# A Unified Data Platform for Big Data & Deep Learning

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IBM Spectrum Scale Development and Client Adoption

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#### Agenda

- Challenges for Storage System in Big Data and Analytics
- Spectrum Scale Big Data and Analytics Solution
- Spectrum Scale for Deep Learning
- Customer Case Sharing



# Challenges for Storage System in Big Data and Analytics

### Typical Challenges in Big Data Analytics (1/4)

#### Inefficient data movements and many copies

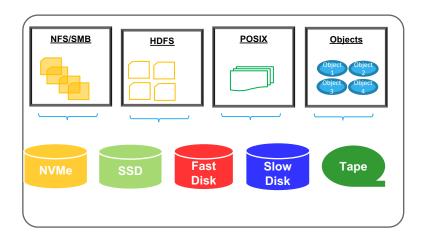
<u>NFS/SMB</u>	<u>HDFS</u>	<u>POSIX</u>	<u>Objects</u>
			Object 1 2 Object 3 Object 4
8888	6666	8888	8888
Storage System #1	Storage System #2	Storage System #3	Storage System #4

- Different analytics workloads might need different interface
  - ✓ map/reduce jobs need HDFS interface
  - ✓ Data ingestion clients need NFS/SMB interface
  - Traditional workloads(such as data warehouse, HPC etc) requires POSIX interface
- Unrich data accessing interfaces from storage system make customers have to set up siloed infrastructure
- Siloed data lakes bring inefficient data analysis(e.g. long time analysis because of data movement)
- Siloed data lakes bring inefficient storage space utilization(e.g. many copies on different storage systems)
- Multiple siloed storage systems bring further issues on cost, management and scaling.



### Typical Challenges in Big Data Analytics (2/4)

#### Inefficient data lifecycle management

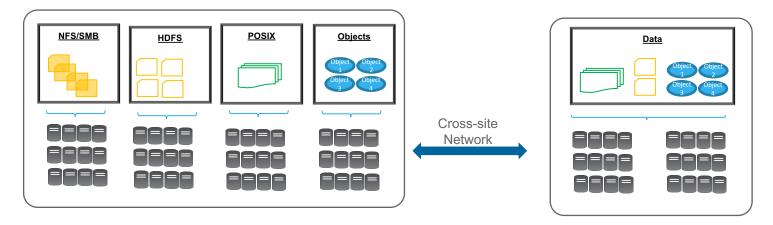


- Different tiers for different data(hot/warm/cold data)
- Leverage different disks from NVMe to Tape according to performance/cost
- Policy-based data movement automatically
- Easily move cold data back when needed



### Typical Challenges in Big Data Analytics (3/4)

#### **Cross-site HA or DR for business continuity**

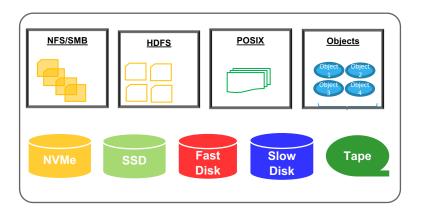


- Flexible options for different business continuity requirements
  - ✓ Different RTO(Recovery Time Object)/RPO(Recovery Point Object)
- Different options for low/medium/high cost for cross-site HA or DR



### Typical Challenges in Big Data Analytics (4/4)

#### Inefficient Data Archive and Share to Cloud Storage



- Efficient data sharing cross different sites
- Efficient data sharing between off-premise and onpremise
- Flexible options considering different cost/requirements



# Spectrum Scale Big Data and Analytics Solution

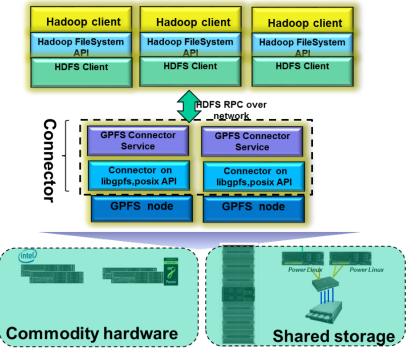
### Spectrum Scale Supports Big Data and Analytics

- Spectrum Scale completely transparent to Hadoop
  - Spectrum Scale HDFS Transparency Connector makes Spectrum Scale transparent to Hadoop
  - HDFS Transparency Connector Works with Spectrum Scale 4.1.X/ 4.2.x/5.0.x
  - Shipped with Spectrum Scale 4.2.x/5.0.x
    - Download the latest from <u>IBM developerWorks GPFS</u> wiki
- HDFS Transparency and Spectrum Scale services could be managed from HortonWorks HDP Ambari GUI
- Spectrum Scale and Hortonworks Data Platform are certified on both IBM Power and X86.



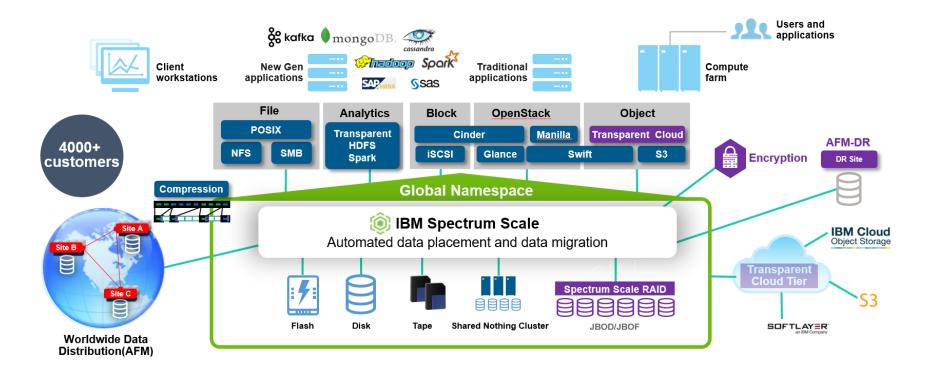


#### hdfs://hostnameX:portnumber



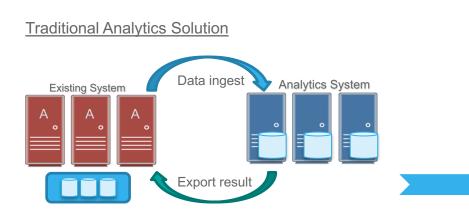


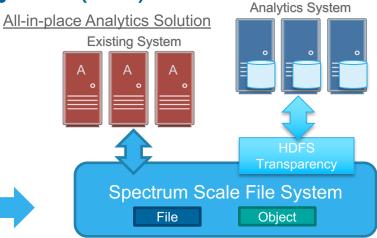
### **Unified Data Platform for In-lace Analytics**





### Spectrum Scale In-Place Analytics (1/2)





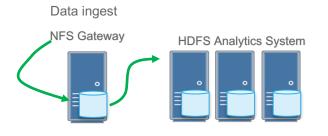
- Build analytics system from scratch, not only for compute but also for storage
- Add storage and compute resource at the same time no matter it's required
- Native HDFS doesn't support native POSIX
- Lacks of enterprise data management and protection capability

- ✓ Can leverage existing Spectrum Scale storage
- ✓ Unified interface for File and Object analytics
- POSIX compatibility
- Mature enterprise data management and protect solutions derived from Spectrum Storage family and 3<sup>rd</sup> part components



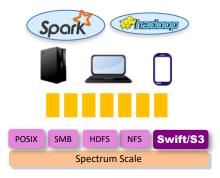
### Spectrum Scale In-Place Analytics (2/2)

#### NFS interface from HDFS



- HDFS doesn't support random read/write, only append mode
- HDFS NFS Gateway has to write data from clients to the local disks first and then move it to HDFS to handle the out-of-order write requests
- No HA for NFS Gateway so far

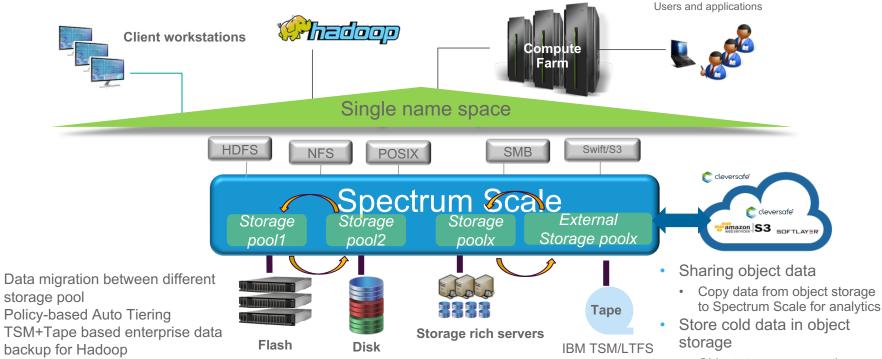
#### Spectrum Scale In-place Analytics Solution



- Rich data ingest interface(SMB, NFS, HDFS, swift/S3 etc)
- ✓ Spectrum Scale Protocol HA
- Random read/write support
- Efficient data ingest because of no data movement from local disk to Spectrum Scale
- Only one data copy and all data are visible immediately from all interface



### Spectrum Scale ILM for Data Analytics



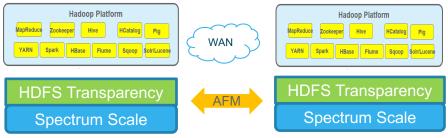
- Object storage as one tier
- Policy-driven object migration



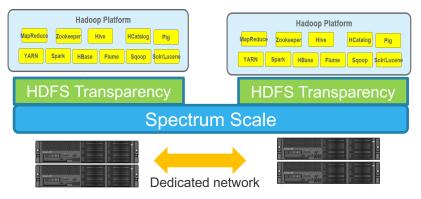
### Spectrum Scale DR Solutions for Analytics

- Hbase Cluster Replication
  - WAL-based asynchronous way
  - All nodes in both cluster should be accessible for each other
  - Both clusters could provide Hbase service on the same time
  - Only available for Hbase
  - Supported over HDFS Transparency + Spectrum Scale
- Spectrum Scale AFM IW-based replication
  - AFM IW-based replication from production cluster(cache site) to standby cluster(home site)
  - Both sites should be Spectrum Scale cluster for Hadoop application failover
  - Only one cluster can provide Hbase service(conflict in assigning region servers if Hbase is up on both cluster)
- Spectrum Scale Active-Active DR
  - 2 replica in production cluster; another replica in standby cluster
  - Dedicated network for two clusters(10Gb bandwidth)
  - Distance between two clusters is less than 100Km
  - Can achieve RPO=0 in DR



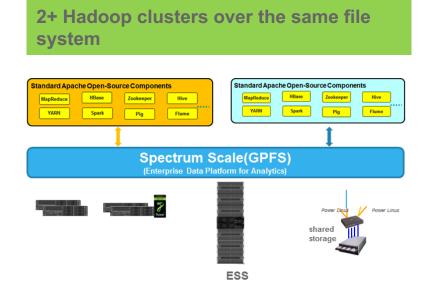




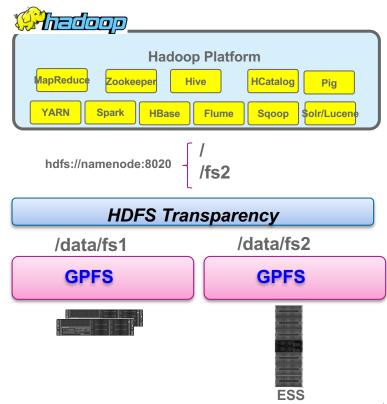




#### Spectrum Scale BD&A Solution New Features



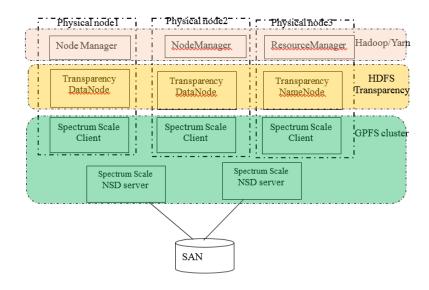
#### **Natively Support 2+ File Systems**



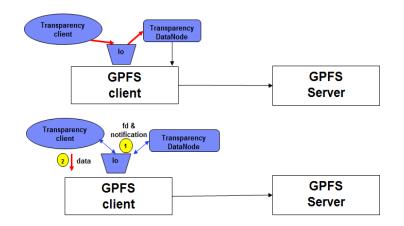


#### Spectrum Scale BD&A Solution New Features

#### **Short Circuit Write**



- All Hadoop nodes are Spectrum Scale nodes/HDFS Transparency nodes
- Short circuit write reduces the traffic between client and DataNode on the loop lo adapter





# Spectrum Scale for Deep Learning

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### Spectrum Scale for Deep Learning

#### Spectrum Scale for Deep Learning

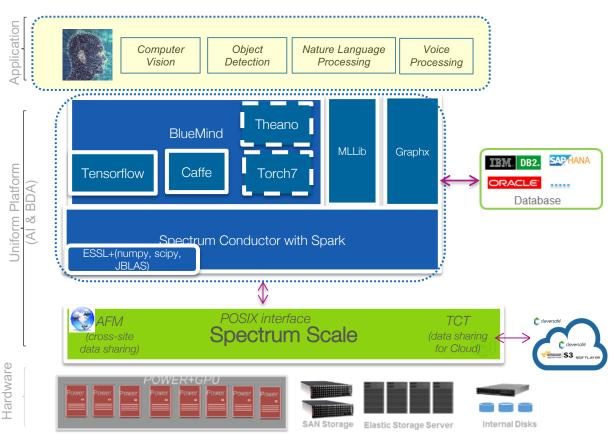
- SAN or internal SSD disks for entry level customers
- GSxS/SSD could be for customers that have large data size or whose data will increase in the near future

#### Advantages from Spectrum Scale

- ✓ Data read/write in parallel for performance
- ✓ Access the same data from all nodes
- ✓ SSD/NVMe

#### **Requirements for Spectrum Scale**

- $\checkmark$  High IOPS for small IO with low cost
- Take LROC to prefetch and cache the training data
- Workloads read the training data multiple times from local LROC(IO read acceleration)

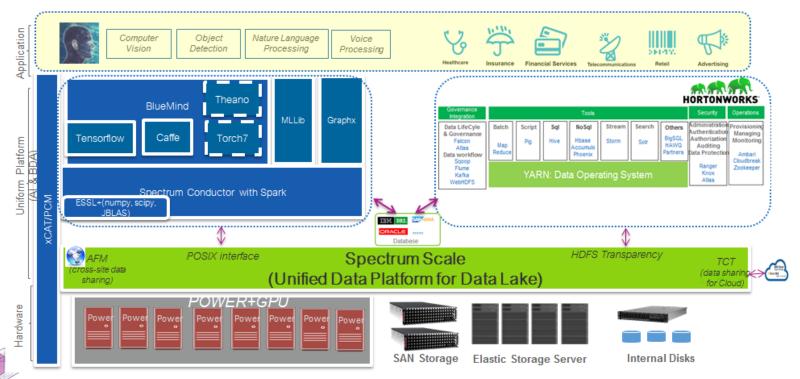




### Spectrum Scale for Cognitive and BD&A Solution

Spectrum Scale Key Advantages:

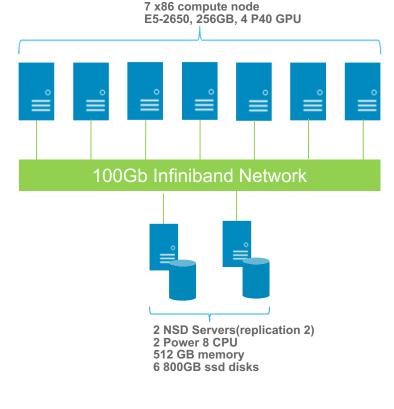
- · Support long-term rapid increasing big data with extremely scaling for file system
- · Fast analytics results from in-place analytics without data movement
- Easy maintenance from centralized storage management for multiple Hadoop cluster
- Support internal disk based for entry level customer(less than 100TB data size) and scale to PB level in ESS



# **Customer Case Sharing**

### Case 1: Spectrum Scale Services Caffe

- Deep Learning Configurations
  - Take Caffe framework in the PoC
  - 16KB ~ 100KB per picture
  - 20GB totally(around 1M pictures)
- Spectrum Scale configuration
  - RDMA is not enabled yet
  - Sing SSD IOPS: 80K IOPS for 4K random read; 60K IOPS for 4K random write
- Requirements
  - 7000 pics/s for each single client with keeping GPU at full load
- POC Results
  - 7200 pics/s for single client
  - 50000 pics/s for 7 clients concurrently(only 40% Spectrum Scale disk bandwidth used)
  - Customer is satisfied with the results(They also tried other vendors solutions but cannot meet their performance requirements)

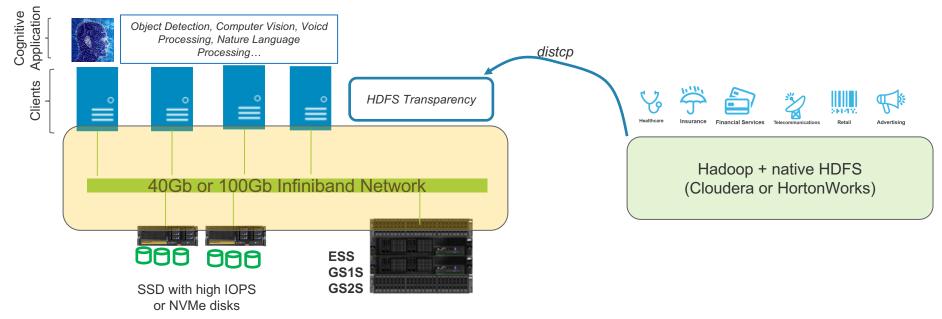




### Case 2: Incremental Data Ingest from Native HDFS

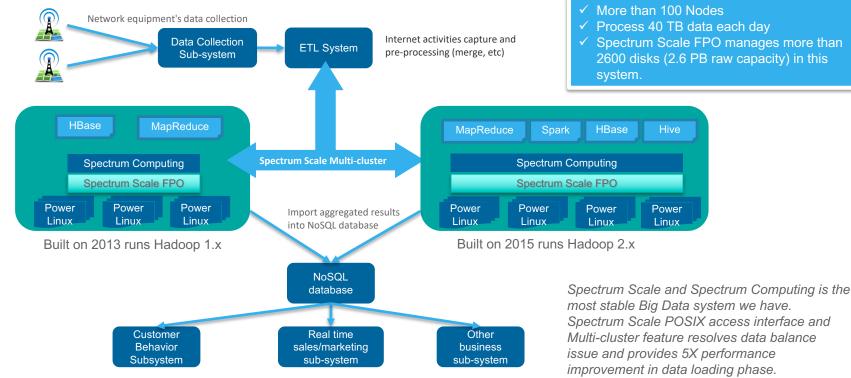
Customer:

- Save data in HDFS
- Take IBM DLI with Spectrum Scale for their cognitive workloads
- · They require a solution to do incremental data ingest from native HDFS
  - Snapshot diff based distcp





### **Bill Analytics System for a Telecom Operators**



-- Client IT Architecture



# Thank You