



Turn Data into Insights with Data Lakes and Analytics on AWS

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Data is a strategic asset for every organization

“The world’s most valuable resource is no longer oil, but **data**.*”



*Copyright: The Economist, 2017, David Parkins

Data Analytics Workflow



Traditionally data Analytics revolved around data warehouses



Customers want more value from their data



Growing
exponentially



From new
sources



Increasingly
diverse

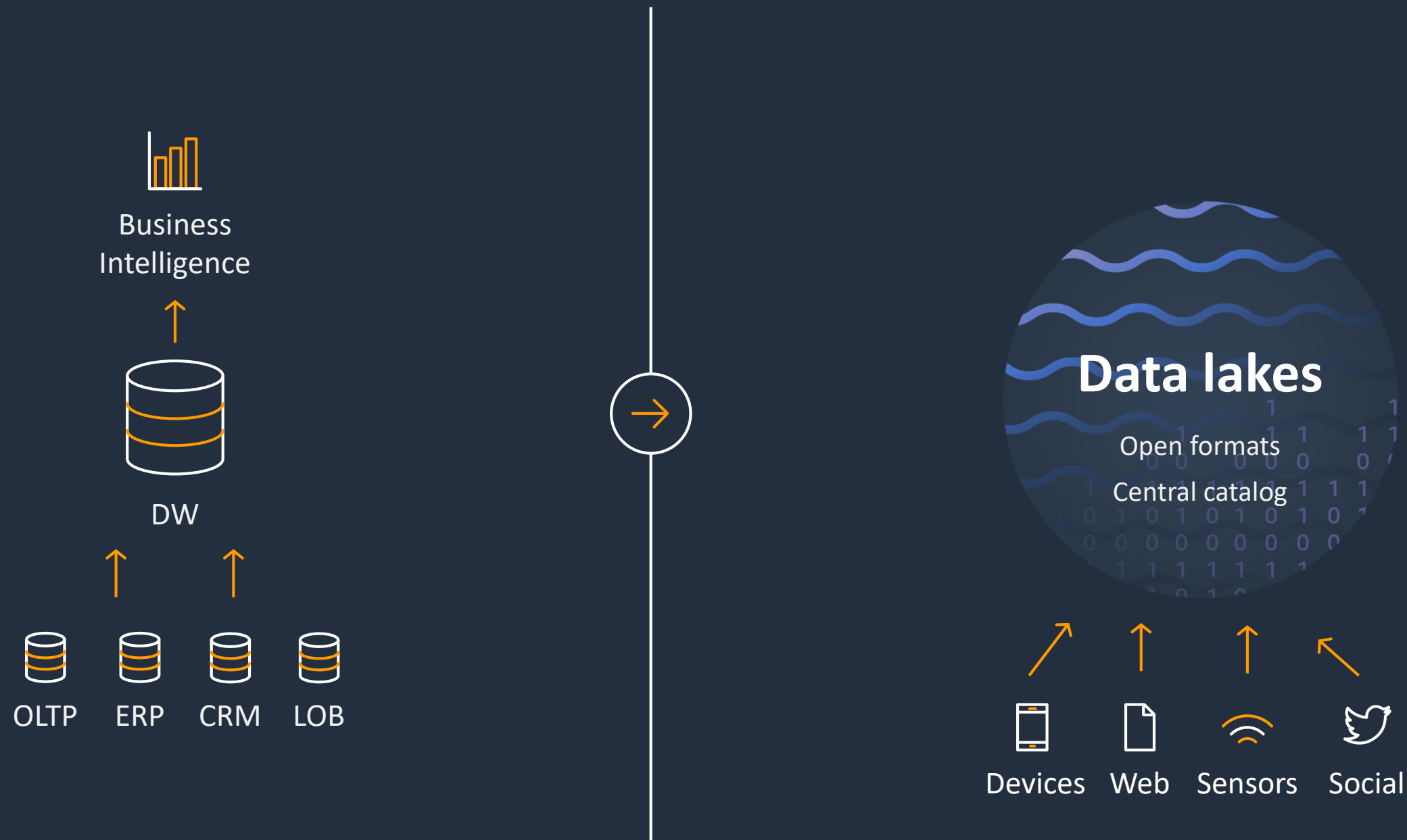


Used by
many people



Analyzed by
many applications

Customers moving to data lake architectures



Bringing together the best of both worlds



Extends or evolves DW architectures

Store any data in any format

Durable, available, and exabyte scale

Secure, compliant, auditable

Run any type of analytics from DW to Predictive

Why choose AWS for data lakes and analytics?

1. Easiest to build data lakes and analytics

Amazon S3: Highly durable, secure and scalable single storage layer for all analytics and ML

AWS Lake Formation: A service to build secure data lakes in days instead of months.

Deep integration across analytics & infrastructure(including federated queries)

The fastest way to go from zero to insights, covering all
data for all users

2. Most secure infrastructure for analytics

Services for security and governance



Security

Amazon GuardDuty

AWS Shield

AWS WAF

Amazon Macie

VPC



Identity

AWS IAM

AWS SSO

Amazon Cloud Directory

AWS Directory Service

AWS Organizations



Encryption

AWS Certification Manager

AWS Key Management Service

Encryption at rest

Encryption in transit

Bring your own keys,
HSM support



Compliance

AWS Artifact

Amazon Inspector

Amazon Cloud HSM

Amazon Cognito

AWS CloudTrail

2. Most secure infrastructure: certifications

Global		United States			
	CSA Cloud Security Alliance Controls		CJIS Criminal Justice Information Services		ITAR International Arms Regulations
	ISO 9001 Global Quality Standard		DoD SRG DoD Data Processing		MPAA Protected Media Content
	ISO 27001 Security Management Controls		FedRAMP Government Data Standards		NIST National Institute of Standards and Technology
	ISO 27017 Cloud Specific Controls		FERPA Educational Privacy Act		SEC Rule 17a-4(f) Financial Data Standards
	ISO 27018 Personal Data Protection		ISO FFIEC Financial Institutions Regulation		VPAT/Section 508 Accountability Standards
	PCI DSS Level 1 Payment Card Standards		FIPS Government Security Standards	Asia Pacific	
	SOC 1 Audit Controls Report		FISMA Federal Information Security Management		
	SOC 2 Security, Availability, & Confidentiality Report		GxP Quality Guidelines and Regulations		IRAP [Australia] Australian Security Standards
	SOC 3 General Controls Report		HIPPA Protected Health Information		K-ISMS [Korea] Korean Information Security
					MTCS Tier 3 [Singapore] Multi-Tier Cloud Security Standard
					My Number Act [Japan] Personal Information Protection
				Europe	
					C5 [Germany] Operational Security Attestation
					Cyber Essentials Plus [UK] Cyber Threat Protection
					G-Cloud [UK] UK Government Standards
					IT-Grundschutz [Germany] Baseline Protection Methodology



3. Most comprehensive and open

Data, visualization, engagement, & machine learning



Data



Dashboards



Digital User Engagement



Predictive Analytics

Analytics



Data Warehousing



Big Data Processing



Serverless Data processing



Interactive Query



Operational Analytics



Real time Analytics

Data lake infrastructure & management



Infrastructure



Security & Management



Data Catalog & ETL

Data movement

Migration & Streaming Services

3. Most comprehensive and open

Data, visualization, engagement, & machine learning

NEW



Data Exchange



QuickSight



Pinpoint



SageMaker



Comprehend



Lex



Polly



Rekognition



Translate

+ many more

Analytics



Redshift



EMR (Spark & Hadoop)



AWS Glue
(Spark & Python)



Athena



Elasticsearch
Service



Kinesis Data
Analytics

Data lake infrastructure & management



S3/Glacier



Lake
Formation



AWS Glue

Data movement

Database Migration Service | Snowball | Snowmobile | Kinesis Data Firehose | Kinesis Data Streams | Managed Streaming for Apache Kafka

3. Open standards, formats, and Apache open source

Flink	Mahout	PyTorch
Ganglia	MapReduce	R
Hbase	MxNET	Scala
HCatalog	MySQL	Spark
HDFS	Oozie	Sqoop
Hive	ORC	SQL
Hudi	Parquet	TensorFlow
Java	Phoenix	Tez
JupyterHub	Pig	YARN
Kafka	Presto	Zeppelin
Livy	Python	Zookeeper

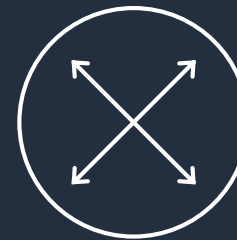
4. Most scalable, cost-effective, high-performance infrastructure for analytics



On-demand,
Reserved, and
Spot instances to
reduce costs



100 Gbps
bandwidth
network interfaces
for performance



Industry leading choice
of 200+ instance types
to meet workload
needs



Five highly
available storage
tiers and
intelligent tiering

4. Most scalable, cost-effective infrastructure for analytics

Some examples of advanced cost saving capabilities in analytics services



EMR

Autoscaling

57% less than
on-premises
per IDC report



Redshift

Less than 1/10th
of the cost of
traditional, on-
premises solutions



Athena & QuickSight

Serverless pay
only for what is used

Pricing per session
for visualization

Get answers from your data in AWS Data Lake

Query Directly with Amazon Athena

Athena

Query Editor

Saved Queries

History

Data sources

Workgroup : primary

Settings

Tutorial

Help

What's new

Data source

Connect data source

awsdatacatalog

Database

ticketdata

Filter tables and views...

Tables (22)

Create table

mlb_data

name_data

nfl_data

nfl_stadium_data

parquet_person

parquet_sport_location

parquet_sport_team

parquet_sporting_event

parquet_sporting_event_ticket

person

player

seat

seat_type

sport_division

sport_league

sport_location

sport_team

sporting_event

sporting_event_ticket

sporting_event_ticket_1bb4a008b349ed873527a4c2b9f8ac5f

ticket_purchase_hist

ticket_purchase_hist_95f83e3d847527d7c4e84a4949d62d2b

Views (2)

Create view

sporting_event_info

sporting_event_ticket_info

New query 1

New query 2

New query 3

New query 4

1

2

3

SELECT "mlb_id", "mlb_name", "mlb_pos", "mlb_team", "mlb_team_long", "bats", "throws", "birth_year", "bp_id", "bref_id", "bref_name", "cbs_id", "cbs_name", "cbs_pos", "espn_id", "espn_name", "espn_pos", "fg_id", "fg_name", "lahman_id", "nfb_id", "nfb_name", "nfb_pos", "retro_id", "retro_name", "debut", "yahoo_id", "yahoo_name", "yahoo_pos", "mlb_depth"

FROM "ticketdata"."mlb_data"

LIMIT 1000

Run query

Save as

Create

(Run time: 1.77 seconds, Data scanned: 395.04 KB)

Format query

Clear

Use Ctrl + Enter to run query, Ctrl + Space to autocomplete

Results

mlb_id

mlb_name

mlb_pos

mlb_team

mlb_team_long

bats

throws

birth_year

bp_id

bref_id

bref_name

cbs_id

cbs_

1

+5.065600000000000e+05

Alexi Amarista

3B

SD

San Diego Padres

L

R

1989

+5.588900000000000e+04

amarial01

Alexi Amarista

1735053

Alexi A

2

+4.582100000000000e+05

Alexi Casilla

2B

TB

Tampa Bay Rays

S

R

1984

+4.579900000000000e+04

casilal01

Alexi Casilla

1103724

Alexi C

3

+4.683960000000000e+05

Alexi Ogando

P

ATL

Atlanta Braves

R

R

1983

+4.991000000000000e+04

ogandal01

Alexi Ogando

1174266

Alexi O

4

+4.696860000000000e+05

Alfredo Aceves

P

NYN

New York Yankees

R

R

1982

+4.692800000000000e+04

aceveal01

Alfredo Aceves

1638980

Alfredo

5

+4.516280000000000e+05

Alfredo Figaro

P

TEX

Texas Rangers

R

R

1984

+5.245300000000000e+04

figaral01

Alfredo Figaro

1654334

Alfredo

6

+5.540540000000000e+05

Alfredo Gonzalez

C

CWS

Chicago White Sox

R

R

1992

+6.920100000000000e+04

2210083

Alfredo

7

+5.012450000000000e+05

Alfredo Marte

LF

BAL

Baltimore Orioles

R

R

1989

+5.175500000000000e+04

marteal01

Alfredo Marte

1956711

Alfredo

8

+4.305800000000000e+05

Alfredo Simon

P

CIN

Cincinnati Reds

R

R

1981

+4.558100000000000e+04

simonal01

Alfredo Simon

448968

Alfredo

9

+4.553780000000000e+05

Ali Solis

C

LAD

Los Angeles Dodgers

R

R

1987

+5.236200000000000e+04

solisal01

Ali Solis

1946524

Ali Solis

10

+4.888520000000000e+05

Allan Dykstra

1B

TB

Tampa Bay Rays

L

R

1987

+5.781300000000000e+04

dykstal01

Allan Dykstra

1960245

Allan D

11

+5.018000000000000e+05

Allen Craig

1B

BOS

Boston Red Sox

R

R

1984

+5.104300000000000e+04

craigal01

Allen Craig

1661498

Allen C

12

+5.439030000000000e+05

Allen Webster

P

PIT

Pittsburgh Pirates

R

R

1990

+5.878700000000000e+04

webstal01

Allen Webster

1799467

Allen W

13

+6.072370000000000e+05

Amir Garrett

P

CIN

Cincinnati Reds

R

L

1992

+7.094600000000000e+04

2053474

Amir G

14

+4.448430000000000e+05

Andre Ethier

RF

LAD

Los Angeles Dodgers

L

L

1982

+4.596200000000000e+04

ethiean01

Andre Ethier

490390

Andre E

15

+5.165890000000000e+05

Andre Rienzo

P

MIA

Miami Marlins

R

R

1988

+5.663800000000000e+04

rienzan01

Andre Rienzo

2027388

Andre F

16

+5.927430000000000e+05

Andrelton Simmons

SS

LAA

Los Angeles Angels

R

R

1989

+6.711900000000000e+04

simmoan01

Andrelton Simmons

1918584

Andrelt

17

+4.332170000000000e+05

Andres Blanco

2B

PHI

Philadelphia Phillies

S

R

1984

+3.283500000000000e+04

blancan01

Andres Blanco

392116

Andres

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Analyze with Hadoop on Amazon EMR

Create Cluster - Advanced Options [Go to quick options](#)

Step 1: Software and Steps

Step 2: Hardware

Step 3: General Cluster Settings

Step 4: Security

Software Configuration

Vendor ☒ Amazon ☐ MapR

Release

☒ Hadoop 2.6.0

☒ Zeppelin-Sandbox 0.5.5

☐ Ganglia 3.6.0

☐ Pig 0.14.0

☒ Hive 1.0.0

☐ Hue 3.7.1

☐ Presto-Sandbox 0.125

☐ Mahout 0.11.0

☒ Spark 1.5.2

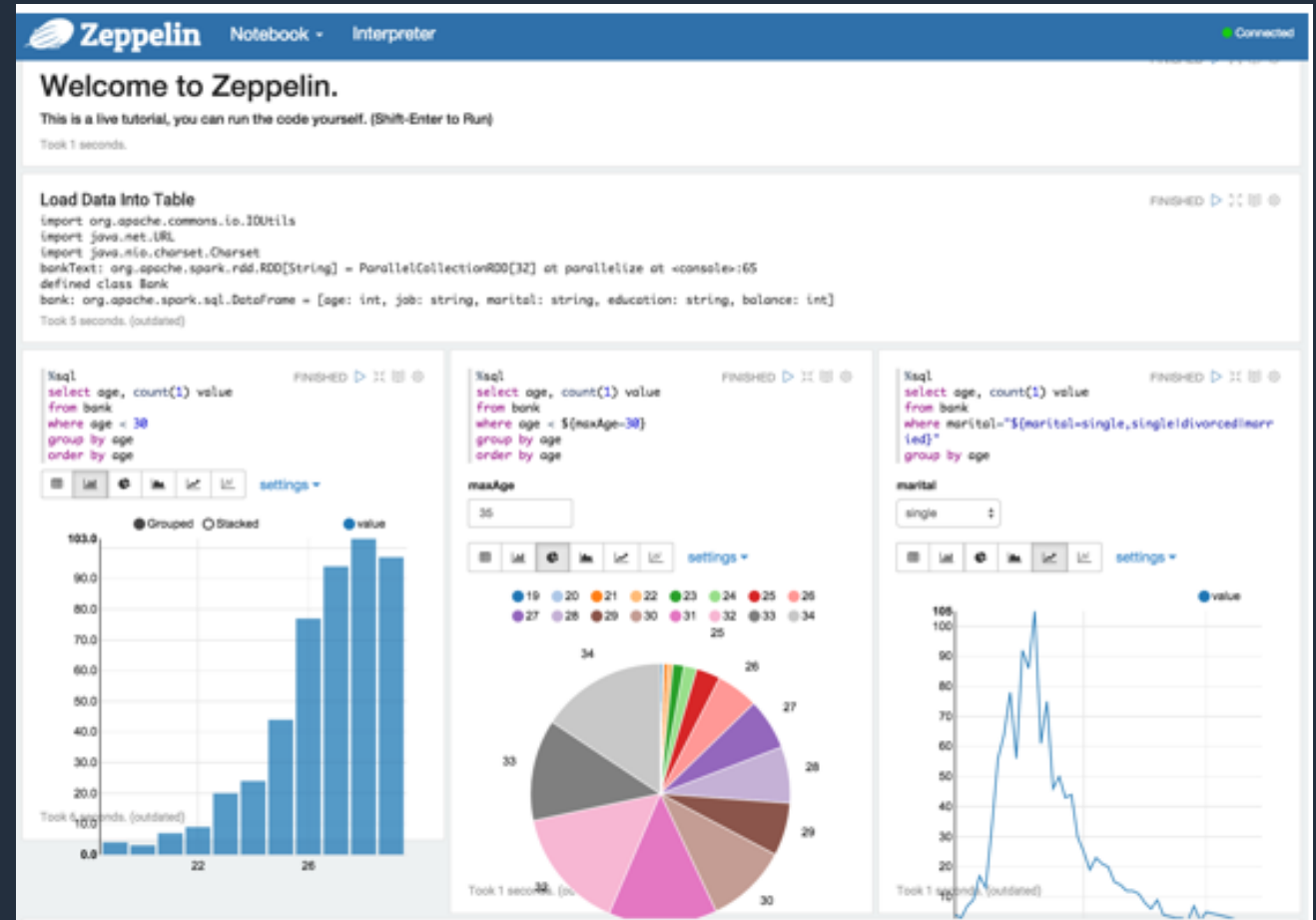
☐ Oozie-Sandbox 4.2.0

Scala

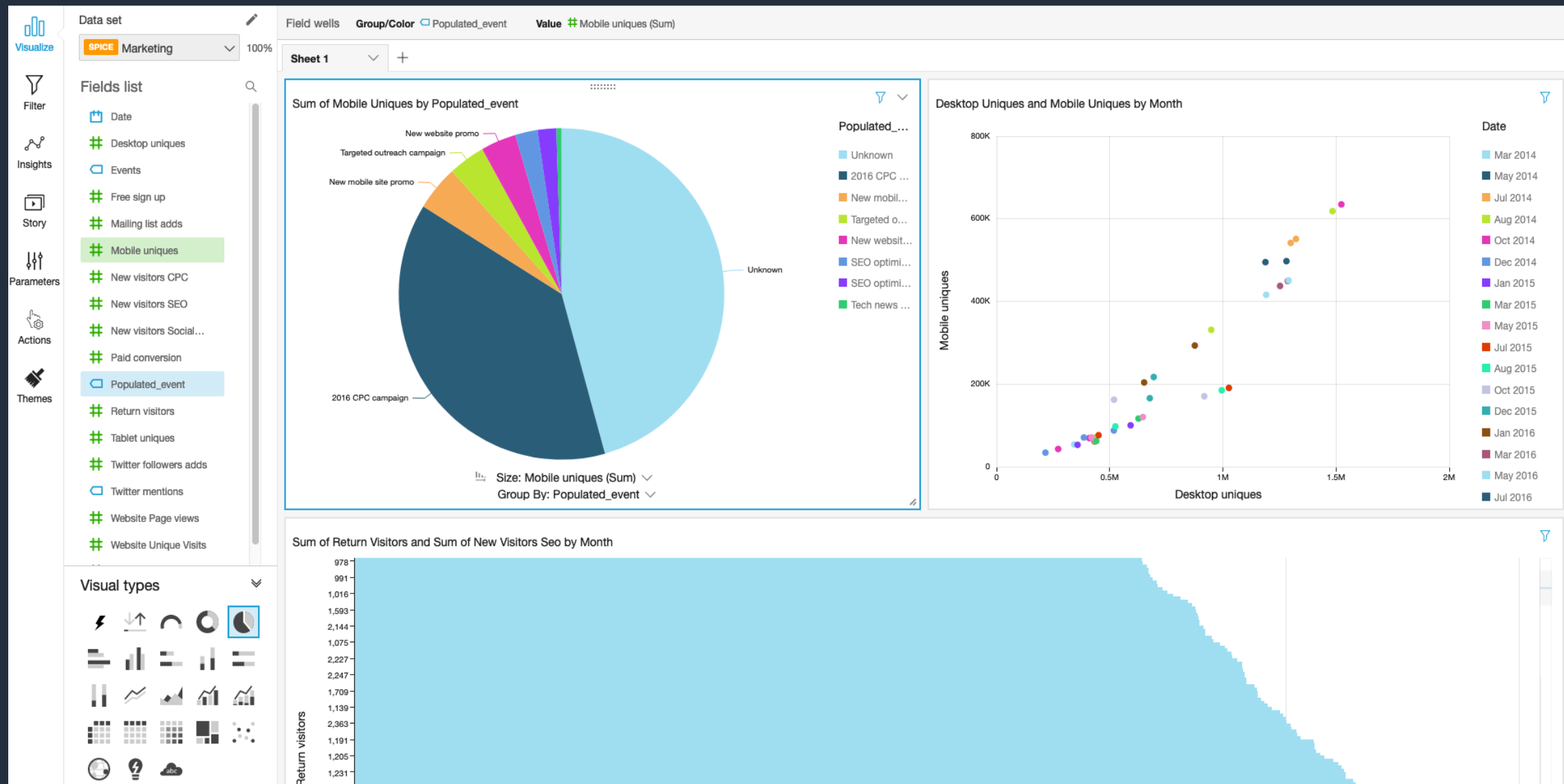
```
val movieLensHomeDir = "s3://emr.examples/movieLens/"

val movies = sc.textFile(movieLensHomeDir + "movies.dat").map { line =>
  val fields = line.split("::")
  // format: (movieId, movieName)
  (fields(0).toInt, fields(1))
}.collect.toMap

val ratings = sc.textFile(movieLensHomeDir + "ratings.dat").map { line =>
  val fields = line.split("::")
  // format: (timestamp % 10, Rating(userId, movieId, rating))
  (fields(3).toLong % 10, Rating(fields(0).toInt, fields(1).toInt, fields(2).toDouble))
}
```



Create Visualizations with Amazon QuickSight



Train ML Models with Amazon SageMaker

Amazon SageMaker

Dashboard

Search

▼ Ground Truth

Labeling jobs

Labeling datasets

Labeling workforces

▼ Notebook

Notebook instances

Lifecycle configurations

Git repositories

▼ Training

Algorithms

Training jobs

Hyperparameter tuning jobs

▼ Inference

Compilation jobs

Model packages

Models

Endpoint configurations

Endpoints

Batch transform jobs

▼ Augmented AI

Human review workflows

Worker task templates

Human review workforces

▲ AWS Marketplace

Amazon SageMaker > Notebook instances > Create notebook instance

Create notebook instance

Amazon SageMaker provides pre-built fully managed notebook instances that run Jupyter notebooks. The notebook instances include example code for common model training and hosting exercises. [Learn more](#)

Notebook instance settings

Notebook instance name

ML-Notebook

Maximum of 63 alphanumeric characters. Can include hyphens (-), but not spaces. Must be unique within your account in an AWS Region.

Notebook instance type

ml.t2.medium

Elastic Inference [Learn more](#)

ml.eia1.medium

▲ Additional configuration

Permissions and encryption

IAM role

Notebook instances require permissions to call other services including SageMaker and S3. Choose a role or let us create a role with the [AmazonSageMakerFullAccess](#) IAM policy attached.

AmazonSageMaker-ExecutionRole-20200113T115452

Root access - optional

☐ Enable - Give users root access to the notebook

☒ Disable - Don't give users root access to the notebook

Lifecycle configurations always have root access

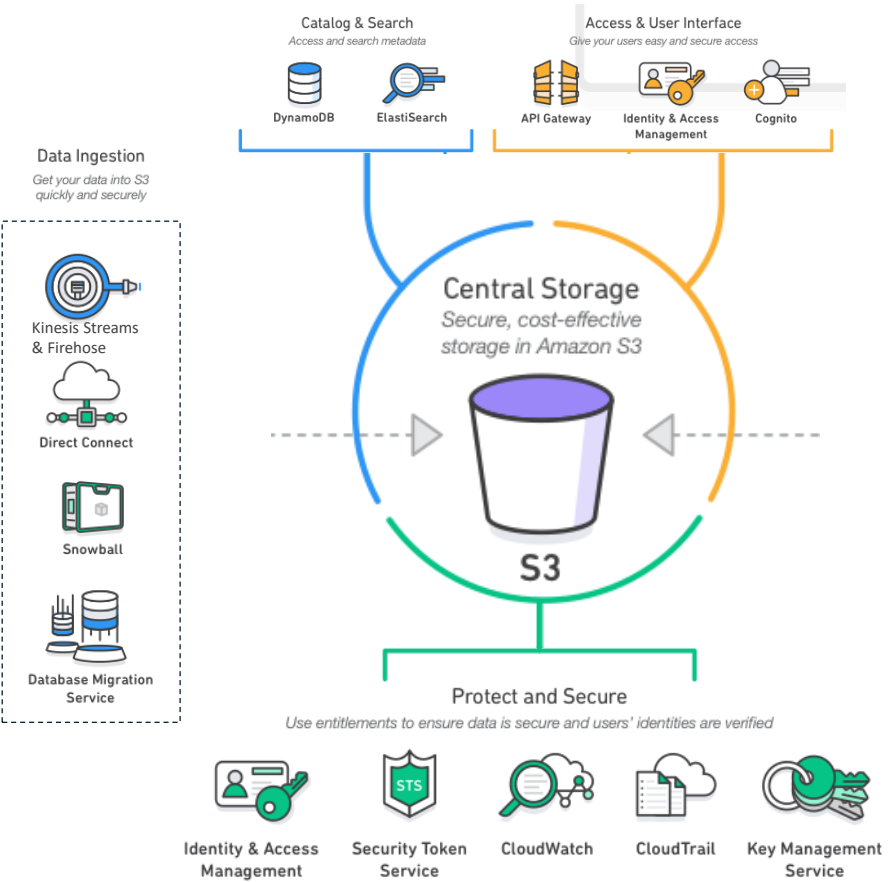
Encryption key - optional

Encrypt your notebook data. Choose an existing KMS key or enter a key's ARN.

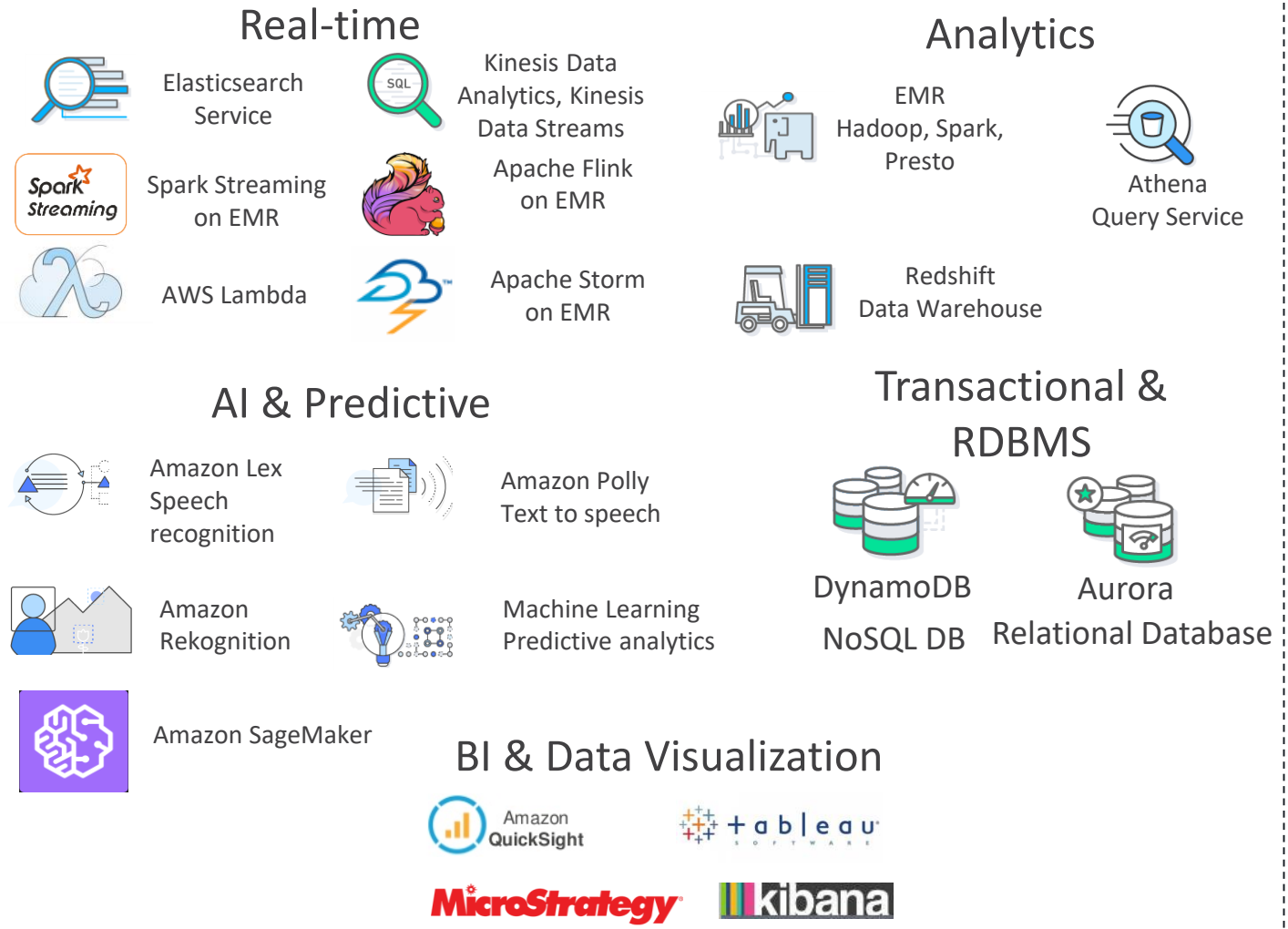
testCMK

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Fully integrate with other AWS Services



Processing & Analytics



More data lakes and analytics than anywhere else

Tens of thousands of data lakes run on AWS across all industries



Complemented by AWS Partner Network (APN) Solutions providers

Collection & preparation



Governance



Visualization



Get help from Data & analytics APN consulting partners

GLOBAL



JAPAN



CHINA



LATAM



NORTH AMERICA



EMEA



APAC



Learn analytics with AWS Training and Certification

Resources created by the experts at AWS to help you build and validate data analytics skills



New free digital course: “Data Analytics Fundamentals”



Classroom offerings, including “Big Data on AWS”, feature AWS expert instructors and hands-on labs



Validate expertise with the “AWS Certified Big Data—Specialty” exam or the new “AWS Certified Data Analytics—Specialty” beta exam

Visit aws.amazon.com/training/paths-specialty/



Customer Success Powered by AWS.



FINRA oversees > 3,000 securities firms doing business in the United States.

Challenge:

FINRA's legacy system did not scale well

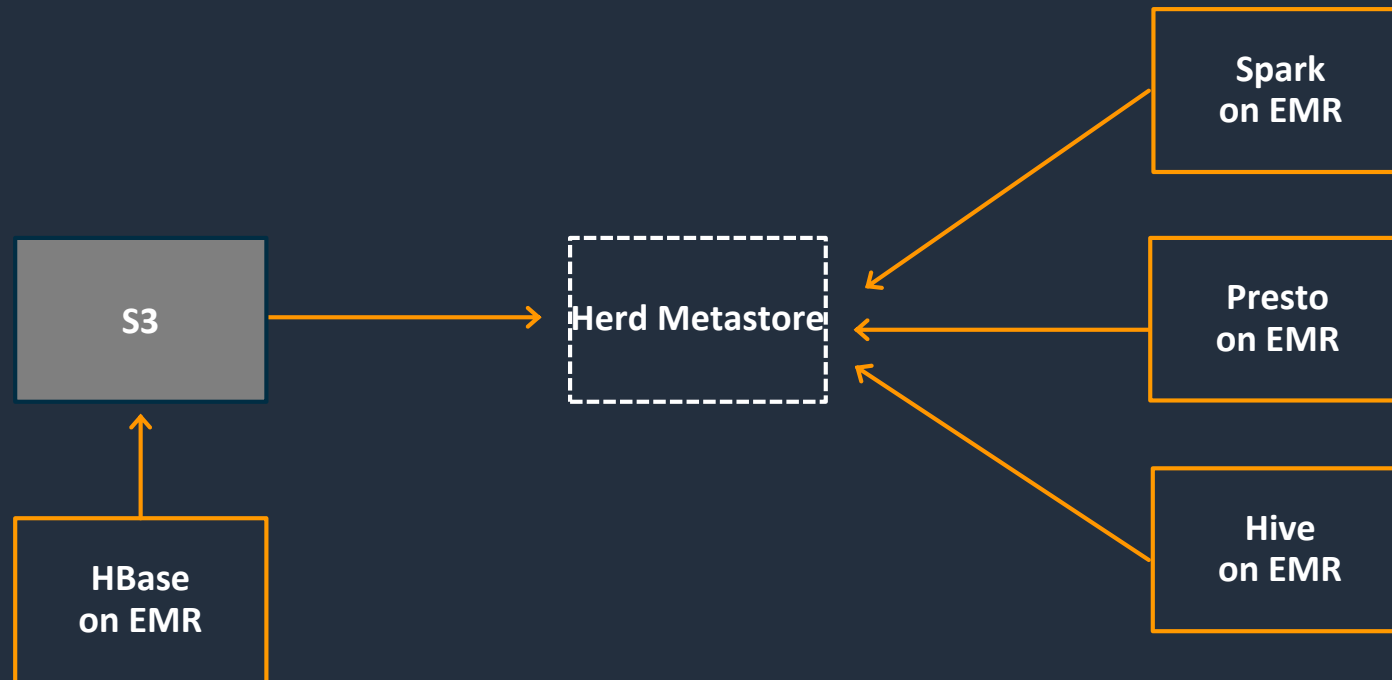
- Up to 75 billion events per day
- Run complex surveillance queries over 20+ PB of data

Solution:

- Migrated their big data appliance to a S3 Data Lake and used EMR for ingestion and processing
- Migrated to RDS and testing Aurora



FINRA uses S3 to Build Data Lake with EMR



- Required fast access across **trillions of trade records (20PB+)**
- Migrated from on-premises system
- Use Apache HBase on Amazon EMR to store and serve this data
- Use EMR engines—Spark, Presto, and Hive to process data
- Lower costs by **60%** over on-premises system



Nasdaq operates financial exchanges around the world, and processes large volumes of data.

Challenge:

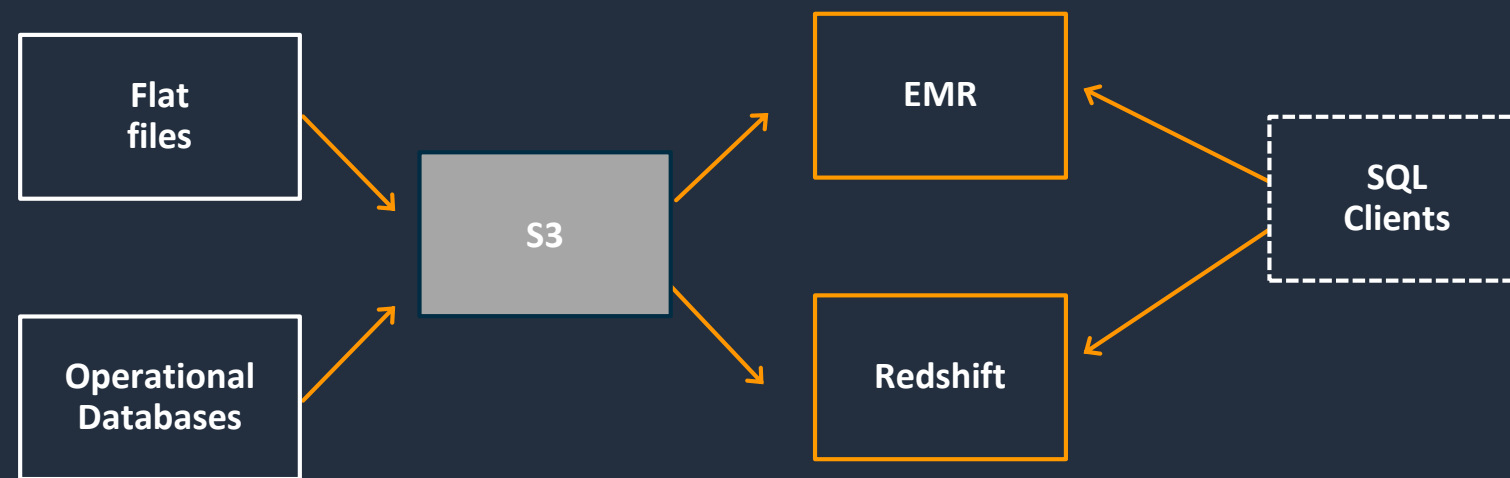
Nasdaq wanted to make their large historical data footprint available to analyze as a single dataset.

Solution:

- Use Amazon Redshift for interactive querying
- Use Amazon S3 as a Data Lake, and Presto on EMR to process historical data



Nasdaq Uses AWS to Build a Data Lake



Data from all 7 exchanges operated by
Nasdaq
(orders, quotes, trade executions)

- Migrate legacy on-premises warehouse to Amazon Redshift
- **4.8B rows** inserted per trading day (orders, trades, quotes)
- Ingest data from multiple sources, validates, and stages in S3
- **Redshift reads data out of S3** for fast queries
- Presto on EMR and S3 used for analysis of massive historical data set



Challenge:

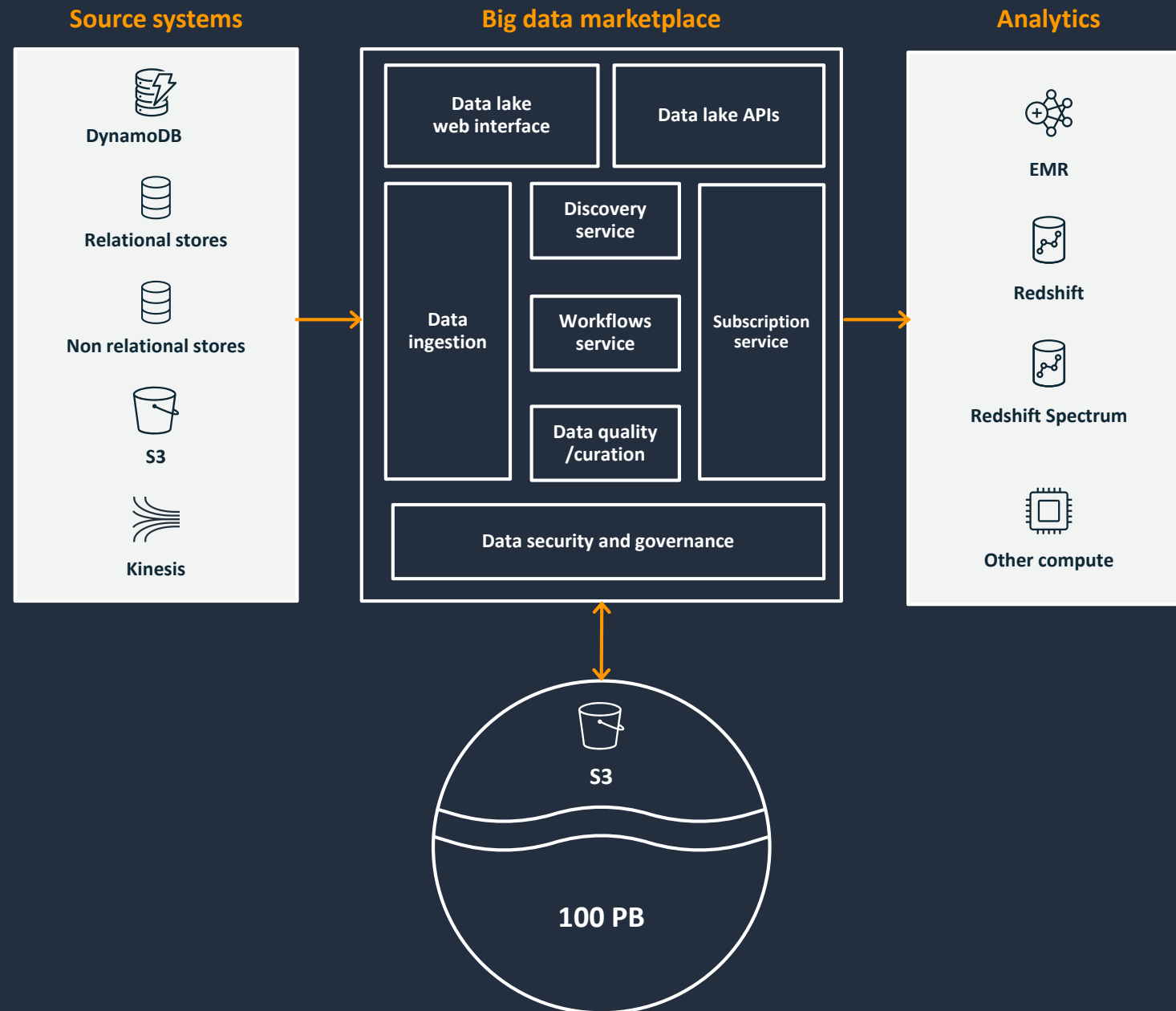
- Amazon needed to analyze a massive amount of data to find insights, identify opportunities, and evaluate business performance.
- The Oracle DW did not scale, was difficult to maintain, and costly.

Solution:

- Amazon deployed a data lake with Amazon S3, and now runs analytics with Amazon Redshift, Redshift Spectrum, and Amazon EMR.



amazon.com Amazon uses an AWS Data Lake



- Migrated system with 50 PB of analytics data
- Optimized scale and cost
- Eliminated Oracle **licensing cost**
- Expanded analytics toolset
- Doubled dataset used in analytics from **50PB to 100PB**

EQUINOX

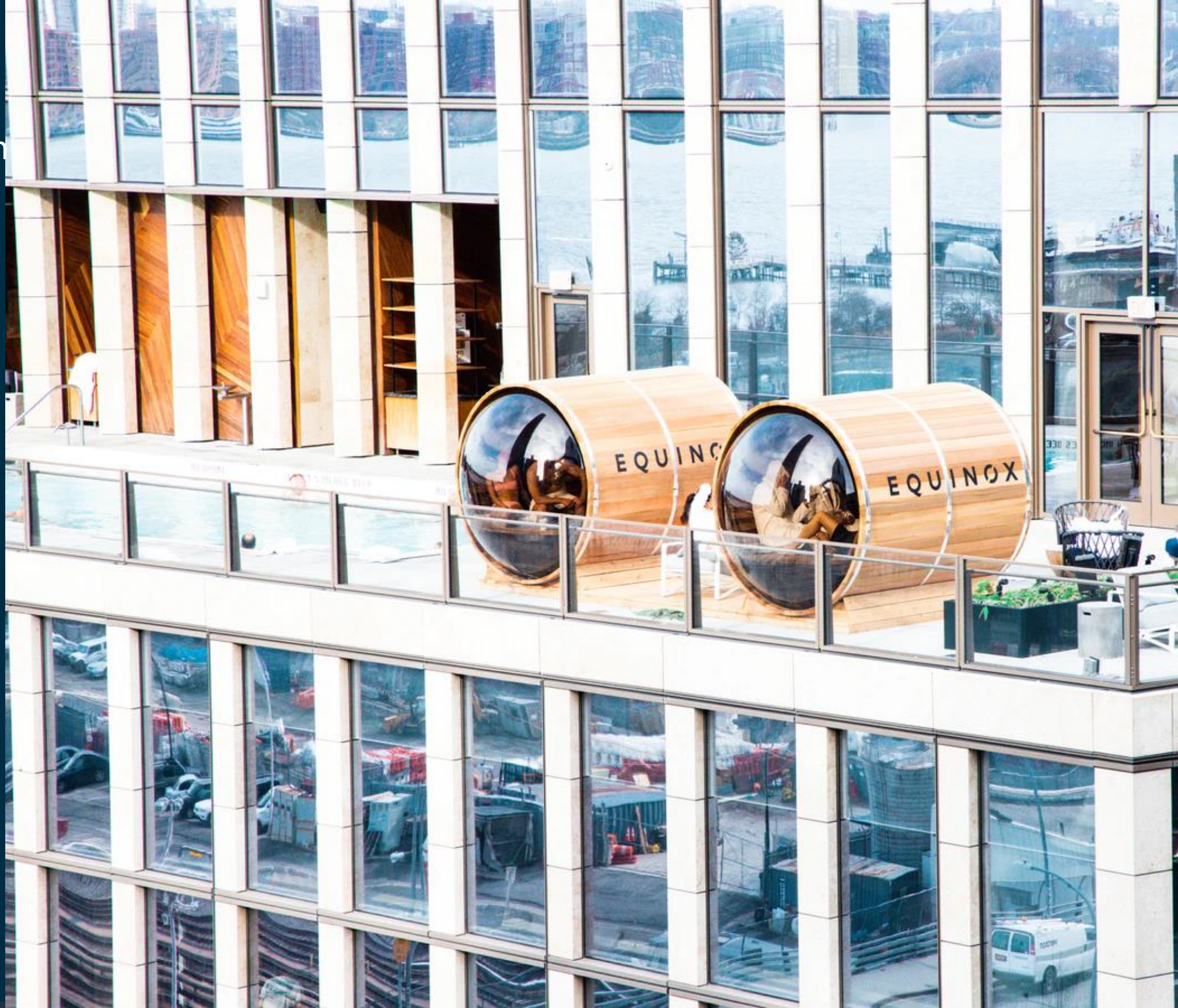
Equinox Fitness has a number of health and fitness brands

Challenge:

- Their data warehouse had limited integration.
- They needed to reduce administration and costs, blend structured and semi-structured data for analytics, and evolve into a data lake strategy.

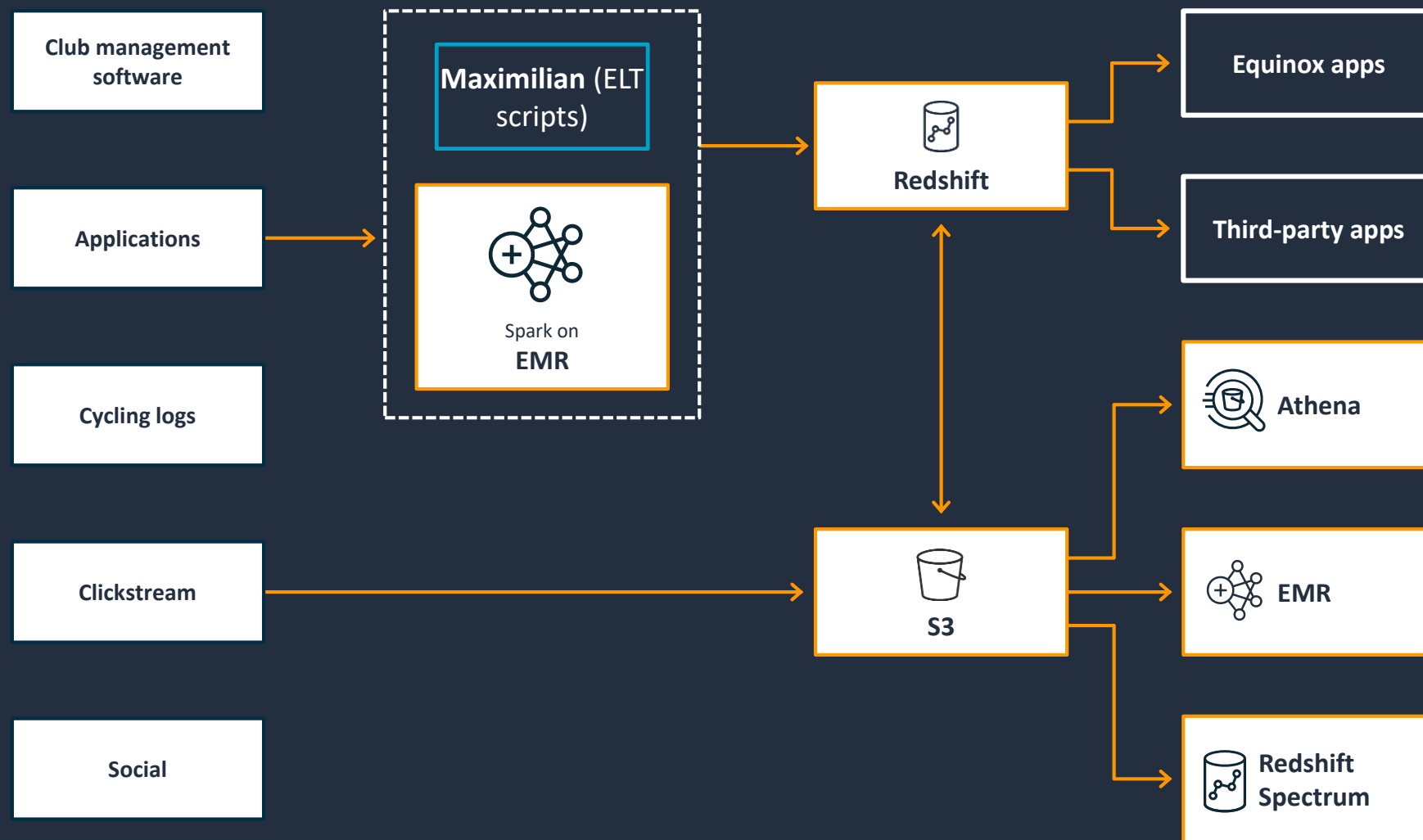
Solution:

- Migrated to Amazon Redshift to combine data from disparate sources.
- They land data directly in an Amazon S3 data lake and perform analytics using Amazon Redshift, Redshift Spectrum, and Amazon EMR.



80% cost savings by migrating to Amazon Redshift

EQUINOX



- Migrated from Teradata data warehouse
- Built a DW with Redshift and data lake with S3
- Analytics on data lake with Amazon Athena, Amazon Redshift Spectrum, and Amazon EMR
- Increased user productivity to move faster
- Amazon Redshift **costs ~20%** of its original Teradata maintenance and support
- Report time reduced **from months to days**

Next steps...

1

Sign up for an AWS account

Instantly get access
to the AWS Free Tier

2

Learn with 10-minute tutorials

Explore and learn with
simple tutorials

3

Start building with AWS

Begin building with step-by-step
guide to help you launch
your AWS project

AWS Data Exchange

Easily find and subscribe to 3rd-party data in the cloud

Quickly find diverse data in one place



>1,000 data products

>80 data providers including include Dow Jones, Change Healthcare, Foursquare, Dun & Bradstreet, Thomson Reuters, Pitney Bowes, Lexis Nexis, and Deloitte

Easily analyze data



Download or copy data to S3

Combine, analyze, and model with existing data

Analyze data with EMR, Redshift, Athena, and AWS Glue

Efficiently access 3rd party data



Simplifies access to data: No need to receive physical media, manage FTP credentials, or integrate with different APIs

Minimize legal reviews and negotiations

Feedback Survey

Please let us know what you think about the session.

Get help from AWS with your analytics project.



Scan QR code to
complete survey

As a token of appreciation, we'll mail you an
AWS souvenir

Thank you!

iosemeke@amazon.com