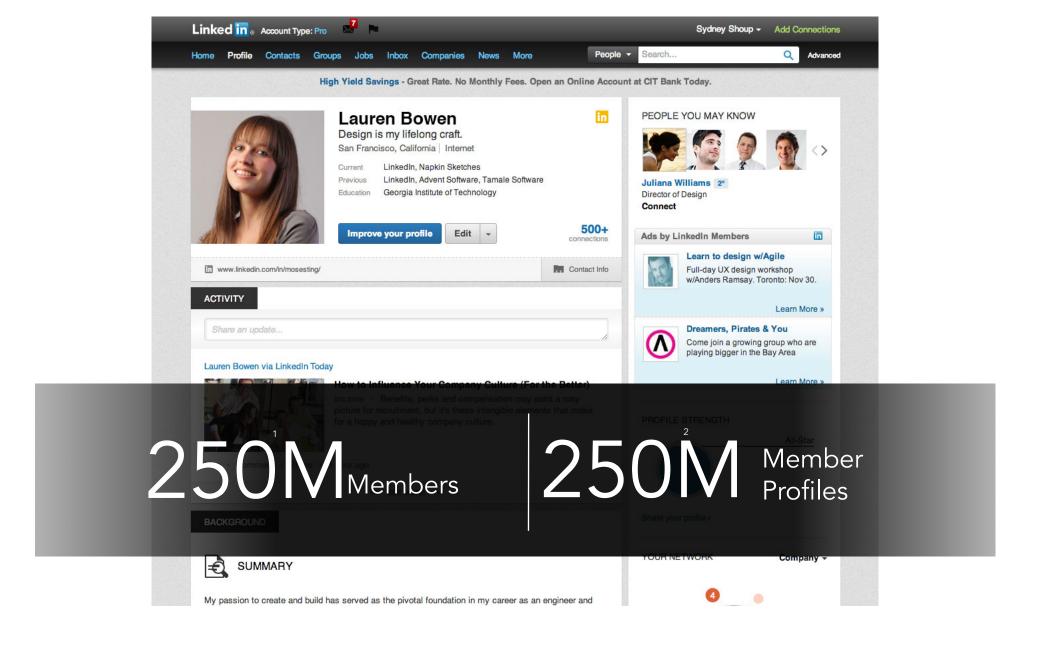


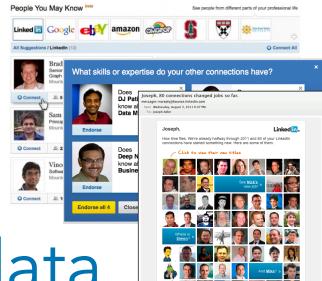
**Turning data into products** 

Sam Shah



### LinkedIn: the professional profile of record



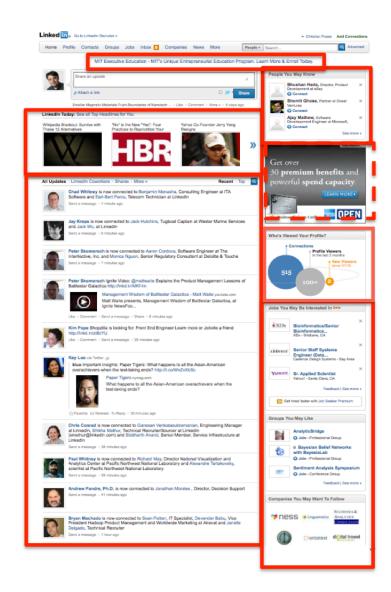


We have a lot of data.

We want to leverage this data to build products.

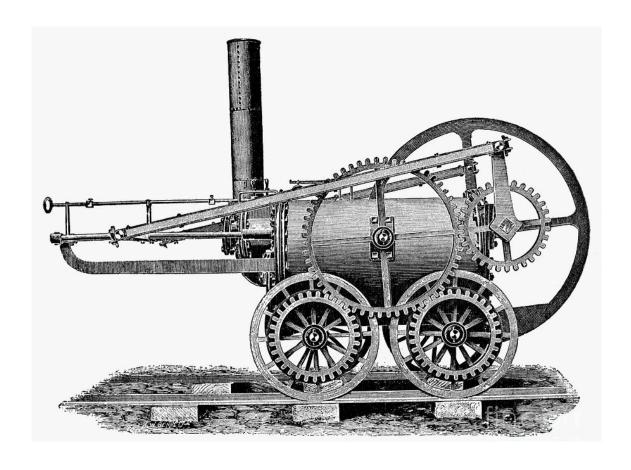


# Homepage powered by data



## We're in the "pre-industrial age" of Big Data

Need "bridges & railways"



Algorithms

#### Year in Review



- Steps to make the email
  - Collect job changers
  - Figure out who is connected to them
  - Rank job changes

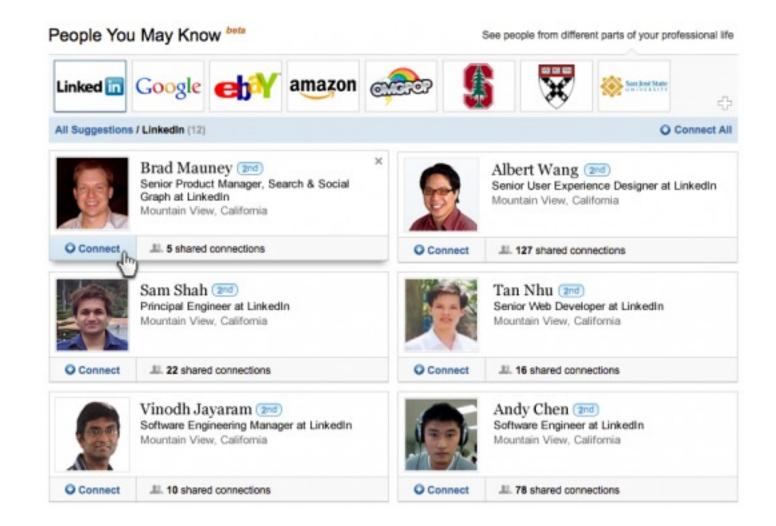
#### Example: Year in Review



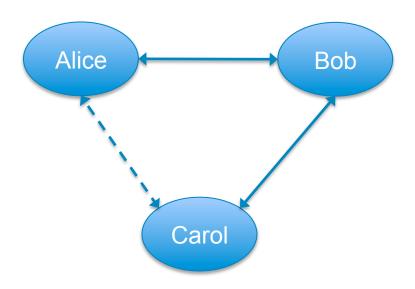
```
memberPosition = LOAD '$latest positions' USING BinaryJSON:
memberWithPositionsChangedLastYear = FOREACH (
  FILTER memberPosition BY ((start_date >= $start_date_low ) AND
    (start_date <= $start_date_high))</pre>
) GENERATE member_id, start_date, end_date;
allConnections = LOAD '$latest_bidirectional_connections' USING
BinaryJSON;
allConnectionsWithChange_nondistinct = FOREACH (
   JOIN memberWithPositionsChangedLastYear BY member_id,
   allConnections BY dest
 ) GENERATE allConnections::source AS source,
  allConnections::dest AS dest;
allConnectionsWithChange = DISTINCT
  allConnectionsWithChange_nondistinct;
memberinfowpics = LOAD '$latest memberinfowpics' USING
  BinaryJSON;
pictures = FOREACH ( FILTER memberinfowpics BY
  ((cropped picture id is not null) AND
  ( (member_picture_privacy == 'N') OR
    (member_picture_privacy == 'E')))
 ) GENERATE member_id, cropped_picture_id, first_name as
   dest_first_name, last_name as dest_last_name;
resultPic = JOIN allConnectionsWithChange BY dest, pictures
 BY member_id;
connectionsWithChangeWithPic = FOREACH resultPic GENERATE
  allConnectionsWithChange::source AS source id.
  allConnectionsWithChange::dest AS member_id,
  pictures::cropped_picture_id AS pic_id,
  pictures::dest_first_name AS dest_first_name,
  pictures::dest_last_name AS dest_last_name;
```

```
ioinResult = JOIN connectionsWithChangeWithPic BY source id.
memberinfowpics BY member_id;
withName = FOREACH joinResult GENERATE
 connectionsWithChangeWithPic::source_id AS source_id,
 connectionsWithChangeWithPic::member_id AS member_id,
 connectionsWithChangeWithPic::dest_first_name as first_name,
 connectionsWithChangeWithPic::dest_last_name as last_name,
 connectionsWithChangeWithPic::pic_id AS pic_id,
memberinfowpics::first_name AS firstName,
 memberinfowpics::last_name AS lastName.
 memberinfowpics::gmt_offset as gmt_offset,
 memberinfowpics::email_locale as email_locale,
 memberinfowpics::email_address as email_address;
resultGroup0 = GROUP withName BY (source_id, firstName,
 lastName, email_address, email_locale, gmt_offset);
-- get the count of results per recipient
resultGroupCount = FOREACH resultGroup@ GENERATE group,
withName as toomany, COUNT_STAR(withName) as num_results;
resultGroupPre = filter resultGroupCount by num_results > 2;
resultGroup = FOREACH resultGroupPre {
  withName = LIMIT toomany 64;
  GENERATE group, withName, num results:
x_in_review_pre_out = FOREACH resultGroup GENERATE
FLATTEN(group) as (source_id, firstName, lastName,
email_address, email_locale, gmt_offset),
 withName.(member_id, pic_id, first_name, last_name) as jobChanger, '2011' as changeYear:chararray,
 num_results as num_results;
x_in_review = FOREACH x_in_review_pre_out GENERATE
 source_id as recipientID, gmt_offset as gmtOffset,
firstName as first_name, lastName as last_name, email_address,
 email_locale,
 TOTUPLE( changeYear, source_id, firstName, lastName,
  num_results,jobChanger) as body;
rmf $xir:
STORE x_in_review INTO '$xir' USING BinaryJSON('recipientID');
```

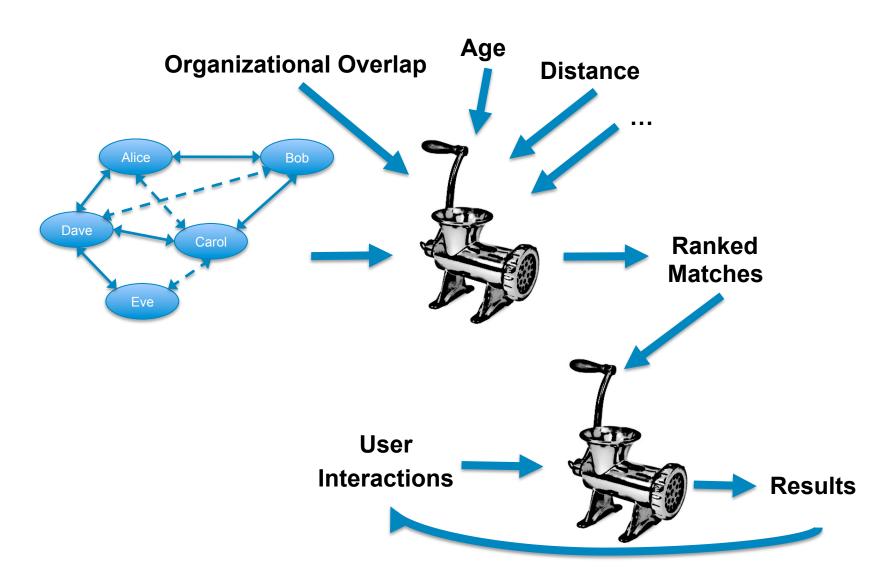
#### People You May Know



# People You May Know

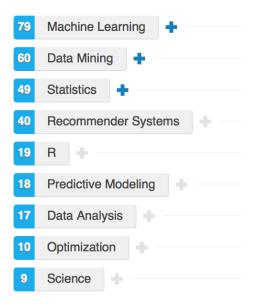


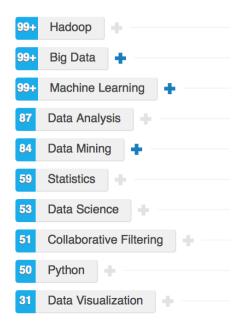
### People You May Know

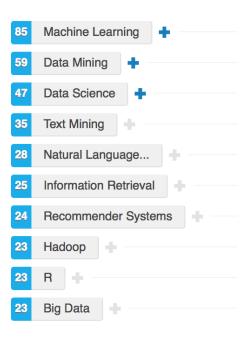


Infrastructure

### Skill sets







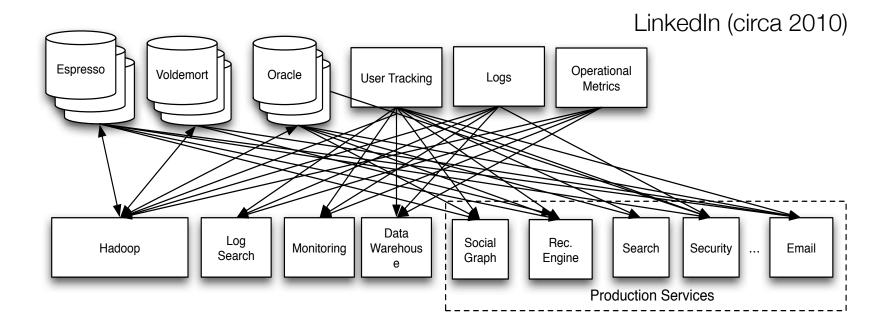
#### Top Complaints from Data Scientists

- Discovery: where is the data?
- Wrangling: can I make sense of the data?
- Verifying: is the data correct?
- Scaling: how can I scale my computation?
- Workflow: how can I operate my processing?
- Publishing: how can I get my results into production?

#### Top Complaints from Data Scientists

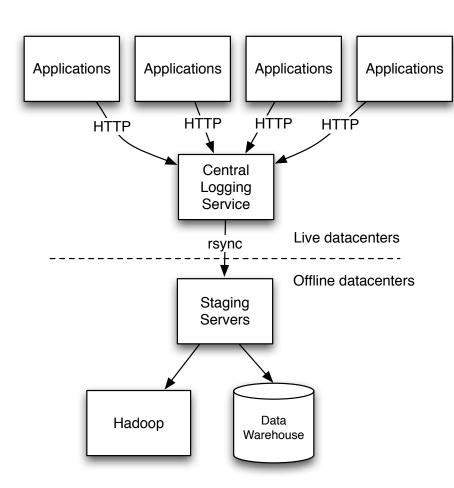
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# Discovery: where is the data?



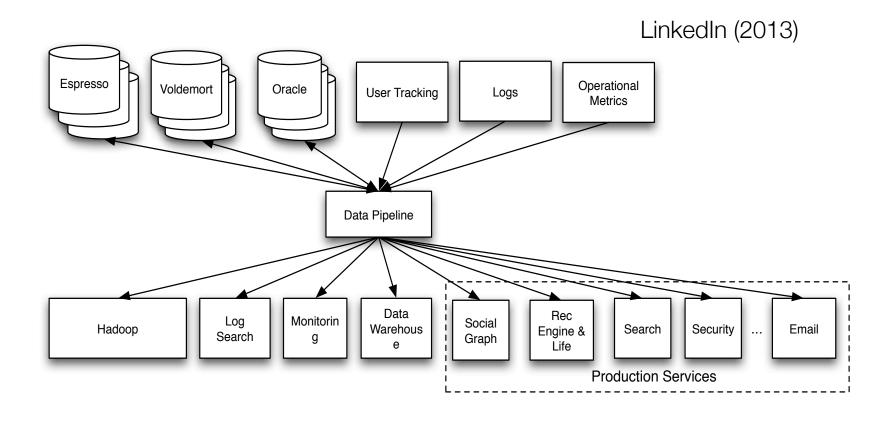
O(n²) point-to-point data integration complexity

# Infrastructure fragility



- Can't get all data
- Hard to operate
- Multi-hour delay
- Labor intensive
- Slow
- Does it work?

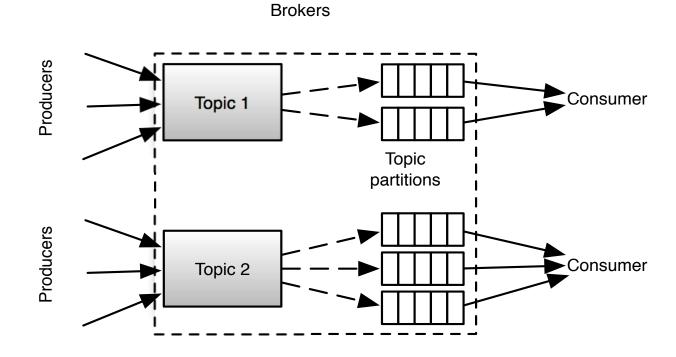
# Ingress - O(n) data integration



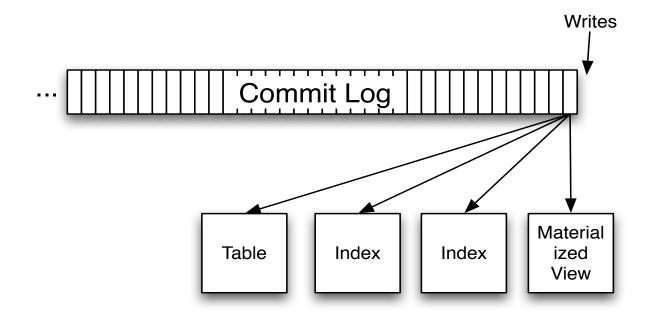
# Ingress - Apache Kafka

- Multi-broker publish/subscribe system
- Categorized topics
  - "PeopleYouMayKnowTopic"
  - "ConnectionUpdateTopic"





# What is a commit log?



#### Top Complaints from Data Scientists

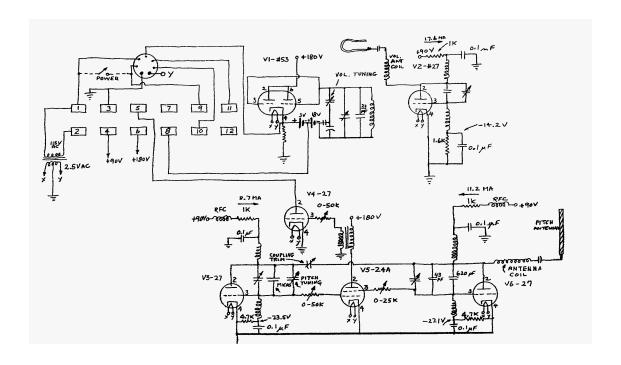
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### Data model

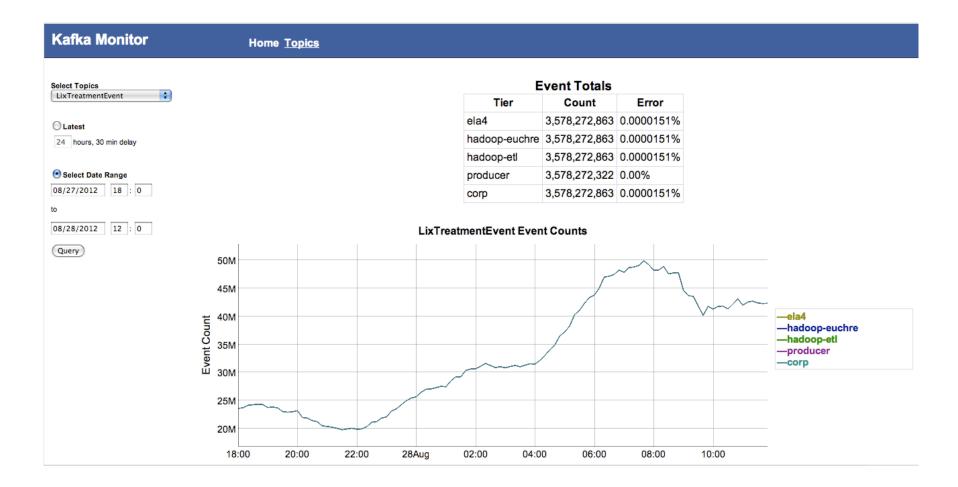
```
LinkedIn (circa 2010)
  tracking code=null,
  session id=42,
  tracking time=Tue Jul 31 07:27:25 PDT 2010,
  error key=null,
  locale=en us,
  browser id=ddc61a81-5311-4859-be42-ca7dc7b941e3,
  member id=1214,
  page key=profile,
tracking info=Viewee=1213,lnl=f,nd=1,o=1214,^SP=pId-'pro_stars',rslvd=t,vs=v
, vid=1214, ps=EDU | EXP | SKIL | ,
  error id=null,
  page type=FULL PAGE,
                                                                   Reid Hoffman
                                                                                                  2nd
                                                                   Entrepreneur. Product Strategist. Investor.
  request path=view
                                                                   San Francisco Bay Area | Internet
   . . .
                                                                         Greylock, LinkedIn, Wrapp
                                                                         College Eight, UCSC, Six Apart, Tagged
                                                                         Oxford University
                                                                   Education
                                                                            Know Reid?
                                                                                                184,444
                                                                    Follow
                                                     Contact Info
                                                                                      in www.linkedin.com/in/reidhoffman
```

# Schemas

- Schemas are the contract
  - DDL for data definition and schema
- Central versioned registry of all schemas
- Schema evolution with programmatic checks



# Audit trail



#### Top Complaints from Data Scientists

- Discovery: where is the data?
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#### Models of computation

Online Nearline Offline

- Sub-second processing
- Harder to scale
- Must handle failures gracefully
- Computationally intensive
- Easier to scale
- Easier to tolerate failures
- Faster iteration

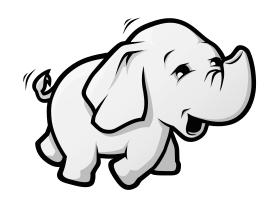
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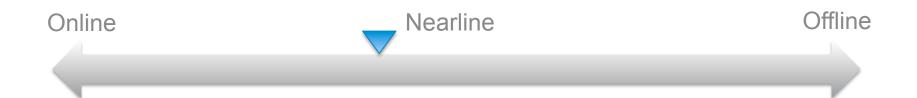
### Hadoop

#### Why we use Hadoop



- Simple programmatic model
- Rich developer ecosystem
  - Languages: Pig, Hive, Crunch, Cascading, ...
  - Libraries: Mahout, DataFu, ElephantBird, ...
- DataFu
  - Large-scale machine learning and statistical operations
- Horizontal scalability, fault tolerance, multi-tenancy
  - Reliably process multiple TB of data

#### Models of computation



- Sub-second processing
- Harder to scale
- Must handle failures gracefully
- Computationally intensive
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- Easier to tolerate failures
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## Apache Samza – "MapReduce for streams"

samza

Samza API

MapReduce

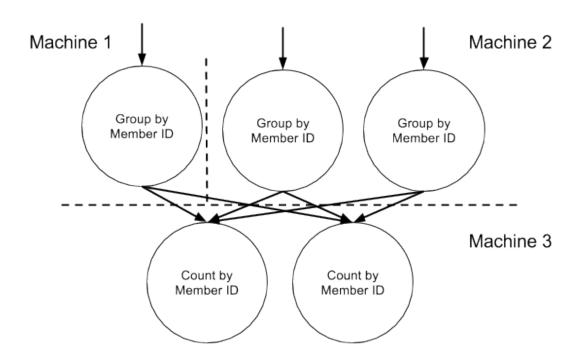
YARN

Kafka

YARN

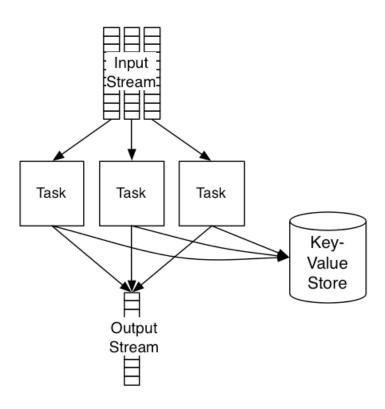
**HDFS** 

#### Samza

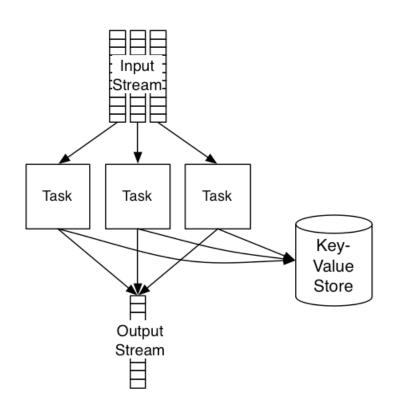


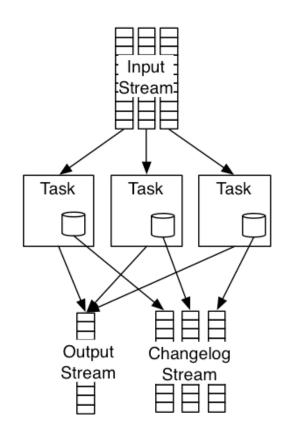
SELECT COUNT(\*) FROM PageViewEvent GROUP BY member\_id

## Samza: State Management



#### Samza: State Management

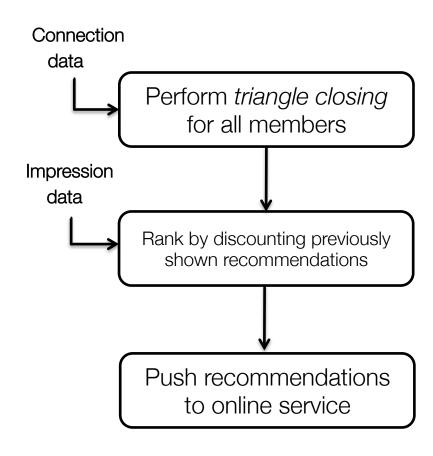




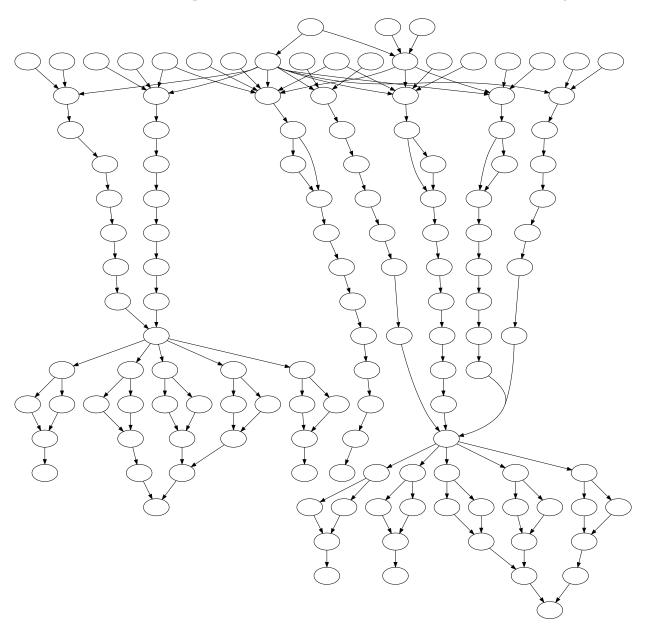
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# People You May Know - Workflow



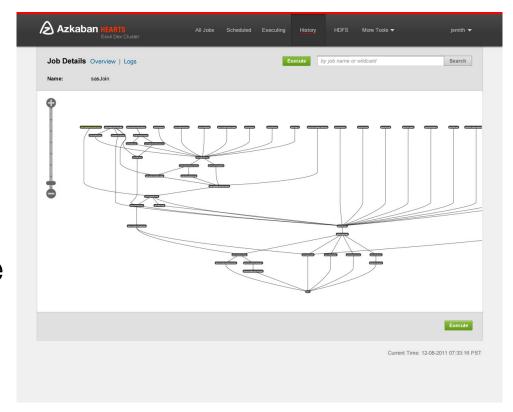
# People You May Know – Workflow (in reality)



# Workflow Management - Azkaban



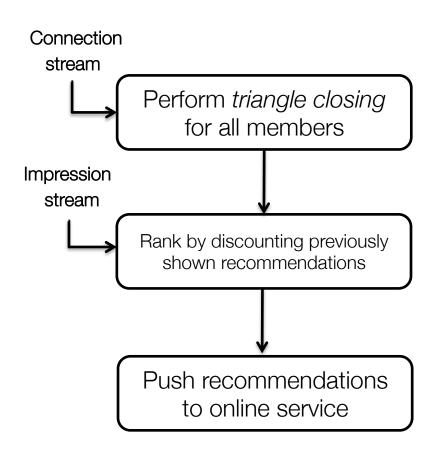
- Dependency management
- Diverse job types
- Scheduling
- Monitoring
- Visualization
- Configuration
- Retry/restart on failure
- Resource locking



#### Top Complaints from Data Scientists

- Discovery: where is the data?
- Wrangling: can I make sense of the data?
- Verifying: is the data correct?
- Scaling: how can I scale my computation?
- Workflow: how can I operate my processing?
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# People You May Know - Workflow



Member Id 1213 =>

[Recommended member id 1734,

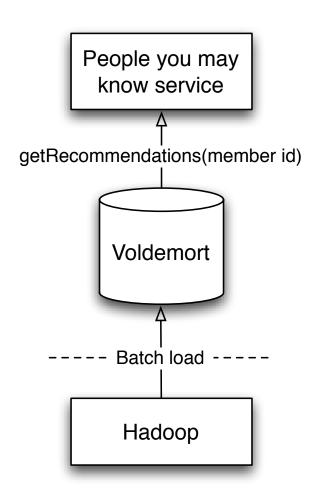
Recommended member id 1523

. . .

Recommended member id 6332]

# Egress - Key/Value

- Voldemort
  - Based on Amazon's Dynamo
- Distributed and elastic
- Horizontally scalable



#### Systems (all open source)

- Apache Kafka: publish/subscribe commit log
- DataFu: Common data routines
- Apache Samza: stream processing framework
- Azkaban: workflow management
- Voldemort: key/value store

Empowers data scientists and engineers to focus on new product ideas, not infrastructure Learning More

data.linkedin.com