# ExaO: Traffic Optimization for ExaScale Science Applications

Justas Balcas<sup>2</sup>, Greg Bernstein<sup>3</sup>, Haizhou Du<sup>4</sup>, Azher Mughal<sup>1</sup>, Harvey Newman<sup>1</sup>, **Qiao Xiang**<sup>5</sup>, Y. Richard Yang<sup>5</sup>, Jingxuan Zhang<sup>4</sup>

<sup>1</sup> California Institute of Technology, <sup>2</sup> CERN, <sup>3</sup> Grotto Networking,
 <sup>4</sup> Tongji University, <sup>5</sup> Yale University
 March 31st, 2017, IETF98, Chicago

# Next-Gen Integrated Systems for Exascale Science

- Vision: resource in worldwide-distributed environments should be deployed flexibly to meet the demands of exascale science applications.
- Goal:
  - Production deployments of a new class of intelligent, softwaredefined global systems for data-intensive science programs involving a worldwide ensemble of sites and networks.
  - A game-changer for next-generation science data flow orchestration that shapes the future architecture and operational modes of exascale computing facilities.
- **Members**: worldwide multi-organizational collaboration between Caltech, CERN, Tongji University and Yale University.

# Next-Gen Integrated Systems for Exascale Science

#### • Timeline:

- Pre-production deployment in 2017.
- Production deployment in ~2018-19.
- Worldwide deployment of such systems in ~2020-24.

### LHC: Large Hadron Collider



LHC Computing Model Uni x Lab m regional group Uni a Tier 1. CERN Tier 1 .ab a UK USA Tier 0 France The LHC Computing Centre Physics Department Tier2 CERN Uni n Italy Desktop Germany NL Lab c physics group .... Uni y Uni b α.

Figure source: cern.ch

# The Compact Muon Solenoid Computing Model



### **Problem Settings**



Figure source: cern.ch

- A multi-domain data center network
- Heterogeneous resource /system provision
- Various jobs
  - Exascale dataset transfers
  - MapReduce analytics
  - MPI analytics
  - etc.

#### Takeaway from Version 00

- Use ALTO topology services (path-vector and routing state abstraction) to provide on-demand fine-grained network information from different sites/domains.
- Use such information to orchestrate dataset transfer scheduling.
- Prototype demo at SC'16.
- Currently under production development.

### Update in Version 01

- Dataset transfer is only one application in CMS.
- ALTO has the potential to improve the performance of analytic applications.
  - In CMS DC networks, network resource is not always the bottleneck.
- ExaO: a multi-resource orchestrator for CMS applications.

#### Impact on ALTO Working Group

- ExaO is a representative ALTO application in data center networks, which is a major use case of ALTO listed in the WG Charter.
- We expect the success of ExaO to provide key insights and experience in deploying ALTO services and developing ALTO applications.

#### Overview

- Deploy ALTO clients and servers at sites and networks to retrieve endpoint properties and topology information.
- Expand the capability of abstract network element (ane) to provide an abstract view of computing, storage and networking resources.
- Use such views for deep site orchestration among virtualized clusters, storage subsystems and subnets to successfully coschedule CPU, storage and networks.

#### ExaO Architecture



### **Resource Abstraction Agent**

- 1. Orchestrator sends input dataset name and job information to resource locator.
- Locator talks to resource management system to get endpoint addresses of available resources for the job and sends to ALTO client.
- 3. ALTO client issues queries to ALTO server.
  - EPS query to get properties of resource nodes.
  - PV-based query to get properties of ane connecting resource nodes.



#### **Resource Abstraction Agent**

- 4. ALTO server executes the query and send the response to ALTO client.
- 5. ALTO client sends the information of resource nodes and ane to ANE aggregator.
- ANE aggregator encodes the received information into a single ane and sends back to the orchestrator.



#### ANE Aggregator: Provide a One-Big-ANE View

- Motivation: existing resource node abstractions, e.g., HTCondor and YARN, only provide coarse-grained resource information, e.g., # of cores and size of memory.
- Basic idea:
  - Abstract the set of resource nodes and the connecting network available to the job into a single ane.
  - The output ane provides an abstract view of computing, storage and networking resources.
  - Each property of each resource is encoded into a property of ane.
  - Properties from different resources are merged to reduce information overhead and privacy exposure.

#### ANE Aggregator: An Example



- ane.dtbw = min{s.iobw, c.membw, l.bw}
- ane.delay = s.skdelay+l.delay
- Property merging is motivated by ALTO-PV and ALTO-RSA and is crucial for privacy preserving.

### Design Issue: Why ANE Instead of PID?

- The abstraction requested by job is data-oriented and highly dynamic.
- PID is a static abstraction decided only by sites.
- ANE is a dynamic abstraction decided by both sites and applications.
  - Sites: policy, regulation, etc.
  - Applications: dataset name, job properties, preferred resource, etc.

#### Resource Abstraction Agent: An Example

- Job J needs dataset X as input.
- Data center A and B eachh has a copy of X.
- Resource locator in data center
   A finds J would be placed on a different node from X's location.
   Both nodes are in the same rack.



#### Data Center A: EPS Query of Storage Node

Response

#### Request

{

}

```
"endpoints": ["ipv4:10.0.0.1"],
"properties": ["iobw", "skdelay"]
```

```
ł
  "meta" : {
     "dependent-vtags" : [...]
  },
  "endpoint-properties": {
     "ipv4:10.0.0.1" : {
          "iobw": "150",
          "skdelay": "30"
     }
```

#### Data Center A: EPS Query of Computing Node

Response

#### Request

{

}

```
"endpoints": ["ipv4:10.0.0.5"],
"properties": ["membw"]
```

```
{
    "meta":{
        "dependent-vtags":[...]
    },
    "endpoint-properties":{
        "ipv4:10.0.0.5":{
        "membw":"200"
        }
    }
}
```

#### Data Center A: PV-based ECS Query

#### Request

ł

```
"cost-type": {
    "cost-mode": "path-vector",
    "cost-metric": "ane"
},
"endpoints": {
    "srcs":["ipv4:10.0.0.1"],
    "dsts":["ipv4:10.0.0.5"]
}
```

# Response "meta" : { "vtag": [... "query-id":"query\_0"], "dependent-vtags": [...], "cost-type": { "cost-mode": "path-vector", "cost-metric": "ane" }, "endpoint-cost-map": { "ipv4:10.0.0.1" : { "ipv4:10.0.0.5": ["ane:l1"] }

#### Data Center A: PV-based ANE Property Query

#### Request

{

}

```
"query-id": "query_0",
"entities": ["ane:l1"],
"properties": ["bw", "delay"]
```

```
Response
{
    "meta":{
        "dependent-vtags":[...]
    },
    "property-map":{
        "ane:l1":{"bw":"100",
        "delay":"40"}
    }
}
```

#### Data Center A: ANE Aggregator



### Data Center B: ALTO Query + ANE Aggregator

 Resource locator in data center B finds J would be placed on the same node where X is stored.



### Design Issue: Aggregator Inside/Outside ALTO

- ANE property aggregation is motivated by ALTO-RSA.
- How different resources are further encoded into ane is dynamic and application-specific.
- ANE aggregator as an ALTO service?
  - Pro: enrich ALTO's control capability on privacy leakage.
  - Con: ALTO is supposed to be agnostic of applications.

#### Current design

- Resource locator only passes endpoint address to ALTO client.
- ANE aggregator works as an independent module instead of an ALTO service.
- This design is modular.

#### Make and Enforce Resource Orchestration Decisions

- Decisions include
  - where to place analytic processes, and
  - where to transfer/store intermediate/final results.
- Under certain cases, decisions also include copy a hot dataset to another location before placing analytic processes.
- Decision enforcers are implemented on top of current resource management systems in each site.
- ALTO provides fine-grained resource information to improve the performance of these two components.

#### Summary

- ExaO expands the capability of abstract network element (ane) to provide an abstract view of computing, storage and networking resources, which supports the efficient resource orchestration of data-intensive applications.
- We are also exploring the feasibility and benefit of applying costcalendar and flow cost service in supporting next generation science data flow orchestration.

#### Milestones

- Pre-production deployment of ExaO by IETF 100.
- Production deployment by IETF 102-103.

**Backup Slides** 

# CMS Science Network

- Geographically distributed
- Multi-domain
  - Each domain has its own policy.
  - Resource information is private.
- Heterogeneous

### Heterogeneity

- Heterogeneous resources
  - CPU: AMD, Intel (various specs...)
  - Storage: Tape, SATA hard-drive, SSD, etc.
  - Networking: 1/10/40/100Gbps
- Heterogeneous systems
  - File systems: HDFS in US sites, dCache/NFS in European sites, EOS/CephFS at CERN, etc.
  - Schedulers: even Hadoop provides different scheduling policies
- Heterogeneous jobs
  - Dataset transfers
  - MapReduce analytic
  - MPI analytic

#### Example 2: File System Becomes Bottleneck

- A MapReduce job J needs dataset X as input
- X has a copy in site A using HDFS and another copy in site B using NFS
- J should run at site A since HDFS provides a better data throughput, which is a key factor for MapReduce job latency.

#### Example 3: Copy Data First?

- Two jobs J1 and J2 need a dataset X as input.
- X only has one copy at site A and A has abundant resources.
- Scheduling Option 1: executing J1 and J2 sequentially.
- Scheduling Option 2: assign the nodes storing X to J1, and assign nearby nodes to J2.
- Scheduling Option 3: make an additional copy of X to a set of idle nodes in A. J1 and J2 can then run at the same time.
- Which option is better depends on
  - How long would each job run?
  - How large is X? (In other words, how long does it take to make another copy)