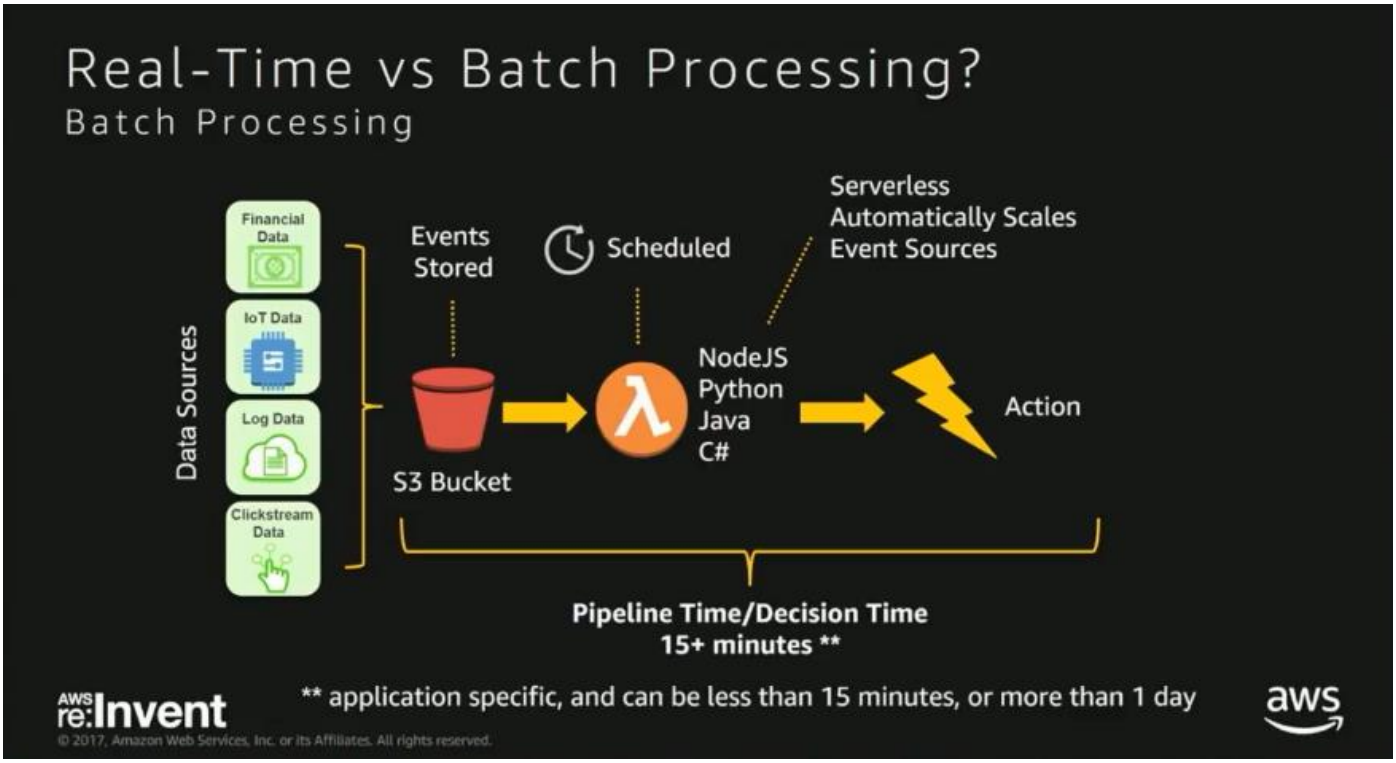
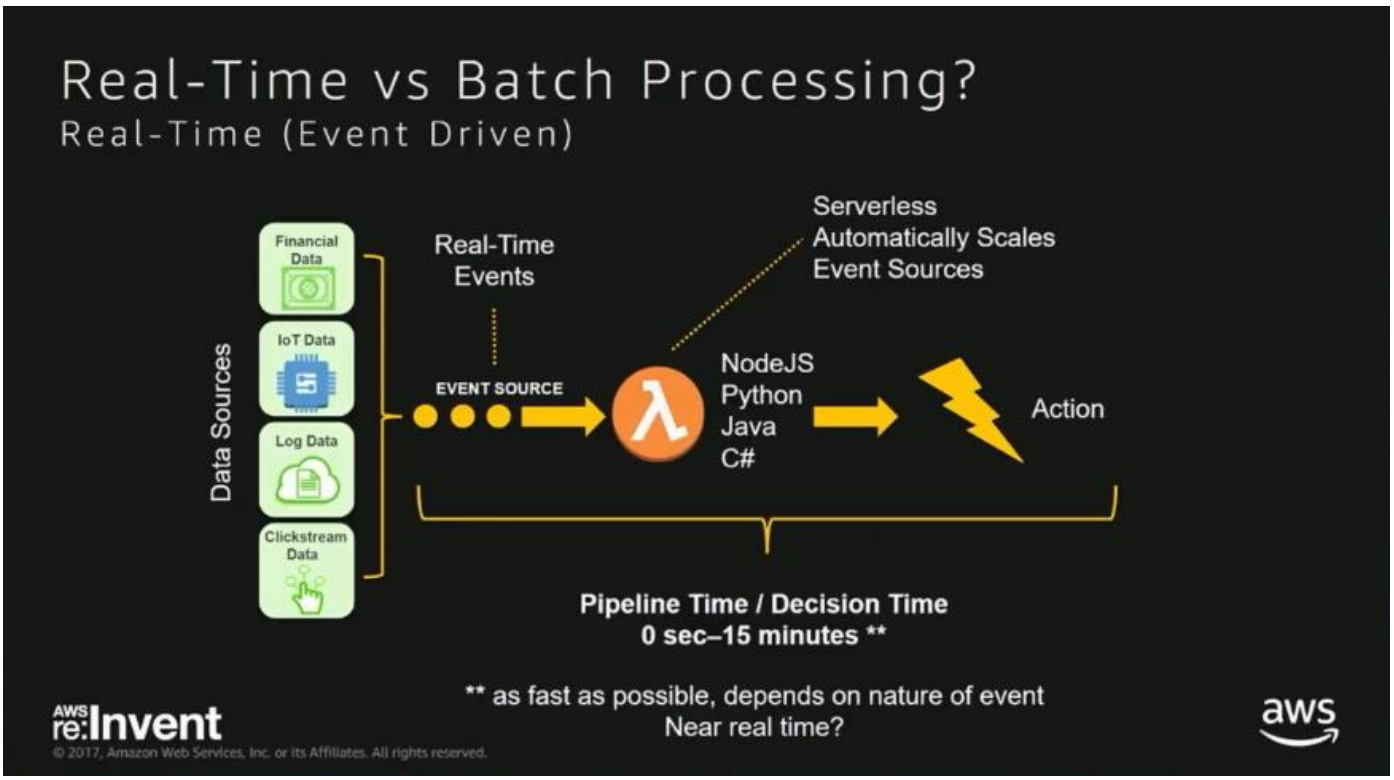
Real-time data processing is a powerful technique that allows businesses to make agile automated decisions. This process is particularly powerful when applied to workloads like ***security***, analyzing access logs, parsing audit logs, and monitoring API activity to detect behavior anomalies. Combined with automation, business can quickly take action to remediate security concerns, or even train a machine learning (ML) model. We explore different techniques for analyzing real-time streams on AWS using ***Lambda***, ***Amazon Kinesis***, Spark with ***Amazon EMR***, and ***Amazon DynamoDB***. We also cover best practices around short- and long-term storage and analysis of data and, briefly, the possibility of leveraging ML.



Batch processing is about collecting events and processing them in some form of scheduled manner. ***Clickstream data*** events can be stored in an ***S3 bucket*** and then use lambda as the compute source to run specialized analytics on the events. All these data can be pulled and analyzed on some specified schedule.



For ***Real-Time processing***, we are going to be processing the message in real-time before storing it in ***S3*** buckets.

AWS Lambda
Serverless Compute

No Infrastructure to manage: Compute without managing infrastructure like Amazon EC2 instances and Auto Scaling groups.

Bring your own code: Stateless, event-driven code with native support for Node.js, Java, Python, and C# languages.

Triggered by events: Direct Sync & Async API calls, AWS service integrations, and third-party triggers.

Efficient performance at scale: Easy to author, deploy, maintain, secure, and manage. Focus on business logic to build back-end services that perform at scale.

Cost-effective: Automatically matches capacity to request rate. Compute cost 100 ms increments.

Data Sources and Applications

Parse Load Balancer Access Logs looking for anomalous patterns

Load Balancer

Parse VPC Flow Logs to detect anomalies and abuse

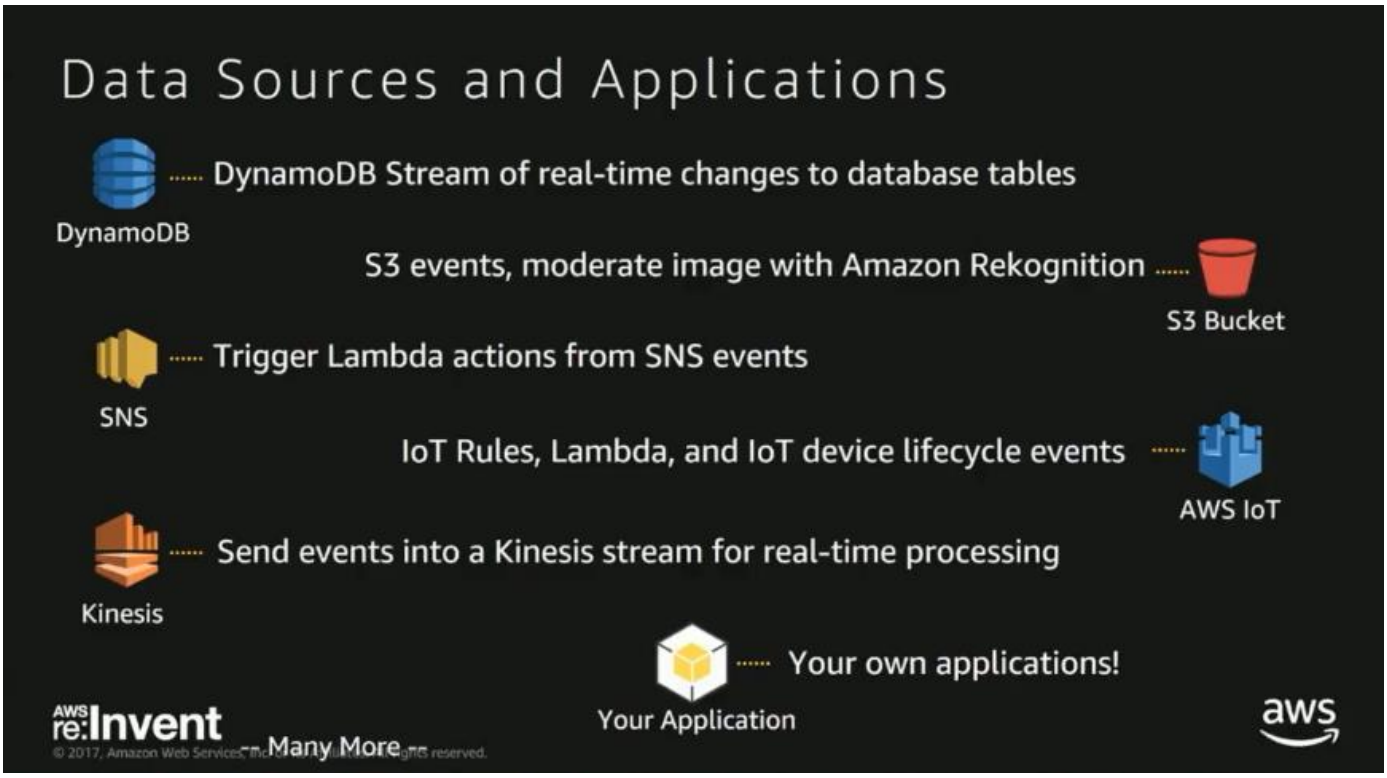
Flow Logs

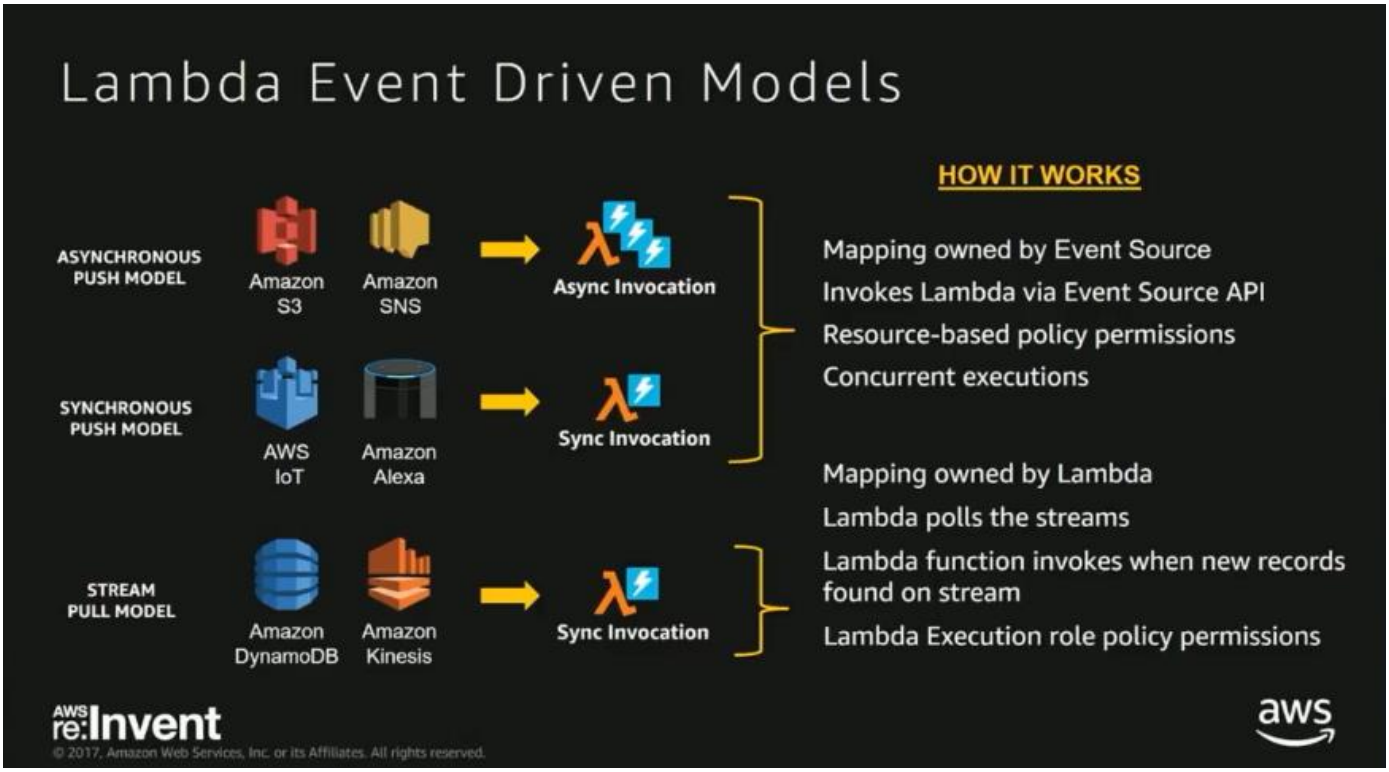
Parse CloudTrail events to detect suspicious patterns such as logins from another country

CloudTrail

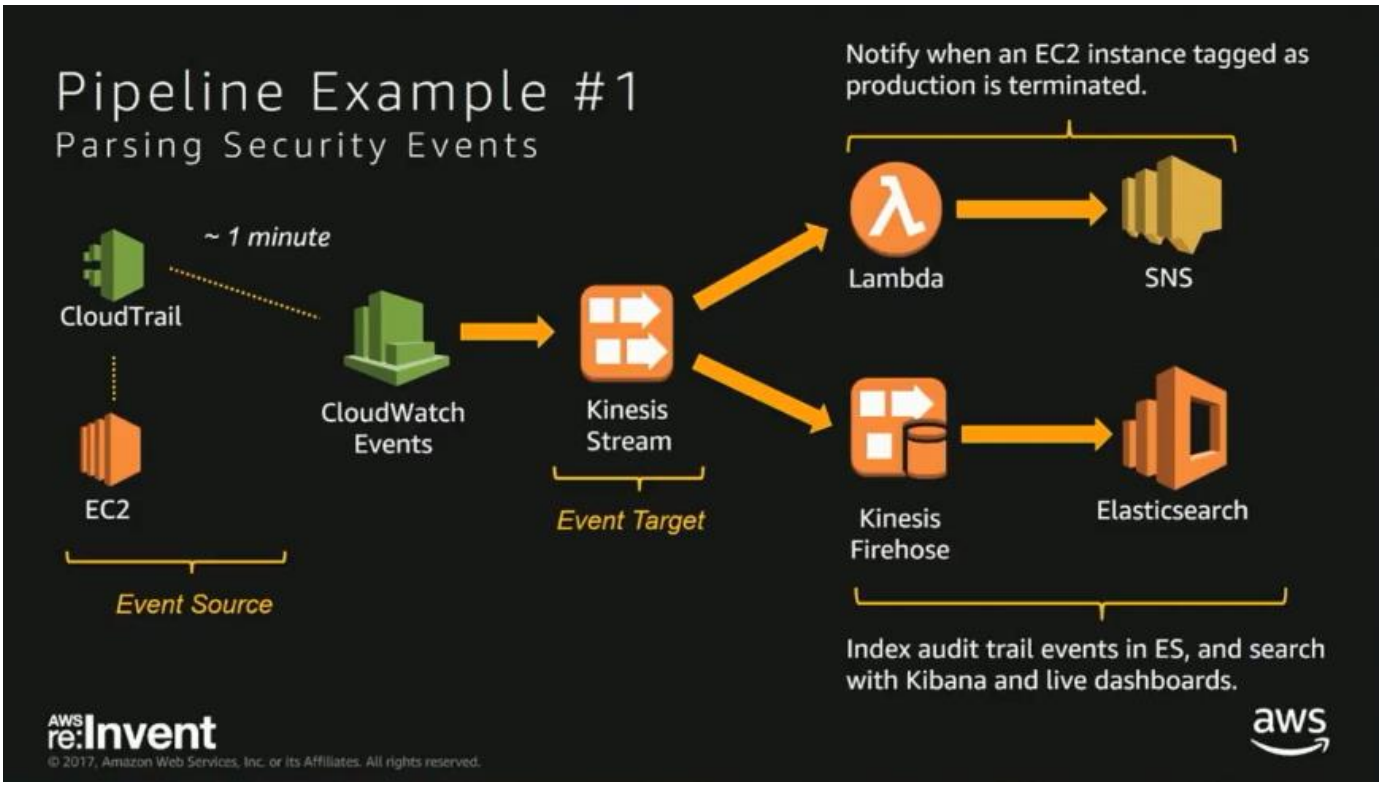
Configure event triggering and alarms from sources such as CloudTrail or other event sources

CloudWatch
(logs, events, alarms)





These are the 2 types of invocations models for triggering a lambda function. Sync invocation is when you have to wait on the lambda to get the result back and use further down the pipeline.



We are going to turn on and use ***CloudTrail*** monitor all your AWS ***EC2*** calls, this will let us know if something was turned off and by whom. We then set up a ***CloudWatch event*** on a specific event like ***EC2 Prod instance termination*** from the ***CloudTrail logs*** and put that event in a ***Kinesis Stream***. We then have 2 consumers that are going to be consuming data coming from the Kinesis stream, a ***lambda function*** will put the message into an ***SNS*** topic, and a ***Kinesis Firehose*** that will put the event into ***ElasticSearch*** for searching and display using Kibana

Configuring CloudWatch Event Rules

Parsing Security Events

CloudWatch Event Source Pattern

```
{
  "source": [
    "aws.ec2"
  ],
  "detail-type": [
    "AWS API Call via CloudTrail"
  ],
  "detail": {
    "eventSource": [
      "ec2.amazonaws.com"
    ],
    "eventName": [
      "TerminateInstances"
    ]
  }
}
```

- Will get EC2 CloudTrail audit events on the account
  - CloudTrail audit events track API events, which includes the principal, target resource, and action that took place
  - Specifically filtering for "TerminateInstances" events
  - On average, events are emitted in ~1–2 minutes

What Does the Event Look Like?

Parsing Security Events

What do we need?

- IAM user that terminated the instance
- When the instance was terminated
- IP address of user that terminated the instance
- What instance was terminated

Summarized

```
{
  "id": "aa7d2051-b082-6b8e-8d36-f7e72064ab5c",
  "detail-type": "AWS API Call via CloudTrail",
  "source": "aws.ec2",
  "account": "XXXXXXXXXXX",
  "time": "2017-10-03T19:46:45Z",
  "region": "us-east-1",
  "detail": {
      "userIdentity": {
          "type": "IAMUser",
          "arn": "arn:aws:iam::332631645289:user/jcv123",
          "accountId": "XXXXXXXXXXX",
          "userName": "jcv123"
      },
      "eventTime": "2017-10-03T19:46:45Z",
      "eventSource": "ec2.amazonaws.com",
      "eventName": "TerminateInstances",
      "awsRegion": "us-east-1",
      "sourceIPAddress": "324.21.198.71",
      "requestParameters": {
          "instancesSet": {
              "items": [ { "instanceId": "i-098aa789cd6b411c0" } ]
          }
      },
      "requestID": "3b6aeef7-c704-40f8-9886-a2091f04d290",
      "eventID": "e9a6003d-12e7-4efa-b71d-a13f4f171fd5",
      "eventType": "AwsApiCall"
  }
}
```

Lambda Function

Parsing Security Events

- Send SMS message to administrator when a user terminates an instance tagger as a production instance
- We could also send a message to a Slack Channel, or Pager Duty, or send an email
- Got notified in 30 seconds!

```
import base64
import json
import boto3

ec2_resource = boto3.resource('ec2')
sns_client = boto3.client('sns')

def handler(event, context):
  for record in event['Records']:
    # kinesis data is base64 encoded so decode here
    payload = json.loads(base64.b64decode(record['kinesis']['data']))
    print('Payload: {}'.format(payload))
    # process event
    user_name = payload['detail']['userIdentity']['userName']
    instance_id = payload['detail']['requestParameters']['instancesSet']['items'][0]['instanceId']
    ec2_instance = ec2_resource.Instance(instance_id)
    for tags in ec2_instance.tags:
      if tags['Key'] == 'Type' and tags['Value'] == 'Production':
        print('Production instance was terminated, notifying!')
        # a production instance was terminated, let's send a notification
        sns_client.publish(
          PhoneNumber='+15552562562',
          Message='User {} has terminated production instance {}'.format(user_name, instance_id))
```

Imports

Boto3 clients

Parse Kinesis Record

Extract Username and InstanceId

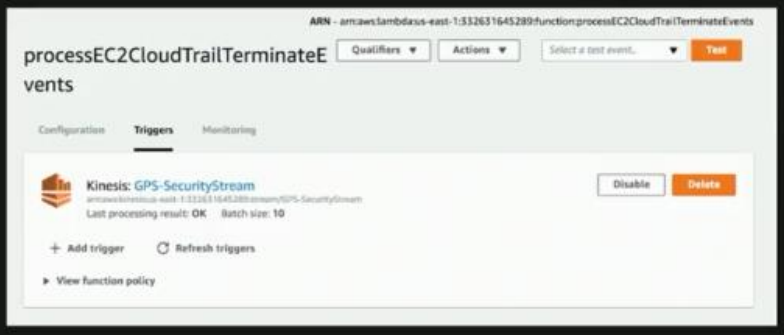
Check if instance is Production

Send SMS!

Executing the Lambda Function

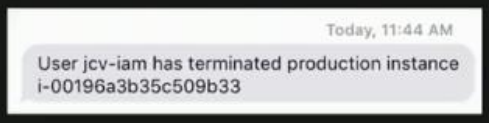
Parsing Security Events

- Configure a Kinesis trigger for our Lambda function
- Lambda service handles the Kinesis ingestion and execution of the Lambda function
- Also handles errors and retries

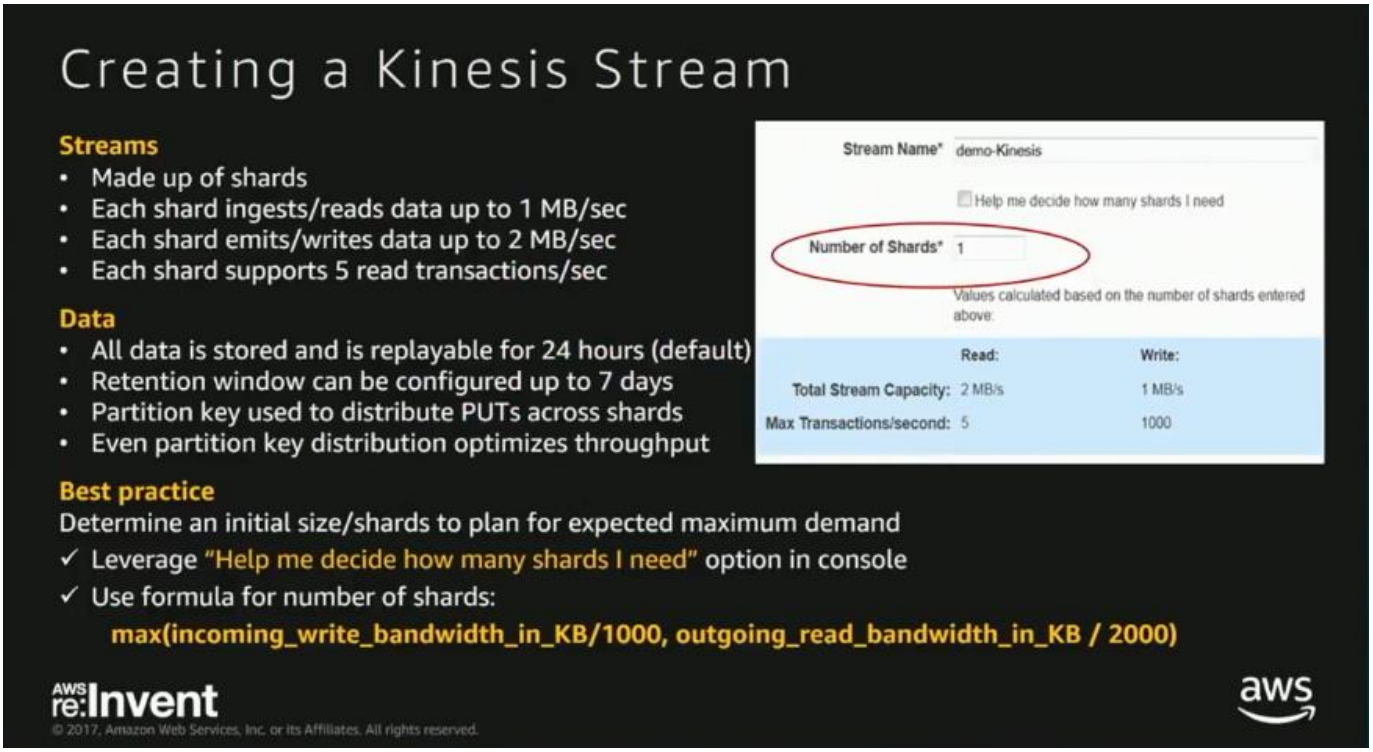


- Finally, we get our text message when an instance is terminated

Huzzah!



Lambda functions are synchronously invoked, meaning that there is a guarantee or message order processing. This is why there is the concept of retries to enable the stoppage of the real-time processing of your pipeline messages before they fall off the Trim Horizon. You can fix the issue with retries before messages start getting dropped at the trim horizon, then your processing will continue as expected.



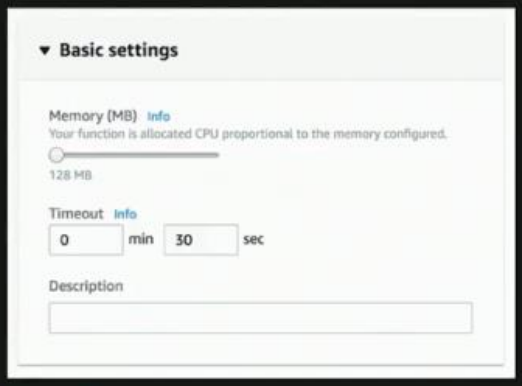
Creating Lambda Function

Memory
CPU allocation proportional to the memory configured
Increasing memory makes your code execute faster (if CPU bound)
Increasing memory allows for larger record sizes processed

Timeout
Increasing timeout allows for longer functions, but longer wait in case of errors

Permission model
Execution role defined for Lambda must have permission to access the stream

Retries
With Amazon Kinesis, Lambda retries until the data expires (24 hours)



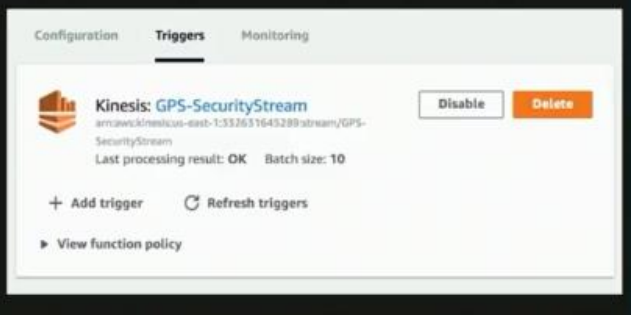
Best practice
Write Lambda function code to be stateless
Instantiate AWS clients and database clients **outside** the scope of the function handler to take advantage of connection re-use.

Configuring Event Trigger

Amazon Kinesis mapped as event source in **Lambda**

Batch size
Max number of records that Lambda will send to one invocation
Not equivalent to effective batch size
Effective batch size is every 250 ms—calculated as:
MIN(records available, batch size, 6MB)
Increasing batch size allows fewer Lambda function invocations with more data processed per function

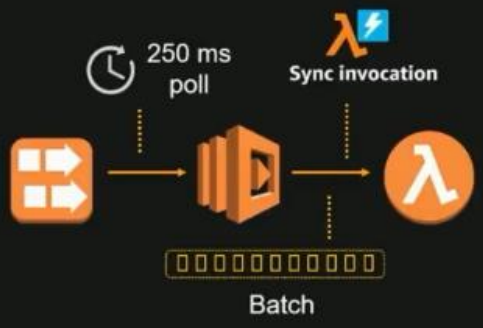
Best Practices
Set to "Trim Horizon" for reading from start of stream (all data)
Set to "Latest" for reading most recent data (LIFO) (latest data)



How It Works

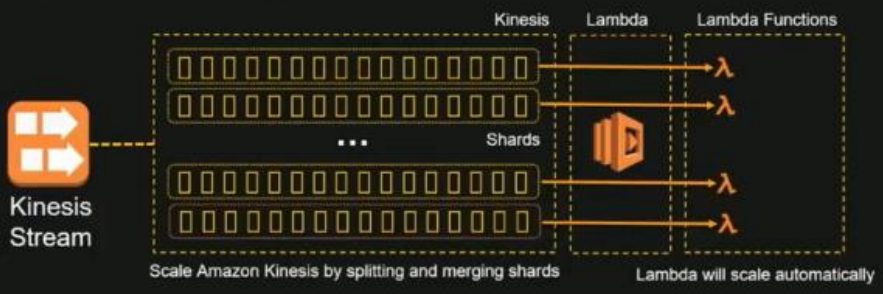
Polling
Concurrent polling and processing per shard
Lambda polls every 250 ms if no records found
Will grab as much data as possible in one GetRecords call (Batch)

Batching
Batches are passed for invocation to Lambda through function parameters
Batch size may impact duration if the Lambda function takes longer to process more records
Sub batch in memory for invocation payload



Synchronous invocation
Batches invoked as synchronous RequestResponse type
Lambda honors Amazon Kinesis at least once semantics
Each shard blocks in order of synchronous invocation

Tuning Throughput

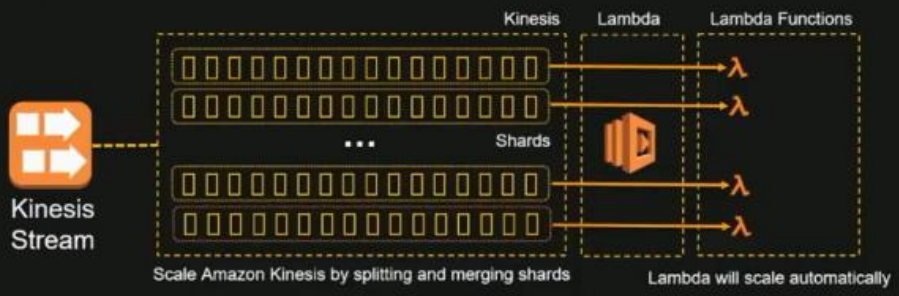


If put/ingestion rate is greater than the theoretical throughput, your processing is at risk of falling behind

Maximum theoretical throughput
shards * 2 MB / Lambda function duration (s)

Effective theoretical throughput
shards * batch size (MB) / Lambda function duration (s)

Tuning Throughput with Retries



Retries

Will retry on execution failures until the record is expired

Throttles and errors impact duration and directly impact throughput

Best Practice

Retry with exponential back-off of up to 60 s

Effective theoretical throughput with retries

(# shards * batch size (MB)) / (function duration (s) * retries until expiry)

Monitoring Your Real-Time Pipeline

Monitoring Amazon Kinesis Streams

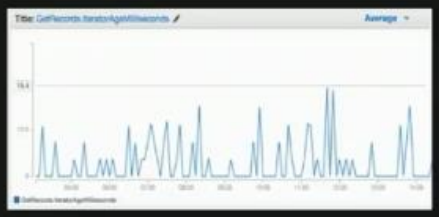
- GetRecords: (effective throughput)
- PutRecord: bytes, latency, records, etc.
- GetRecords.IteratorAgeMilliseconds: how old your last processed records were

Monitoring Lambda functions

- Invocation count: time function invoked
- Duration: execution/processing time
- Error count: number of errors
- Throttle count: number of time function throttled
- Iterator age: time elapsed from batch received and final record written to stream

Debugging

- Review all metrics
- Make custom logs
- View RAM consumed
- Search for log events



Example Debugging with X-Ray

AWS X-Ray

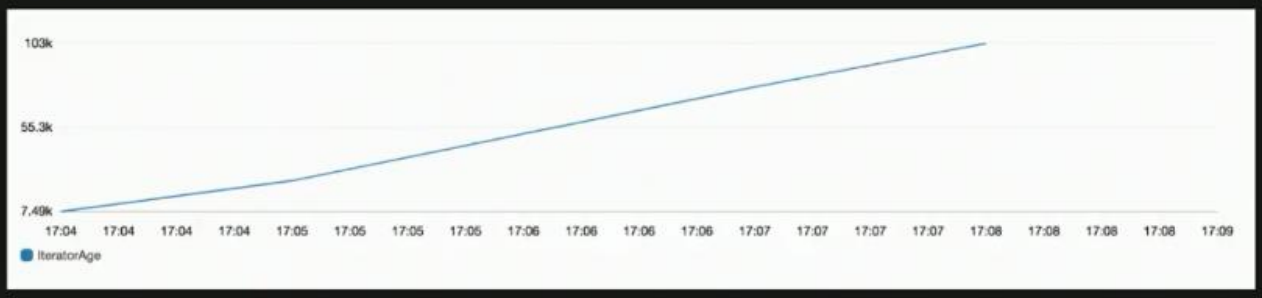
Retries with Lambda/Exponential Back-offs

- Total Duration = 6 seconds (five retries before success)
- Lambda Invocation Duration = 2.2 seconds
- Actual Throughput = 1 * (10000/6) = 1666 records/s
- Ideal Throughput = 1 * (10000/2.2) = 4545 records/s

2.72x difference

Falling Behind

Not Processing Fast Enough

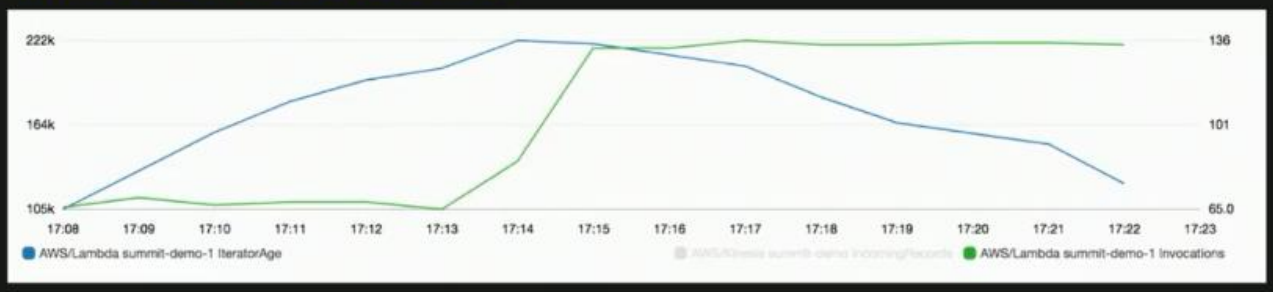


- Producers are inserting 100k records per minute
- 2 shards
- Lambda functions take 1.7 seconds to process 1000 records
- 2 * (1000/1.7) = 1176 records/s = ~70k/minute

100k produced per minute > ~70k processed per minute

Resharding to Increase Throughput

Increase Number of Shards to Catch Up



- Increased number of shards to 4 (doubled)
- Lambda functions take 1.7 seconds to process 1000 records
- 4 * (1000/1.7) = 2352 records/s = ~141k / minute

100k produced per minute < ~141k processed per minute
REAL TIME!

Netflix and AWS

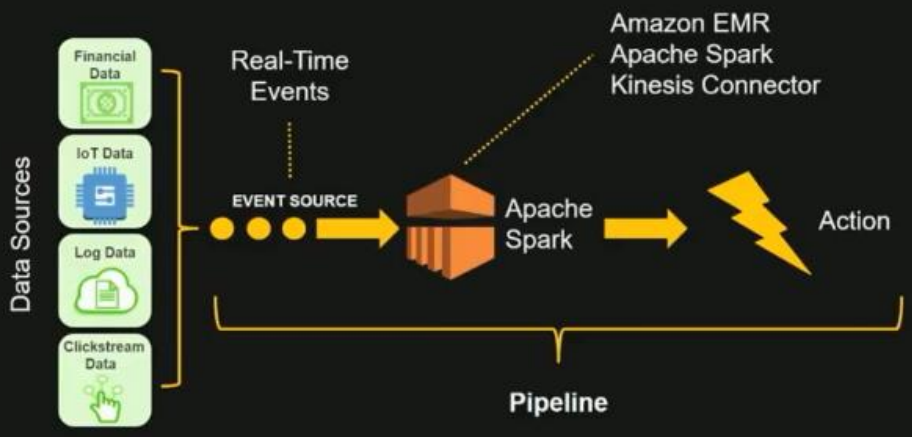
"Amazon Kinesis Streams processes multiple terabytes of log data each day, yet events show up in our analytics in seconds. We can discover and respond to issues in real time, ensuring high availability and a great customer experience."

–John Bennett
Senior Software Engineer, Netflix

NETFLIX

- Netflix needed a solution for ingesting, augmenting, and analyzing the multiple terabytes of data its network generates daily in the form of virtual private cloud (VPC) flow logs.
- Netflix uses AWS to analyze billions of messages across more than 100,000 application instances daily in real time, enabling it to optimize user experience, reduce costs, and improve application resilience.

What about EMR?



Why EMR + Apache Spark?

- Mature features and libraries
- Machine learning functionality with SparkML
- Interact with Data using SparkSQL
- Spark Streaming provides support for operators states such as sliding windows
- Fault tolerance/recovery out of the box
- Ability to combine batch and streaming
- Multiple languages (Scala, Java, Python, R, more)

SparkSQL (Python)

```
context = HiveContext(sc)
results = context.sql(
  "SELECT * FROM people")
names = results.map(lambda p: p.name)
```

SparkML (Python)

```
data = spark.read.format("libsvm")\
.load("hdfs://...")

model = KMeans(k=10).fit(data)
```

GraphX (Scala)

```
graph = Graph(vertices, edges)
messages = spark.textFile("hdfs://...")
graph2 = graph.joinVertices(messages) {
  (id, vertex, msg) => ...
}
```

Spark

Zillow and AWS

> "We can compute Zestimates in seconds, as opposed to hours, by using Amazon Kinesis Streams and Spark on Amazon EMR."
>
> *–Jasjeet Thind*
> *Vice President of Data Science and Engineering*

Zillow provides online home information to tens of millions of buyers and sellers every day.

- Needed to provide timely home valuations for all new homes
- Runs Zestimate, its machine learning-based home-valuation tool, on AWS
- Performs machine-learning jobs in hours instead of a day
- Gives customers more accurate data on more than 100 million homes
- Scales storage and compute capacity on demand

Next Steps

- ✓ Learn more about AWS Serverless at https://aws.amazon.com/serverless
- ✓ Explore the AWS Lambda Reference Architecture on GitHub:
  - Real-Time Streaming: https://github.com/awslabs/lambda-refarch-streamprocessing
  - Distributed Computing Reference Architecture (serverless MapReduce) https://github.com/awslabs/lambda-refarch-mapreduce

Next Steps

- ✓ Create an Amazon Kinesis stream. Visit the Amazon Kinesis Console and configure a stream to receive data Ex. data from Social media feeds.
- ✓ Create and test a Lambda function to process streams from Amazon Kinesis by visiting Lambda console. First 1M requests each month are on us!
- ✓ Read the Developer Guide and try the Lambda and Amazon Kinesis Tutorial:
  - http://docs.aws.amazon.com/lambda/latest/dg/with-kinesis.html
- ✓ Send questions, comments, feedback to the AWS Lambda Forums

AWS re:Invent

aws