## Spectrum Scale and Compute

Data Architecture for Big Data, Analytics, and Cognitive Clouds

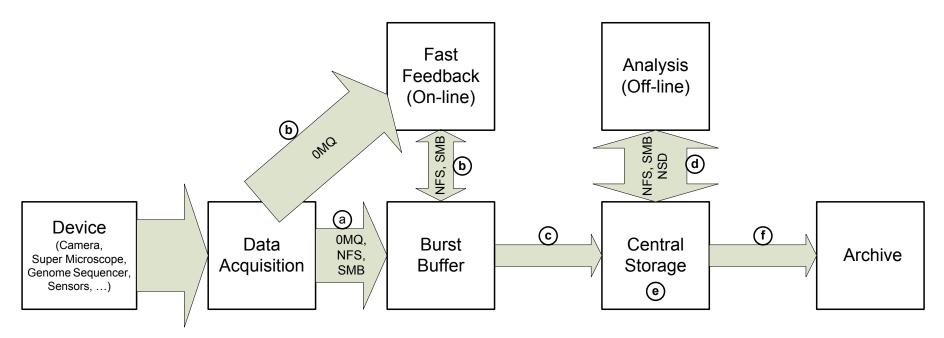
# Agenda

- Data Centric Use Cases
- Data Intensive Science Challenges
- Al Hierarchy of Needs
- Data Architecture

## Spectrum Scale Data Intensive Use Cases

- Instrument Driven Science and Healthcare
  - ALS and CryoEM
  - Radio Astronomy
  - Weather
  - Genomics
- Data Driven Engineering
  - Modeling and simulation results (i.e. visualization)
  - Sensor data analysis
  - Financial market data analysis and reporting
  - Supply chain efficiency
- Big Data
  - Market Insights
  - Operational efficiency (including IT)
- Cognitive
  - Personalized Medicine
  - Autonomic Driving
  - Natural Language Processing

### Typical workflow



a) Real-time data ingest (data acquisition)

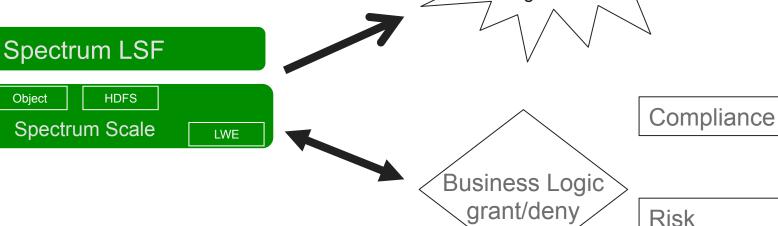
- b) Visualization and near real-time analysis (online processing)
- c) Data movement from Burst Buffer to Central Storage
- d) Deep analysis (offline processing)
- e) Data management of Central Storage
- f) Long-term data archiving

Note: User/Scientists need access to data during each stage of the workflow.

Note: The workflow distills data to the relevant insight by increasing the ratio of content/data (=Pipeline of forgetting the unimportant).

## Data and Workflow Challenges

- Provenance
  - Reproducibility
    - Scientific Workflows: transformations, interpretations, analyses
  - Lineage
    - Origin
    - Ownership
    - Usage
- Governance
  - Compliance, Retention, Data Integrity
  - Legal Hold, Defensible Disposal
- Audit Logging and Intrusion Detection



Record Workload,

Provenance, and

Assessment

Lineage



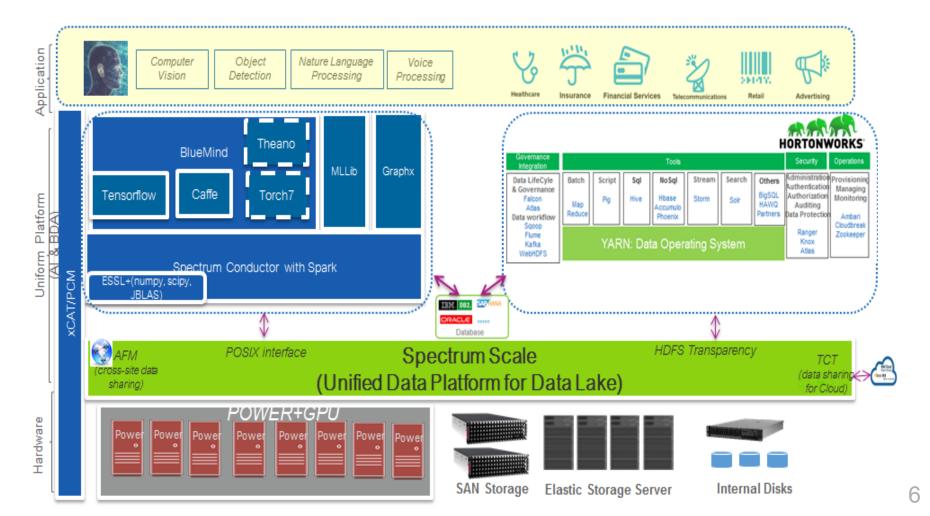
Object

POSIX

#### Spectrum Scale Solution for Cognitive and BDA (Best Practices)

Solution Key Values:

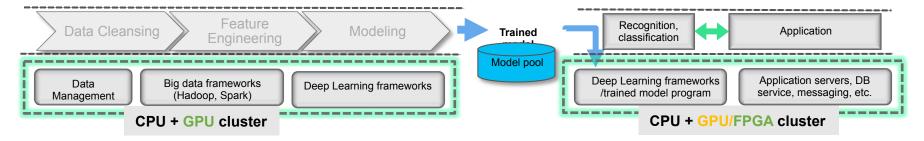
- · Support long-term rapid increasing big data with extreme scaling for file system
- · Fast analytics results from in-place analytics without data movement
- Easy maintenance from centralized storage management for multiple Hadoop cluster
- Support internal disk based for entry level customer(less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



### Cognitive Workloads with Spectrum Scale Data Ocean

#### Training (Research/Development)

#### Inference (Deployment/Production)



Data Hierarchy in Deep Learning		Phase	Tools and Applications
Al/DL Al/DL A/B Testing, ML Algorithms Analytics, Segment, Aggregates, Features, Training Data Cleaning, Anomaly Detection, Preparing Instrumentation, Logging, Sensors, External Data, User Generated Content	Deep Learning	Deep Learning	<ul> <li>Frameworks: TensorFlow(Apache), Caffe(BSD), Torch(BSD), Theano(free), CNTK(free), Neon(Apache)</li> <li><i>IDE</i>: *IBM Power Al Enterprise/CwS with deep learning, Nvidia Digits, *Watson</li> <li>Spectrum Scale: POSIX for Power Al; need to evaluate performance</li> </ul>
	Transform/ Prepare	Transform/Prepare	Machine Learning: IBM SPSS, *IBM DSX, SAS Spectrum Scale: SPSS works over GPFS POSIX/HDFS; SAS works over GPFS POSIX
	Move/Store Collect	Move/Store	Platform: Hadoop, Spark, POSIX/NFS/SMB         ETL: *Talend         Spectrum Scale: POSIX/HDFS interface; need to evaluate performance
IBM Confidential Ref: Monica Rogati – Al Hierarchy of Needs		Collect	Variable data ingestion end devices

