Wolfram Wingerath

Big Data Analytics With AWS Athena

Big Data



Big Data Analytics With AWS Athena

Wolfram Wingerath, Code.Talks 2019

🖋 Baqend

I Am Wolle



Research:

- Stream Processing
- Real-Time Databases
- NoSQL & Cloud Systems
- ...





Practice:

- Web Caching •
- Big Data Analytics •
- Anger Management •

•••



I Like Real-Time Stuff

2018

Real-Time Processing Explained

A Survey of Storm, Samza, Spark & Flink

Real-Time Databases Explained

Why Meteor, RethinkDB, Parse & Firebase Don't Scale

2017





METE R 🕅 Rethink DB 🎽 Firebase Ballend

Let's Talk About **Batch** Analytics



What's in the Data?





Accelerating Personalized HTML



Split Testing for Web Performance



Goal: Performance & Business Insights



 Time-to-First-Byte First (Contentful) Paint DOM Timer First Input Delay 	Performance
 Session Length Time on Site First User Interaction Bounce Rate 	User Engagement
 Cart Size Transactions Conversion Rate Revenue 	Business KPIs
 Page Views & Sessions Browser Distribution JavaScript Errors Caching Insights 	QA Metadata

How to Monitor Performance?





- Logging requests is not enough:
 - X User? Rendering? ...
 - X Browser cache (invisible)
 - X Origin requests (no logs)
 - CDN requests
- Solution: **Tracking every Pl** (page impression)

When to Send Data Beacons?





Let's Analyze the Data

Tracking Data in MongoDB



- Debugging with OLTP **queries**
- Analyses with **aggregation pipeline**, e.g.:
 - Average session length
 - Uplift vs. Acceleration (example: Conversion rate vs. first paint uplift)
 - Different browser timings by device/browser/...





- Automation via Jenkins jobs
- **Caching-related** statistics, e.g.:
 - Request latency distribution (histogram)
 - Image optimization efficiency

Problems I: CDN Data Import



- 1. Partitioning by hour, but <u>not</u> by customer
- → Not scalable!
- 2. Indexing & database import:
 - a) Import into indexed table
 - b) Drop index \rightarrow import \rightarrow create index
 - c) Query table without index

- each takes forever

Problems II: Aggregation Pipeline



Indexing

Queries over non-indexed

attributes were infeasible



Runtime

Even with indexes in place,

queries could take 30+ min.



Scalability

Queries got slower with

increasing amounts of data



MongoDB aggregation pipelines

become sophisticated quickly

Problems III: Reporting



- Scheduling:
 - 1. On-demand
 - 2. Periodic job
- Problems:
 - Cumbersome to build & maintain
 - Awkward to extend
 - Unreliable

Problems IV: Joins in MongoDB



Joins infeasible for data-intensive queries!

- → No conversion analysis
- → No business uplift validation
- → <u>Not acceptable!</u>



Fixing My Life With Flexe Athena

The "A" Stands for " AWS ome"

- Desperate **attempt**:
 - 1. Dump MongoDB collection
 - 2. Upload to S3
 - 3. Query with Athena

Typical analysis:



- 1 equi-join
- 3 mio. Pls
- ~15+ min.

The "A" Stands for " AWS ome"

Desperate attempt: New best practice:

- 1. Dump MongoDB collection
- 2. Upload to S3
- 3. Query with Athena



- Typical analysis:
 - 1 equi-join
 - 3 mio. Pls
 - ~10 <u>seconds</u>

What's an Athena?



- Managed Presto:
 - Interactive analytics with SQL
 - Heterogeneous datastores
 - Petabyte-scale (Facebook)

- **Pricing** by scanned data volume:
 - → Efficient storage formats!
 - → Partitioning or clustering!
 - → Careful query design!

Upgrading Our ETL Pipeline



- Scalability & efficiency:
 - Hundreds of gigabytes scanned in a query
 - Response time on the order of seconds

Processing Stages & Latency



- Simple metrics
 - Counters
 - Extreme values
 - Specific errors

- Complex aggregations
 - Conversion rate
 - Performance by day
 - Seasonal effects

Stage 0: Data Preparation

```
{
    "_id": "ABC",
    "loadEvent": {
        "$numberLong": "1571101211368"
    },
    "createdAt": {
        "$date":"2019-10-15T01:00:11.462Z"
    },
    ...
}
```



- Schema Definition
 - **1. Tables** for raw data
 - 2. Views on top to hide artifacts

Example: Timestamps



- 1. Extract UNIX timestamp from JSON
- 2. Cast to varchar
- 3. Cast to decimal
- 4. Divide by 1000
- 5. Convert to timestamp

Example: Timestamps



- 1. Extract UNIX timestamp from JSON
- 2. Cast to varchar
- 3. Cast to decimal
- 4. Divide by 1000
- 5. Filter out rubbish
- 6. Convert to timestamp

Stage 1: Join Beacons



- Consolidate PI data into single rows
- Data cleaning (e.g. nullify when loadEnd < navigationStart)</p>

Stage 2: Resolve User Agents



WhatIsMyBrowser.com

- **Paid service**: interpreting user agents is complex!
- Fallback: simple case-when logic for browsers and device type
- **Simplification** required, e.g.:
 - Device: mobile/desktop/tablet/server/game console/wearable/vehicle/...
 - Browser: Chrome/Firefox/Safari/Opera/IE/Netscape/Tesla Browser/...





- Session length, bounces
- Time on site
- Performance (e.g. median FCP)
- Conversions

Stage 4: Materialized Views



Business uplift during Proof of Concept (PoC) Aggregation over days or weeks

Performance and business trend analysis

Reporting: The Right Tool for the Job





Unified analytics

Reporting: The Right Tool for the Job



- Requirements:
 - Automation
 - Easy data exploration
 - Robustness






Worst Practices

Why, Oh, Why, Quick Sight?

		Last X Days (Inc	cluding Yester	rday) > Device Type		
		i ‡ 14			r.	
skgroup						
SKHTML	Sessions	all 0	Mobile 0	Desktop		
SKITPL	Orders	0	0	0		
	Conversions	0	0	0		
	Order Value	€0	€0	€0		
	skewed Order					
	Order Uplift					
	skewed Order					
	Order Value U					
	Load Uplift					
	First Paint Up					
	Session Lengt					
	Return Rate					
	Bounces Impr					
	AOV Uplift					
	# PI Uplift					
	# Session Upl					
	# User Uplift					
	addToCart Val					
	# addToCart					

Why, Oh, Why, Quick Sight?

ා Reset ∇ Filter Controls App none Last X Days 14 Speed Kit PoC Analysis Dashboard Executive Summary of Uplifts (Positive Value = Improvement) Last X Days (Including Yesterday) > Device Type ï...‡ 14 skgroup all Mobile Desktop SKHTML Sessions Orders Conversions Order Value €0 €0 €0 skewed Order.. Order Uplift skewed Order ... Order Value U... Load Uplift First Paint Up.. Session Lengt. Return Rate Bounces Impr.. AOV Uplift # PI Uplift # Session Upl... # User Uplift addToCart Val.. # addToCart ..

Time Works Differently in AWS Dashboards







Time Works Differently in AWS Dashboards





Click Once, Pay Twice

rum.response_cause_by_day ×	🧪 app	🖍 days_in_past	🖍 responsecau	🧪 total
SPICE Data Set 53.6KB mport complete: 00% success 00% success i80 rows were imported to SPICE 0 rows were skipped rew summary ast refreshed: 20 hours ago Schedule refresh		e preview data from the so		# Int Please try again or
ta source name: rum.responseCauseByDay tabase name: ATHENA Delete to the solution of the solut	Reload previe	every 234	iew.	
haven't updated the data set, vet.	Sight c26063a5-b35a-468 a1f */ SELECT "app", past", "responsecause",		5	Succeeded 55.

But There is More !



Questionable Limits

Diagrams in the dashboard, query timeouts, etc.



Exhausted Resources in Athena Queries fail depending on time, day, and weather



Implicit Query Rewriting in QuickSight Hard to predict, hard to circumvent, hardly helpful







Real-Time Analytics & Reporting



- Continuous ingestion
- Streaming analytics & real-time dashboards
- Drop QuickSight



Thanks! Any Questions?



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Thanks! Any Questions?

Right Here Right Now GDPR Panel Discussion

Kino 7 Friday, 11:00 Web Performance Talk

Kino 7 Friday, 16:00 Service Worker Hackathon

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