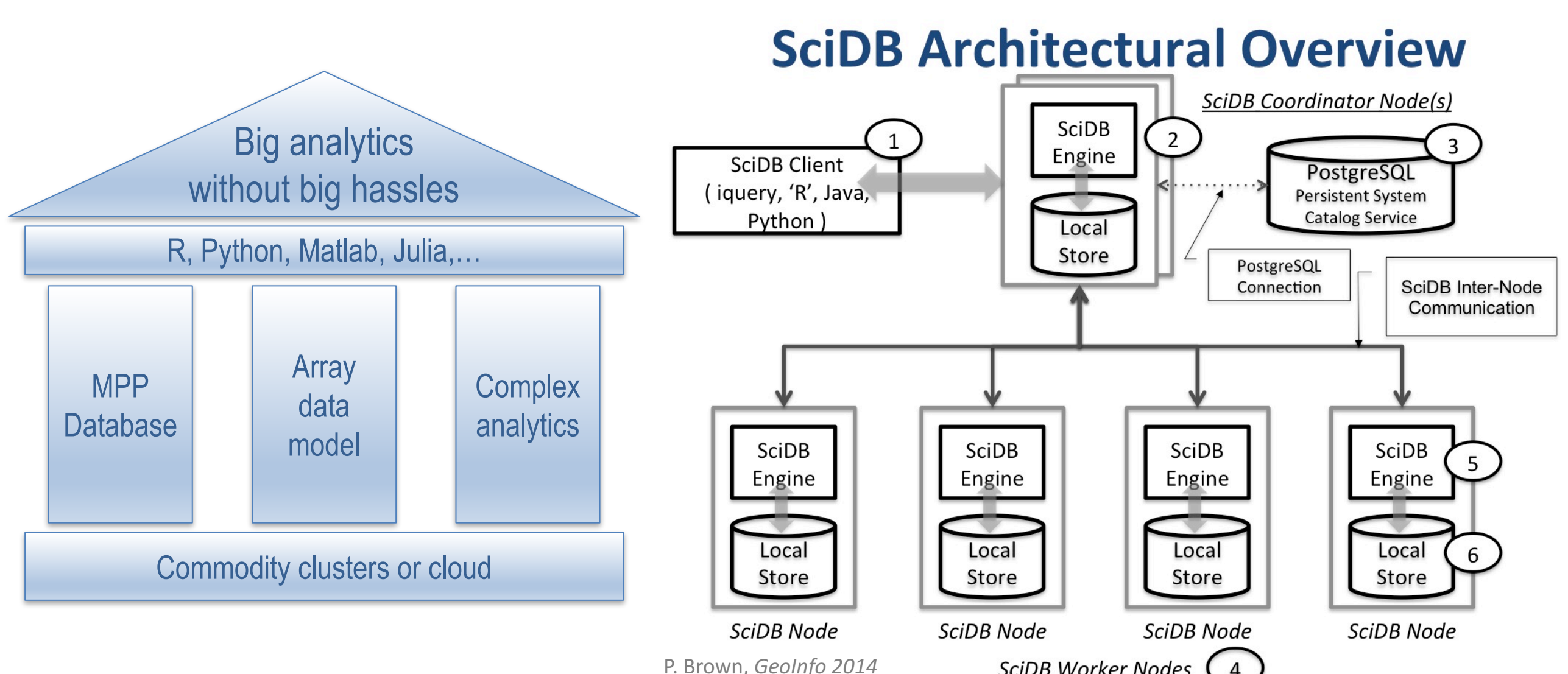
### STARE example metadata for a MODIS granule

```

<notional-modis-stare-metadata>
<TemporalHex> x404e0052c0003
</TemporalHex>
<SpatialHex>
x4a00000000000004,
x4a20000000000004 x4a3fffffffffff,
x4a50000000000004,
x4a80000000000004 x4adfffffffffff,
x4af0000000000004,
x4b10000000000004,
x4b60000000000004,
x4be0000000000004,
x4c00000000000004 x4c1fffffffffff,
x4c30000000000004 x4c7fffffffffff,
x4ca0000000000004,
x4cc0000000000004,
x4ce0000000000004 x4cfffffffffff,
x4f80000000000004,
x4fa0000000000004 x4bfffffffffff
</SpatialHex>
</notional-modis-stare-metadata>
    
```

Resolution level reduced for clarity

- ### Why SciDB?
- #### Resource Consumption Advantages
- Minimize download and local data management
  - Free end-user resources for research and science
- #### Performance Advantages
- Array data model is **better suited for scientific data than relational databases.**
  - Tightly coupled analysis and storage layers allows **better optimization than Spark.**



### SciDB Query

To spatiotemporally "join" the two datasets with STARE indexing (5-min at level 7, i.e. ~78-km resolution):

```

join( nmq_precip, trmm1_2B31 ); - Magic!
    
```

To subset temporally for visualization:

```

select *
into nmq_trmm_09120303
from nmq_trmm1_result - result from previous join
where
tIndex= temporalIndexFromString("2009-11-03 03:00:00.000 (00)") or
tIndex= temporalIndexFromString("2009-11-03 03:04:16.000 (00)");
    
```

Joining a week of NMQ and TRMM 2B31 takes **less than 1 minute** on MAS cluster.

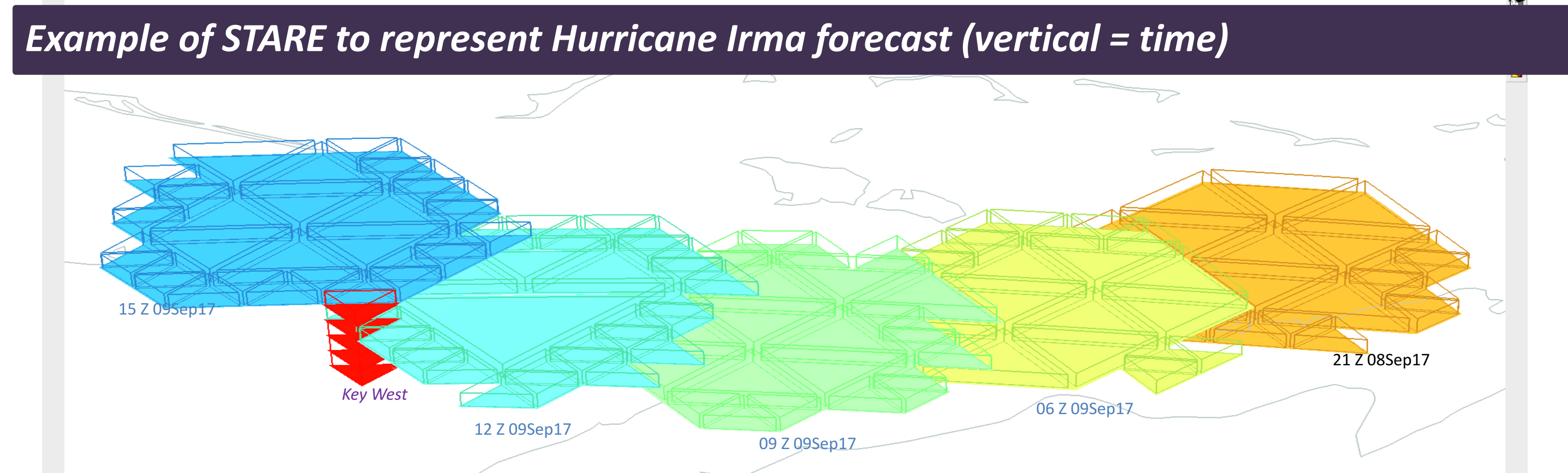
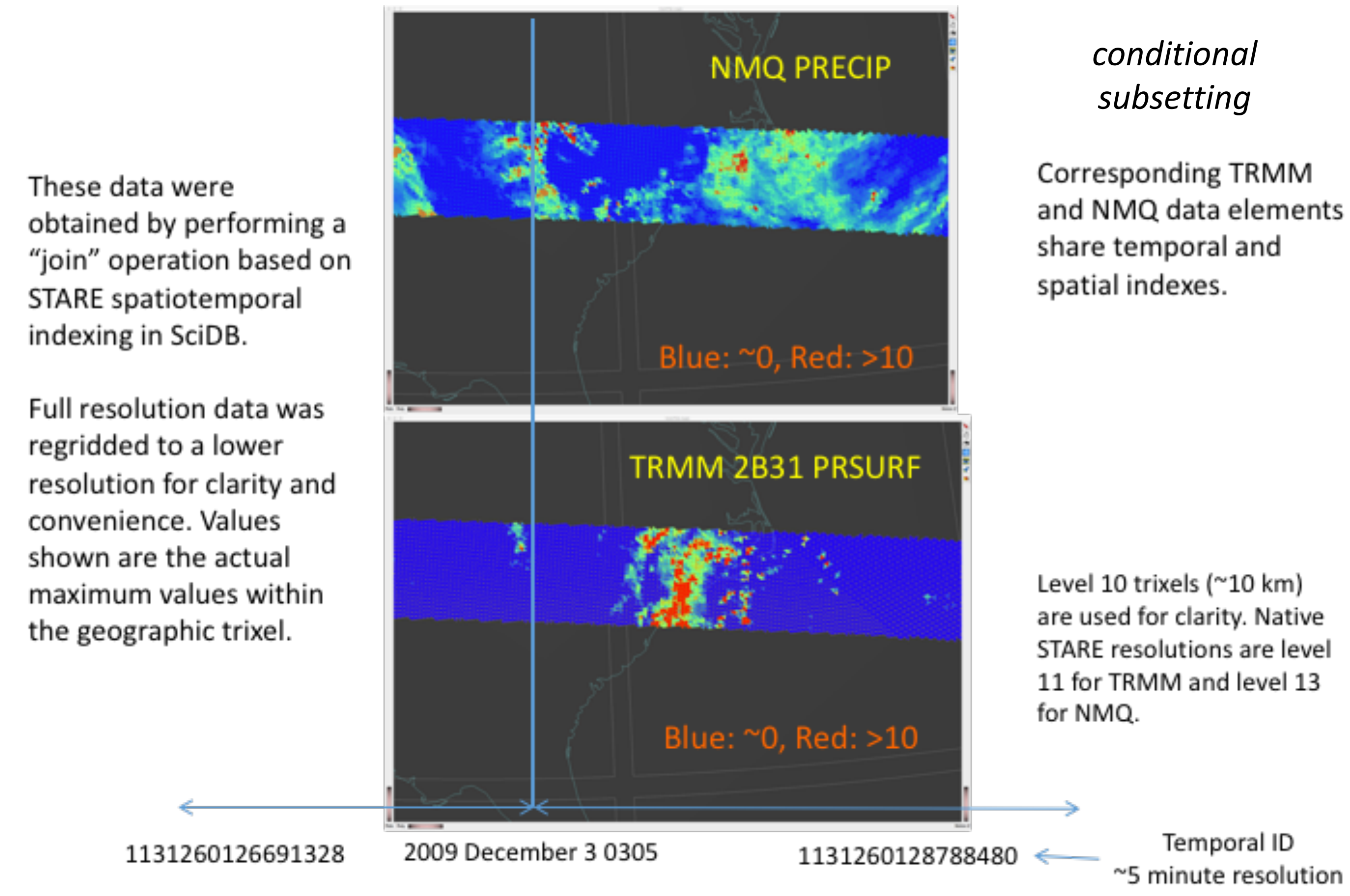
NMQ: 2016(=12\*24\*7) 5-min time slices of 7000x3500 2D floating-point array.

TRMM 2B31: 110 orbit granules of ~9000x49 2D floating-point array (multiple attributes)

No more painstakingly reading and filtering numerous files from various datasets!

Operations can be straightforwardly interfaced with a GUI for *Visual Analytics!*

One remaining performance hurdle is *visualization* – data movement! (see summary for related presentations)



- STARE for spatiotemporal regions for data search and combination
  - Multi-resolution shows flexibility of the scheme
  - Geodesic edges speed geometric calculations
  - Very general regions and temporal structure can be supported
  - Useful for metadata and for general spatiotemporal specification
  - Memory and compute efficient
  - Naturally supports efficient data placement and parallel computing
  - Supports processing closer to where data are stored

### Summary

#### SciDB on STARE

- Provides a unifying scheme for comparing and combining diverse data sets
- Naturally supports data placement alignment for efficient use of SciDB
- Set and logic operations are efficient, straightforward to code
- Transparent use of high-end parallel/distributed compute & storage
- Scientists can work with data via high-level queries
- Growing set of functions for representation, regridding in SciDB enabled by STARE

**Related Presentations:**  
 STARE in Visualization: IN23F-07, IN33C-0141; in the SciDB array database: IN41B-0035 (this work), IN33E-04, and the path to enabling machine learning: IN11E-07.

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