# Monitoring the Dynamic Resource Usage of Scala and Python Spark Jobs in Yarn

Ed Barnes, Ruslan Vaulin and Chris McCubbin Sqrrl Data







#### **DATA SOURCES**

#### **BEHAVIOR GRAPH**

#### THREAT HUNTING

#### SECURITY DATA







Alerts

Threat Intelligence

#### **NETWORK DATA**



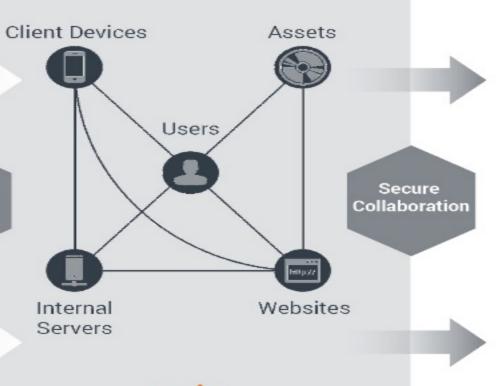
Proxy





NetFlow

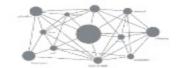
Dynamic Knowledge Extraction



SEARCH



#### **EXPLORATION**



REPORTS



ANOMALIES



#### **ENDPOINT/IDENTITY DATA**





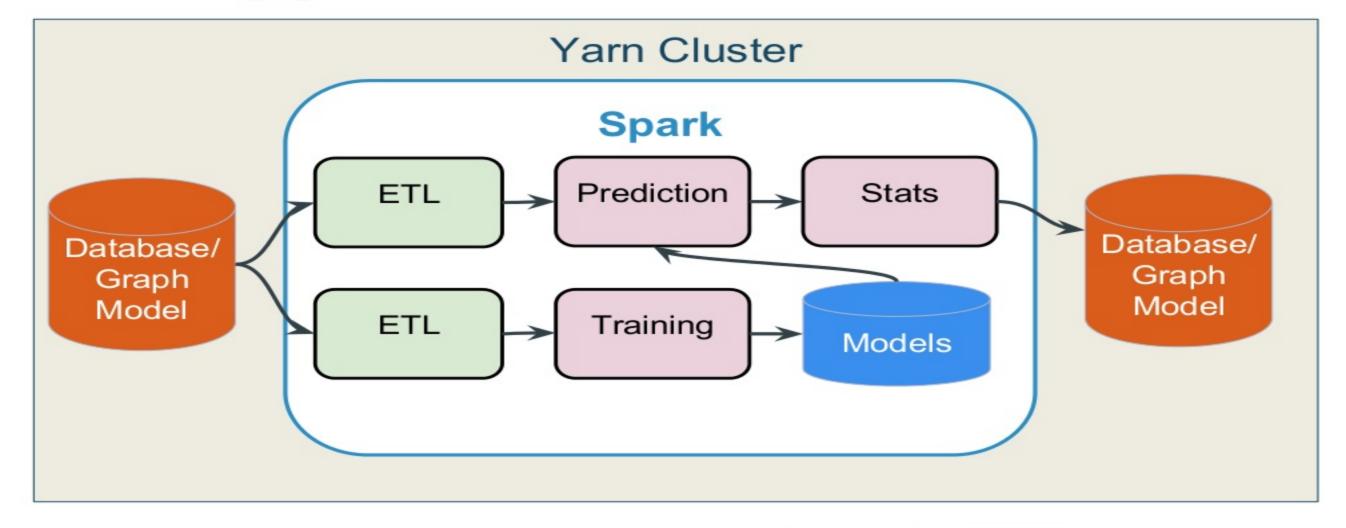


Authentication Processes



**MACHINE LEARNING** 

#### **ML Application Workflow**





## Taking Spark Applications into Production

- Requires execution framework
- Scalable, Robust, Tested
- Test at scale
- Many issues show up only at scale
  - Performance
  - Memory requirements
  - Failures
  - Scaling
- Debugging distributed applications is really hard!

## Spark UI: job level



Jobs Stages

Storage Environment

SPARK: DGA - training for att dn... application UI

#### Spark Jobs (?)

Total Uptime: 1.6 h Scheduling Mode: FIFO Completed Jobs: 10 Failed Jobs: 1

▶ Event Timeline

#### Completed Jobs (10)

Job Id	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
9	reduce at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 23:06:31	9.6 min	1/1	40/40
8	reduce at /data7/yarrv/local/usercache/yarrv/appcache/application_1485380275144_0028/container	2017/01/25 22:58:14	8.0 min	2/2	80/80
7	count at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 22:50:39	7.6 min	1/1	40/40
6	collect at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 22:50:21	18 s	1/1	40/40
5	collect at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 22:31:21	19 min	2/2	80/80
4	collect at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 22:24:50	6.5 min	1/1	40/40
3	count at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 22:18:18	6.5 min	1/1	40/40
2	stats at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container_1	2017/01/25 22:11:30	6.8 min	1/1	40/40
1	runJob at PythonRDD.scala:393	2017/01/25 22:11:25	5 s	1/1	1/1
0	take at SerDeUtil.scala:201	2017/01/25 22:11:22	3 s	1/1	1/1

#### Failed Jobs (1)

Job Id	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
10	reduce at /data7/yarn/local/usercache/yarn/appcache/application_1485380275144_0028/container	2017/01/25 23:16:24	30 min	0/1 (1 failed)	39/40 (4 failed)

## Spark UI: task level



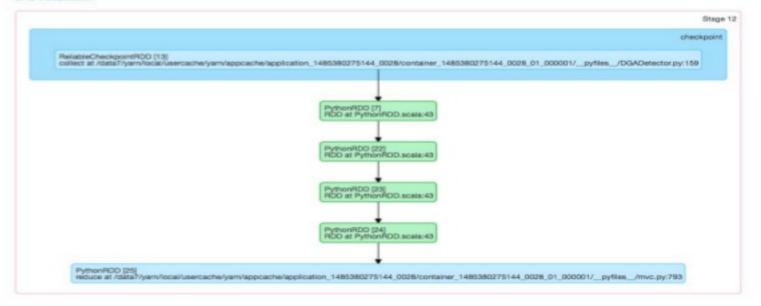
Jobs Stages Storage Environment Executors

SPARK: DGA - training for att dn... application UI

#### Details for Stage 12 (Attempt 0)

Total Time Across All Tasks: 2.4 h Locality Level Summary: Node local: 2: Process local: 41 Input Size / Records: 2.1 GB / 50932

- DAG Visualization



#### Tasks

Index	ID	Attempt	Status	Locality Level	Executor ID / Host		Duration	GC Time	Input Size / Records	Errors
0	407	0	SUCCESS	PROCESS_LOCAL	8/	2017/01/25	8.2 min	90	104.9	
32	441	0	FAILED	PROCESS_LOCAL		2017/01/25 23:25:08	21 min	2 s	362.9 MB (memory) / 8211	java.lang.OutOfMemoryError: Java heap space +details

## Spark UI: task level



Jobs Stages Storage Environment Executors

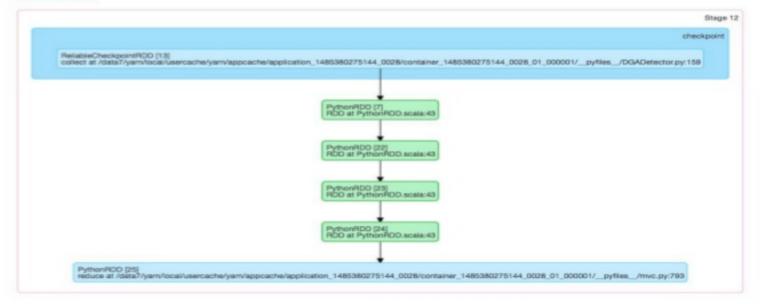
SPARK: DGA - training for att dn... application UI

OOM

#### Details for Stage 12 (Attempt 0)

Total Time Across All Tasks: 2.4 h Locality Level Summary: Node local: 2: Process local: 41 Input Size / Records: 2.1 GB / 50932

- DAG Visualization



#### Tasks

Index	ID	Attempt	Status	Locality Level	Executor ID / Host		Duration	GC Time	Input Size / Records	Errors	Excep
0	407	0	SUCCESS	PROCESS_LOCAL	8/	2017/01/25	8.2 min	90	104.9		
32	441	0	FAILED	PROCESS_LOCAL		2017/01/25 23:25:08	21 min	25	362.9 MB (memory) / 8211	space	MemoryError: Java heap +details

#### Case Study: Py4J Issue

 Testing engineer - "Your code blows up with OOM when processing X amount of data!"

```
17/01/30 12:01:18 INFO ContextCleaner: Cleaned accumulator 2

Exception in thread "Thread-7" java.lang.OutOfMemoryError: Java heap space
at java.util.Arrays.copyOf(Arrays.java:2367)
at java.lang.AbstractStringBuilder.expandCapacity(AbstractStringBuilder.java:130)
at java.lang.AbstractStringBuilder.ensureCapacityInternal(AbstractStringBuilder.java:114)
at java.lang.AbstractStringBuilder.append(AbstractStringBuilder.java:587)
at java.lang.StringBuilder.append(StringBuilder.java:214)
at py4j.Protocol.getOutputCommand(Protocol.java:322)
at py4j.commands.CallCommand.execute(CallCommand.java:82)
at java.lang.Thread.run(Thread.java:745)
```

- Why?!!!!
- Nothing obvious in Spark UI!

#### Case Study: Py4J Issue

 Testing engineer - "Your code blows up with OOM when processing X amount of data!"

```
17/01/30 12:01:18 INFO ContextCleaner: Cleaned accumulator 2

Exception in thread "Thread-f" java.lang.OutOfMemoryError: Java heap space
    at java.util.Arrays.copyof(Arrays.java:2367)
    at java.lang.AbstractStringBuilder.expandCapacity(AbstractStringBuilder.java:130)
    at java.lang.AbstractStringBuilder.ensureCapacityInternal(AbstractStringBuilder.java:114)
    at java.lang.AbstractStringBuilder.append(AbstractStringBuilder.java:587)
    at java.lang.StringBuilder.append(StringBuilder.java:214)
    at py4j.Protocol.getOutputCommand(Protocol.java:322)
    at py4j.commands.CallCommand.execute(CallCommand.java:82)
    at java.lang.Thread.run(Thread.java:745)
```

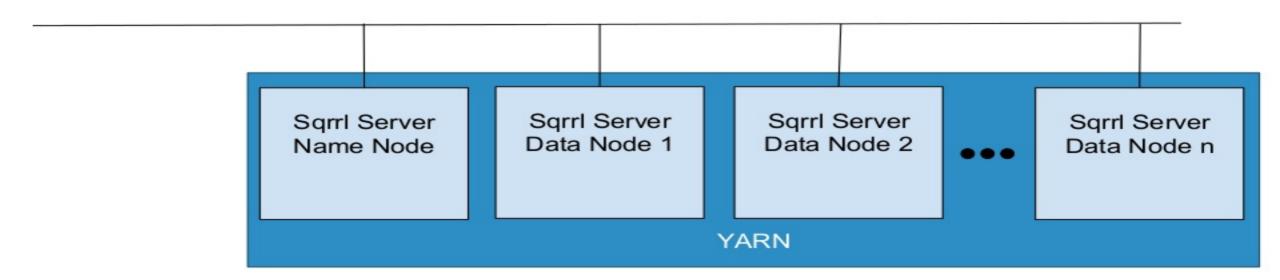
OOM

Exception

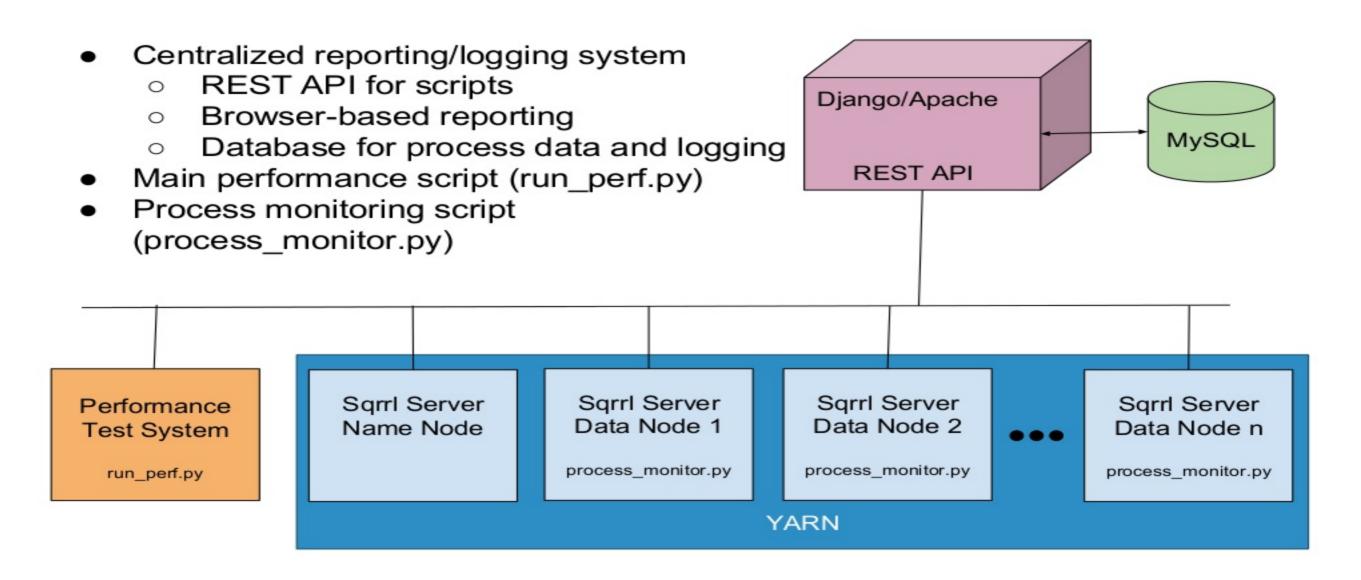
- Why?!!!!
- Nothing obvious in Spark UI!

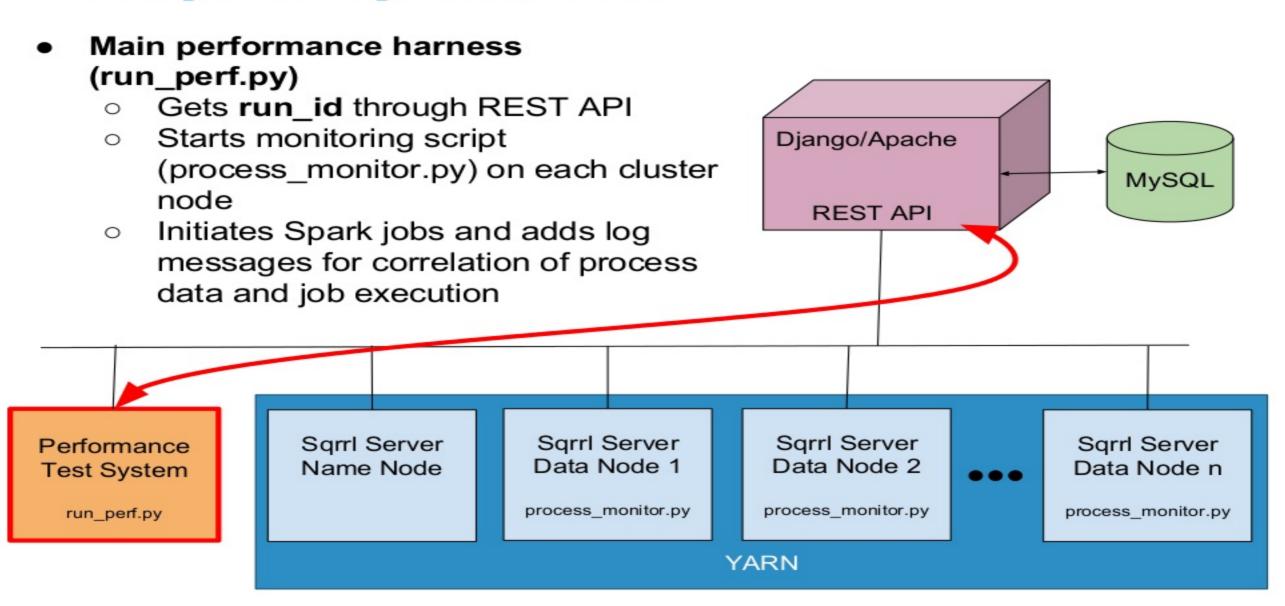
## **Tooling Approach Requirements**

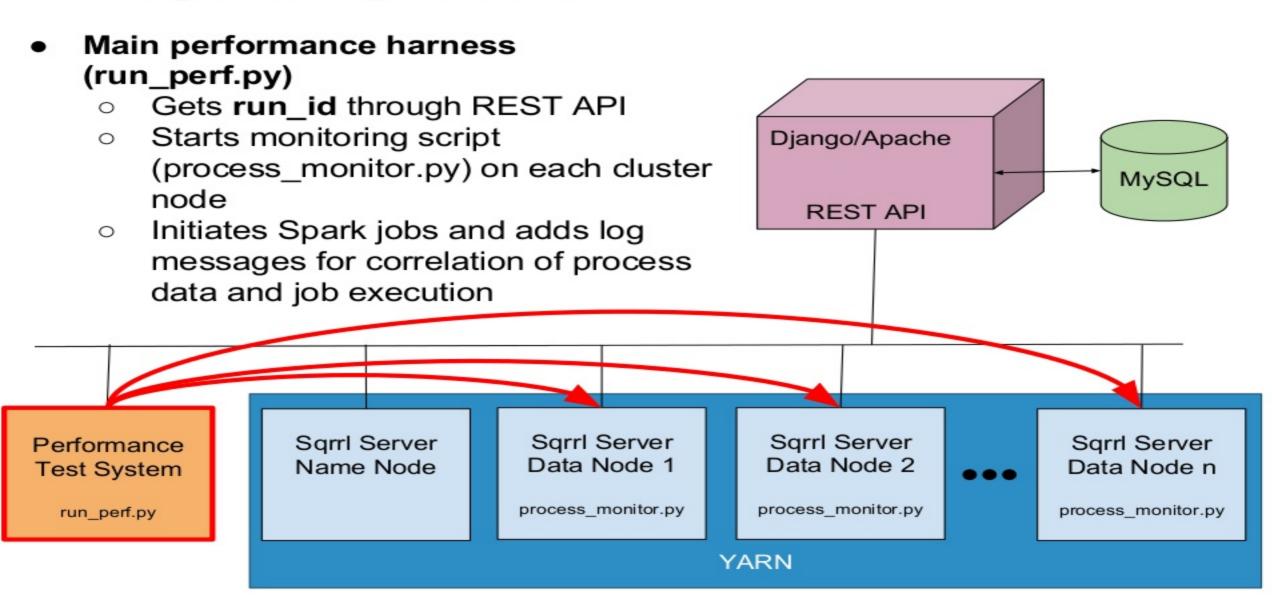
- Cluster-wide process monitoring
- Per-node, per-process statistics
- Identification of high CPU, memory usage
- Record process hierarchy/timing of Spark jobs under YARN
- Characterize scaling behaviors for production/releases
- Integrate with internal processes/harnesses for general development and test use

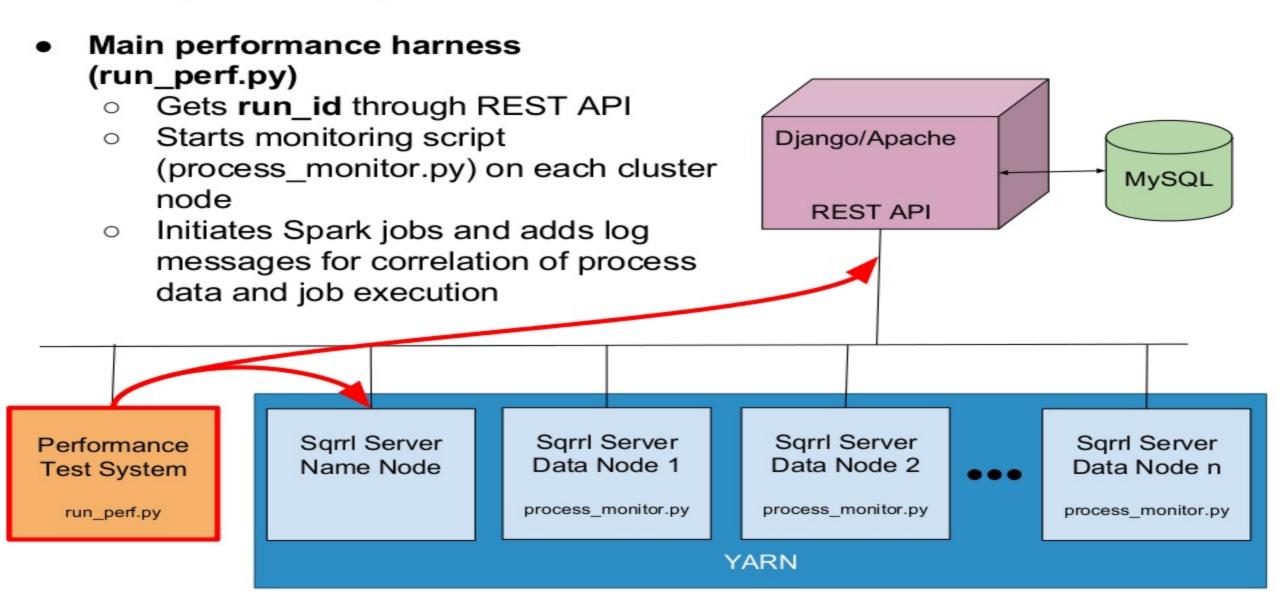


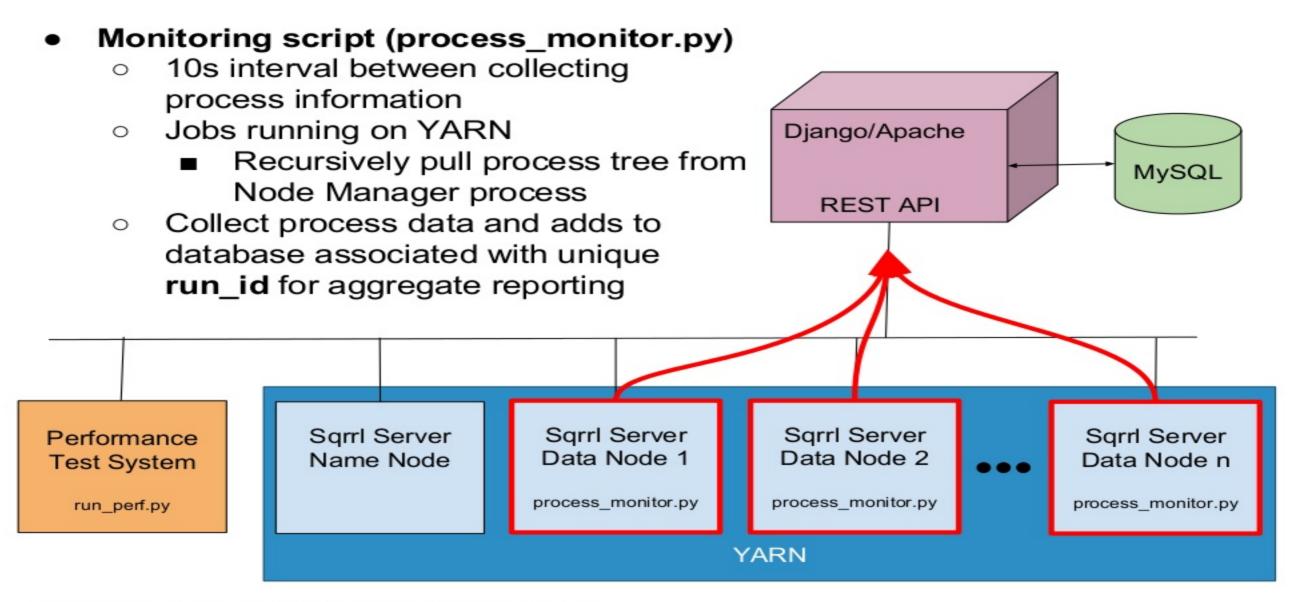
## **Overall System Design**



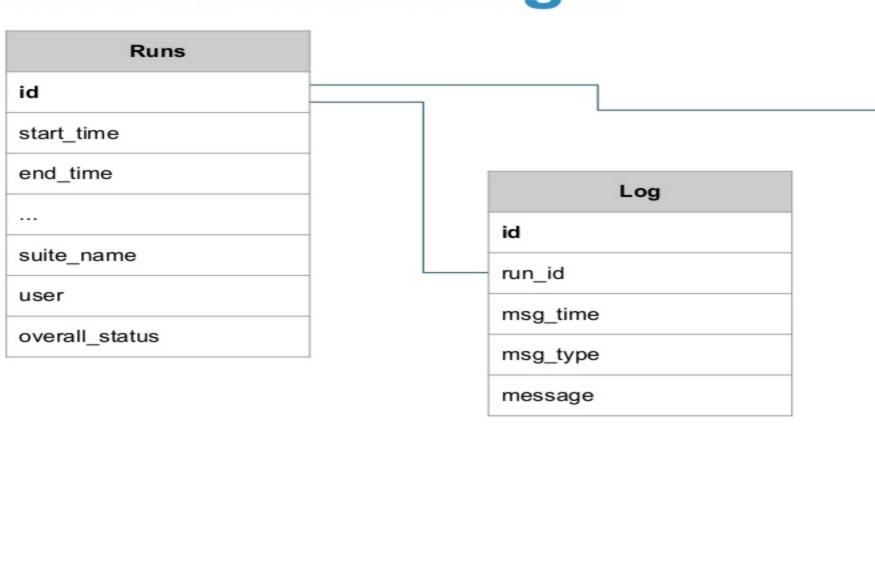








## Database Design



#### **ProcessMonitor** id time\_stamp monitor run id pid ppid pcpu pmem cpu time elapsed time resident size mapped private shared short command long\_command

Dashboard

Results

Longevity Dashboard

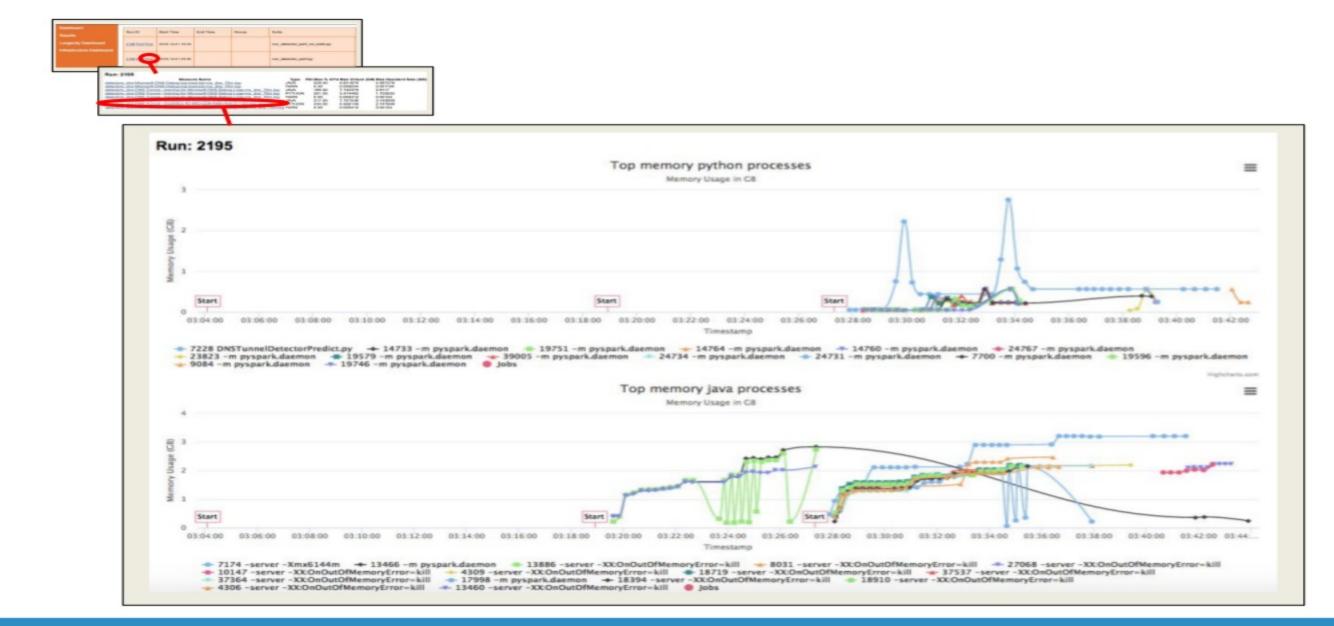
Infrastructure Dashboard

Run ID	Start Time	End Time	Group	Suite
2196 Perf Proc	2016-12-01 16:40			run_detector_perf_no_ootb.py
2195 Perf Proc	2016-12-01 03:00			run_detector_perf.py

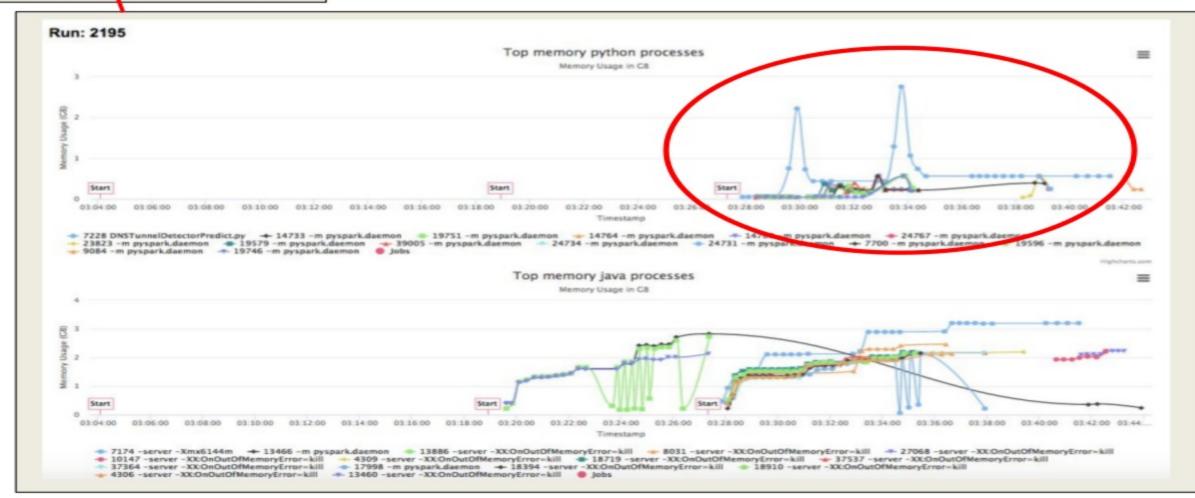
Run: 2195

Suite		
pert_no_cotb.py		
perf.py		
p		

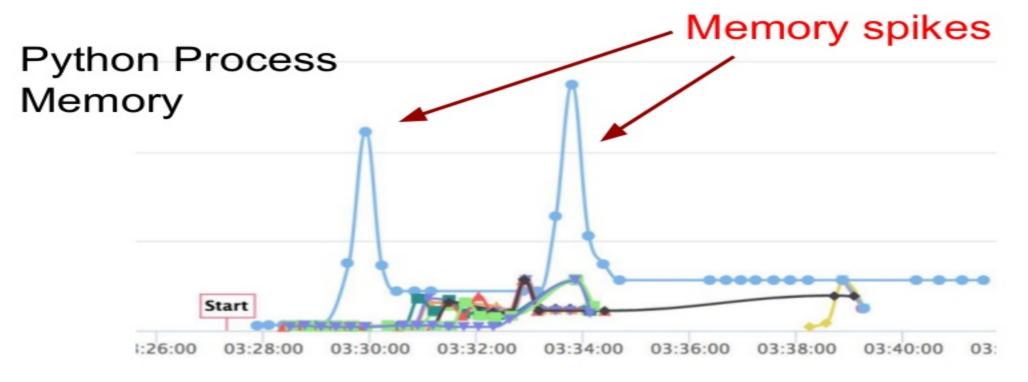
ı	Ruii. 2193				
ı	Measure Name T	ype	PID Max % CP	U Max Virtual	(GB) Max Resident Size (GB)
ı	detectors dns-Microsoft-DNS-Debug-log-load-job-ms dns 70m.log	/A	529.00	5.841876	4.997276
ı	detectors dns-Microsoft-DNS-Debug-log-load-job-ms dns 70m.log	RN	0.40	0.009244	0.001104
ı	detectors dns-DNS-Tunneltraining-for-Microsoft-DNS-Debug-Logs-ms dns 70m.log JAV	/A	185.00	7.740376	2.8117
ı	detectors dns-DNS-Tunneltraining-for-Microsoft-DNS-Debug-Logs-ms dns 70m.log PYT	THON	251.00	3.474492	1.702632
ı	detectors dns-DNS-Tunneltraining-for-Microsoft-DNS-Debug-Logs-ms dns 70m.log YAF	RN	0.00	0.009412	0.00124
ı	detectors dos DNS Tunnel prediction for Microsoft DNS Debug Logs ms dos 70m.log JAV	/A	217.00	7.757036	3.193608
ŧ	detectors dns-DNS-Tunnelprediction-for-Microsoft-DNS-Debug-Logs-ms dns 70m.log>Y1	THON	434.00	4.556148	2.747848
	detectors dris-DINS-Turmel - prediction for Microsoft DNS-Debug-Logs-ms dns 70m.log YAF	RN	0.00	0.009412	0.00124





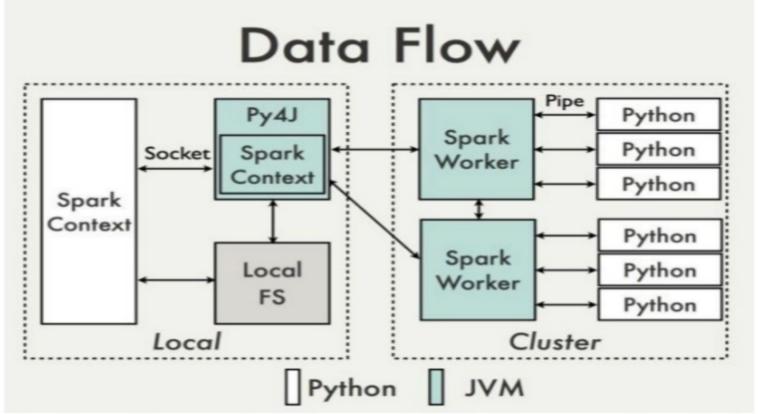


#### Py4J Issue: Evidence



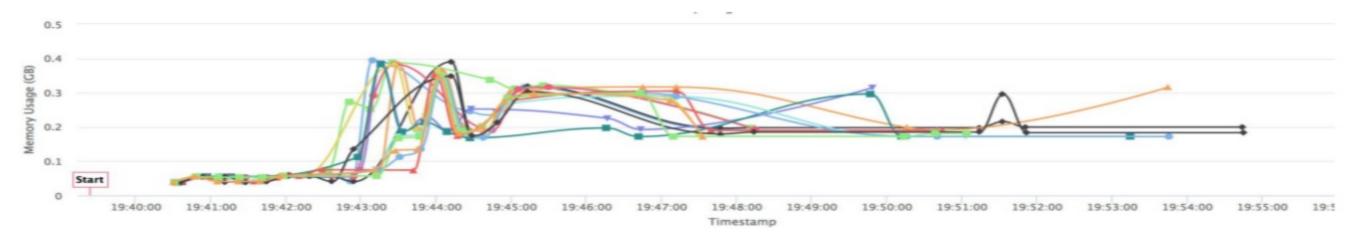
- Memory spiked during loading of trained ML models into python process.
- Only on driver!

## Diagnosis and Conclusions



- Loading trained model from HDFS required decryption with java libraries and Py4J for loading into python.
- Py4J protocol inflated size by almost 100x!
- Solution: implement custom protocol for loading data from jvm to python via socket.

#### Py4J issue: After Fix



- Memory spikes are gone!
- Ready for production!

## Recommendations & Lessons Learned

- Do not take scalability for granted!
- Understand Spark's architecture
  - Python/JVM interaction
- Follow best practices
  - Iterators not Lists
  - Careful with joins
- Understand your computing demands
- Test at scale
- Invest in tools
- Think distributed and your code will shine!

## Thank You.



sqrrl.com @SqrrlData

ebarnes@sqrrl.com ruslan@sqrrl.com chris@sqrrl.com

