Application-Centric SSD Cache Allocation for Hadoop Applications

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SSD vs HDD



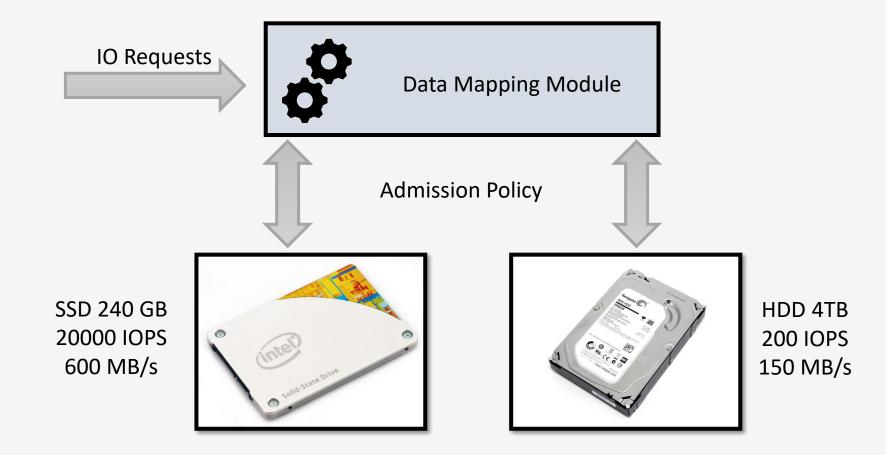
- •) Higher Performance, especially in random access
- Expensive in Per GB Capacity



- Large Capacity
- Lower random IOPS (I/O Operations Per Second)

SSD Caching System is a balance between **Cost** and **Performance**

SSD Caching System



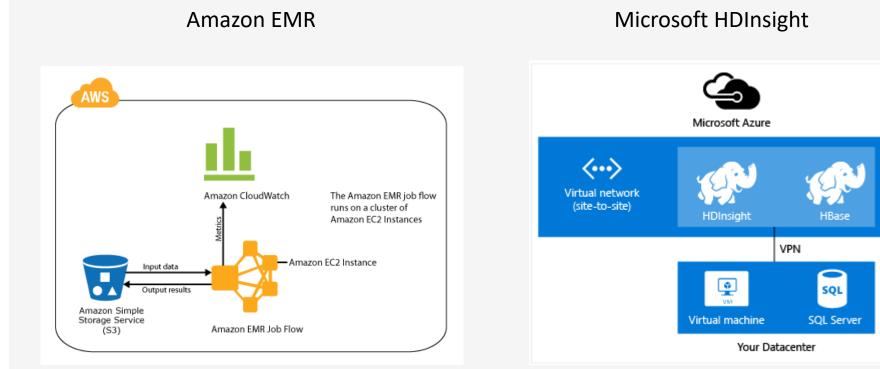
The Data Mapping Module is the Fundamental Part

SSD Cache is Widely Used in the Virtualization Environment

Amazon Elastic Block Store (EBS) Supporting Amazon EC2

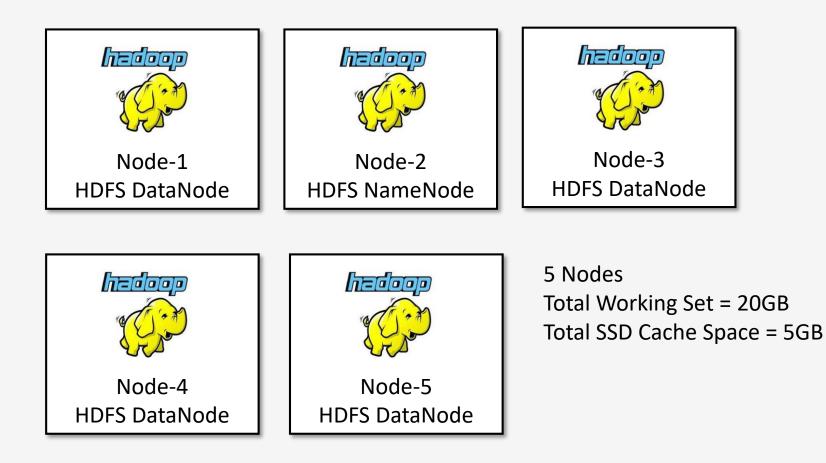
	Solid State	Drives (SSD)	Hard Disk	Drives (HDD)		EBS Magnetic
Volume Type	EBS Provisioned IOPS SSD (io1)	EBS General Purpose SSD (gp2)*	Throughput Optimized HDD (st1)	Cold HDD (sc1)	Volume Type	EBS Magnetic
	Highest performance SSD volume designed for	General Purpose SSD volume that balances price	Low cost HDD volume designed for frequently	Lowest cost HDD volume designed for	Use Case	Infrequent Data Access
Short Description	latency-sensitive transactional workloads	performance for a wide variety of transactional workloads	accessed, throughput intensive	less frequently accessed workloads	API Name	standard
			workloads		Volume Size	1 GB - 1 TB
Use Cases	I/O-intensive NoSQL and relational databases	Boot volumes, low-latency interactive apps, dev & test	Big data, data warehouses, log processing	Colder data requiring fewer scans per day	Max IOPS/Volume	40-200
API Name	io1	gp2	st1	sc1	Max IOPS Burst Performance	-
Volume Size	4 GB - 16 TB	1 GB - 16 TB	500 GB - 16 TB	500 GB - 16 TB	Max Throughput/Volume	40-90 MB/s
Max IOPS**/Volume	20,000	10,000	500	250	Wax mioughpub volume	40-00 MD/3
Max Throughput/Volume	320 MB/s	160 MB/s	500 MB/s	250 MB/s	Max Throughput Burst Performance	-
Max IOPS/Instance	75,000	75,000	75,000	75,000	Max IOPS/Instance	48,000
Max Throughput/Instance	1,750 MB/s	1,750 MB/s	1,750 MB/s	1,750 MB/s		
Price	\$0.125/GB-month	\$0.10/GB-month	\$0.045/GB-month	\$0.025/GB-month	Max Throughput/Instance	800 MB/s
	\$0.065/provisioned IOPS				Disc	\$0.05/GB-month
Dominant Performance Attribute	IOPS	IOPS	MB/s	MB/s	Price	\$0.05/million I/O

And Supports the Elastic Hadoop Clusters



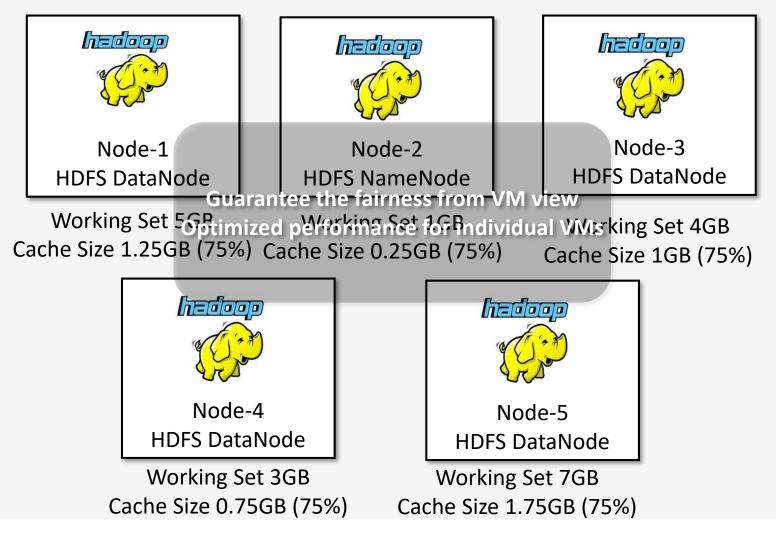
The VM View and the Application View

VM view: IO Latency Application view: **Job Completion Time**

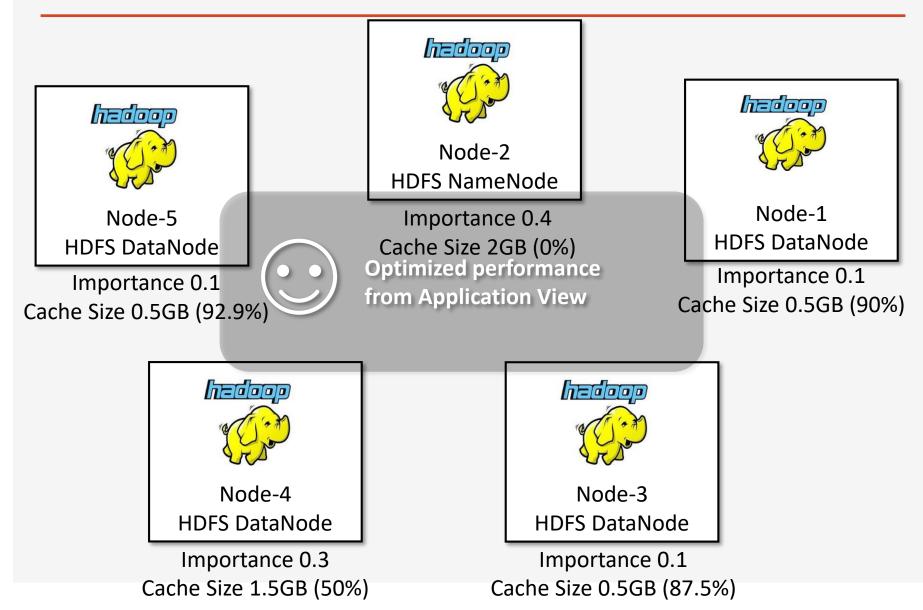


VM-centric Approach

Allocating the SSD cache according to the working set , aiming to minimize per-VM IO latency

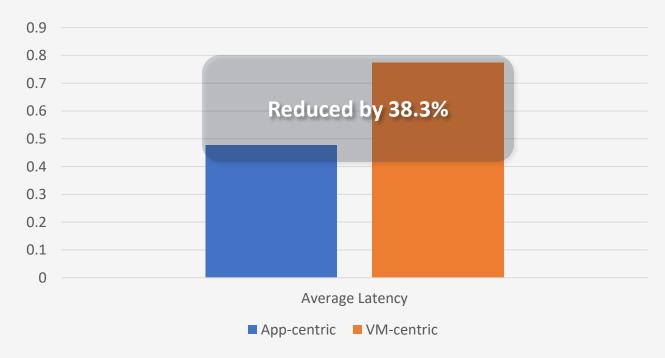


However, as the importance of nodes are different



Improve Application-Level Performance

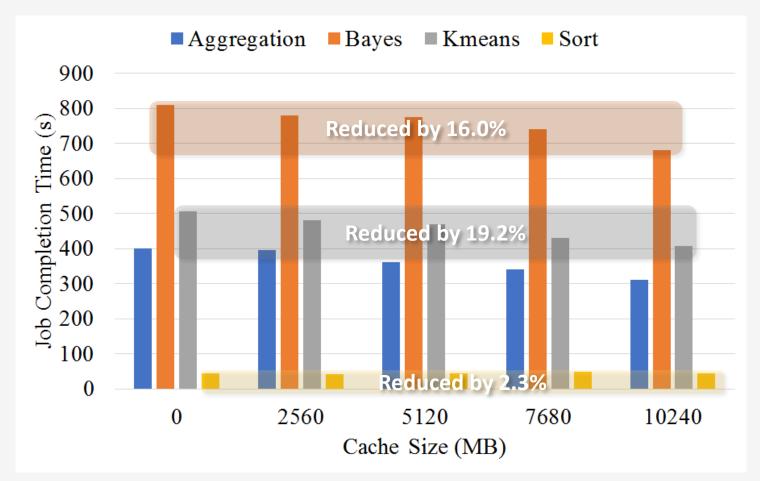
- Latency related to job execution = Importance(VM) * Latency(VM)
 - VM-centric: 0.775I_{HDD}
 - App-centric: 0.47836l_{HDD}
- Average latency related to job execution **reduced by 38.3%**



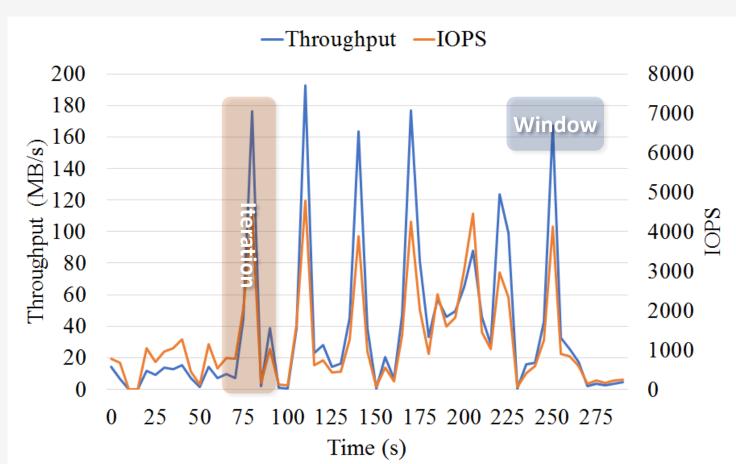
The relationships among VMs inside the application cannot be ignored

Two Principals

Allocating too much SSD Cache is unnecessary due to different importance



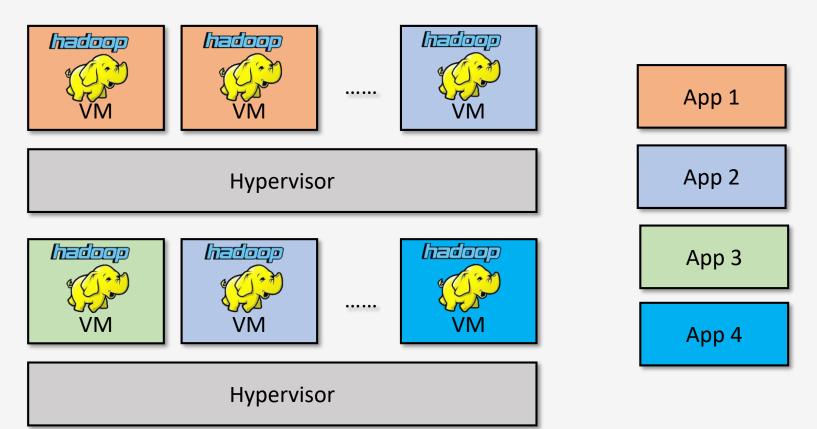
Two Principals



For different stages of the workload, the requirements may be different

We Also Need Global View

One application may be deployed on multiple hypervisors Multiple applications may be deployed on one hypervisor We need SSD cache allocation **from the global view**



How to allocate SSD cache from application view?

- Problem: How to allocation per-VM SSD cache for elastic Hadoop clusters, to reduce the job completion time?
- Solution: Application-centric SSD cache allocation (AC-SSD)
- Two challenges
 - How to **allocate** appropriate SSD cache resources (space and IOPS) for virtual machines inside the Hadoop cluster from the application view?
 - How to **dynamically change** the plan to adapt to continuously and dynamically changing workloads?

Application-Centric SSD Cache Allocation (AC-SSD)

- To figure out the importance and allocate per VM SSD cache
 - Use genetic algorithm to calculate the nearly optimal weights for VMs based on importance
 - Weight based SSD cache allocation
- To react to rapidly changing workload
 - Use closed loop adaptation

Genetic Algorithm

- Proving that the per VM SSD cache allocation is NP complete (more details in paper)
- Definition
 - Chromosome: Tuple {*w_{Storage}*, *w_{IOPS}*}, indicates weights of SSD cache space and IOPS
 - **Genome**: A set of chromosomes, indicates the SSD cache allocation plan
 - **Selection**: Select genomes randomly by fitness
 - Crossover: Select random numbers of chromosome of two genomes and swap them
 - Mutation: Change the tuple of chromosome within a specific range

Fitness Calculation

- We calculate the fitness from 3 levels
 - Importance Factor: Indicates the contribution of VMs.
 - IO time: Indicates whether the workload is IO sensitive.
 - Average request size: indicates the access pattern, whether sequential or random.

Importance Factor

- For the execution of Hadoop application
 - The nodes require data locally and from other nodes
 - The data dependency indicates the contribution to the job completion time
- Based on the observation
 - We use the **ratio** of network throughput to IO throughput the importance

factor

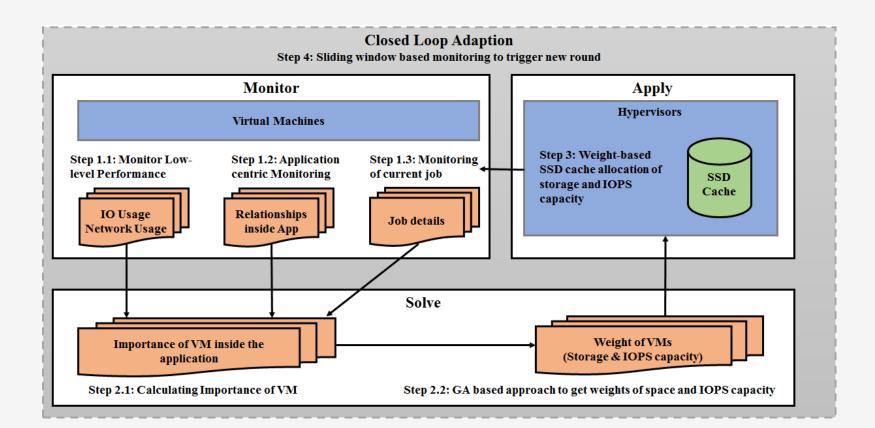
VM 1 IO throughput = 500 MB Network throughput = 250 MB Ratio=0.5

VM 2 IO throughput = 1000 MB Network throughput = 200 MB Ratio=0.2

VM 3 IO throughput = 800 MB Network throughput = 640 MB Ratio=0.8

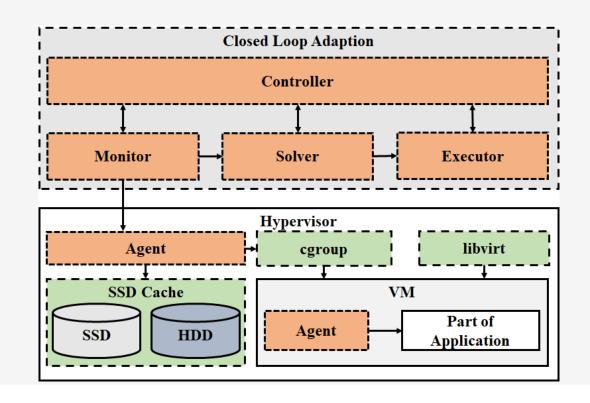
Closed Loop Adaptation

• Consists of 3 main steps: Monitor, Solve and Apply



Implementation

- We implement AC-SSD on Xen hypervisor
 - Supporting closed loop adaptation: Java based controller, monitor, solver, executor; agents on VMs and hypervisor
 - Use cgroup to control the weight of IOPS
 - LRU based SSD cache

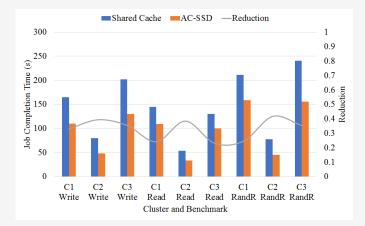


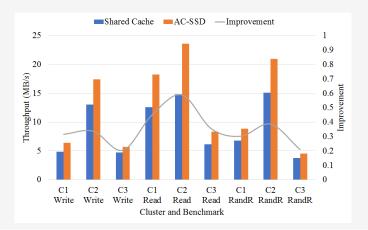
Experiment Setup

- Environment
 - 20 VMs hosted on 4 hypervisors
 - 3 Clusters of 5, 10 and 5 nodes, with different VM placements
 - 640MB SSD cache for each hypervisor
- Benchmarks
 - IO Sensitive (TestDFSIO)
 - Hadoop micro benchmarks (Sort, Terasort, Wordcount)
 - SQL micro benchmarks (Aggregation, Join, Scan)
 - Machine learning micro benchmarks (Bayes, KMeans, PageRank)

Result - IO sensitive

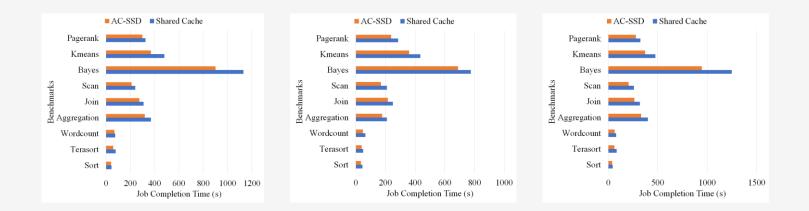
- Compared to shared cache
- For IO sensitive workloads (TestDFSIO)
 - Job completion time reduced by 31% in average
 - Throughput improved by 35% in average
 - Better for read workloads





Result - Benchmarks

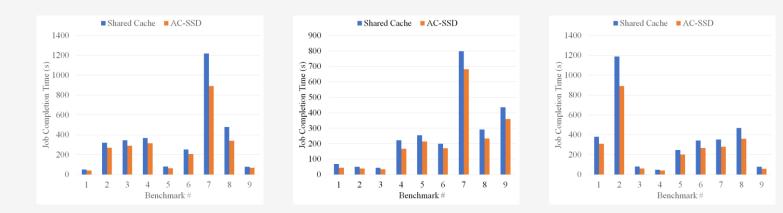
- Compared to shared cache
- For micro benchmarks
 - Job completion time reduced by 14.3%~17.8%
 - Works better for Kmeans and Bayes benchmark



Result - Self Adaptation

- Compared to shared cache
- For rapid changing workloads
 - Reduced by ~20% for rapidly changing workloads, for 3 clusters

No.	Workload - C1	Workload - C2	Workload - C3
1	Sort	Wordcount	Aggregation
2	Join	Terasort	Bayes
3	Pagerank	Sort	Wordcount
4	Aggregation	Aggregation	Sort
5	Terasort	Join	Scan
6	Scan	Scan	Join
7	Bayes	Bayes	Pagerank
8	KMeans	Pagerank	KMeans
9	Wordcount	KMeans	Terasort



Conclusion

- We present AC-SSD, an Application-Centric SSD caching system
 - Present the importance factor of VMs inside the application, based on the network throughput and disk IO
 - Use genetic algorithm to calculate the nearly optimal weight of VMs
 - Use closed loop adaptation to react to rapidly changing workloads
 - The evaluation shows that it reduces the job completion time comparing to the shared cache

Thanks

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