AWS State, Local, and Education Learning Days

North Carolina





AI/ML For Data and Analytics Session subtitle

Jesse Roberts (he/him)

Principal Solutions Architect, Lead SA SLG/EDU US East Amazon Web Services

slgjesse@amazon.com



© 2024, Amazon Web Services, Inc. or its affiliates. All rights reserved.

Challenges we're hearing

Demand is rising, while resources and capacity to deliver them aren't keeping pace

Expectations are rising, with constituents demanding the same digital experiences they get from the private sector

Infrastructure is aging, creating friction across the data lifecycle (capture, storage, management, leveraging

Requirements are becoming more complex, creating barriers to change and blocking adoption of "built for change" solutions

Change is slowing, with risk-averse cultures introducing inertia and thwarting innovation

What internal and external customers are asking for

Better experiences that minimize time to access, provide self-service capabilities, and deliver in joined-up ways

More productive staff who have answers at their fingertips and aren't stuck doing manual work

Empowered teams that focus on outcomes, use make data-informed decisions, and meet legal and policy requirements

The Data Driven Organization



"An organization that harnesses data as an asset drive sustained innovation and create actionable insights to improve policy making decisions that reflect outcomes constituents care about, that builds greater trust."

Key Characteristics



Set 'Think Big' goals

ଚୁଦିତ୍ର

Shared leadership conviction and Business-IT alignment on data ownership



Upskilled and empowered producers and consumers who self-serve



Focus on delivering policy priorities with quality



Strong collaboration and agility concerning data products across data producers and consumers



Privacy, security, compliance and federated governance without impeding innovation





A Modern, Foundational Data Architecture

	Query, Sea	rch & Visualizatio	n Al/N	AI/ML, Data Analytics & Processing				
		lt Storage & bases		ilt Storage & bases	Simple Curated Storage			
		Data Preparation	, Matching, Enricl	nment & Curatio	'n			
evOps / ataOps			Data Governance			External Data		
	Data Ir	ngestion, Discover	y & Movement		Data Catalog	Sources & Products		
		Security, Id	entity & Access M	anagement				
aws	Use Case	Use Case	Use Case	Use Case	Use Case			
	4, Amazon Web Services, Inc. or its affiliates. Al	l rights reserved.				7		

De Da



Machine learning

- Data Discovery
- Data Attribution
- Metadata Tags
- Schema Definitions
- Data Catalogues
- Data Protection





A Modern, Foundational Data Architecture

	Query, Sea	arch & Visualizatio	n Al/	ML, Data Analytics	& Processing	
		ilt Storage & bases		uilt Storage & abases	Simple Curated Storage	
		Data Preparation	, Matching, Enric	chment & Curatior		
			Data Governanc	e		
ps / Ops	Data I	ngestion, Discover	y & Movement	D	ata Catalog	
		Security, Ide	entity & Access N	anagement		
۲ <u>S</u>	Use Case	Use Case	Use Case	Use Case	Use Case	

Extract insights from unstructured content

Extract insights from unstructured documents and forms, like images, PDFs, and audio

- Analyze text with natural language processing (NLP) to identify topics, extract entities, understand sentiment, and classify documents with Amazon Textract, Amazon Rekognition, and Amazon Comprehend
- Translate content at scale with Amazon Translate



Content analysis and object detection

Extract insights and identify objects of interest from large volumes of images and videos with Amazon Rekognition

- Detect personal protective equipment (PPE) to improve worker safety
- Analyze vehicle traffic and pedestrian and bicycle safety
- Detect objects of interest in video and reduce human effort required to review footage



Text to Speech Transcription, Classification & Analytics

AN EXAMPLE FROM AWS POST CALL ANALYTICS AND CALL SUMMARIZATION

\$50 One second 10 15 20 years one				
ienAl Transcript Summary	Call Analytics Sum	imary		
ummary ne customer asked about the cashback rate of their rewards card, and paid \$50 to the ccount balance.	Issue I'm calling about my rewards ca	ırd.		
I Customer Sentiment ositive 🥹	Action Items No action items detected.	Home > Call List > Call Details		
l Agent Sentiment	Outcomes	Call Details Info		Swap Agent/Call
eutral 😀	your card balance is \$75.34.			
ow could the overall experience be improved? ne agent could have not assumed the customer's pronouns, and also asked for their		Call Metadata	Transcribe Details	Sentiment
ddress.		Timestamp	туре	Customer Agent
		2023-02-22 12:23:49	Transcribe Call Analytics	5
ranscript		Guid	Job Id	2
		102	Card2_GUID_102_AGENT_AndrewK_DT_2023-02-22T12- 23-49.wav	0 0
Agent - 00:03		Agent		
Gent - 00:03 Hello, thanks for calling Bank. Uh my name is, how can I help you?		AndrewK	File Format wav	-5 Q1 Q2 Q3
Customer 00:07		Call Duration	Sample Rate	Call Quarter
Customer - 00:07 Hi, <mark>[Issue]: I'm calling about my rewards card.</mark>		03:02	8000	Speaker Time
Agent - 00:11		Entity Recognizer Name sample-entities.csv	PII Redaction	Agent dustomer discussioner
Agent - 00:11 Excellent. Give me a second to look up your customer account for verification purpose:	s. Can you please uh state your first nar		Enabled	100.0% 90.0%
Customer - 00:20		Language Model en-US	Custom Vocabulary	80.0%
My first name is [DII]			-	70.0% 60.0%
		Agent Sentiment	Vocabulary Filter	50.0%
		Sentiment: ^{CD} Trend: ~7	-	40.0%
		Customer Sentiment	Average Word Confidence	20.0%
		Sentiment: Trend: ~	Average word confidence	10.0%

A Modern, Foundational Data Architecture

	Purpose Bui	rch & Visualizatio It Storage & bases			L, Data Analytic t Storage & ases	s & Processing Simple Curated Storage	
		Data Preparation					External
Ops / aOps	Data Ir	ngestion, Discove	ry & Mov	ement		Data Catalog	Data Sources & Products
WS	Use Case	Security, Id Use Case	-	Access Ma Case	nagement Use Case	Use Case	



Data Governance

Data Protection

ML-powered sensitive data identification and redaction

Amazon Macie is a data security service that uses machine learning (ML) and pattern matching to discover and help protect your sensitive data.

Data Quality

Use ML to detect anomalies and hard-to-detect data quality issues

AWS Glue Data Quality

learns patterns on data statistics gathered over time using ML algorithms. It detects anomalies, unusual data patterns and alerts users. It also auto-creates rules to monitor these specific patterns so that you can progressively build data quality rules.

Entity Resolution & Data Matching

Match, link, and enhance related customer, product, business, or healthcare records stored across multiple applications, channels, and data stores

AWS Entity Resolution and AWS LakeFormation FindMatches enable you to identify matching records in your dataset even when the records to not have a common unique identifier and no fields match exactly.

ML-powered Matching & Entity Resolution



A Modern, Foundational Data Architecture

	Query, Sea	rch & Visualizatio	n Al/N	1L, Data Analytic	s & Processing	
	Purpose Bui Data	lt Storage & bases		lt Storage & bases	Simple Curated Storage	
		Data Preparation	, Matching, Enrich	nment & Curatior	٦	
			Data Governance			
iOps	Data Ir	gestion, Discover	y & Movement		ata Catalog	
		Security, Ide	entity & Access M	anagement		
VS and the second	Use Case	Use Case	Use Case	Use Case	Use Case	17

aw

Data Preparation, Matching, Enrichment & Curation

ML-based data transformations

ML-generated data insights & data augmentation

• Trends, anomalies, categorization, classification, labeling, sentiment, summarization, etc.

ML services for can be integrated into data processing jobs

• E.g., translation, inference, labeling, etc.

Generative AI powered data integration coding assistance



Generative AI assistance for data integration

aws Ervices Q Search	[Option+S]		Ъ 🗘 🕐 🙆 Tokyo ▼	Username @ 1234-5678-9012 🔻
AWS Glue	AWS Glue > Jobs			() ()
Getting started	AWS Glue Studio Info			
ETL jobs Visual ETL	Create job Info			
Notebooks Job run monitoring Data Catalog tables Data connections	Author in a visual interface focused on data flow. Visual ETL	Author using an interactive code notebook.	Author code with a script edit Script editor	tor.
Workflows (orchestration)	Example jobs Info		Crea	ate example job
 Data Catalog Data Integration and ETL 				
 Legacy pages 	Your jobs (2) Info		C Actions V	
	Q, demo	X 2 matches		< 1 > ©
What's New 亿 Documentation 亿	□ Job name	Last modified	▼ AWS Glue version	
AWS Marketplace	q-demo-taxi Glue ETL q-demo Glue ETL	4/26/2024, 1:19:07 PM 4/25/2024, 3:41:38 PM	4.0	
Enable compact mode				
Enable new navigation				
CloudShell Feedback			24, Amazon Web Services, Inc. or its affiliates. Privac	cy Terms Cookie preferences

A Modern, Foundational Data Architecture

	Query, Sea	rch & Visualizatio	n Al/N	1L, Data Analytics	& Processing	
		lt Storage & bases	-	ilt Storage & bases	Simple Curated Storage	
		Data Preparation	, Matching, Enric	nment & Curatior	ו	
			Data Governance	9		
Ops	Data Ir	gestion, Discover	y & Movement	D	ata Catalog	
		Security, Ide	entity & Access M	anagement		
/S	Use Case	Use Case	Use Case	Use Case	Use Case	



AWS brings ML closer to data



ML Purpose-built databases

- Databases with in-built AI/ML support like vector stores (Amazon Aurora PostgreSQL-Compatible & RDS for PostgreSQL, Amazon Neptune ML, Amazon MemoryDB for Redis, Amazon Document DB, and Amazon OpenSearch
- Integrate ML into queries using run-time inference inside SQL as done in Amazon Redshift ML and Athena ML

Inference at query time – Link Prediction

Without inference

g.V().has('name', 'Bob').
out('liked').
hasLabel('movie').values('title')

 $\Rightarrow A Bugs Life$ $\Rightarrow Monsters Inc.$

With inference

```
g.with("Neptune#ml.endpoint","ENDPOINT").
V().has(`name', `Bob').
out('liked').with("Neptune#ml.prediction").
hasLabel(`movie').values(`title')
```



 \Rightarrow Toy Story



Train & use models in Amazon Redshift ML

TRAIN





PREDICT



Example: Train a model on data from before 2020-01-01, then use the prediction function on the testing set. The following query displays the predictions of whether customers who signed up after 2020-01-01 will go through churn or not.

A Modern, Foundational Data Architecture

	Query, Search & Visualizat	ion Al/N	1L, Data Analytic	s & Processing	
	Purpose Built Storage & Databases		lt Storage & bases	Simple Curated Storage	
	Data Preparatio	on, Matching, Enric	hment & Curatio	n	
		Data Governance			
ps / Dps	Data Ingestion, Discove	ery & Movement		Data Catalog	
	Security, I	dentity & Access M	anagement		
S	Use Case Use Case	Use Case	Use Case	Use Case	

AI/ML in Query, Search & Visualization

- Natural Language insights from GenAI database query
 - text-to-sql
 - natural language search
 - Visualization generation
- Content & data summarization & classification
 - AI-assisted data story telling



Al-powered dashboard authoring experience

A NEW DASHBOARD BUILDING EXPERIENCE POWERED BY GENERATIVE BI

Build visuals

Use natural language to quickly build visuals for dashboards and reports

Create calculations

Build calculations using natural language without looking up or learning specific syntax

Refine visuals

Quickly update visuals by describing desired formats using natural language



Comp Add a dimension Comp ADD: Comp Comp ADD: Comp	Acked > Vehicle registration Stat	I Ton Pickup Interpreted as: Total Registrations by SEGMENT.
Dataset 100% SPICE REGISTRATIONS_G ▼ Search fields Q + CALCULATED FIELD Horizontal states Location Y AXIS O STATE_NAME Model O DMA_NAME Add a dimension AGE VALUE	acked > Full Size 3qtr to Subco Lower Mid Size I M SUB Full Size I M SUB Full Size I M SUB Cor	I Ton Pickup npact Utility uxury Utility d Size Sport e Luxury Car pact Pickup Full Size Car uxury Sport
ETHNICITY Add a measure GROUP/COLOR GROUP/COLOR	Full Size Lu	mpact Sport
= Luxury Count = Luxury Ratio	Totat venictes registered	0 200K 400K 600K 800K 1,000K ADD TO ANALYSIS
LUXURY_STATUS SMALL MULTIPLES Add a dimension)	Did you mean
MDL2 Model	45-54 YEARS	MoM % Change in Registrations by SEGMENT.
 MoM % Change in Registrations Most common brands Registrations 	- 55-64 YEARS	
REPORT_DATE SEGMENT	65-74 YEARS	



AI–assisted storytelling

IMPACTFUL DATA STORYTELLING TO DRIVE ACTIONS

Interpret data for others

Help others derive meaning from data and reach conclusions to drive decisions

Generate stories using Al

Produce cohesive, powerful, and insightful narratives by analyzing only a few words of data

Create refined content

aws

Control AI verbosity, customize narrative text, and apply stunning visual themes to bring content to life

Share up-to-date governed data Quickly update and disseminate data at any time



Al answers to questions of data on demand

DEEP INSIGHTS AT YOUR FINGERTIPS

Executive summaries of dashboards

Instant summaries of key dashboard insights in natural language explaining top movers, outliers, and more

Powerful Q&A for nonexperts

Suggested questions and "what's in my data" show what can be asked

Multivisual answers with narrative insight summaries explain answer context

Support for vague questions and "did you mean" alternatives enable iterative fact-finding





Demo - GenAl Text to SQL

This market research assistant utilizes a modern data architecture and generative AI techniques to empower an organization's market research efforts.

By integrating various data sources and leveraging natural language capabilities, the assistant allows non-technical users to easily query and analyze market research data using conversational natural language.



Demo – Conversationally Interact with EHR Data

By combining HealthLake with a large language model, healthcare providers can interact conversationally with their data, gaining insights and making decisions faster than ever before.

One way to use a large language model with HealthLake is through a chatbot interface. Users can ask questions about both their structured data (e.g., Electronic Health Records (EHR)) or their unstructured data (e.g., doctor's notes).

	$\leftarrow \ \rightarrow$	C ŵ	🔿 ลิ ତଂ https://d3nmipk0gd49rk.cloudfront.net/index.html	120%	🖞 🥶 🔞	=
	≡	Amazon	HealthLake ChatBot[hcls_ai_tester]			0
ers ally			o with healthcare insights. You can search structured data by asking about the number of patients with a given condition or taking a you can search unstructured data by typing "Search Doctor's notes for" for some information.			
an						
ed	т,	r click on the n	nic			Ŷ
,	_					

A Modern, Foundational Data Architecture

	Query, Sea	rch & Visualizatic	n Al/N	ባL, Data Analytics	& Processing	
	Purpose Buil Datab			lt Storage & bases	Simple Curated Storage	
		Data Preparatior	n, Matching, Enric	nment & Curation		
			Data Governance			
ps / Dps	Data In	gestion, Discover	ry & Movement	D	ata Catalog	
		Security, Id	entity & Access M	anagement		
S	Use Case	Use Case	Use Case	Use Case	Use Case	
7						71



Fraud detection and prevention

Detect and prevent fraud, waste, and abuse

- Enhance accuracy and speed to help detect and prevent waste fraud and abuse
- Managed service approach with prebuilt ML models for fraud detection
- Supervised and unsupervised models for developing highly targeted models to utilize customer data as part of fraud prevention efforts





Predictions and forecasts from IoT and sensor data

Leverage data from smart cities and facilities

• Smart cities

- Predictive maintenance
- Facility management



Machine learning to forecast trends and support decisions

Accurate, time series forecast with machine learning

- Predicting service demand or program activities
- Allocating resources to optimize impact and outcomes for citizens
- Financial planning and revenue / cost forecasts



No Code Predictive Analytics Value Proposition

Accelerate data science teams

Do more with your current team by using low-code machine learning tools in order to get to the desired outcomes faster.



Enable business teams

Give business teams the ability to do ML without any code, scaling the number of people who can create ML powered insights, forecasts, and predictions

Collaborate together

Leverage the integrated capabilities of Amazon QuickSight and Amazon SageMaker Canvas making it easy for business users to use machine learning or for data scientists to make changes on the models business users build and creating one place for all the analytics and machine learning in a team or organization.

What can we do to maximize the opportunities presented by AI/ML and GenAI?

- Develop and begin implementing data modernization strategy & modern data architecture
- Encourage learning & managed experimentation
- Participate in public sector efforts to develop best practices
- Adopt an iterative approach: Think Big, Start Small, Scale Fast



AWS is here to help





Thank you!

Jesse Roberts (he/him)

Principal Solutions Architect, Lead SA SLG/EDU US East Amazon Web Services <u>slgjesse@amazon.com</u> Please complete the survey for this session



Track Name: Data & Analytics

Session Name: AI/ML for Data and Analytics

