# re:Invent

Getting Maximum Performance from Amazon Redshift: High Concurrency, Fast Data Loads, and Complex Queries







November 13, 2013



#### **Amazon Redshift**

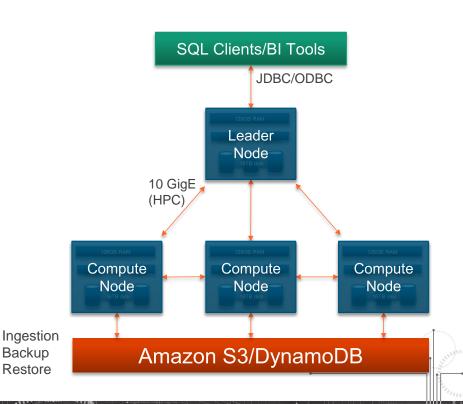


Fast, simple, petabyte-scale data warehousing for less than \$1,000/TB/Year



#### **Amazon Redshift architecture**

- Leader Node
  - SQL endpoint
  - Stores metadata
  - Coordinates query execution
- Compute Nodes
  - Local, columnar storage
  - Execute queries in parallel
  - Load, backup, restore via Amazon S3
  - Parallel load from Amazon DynamoDB
- Single node version available



#### Amazon Redshift is priced to let you analyze all your data

	Price Per Hour for HS1.XL Single Node	Effective Hourly Price per TB	Effective Annual Price per TB
On-Demand	\$ 0.850	\$ 0.425	\$ 3,723
1 Year Reservation	\$ 0.500	\$ 0.250	\$ 2,190
3 Year Reservation	\$ 0.228	\$ 0.114	\$ 999

#### **Simple Pricing**

Number of Nodes x Cost per Hour No charge for Leader Node No upfront costs Pay as you go



# re:Invent

Getting Maximum Performance from Amazon Redshift: High Concurrency

Ben Myles, Desk.com (@benmyles)

November 13, 2013





Delight your customers with awesome Customer Support

TRY IT FOR FREE TODAY No credit card required.

GET STARTED ▶ REQUEST A DEMO



"With everything that we do, Desk.com makes us better."

Square





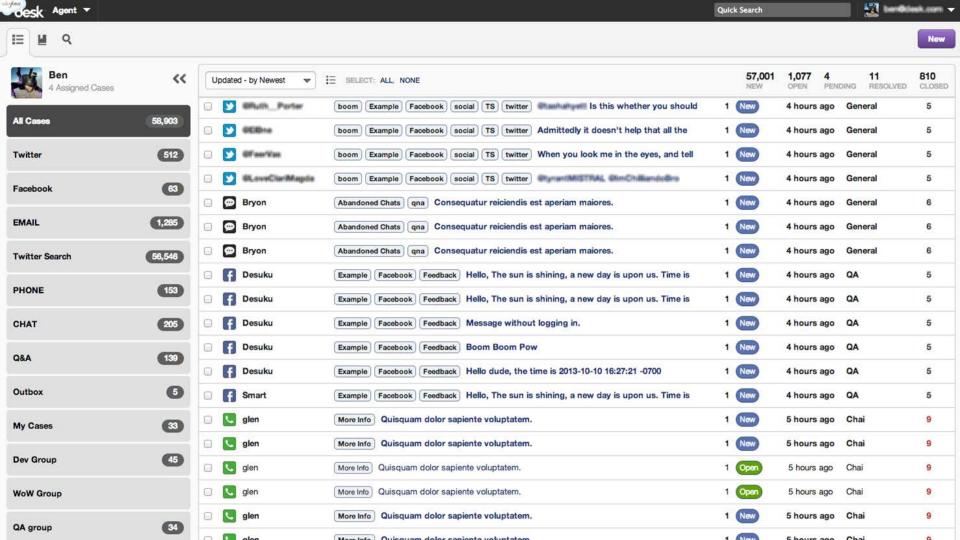






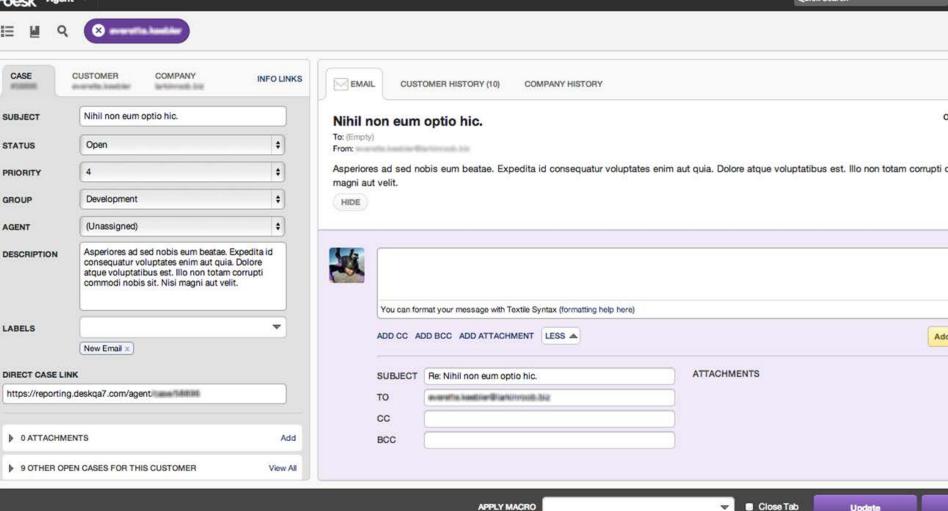


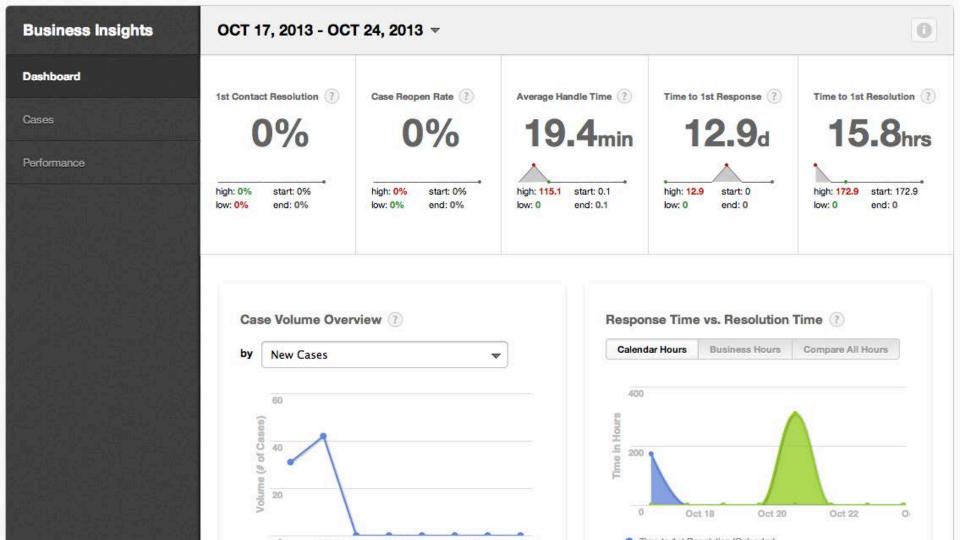


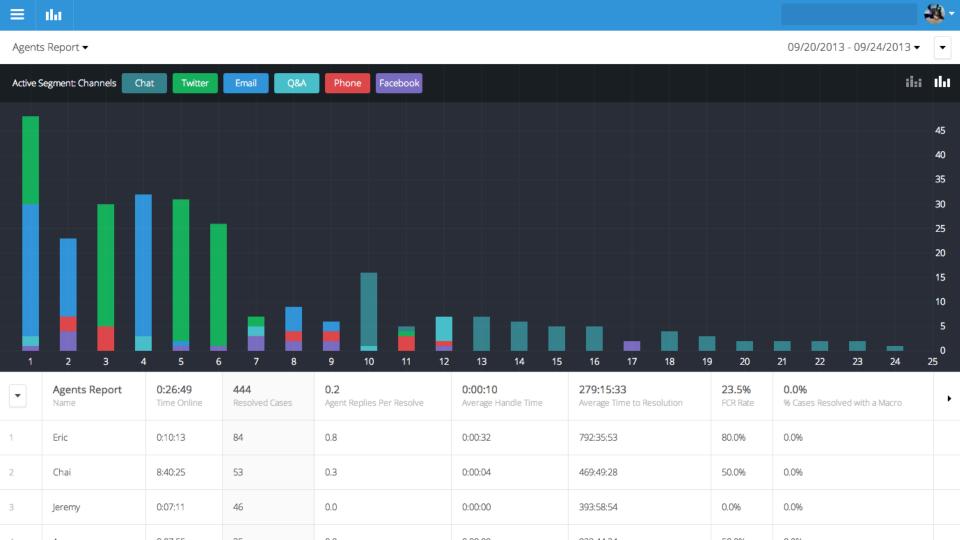


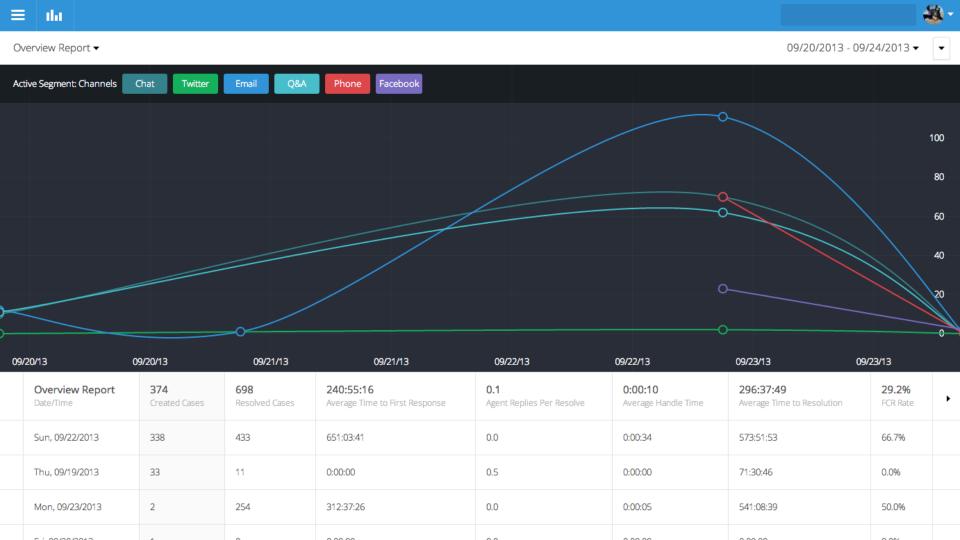


Update









# **Our Data**

- Primarily raw events
  - E.g., 'case closed', 'case received', 'case resolved'

Nearly 3B event rows

About 200K+ new events per hour



# **Amazon Redshift Environment**

Read Cluster: 2+1 Node 8XL

• Write Cluser: 2+1 Node XL

< 1TB disk space used (compression ftw!)</li>

High CPU and I/O utilization



User-facing portal

#### Requirements:

- All queries must execute in < 30 seconds</li>
- Target is < 5 seconds</li>



## **Technique 1: Less Processing at Query Time**

- Generate events from events to simplify queries
- Example: case\_first\_open from case\_open
- However, increases resources consumed and time taken by hourly data updates



```
SELECT
  DATE_TRUNC('day', e. created_at) AS t,
  CASE WHEN event_type =
                                       1 THEN
      COUNT(*) ELSE NULL END AS case_opens,
  CASE WHEN event_type = /*case_first_open*/2 THEN
      COUNT(*) ELSE NULL END AS case_first_opens
FROM events e
WHERE ...
```



#### Technique 2: Read + Write Cluster

- Originally had single cluster
- Each hourly data update used significant CPU + I/O and ran for ~30mins
- Huge impact on query performance



Technique 2: Read + Write Cluster

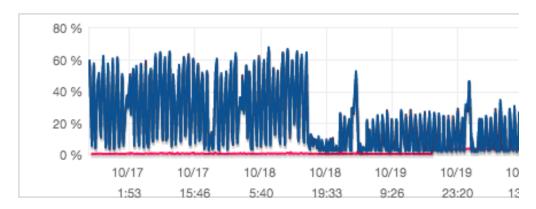
 Added a 'Write' cluster just for processing data updates

MySQL -> S3 -> 'Write' Cluster -> S3 -> 'Read' Cluster



#### Technique 2: Read + Write Cluster

#### **CPU Utilization**

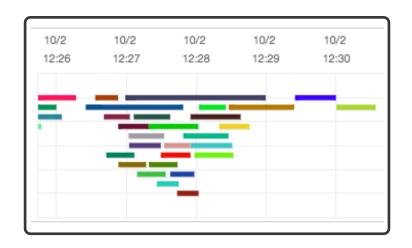




# **Concurrency Challenges**

 Queries execute when endusers load reports

- Max of 15 concurrent queries in Amazon Redshift
- Single user rapidly hitting refresh could have big impact

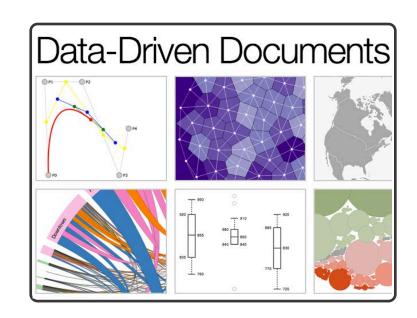




# Solution: Larger Datasets, Fewer Queries

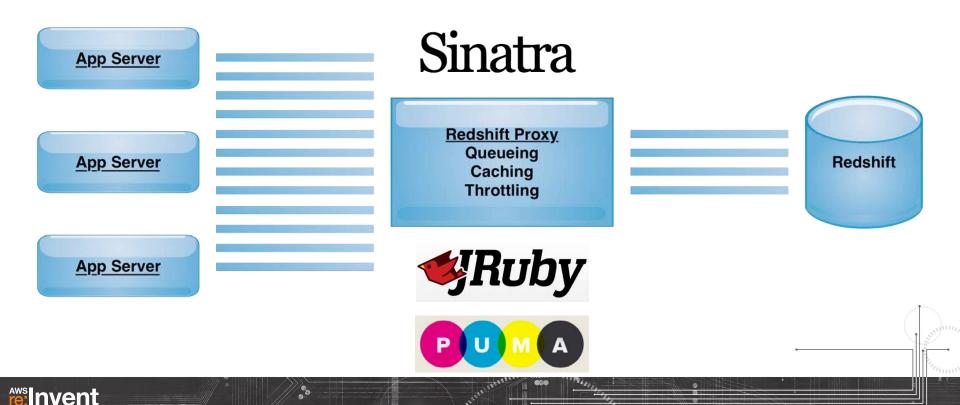
D3 allows binding MBs of data in browser

 Paging, sorting, filtering is done client-side – instantly





# Solution: Amazon Redshift Proxy



# **Amazon Redshift Proxy: Queueing**

- Queue requests outside Amazon Redshift
- Minimize (or control) contention
- Abandoned requests don't ever get to Amazon Redshift
- Future: provide ETAs and other queue statistics



# **Amazon Redshift Proxy: Caching**

- Data only updates once per hour
- Cache all reports (JSON) for duration of hour
- Every cache hit is a big win
- Just use memcached



# **Amazon Redshift Proxy: Throttling**

- We can rate limit reports on per-customer basis
- Ensures single customer cannot monopolize resources



# **Amazon Redshift Proxy: Example**

```
curl http://localhost:9292/query
    -H 'Accept: application/j son'
    -H 'Content-Type: application/j son'
    -d '{"sql":"SELECT * FROM events LIMIT
?", "params":[2]}'
```



# **Amazon Redshift Proxy: Example**

```
"header":
   ["site_id", "Int"],
   ["event_type", "Int"],
   ["created_at", "Timestamp"],
"rows"
   [1, 1 "2013-07-05T19: 01: 37-07: 00",
```



# **Summary**

- Amazon Redshift allowed us to rapidly build and ship our next-gen customer-facing reporting portal
- It's always getting better improvements and new features nearly every week
- Awesome support and guidance from AWS team



# re:Invent

Getting Maximum Performance from Amazon Redshift: Fast Data Loads

Niek Sanders, HasOffers

November 13, 2013







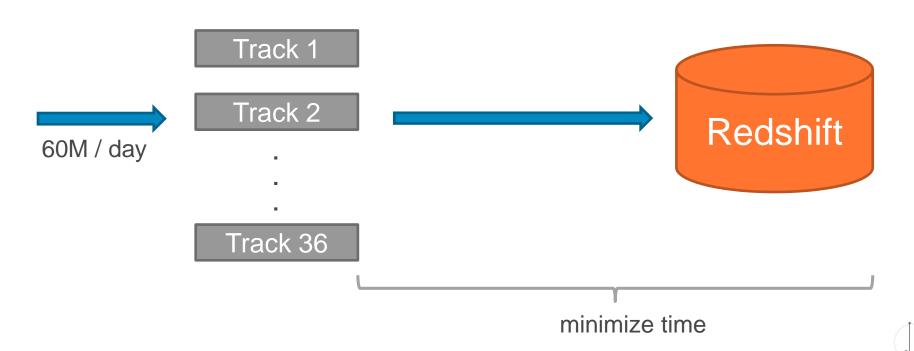
- Attribution for web & mobile marketing
- Tons of data
- Ad-hoc analysis
- Near real-time expectation



# Live(ish) Amazon Redshift Data Loading

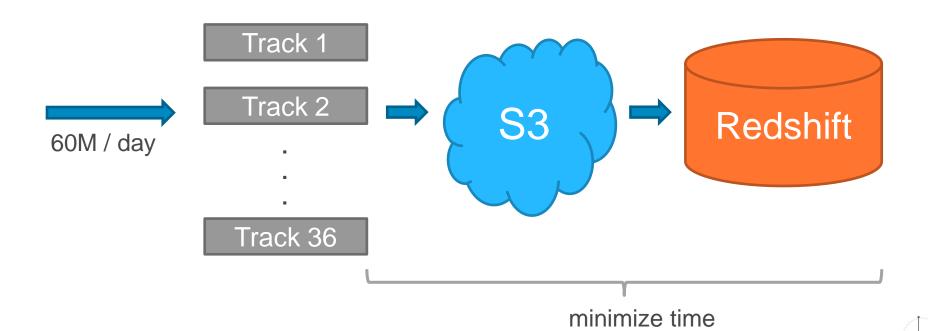


# **Objective**



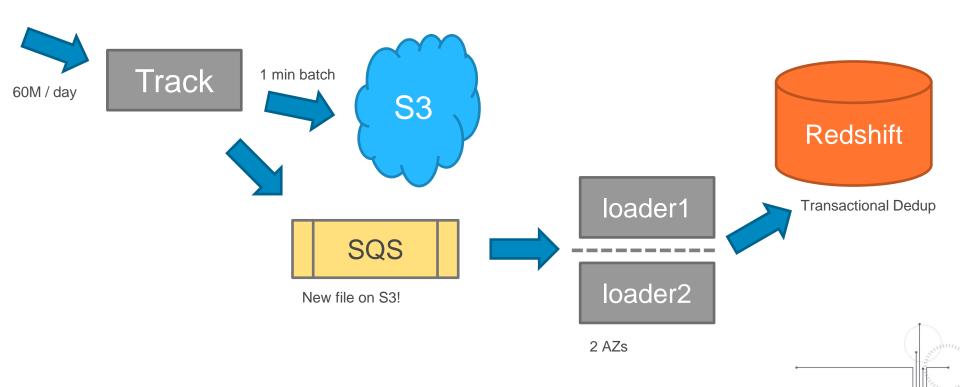


# Requires Amazon S3, DML Inserts Slow





## Core Idea





# High Availability Track Redshift loader1 loader2

- 2 loaders, separate AZs
- Shared-nothing on loaders
- Amazon SQS re-queue if loader dies
- Idempotent, transactional de-duplicator



# Four Big Ideas

- 1. Micro-batching and Amazon S3
- 2. Amazon SQS for work queue
- 3. Bulk Amazon S3 COPY
- 4. Transactional de-duplicator



## Four Big Ideas

- 1. Micro-batching and Amazon S3
- 2. Amazon SQS for work queue
- 3. Bulk Amazon S3 COPY
- 4. Transactional de-duplicator



#### **Amazon S3 COPY**

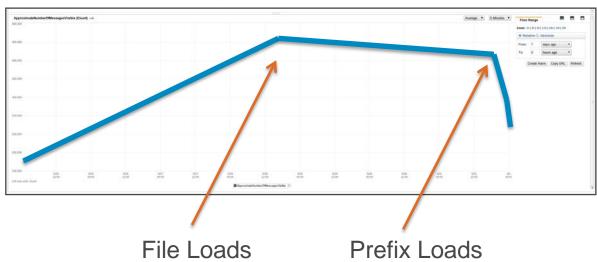
COPY clicks FROM 's3://foo/2014/01/'

Single file or path prefix.



#### **Bulk Amazon S3 COPY**

Amazon SQS Work Queue

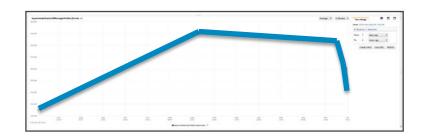




#### **Bulk Amazon S3 COPY**

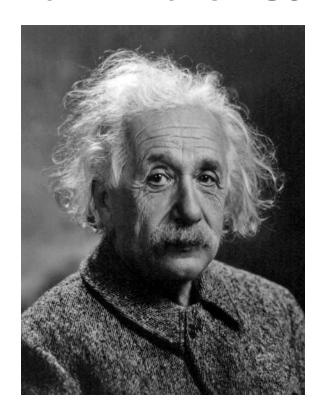
Bulk loads = 10x speed

- Cluster size, parallelism
  - Roughly equal file size
  - XL nodes → 2 files per node
  - 8XL nodes → 16 files per node





#### **Bulk Amazon S3 COPY: Common Prefixes**



s3://foo/customerA/2013/03/clicks81.gz s3://foo/customerB/2013/03/clicks93.gz s3://foo/customerB/2013/03/clicks30.gz s3://foo/customerC/2013/03/clicks01.gz



s3://foo/common\_prefix/



#### **Bulk Amazon S3 COPY: Common Prefixes**





## Four Big Ideas

- 1. Micro-batching and Amazon S3
- 2. Amazon SQS for work queue
- 3. Bulk Amazon S3 COPY
- 4. Transactional de-duplicator



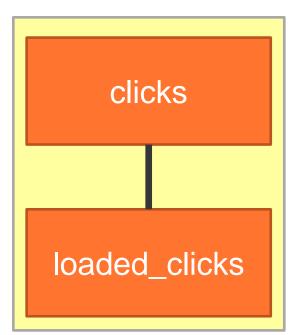


# Transactional De-duplicator SQS | loader2 | Redshift | Redshift

- Prevent duplicate loads
- Redshift transactions
- Idempotent load flow



### **Transactional Deduplicator**



- Batch in loaded\_clicks? Stop
- Begin txn
  - COPY
  - Batch in loaded\_clicks? Rollback
  - Insert batch to loaded\_clicks
- Commit



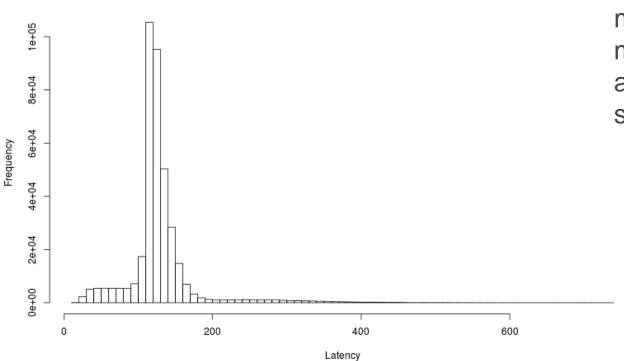
## Four Big Ideas

- 1. Micro-batching and Amazon S3
- 2. Amazon SQS for work queue
- 3. Bulk Amazon S3 COPY
- 4. Transactional de-duplicator



#### Results





min = 15

max = 733

avg =  $129 \sec$ 

stddev = 47







niek@hasoffers.com

# Thank you!



# re:Invent

Getting Maximum Performance from Amazon Redshift: Complex Queries

Timon Karnezos, Aggregate Knowledge

November 13, 2013



#### Study: Facebook Leads to 24% Sales Boost

Social network brings new cost, per Aggregate Knowl

channel to see a user before they convert. That last-click attribution, for better or worse, is what helped Google establish a multibillion dollar advertising business. "Where Facebook gets last-touch credit is where it's the only place on the planet that reached that user," Jakubowski said. Facebook tends to factor in a couple days or weeks before a user converts, he noted. Smallwood emphasized that marketers shouldn't look at Facebook in silo but consider an entire media mix and attribute success through multi-touch attribution.

# Meet the new boss

Multi-touch Attribution



#### Examples and real world applications [edit]

Data shows that a large percentage of users using a certain eCommerce platform found it by searching for "Thai food" on Google Food" page and then logged off without placing an order. Looking at each of these events as separate data points does not repre However, viewing these data points as a representation of overall user behavior enables one to interpolate how and why users at

Behavioral analytics looks at all site traffic and page views as a timeline of connected events that did not lead to orders. Since r disconnect between what they are searching for on Google and what the "Asian Food" page displays. Knowing this, a quick look prominently and thus people do not think it is actually offered, even though it is.

Behavioral analytics is becoming increasingly popular in commercial environments. Amazon.com is a leader in using behavioral

# Same as the old boss

Behavioral Analytics



#### Examples [edit source]

Market basket analysis might tell a retailer that customers often purchase shampoo and conditioner together, so putting both items on promotion at the same time would not create a significant increase in profit, while a promotion involving just one of the items would likely drive sales of the other.

Market basket analysis may provide the retailer with information to understand the purchase behavior of a buyer. This information will enable the retailer to understand the buyer's needs and rewrite the store's layout accordingly, develop cross-promotional programs, or even capture new buyers (much like the cross-selling concept). An apocryphal early illustrative example for this was when one super market chain discovered in its analysis that customers that bought diapers often bought beer as well, have

# Same as the old boss

Market Basket Analysis







# We know how to do this in SQL\*!

\* SQL:2003



#### Here it is

```
SELECT record_date, user_id, action,
       site, revenue,
       SUM(1) OVER
             (PARTITION BY user_id
              ORDER BY record_date ASC)
       AS position
FROM user_activities;
```



# So why is MTA hard?



# "Web Scale"

#### Queries

- > 30 queries
- ➤ 1700 lines of SQL
- ➤ 20+ logical phases
- > GBs of output

#### **Data**

- > ~109 daily impressions
- > ~10<sup>7</sup> daily conversions
- > ~10<sup>4</sup> daily sites
- > x 90 days

per report



# So, how do we deliver complex reports over "web scale" data?

(Pssst. The answer's Amazon Redshift. Thanks, AWS.)



# Write (good) queries

Organize the data

Optimize for the humans



# Write (good) queries

Remember: SQL is code



# Software engineering rigor applies to SQL

**Factored** 

Concise

Tested



# **Common Table Expression**



Factored
Concise
Tested

```
ON fp.inventory_placement_id = i.inventory_placement_id_AND
             fp.campaign id = i.campaign id
    30IN campaign_metadata_2013_07_17 campaign_metadata
    JOIN tracking_campaign_metadata_2013_07_17 tracking_campaign_metadata
          ON tracking campaign metadata.campaign id = campaign metadata.campaign id
   JOIN inventory_placement_metadata_2013_07_17 inventory_placement_metadata
          ON inventory placement metadata.tracking data_provider_id = tracking_campaign_me
inventory_placement_metadata.inventory_placement_id = 1.inventory_placement_i
                                                                                                dadata.tracking_data_proylder_id AND
    JOIN derived_inventory_provider_name_lookup_2013_07_17 inventory_provider_name_lookup
          ON inventory_provider_name_lookup.inventory_provider_name = inventory_placemen
                                                                                                metadata.inventory.
 SELECT record_date,
          request_id,
          c.advertiser_id AS advertiser_id,
          attributed revenue
 FROM conversions unified c
 JOIN derived filter conversions 2013 07 17 fc ON fc.tag id = c.tag id AND
          DATE_TRUNC('day', c.record_date)
                                                         AS conversion_record_date,
                                                         AS ak_user_id,
          c.request_id
                                                         AS conversion_request_id,
          i.request_id
                                                         AS impression_request_id,
          1. Inventory provider name surrogate id
                                                        AS inventory provider name surrogate id,
         c.attributed revenue
                                                         AS conversion attributed revenue,
            - DATEDIFF doesn't need a DATE_TRUNC bec
          DATEDIFF(day, i.record_date, c.record_date) AS day_offset,
          SUM(I) OVER (PARTITION BY 1, advertiser id, 1, ak user id, c, request id ORDER BY 1, record date DESC ROWS UNBOUNDED PRECEDING) AS position
 JOIN targeted_conversions c
          ON f.ak_user_id = c.ak_user_id AND
              1.advertiser_id = c.advertiser_id AND
             1.record_date < c.record_date AND
1.record_date > (DATE_TRUNC('day', c.record_date) = interval '45 days')
SELECT . FROM chains;
```



### Window functions



- -- Position in timeline
  SUM(1) OVER (PARTITION BY user\_id
  ORDER BY record\_date DESC ROWS UNBOUNDED PRECEDING)
- -- Event count in timeline

  SUM(1) OVER (PARTITION BY user\_id

  ORDER BY record\_date DESC BETWEEN UNBOUNDED PRECEDING AND UNBOUNDED FOLLOWING)
- -- Transition matrix of sites

  LAG(site\_name) OVER (PARTITION BY user\_id

  ORDER BY record\_date DESC)
- -- Unique sites in timeline, up to now

  COUNT(DISTINCT site\_name) OVER (PARTITION BY user\_id

  ORDER BY record\_date DESC

  ROWS UNBOUNDED PRECEDING)



#### Window functions

Scalable, combinable

Compact but expressive

Simple to reason about



# Organize the data



#### Leverage Amazon Redshift's MPP roots

Fast, columnar scans, 10

Fast sort and load

Effective when work is distributable



#### Leverage Amazon Redshift's MPP roots

Sort into multiple representations

Materialize shared views

Hash-partition by user\_id



# Optimize for the humans



#### Operations should not be the bottleneck

Develop without fear

Trade time for money

Scale with impunity



#### Operations should not be the bottleneck

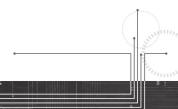
Fast Amazon S3 = scratch space for cheap

Linear query scaling = GTM quicker

Dashboard Ops = dev/QA envs, marts, clusters with just a click



# But, be frugal





## Quantify and control costs

Test across different hardware, clusters

Shut down clusters often

Buy productivity, not bragging rights



# Thank you!

#### References

http://bit.ly/rs\_ak

http://www.adweek.com/news/technology/study-facebook-leads-24-sales-boost-146716

http://en.wikipedia.org/wiki/Behavioral\_analytics

http://en.wikipedia.org/wiki/Market\_basket\_analysis



#### Amazon Redshift Customers at re:Invent

DAT306 - How Amazon.com is Leveraging Amazon Redshift

Thursday 11/14: 3pm in Murano 3303



DAT205 - Amazon Redshift in Action: Enterprise, Big Data, SaaS Use Cases

Friday 11/15: 9am in Lido 3006







# re:Invent

Please give us your feedback on this presentation

#### **DAT305**

As a thank you, we will select prize winners daily for completed surveys!



