# Data Pipeline at Tapad

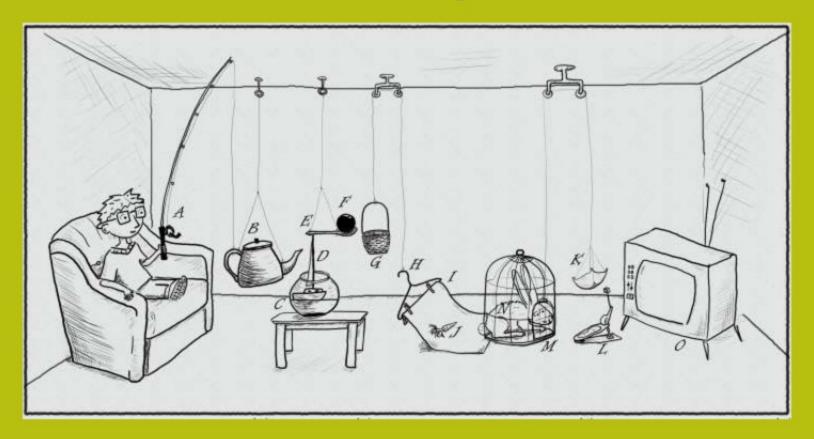
@tobym
@TapadEng

### Who am I?

Toby Matejovsky
First engineer hired at Tapad 3+ years
ago
Scala developer

@tobym

# What are we talking about?



## Outline

- What Tapad does
- Why bother with a data pipeline?
- Evolution of the pipeline
- Day in the life of a analytics pixel
- What's next

### What Tapad Does

Cross-platform advertising and analytics Process billions of events per day

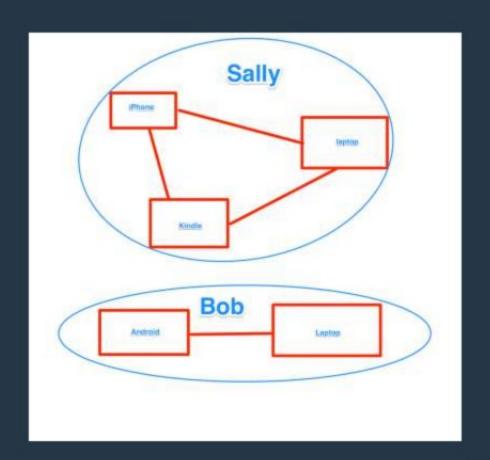
### Cross platform?

Device Graph

Node=device edge=inferred connection

Billion devices Quarter billion edges

85+% accuracy



### Why a Data Pipeline?

Graph building
Sanity while processing big data
Decouple components
Data accessible at multiple stages

## Graph Building

Realtime mode, but don't impact bidding latency Batch mode

### Sanity

Billions of events, terabytes of logs per day Don't have NSA's budget Clear data retention policy Store aggregations

### Decouple Components

Bidder only bids, graph-building process only builds graph

Data stream can split and merge

### Data accessible at multiple stages

Logs on edge of system Local spool of data Kafka broker Consumer local spool HDFS

### Evolution of the Data Pipeline

Dark Ages: Monolithic process, synchronous process

Renaissance: Queues, asynchronous work in same process

Age of Exploration: Inter-process comm, ad hoc batching

Age of Enlightenment: Standardize on Kafka and Avro

### Dark Ages

Monolithic process, synchronous process

It was fast enough, and we had to start somewhere.

#### Renaissance

Queues, asynchronous work in same process

No, it wasn't fast enough.

### Age of Exploration

Inter-process communication, ad hoc batching

Servers at the edge batch up events, ship them to another service.

# Age of Enlightenment

Standardize on Kafka and Avro

Properly engineered and supported, reliable

# Age of Enlightenment

Standardize on Kafka and Avro

Properly engineered and supported, reliable

### Tangent!

Batching, queues, and serialization

### Batching

Batching is great, will really help throughput

Batching != slow

#### Queues

**Queues** are amazing, until they explode and destroy the Rube Goldberg machine.

"I'll just increase the buffer size."

- spoken one day before someone ended up on double PagerDuty rotation

## Care and feeding of your queue

Monitor

Back-pressure

Buffering

Spooling

Degraded mode

### Serialization - Protocol Buffers

```
Tagged fields
Sort of self-describing
required, optional, repeated fields in schema
"Map" type:
message StringPair {
   required string key = 1;
   optional string value = 2;
```

#### Serialization - Avro

Optional field: union { null, long } user\_timestamp = null;

Splittable (Hadoop world)

Schema evolution and storage

### Day in the life of a pixel

Browser loads pixel from pixel server

Pixel server immediately responds with 200 and transparent gif,

then serializes requests into a batch file

Batch file ships every few seconds or when the file reaches 2K

### Day in the life of a pixel

Pixel ingress server receives 2 kilobyte file containing serialized web requests.

Deserialize, process some requests immediately (update database), then convert into Avro records with schema hash header, and publish to various Kafka topics

### Day in the life of a pixel

Producer client figures out where to publish via the broker they connect to

Kafka topics are partitioned into multiple chunks, each has a master and slave and are on different servers to survive an outage.

Configurable retention based on time

Can add topics dynamically

### Day in the life of a pixel

Consumer processes are organized into groups
Many consumer groups can read from same Kafka topic
Plugins:
trait Plugin[A] {
 def onStartup(): Unit
 def onSuccess(a: A): Unit
 def onFailure(a: A): Unit
 def onShutdown(): Unit

GraphitePlugin, BatchingLogfilePlaybackPlugin, TimestampDrivenClockPlugin, BatchingTimestampDrivenClockPlugin, ...

### Day in the life of a pixel

```
trait Plugins[A] {
  private val _plugins = ArrayBuffer.empty[Plugin[A]]
  def plugins: Seq[Plugin[A]] = _plugins
  def registerPlugin(plugin: Plugin[A]) = _plugins += plugin
}
```

### Day in the life of a pixel

```
object KafkaConsumer {
    sealed trait Result {
        def notify[A](plugins: Seq[Plugin[A]], a: A): Unit
    }

    case object Success extends Result {
        def notify[A](plugins: Seq[Plugin[A]], a: A) {
            plugins.foreach(_.onSuccess(a))
        }
    }
}
```

```
/** Decorate a Function1[A, B] with retry logic */
case class Retry[A, B](maxAttempts: Int, backoff: Long)(f: A => B){
 def apply(a: A): Result[A, B] = {
  def execute(attempt: Int, errorLog: List[Throwable]): Result[A, B] = {
   val result = try {
     Success(this, a, f(a))
   } catch {
     ... Failure(this, a, e :: errorLog) ...
    result match {
     case failure @ Failure(_, _, errorLog) if errorLog.size < maxAttempts =>
      val _backoff = (math.pow(2, attempt) * backoff).toLong
      Thread. sleep(backoff) // wait before the next invocation
      execute(attempt + 1, errorLog) // try again
     case failure @ Failure(_, _, errorLog) =>
      failure
  execute(attempt = 0, errorLog = NiI)
```

### Day in the life of a pixel

Consumers log into "permanent storage" in HDFS.

File format is Avro, written in batches.

Data retention policy is essential.

### Day in the life of a pixel

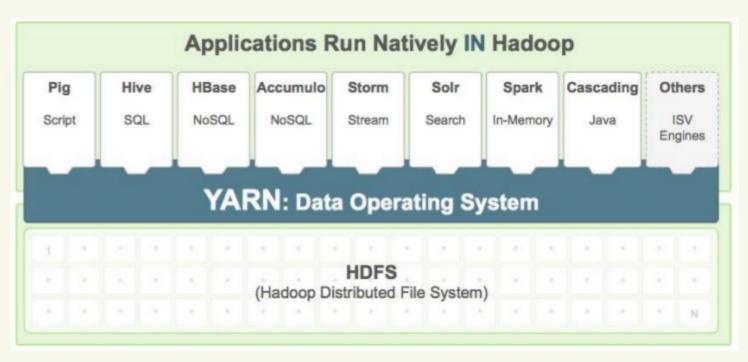
Hadoop 2 - YARN

Scalding to write map-reduce jobs easily

Rewrite Avro files as Parquet

Oozie to schedule regular jobs

#### YARN



### Scalding

```
class WordCountJob(args : Args) extends Job(args) {
  TextLine( args("input") )
    .flatMap('line -> 'word) { line : String => tokenize(line) }
    .groupBy('word) { _.size }
    .write( Tsv( args("output") ) )

// Split a piece of text into individual words.
    def tokenize(text : String) : Array[String] = {
        // Lowercase each word and remove punctuation.
        text.toLowerCase.replaceAll("[^a-zA-Z0-9\\s]", "").split("\\s+")
    }
}
```

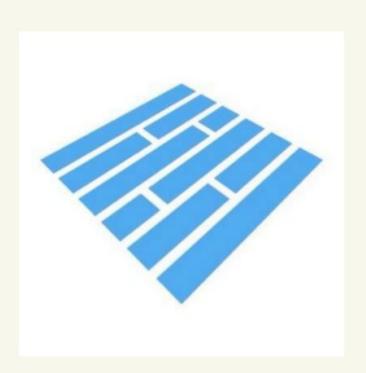
### Parquet

Column-oriented storage for Hadoop

Nested data is okay

Projections

**Predicates** 



### Parquet

### Oozie

```
<workflow-app name="combined queries" xmlns="uri:oozie:workflow:0.3">
    <start to="devices-location"/>
    <!--<start to="export2db"/>-->
    <action name="devices-location">
        <shell xmlns="uri:oozie:shell-action:0.1">
            <job-tracker>${jobTracker}</job-tracker>
            <name-node>$ (nameNode) </name-node>
            <exec>hadoop</exec>
            <argument>fs</argument>
            <argument>-cat</argument>
            <argument>${devicesConfig}</argument>
      <capture-output/>
    </shell>
    <ok to="networks-location"/>
    <error to="kill"/>
  </action>
```

### Day in the life of a pixel

Near real-time consumers and batch hadoop jobs generate data cubes from incoming events and save those aggregations into Vertica for fast and easy querying with SQL.

# Stack summary

Scala, Jetty/Netty, Finagle Avro, Protocol Buffers, Parquet Kafka Zookeeper Hadoop - YARN and HDFS Vertica Scalding Oozie, Sqoop

### What's next?

Hive Druid Impala Oozie alternative



@tobym

@TapadEng

yes, we're hiring!:)

Toby Matejovsky, Director of Engineering toby@tapad.com
@tobym

