

Real-time Machine Learning Analytics Using Structured Streaming and Kinesis Firehose

Caryl Yuhas (@ckred)

Myles Baker (@mydpy)

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About Databricks

TEAM

Started Spark project (now Apache Spark) at UC Berkeley in 2009

MISSION

Making Big Data Simple

PRODUCT

Unified Analytics Platform

Impact of Real-Time Analytics

- Capturing customer interactions, user behavior, and sensor readings is **rapidly increasing**
- Businesses need to respond immediately to new information **as it arrives**
- Real-time analytics is at the **core** of next-generation IT systems

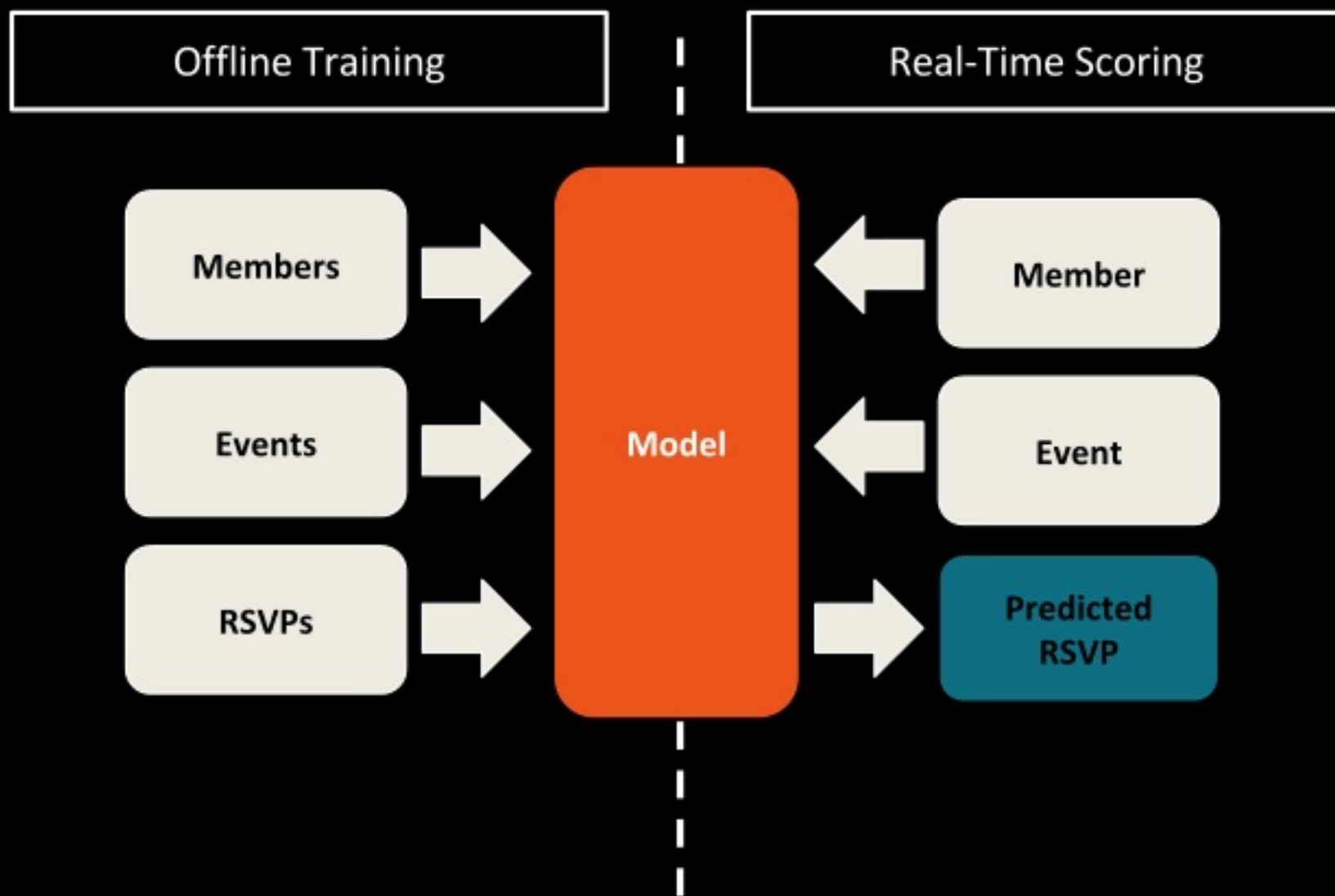
Challenges Building a Solution

- Performant, scalable real-time analytics requires **connecting multiple tools**
- Streaming data comes with all of the **problems** of static data with **added complexity**
- Machine learning models need to be trained on historical data and **scored with real-time data**

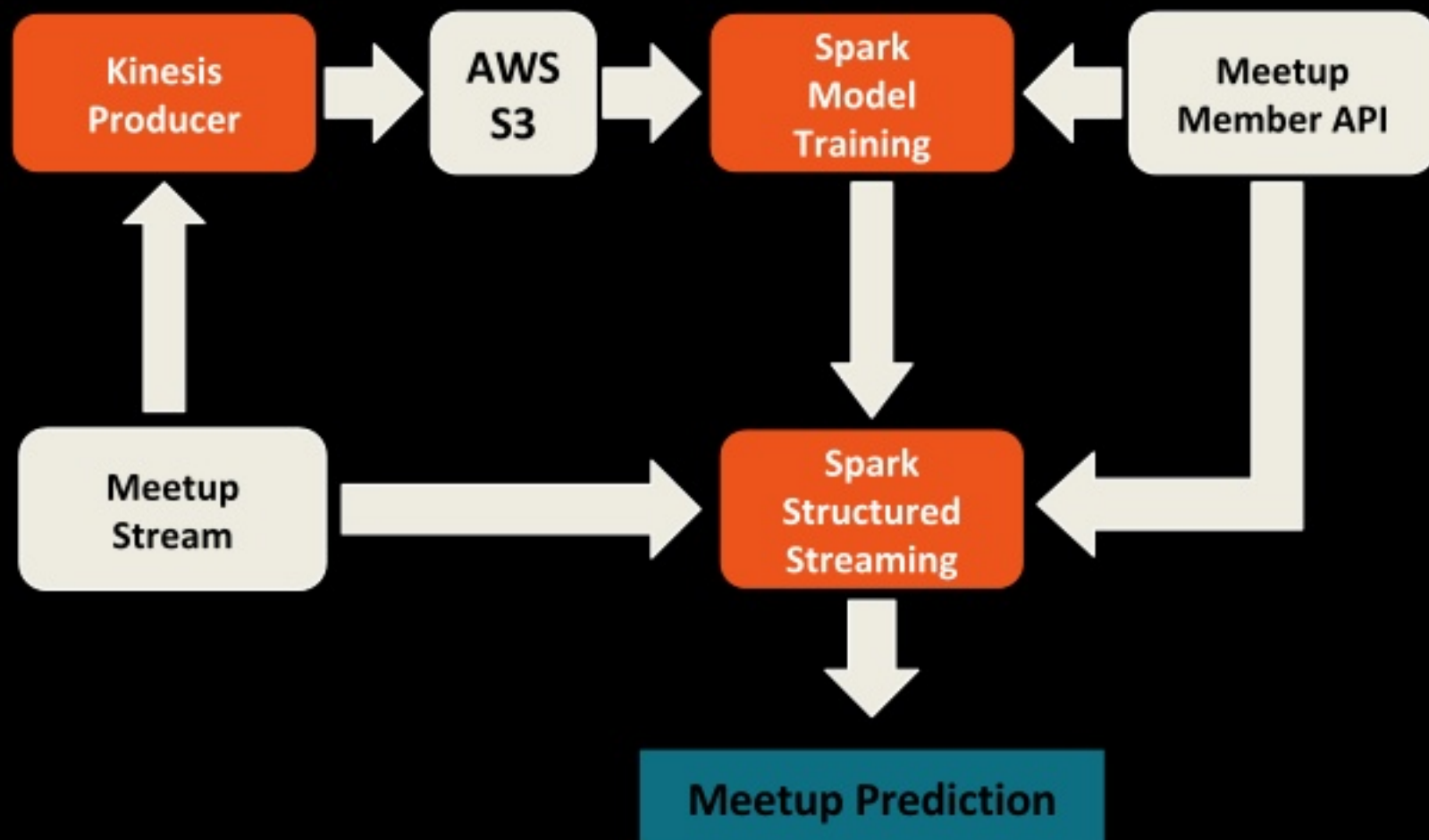
The Meetup Streaming API

- Can we explore Meetup data in real-time?
- Can we predict RSVPs for new Meetups using streaming data from the [Meetup API](#)?
 - Members
 - Events
 - RSVPs

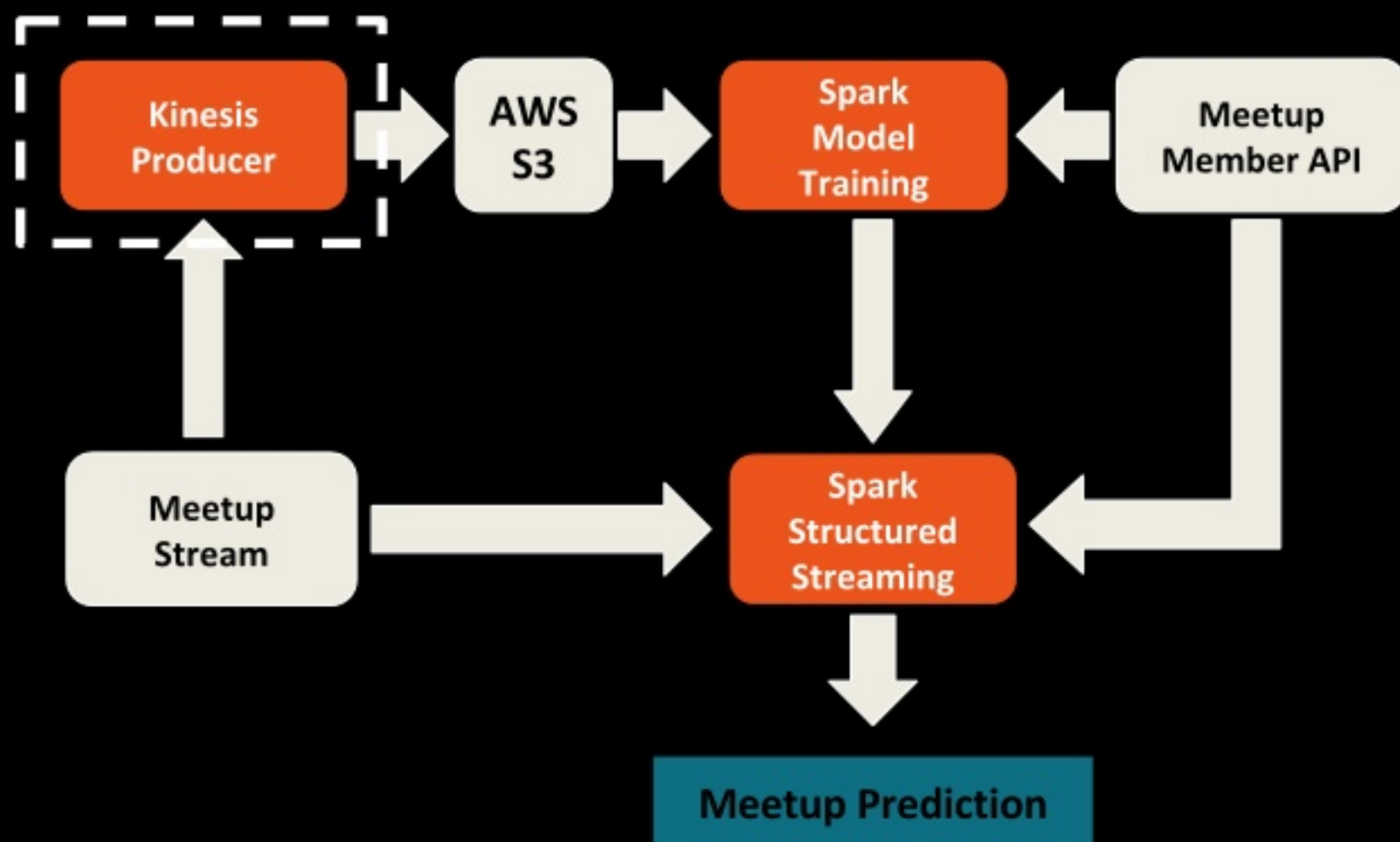
A Data Model for Training and Scoring



Component Integration and Serving



Component Integration and Serving



Producing the Kinesis Firehose Stream

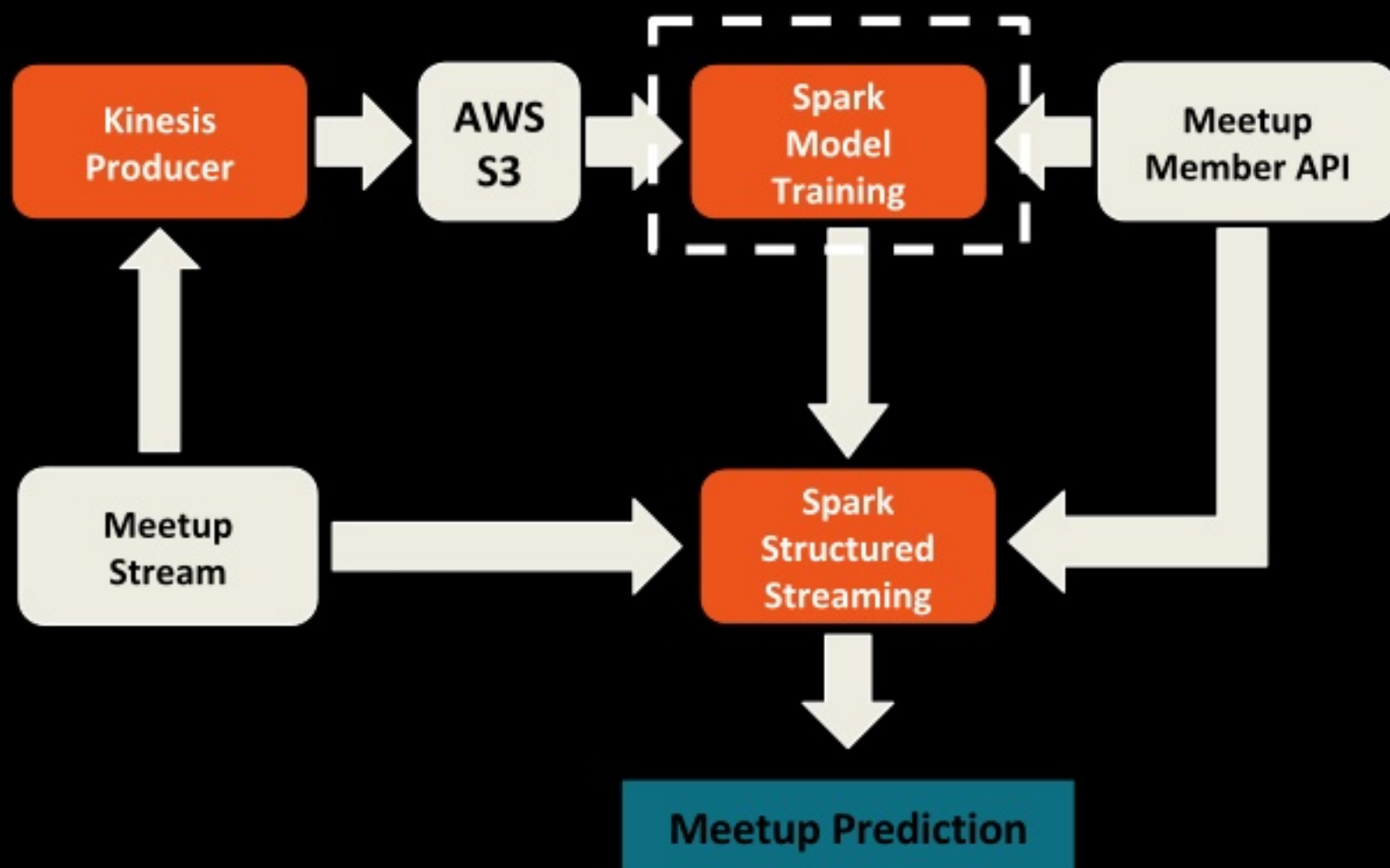
```
requests.get(apiURL, stream = True)  
kinesis = boto3.client('firehose')
```

```
kinesis.put_record_batch(  
    DeliveryStreamName='meetup',  
    Records=rsvps)
```

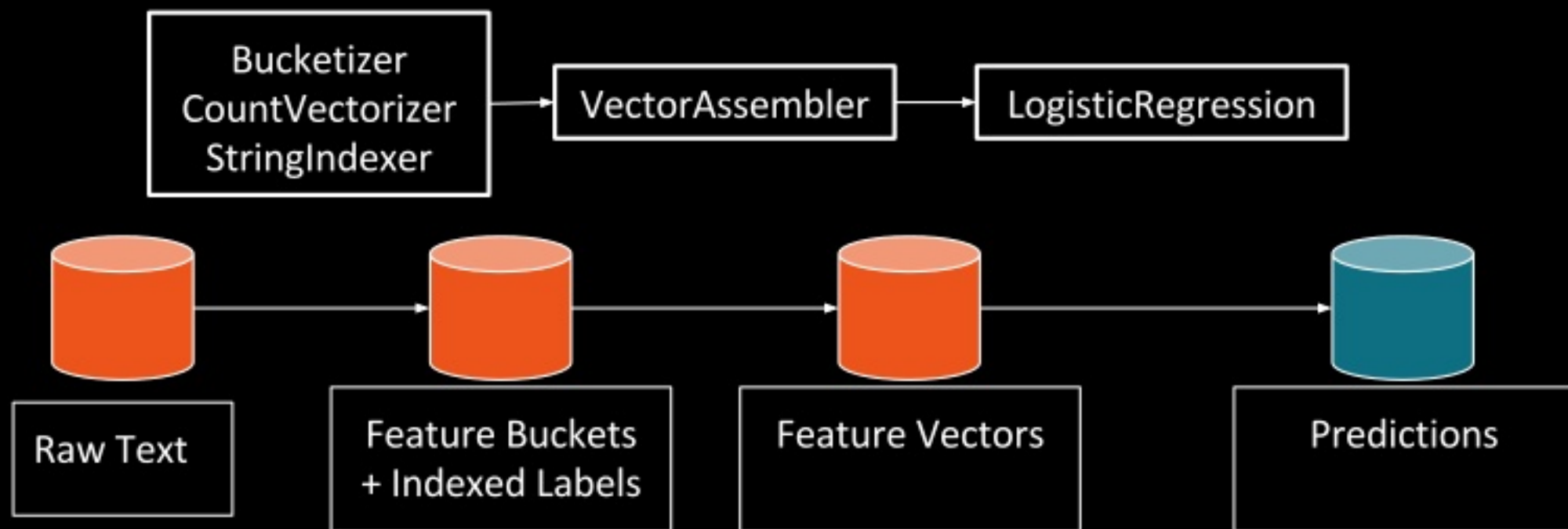
requests.get() makes a request to the **Meetup API**, keeping the **stream** open
boto3.client creates a **firehose kinesis** client

kinesis.put_record_batch() writes the records streamed to S3 using the Kinesis **Firehose** delivery stream **'meetup'**

Component Integration and Serving



Our Meetup ML Pipeline



Create an ML Pipeline

```
val pipeline = new Pipeline()  
  .setStages(Array(  
    transformers,  
    estimators,  
    models))
```

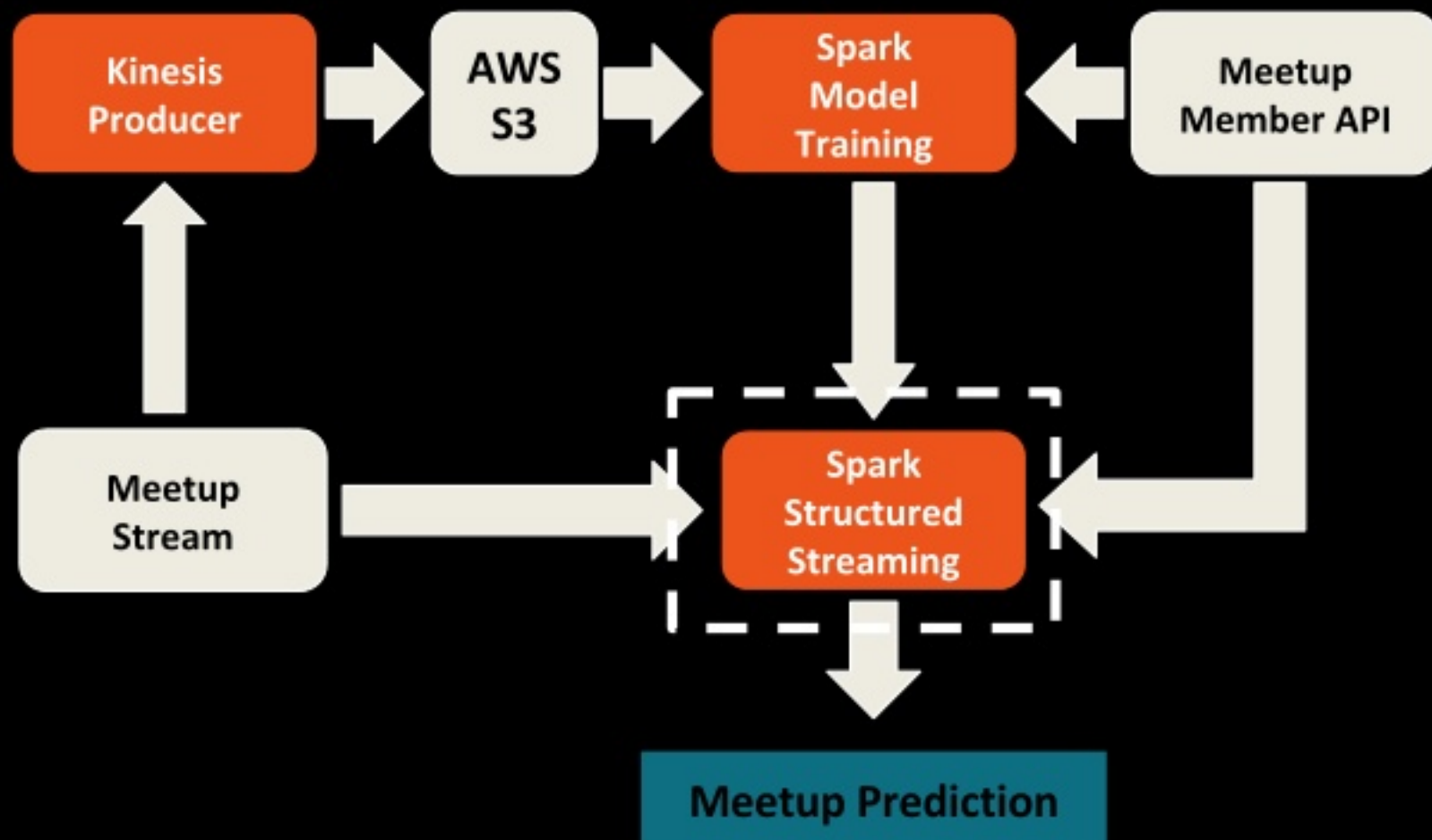
A **Pipeline** allows us to simply chain a series of transformations and estimators

```
val model = pipeline.fit(meetup)  
  
model.write.overwrite().save(...)
```

Fit a **model** based on the **pipeline**

Save the **model** to disk for scoring

Component Integration and Serving



Scoring the Model in Real-time

```
val model = PipelineModel.load(...)
```

Load the **trained** model

```
val events = spark.readStream  
    .parquet(...)
```

Stream meetup **event** data

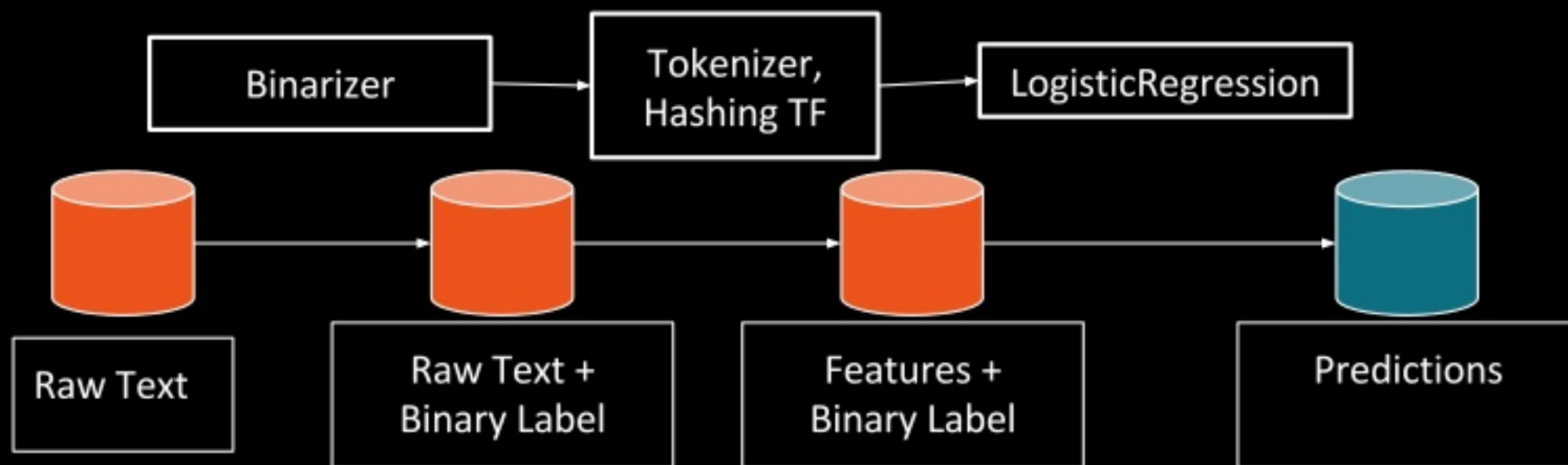
```
model.transform(members)
```

Score the model

ML Limitations in Structured Streaming

- Structured streaming does not support operations needed by ML methods
 - count, collect, round, aggregate*, etc.
- Many models, transformers, and estimators are not supported
 - K-Means, SVM, CountVectorizer, VectorAssembler, StringIndexer, etc.

Our Streaming Meetup ML Pipeline



Alternative Scoring: Model Export

1) Fit ML model in Databricks using Spark MLlib.

2) Export model (as JSON files) in Databricks

```
val lrModel = new LogisticRegression().fit(myData)
ModelExporter.export(lrModel, "s3a:/...")
```

3) Deploy model in external system

```
import com.databricks.ml.local.ModelImport
val lrModel = ModelImport.import("s3a:/...")
val jsonInput = json(...)
val jsonOutput = lrModel.transform(jsonInput)
```

Try Apache Spark in Databricks!

UNIFIED ANALYTICS PLATFORM

- Collaborative cloud environment
- Free version (community edition)

DATABRICKS RUNTIME 3.0

- Apache Spark - optimized for the cloud
- Caching and optimization layer - DBIO
- Enterprise security - DBES

Try for free today.
databricks.com

#SFexp6



mydpy/ss-2017-structured-streaming

Thank you

caryl@databricks.com
mbaker@databricks.com



Common Questions

- Can I consume data from Kinesis **directly** with Structured Streaming?
- Does **MLlib** support streaming data frames?
- Why did you use **Boto3** to produce the Kinesis stream?
- How well does this **scale**? Can we test **volume**?