# FUSING APACHE SPARK AND LUCENE FOR NEAR-REALTIME PREDICTIVE MODEL BUILDING

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#### Contributors

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#### **Data Overview**

#### Location data

- Each srclp defined as unique row key
- Provides approximate location of each key
- Timeseries containing latitude, longitude, error bound, duration, timezone for each key

#### Clickstream data

- Contains clickstream data of each row key
- Contains startTime, duration, httphost, httpuri, upload/download bytes, httpmethod
- Compatible with IPFIX/Netflow formats



## Marketing Analytics

Anonymous aggregate analysis for customer insights





Lookalike modeling





Churn reduction



Competitive analysis



Increased share of stomach

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#### **Data Model**

Dense dimension, dense measure

Schema: srcip, date, hour, tld, zip, tldvisits, zipvisits Data: 10.1.13.120, d1, H2, macvs.com, 94555, 2, 4

Sparse dimension, dense measure

Schema: srcip, date, tld, zip, clickstreamvisits, zipvisits
Data: 10.1.13.120, d1, {macys.com, kohls.com}, {94555, 94301}, 10, 15

Sparse dimension, sparse measure

Schema: srcip, date, tld, zip, tldvisits, zipvisits
Data: 10.1.13.120, d1, {macys.com, kohls.com}, {94555, 94301}, {macys.com;4, kohls.com;6}, {94555;8, 94301;7}
Schema: srcip, week, tld, zip, tldvisits, zipvisits
Data: 10.1.13.120, week 1, {macys.com, kohls.com}, {94555, 94301}, {macys.com;4, kohls.com;6}, {94555;8, 94301;7}

Sparse dimension, sparse measure, last N days

Schema: srcip, tld, zip, tldvisits, zipvisits
Data: 10.1.13.120, {macys.com, kohls.com}, {94555, 94301}, {macys.com;4, kohls.com;6}, {94555;8, 94301;7}

Competing technologies: PowerDrill, Druid, LinkedIn Pinot, EssBase



#### **Document Dataset Representation**

Example

Schema: srcip, tld, zip, tldvisits, zipvisits
Data: 10.1.13.120, {macvs.com, kohls.com}, {94555, 94301}, {macvs.com: 4, kohls.com: 6}, {94555.8, 94301:7}

DataFrame row to Lucene Document mapping

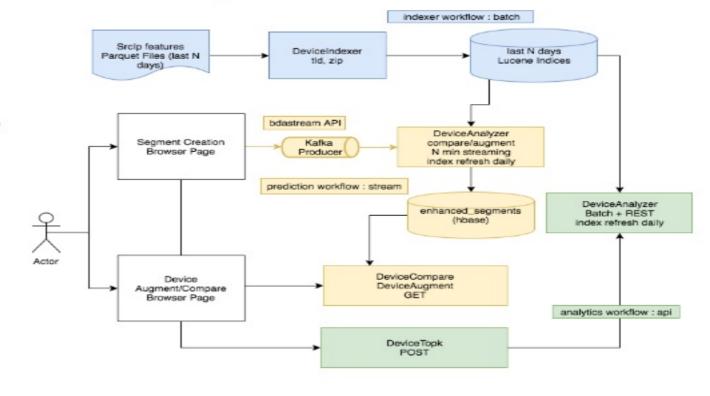
Store/schema	Row	Document
srcip	primary key	docld
tld	String	SingleValue/MultiValue
zip	Array[String]	Indexed Fields
tldvisits	Double	SparseVector
zipvisits	Map[String, Double]	StoredField

- Distributed collection of srclp as RDD[Document]
  - ~100M srcip, 1M+ terms (sparse dimensions)



## DeviceAnalyzer

- DeviceAnalyzer goals
  - Search and retrieve devices that matched query
  - Generate statistical and predictive models on retrieved devices

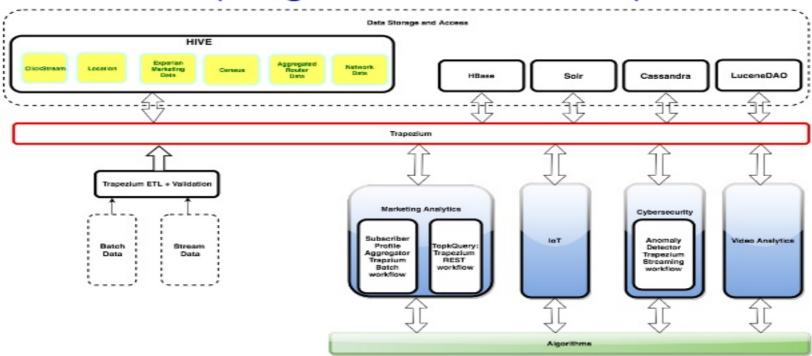




#### What is Trapezium?

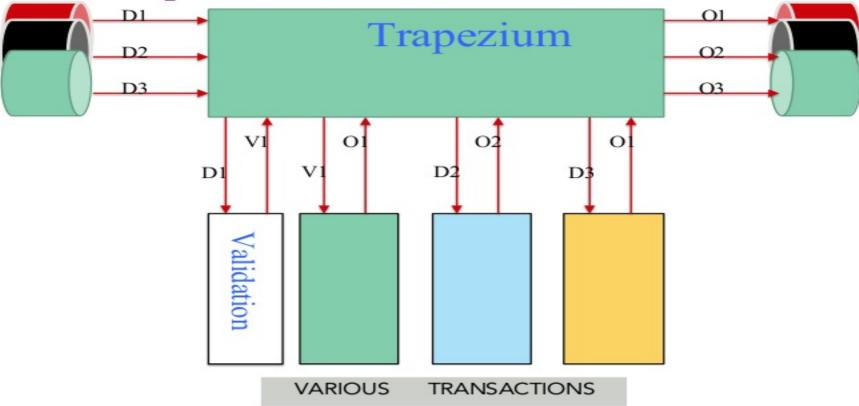
DAIS Open Source framework to build batch, streaming and API services

https://github.com/Verizon/trapezium





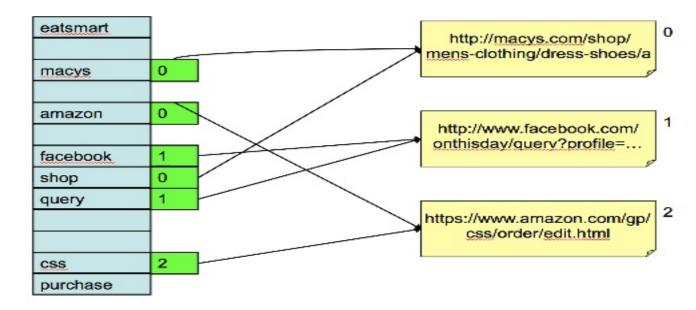
## Trapezium Architecture





#### Lucene Overview

- Scalable, full-text search library
- Focus: Indexing + searching documents





## Trapezium LuceneDAO

- SparkSQL and MLlib optimized for full scan, column indexing not supported
- Why Spark + Lucene integration
  - Lucene is battle tested Apache Licensed Open Source Project
  - Adds column search capabilities to Spark
  - Adds spark operators (treeAggregate, treeReduce, map) to Lucene
- LuceneDAO features
  - Build distributed lucene shards from Dataframe
  - Save shards to HDFS for QueryProcessor (CloudSolr)
  - Access saved shards through LuceneDAO for ML pipelines



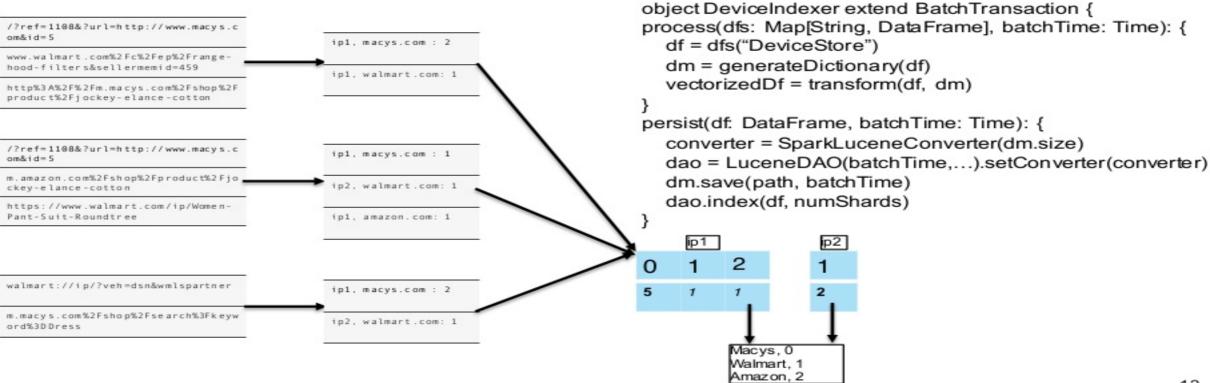
#### **Trapezium Batch**

```
runMode = "BATCH"
dataSource = "HDFS"
dependentWorkflows={
    workflows=[aggregate]
    frequencyToCheck=100
}
hdfsFileBatch = {
    batchTime = 86400
    timerStartDelay = 1
    batchInfo = [{
        name = "DeviceStore"
        dataDirectory = {saiph-devqa=/aggregates}
        fileFormat = "parquet"
    }]
}
```

```
transactions = [{
           transactionName = "DeviceIndexer"
           inputData = [{name = "DeviceStore"}]
           persistDataName = "indexed"
         }]
                                        indexer workflow: batch
  Srclp features
                          DeviceIndexer
                                                        last N days
Parquet Files (last 35
                             tld, zip
                                                       Lucene Indices
                                                DeviceAnalyzer
                                                compare/augment
                                                N min streaming
                                                index refresh daily
                                                             DeviceAnalyzer
                                                             Batch + REST
                                                            index refresh daily
```



## DeviceAnalyzer: Indexing





#### LuceneDAO Index Size

300.0 ---

InputSize(gb)

IndexSize(gb)

rows	In putSize(gb)	IndexSize(gb)
1M	4.0	5.1
4M	14.4	19.0
8M	27.9	35.7
16M	58.8	63.2
73M	276.5	228.0
73M all	276.5	267.1





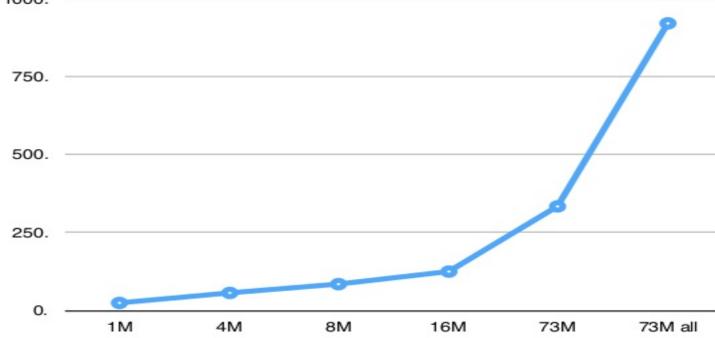
#### LuceneDAO Shuffle Size

1000.

Dictionary(mb)



	Sh. ifflo\\/sito/mh\	Distinger (mb)
rows	ShuffleWrite(mb)	Dictionary(mb)
1M	25	22.0
4M	56	30.0
8M	85	31.6
16M	126	32.2
73M	334	32.4
73M all	921	146.5





#### LuceneDAO Index Runtime

Runtime (s)

20 executors 16 cores Executor RAM 16 GB Driver RAM 8g

3000 -	
2250	
1500 -	
750	

8M

rows Runtime (s)

1M 135

4M 228

8M 434

16M 571

73M 1726

73M all 2456



4M

1M

16M

#rows

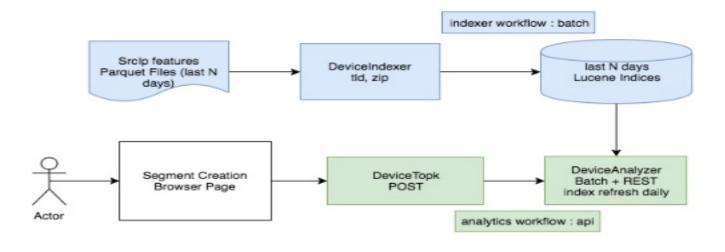
73M

73M all

15

#### Trapezium Api

```
runMode = "BATCH"
dataSource = "HDFS"
httpServer = {
  provider = "akka"
  hostname = "localhost"
  port = 19999
  contextPath = "/"
  endPoints = [{
    path = "analyzer-api"
    className = "TopKEndPoint"
  }]
}
```





#### DeviceAnalyzer: Topk

- Given a query select \* from devices where tld='macys.com' OR 'nordstorm.com' AND (city='SanFrancisco' OR 'Brussels') AND (device='Android') ...
  - ML: Find topk dimensions highly correlated with selected device
  - BI: group by tld order by sum(visits) as tldVisits limit topk



#### **Trapezium Stream**

```
runMode = "STREAM"

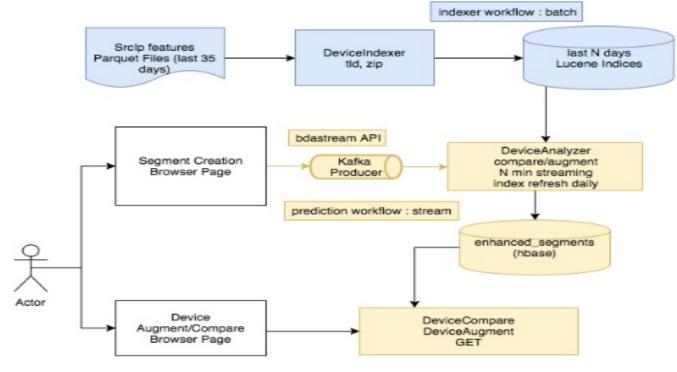
dataSource = "KAFKA"

kafkaTopicInfo = {
    consumerGroup = "KafkaStreamGroup"
    maxRatePerPartition = 970
    batchTime = "5"
    streamsInfo = [{
        name = "queries"
        topicName = "deviceanalyzer"

}]

transactions = [{
    transactionName = DeviceAnalyzer"
    inputStreams = [{name: "queries"}]
    persistStreamName = "deviceanalyzer"
    isPersist = "true"

}]
```





## DeviceAnalyzer: Compare

- Given two queries
   select \* from Devices where
   tld='macys.com' OR 'nordstorm.com' AND (city='SanFrancisco') AND (device='Android')
   select \* from Devices where
   tld='macys.com' OR 'nordstorm.com' AND (city='Brussels') AND (device='Android')
- Find the dimensions that discriminate the devices associated with two groups



## DeviceAnalyzer: Augment

- Given a query
- select \* from Devices where tld='macys.com' OR 'nordstorm.com' AND (city='SanFrancisco' OR 'Brussels') AND (device='Android')...
  - Find devices similar to seed as lookalikes
  - Find dimensions that represent lookalikes

```
object DeviceAnalyzer extends StreamingTransaction {
converter = SparkLuceneConverter(dm.size)
batchTime = Trapezium.getSyncTime("indexer")
dao = LuceneDAO(batchTime...)
   .setConverter(converter).load(sc, indexPath)
dict = loadDictionary(sc, indexPath, batchTime)
all = dao.search("*:*")
def processStream(streams: Map[String, DStream[Row]]):
 streams("queries").collect().map{ request =>
  audience = dao.search(request)
  response = getLookalikeDimensions(all, audience, dict)
        Sparse weighted least squares using
         Breeze QuadraticMinimizer
         L2 regularized linear regression
```



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#### **FastSummarizer**

- Statistical and predictive operators
  - sum: sum over numeric measures
  - support: sum over distinct docID
  - sumSquared: L2 norm
  - gram: Uses BLAS sspr
  - solve: Uses BreezeQuadraticMinimizer to support L1
- Implemented using Array[Float] for shuffle opt
- Scala/Java for Level1 operations
- OpenBLAS for Level3 operations



## Sync API Benchmark

73M rows 1M+ search terms
1 measure on 250K sparse dimensions
20 executors 8 cores
32 GB driver RAM 16 GB executor RAM
akka-http cores: 24 default

runtime(s)

10.5

7.

3.5

0. 1 5 10 20 40 qps

#### topk

qps	runtime(s)
1	1.389
5	1.663
10	3.214
20	5.992
40	12.174



## **Async API Benchmark**

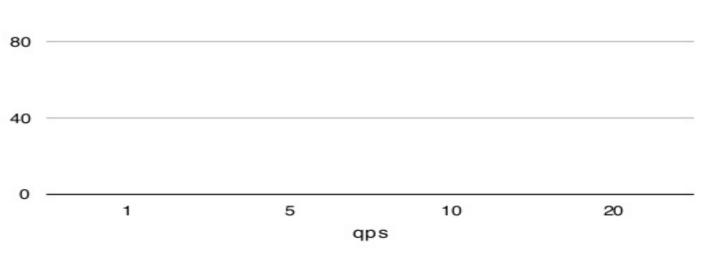
160 -

120 -

compare

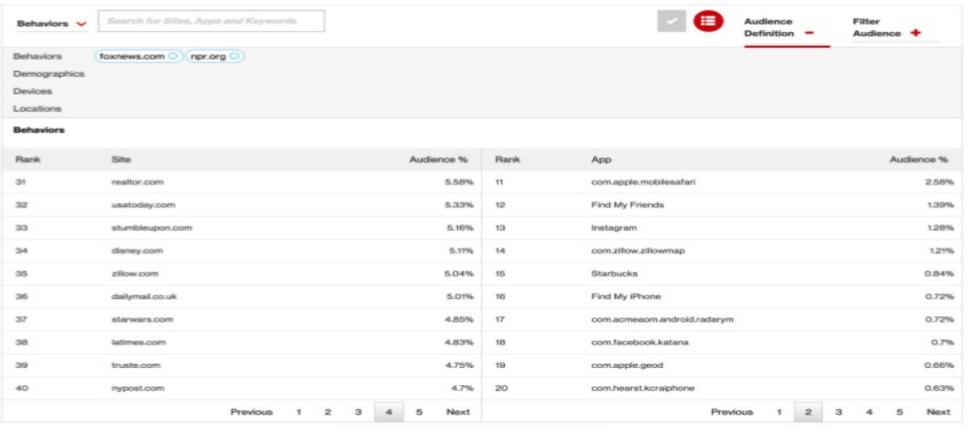
73M rows, 1M+ search terms
1 measure on 250K sparse dimensions
20 executors 8 cores
32 GB driver RAM 16 GB executor RAM
forkjoinpool = 40
Kafka Fetch + compare/augment + HBase Persist

predictions				
qps	compare(s)	augment(s)		
1	9	16		
5	13	36		
10	23	70		
20	42	142		



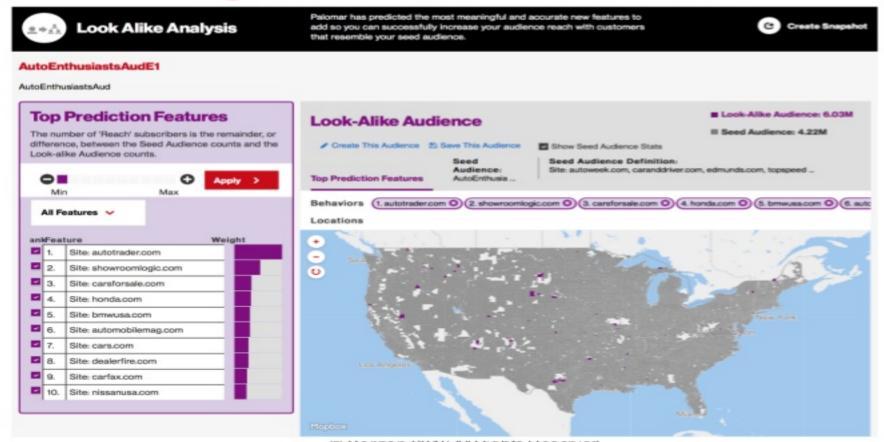


#### topk tld + apps





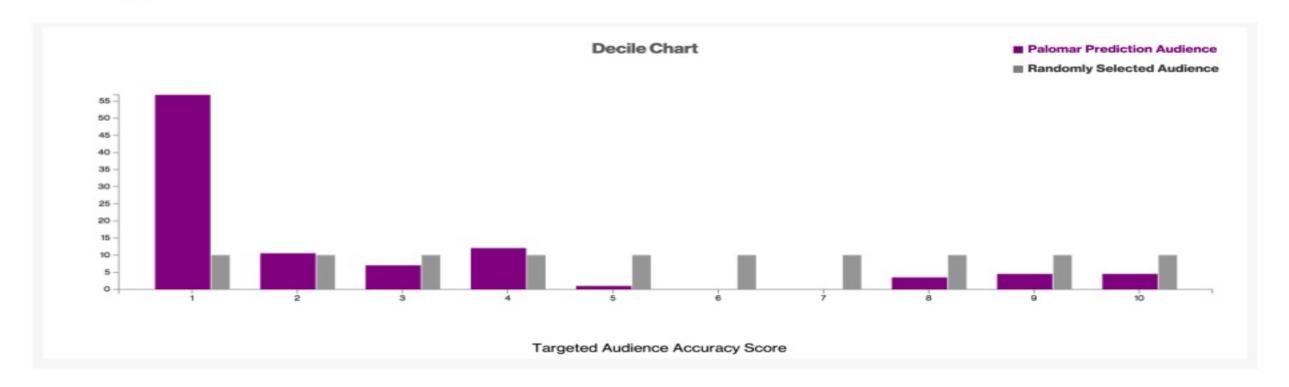
#### Augment: Auto Enthusiastic





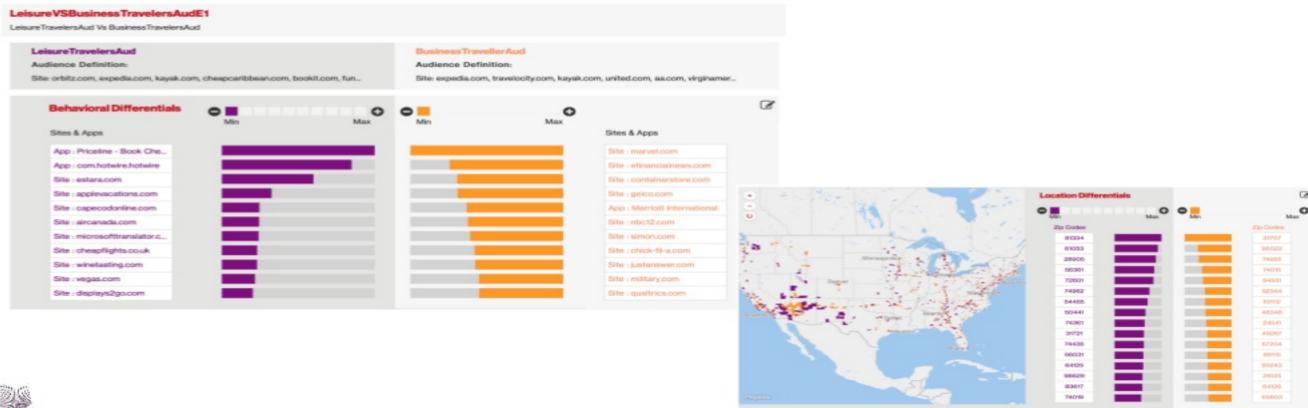
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#### **Augment Model Performance**



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#### Compare: Leisure vs Business Travellers





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