



# Spark and Spark Streaming @ Netflix

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# Mission

- Enable **rapid** pace of **innovation** for Algorithm Engineers
- **Business Value** – More A/B tests

# Experiments



Users with plays



Feature selection



Large sample size

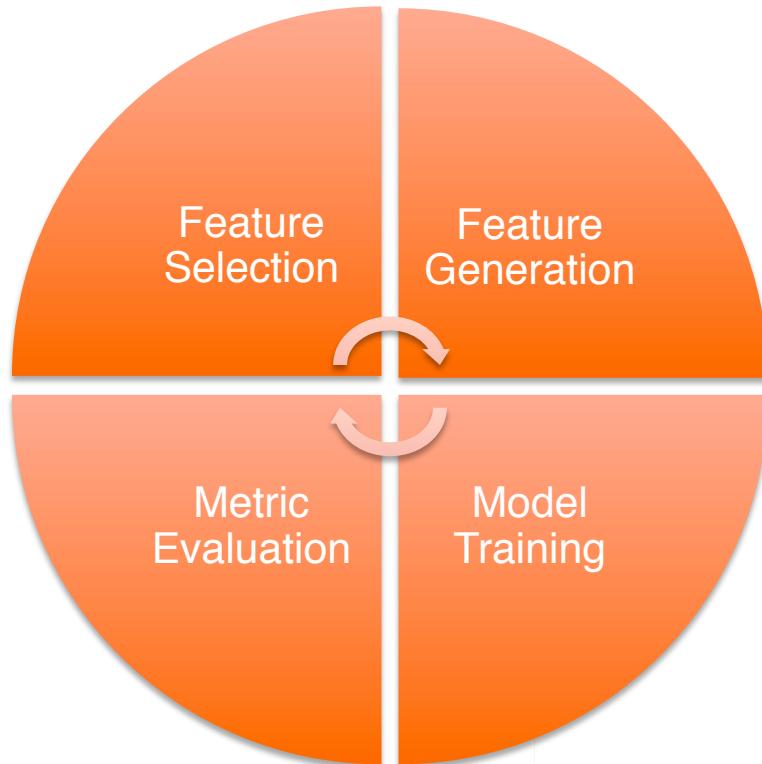


Turn back time

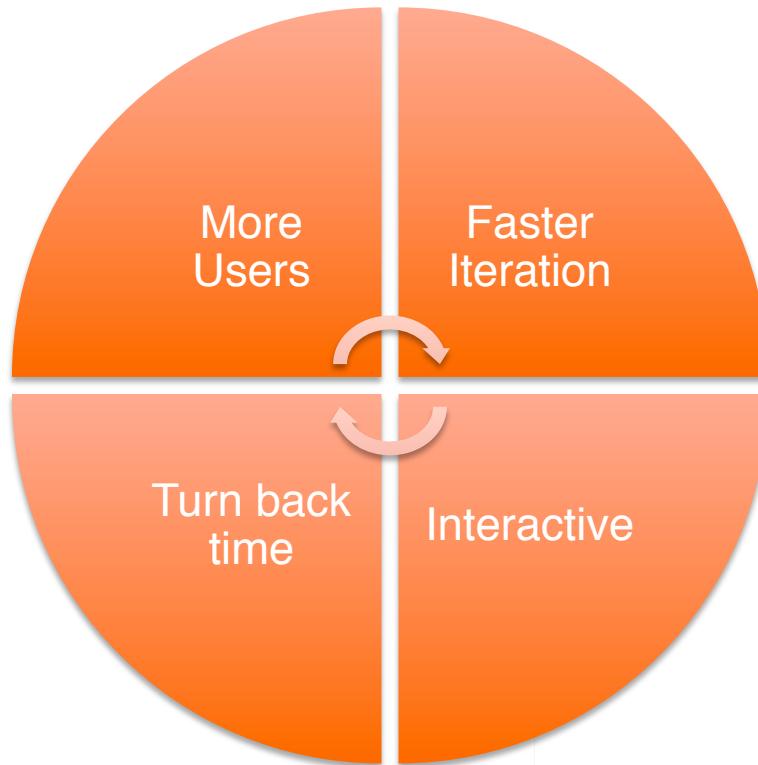


Multiple ideas

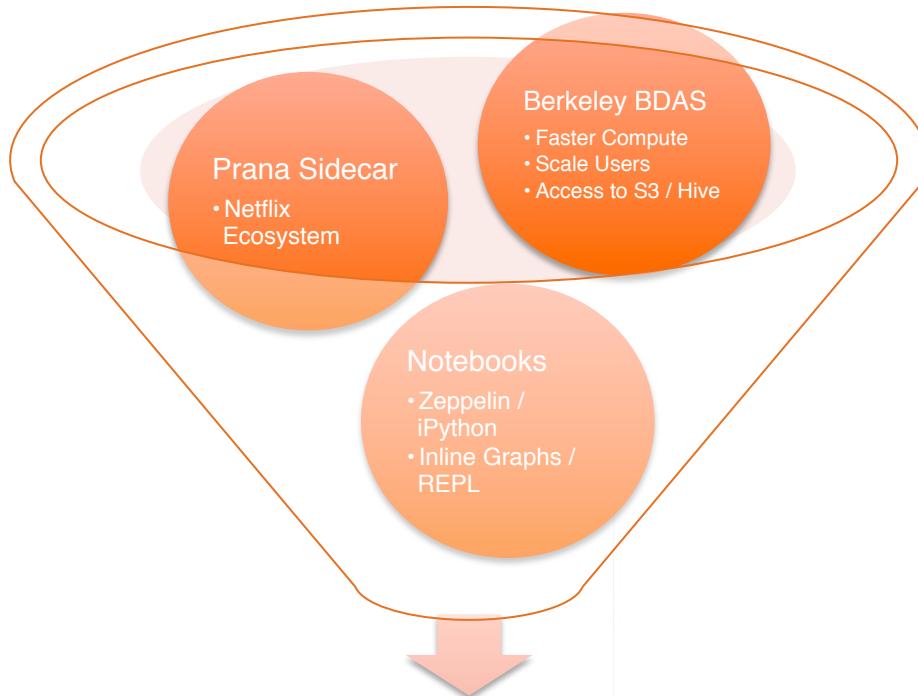
# Use Cases



# Use Cases



# Solution – Netflix BDAS



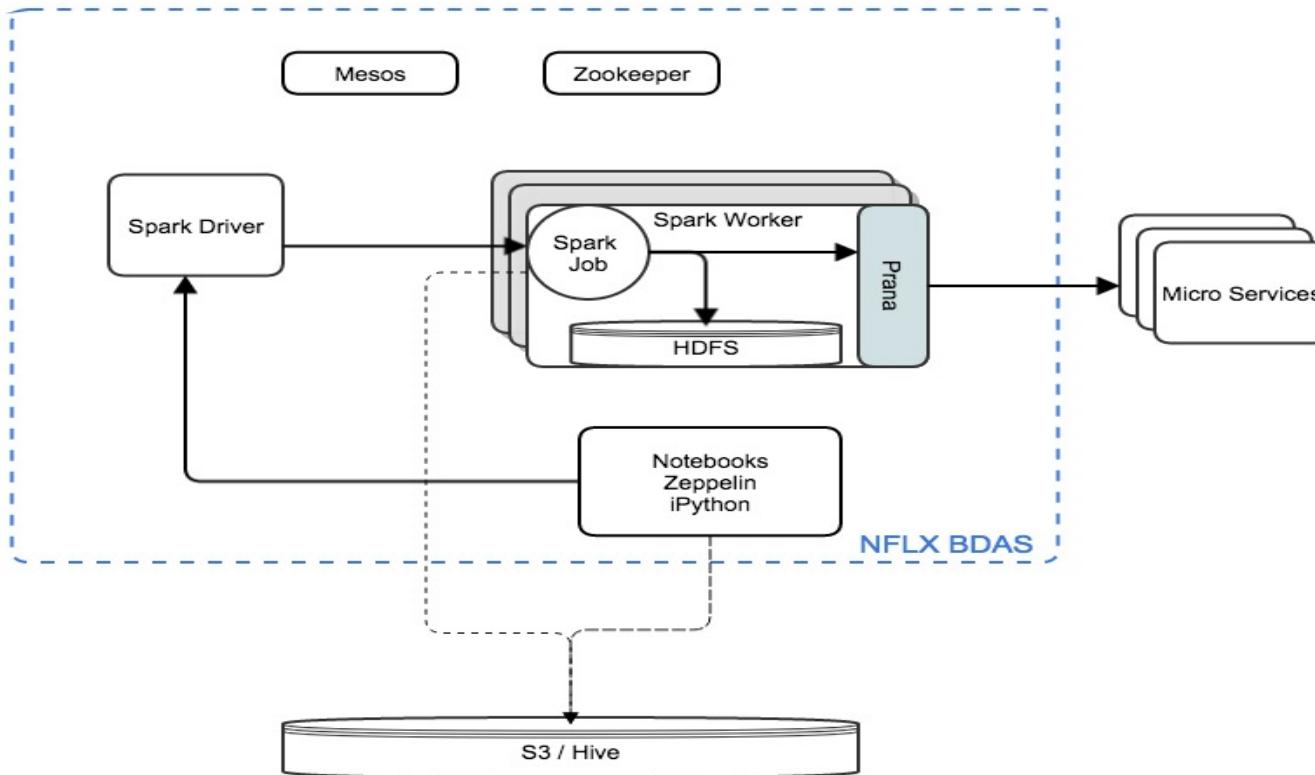
# Netflix BDAS - Features

- Simplicity
  - Individual cluster
- Prana - Netflix ecosystem
  - Automatic configuration
  - Classloader isolation
  - Discovery & healthcheck

# Netflix BDAS - Features

- Ad-hoc experimentation
- Time machine functionality
- Access to Hive data and micro services from single place
  - Access to multiple AWS buckets (S3)

# Netflix BDAS – Sample Deployment



# Wins

- 8X the number of users
- 5x - 9x faster
- Interactive



Turn back time

# Learnings

- Easy to bring down an online system
- Almost killed 1000's of ETL jobs
  - hive metastore update



- Too many systems and configuration
- Playing catch up with libraries and tools
  - Hadoop, iPython, Zeppelin

- Scala / Spark learning curve
- Debugging
  - files open, no resources etc.



# Increased Adoption

Adoption increasing amongst teams

- Multiple Algorithmic Eng. teams
- Personalization Infrastructure
- Marketing
- Security
- A/B Test Engineering

# Looking Ahead

- Spark-R / Dataframes support
- Multi-tenancy
  - Job specific configurations
- Debuggability
- Newer notebooks
- Spark Streaming
  - Lambda Architecture
  - Real-time algorithms (trending now)



A photograph of the Golden Gate Bridge in San Francisco, California, taken at sunset. The bridge's towers and cables are bathed in a warm, orange glow, contrasting with the cool blue of the San Francisco Bay and the city skyline in the background.

# Netflix Streaming Event Data Pipeline

Monal Daxini

# Netflix Event Data Pipeline

Event Streams

Stream processing

Publish

Collect

Process

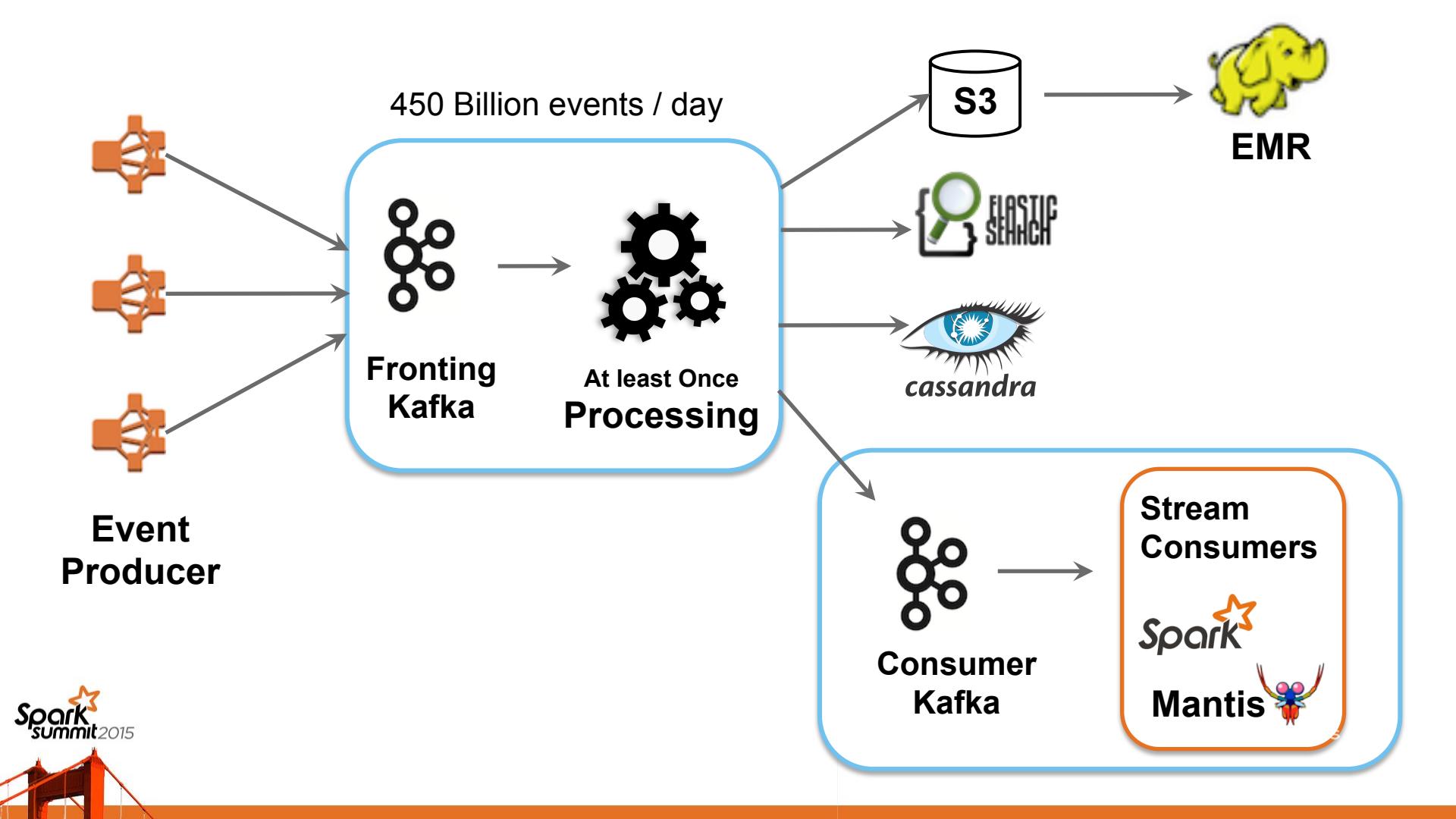
Move

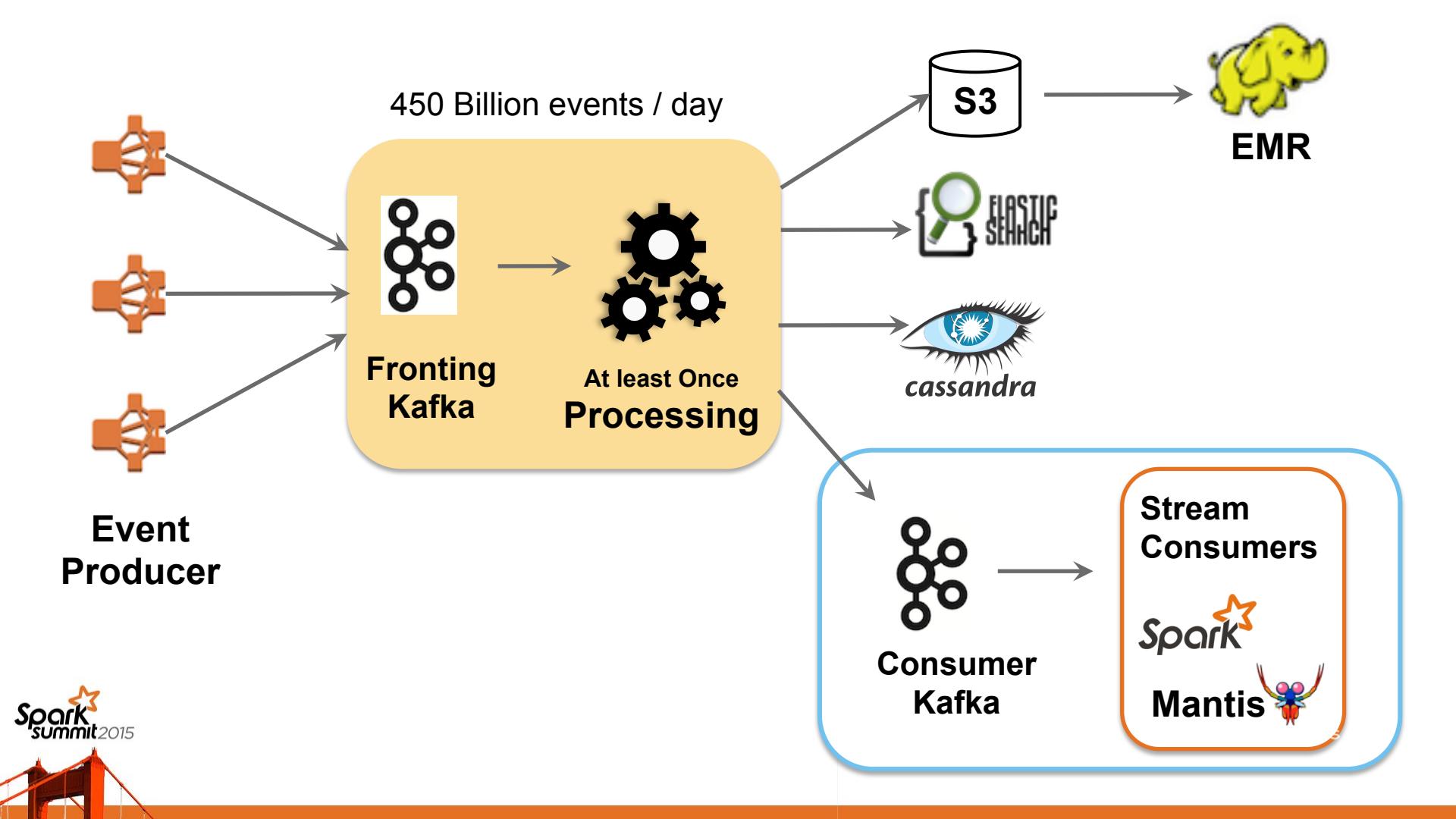
Events @ Cloud Scale

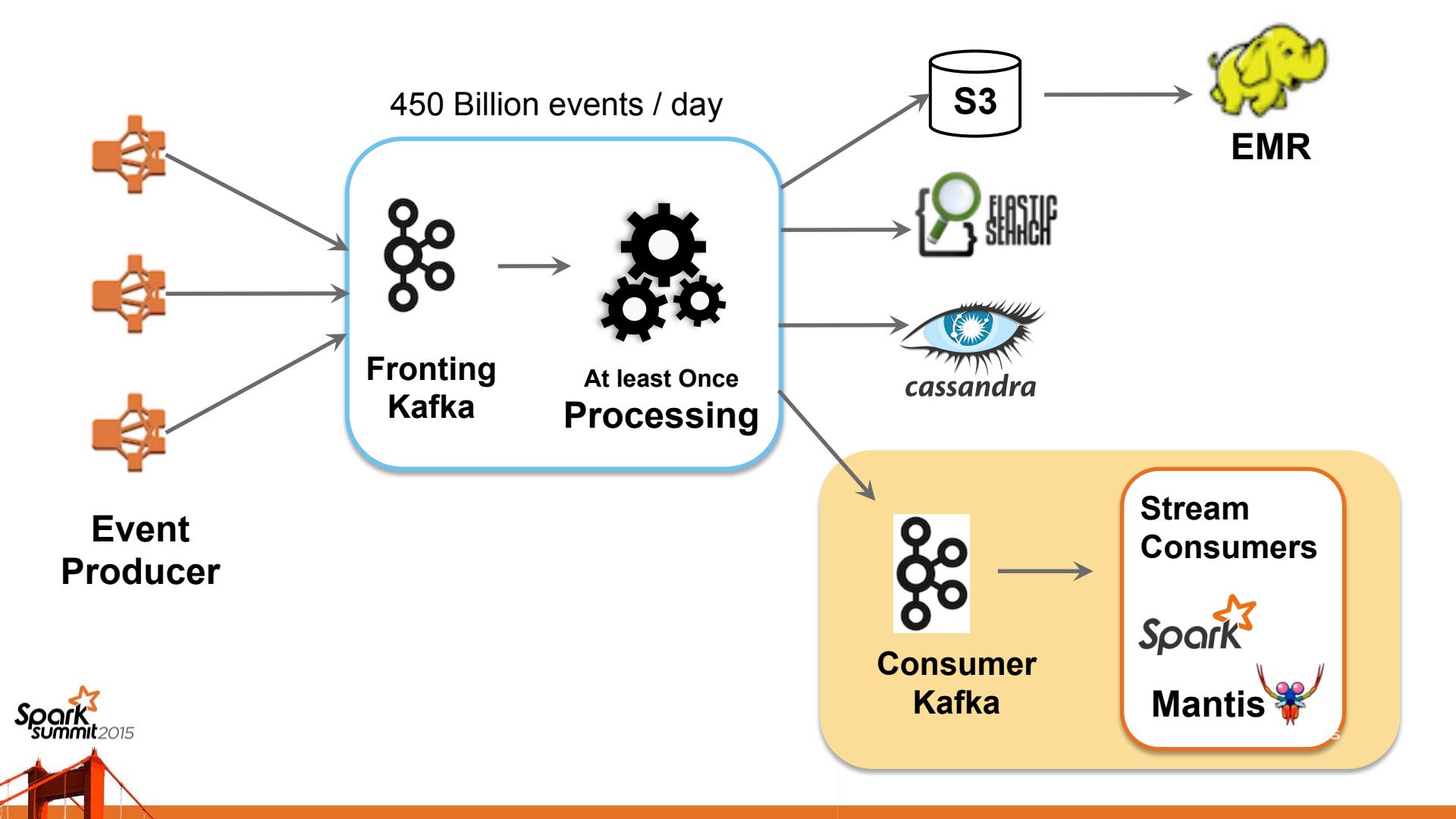


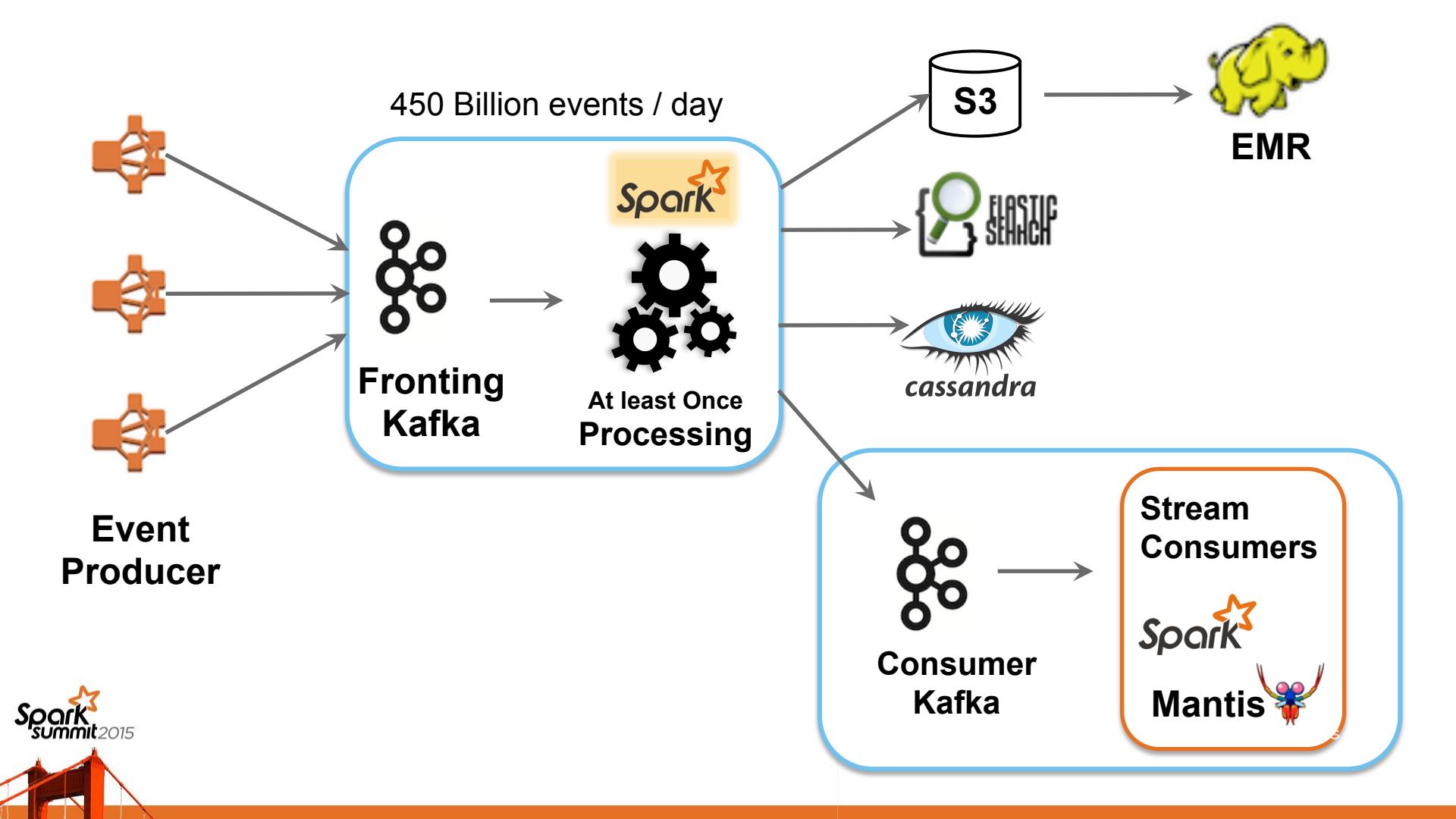
450 Billion Events per Day

8 Million (17 GB) per sec peak

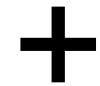








## What's Missing?



Backpressure



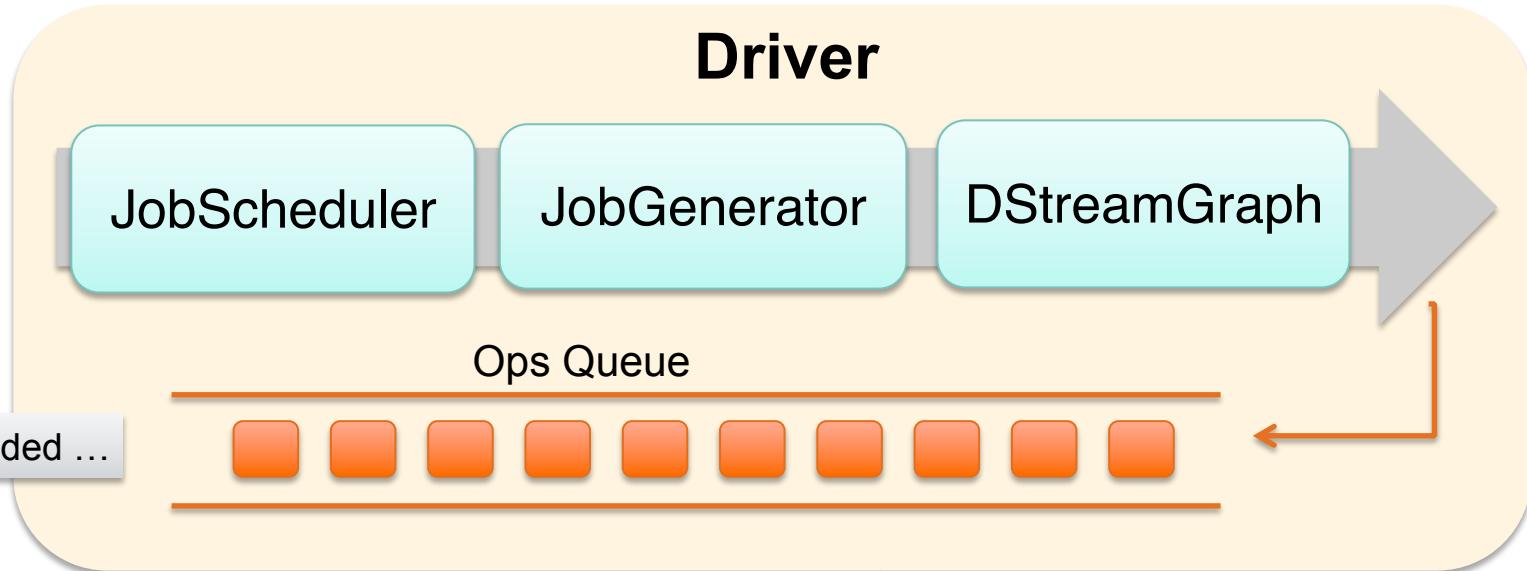
Direct API for Kafka



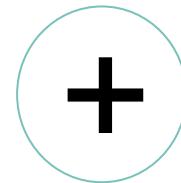
Improve Cloud Multi-tenancy



# Backpressure



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# Backpressure

[SPARK-7398](#) – Add backpressure to Spark streaming

[SPARK-6691](#) – Add dynamic rate limiter to Spark streaming



# Backpressure

Backpressure implementation slated for  
Spark 1.5 release



# Direct API for Kafka

Spark 1.3  
Kafka Integration

2x  
Faster

Spark 1.2  
Receiver based  
Kafka Integration



# Direct API for Kafka

Enhance



Prefetch messages

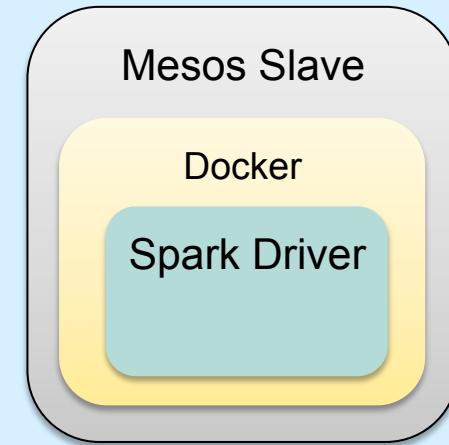
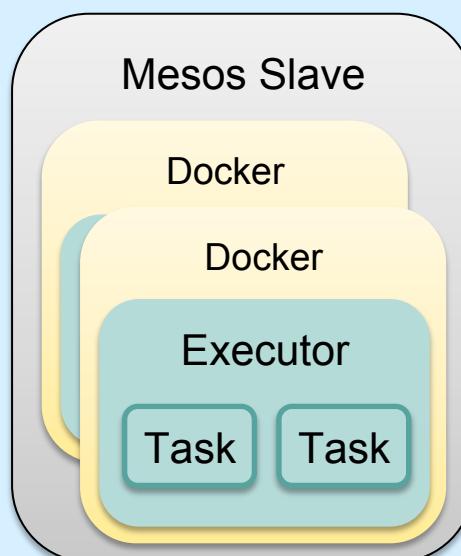
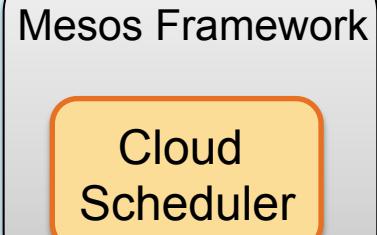


Connection reuse (pooling)

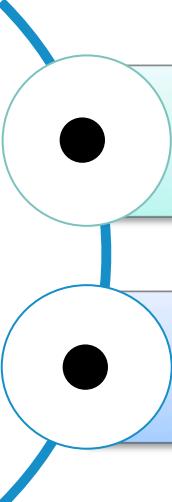


# Cloud Multi-tenancy

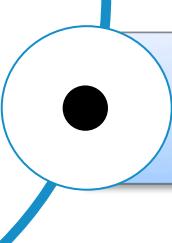
Improve



# Next?



Measuring Spark Streaming Latencies



Spark Streaming Cloud Multi-tenancy

