

AWS

S U M M I T

# Building Your Data Lake on AWS

Ben Snively, Senior Solutions Architect, AWS

August 14, 2017



# What is a Data Lake?

An **architectural approach** that allows you to store massive amounts of “**raw**” data into a central location

It's **readily available** to be categorized, processed, analyzed, and consumed by **diverse groups**

# Why use a Data Lake?

Leverage **all data** within  
your organization



# Legacy data architectures exist as isolated data silos



Hadoop  
cluster



Data  
warehouse  
appliance



SQL  
database

# Navigating the Data Lake...

Data Lake is a new and increasingly popular architecture to store and analyze massive volumes and heterogenous types of data in a centralized repository



# Building a Data Lake on AWS

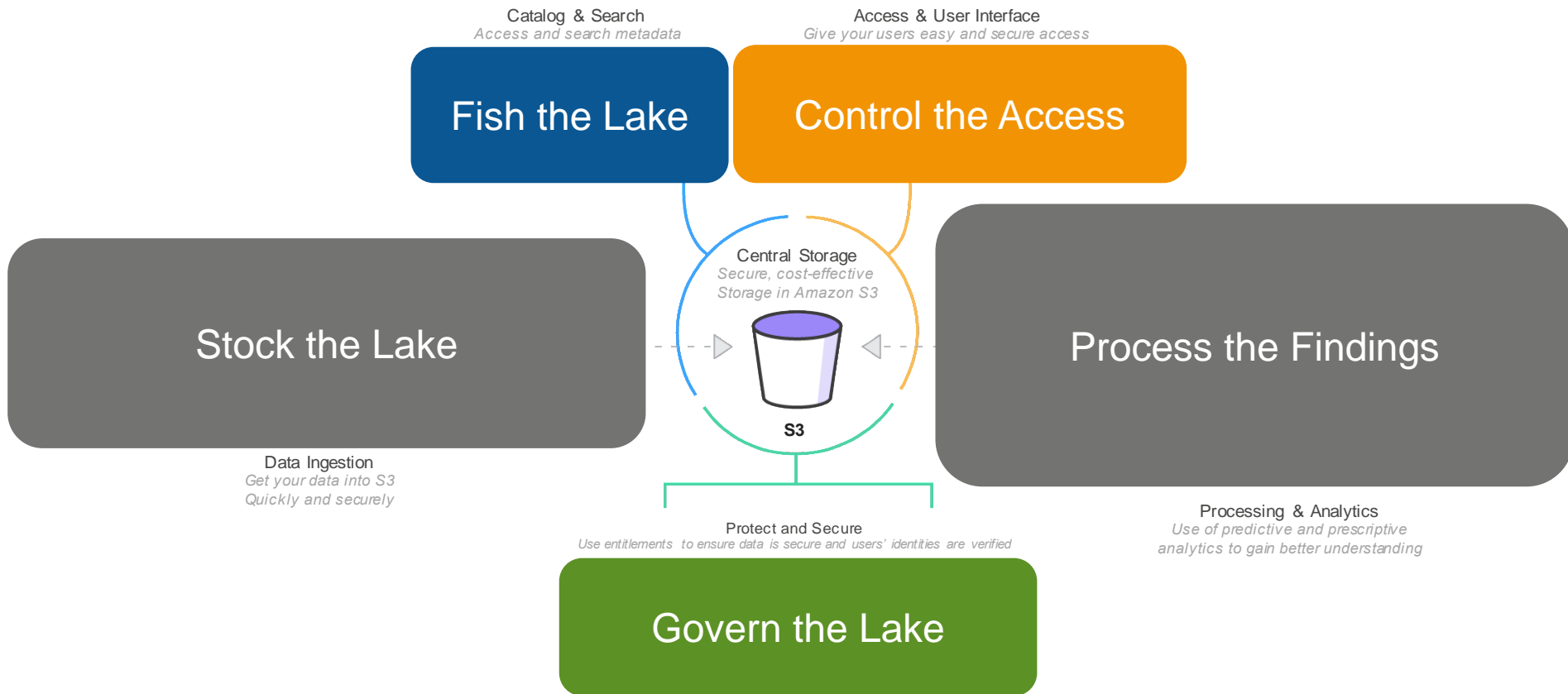
# Why AWS?

*Implementing a Data Lake architecture requires a broad set of tools and technologies to serve an increasingly diverse set of applications and use cases.*

Boils down to:

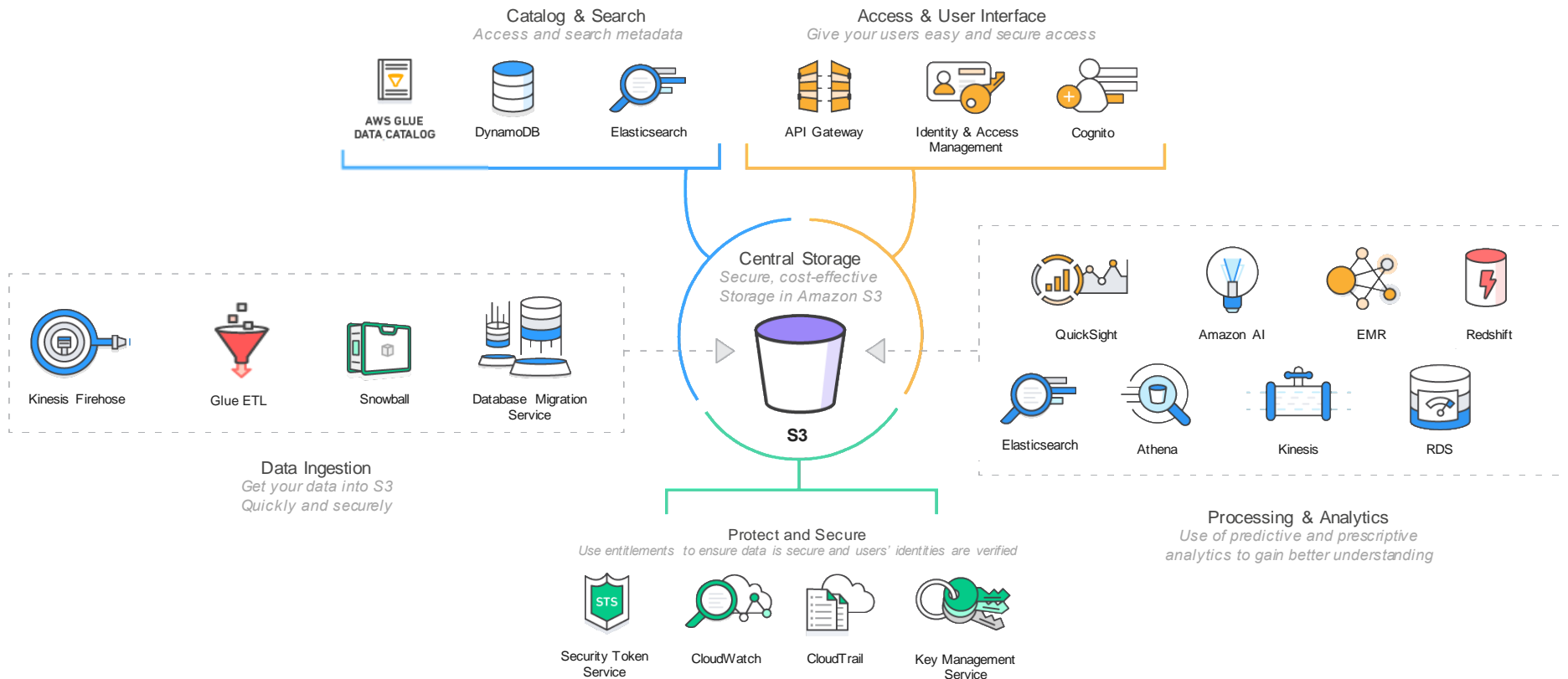
Picking the right tool for the right job...on a consumption-based model...

# Data Lake reference architecture





# Data Lake reference architecture



# S3 – Center of the Data Lake

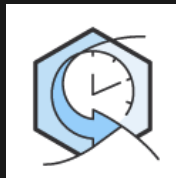


# Why Amazon S3 for Data Lake?



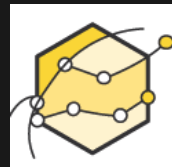
## Durable

Designed for 11 9s of durability



## Available

Designed for 99.99% availability



## High performance

- Multiple Upload
- Range GET



## Easy to use

- Simple REST API
- AWS SDKs
- Read-after-create consistency
- Event notification
- Lifecycle policies



## Scalable

- Store as much as you need
- Scale storage and compute independently
- No minimum usage commitments



## Integrated

- Amazon EMR
- Amazon Redshift
- Amazon DynamoDB

# Stock the Lake – Data Ingestion



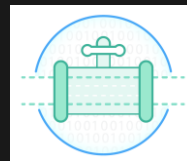
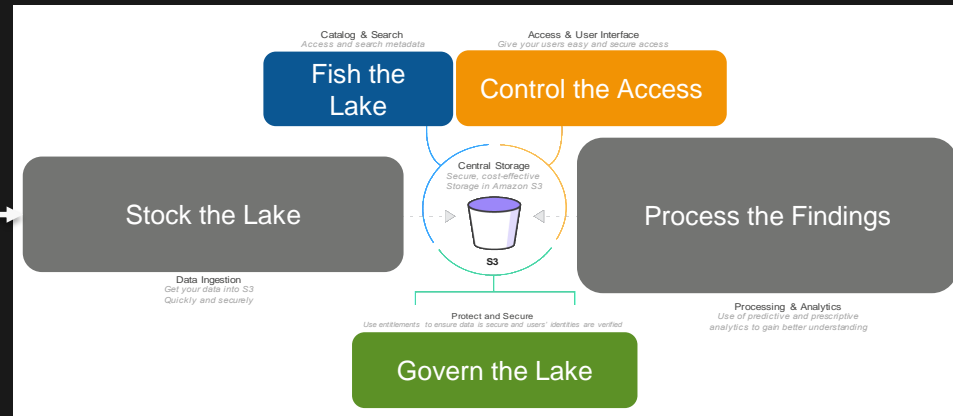
## Amazon Glue ETL

- Serverless ETL engine generates Python code that is entirely customizable, reusable, and portable.



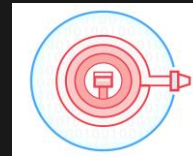
## Database Migration Service

Migrate your RDBMS into S3  
(as well as other targets)



## Amazon Kinesis Streams

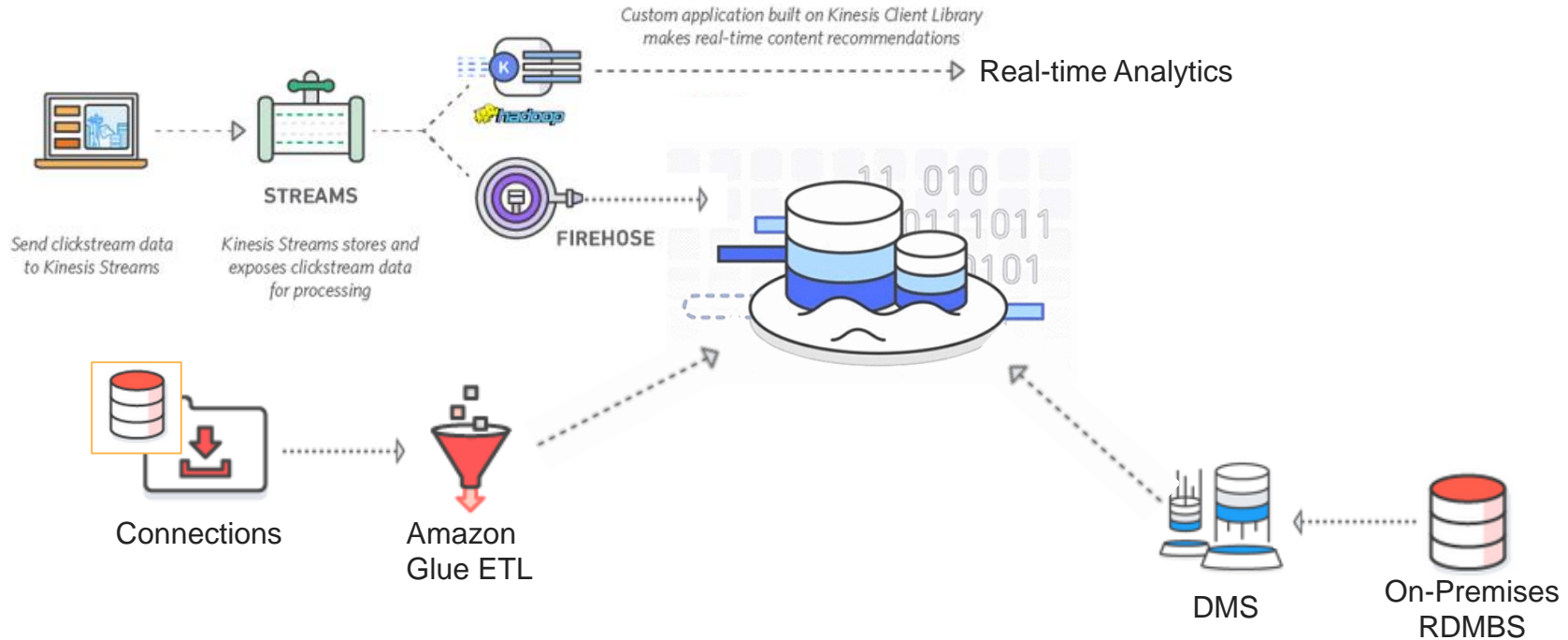
- Build your own custom applications that process or analyze streaming data



## Amazon Kinesis Firehose

- Easily load massive volumes of streaming data into S3, Amazon Redshift, and Amazon Elasticsearch Service

# Stock the Lake – Data Ingestion



# Demonstration

**What is the speed of the data?**

**What is the source of the data?**

# Fishing the Lake – Catalog/Search



Glue Data  
Catalog

Structural and  
Operational  
metadata

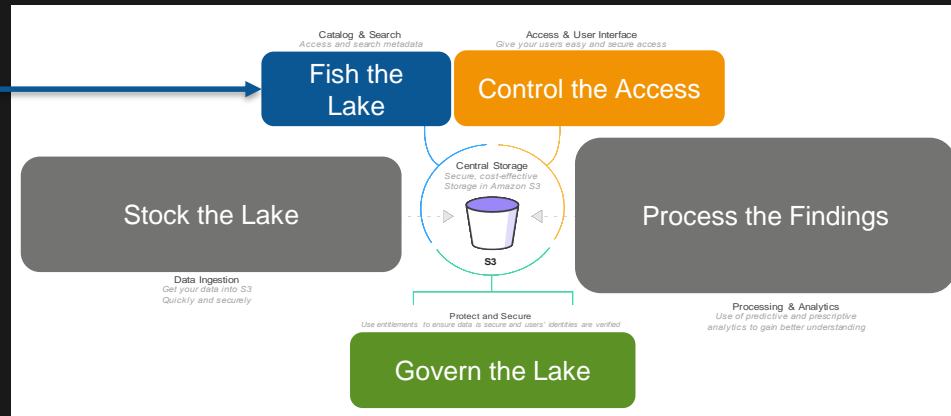


Amazon  
DynamoDB

Metadata/Tag  
Lookup



Search Content



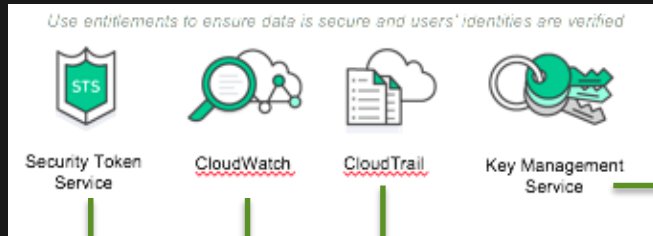
# Demonstration

**What type of data?**

**How is the data being queried?**



# Govern the Lake

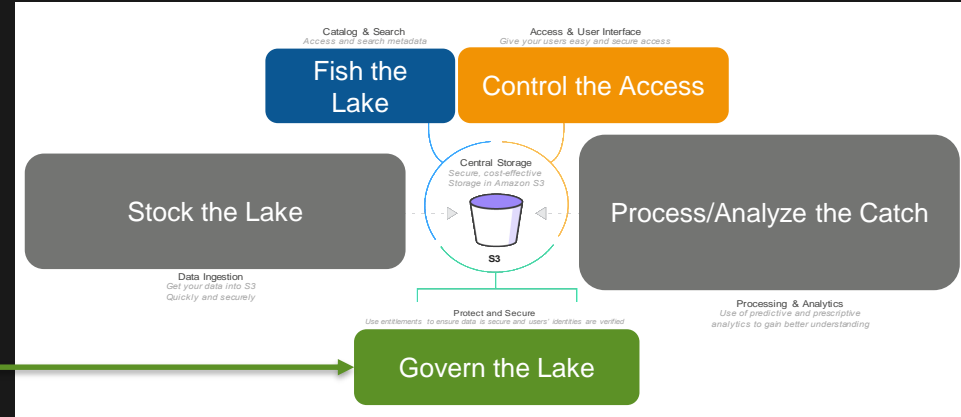


Temporary  
Tokens

Performance

Auditing

Encryption



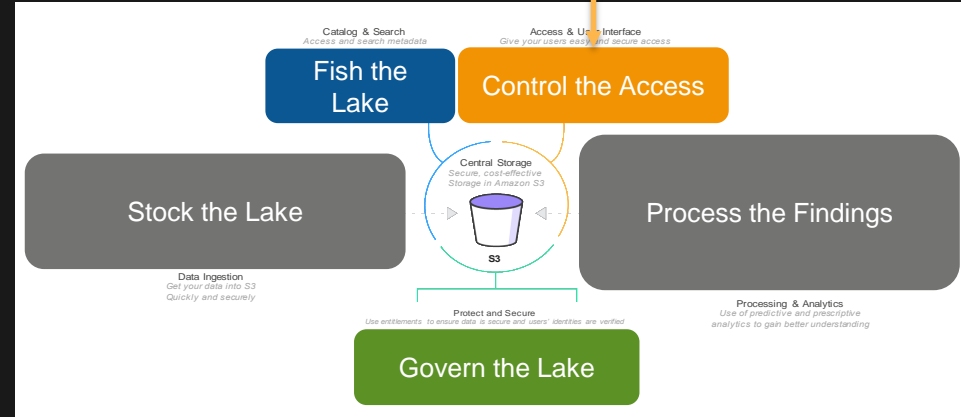
# Control the Access



Interfaces

Identity/Access

User Authentication



# Process/Analyze the Catch

## Processing & Analytics

### Real-time



ElasticSearch Service



Kinesis Analytics,  
Kinesis Streams



Spark Streaming on EMR



AWS Lambda



Apache Flink on EMR



Apache Storm on EMR

### Batch



EMR Hadoop, Spark,  
Presto



Redshift Data Warehouse



Athena Query Service

### AI & Predictive



Amazon Lex



Amazon Polly



Amazon Rekognition



Machine Learning

### Transactional & RDBMS

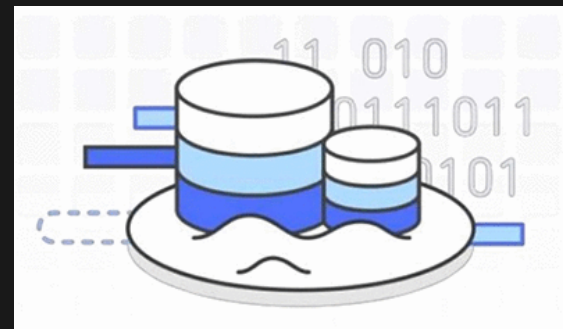
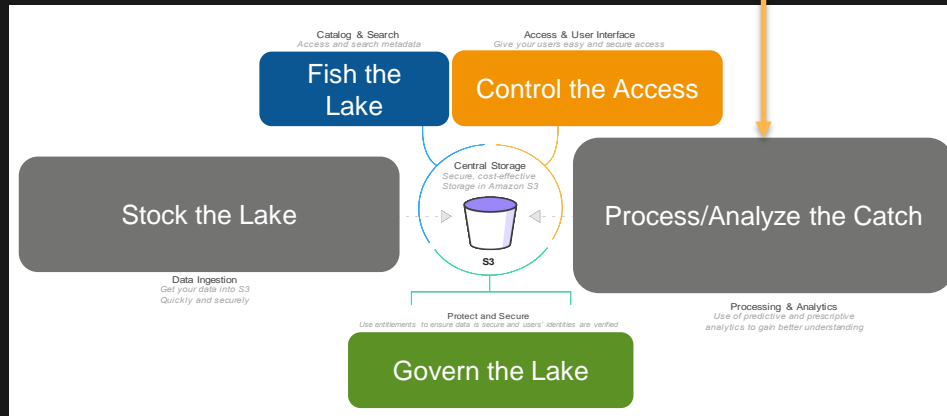


DynamoDB, NoSQL DB

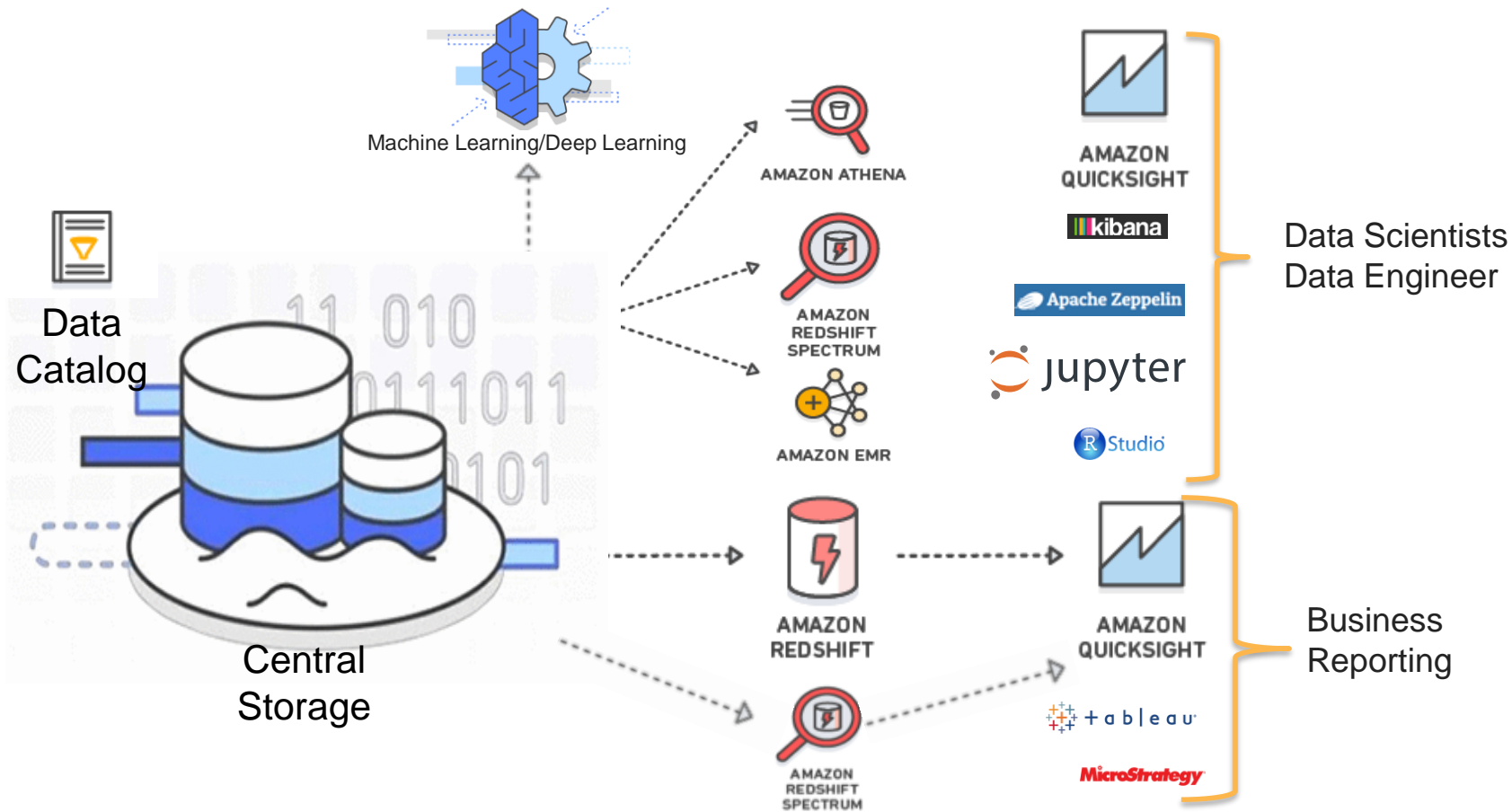


Aurora Relational Database

### BI & Data Visualization



# Process/Analyze the Catch



# Interactive query service



Amazon  
Athena

- **Query directly from Amazon S3**
- **Use ANSI SQL**
- **Serverless**
- **Multiple data formats**
- **Pay per query**



Amazon  
Elastic  
MapReduce

**Hadoop/HDFS clusters**

**Hive, Pig, Impala, Hbase, Spark, Presto**

**Easy to use, fully managed**

**On-demand, reserved instance, and**

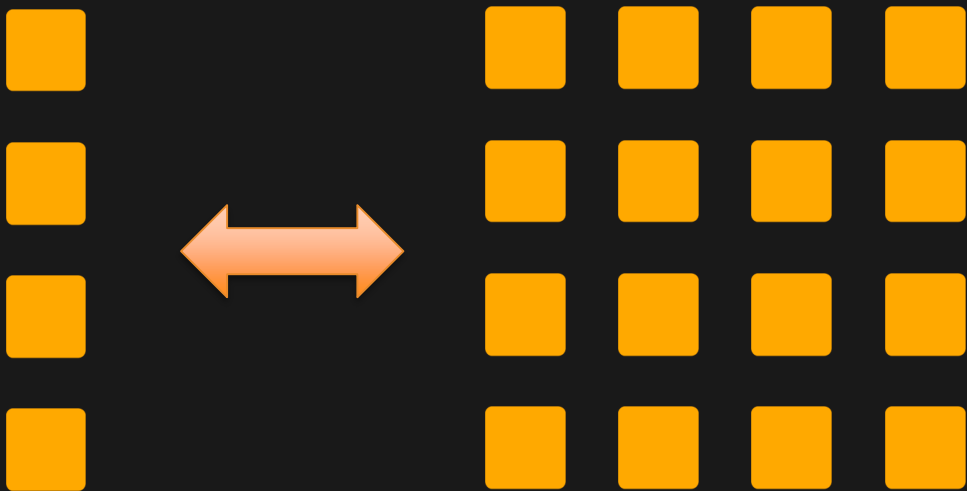
**Spot pricing**

**Tight integration with Amazon S3,**

**DynamoDB, and Kinesis**

# Resizable clusters

Easy to add and remove compute capacity on your cluster.



# ON A SINGLE MACHINE



COST:  $4\text{h} \times \$1.06 = \$4.24$

PROCESSING TIME: 4h



# ON MULTIPLE MACHINES



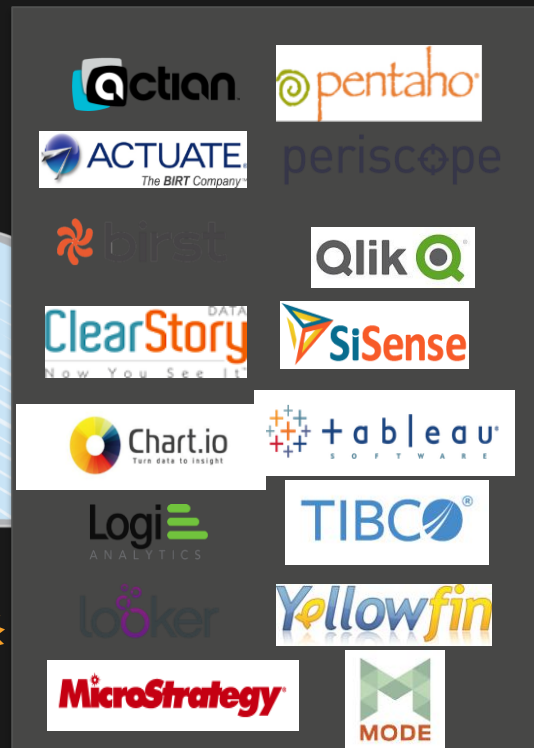
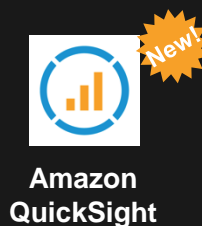
COST:  $4 \times 1\text{h} \times \$1.06 = \$4.24$

PROCESSING TIME: 1h

# Amazon Redshift works with third-party analysis tools



JDBC/ODBC



# Amazon Redshift has security built in

SSL to secure data in transit

Encryption to secure data at rest

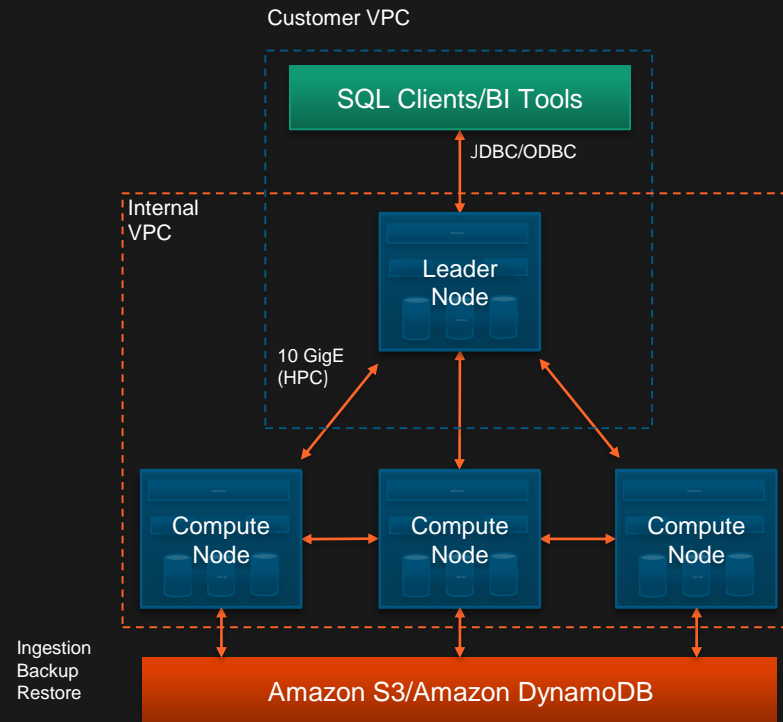
- AES-256; hardware accelerated
- All blocks on disks and in Amazon S3 encrypted
- HSM support

No direct access to compute nodes

Audit logging, AWS CloudTrail, AWS KMS integration

Amazon VPC support

SOC 1/2/3, PCI-DSS Level 1, FedRAMP, HIPAA



# Redshift Spectrum

Leverages Amazon Redshift's advanced cost-based optimizer

Pushes down projections, filters, aggregations and join reduction

Dynamic partition pruning to minimize data processed

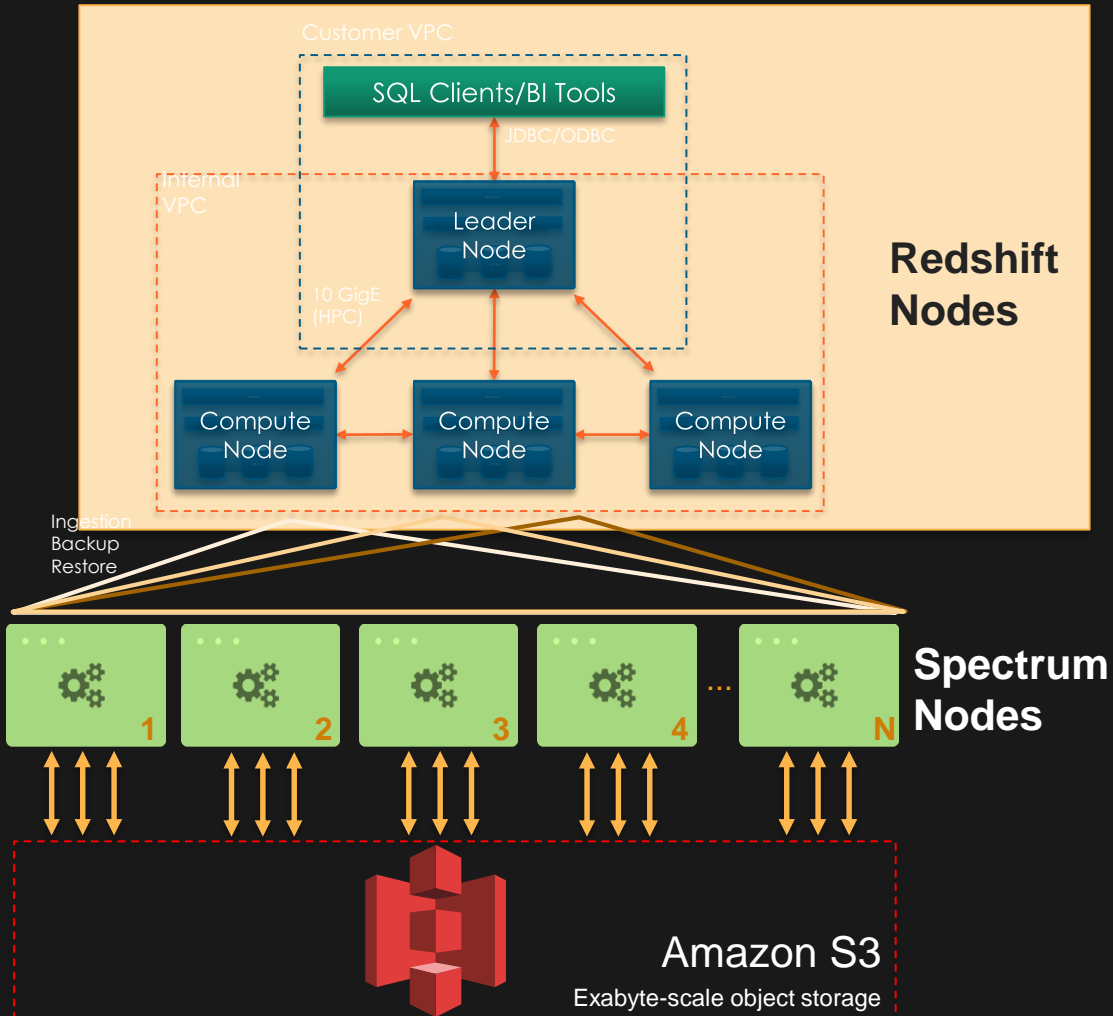
Automatic parallelization of query execution against Amazon S3 data

Efficient join processing within the Amazon Redshift cluster



**Data Catalog**

Apache Hive Metastore





# Amazon AI

Intelligent services powered by deep learning

## AI SERVICES



AMAZON  
REKOGNITION  
IMAGE RECOGNITION



AMAZON  
POLLY  
TEXT-TO-SPEECH



AMAZON LEX  
VOICE AND TEXT  
CHATBOTS

## AI PLATFORMS



AMAZON  
MACHINE  
LEARNING



AMAZON  
EMR



SPARK &  
SPARKML

## AI FRAMEWORKS

AWS DEEP LEARNING AMI

APACHE  
MXNET

TENSOR-  
FLOW

CAFFE

TORCH

THEANO

CNTK

KERAS

## AI INFRASTRUCTURE

AMAZON EC2 P2  
AND G2 GPU

AMAZON EC2  
CPUS

AWS LAMBDA

ENHANCED  
NETWORKING

AWS IOT AND  
AWS GREENGRASS

# Proven customer success

The vast majority of big data use cases deployed in the cloud today run on AWS.



HEARST



NETFLIX



REDFIN



SIEMENS





## Case study: Re-architecting compliance

*“For our market surveillance systems, we are looking at about 40% [savings with AWS], but the real benefits are the business benefits: We can do things that we physically weren’t able to do before, and that is priceless.”*

- Steve Randich, CIO



### What FINRA needed

- Infrastructure for its market surveillance platform
- Support of analysis and storage of approximately 75 billion market events every day

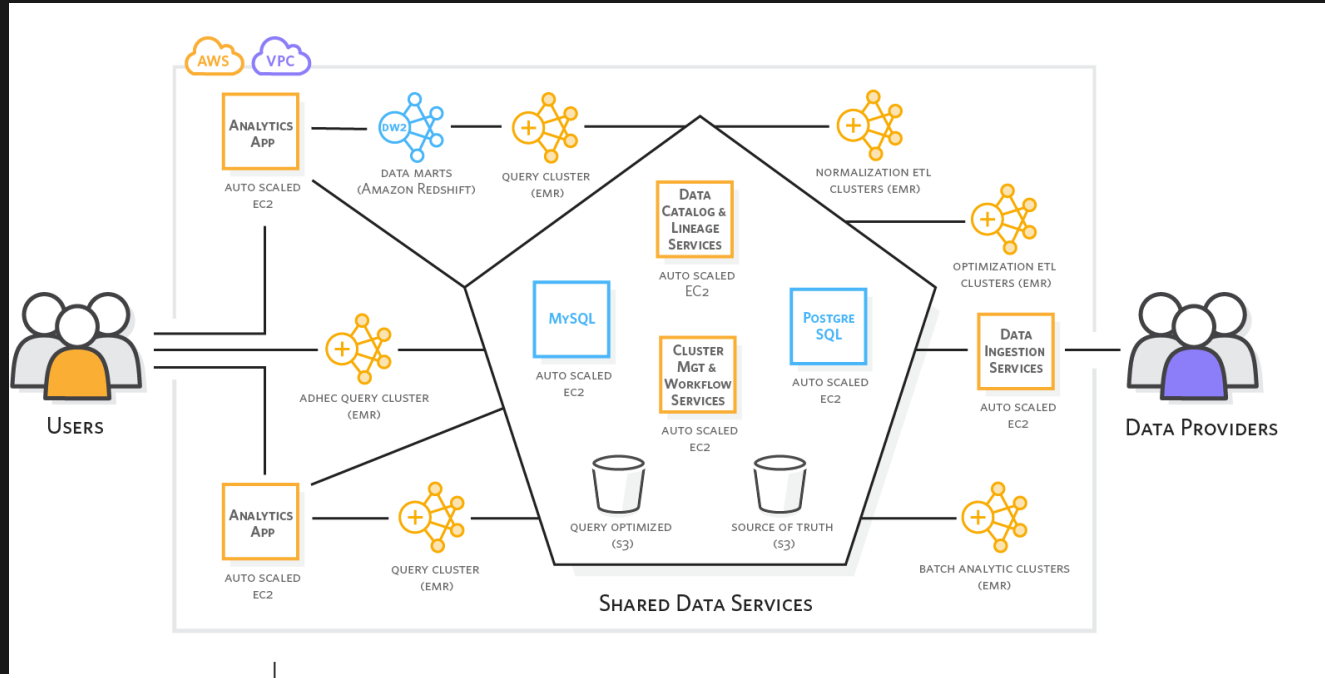
### Why they chose AWS

- Fulfillment of FINRA’s security requirements
- Ability to create a flexible platform using dynamic clusters (Hadoop, Hive, and HBase), Amazon EMR, and Amazon S3

### Benefits realized

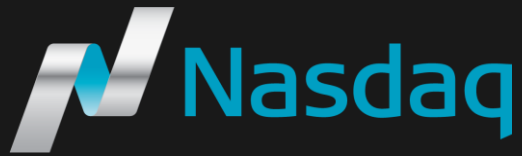
- Increased agility, speed, and cost savings
- Estimated savings of \$10-20M annually by using AWS

# Fraud detection



FINRA uses Amazon EMR and Amazon S3 to process up to 75 billion trading events per day and securely store over 5 petabytes of data, attaining savings of \$10-20M per year.





- Nasdaq implements an Amazon S3 data lake + Amazon Redshift data warehouse architecture
- Most recent two years of data is kept in the Amazon Redshift data warehouse and snapshotted into Amazon S3 for disaster recovery
- Data between two and five years old is kept in Amazon S3
- Presto on Amazon EMR is used to ad-hoc query data in Amazon S3
- Transitioned from an on-premises data warehouse to Amazon Redshift & Amazon S3 data lake architecture
- Over 1,000 tables migrated
- Average daily ingest of over 7B rows
- ***Migrated*** off legacy DW to AWS (start to finish) ***in 7 man-months***
- AWS costs were **43%** of legacy budget for the same data set (~1100 tables)

# Building a Data Lake

An **architectural approach** that allows you to store massive amounts of “**raw**” data into a central location

It's **readily available** to be categorized, processed, analyzed, and consumed by diverse groups



AWS

S U M M I T

Thank you!

