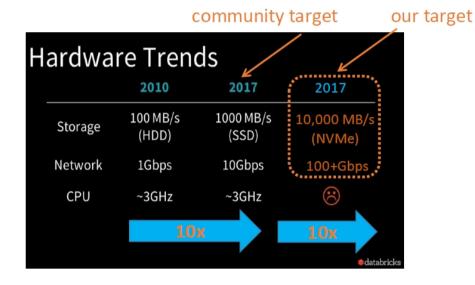
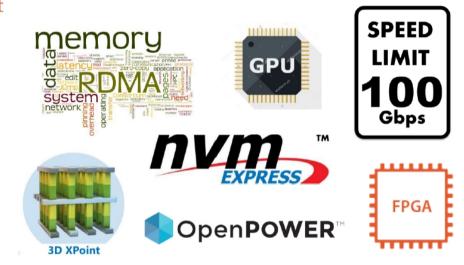
Spark Terasort and CRAIL www.crail.io

Peter Hofstee IBM Research Austin

I/O Hardware Trends

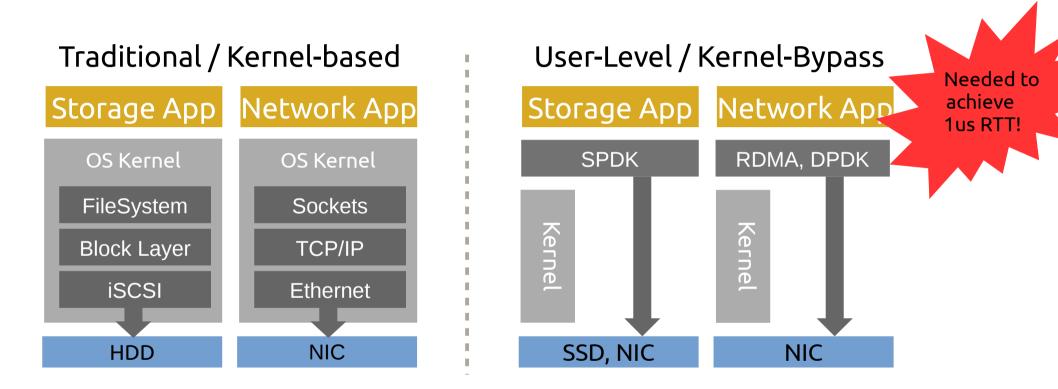
Speed Diversity





- Network interconnects have evolved from
 - Gbps bandwidth to 100Gbps
 - 100us delay to 1us delay
- Storage technology has evolved
 - Factor 100x-1000x

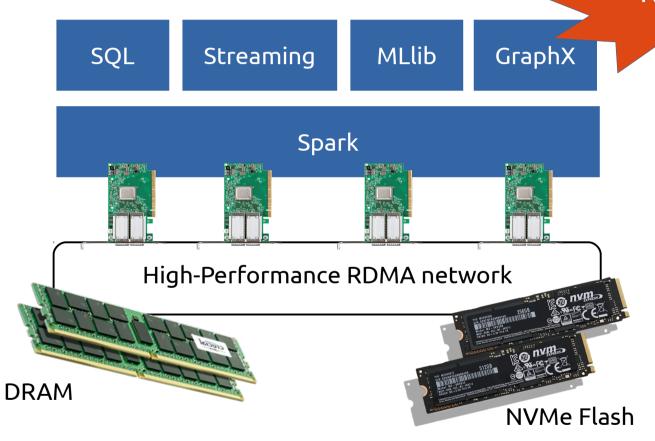
User-Level APIs



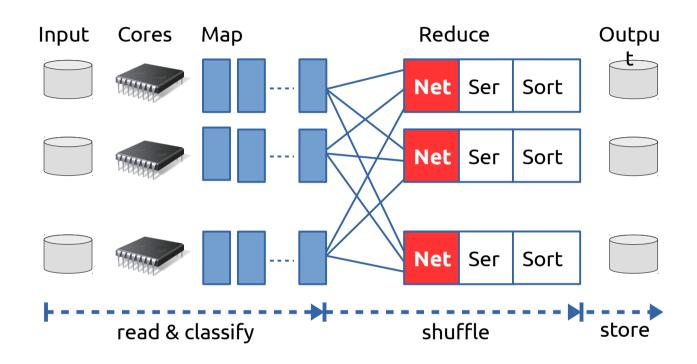
Modern APIs for Networking and Storage offer asynchronous non-blocking user-level access to hardware

Let's Use it!

Performance! Realtime!



Case Study: Sorting in Spark

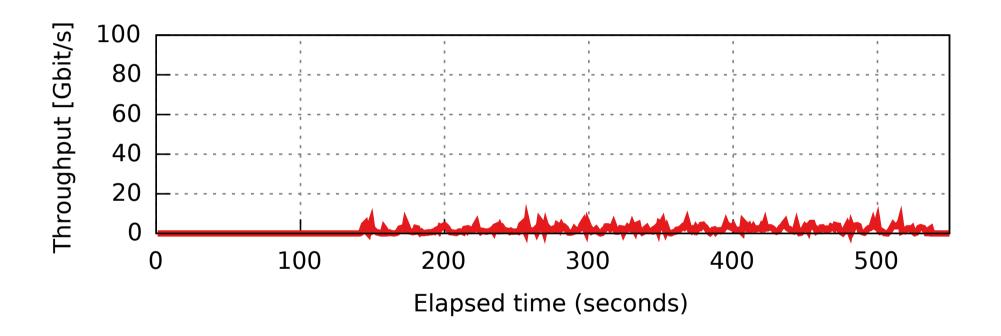


- Map task classify data into local files (typically absorbed by buffer cache)
- Reduce task fetch remote files over the network
- Sorting requires the entire data set to be shuffled over the network

Experiment Setup

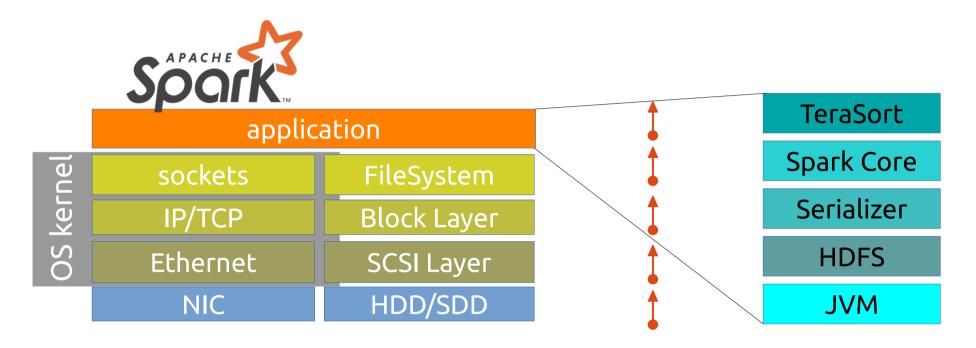
- Total data size: 12.8 TB
- Cluster size: 128 nodes
- Cluster hardware
 - DRAM: 512 GB DDR 4
 - Storage: 4x 1.2 TB NVMe SSD
 - Network: 100GbE Mellanox RDMA
- Software
 - Spark 2.0.0

How is the Network Used?

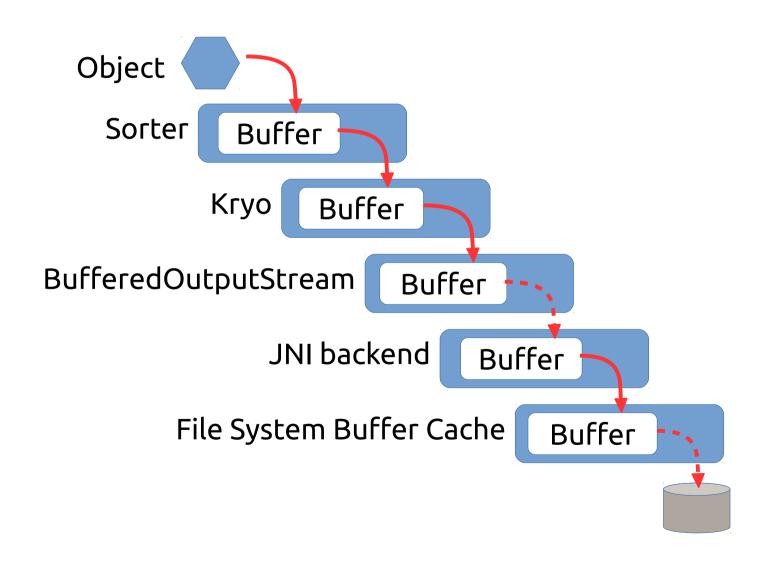


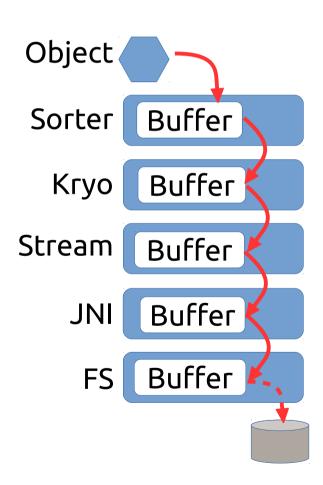
Only 5-10 Gpbs of the network is being used

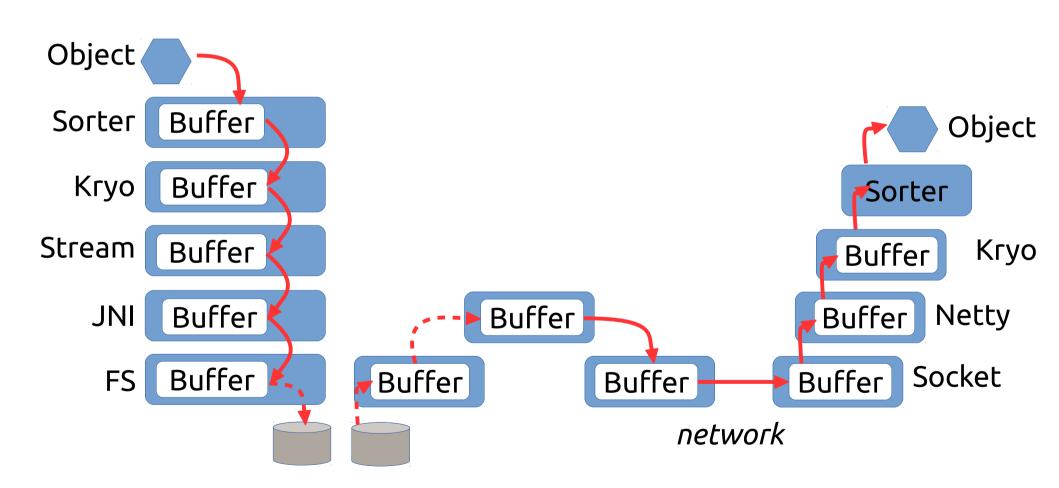
What is the Problem

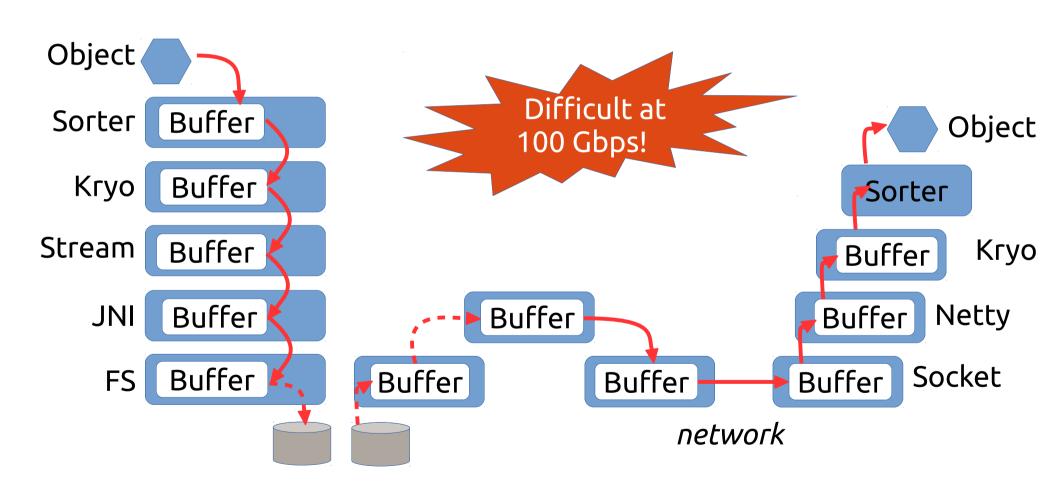


- Application use the legacy APIs
- Applications themselves are heavily layered!
- Overhead during local file system writing
- Overhead during network processing
 - Data copies, context switches, cache pollution, etc





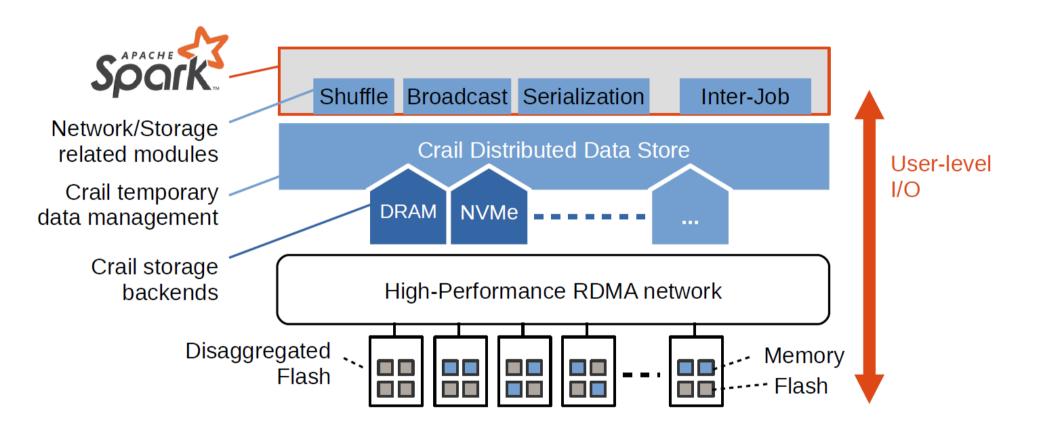




How can we fix this...

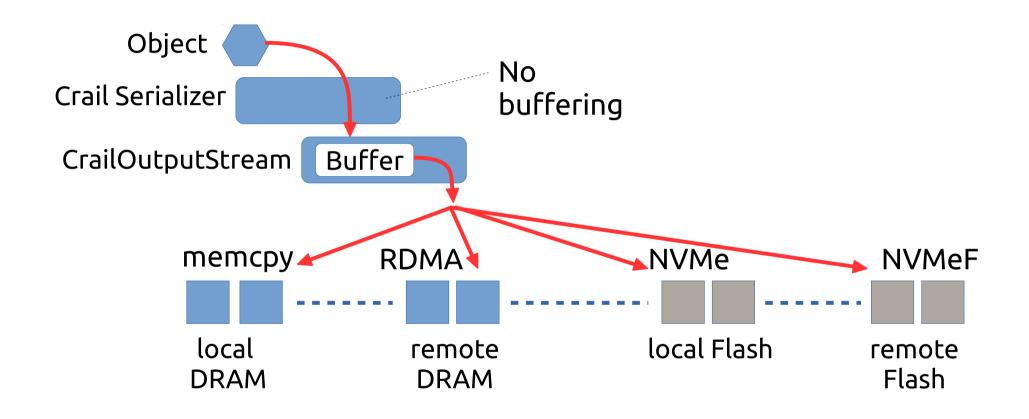
- Not just for shuffle
 - For broadcast, RDD transport, inter-job sharing, etc.
- Not just for RDMA and NVMe
 - For any future high-performance I/O hardware
- Not just for co-located compute/storage
 - Also for disaggregated storage, heterogeneous resource distribution, etc.
- Not just improve things
 - Make it perform at the hardware limit

The CRAIL Approach



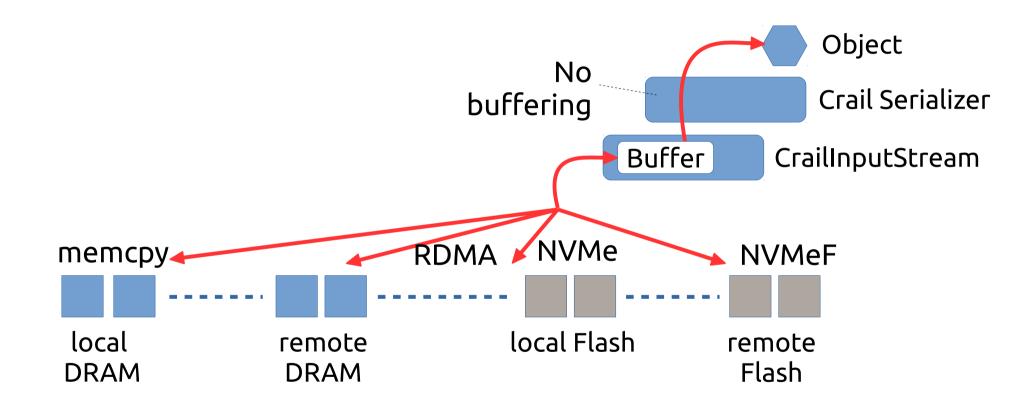
Re-think how I/O is handled in case of fast networking and storage hardware

Example: Crail Shuffle (map)



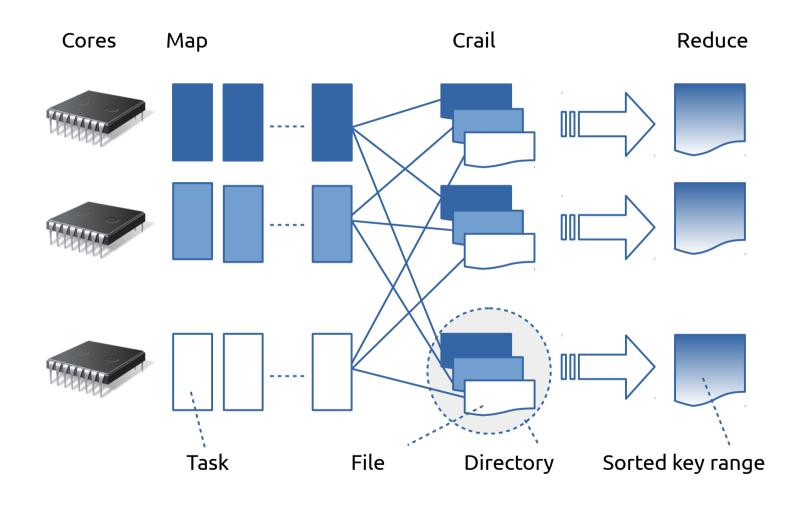
Higher-performing tiers are filled up across the cluster prior to using lower performing tiers

Example: Crail Shuffle (reduce)



Other Spark I/O operations such as broadcast, SQL join, etc., are implemented similarly

Crail Shuffle: File System Layout



Evaluation – Terasort

128 nodes OpenPOWER cluster

- 2 x IBM POWER8 10-core @ 2.9 GHz
- DRAM: 512GB DDR4
- 4 x 1.2 TB NVMe SSD
- 100GbE Mellanox ConnectX-4 EN (RoCE)
- Ubuntu 16.04 (kernel 4.4.0-31)
- Spark 2.0.2

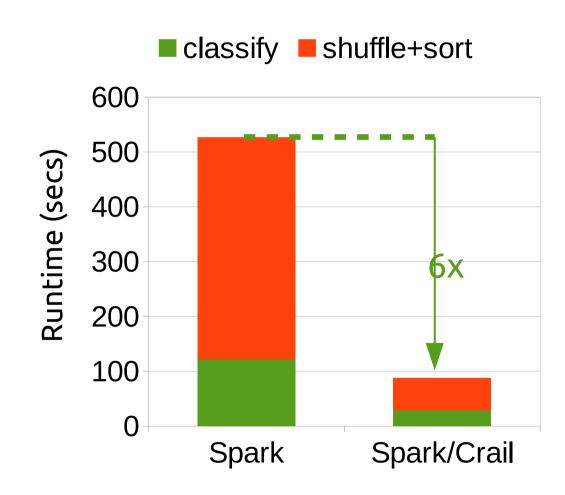
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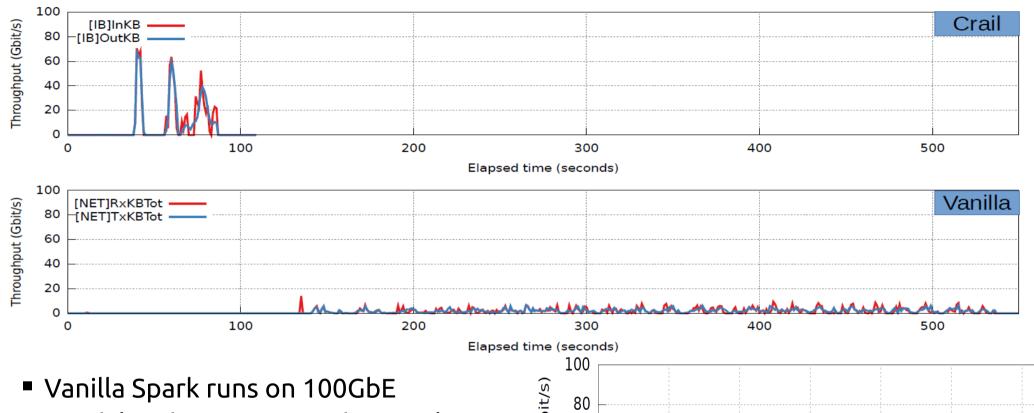
Performance gain: 6x

- Most gain from reduce phase:
 - Crail shuffler much faster than Spark build-in
 - Dramatically reduced CPU involvement
 - Dramatically improved network usage
- Map phase: all activity local
 - Still faster than vanilla Spark

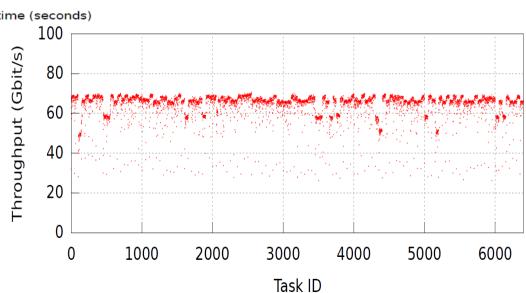


12.8 TB data set, TeraSort

Evaluation – Network IO



- Spark/Crail runs on 100Gb RoCE/RDMA
- Vanilla Spark peaks at ~10Gb/s
- Spark/Crail shuffle delivers ~70Gb/s

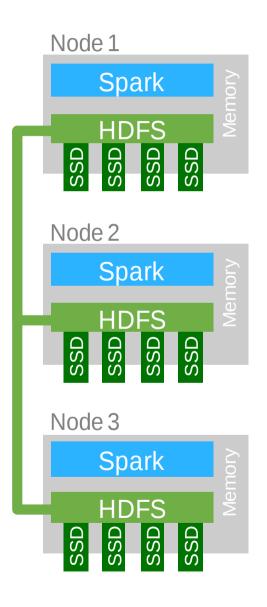


Sorting Comparison

| | Spark + Crail | Spark 2.0.2 | Winner 2014 | Winner 2016 |
|---------------------|------------------|----------------|----------------|----------------|
| Size TB | 12.8 | | 100 | |
| Time sec | 98 | 527 | 1406 | 134 |
| Cores | 2560 | | 6592 | 10240 |
| Nodes | 128 | | 206 | 512 |
| NW Gb/s | 100 | | 10 | 100 |
| Rate TB/min | 7.8 | 1.4 | 4.27 | 44.78 |
| Rate/core GB/min | 3.13 | 0.58 | 0.66 | 4.4 |

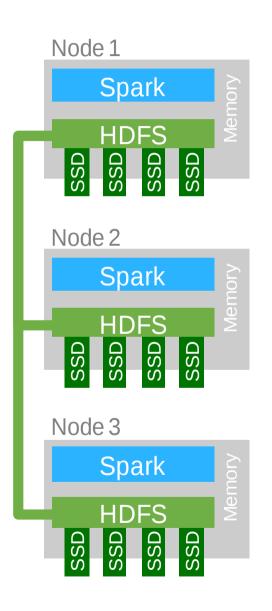
- Spark/Crail CPU efficiency is close to 2016 sorting benchmark winner: 3.13 vs. 4.4 GB/min/core
- 2016 winner runs native C code!

Storage Disaggregation

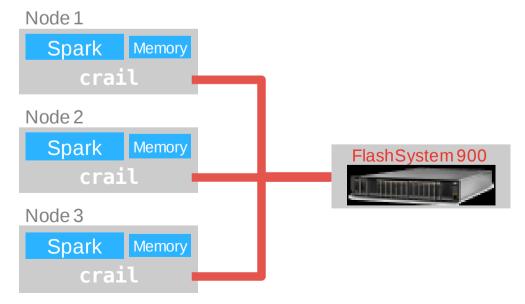


- Why disaggregation?
 - Independent scaling of compute and storage
 - Higher utilization due to less fragmentation
 - Easier maintenance
- Challenges:
 - Systems like Hadoop/Spark have been designed for local storage
 - But: new <u>fast networks</u> may permit storage disaggregation

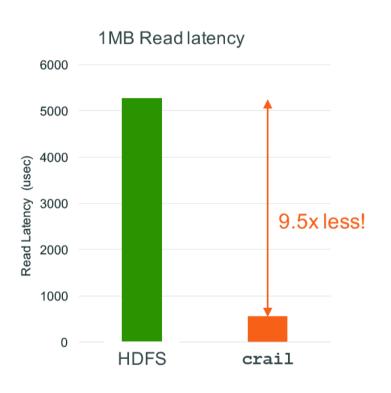
Storage Disaggregation



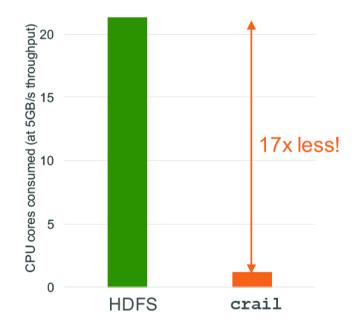
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IBM Flashsystem: Crail vs HDFS



Total CPU utilization @ 5 GB/s throughput



HDFS setup

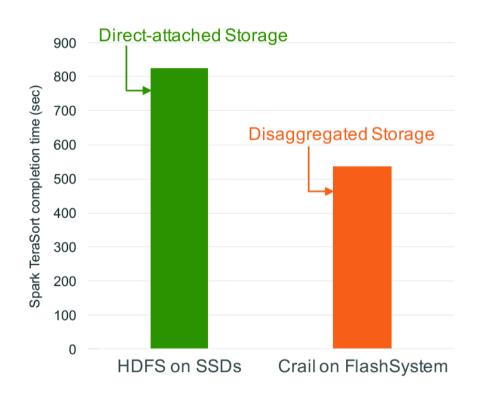
- 10 node cluster
- 56 Gbit Infiniband network
- 2 x 1TB SSDs / node
- No replication

crail setup

- 10 node cluster
- 56 Gbit Inifiniband network
- 1 x FlashSystem 840
 - 8 Flash cards
 - 23TB usable capacity

The two systems have the same bandwidth from Flash (~10 GB/s) and about the same total capacity.

IBM Flashsystem: TeraSort with HDFS vs Crail



Experimental Setup

- Sorting 400GB of data using Spark
- HDFS setup
 - 10 node cluster, 56 Gbit Infiniband network
 - 2 x 1TB SSDs / node, 2-way replication
 - HDFS is using host memory (OS page cache)
- crail setup
 - 10 node cluster, 56 Gbit Inifiniband network
 - 1 x FlashSystem 840 (8 Flash cards, 23TB usable)
 - Crail is not using host memory
- The two systems have the same bandwidth from Flash (~10 GB/s) and about the same total Flash capacity.

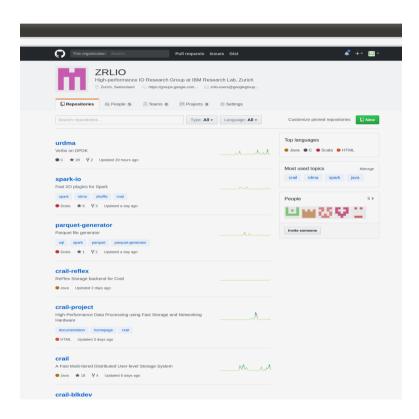
Crail + FlashSystem achieves 40% performance improvement with lower TCO and all the benefits of disaggregation

Crail is Open Source!

www.crail.io



https://github.com/zrlio



Related Work

Three classes of related work:

- New Data Processing Systems for High-Performance Network & Storage Hardware
 - FARM, RamCloud, HERD, etc
 Fast, but require applications to be written from scratch
- Updates/patches to existing Systems
 - Ohio Spark/Hadoop Distro

Slow because no radical changes possible: fetrofitting RDMA/Flash integration into existing file/socket based I/O stacks

- Memory/Flash caches/stores
 - Example: Tacyon

Slow because not designed for high-performance hardware

Conclusion

Today's open source analytics stacks:

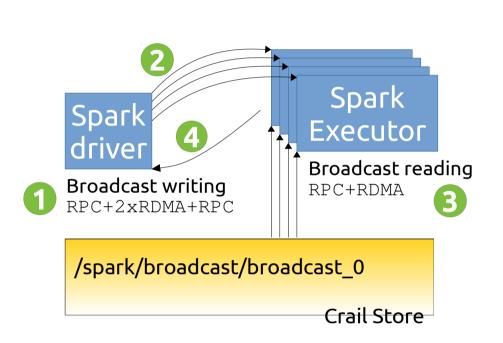
- Existing analytics stacks designed for yesterday's commodity hardware
- Performance on high-end hardware inhibited by heavy-layered stack architecture

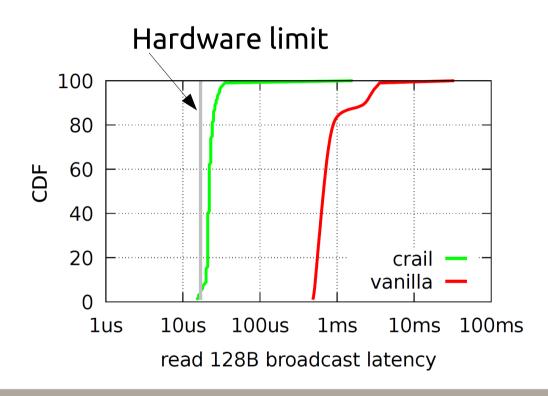
The Crail Approach:

- Radical re-design of I/O (network & storage) for analytics by exploiting modern hardware
 - RDMA, NVMe & NVMe over fabrics
- Enable high-performance disaggregated storage for analytics
- Extend Spark operation to take advantage of Crail
- Crail is open source: www.crail.io

Backup

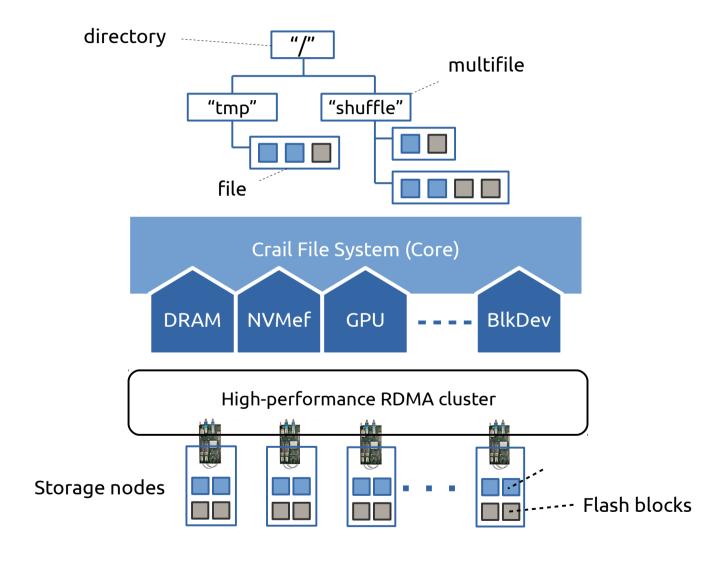
Spark/Crail Broadcast





```
val bcVar = sparkContext.Broadcast(new Array[Byte](128))
sparkContext.parallelize(1 to tasks, tasks).map(_ => {
   bcVar.value.length
}).count
```

The Crail Store



Crail Storage Tiering

