

Video Analytics in the Big & Fast Streaming Data Era

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Strata + HADOOP
CONFERENCE WORLD
TOOLS AND TECHNIQUES THAT MAKE DATA WORK

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Video Analytics Trends

1. City-wide surveillance deployments
2. Increasing camera resolutions
3. Structured data: Correlation & Smart cameras
4. Mobile handsets as sensors
5. Advanced Video Content Analysis (VCA)



Motivations

- Video investigation in action
 - Hunting Criminal Zhou Kehua, China, 2012
 - Boston Marathon Bombing, USA, 2013
- Traditional approaches do not work
 - Challenges
 - Status quo

Hunting Criminal Zhou in 2012

- 1000+ Policemen X 2 months.
- 300TB Surveillance videos
 - Equivalently **830,000 movies.** ☹
- Collecting the local video storages was challenging.
- Video analytics:
 - Face detection & recognition,
 - Video summarization, indexing , retrieval



Boston Marathon Bombing in 2013

- 120 FBI Agents sorted through 13,000+ videos (10 TB's).
- Crowd sourcing videos on Internet examined by thousands of volunteers.
- Video analytics: “to compress video, pinpoint areas of interest, look for anomalies and find relevant details...”



More on Industrial Internet

Deep Water Horizon oil spill monitoring

- BP oil spill at Mexico Gulf in 2010.
- Deep water cameras recorded the evidence.
- Could be better monitored & prevented w/ analytics.

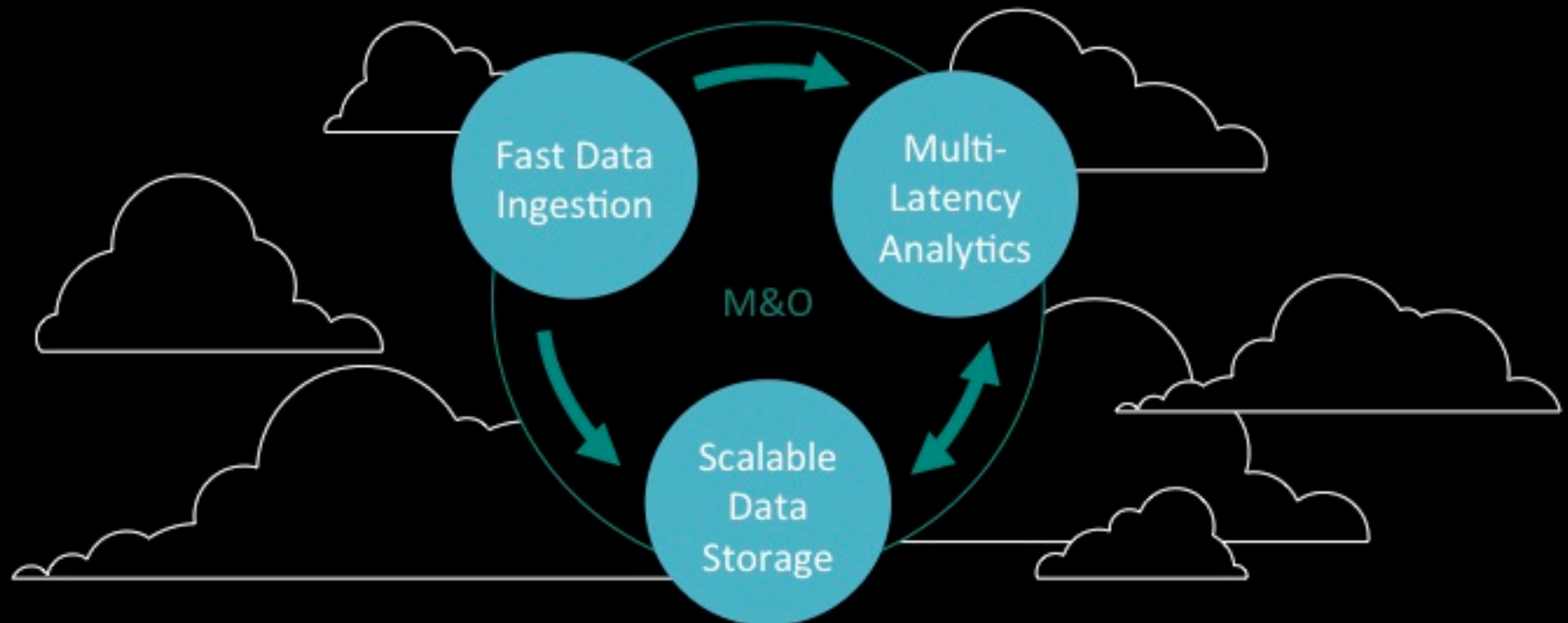


Deep Space Exploration : new international initiative

- SKA : Square Kilometre Array in 2018
- 700 TB / second
- “How big data is fueling a new age in space exploration” – VentureBeat, Oct 2014.



Challenges in Big & Fast Data Era



Challenges in Video Data

Video Stream Formats

- Different real time protocols: MJPEG, H264, MPEG, RTSP, Proprietary, etc.
- Various bit rates consist of streaming packets.
- 100KB – MB/s.

Video File Formats

- Different compression standards: H264, MPEG, Proprietary, etc.
- Large volume files.
- GB - TB



The Tower of Babel -- Pieter Bruegel, 1563

Status Quo

Program Director, National Science Foundation, May 2014:

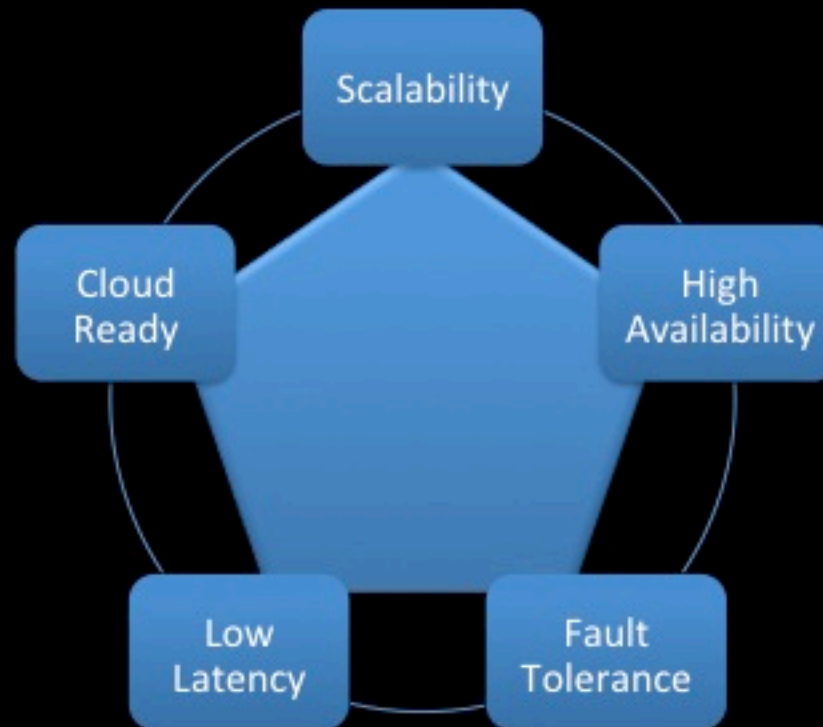
"(In Video Analytics,) People are working hard, but we are not really there yet.

Although the current applications have proven useful, today's technology has significant limitations.

*Government, academic and industry researchers confront a number of challenges that range from exploring ways to more efficiently comb through huge stores of video -- essentially a **big-data problem** ...*

*Furthermore, developments in software have **failed to keep pace with the explosion in video camera hardware.**"*

Desired Platform Features



What it could look like...



CROWDSOURCING

= PROBLEM SOLVED!



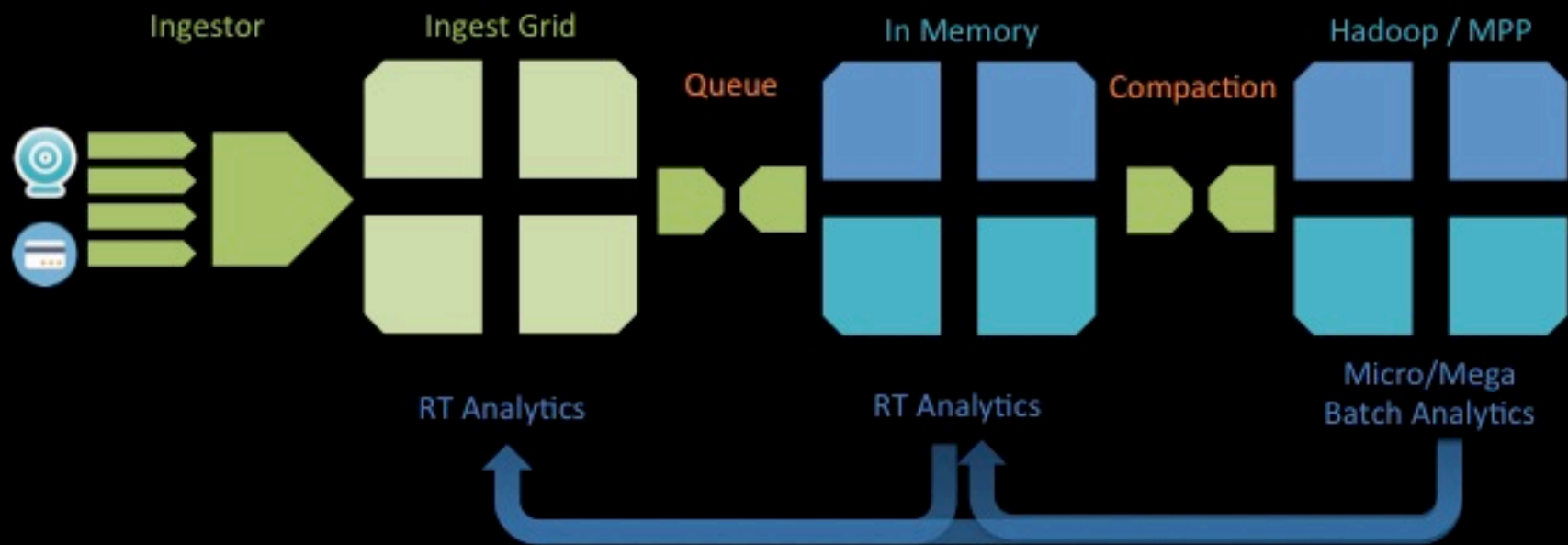
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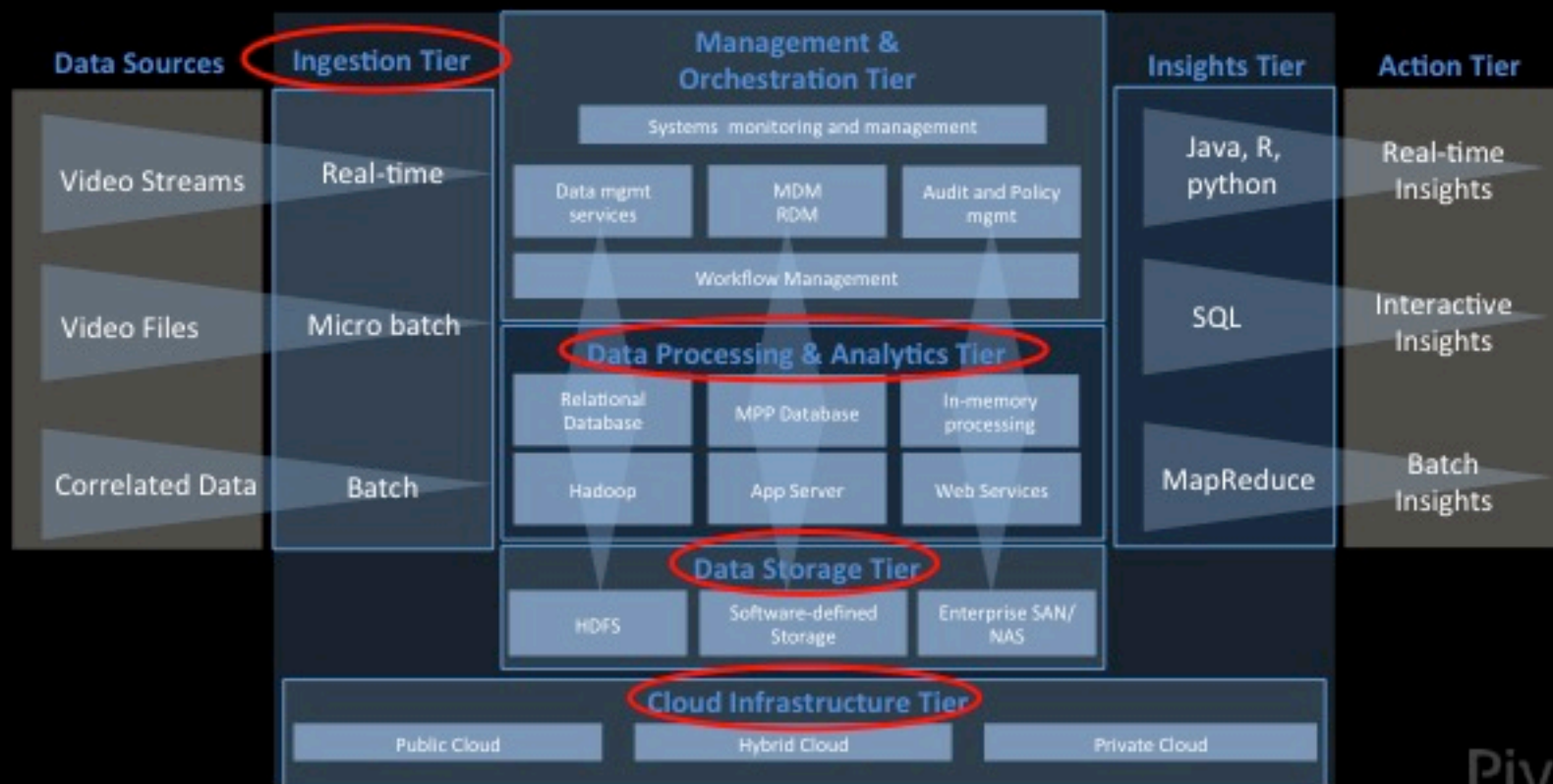
Introducing **Video Analytics Data Lake**

Disruptive Technology for Video Analytics Platform

VADL Conceptual Architecture

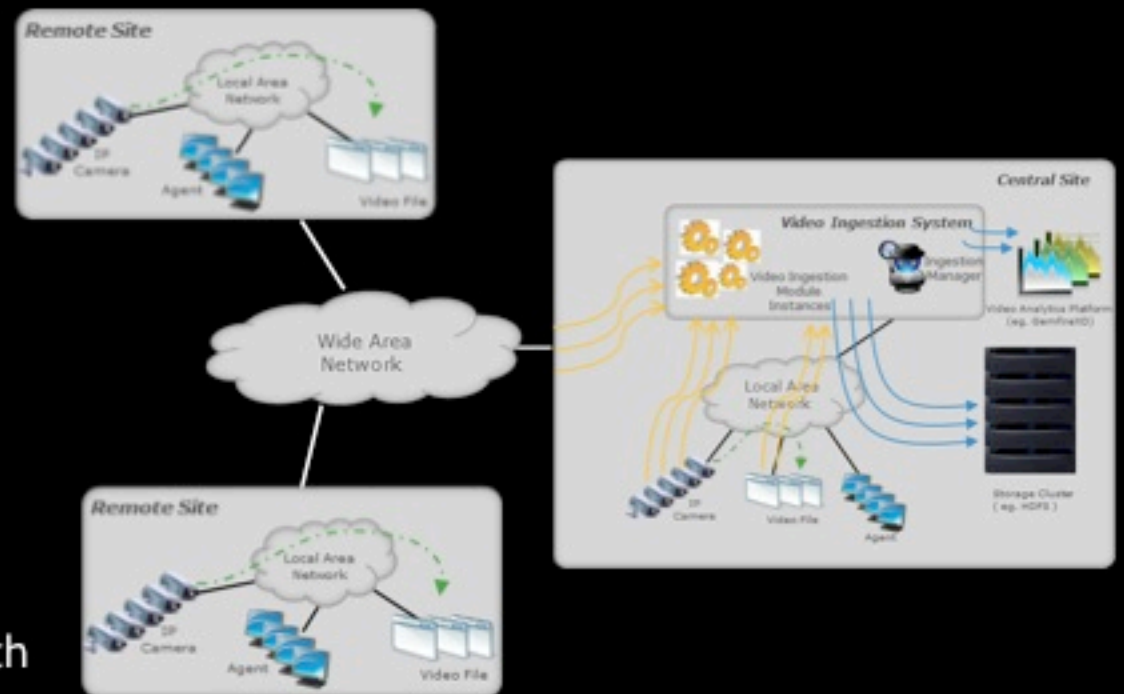


VADL Building Blocks

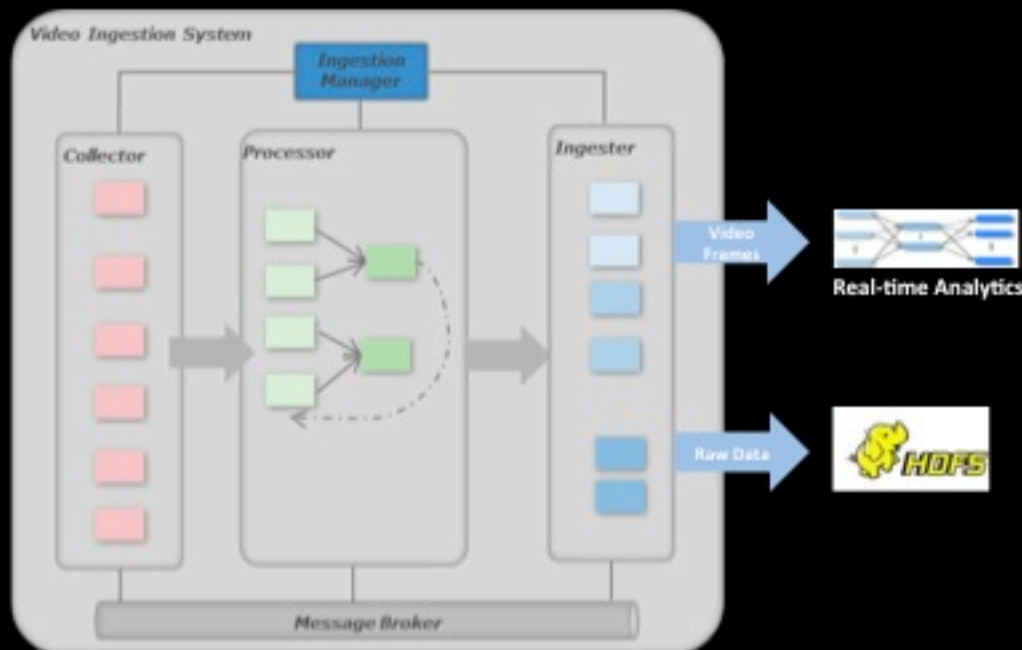


Video Data Ingestion

- Video Stream
 - From campus camera
 - Ingested into data storage
 - Pipelined to real-time analytics
- Video File
 - From remote site
 - Ingested into data storage
 - For batched analytics
- 3rd-Party Data
 - For correlated analytics with video data



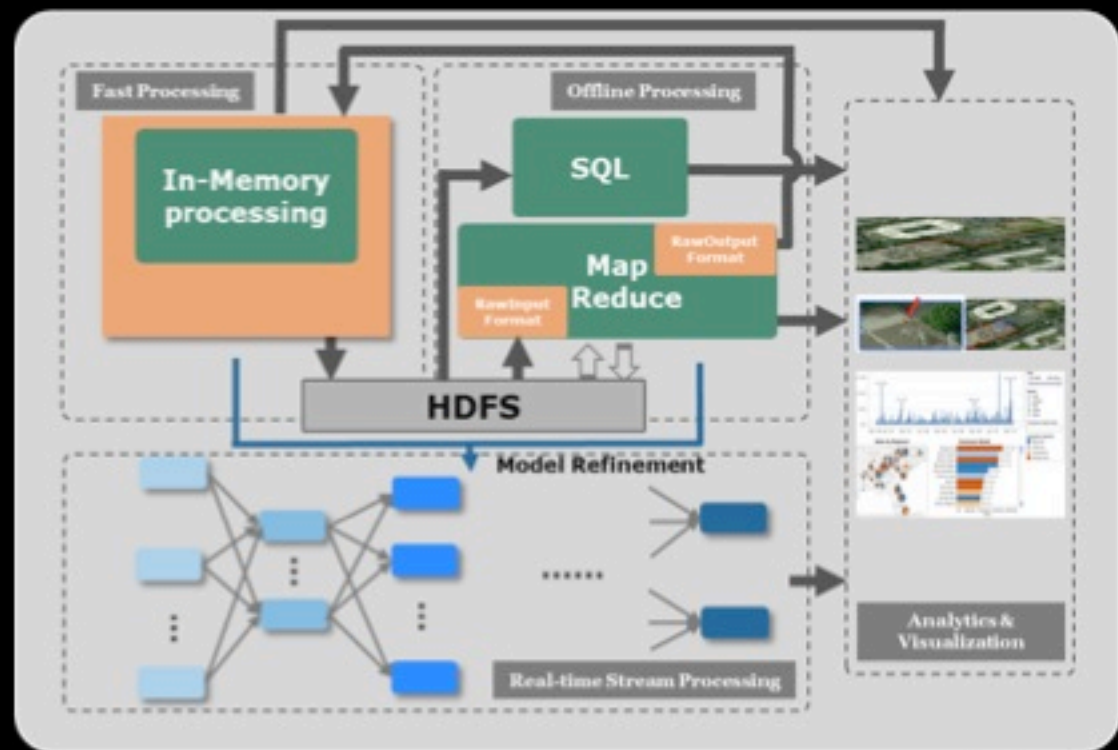
Ingestion System Design



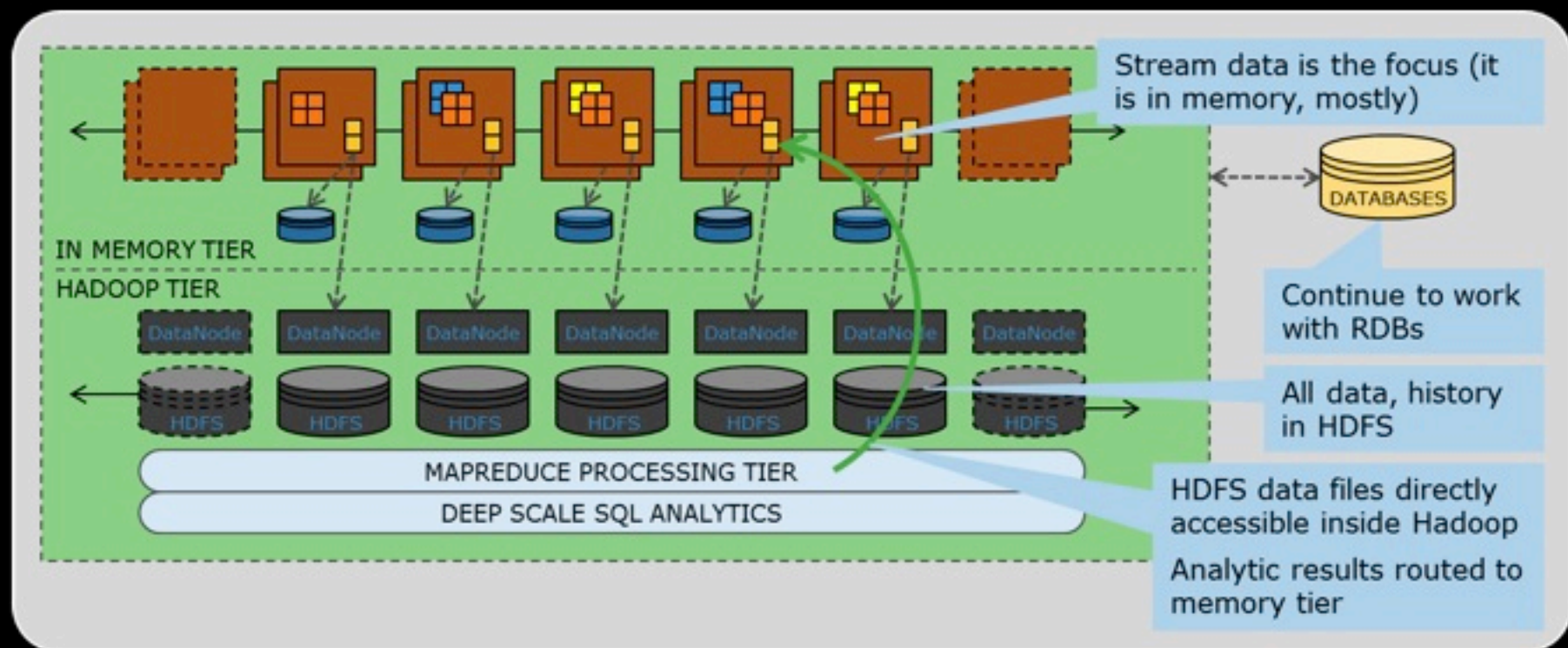
- Ingestion Modules
 - Collector
 - Processor
 - Ingestor
- Message Broker
 - Data communication between modules via messaging middleware (e.g. RabbitMQ)
- Ingestion Manager
 - Topology management
 - Module registration
 - Job Launching
 - HA, load balancing, fault tolerance, etc

Multi-Latency Video Analytics

- Real-Time Stream Processing
 - Sub-second
 - In memory
- Fast Processing
 - Seconds to minutes
 - In memory
- Offline Processing
 - Minutes to hours
 - Persistent data storage
- Analytics Visualization
 - Actionable insights



Scalable Data Storage



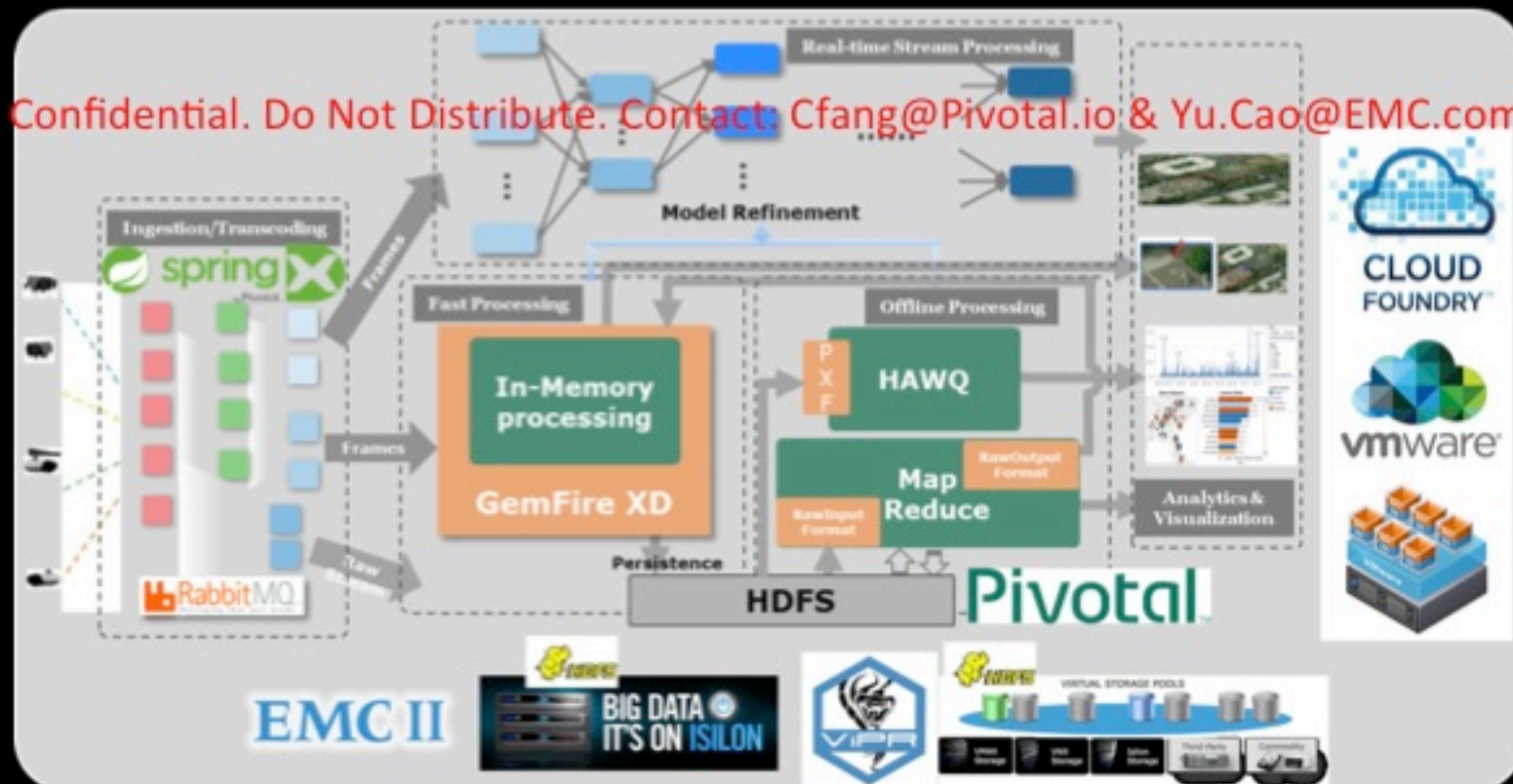
Up in the Cloud

- Provision and auto-scale the computing clusters with IaaS like VMware or OpenStack
- Leverage cloud storage to share or archive non-critical data
- Deploy VCA apps at PaaS like Cloud Foundry to enable Internet and mobile access
- VCA-as-a-Service



An EVP-Powered VADL Solution

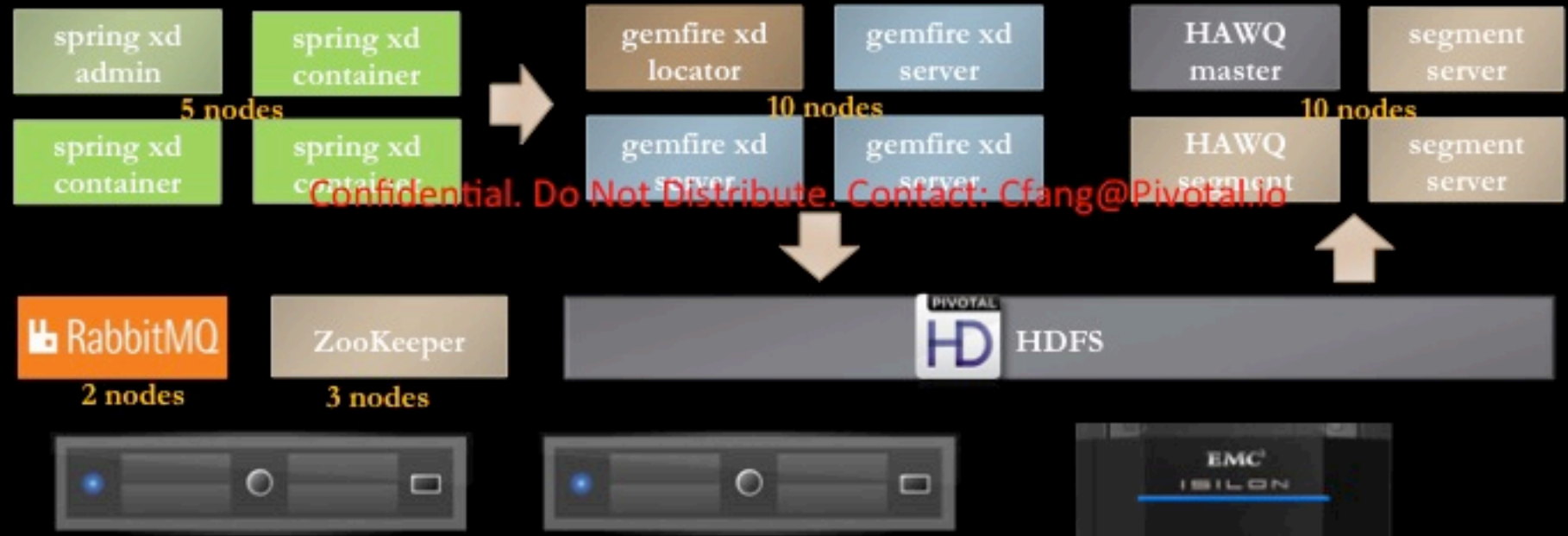
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Prototype Deployment



Gemfire XD



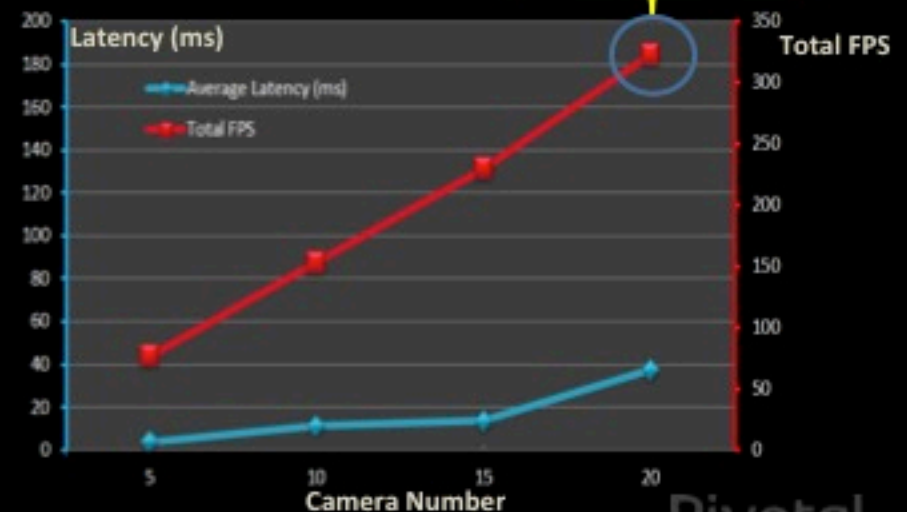
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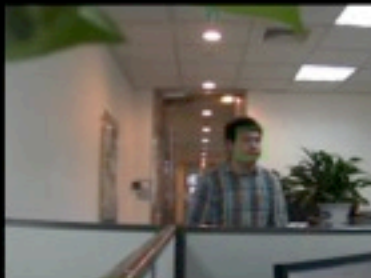
Performance Benchmark Test

- Increment the number of ingested camera streams, until the server stresses out:
 - Average FPS of each camera is about 15, frame size 640*480
 - Server hardware:
 - 8 * 4-core 2.40GHz Intel(R) Xeon(R) CPU
 - 16G memory
- Each GFXD server ingests video streams from multiple cameras, up to 20
 - Low end-to-end latency (<50ms)
 - Without loss of frame
- Overall performance is linearly scalable to cluster size



Use Case: Face Detection & Traffic Monitoring

- Face Detection in real time.
- Face Tracking via motion prediction & Adaptive Online Machine Learning to lower down false positives.
- Automatically monitor traffic



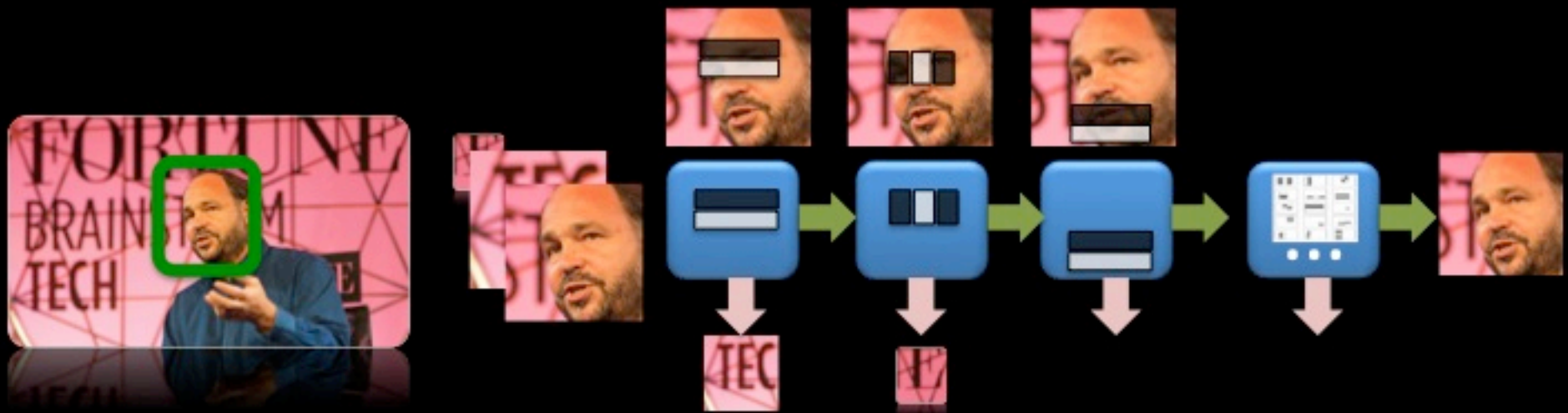
Demo Video

Demonstration:

Face Detection & Traffic Monitoring

Face Detection in Real-Time Stream

- AdaBoost $w_{t+1,i} = w_{t,i} \beta_t^{1-e_i}$, $\beta_t = \frac{e_t}{1-e_t}$ + Cascaded Trees + 160K Haar-Like features *
- Multi-stage Ensemble classifiers promptly reject 99%+ of sub-images
- Real time detection : ~150 ms / frame in Java



Use Case: Real-time Daylight World Map



- Real-time Streaming **15** live video feeds distributed worldwide.
- **Dominant color extraction** for each video: 50ms/frame.
- Real-time Daylight World Map Visualization App on Cloud Foundry.
- Next will demo the results for videos taken in 24 hours.

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Video demo



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For Data Scientists...

Technology Enablement for Data Science

- Batch processing
 - Off line training : MapReduce ...
- Sequential processing
 - Online machine learning : SpringXD, GemfireXD ...
 - Object Tracking, Online anomaly detection, etc.

Towards the Digital Brain & Beyond

VADL is another attempt towards Digital Brain

- Kaushik Das, *"Data Science and Smart Systems: Creating the Digital Brain"*, Strata, Feb 2014, Santa Clara.



Strata

Data Science inspired by Brain Science :

- **This is just the beginning**
- *Deep Learning: The Deep End of Deep Learning*
- *Incognito* : What is in a conscious mind?
- *On Intelligence*: Human brains are constantly doing Prediction & Anomaly Detection.



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Q/A?

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