AWS INNOVATE



Databases and Analytics on the AWS Cloud

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What to expect from the session



- Challenges and architectural principles
- How to simplify big data processing
- What database technologies should you use?
 - Why and When?
 - How?
- Architectural patterns and Customer Examples



Key Ideas



Data is your organization's most valuable resource

All data has the potential to be big data

Databases are no longer the center of analytics*

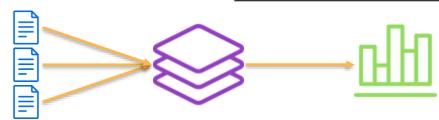
*but they play a critical role!



Evolution of Analytics

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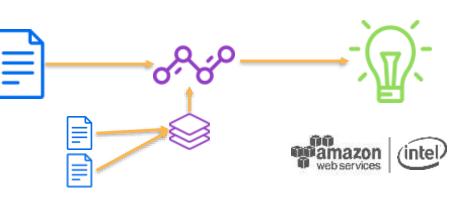
Batch analytics



Real-time analytics



Predictive/Adaptive analytics



A plethora of tools













EMR



S3



DynamoDB



SQS

cascading











Amazon

Glacier



RDS



ElastiCache













Amazon Kinesis

Streams app













Amazon Kinesis

Lambda











Amazon Kinesis Streams **Analytics**







Apache Zeppelin











Challenges



• Is there a reference architecture?

What tools should I use?

• How?

• Why?



Architectural Principles



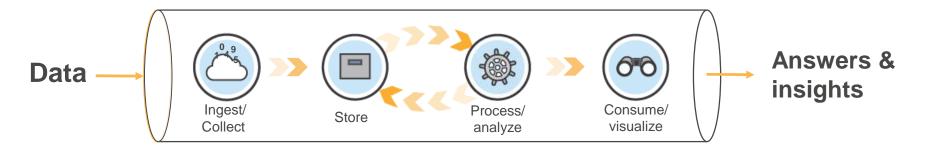
Build decoupled systems

- Data → Store → Process → Store → Analyze → Answers
- Use the right tool for the job
- Data structure, latency, throughput, access patterns
 Leverage AWS managed services
- Scalable/elastic, available, reliable, secure, no/low admin
 Use log-centric design patterns
- Immutable logs, materialized views (schema-on-read)
- Be cost-conscious
- Big data ≠ big cost



Simplify Big Data Processing





Time to answer (Latency)

Throughput

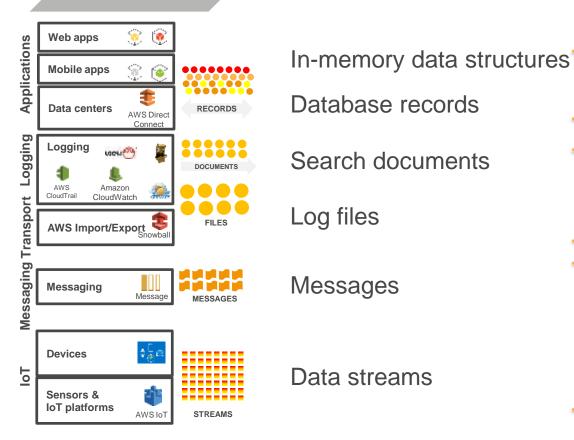
Cost



Types of Data

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Transactions

Files

Events



Data Temperature

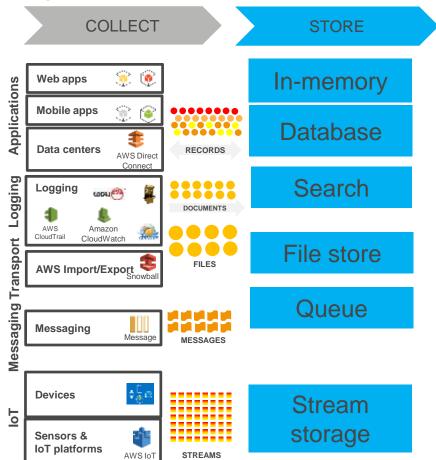


	Hot	Warm	Cold
Volume	MB-GB	GB-TB	PB-EB
Item size	B–KB	KB-MB	KB-TB
Latency	ms	ms, sec	min, hrs
Durability	Low-high	High	Very high
Request rate	Very high	High	Low
Cost/GB	\$\$-\$	\$-¢¢	¢
	Hot data	Warm data	Cold data



Types of Data Stores





Caches, data structure servers

SQL & NoSQL databases

Search engines

File systems

Message queues

Pub/sub message queues



Message & Stream Storage

Streams



COLLECT



Amazon SQS

Managed message queue service

Apache Kafka

 High throughput distributed streaming platform

Amazon Kinesis Streams

Managed stream storage + processing

Amazon Kinesis Firehose

Managed data delivery

Amazon DynamoDB

- Managed NoSQL database
- Tables can be stream-enabled



Why Stream Storage?

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Decouple producers & consumers

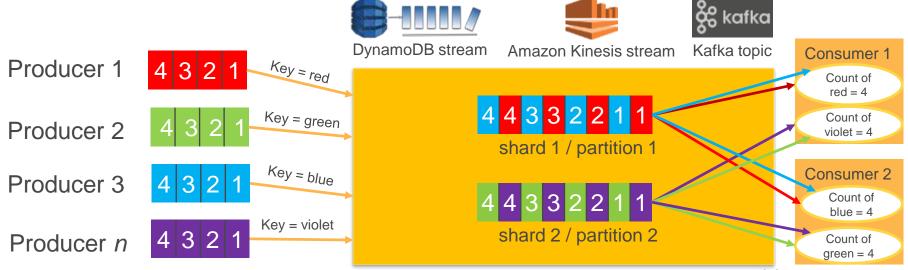
Persistent buffer

Collect multiple streams

Preserve client ordering

Parallel consumption

Streaming MapReduce





File Storage **AWS INN@VATE** COLLECT **STORE** • Web apps Applications In-memory Mobile apps . Database **Data centers** RECORDS AWS Direct Connect Logging Logging Search DOCUMENTS AWS Amazon Amazon S3 **Transport** CloudTrail CloudWatch Amazon S3 AWS Import/Export 🐺 FILES Messaging **Amazon SQS** Messaging Message MESSAGES % Apache Kafka **Devices Amazon Kinesis** æ **Streams** Sensors & **Amazon Kinesis** IoT platforms **Firehose** STREAMS AWS IoT

Amazon DynamoDB

Streams



Why Is Amazon S3 Good for Analytics? AWS INN⊕VATE

- Natively supported by big data frameworks (Spark, Hive, Presto, etc.)
- No need to run compute clusters for storage (unlike HDFS)
- Can run transient Hadoop clusters & Amazon EC2 Spot Instances
- Multiple & heterogeneous analysis clusters can use the same data
- Unlimited number of objects and volume of data
- Very high bandwidth no aggregate throughput limit
- Designed for 99.99% availability can tolerate zone failure
- Designed for 99.99999999% durability
- No need to pay for data replication
- Native support for versioning
- Tiered-storage (Standard, IA, Amazon Glacier) via life-cycle policies
- Secure SSL, client/server-side encryption at rest
- Low cost



File Storage COLLECT **STORE** • Web apps Applications In-memory Mobile apps . Database **Data centers** RECORDS AWS Direct Connect Logging Logging Search DOCUMENTS AWS Amazon **Transport** CloudTrail CloudWatch Amazon S3 AWS Import/Export 🐺 FILES Messaging **Amazon SQS** Messaging Message MESSAGES % Apache Kafka **Devices Amazon Kinesis** æ **Streams** Sensors & **Amazon Kinesis** St -----IoT platforms **Firehose** STREAMS AWS IoT

Amazon DynamoDB

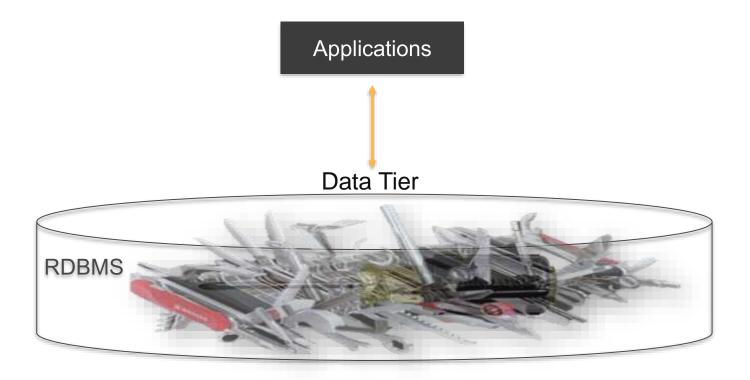
Streams





Database Access Anti-Pattern

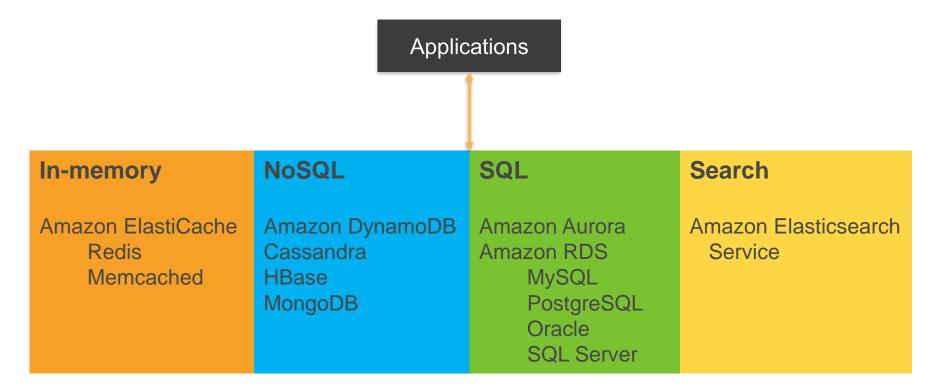






Best Practice: Use the Right Tool for the Job AWS INN@VATE











ElastiCache

Microsecond Real-Time Performance

Fully Managed

Redis Automatic Failover = NoOps

Enhanced Redis Engine

No Cross-AZ Data Transfer Costs

Easy to Deploy, Use and Monitor

Open-Source Compatible







DynamoDB

Fully managed NoSQL

Document / Key-Value store

Single-digit millisecond latency

Massive and seamless scalability

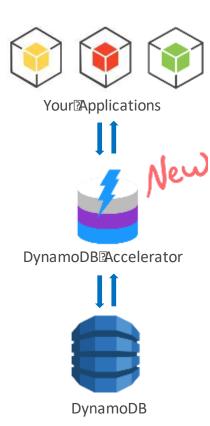
Event-driven programming



DynamoDB Accelerator (DAX) Preview







Features

- Fully managed, highly available: handles all software management, fault tolerant, replication across multi-AZs within a region
- **DynamoDB API compatible:** seamlessly caches DynamoDB API calls, no application re-writes required
- Write-through: DAX handles caching for writes
- **Flexible**: Configure DAX for one table or many
- **Scalable**: scales-out to any workload with up to 10 read replicas
- Manageability: fully integrated AWS service: Amazon CloudWatch, Tagging for DynamoDB, AWS Console
- Security: Amazon VPC, AWS IAM, AWS CloudTrail, AWS **Organizations**





RDS

Automated backups (with point-in-time recovery)

Cross-region snapshot copies

Automated patch management

Automated Multi-AZ replication

Scale up / Scale down instance types

Scalable storage on demand

"License included" and BYOL models









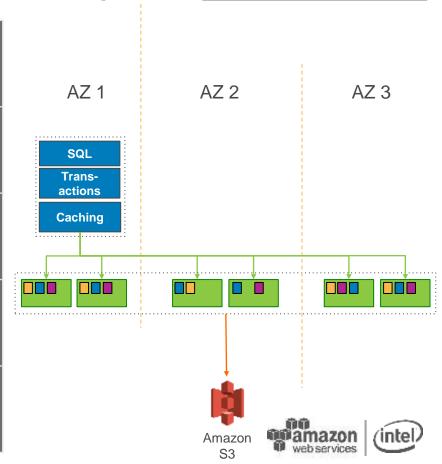




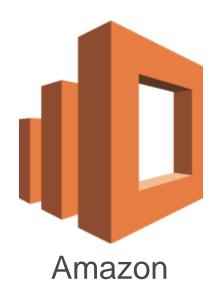


Amazon Aurora: MySQL andPostgreSQL-compatible AWS INN®VATE

- 5x faster than MySQL on same hardware
- SysBench: 100 K writes/sec and 500 K reads/sec
- Designed for 99.99% availability
- 6-way replicated storage across 3 AZs
- Scale to 64 TB and 15 Read Replicas







Elasticsearch

Service

Distributed search and analytics engine

Managed service using Elasticsearch and Kibana

Fully managed – zero admin

Highly available and reliable

Tightly integrated with other AWS services







Which Data Store Should I Use?



Data structure

Fixed schema → SQL, NOSQL

Schema-Free (JSON) → NoSQL, Search

(Key, Value) → In-Memory, NoSQL

Access patterns → Store data in the format you will access it

Put/Get (Key, Value) → In-Memory, NoSQL

Simple relationships (1:N, M:N) → NoSQL

Complex relationships (Multi-table joins, transactional) → SQL

Faceting, Search → Search

Data characteristics → Hot, warm, cold

Cost → Right cost



Analytics Types & Frameworks

PROCESS / ANALYZE

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Batch

Takes minutes to hours

Example: Daily/weekly/monthly reports

Amazon EMR (MapReduce, Hive, Pig, Spark)

Interactive

Takes seconds

Example: Self-service dashboards

Amazon Redshift, Amazon Athena,

Amazon EMR (Presto, Spark)

Message

Takes milliseconds to seconds

Example: Message processing

Amazon SQS applications on Amazon EC2

Stream

Takes milliseconds to seconds

Example: Fraud alerts, 1 minute metrics

Amazon EMR (Spark Streaming, Flink), Amazon Kinesis

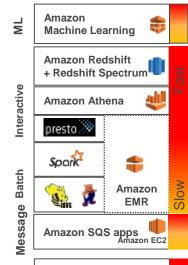
Analytics, KCL, Storm, AWS Lambda

Machine Learning

Takes milliseconds to minutes

Example: Fraud detection, forecast demand

Amazon ML, Amazon EMR (Spark ML)









What About ETL?







PROCESS / ANALYZE

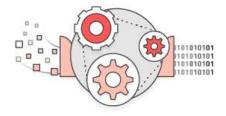
Data Integration Partners

alooma

Reduce the effort to move, cleanse, synchronize, manage, and automatize data related processes.







AWS Glue is a fully managed ETL service that makes it easy to understand your data sources, prepare the data, and move it reliably between data stores



https://aws.amazon.com/big-data/partner-solutions/

Data Consumption

Applications & API

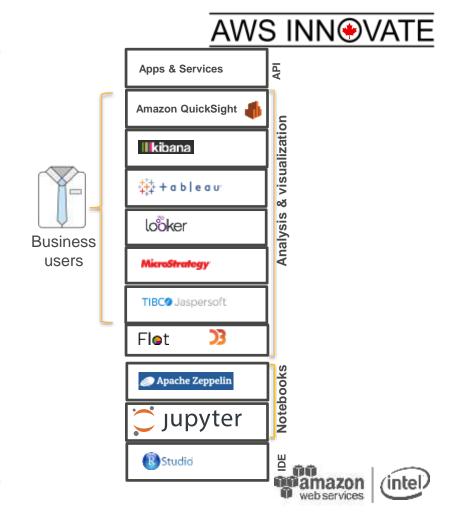
Analysis and visualization

Data scientist,

developers

Notebooks

IDE



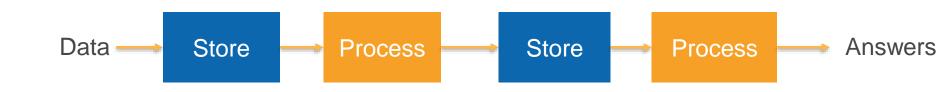
Design Patterns and Customer Examples



Primitive: Decoupled Data Bus



Storage decoupled from processing Multiple stages



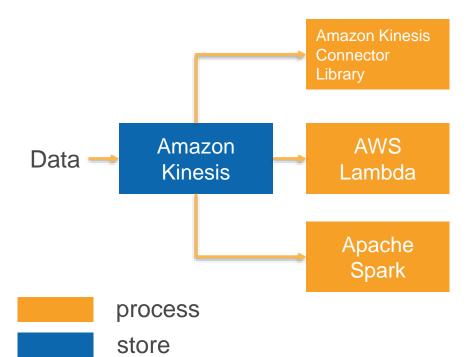




Primitive: Pub/Sub



Parallel stream consumption/processing

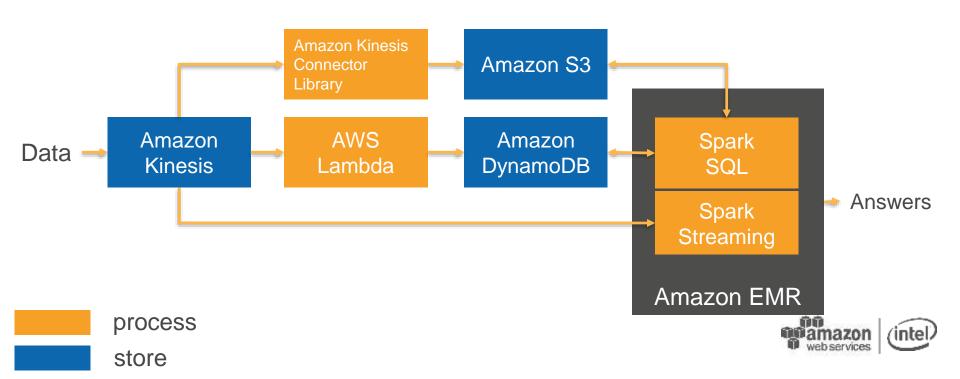




Primitive: Materialized Views

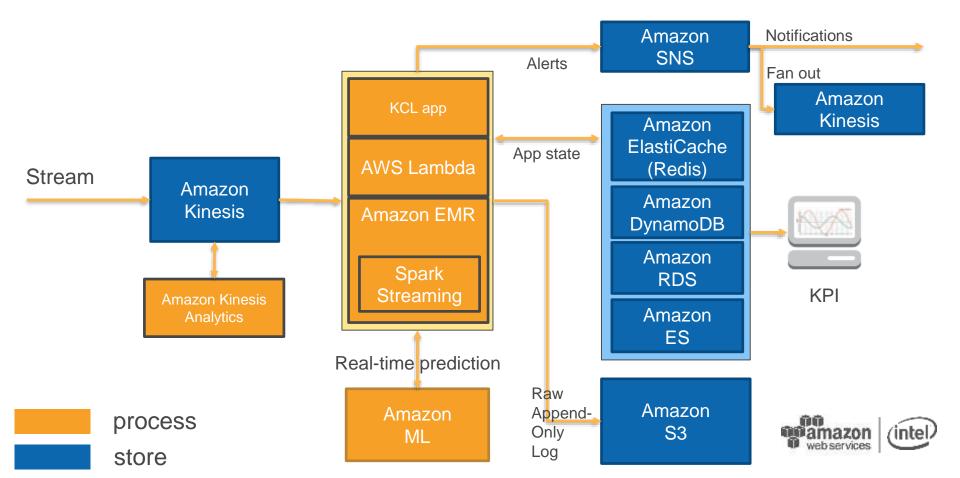


Analysis framework reads from or writes to multiple data stores



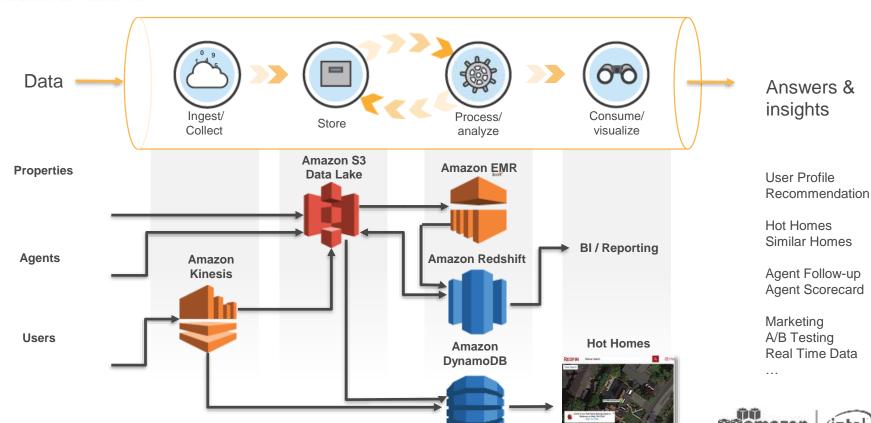
Real-time Analytics

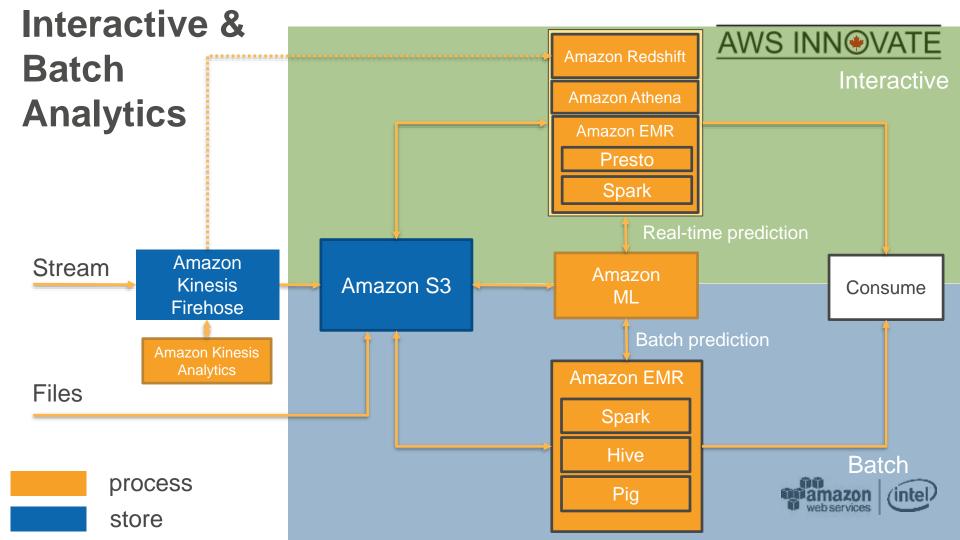








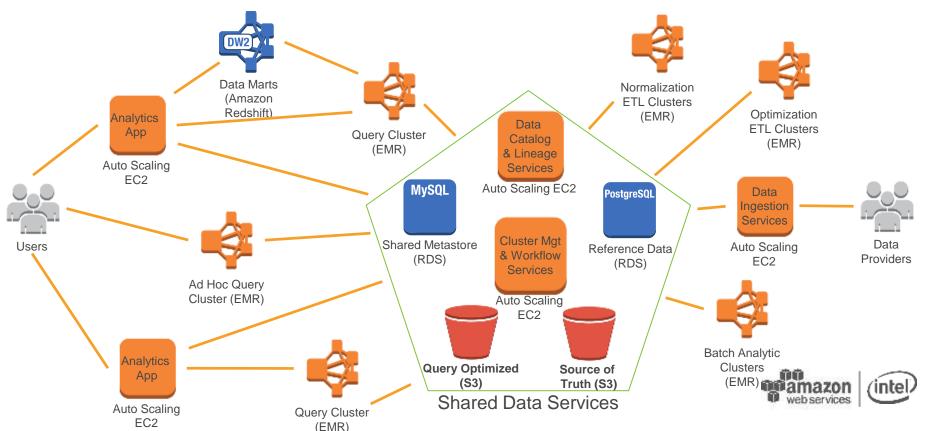






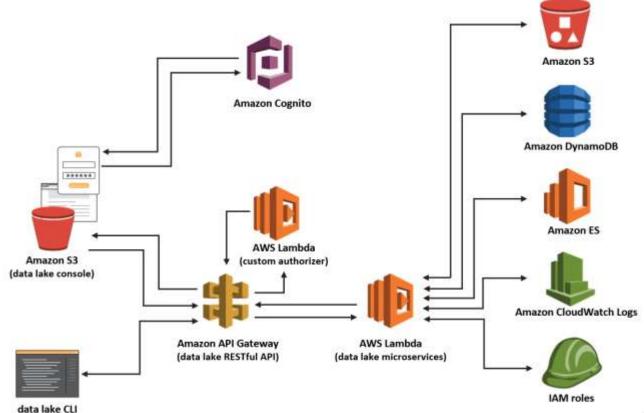


>5 PB, up to 75 billion events per day



Data Lake

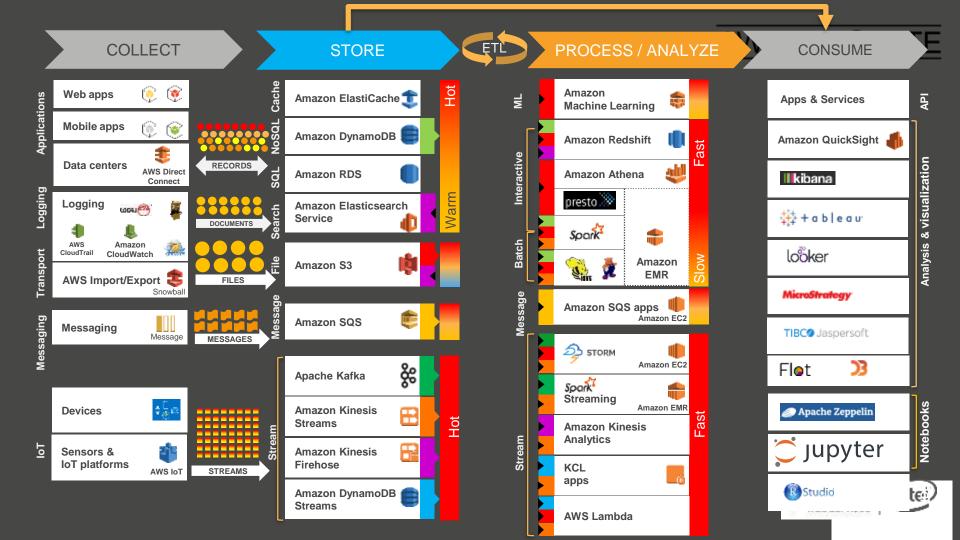




(RESTful API access)

Putting It All Together





Architectural Principles



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