

Druid pour l'analyse de données en temps réel

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Intro

Plan

- ▶ Introduction; why?
- ▶ How?

Experience

- ▶ Real Time Social Media Analytics

Real Time?

- ▶ Ingestion Latency: seconds
- ▶ Query Latency: seconds

Demande

- ▶ Twitter: 20k msg/s, 1msg = 10ko pendant 24h
- ▶ Facebook public: 1000 à 2000 msg/s en continu

En pratique

- ▶ Twitter: 400 msg/s en continu, pics à 1500

Origine (PHP)



Introduction

- ▶ Traitement de donnée gros volume + faible latence
- ▶ Typiquement pulse

DEMO

Pre Considerations

Discovered vs Invented

Try to conceptualize (events)

Scalable + Real Time + Fail safe

- ▶ timeseries
- ▶ alerting system
- ▶ top N
- ▶ etc...

In the End

Druid concepts are always emerging naturally

Druid

Who

Metamarkets

Goal

Druid is an open source store designed for real-time exploratory analytics on large data sets.

hosted dashboard that would allow users to arbitrarily explore and visualize event streams.

Concepts

- ▶ Column-oriented storage layout
- ▶ distributed, shared-nothing architecture
- ▶ advanced indexing structure

Features

- ▶ fast aggregations
- ▶ flexible filters
- ▶ low latency data ingestion

**arbitrary exploration of billion-row tables
tables with sub-second latencies**

Storage

- ▶ Columnar
- ▶ Inverted Index
- ▶ Immutable Segments

Columnar Storage

Index

- ▶ Values are dictionary encoded

```
{"USA" 1, "Canada" 2, "Mexico" 3, ...}
```

- ▶ Bitmap for every dimension value (used by filters)

```
"USA" -> [0 1 0 0 1 1 0 0 0]
```

- ▶ Column values (used by aggregation queries)

```
[2,1,3,15,1,1,2,8,7]
```

Data Segments

- ▶ Per time interval
- ▶ skip segments when querying
- ▶ Immutable
- ▶ Cache friendly
- ▶ No locking
- ▶ Versioned
- ▶ No locking
- ▶ Read-write concurrency

Real-time ingestion

- ▶ Via Real-Time Node and Firehose
- ▶ No redundancy or HA, thus not recommended
- ▶ Via Indexing Service and Tranquility API
- ▶ Core API
- ▶ Integration with Streaming Frameworks
- ▶ HTTP Server
- ▶ **Kafka Consumer**

Batch Ingestion

- ▶ File based (HDFS, S3, ...)

Real-time Ingestion

Task 1: [Interval] [Window]
Task 2: []
----->
time

Minimum indexing slots =
Data Sources × Partitions × Replicas × 2

Querying

Query types

- ▶ Group by: group by multiple dimensions
- ▶ Top N: like grouping by a single dimension
- ▶ Timeseries: without grouping over dimensions
- ▶ Search: Dimensions lookup
- ▶ Time Boundary: Find available data timeframe
- ▶ Metadata queries

Tip

- ▶ Prefer `topN` over `groupBy`
- ▶ Prefer `timeseries` over `topN`
- ▶ Use limits (and priorities)

Query Spec

- ▶ Data source
- ▶ Dimensions
- ▶ Interval
- ▶ Filters
- ▶ Aggergations
- ▶ Post Aggregations
- ▶ Granularity
- ▶ Context (query configuration)
- ▶ Limit

Example(s)

TODO

Caching

- ▶ Historical node level
- ▶ By segment
- ▶ Broker Level
- ▶ By segment and query
- ▶ `groupBy` is disabled on purpose!
- ▶ By default - local caching

Load Rules

- ▶ Can be defined
- ▶ What can be set

Components

Druid Components

- ▶ Real-time Nodes
- ▶ Historical Nodes
- ▶ Broker Nodes
- ▶ Coordinator
- ▶ For indexing:
 - ▶ Overlord
 - ▶ Middle Manager
 - ▶ Deep Storage
 - ▶ Metadata Storage
 - ▶ Load Balancer
 - ▶ Cache

Coordinator

Manage Segments

Real-time Nodes

- ▶ Pulling data in real-time
- ▶ Indexing it

Historical Nodes

- ▶ Keep historical segments

Overlord

- ▶ Accepts tasks and distributes them to middle manager

Middle Manager

- ▶ Execute submitted tasks via Peons

Broker Nodes

- ▶ Route query to Real-time and Historical nodes
- ▶ Merge results

Deep Storage

- ▶ Segments backup (HDFS, S3, ...)

Considerations & Tools

When *not* to choose Druid

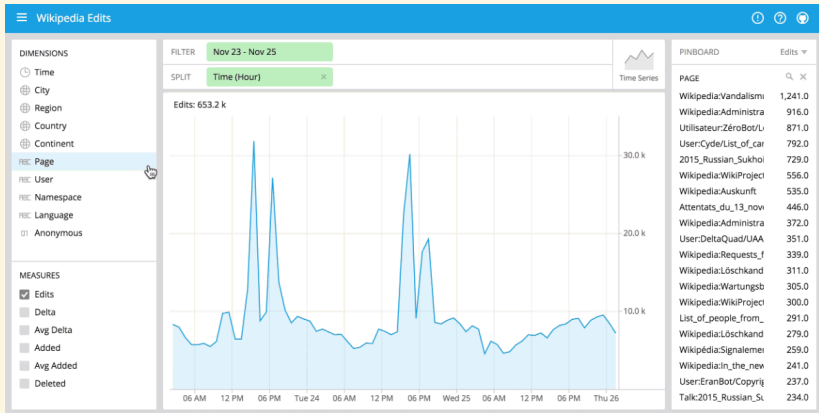
- ▶ Data is not time-series
- ▶ Cardinality is *very* high
- ▶ Number of dimensions is high
- ▶ Setup cost must be avoided

Graphite (metrics)



Graphite

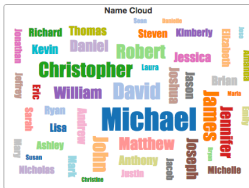
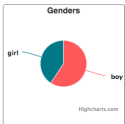
Pivot (exploring data)



Pivot

Caravel (exploring data)

Births



Title

Birth Names Dashboard

The source dataset came from [here](#)

Highcharts.com

Boys

name	sum_num
Michael	2886323
David	1974570
James	1874531
John	1820660
Christopher	1813681
Robert	1669271
Matthew	1390512
William	1344166
Daniel	1285377
Joseph	1263739
Joshua	1073375
Brian	990245

Girls

name	sum_num
Jennifer	1387536
Jessica	997084
Lisa	838689
Elizabeth	802763
Ashley	789125
Sarah	757701
Kimberly	747160
Michelle	724897
Amanda	719625
Melissa	690995
Mary	683738
Stephanie	657908

