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#### Enabling Near-Data Processing in Distributed Object Storage Systems

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### NDP's Data Distribution Problem (1/2)



- Near Data-Processing (NDP) is a simple concept
  - Move compute to the data to reduce IO traffic
- But, modern distributed storage make this tricky
  - Sharding/striping systems don't respect semantic boundaries

## NDP's Data Distribution Problem (2/2)

S3 Select





- Some systems punt on the issue...
  - Assume replicas, or data collation...
  - Erodes NDP goodness, makes life hard for computational storage devices
- Others go for application specific approaches
  - Ceph Skyhook application has awareness for data placement of tables
  - Some MapReduce apps have record boundary awareness for inter-node reads
  - Can be complex to realize in practice, difficult to retrofit or extend general purpose systems

#### Can we just... you know..... not do that?

#### More concretely:

- 1) Can we lay out data in such a way to avoid **boundary conditions**?
- 2) Can we do so without *application specific changes* to the underlying systems?
- 3) Can we do so in a way that **enables simple NDP** without a lot of storage system awareness?

## Solution: Data Alignment

Adjust data locations elements within shards and stripes

## Background and Observations

- Lots of useful operations can be done on data elements
  - Keyframes in video, row groups of columnar data...
- Distributed storage shards data at relatively coarse granularity
  - In contrast to traditional RAID, MiB ranges are common
  - Fit many data elements in a shard
- Storage systems lay out data in a predictable manner
  - Shard and stripe boundaries are predictable







#### Data Alignment Intuition



7

#### Hints for Striped Environments

![](_page_7_Figure_1.jpeg)

- Stripes are written across shards, and are predictable in their sizes
  - Stripe units reside on one shard
  - We can use stripe units sizes to generate *alignment hints*.
- Alignment hints are simply byte offsets that delimit a stripe unit border
  - With these hints, we can tell when a data element will be in a boundary condition

### Generating Alignment Hints

- Hint generator inputs
  - Maximum stripe size
  - Number of data shards
  - Estimated size of next batch of data elements (DE)
- Data elements are laid out within the bounds of the hints
  - Unused space is padded out
  - If the batch size is less than a stripe, we assume a dynamic stripe resizing.
    - Overhead to ensure space for padding

![](_page_8_Figure_9.jpeg)

#### Layout Example

![](_page_9_Figure_1.jpeg)

# Proof of Concept: CSV Data Filtering

Aligned data stored in MinIO, then processed via parallel containers

![](_page_10_Picture_2.jpeg)

### Experimental Overview

- 4 Node (2+2) MinIO Deployment
- 1.7 GiB CitiBike dataset
  - Used alignment hints to pad data

![](_page_11_Picture_4.jpeg)

- Simple filter query that selects ~25k records
  - Container service offers SQLite functionality
- Compare to built in S3-Select
  - Using unaligned data
- Queries issued from separate node from MinIO

![](_page_11_Picture_10.jpeg)

## CSV Data Filtering + MinIO

![](_page_12_Figure_1.jpeg)

### Results Overview

#### It works!

- We can trivially parallelize many filtering operations
- Compared to built in filtering, significantly reduces data movement
  - Will vary on selectivity
  - Some collation may still be needed, e.g. a SUM
- Low overhead (for CSV)
  - ~8 KiB for padding and extra metadata

#### Data Movements for CSV Query S3 Select vs Co-Located Containers

![](_page_13_Figure_9.jpeg)

#### Lots More To Do!

We started with a very simple example, a good start, but...

- What about more sophisticated data types?
- More native support for NDP
- Quality of service (QOS)
  - We have a complex, distributed scheduling problem

Thanks for your time! Please reach out with questions or comments!

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