# Evolution of Big Data Architectures@ Facebook

Architecture Summit, Shenzhen, August 2012
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#### About Me

- Currently Co-founder/CEO of Qubole
- Ran the Data Infrastructure Team at Facebook till 2011
- Co-founded Apache Hive @ Facebook

#### Outline

- Big Data @ Facebook Scope & Scale
- Evolution of Big Data Architectures @ FB
- Qubole

### Big Data @ FB(2011): Scale

- 25 PB of compressed data ~ 150 PB of uncompressed data
- 400 TB/day (uncompressed) of new data
- I new job every second

### Big Data @ FB: Scope

- Simple reporting
- Model generation
- Adhoc analysis + data science
- Index generation
- Many many others...

### A/B Testing Email #1

#### facebook

Hi Denise,

You haven't been back to Facebook recently. Here are just a few things that have been happening while you were gone:

Sign in to Facebook and start connecting

Sign In



#### John Pingel







Tara Peters is at work, and sleepy

April 22 🕎 1



Eric Martin yay les gigantes killing peavy and the sharks staying alive. It's earth day too bitches, take the train.

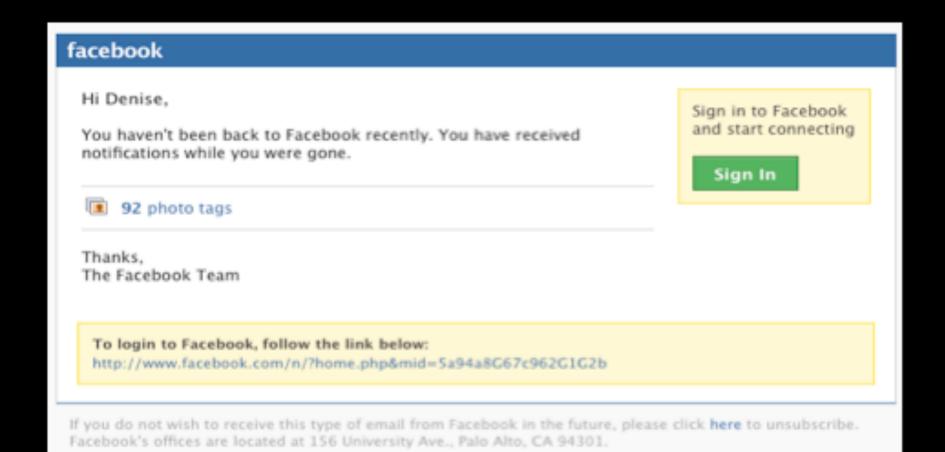
April 22 🖒 2 🖵 3

Thanks, The Facebook Team

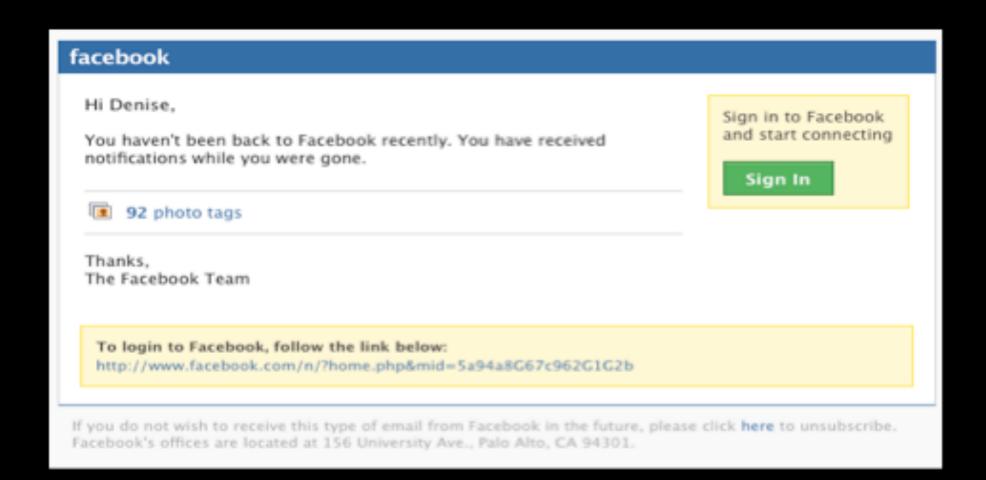
To login to Facebook, follow the link below:

http://www.facebook.com/n/?home.php&mid=5a94e3G67c962G1G2b

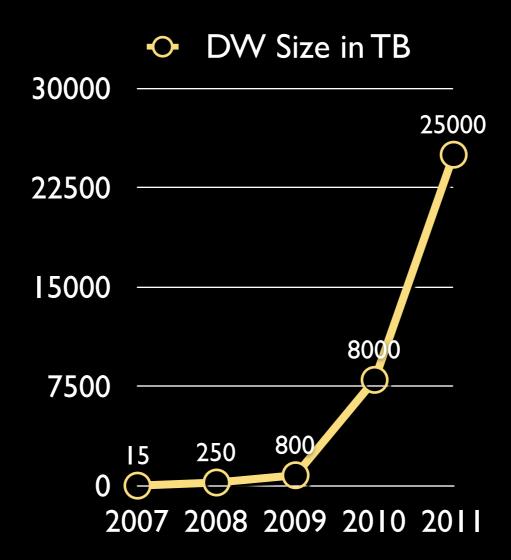
### A/B Testing Email #2



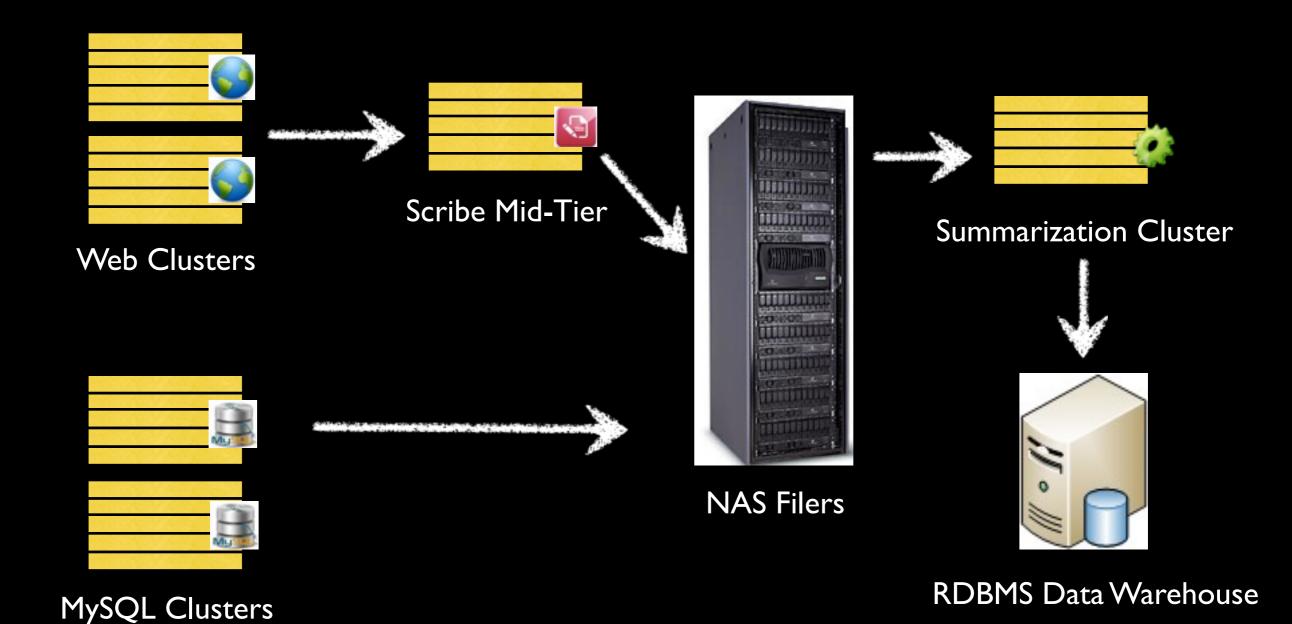
### A/B Testing Email #2 is 3x Better



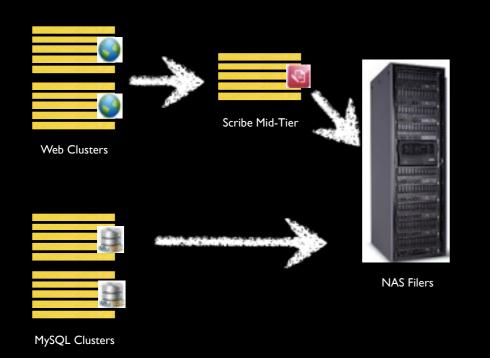
### Evolution: 2007-2011



#### 2007: Traditional EDW



#### 2007: Pain Points



 compute close to storage (early map/reduce)



Summarization Cluster



- daily ETL > 24 hours
- Lots of tuning/indexes etc.
- Lots of hardware planning



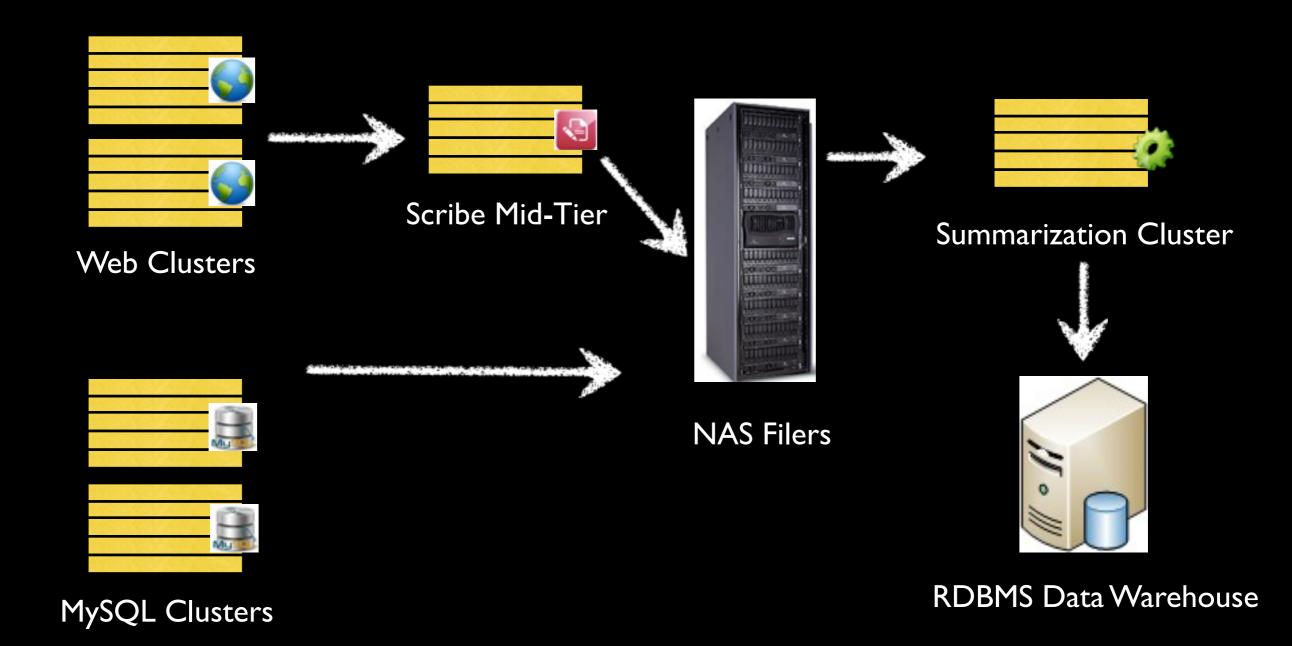
**RDBMS** Data Warehouse

#### 2007: Limitations

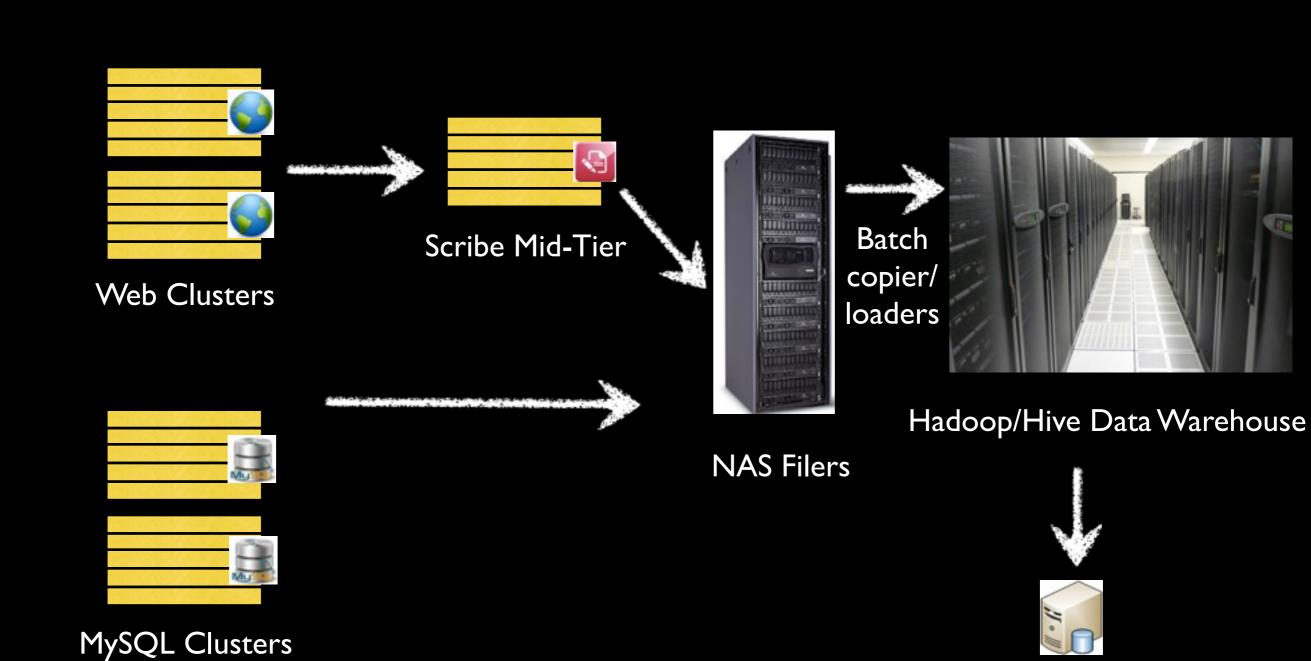
- Most use cases were in business metrics data science, model building etc. not possible
- Only summary data was stored online details archived away



### 2008: Move to Hadoop



### 2008: Move to Hadoop



**RDBMS** Data Mart

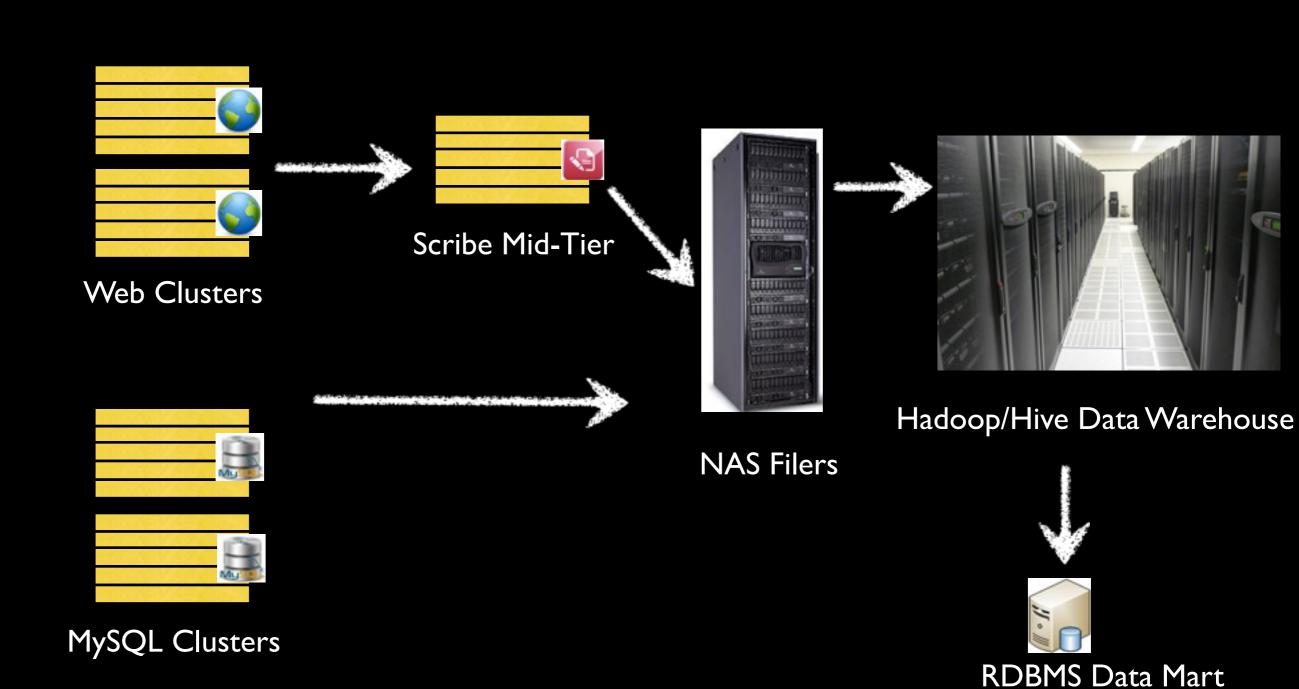
#### 2008: Immediate Pros

- Data science at scale became possible
- For the first time all of the instrumented data could be held online



Use cases expanded

### 2009: Democratizing Data



### 2009: Democratizing Data

Databee & Chronos: Data Pipeline 
Framework

HiPal: Adhoc Queries + Data Discovery

Hadoop/Hive Data Warehouse

Nectar: instrumentation & schema aware data collection

Scrapes:
Configuration
Driven

## 2009: Democratizing Data(Nectar)

- Typical Nectar Pipeline
  - Simple schema evolution built in
  - json encoded short term data
  - decomposing json for long term storage

```
This event has application name 'mobilelog' and app event type
  NOTE: Make sure you use only one application name per new application.
 Also, app event type should not have any special characters or spaces,
 use underscores instead. Ssampling_rate is the scribe sampling rate and
 has a value between 0 and 100 - sampling is on userid
NectarAppSpecificEvent('mobilelog', 'email_mms_upload', $sampling_rate)
 ->addToOdsKeys(array('k1', 'k2'))
                                          // if you want to add additional
                                          // ODS keys
 ->setODSSamplingRate(1)
                                           // default is 10000, meaning 1
                                           // in 10000 events is sent to ODS
 ->addToAppSection(array("key" => "val")) // can add different key value
                                          // pairs for different eventtypes
 ->log();
                                          // need to explicit log app
                                          // specific events
```

## 2009: Democratizing Data (Tools)

 HiPal - data discovery and query authoring



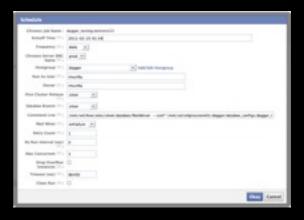
 Charting and dashboard generation tools





## 2009: Democratizing Data (Tools)

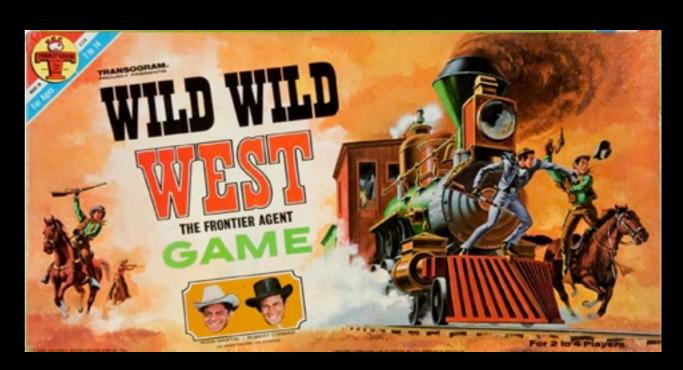
- Databee: Workflow language
- Chronos: Scheduling tool





### 2009: Cons of Democratization

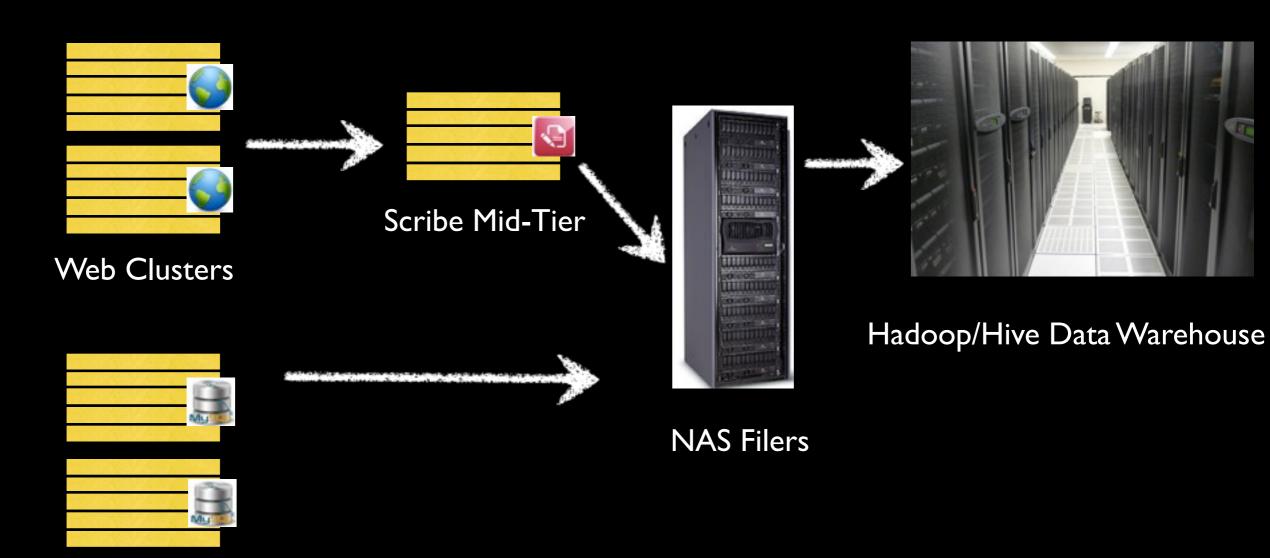
- Isolation to protect against Bad Jobs
- Fair sharing of the cluster - what is a high priority job and how to enforce it



### 2010: Controlling Chaos

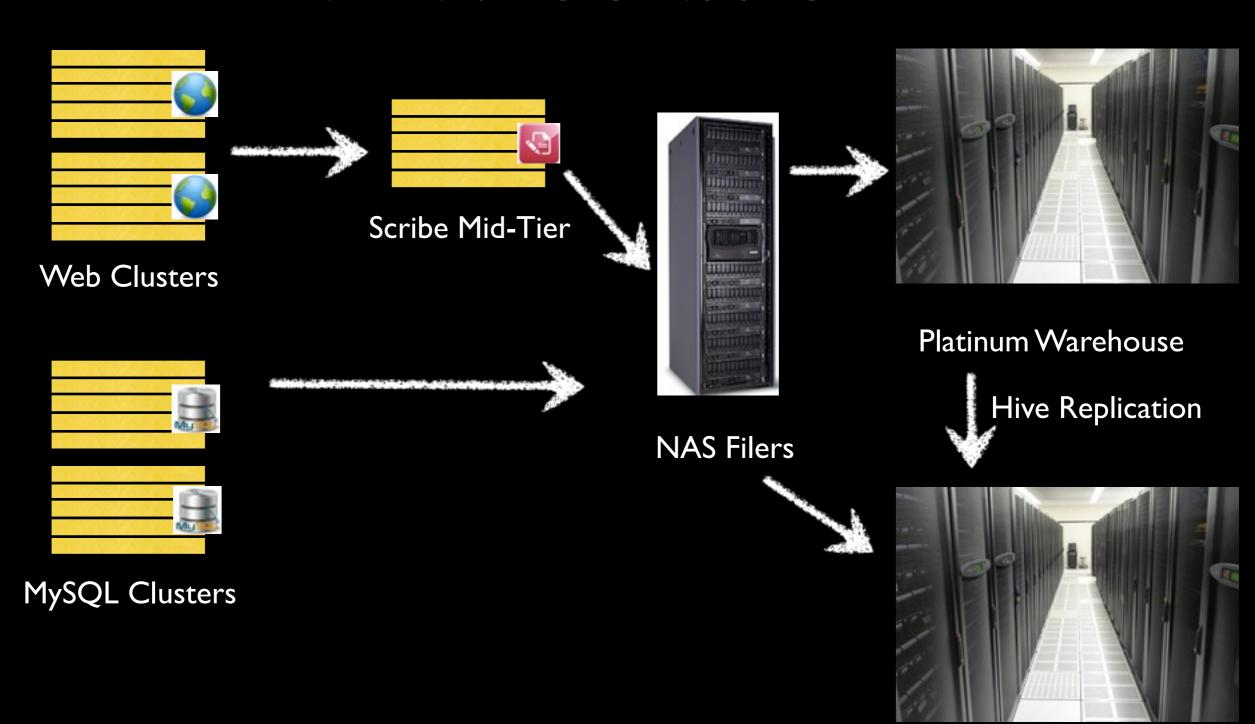
- Isolation
- Reducing operational overhead
- Better resource utilization
- Measurement, ownership, accountability

### 2010: Isolation



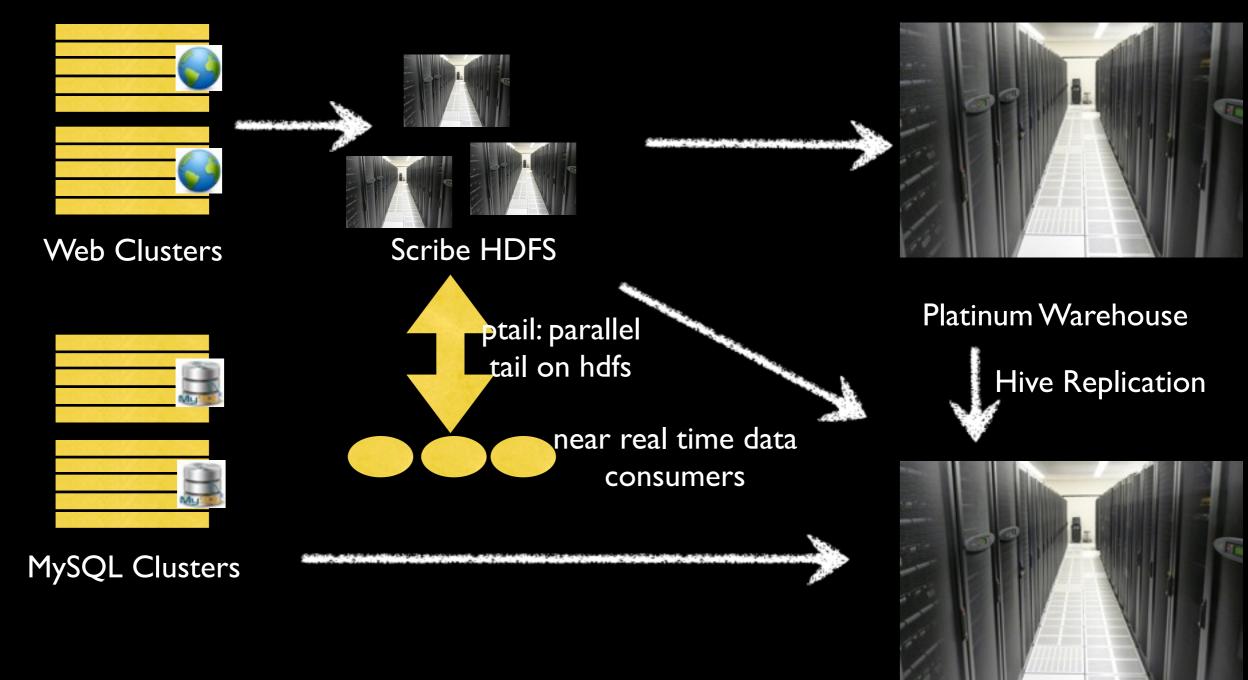
MySQL Clusters

### 2010: Isolation



Silver Warehouse

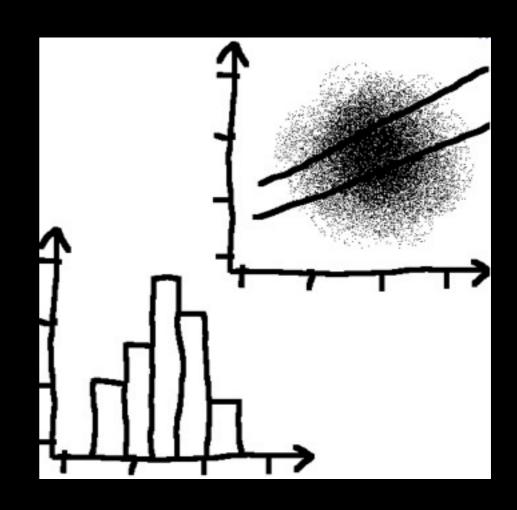
### 2010: Ops Efficiency



Silver Warehouse

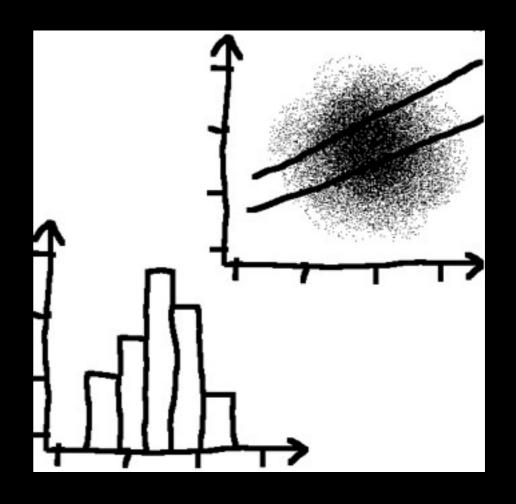
### 2010: Resource Utilization (Disk)

- HDFS-RAID: from 3 replicas to 2.2 replicas
- RCFile: Row columnar format for compressing Hive tables



### 2010: Resource Utilization (CPU)

- Continuous copier/ loaders
- Incremental scrapes
- Hive optimizations to save CPU



### 2010: Monitoring(SLAs)

- Per job statistics rolled up to owner/group/team
- Expected time of arrival vs Actual time of arrival of data
- Simple data quality metrics



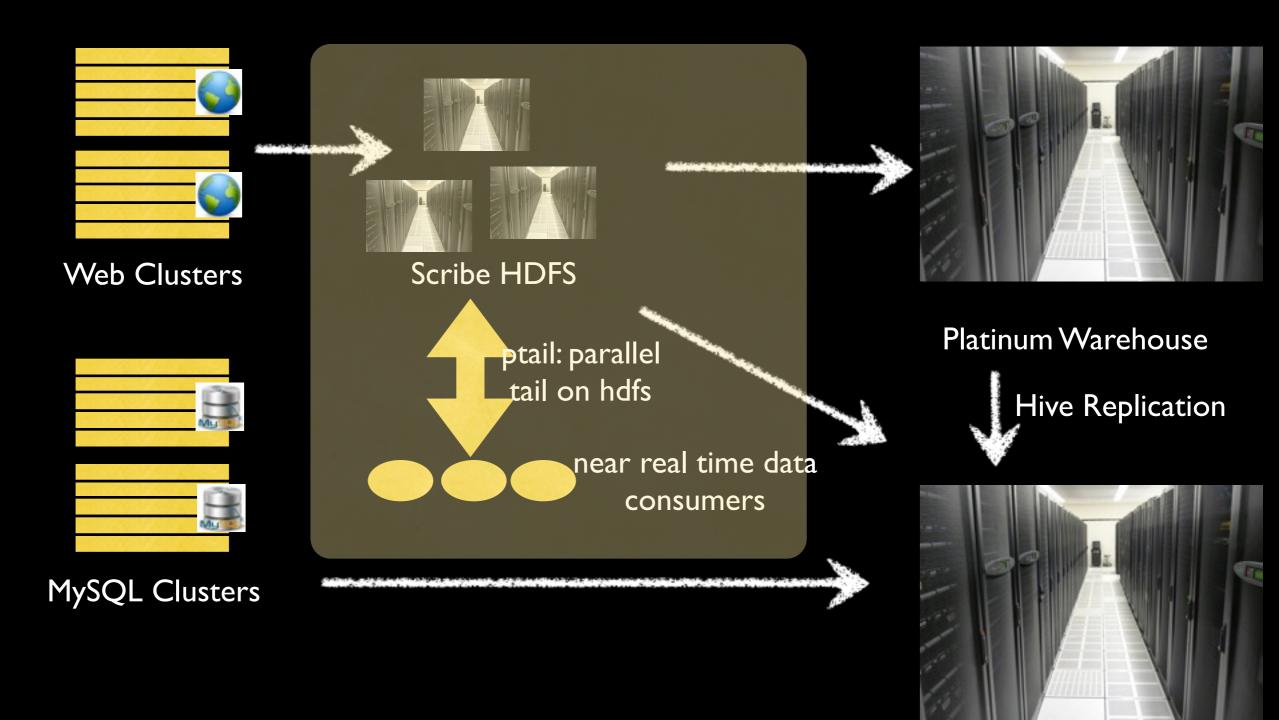
## 2011: New Requirements

- More real time requirements for aggregations
- Optimizing resource utilization

### 2011: Beyond Hadoop

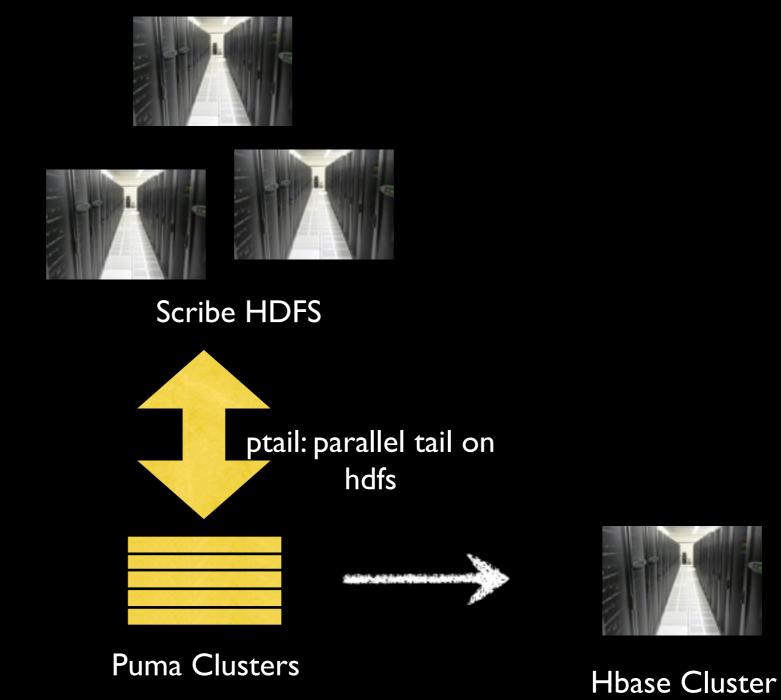
- Puma for real time analytics
- Peregrine for simple and fast queries

### 2011: Puma



Silver Warehouse

### 2011: Puma



### Some takeaways

- Operating and optimizing Data Infrastructure is a hard problem
  - Lots of components from log collection, storage, compute, query processing, tools and interfaces
  - Lots of choices within each part of the stack

### Qubole

- Mission:
  - Data Infrastructure in the Cloud made Easy, Fast and Reliable
  - We take care of operating and optimizing this infrastructure so that you can focus on your data, analysis, algorithms and building your data apps

#### Qubole - Information

- Early Trial(by invitation):
  - www.qubole.com
- Come talk to us to join a small and passionate team
  - jobs@qubole.com
- Follow us on twitter/facebook/linkedin