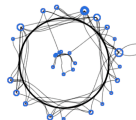


data-driven pipelines high performance computing machine and deep learning relationship and graph analytics streaming data analytics iot technology consulting high performance analytics architecture learned and intelligent systems technology vision, strategy, and practice blockchain and smart contracts data ontology and design data-driven pipelines high performance computing machine and deep learning relationship and graph analytics streaming data analytics iot technology consulting high performance analytics architecture infrastructure intelligent systems real-time event-driven systems blockchain and smart contracts data ontology and design data-driven pipelines high performance computing machine and deep learning relationship and graph analytics streaming data analytics iot technology consulting high performance analytics architecture infrastructure intelligent systems event-driven real-time analytics blockchain and smart contracts data ontology and design data-driven pipelines high performance computing machine messaging topologies relationship and graph analytics streaming data analytics iot technology consulting high performance analytics architecture infrastructure intelligent systems real-time event-driven systems blockchain and smart contracts data ontology and design data-driven pipelines high performance computing machine and deep learning relationship and graph analytics streaming data analytics iot technology consulting high performance analytics architecture infrastructure intelligent systems real-time event-driven systems blockchain and



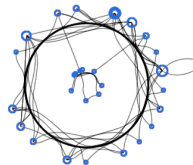
Perspectives On HPC And Enterprise High Performance Data Analytics

Arno Kolster

May 10, 2017



Providentia Worldwide



A Little Bit About Me...

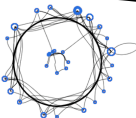
30 years in database architecture and infrastructure operations

Until recently, at PayPal for 13 years with a focus on database, analytics and operations architecture

Career interest turned to HPC to solve real time, web scale analytics problems

Co-recipient of IDC HPC Innovation Award at SC 12 and SC 14 with colleague Ryan Quick

HPC User Forum and Leverage Big Data steering committee member



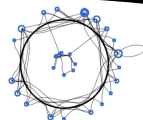






Facebook Admits AI Can't Do Everything Yet and Hires 3,000 Editors

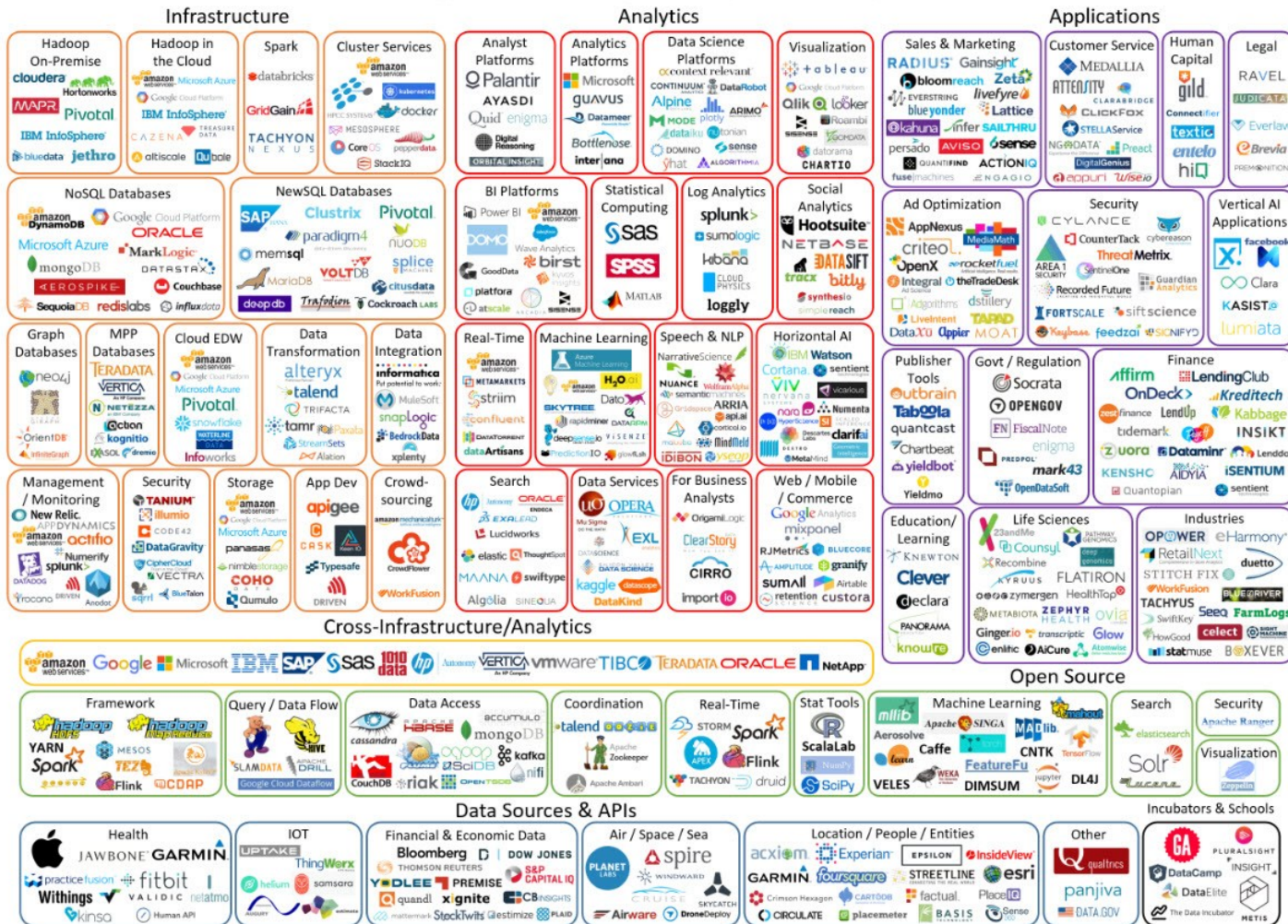
Facebook





LIFE

Big Data Landscape 2016 (Version 3.0)



So What's Happened?

Huge amount of data being generated by humans and machines

Rich social media and data streams

Sensor data from IoT, medical imaging, Fit-Bits, Apple Watch

Autonomous vehicle data

Smart cities

Need to analyze data faster for businesses to stay competitive

Customer insights

Real time marketing

Fraud detection



Sent more than
500 million tweets



Posted over
70 million Instagram photos



Generated more than
4.3 billion Facebook posts



Uploaded more than
4 million hours of YouTube videos



Sent more than
6 billion emails

A FINELY TUNED ECOSYSTEM



Ecosystem: a system involving the interactions between a community of living organisms in a particular area and its nonliving environment

Why The Convergence Of Enterprise And HPC?

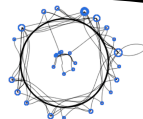
Enterprise is analyzing increasing larger data sets rivaling HPC centers

Hardware and software within budgets to tackle some of these problems

Open source and commercial product availability

Lower latency, real time analytics driven by the need to make quicker decisions

Trend toward Machine Learning, Deep Learning and AI fueling innovation and First-To-Market



Enterprise Needs HPC

Ability to use shared memory for large data sets

Apply modeling and simulation to traditional reactive analytics

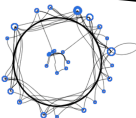
Global, parallel file systems can be accessed from disparate systems

RDMA and distributed memory access

Leverage high speed interconnects to move data faster

Capability of running atomic jobs in parallel

Rich collaboration history



HPC Needs Enterprise

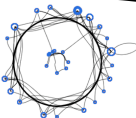
Enterprise has resiliency and availability 'down'

Checkpointing and restarting are antiquated standards and better alternatives exist

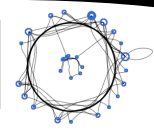
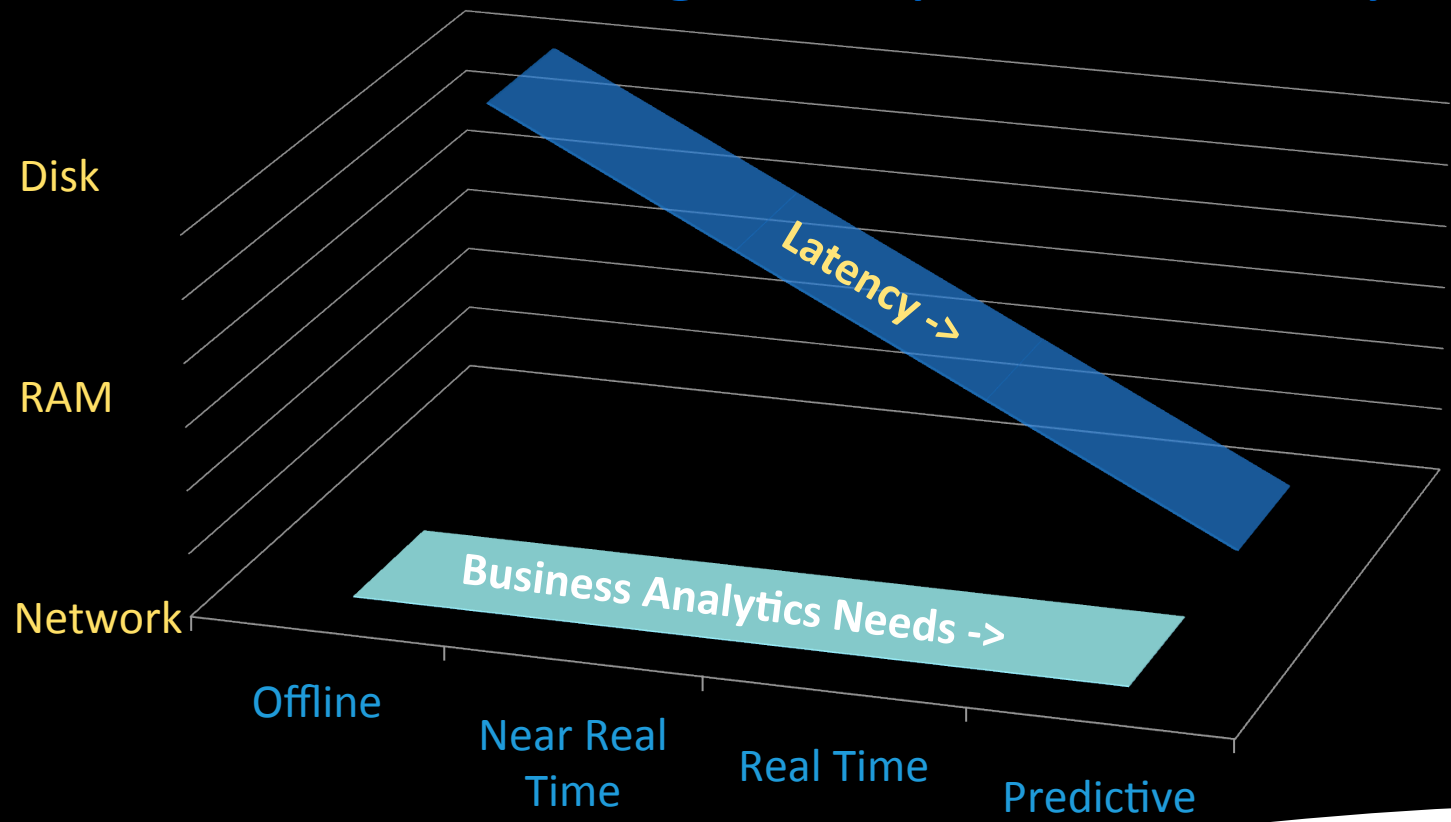
Open source and shift in development paradigms provide much faster time-to-market

Distributed computing i.e. cloud is now second nature

Enterprises by their nature create new markets which provide opportunity to HPC vendors

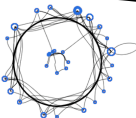


Business Is Driving The Speed Of Analytics



Let's Talk About Real Time

“Complex Event Processing as Digital Signals”



Real-Time Analogy

Everyone likes to go to concerts...



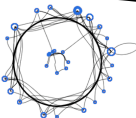
The Concert Experience

You're at the concert listening to your favorite piece of music or song.

You're really enjoying yourself, you've had a glass of wine, you're feeling the groove with the musicians...

Suddenly you hear a bad note.

But it's a concert, the show goes on, you ignore it and you have another sip of wine.



The Concert Experience

But...what just happened?

You analyzed data in real time.

Not a second later,
not a minute later,
not a day later...

But at the instant the event occurred.

You used predictive models to look for anomalies in the event stream, in **REAL TIME** .

So... how do we do that?

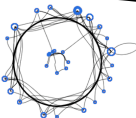
Faster



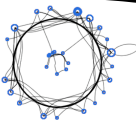
Cheaper



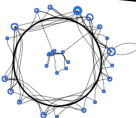
Greener



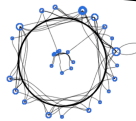
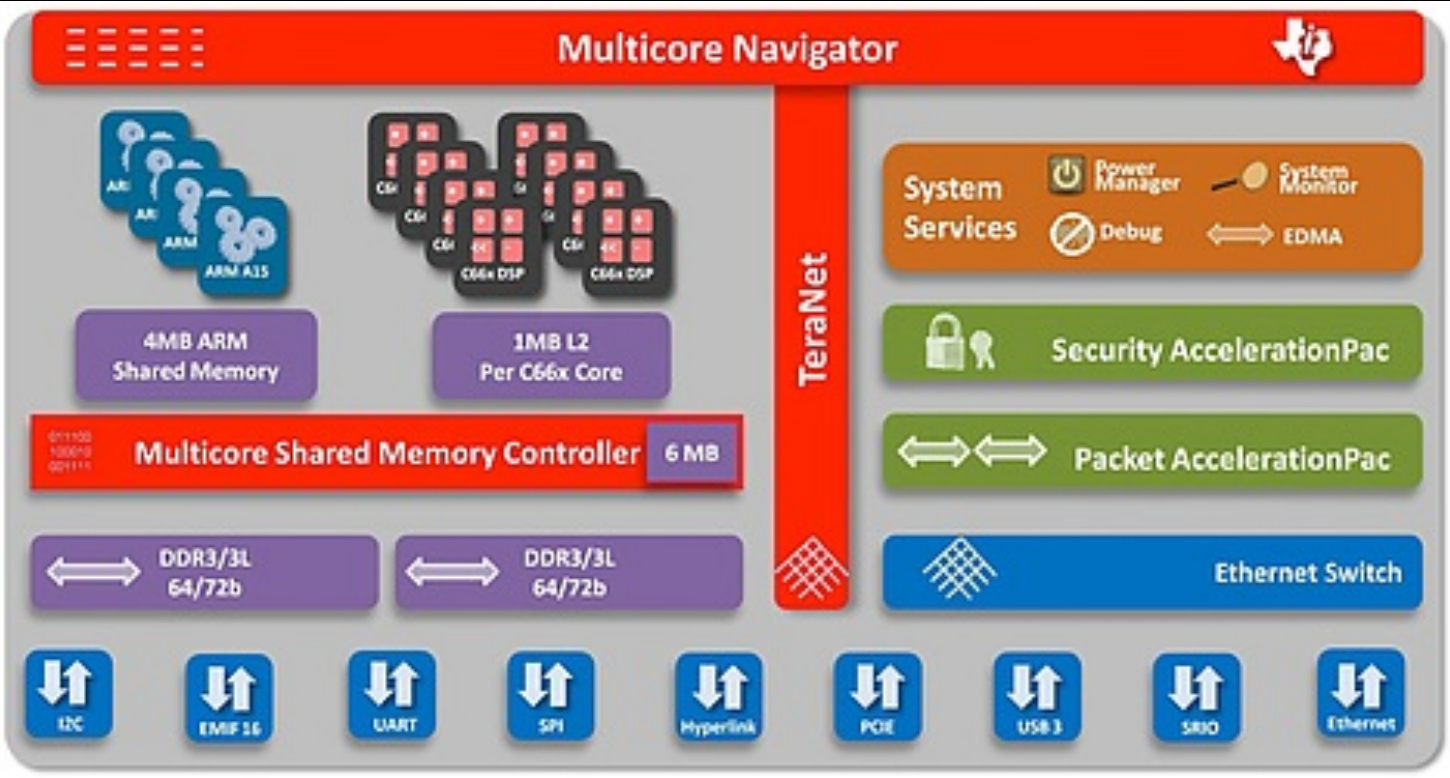
You just add **little** machines to the **BIG** ones...



Innovative idea evolved with HPE's m800 cartridge..



... HPC in a SoC ...



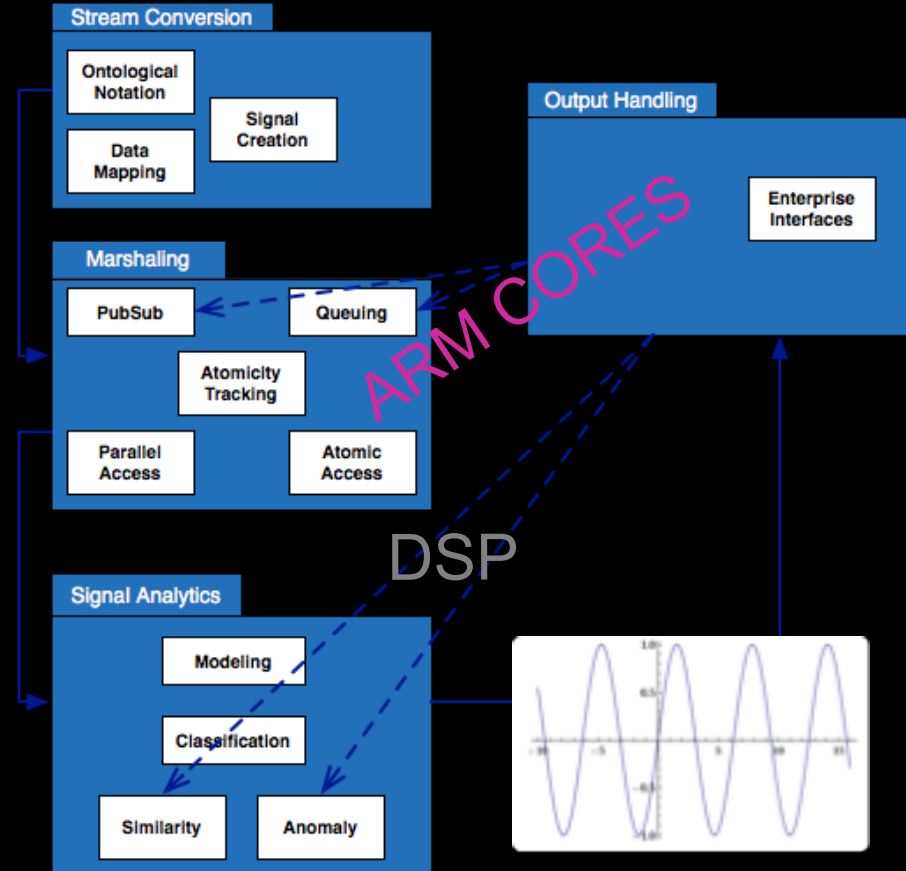
Complex Event Processing as Digital Signals

Familiar Systems Integration (ARM)

- Linux for general purpose work
 - integrating with enterprise systems (databases, marshaling, command & control)
 - short development learning curve (python, java, openCL, openMPI)

Efficient, Real-Time Parallel Processing

- Implement signal analysis in hardware
 - solve encoding, marshaling, atomicity
 - apply both global shared memory and scale-out process best practices
 - leverage cross-platform development to decrease ramp-up and testing time (openCL)



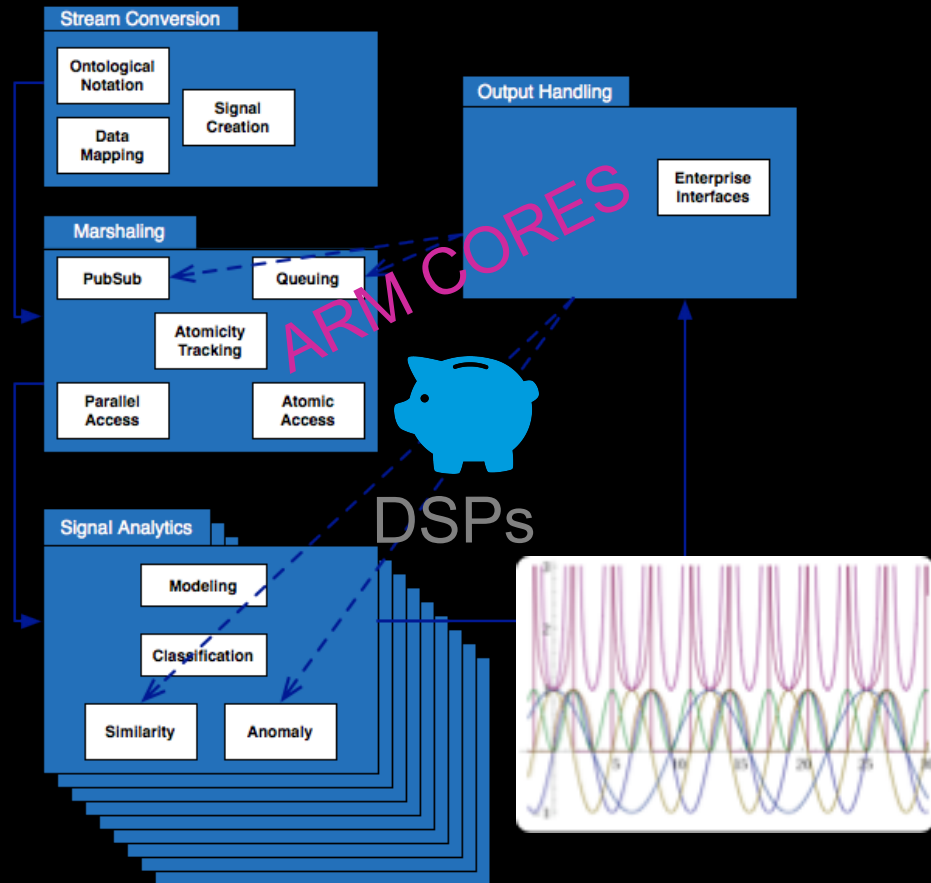
Complex Event Processing as Digital Signals

Parallel, True Real-Time Analytics

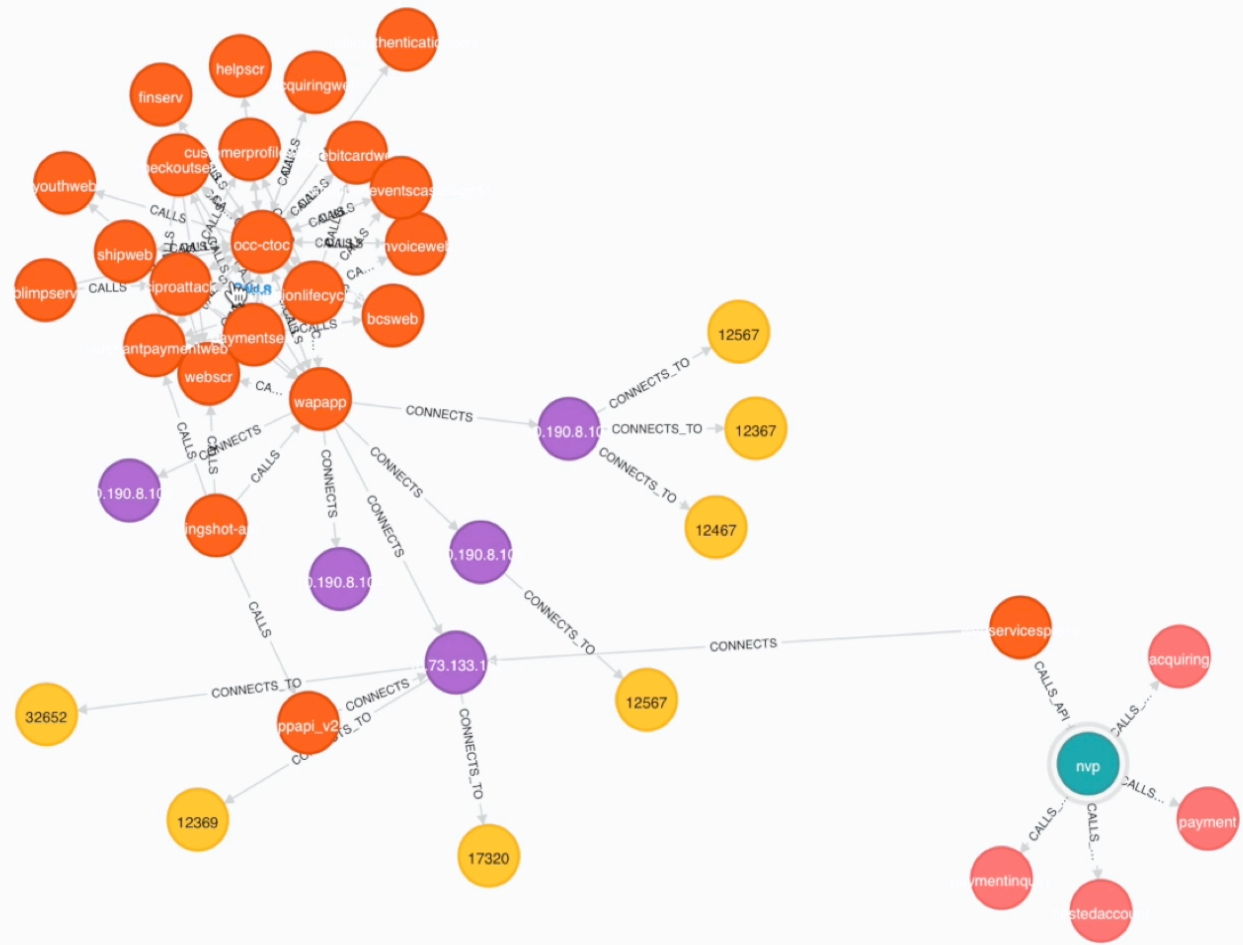
- Multiple filters/atomic event stream
- Multiple streams/filter
- Multiple filters/multiple streams
- Pattern recognition (outliers, clusters, frequency matrices, etc)
- Rich library of functions (notch/high pass/low pass filters, DFT/FFT, z-,bilinear- transform, etc.)

HPC and Enterprise Best Practices

- Multicore implementation
- Tiered shared memory and queuing
- High-speed, low-latency transports inter/intra SoC
- Support for common development libraries and standards (openCL, openMP/MPI)
- Efficient, low-power solutions
- (~55W/cartridge (4 SoCs / cartridge))
- Extreme performance (11.2 GF/watt)



- App
- Ip
- Api
- Method
- Port



Use Case: Flow Analysis/Graphing

Cray Urika-GX



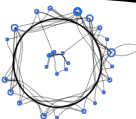
Cray Graph Engine (CGE) - semantic graph, RDF

Supercomputer technology - large shared memory,
Aries fabric

Multi-node clusters up to 48 nodes, 1728 cores

Built-in open source frameworks for Big Data – Hadoop,
Spark, Mesos, OpenStack

Bridging the gap between HPC and enterprise workloads



Graph Analytics

Why are graphs so cool?

They represent relationships between entities

Easily traversed with a variety of different algorithms

Visualization of a graph shows areas of interest quickly

Allow proper modeling and simulation of populations, behaviors, paths flows, etc.

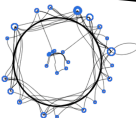
Graph Analytics Challenges

Ingestion rates still not suitable for some real time applications

Temporal analysis of networks still extremely difficult at scale

Disparate offline data sets need to be 'massaged' into the existing graph

Augmentation of retrieved data difficult, requires subsequent processing



The background features a light blue gradient with a network diagram. A central white cloud contains various icons representing data and technology, such as a cursor, a house, a location pin, a speech bubble, a magnifying glass, a calendar, a database, a server, a lightbulb, a gear, and a musical note. This central cloud is connected by a web of white lines to numerous other circular icons scattered across the frame. These icons include a blender, headphones, a video camera, a printer, a microphone, a washing machine, a television, a car, a refrigerator, a lamp, a mobile phone, a camera, a clock, a lightbulb, a drill, a computer monitor, a watch, a desk lamp, a DNA helix, a microwave, a fan, a laptop, a smartphone, a telephone, and a server tower.

The Connected World

Or

We Thought We Had Data Challenges Today....

Just Wait.

Internet of More Stuff

Estimates range from 50 to 100 billion data generating devices by 2021

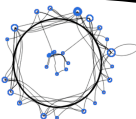
Distributed networks of micro services and imbedded devices

Connected devices - phones, homes, humans and toaster ovens

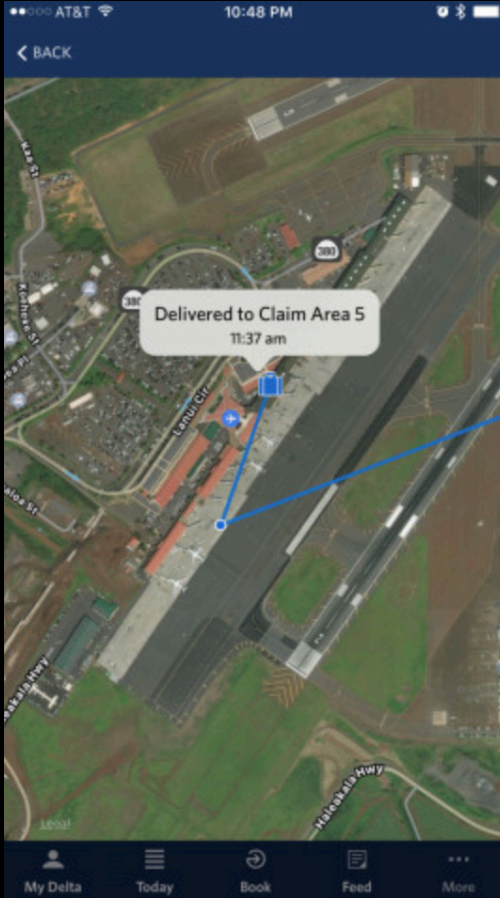
Security vulnerabilities and privacy concerns abound

Standards for wireless protocols and data format specs are still in their infancy

Will we aggregate and analyze at the edge? **Most definitely.**







Delta's Checked Bag Tracker



RFID baggage tags

180 Million* passengers annually

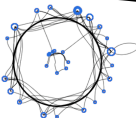
50% of passengers check 1 bag – 90M

50% use the app – 45M

check-in, on plane, off plane, bag claim - 140M recs

~4.5 recs/sec

*<http://news.delta.com/corporate-stats-and-facts>



What Does This Allow Delta To Do?

Minimize bag loss

Customer now able to feel more confident (or not) about bag location

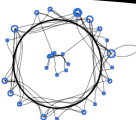
Data Analytics With Enterprise:

Number of bags per passenger / per flight

Near real time reporting on bag delivery

Predictive ground crew resource management

Long term trend analysis



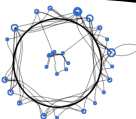
But Add In HPC...

Modeling and simulation of bag handling and impact on carousels, cargo holds

Provide bag manufacturers with information on wear and tear areas based upon analysis of bag images

Cargo loading can be based upon image data for size, shape and cargo hold volume

Discover a potential correlated attack on multiple flights based on similarity of bags



Governance of Data



This will become increasingly important

Curation and provenance of data

Where and how is it being stored?

Where did it come from?

Data integrity, privacy and security

Audit trail logging

Meta-data requirements

Barriers To Acceptance/Social Challenges

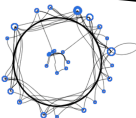
Misunderstanding of what HPC is.

“I just bought a cloud... why do I need to buy this HPC thing now?”

People are comfortable with what they know and very resistant to change.

No ability to apply new technology to existing problems in a different way.
Limited technical vision.

In terms of analytics – shortage of analysts with technical skills worldwide.





What's Next In Analytics?

Paradigm shift in graph technologies to enable real time analytics.

The wire as a data space. "Analytics on data in motion."

Preality. "Pre-reality" – encompassing predictive analytics with real time event processing. ML is driving AI.

More acceptance of HPC in the enterprise with hybrid architectures.

IoT will further drive analytics on the edge.

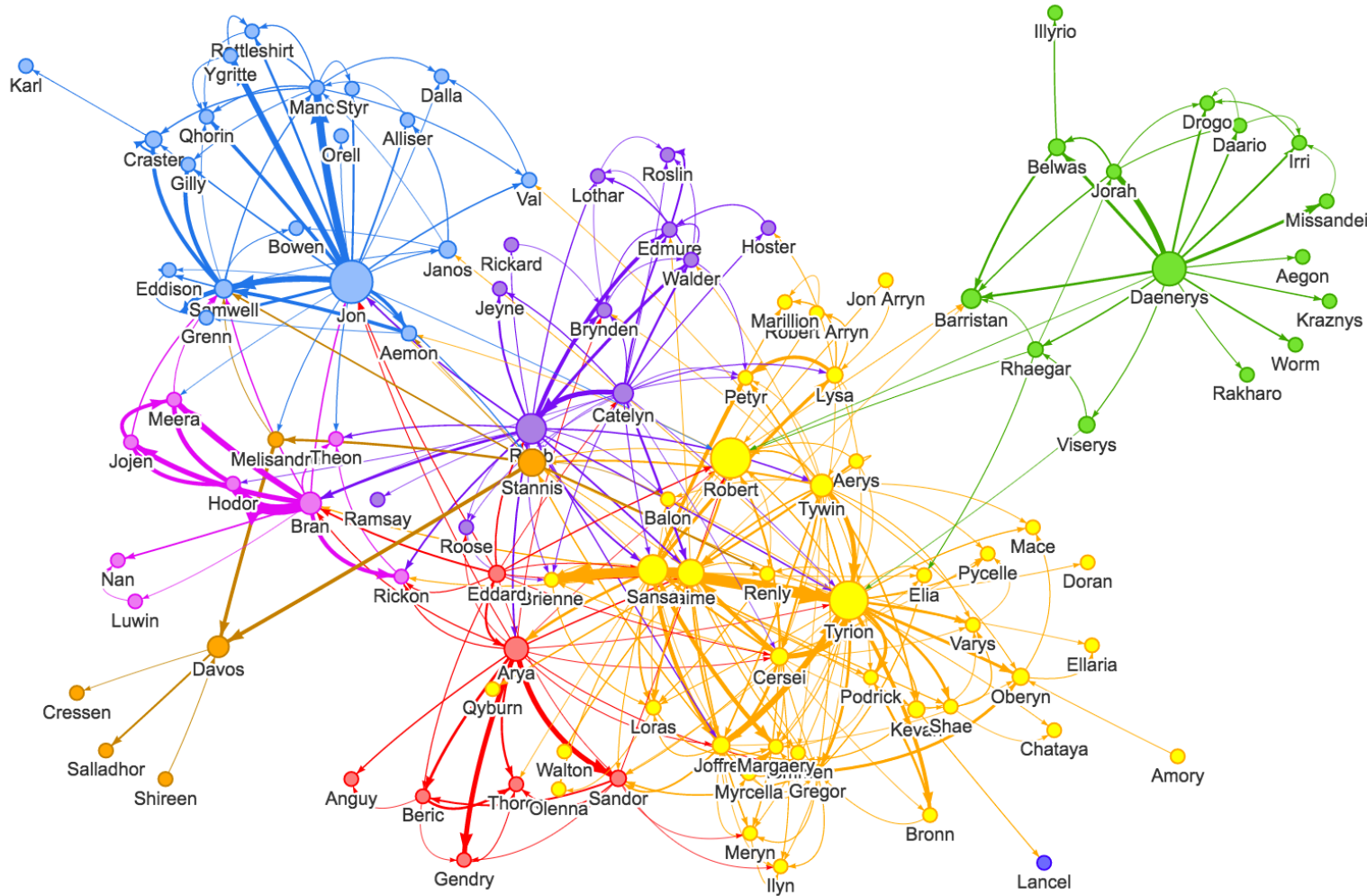
More collaboration between the HPC centers and industry. (NSCI initiative)

Real time streaming
text analytics

Machine learning
Pattern matching

Graph?





Thank you.

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Special Thanks To The **CUG Program Committee** and

David Hancock

Yun (Helen) He

Jim Rogers

