Big Data Analytics on Container-Orchestrated Systems

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July 20, 2017

Outline

Introduction & Background

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Maintenance

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Introduction & Background

Why?

Data growth

IOT & Social networks

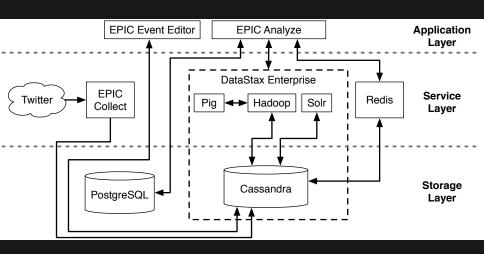
- Increase internet traffic: 3x internet traffic by 2021
- Scale up Big Data Analytics System
- Keeping maintenance at low cost
- Container-orchestrated make infrastructure easier



Keep up with data growth

Background: Project EPIC

- EPIC Collect
- EPIC Analyze



Background: Containerization

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

Development microservices

Background: Container-orchestration systems

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





Background: Microservices Architecture

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

Background: Coreography microservice architecture

- Easier to extend

- Asyncronous
- PubSub interaction
- Messaging system: Apache Kafka



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?

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]

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

Custom components

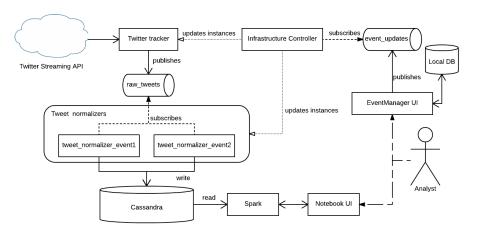
- Event Manager
- Infrastructure Controller

- Twitter tracker
- Twitter Normalizer

Cassandra table structure

```
CREATE TABLE twitter_analytics.tweet (
   id uuid, t_id text, event_kw text, event_name text, hashtags list <text>,
   media_url text, t_coordinates text, t_created_at timestamp,
   t_favorite_count int, t_favorited boolean, t_geo text,
   t_is_a_retweet boolean, t_lang text, t_retweet_count int,
   t_retweeted boolean, t_text text, u_created_at timestamp,
   u_description text, u_favourites_count int, u_followers_count int,
   u_friends_count int, u_geo_enabled boolean, u_id text, u_lang text,
   u_listed_count int, u_location text, u_name text, u_screen_name text,
   u_statuses_count int, u_time_zone text, u_url text, u_uto_offset int,
   um_id text, um_name text, um_screen_name text, urls list <text>,
   PRIMARY KEY (id, t_id))
```

Listing 1: Tweets CQL table script



EPIC EventManager

Smiley

DB code smiley Tokens :D. :J, XD, LOL, WTF, XP, :'L -
Hearth

DB code hearth Tokens <3, love

DB code nba-draft

DB code hoardn Tokens nba, draft, Lonzo, Lavar, basketball, nba draft

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

cassandraspark-samples/analytics

Real time twitter Analytics

Press run above to update stats

Zeppelin

Columns; id, t.id, event, kw, event name, hashtags, media uri, t. coordinates, t. created at, t. favorite. count, t. favorited, t. geo. t. is, a retweet, t. lang, t. retweet count, t. retweeted, t. text, u. created at, u. description, u. favourites, count, u followers count, u friends count, u geo enabled, u id. u lang, u listed count, u location, u name, u screen name, u statuses count, u time zone, u url, u utc offset, um id. um name, um screen name, urls

SOL Context configured

Stats on tweets

Different stats on all the current dataset available on the database

Total tweet count

import org.apache.spark... import com.datastax.spark.connector._

table: com.datastax.spark.connector.rdd.CassandraTableScanRDDFcom.datastax.spark.connector.CassandraRowT = CassandraTableScanRDDF0T at RDD at CassandraRDD.scala:15

rese: IndexedSea[String] = WrappedArray(id, t.id, event_kw, event_name, hashtags, media_url, t_coordinates, t_created_at, t_favorite_count, t_favorited, t_geo, t_is_a_retweet, t_lang, t_retweet_count, t_retweeted, t_text, u_cr

eated_at, u_description, u_favourites_count, u_followers_count, u_friends_count, u_geo_enabled, u_id, u_lang, u_listed_count, u_location, u_name, u_screen_name, u_statuses_count, u_time_zone, u_url, u_utc_offset, um_id, um_nam e, um_screen_name, urls) 35713760

Tweets by event







Languages used in all tweets



Demo time!

Let's track an event...

Event Manager UI

...and analyze it!

Zeppelin Notebook

Evaluation & Results

Reliability

Current vs Prototype

- Threads to monitoring

Kubernetes abstraction

Every minute script check

Auto recovery

- Check log file size

- Rolling update
- Node assignation depending on resources usage

Scalability

Current vs Prototype

Monolyth

 Kubernets replica specification

- Manual process

Stateless microservices

- Shared session state

Independently scalable

Load balancing

- Abstracted infrastructure
- Auto scale

Maintenance

Current infrastructure

- Large & complex
- Strong use of frameworks
- Manual deployment
- Dedicated machines



Prototype

- Easier to mantain: less code, faster development
- Technology flexibility
- Easier to deploy: YAML description files
- Flexibility on cloud provider
- Available built-in tools





Prototype

Twitter tracker 108 lines

Kubernetes controller 145 lines

Event manager 2090 lines

Tweet normalizer 209 lines

Performance

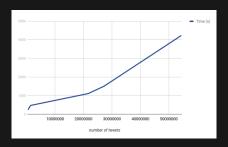
WordCount

Listing 2: WordCount Spark script

```
(1) ShuffledRDD[112] at sortByKey at <console>:31 []
+-(176) MapPartitionsRDD[111] at map at <console>:31 []
| ShuffledRDD[110] at reduceByKey at <console>:31 []
+-(176) MapPartitionsRDD[109] at map at <console>:31 []
| MapPartitionsRDD[108] at flatMap at <console>:31 []
| CassandraTableScanRDD[107] at RDD at CassandraRDD.scala:15 []
```

Listing 3: Debug string rdd

WordCount



Number of tweets	Time (s)	Tweets/sec
490199	238	2060
1400884	469	2987
21680851	1107	19585
27199614	1500	18133
54395957	4228	12866



Future work

- Improve resource specification (CPU, memory)
- Better Cassandra structure
- Unified authentification

Extend the system with new features

Questions?

Thank you!

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Website gerard.space

Repo github.com/casassg/thesis