# Big Data Analytics on Container-Orchestrated Systems

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July 20, 2017

#### **Outline**

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Problem statement

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**Evaluation & Results** 

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Future work

# Why?

# Keeping up with data growth

#### **Problem**

- IOT & Social networks
- Internet traffic
  - Current: 72 petabytes/month
  - Prediction 2021: 232 petabytes/month.



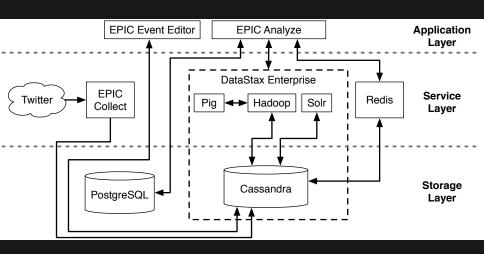
#### Another problem

- Need to scale Big Data Analytics System
- Keeping mantainance at low cost
- Container-orchestrated make infrastructure easier
- Migrate Project EPIC architecture to container-orchestration system

# Background

# **Project EPIC**

- EPIC Collect
- EPIC Analyze



#### Containerization

- Operating-system-level virtualization
- Use host machine system resources
- Docker most used alternative
- Development microservices

#### Container-orchestration systems

- Container interaction abstraction
- Great to deploy microservices architectures
- Apache Mesos vs Kubernetes





#### Microservices Architecture

- Small & specific
- Better scalability
- Loosely-coupled & highly-cohesive
- Orchestration <> Coreography

# Coreography microservice architecture

- Easier to extend

- PubSub interaction
- Messaging system: Apache Kafka
- Asyncronous

#### Problem statement

#### Problem statement

- 1. Advantages and/or limitations from existing infrastructure
  - 1.1 More reliable?
  - 1.2 More scalable?
- 2. Lower maintenance costs than the existing infrastructure?
  - 2.1 Easier to deploy?
  - 2.2 Easier to upgrade?
  - 2.3 More resilient to failures?

# Approach

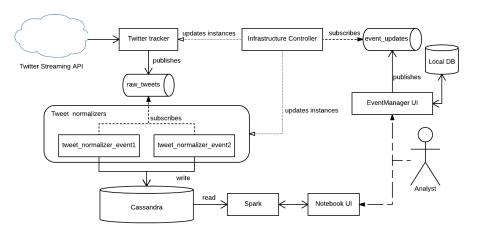
#### **Features**

- Event management
- Real-time collection of streaming Twitter data
- Real-time classification of incoming tweets
- Data Analysis

# Non-functional requirements

- Less code
- Easier deployment
- More flexible

- Better scalability



#### Demo time!

#### Let's track an event...

Event Manager UI

# ...and analyze it!

Zeppelin Notebook

#### **Evaluation & Results**

# Reliability

#### Current vs Prototype

- EPIC Collect: multi threaded, auto-monitoring
- EPIC Analyze: tight dependency to external resources, manual deployment, no recovery procedure

- Kubernetes abstraction
- Controller manager:
  Auto-recovery, replica
  deployment, rolling
  update
- Scheduler: check memory and cpu and assign intance deployment

NAMESPACE	NAME	READY	STATUS	RESTARTS	AGE
cassandra	cassandra-0	2/2	Running	1	2d
cassandra	cassandra-1	2/2	Running	2	11d
cassandra	cassandra-2	2/2	Running	7	22d
cassandra	spark-master-controller-bkdl7	1/1	Running	0	22d
default	hearth-event-parser-2160998245-m19st	1/1	Running	9	2d
default	k8s-controller-3919038388-75tkk	1/1	Running	0	2d
default	smiley-event-parser-3033807940-c0cwj	1/1	Running	9	2d
default	twitter-tracker-2482383360-s0kz1	1/1	Running	5	2d
frontend	eventmanager-ui-3464180876-h605f	1/1	Running	0	23d
frontend	zeppelin-3633522582-t0kng	1/1	Running	0	2d
kafka	kafka-0	1/1	Running	3	11d
kafka	kafka-1	1/1	Running	0	2d
kafka	zoo-0	1/1	Running	0	2d
kafka	zoo-1	1/1	Running	0	2d
kafka	zoo-2	1/1	Running	0	25d
kube-system	heapster-v1.3.0-4211727876-kv0cl	2/2	Running	0	11d
kube-system	kube-dns-806549836-431vx	3/3	Running	0	11d
kube-system	kube-dns-autoscaler-2528518105-2wlsr	1/1	Running	1	27d
kube-system	kube-proxy-gke-development-development-dcfa2eb3-2jhj	1/1	Running	0	2d
kube-svstem	kube-proxv-ake-development-development-dcfa2eb3-15xb	1/1	Runnina	1	26d

# Scalability

#### Current vs Prototype

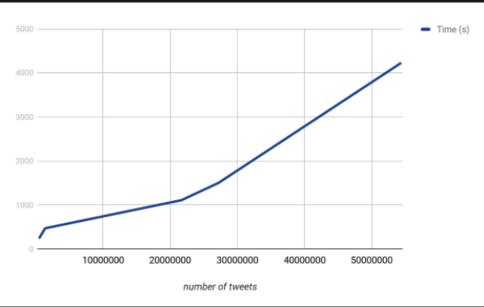
- EPIC Collect: manual, huge load, monolith replication
- EPIC Analyze: monolith, load balancing, shared session state
- Not easy and manual

- Replica specification
- Independently scalable
- Stateless microservices
- Auto scale

Abstracted management interface

### Performance

#### WordCount execution



#### Development and mantenance

#### Current vs Prototype

- Large and complex
- Strong use of frameworks

- Simple and small code (mostly)
- Independent microservices
- Technology flexibility
- YAML deployment description files



#### Results

- High scalability with microservices
- Scale components individually
- Abstraction over system management vs manual
- Monitoring built-in

#### Results

- Easier to mantain: less code, faster development
- Ease of configuration and deployment
- Flexibility on cloud provider deployment
- Resource usage optimizer
- Other features: Rolling-update, auto-scale...



#### Related work

#### Related work

- SMACK: Spark, Mesos, Akka, Cassandra and Kafka [Raul Estrada et al. 2016]
- Hadoop ecosystem [Han Hu et al. 2014]
- Lambda architecture [Zirije Hasani et al. 2014]

#### Future work

#### **Future work**

- Improve resource specification (CPU, memory)
- Unified authentification

Better Cassandra structure

Extend the system to fit current

# Questions?

# Thank you!

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Repo github.com/casassg/thesis