# AWS State, Local, and Education Learning Days

Philadelphia





# **AI/ML For Data and Analytics**

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# 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



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

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# Machine learning

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





# A Modern, Foundational Data Architecture

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### 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



### **Speech to Text Transcription, Classification & Analytics**

#### AN EXAMPLE FROM AWS POST CALL ANALYTICS AND CALL SUMMARIZATION

\$50 One second 10 15 20 years one				
GenAl Transcript Summary	Call Analytics Sum	imary		
<b>iummary</b> 'he 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	rd.		
N Customer Sentiment Positive 🥪	Action Items No action items detected. Outcomes	Home > Call List > Call Details		Swap Agent/Calle
leutral 😐	your card balance is \$75.34.			
ow could the overall experience be improved? he agent could have not assumed the customer's pronouns, and also asked for their ddress.		Call Metadata Timestamp 2023-02-22 12-23:49	Transcribe Details	Sentiment
ranscript		Guid 102	Job Id Card2_GUID_102_AGENT_AndrewK_DT_2023-02-22T12- 23-49.wav	Soon
Agent - 00:03 Hello, thanks for calling Bank. Uh my name is, how can I help you?		AndrewK Call Duration	File Format wav	-5 01 02 03 0
Customer - 00:07 Hi, [Issue]: I'm calling about my rewards card,		03:02	Sample Rate 8000	Speaker Time
excellent. Give me a second to look up your customer account for verification purposes	s. Can you please uh state your first nar	sample-entities.csv	Pli Redaction Enabled	Agent Customer Silence
Customer - 00:20		en-US Agent Sentiment	Custom Vocabulary -	80.0% 70.0% 60.0%
		Sentiment: <sup>(2)</sup> Trend: <sup>~7</sup>	Vocabulary Filter -	50.0% 40.0% 30.0%
		Customer Sentiment	Average Word Confidence	20.0%
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### **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 do not have a common unique identifier and no fields match exactly.

# **ML-powered Matching & Entity Resolution**



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# 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 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**

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### 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







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.

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# 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



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Luxury Count     Luxury Ratio	GENDER :	Total vehicles registered by age	0 200K 400K 600K 800K 1,000K
<ul> <li>LUXURY_STATUS</li> <li>MAKE</li> </ul>	SMALL MULTIPLES	35-44 YEARS	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	
		25-34 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**

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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).

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### 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!

#### **Badri Patel**

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#### **Track Name: Data & Analytics**

Session Name: AI/ML for Data and Analytics

