Large Data Methods: Keeping it Simple on the Path to Big Data

Big Data Business Forum,
San Francisco, November 13, 2012
Jim Porzak, Sr. Dir. Business Intelligence, Minted

What we will cover:

- 1. Who is Minted?
- 2. Our large data challenges.
- 3. Large data solutions.
- 4. Migration to "big data."
- 5. Discussion

About Minted



- A social commerce site.
- Crowd-sourcing graphic designs and art from a global design community.
- Selling those as printed paper products.
- Initially focused on the \$10 billion stationery and \$48 billion art markets.
- Combining community with commerce.
- Built on stellar technology, operations, and customer service.

Minted.com Architecture

~ Classic LAMP
Integrated back office: "MBO"
MySQL holds site & MBO data

Minted Bl Architecture

On Amazon EC2:

- Replicated MySQL site DB
- PostgreSQL BI DB
- Tableau Server

In support of marketing:

Our job is to understand...
the customer,
the whole customer, and
nothing but the customer.

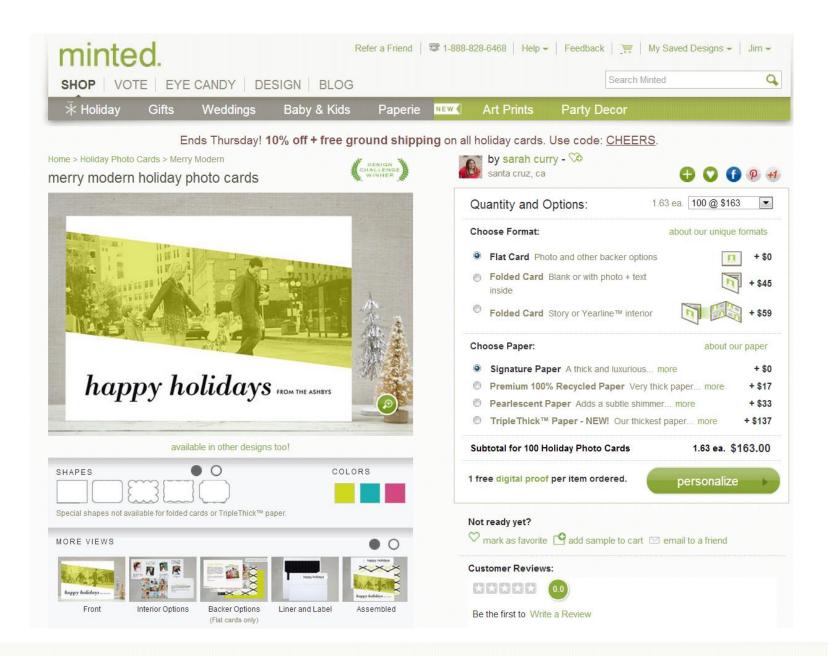
Customer Initiated Actions

CIA's are:

- Ordering
- Other Minted.com actions
- Responding to Minted outreach
 - Email opens, reads, clicks
- Contacting Minted
- Social Behavior
- And more!

Ordering: "How do you sell bread?"

- 1. "You tell me what you want."
- 2. "I give you the bread."
- 3. "I tell you how much it costs."
- 4. "You give me the money."



Other Site CIA's

- 1. Google Analytics
 - Not for individual visitor!
 - f(tagging(t))
- 2. "App-request" logs
 - CIA's & some 2nd level
 - Visitor (cookie) & user ID's
 - 12 months of history

Responses to email:

- 1. Sends
 - Transactional: order ack, ship, ...
 - Marketing: retention, offers
 - Targeted & personalized
- 2. CIA's
 - Bounce, open, click, buy
 - Opt-outs, opt-ins

Other data sources:

- Convertro
- Survey Tools
- Demographic Appends

Roadmap to "Big Data"

- Large Data:
 - PostgreSQL
- Big Data:
 - Some columnar DB
- Bigger Data:
 - Some map-reduce platform

Why PostgreSQL?

- Open source (~free)
- SQL for analytics
 - Window functions, etc.
- Known to scale
- Foundation for many of the columnar DB products

BI in PostgreSQL philosophy

- Data structures as if columnar
 - Big & wide; not star
- Focus on high impact first:
 - Orders
 - Order Detail
 - Customers
 - Site Sessions

Primary wide tables

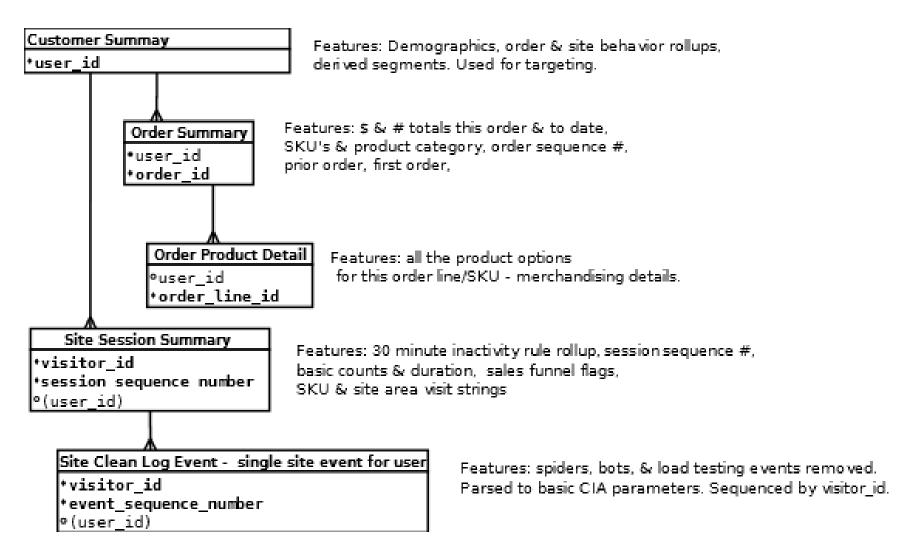


Table Details

- Customer: 35 columns
 - ID, source, acquisition date/source, purchase profiles, site profiles, demo profiles
- Order Summary: 72 columns
 - ID's, date, order seq #, gap, \$'s, #'s, flags, top, prior, first product code/group/class, to-date \$'s by class, geo, source.
- Order Product Detail: 27 columns
 - ID's, timestamp, SKU, \$, #, promo, details of options
- Site Sessions: 27 columns
 - ID's, seq #, # events, duration, timestamps, gap, funnel flags, products, actions, sources, media, campaigns, ...
- Site Clean Events: 21 columns
 - ID's, timestamp, ip, seq #, interval to prior, entry/exit actions, source/medium/campaign, sku, ...

Performance

- Queries off of these tables very fast; typically sub-minute for even the most complex.
- Tableau server has internal columnar engine for interactive analytics performance.
- Nightly refresh & updates in under three hours with no attempt at tuning.

Next Steps:

- Finalize logical design in PostgreSQL based on needs of our business partners over next few months.
- Tune PostgreSQL platform test limits of scale. Estimate when we will need to move to next level. In meantime:
 - POC's on a couple of columnar DBs.
- Migrate to final columnar DB.

Questions? Comments?

Now would be the time!



APPENDIX

Application request logs – deep dive



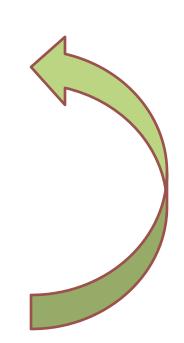
What's an app-request log?

{"time start":1313620339.85, "time end":1313620340.01, "request": {"headers": [["Host", "localhost: 8888"], ["Connection", "keep-alive"],["Cache-Control","max-age=0"],["User-Agent","Mozilla\/5.0 (Macintosh; Intel Mac OS X 10 6 8) AppleWebKitV534.30 (KHTML, like Gecko) ChromeV12.0.742.112 SafariV534.30"],["Accept","textVhtml,applicationVxhtml+xml,applicationVxml;q=0.9,*V*;q=0.8"],["Accept-Encoding", "gzip, deflate, sdch"], ["Accept-Language", "en-US, en; g=0.8"], ["Accept-Charset", "ISO-8859-1, utf-8;q=0.7,*;q=0.3"]],"method":"GET","remote_addr":"127.0.0.1","protocol":"HTTPV1.1","uri":"Vregister"},"response":{"stat us":null,"headers":[["Expires","Sun, 19 Nov 1978 05:00:00 GMT"],["Last-Modified","Wed, 17 Aug 2011 22:32:19 GMT"],["Cache-Control","store, no-cache, must-revalidate"],["Cache-Control","post-check=0, pre-check=0"],["Content-Type", "text\/html; charset=utf-8"], ["X-Powered-By", "PHP\/5.3.2; Qcodo\/0.3.24 (Qcodo Beta 3)"], ["Set-Cookie", "minted tr=UQ%BDn%830%10%7E%17%EF%04%8C%9D41S%D4%A1%EA%90%8C%5D%91%03%26% B5%0Ag%E5%3B%90h%94w%EF%9D%0BR3%FA%BE%DF%3B%5B%23%B5%B9%83QF4%16EeMa%EE%8F% F4%9E%5C%14%957%B2%02%B3Oh%0D%7D%40%1E%95%5BEC%29%8D%E8%7C%04%AC%27%0F%3E%01 %B2%D4%3B%BD%D7%07%A9%19%2F%8C%E8%ED%13%AC%A4%DA%95%85%D2%05%C3z%95G%D7%B9 %189%0D%8C%A6%E0%8BC%AB6%83%BF%A1k7M%18%F2f%04%0C%83%FFq1%87%AF1%3F%BD%9F%B3 %E3%DBkv%FA8%B2D%AA%25%E7%BF%0Fe%27%AC%5CC+%8C%B1g%A2%3A%2F%CD%93%E2iH%D4%D 44%C4%1A%E7o%27%96%BDS%F6%9F%29%27%AD%94%86%0C%80%EFU%F2%F9%B6%04%04d%D6e%3D X%EB%C0 o%EB%88NJ%FEC%CF%0A%C9%8A%17ftv%EC%D3e%0A%26%1C%D6%AE%E9%27H%40.C%C0 O%17%21k%ED%9C%5D%7D%87%90%01Z%F4%81%8C%E7%A5a%DAd%91pC%B0%93%CBHE%1A%A4-%1E%BF; expires=Mon, 13-Feb-2012 22:32:20 GMT; path=V"]]}}



Making app-requests useful

- 1. Parse to .csv (Python)
- 2. Load to BI PostgreSQL DB
- 3. Clean & more parsing
- 4. Sessionize
- 5. Analyze



Sessionization in PostgreSQL Part 1

```
-- get event sequence #s & seconds after prior event
CREATE TABLE v sessions1 AS
SELECT *,
       ROW NUMBER() OVER(Members) AS event seq number,
       event at - LAG(event at) OVER(Members)
          AS interval to prior
  FROM v events
WINDOW Members
    AS (PARTITION BY member_id -- unique member ID
            ORDER BY event at -- timestamp of event
```

Sessionization in PostgreSQL Part 2

```
-- update with session sequence #
CREATE TABLE v sessions2 AS
SELECT *,
       COUNT (CASE WHEN interval to prior IS NULL OR
                       interval to prior > '30 minutes'
                  THEN 1 ELSE NULL END) OVER (Members)
           AS session seq number
 FROM v sessions1
WINDOW Members
   AS (PARTITION BY member id -- unique member ID
            ORDER BY event_seq number -- Session #
```

Sessionization in PostgreSQL Part 3

```
-- now roll up into sessions getting session start, total time in session,
-- site areas explored, other site specific rollups
CREATE TABLE v sessions AS
SELECT member id,
       session seq number,
      MIN(event at) AS session start at,
      COUNT(*) AS num events in session,
       SUM (CASE WHEN interval to prior > '30 minutes'
                THEN NULL ELSE interval to prior END) AS session_duration,
       STRING AGG(DISCRETE site area, ', ') AS site areas visited,
      <other site specific aggregations>
  FROM v sessions2
GROUP BY member id,
          session seq number
ORDER BY member id,
          session seq number
```

How "mining" app-request logs helps us understand the customer:

- 1. Sales funnel analysis by product & YOY.
- 2. Customer's individual interests, preferences, ...
- 3. Customer's evolution in relation w/ Minted
- 4. Usage based customer segments
- 5. And we will discover more!