

Generative AI is Transformative



NEW EXPERIENCES

Create new innovative and engaging ways of interacting with your customers and employees



PRODUCTIVITY

Radically improve productivity across all lines of business



INSIGHTS

Extract insights and clear answers from all your corporate information, enabling faster and better decisions



CREATIVITY

Create new content and ideas, including conversations, stories, images, videos, and music



Gen Al for Maintenance



Gen Al for Maintenance - Chatbot

Example

Step 1: Upload document to platform

Transformers:
Basics, Maintenance, and Diagnostics



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256 Page PDF guide

Step 2: Ask it questions

I've found some burned paper near a transformer. Is this a problem?

How do I perform a DP test?

Step 3: Get the answer

It is unclear whether the burned paper outside the transformer is a problem without further information and testing. The document provides information on how to perform a DP test...

See: pages 138, 162, 197, 216

To perform a DP test, remove a sample of the paper insulation from a convenient location near the top of the center phase

See: page 243



Gen Al for Maintenance - IoT

Step 1: Upload document to platform

Transformers: Basics, Maintenance, and Diagnostics



U.S. Department of the Interior Bureau of Reclamation Technical Service Center Infrastructure Services Division

256 Page PDF guide

Step 2: Pass sensor data

```
{ "sensorData": {
  "timestamp": 1632578627,
  "motorId": "a1b2c3",
  "rpm": 2500,
  "temperature": 150,
  "voltage": 12.5,
  "amperage": 25,
} }
```

Step 3: Get the Analysis

RPM of 2500 is higher than the rated speed of 1445 rpm

>

See: page 243



Gen Al for Maintenance - Images



do you see any rust or cracks or freeze on the pipes in the image?

Yes, upon closer inspection, there appears to be a crack or fracture visible on one of the pipe segments or valve components in the image.

□ Copy ¬ Retry □ ¬

Gen Al for Maintenance - Video

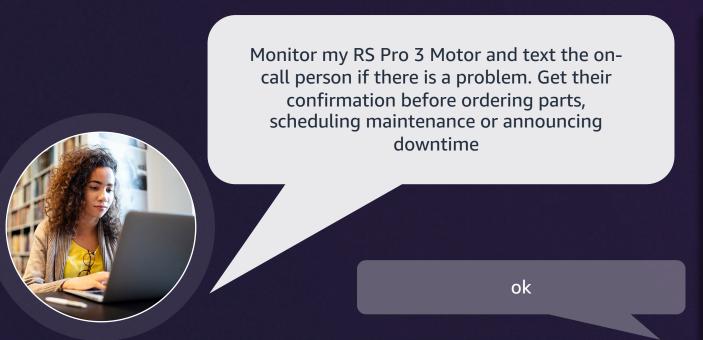


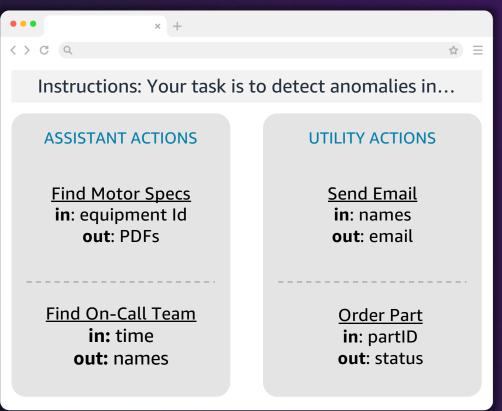
What time did the maintenance trucks arrive?

The trucks arrived late and only got here at 6:45 PM



Gen Al for Maintenance - Agents





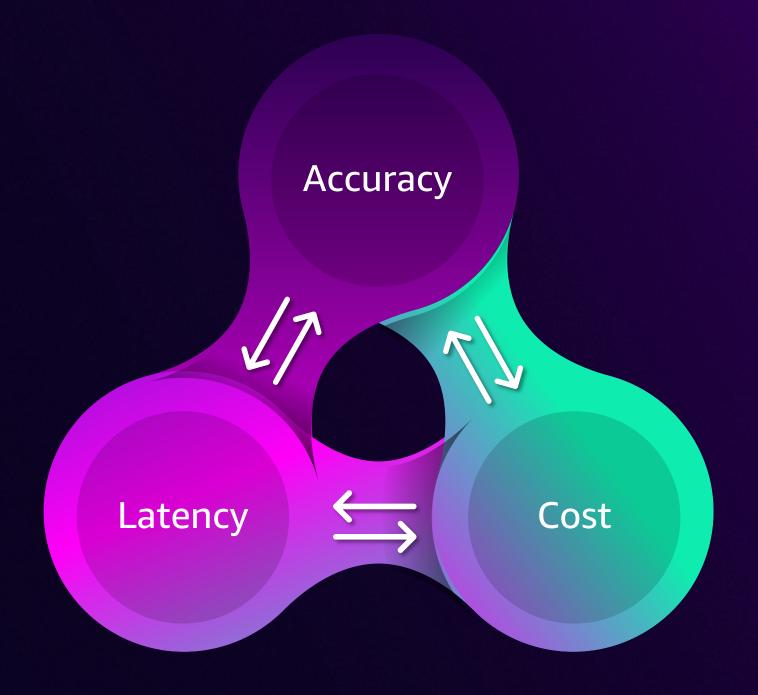
Gen Al questions over the years

2023 2024 2025 The year of POCs The year of Production The year of Business Value How can I leverage agents? What is generative AI? How do I prioritize my projects? How can we go deeper and wider? Is this secure? Do I need to become a prompt engineer? How can I lower my costs? How do I make this real? How can we transform our entire business? How do I choose a model? Where do I get started? What customization method should I use? How can I get all my employees to embrace AI?





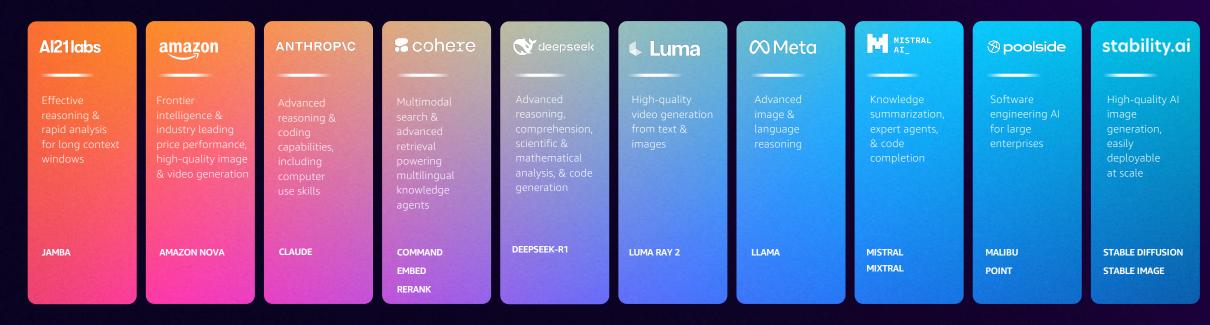
Yes, model selection is important





Amazon Bedrock

BROAD CHOICE OF MODELS



Coming soon

Amazon Bedrock Marketplace enables developers to discover, test, and use over 100 popular, emerging, and specialized foundation models (FMs) alongside the current selection of industry-leading models in Amazon Bedrock.

But your data is your differentiator



Generic generative Al





Generative AI that knows your business and your customers

How to provide your data to gen AI applications?

RAG

Guide foundational models by prompting with contextually relevant data (RAG)

Fine-tuning a Foundation Model

Fine-tune a foundation model using your curated, labeled data

Training your own purpose-built LLM

Train a foundation model using your curated, specialized data

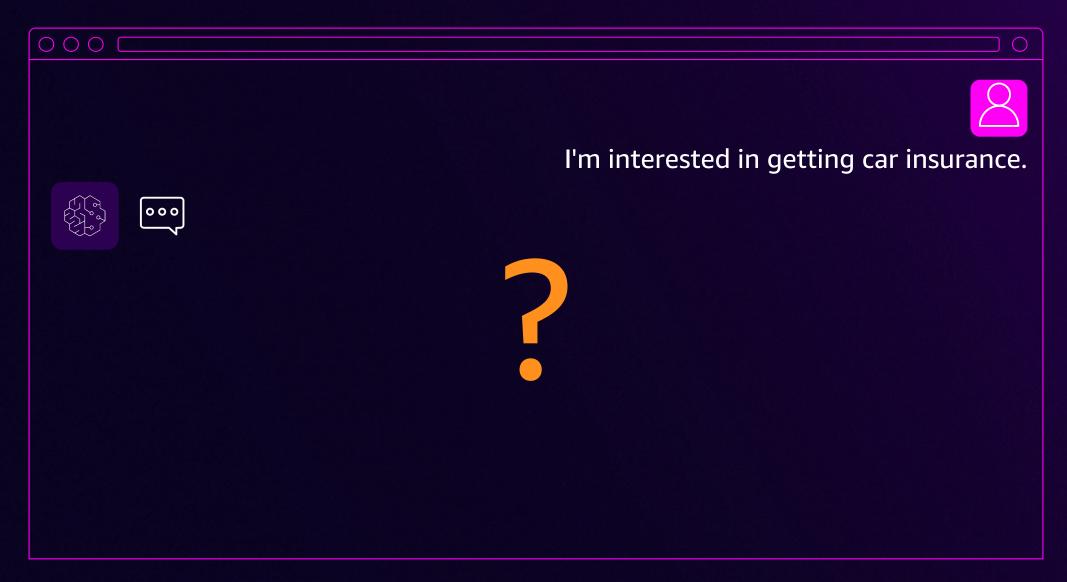


Pat needs Car Insurance

- Pat lives in Nevada and bought a new car!
- This is the 3rd car Pat has owned;
 Pat sold his previous cars
- Pat needs car insurance
- Pat wants to compare different rates and coverages

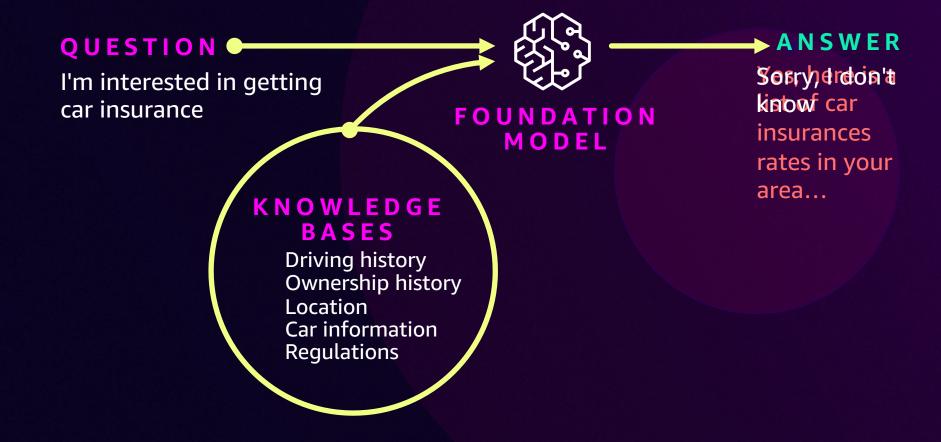


Pat needs car insurance





Retrieval Augmented Generation (RAG)





What type of data do you need for RAG?



USER QUERY

I'm interested in getting car insurance

Regulations, car photos, uploaded claims







Designing an augmented prompt

1. Instructions for the foundation model

2. Situational context

3. Semantic context

4. User input



User input

Prompt engineering

"I'm interested in getting car insurance."

You are a conversation agent for the HorizonGuard insurance marketplace. Responses to questions are written to be helpful and inform in an unbiased manner. Ask human to confirm responses.

Instructions for the model

User input

Prompt engineering

"I'm interested in getting car insurance."

You are a conversation agent for the HorizonGuard insurance marketplace. Responses to questions are written to be helpful and inform in an unbiased manner. Ask human to confirm responses.

The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Instructions for the model

Situational context

User input

Prompt engineering

"I'm interested in getting car insurance."

You are a conversation agent for the HorizonGuard insurance marketplace. Responses to questions are written to be helpful and inform in a unbiased manner. Ask human to confirm responses.

The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Car insurance is one of the most used types of insurance. Most states require purchase of some kind of coverage to legally drive. [...]. Reference: https://www....[...] Instructions for the model

Situational context

Semantic context



User input

Prompt engineering

"I'm interested in getting car insurance."

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The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Auto insurance is one of the most used types of personal insurance. Most states require that you purchase some kind of insurance coverage to drive legally in the state. [...]. Reference: https://content.naic.org/consumer/auto-insurance.htm

T....7

I'm interested in getting car insurance.

Instructions for the model

Situational context

Semantic context

User prompt/question



User input

Prompt augmentation

"I'm interested in getting car insurance."

The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Auto insurance is one of the most used types of personal insurance. Most states require that you purchase some kind of insurance coverage to drive legally in the state. [...]. Reference: https://content.naic.org/consumer/auto-insurance.htm

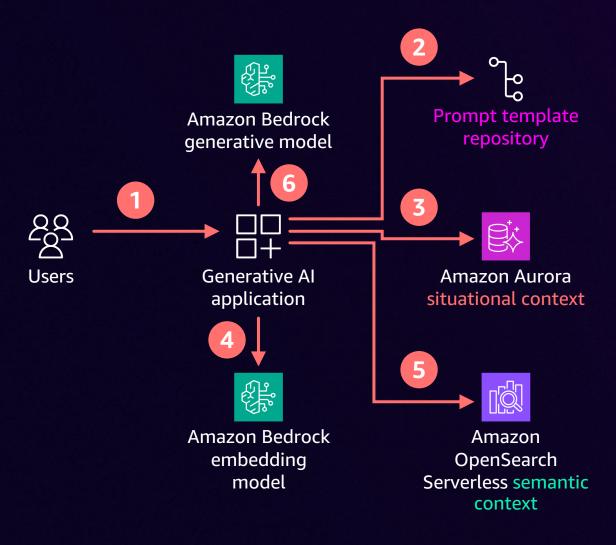
[...]

```
SELECT
  addresses.*, cars.*, dependents.*
FROM cars
JOIN addresses ON
  addresses.user_id = $1
JOIN dependents ON
  dependents.user_id = $1
WHERE cars.user_id = $1;
```

Situational context

Semantic context

Basic Architecture



- 1 User asks question
- 2 Retrieve prompt template (optional, if not baked into app)
- 3 Retrieve situational context from enterprise database
- 4 Use embedding model to create vector for question
- 5 Perform vector search and retrieve semantic context
- 6 Submit augmented prompt to generative model; get response



Impact of augmented prompt on the experience

User input

Prompt engineering

Foundation model response

"I'm interested in getting car insurance."

You are a conversation agent for the HorizonGuard insurance marketplace. Responses to questions are written to be helpful and inform in a unbiased manner. Ask human to confirm responses.

The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Auto insurance is one of the most used types of personal insurance. Most states require that you purchase some kind of insurance coverage to drive legally in the state. [...]. Reference: https://content.naic.org/consumer/auto-insurance.htm

[...]

I'm interested in getting car insurance.

Good morning, Pat, I'd be happy to help you with car insurance. I have your car details on record. You drive a 2024 SUV and live at 123 Noname St, Anytown, NV, USA. Is that correct?



Impact of augmented prompt on the experience

User input

Prompt engineering

Foundation model response

"Yes, that's correct."

You are a conversation agent for the HorizonGuard insurance marketplace. Responses to questions are written to be helpful and inform in a unbiased manner. Ask human to confirm responses.

The human is Pat, lives at 123 Noname St, Anytown, NV, USA and owns a 2024 SUV. Household members include 1 adult and 1 child of legal driving age.

Auto insurance is one of the most used types of personal insurance. Most states require that you purchase some kind of insurance coverage to drive legally in the state. [...]. Reference: https://content.naic.org/consumer/auto-insurance.htm

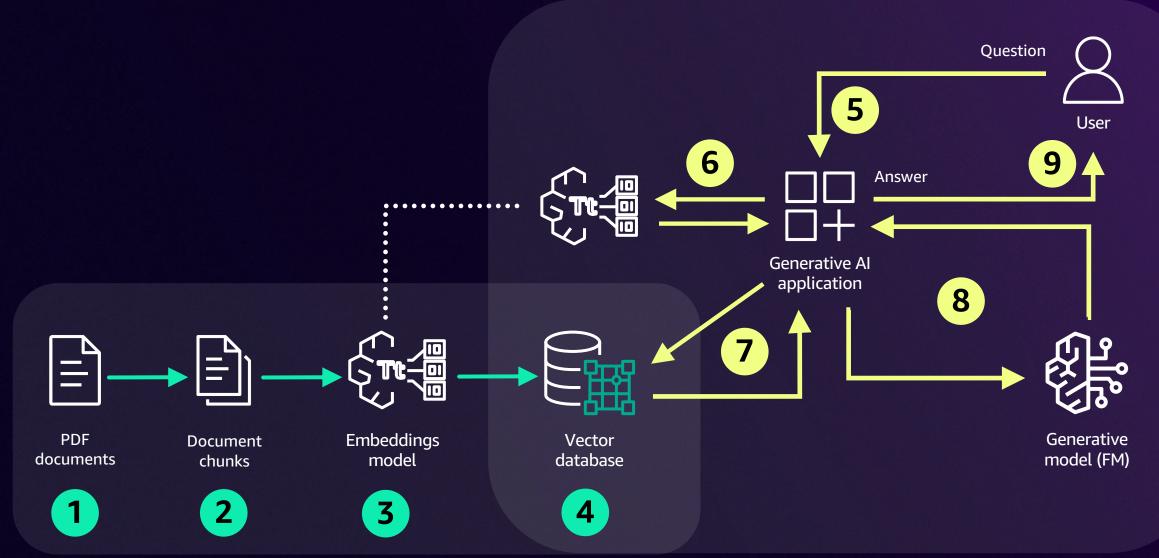
[...]

I'm interested in getting car insurance.

Car insurance in your area costs between \$200 and \$250 per month. Do you want more information?



RAG - Adding Semantic Context

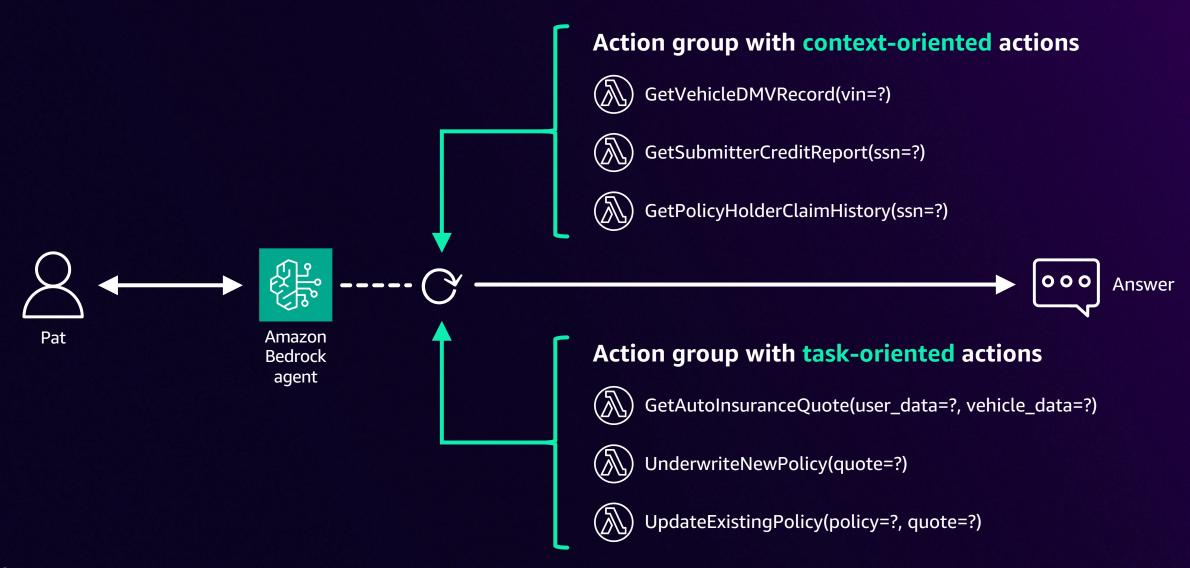




What about Agents?



Continuing the Car Insurance example

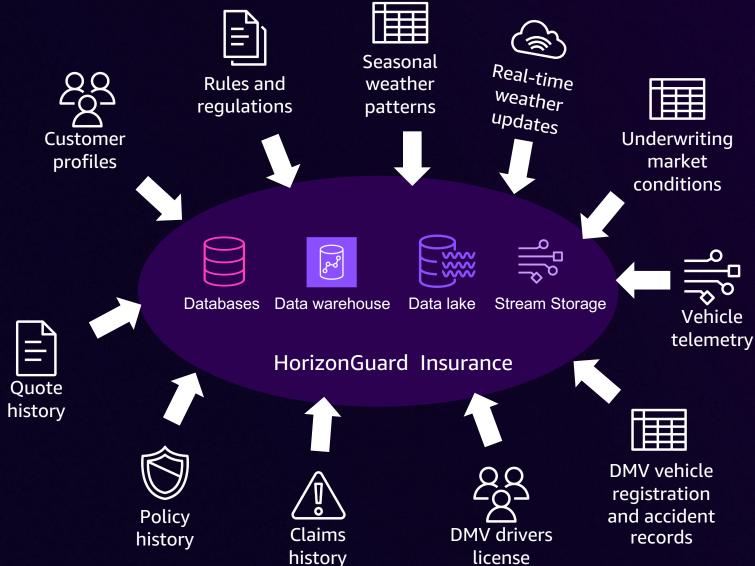


Establishing a GenAl Data Foundation [RAG Centric]



Understand where your RAG Context comes from

registrations



Challenges:

Data silos
Lineage and currency
Data quality
PII
Identifier resolution
Access controls

Types of Source Data



Structured

Existing database

Defined meaning, schema, and query patterns



Semi-structured

May not be under management

JSON, XML, STRUCT



Unstructured

Amazon S3, content management systems

May not be labeled



Enable Unstructured Data for RAG Applications

- Data transform is a must
- Understand data quality
- Define chunking strategy
- Generate embeddings
- Create a pipeline





Break into chunks



What's the most effective strategy?



Generate vectors



Which embedding model to choose?



Store and index



Which vector database to select?



Choose the right data architecture for RAG

Frontend data service layer

Milliseconds



Operational & vector database

User query response time

+

Data structure

+

Access patter

Seconds





Amazon Redshift

Amazon Athena



Data lake architecture

Backend data pipelines

Data freshness (lag from real time)

Data pattern (append, update)

Milliseconds: append or update





Streaming to operational & vector DBs

Seconds: append or update





Streaming to Amazon Redshift

Few minutes, append only



Streaming to data lake

---- Stream / Batch

Multiple minutes, updates





Batch to data lake

Address Data Quality and Lineage

TECHNICAL CHALLENGES

- Data format (e.g. email, phone number, policy ID)
- Data values

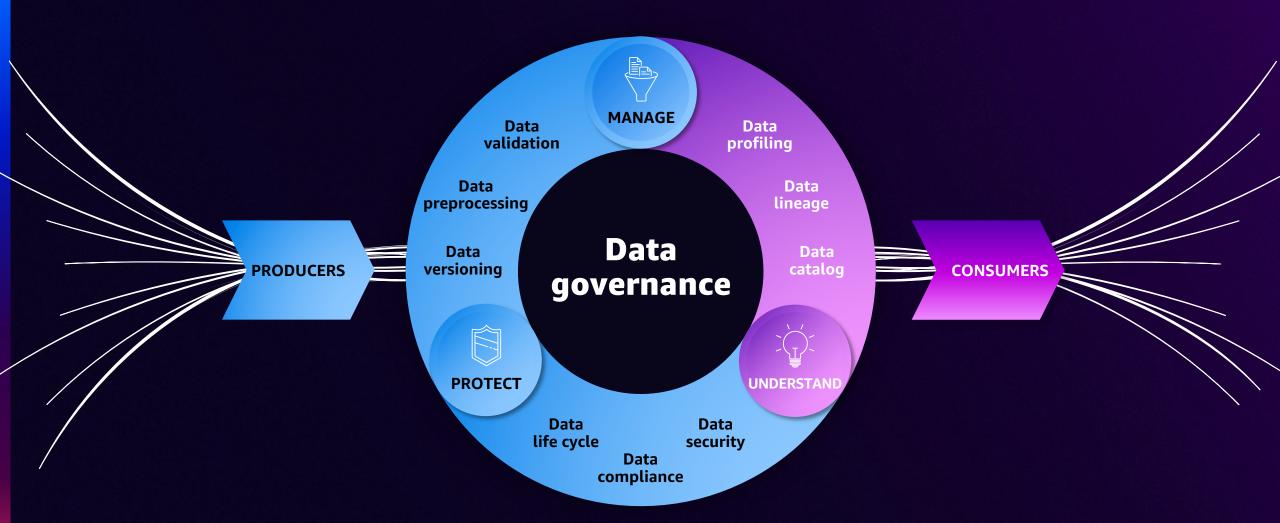
 (e.g. valid ranges, completeness)
- Lack of trust and transparency (e.g. Where did this data come from?)

BUSINESS CHALLENGES

- Meeting business rules (e.g. claim dates, policy limits)
- Regulatory compliance (e.g. minimum liability insurance limits)
- Building transparency and trust

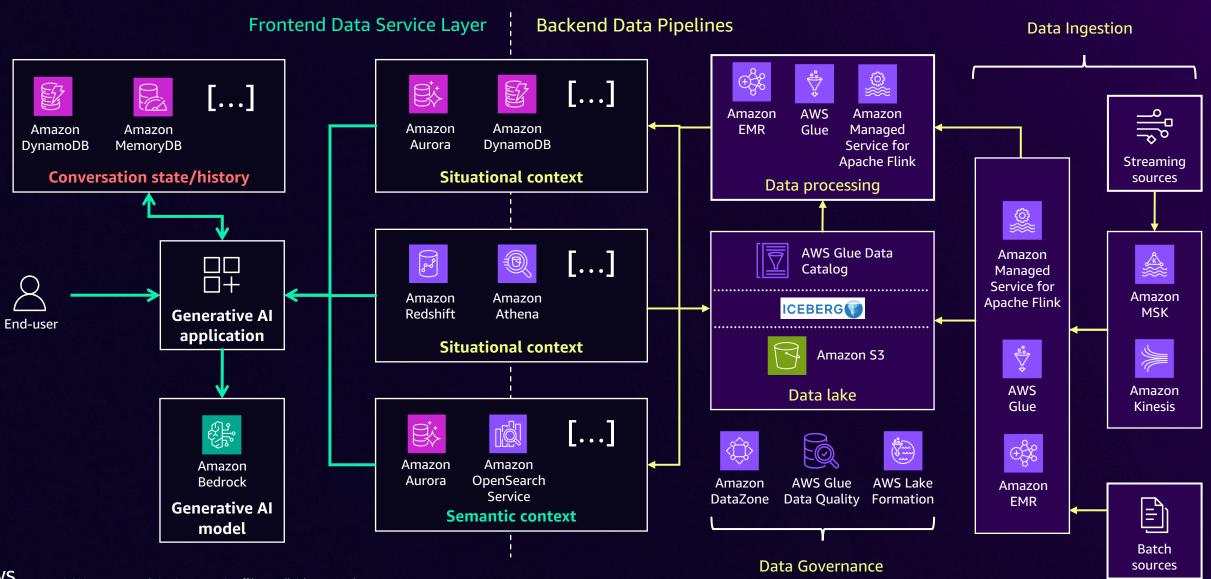


Establish Data Governance





Putting it all together



Adopt Big Data Best Practices



Build decoupled systems

- Data \rightarrow Store \rightarrow Process \rightarrow Store \rightarrow Analyze \rightarrow Answers Use the right tool for the job
- Data structure, latency, throughput, access patterns
 Leverage managed and serverless services
- Scalable/elastic, available, reliable, secure, no/low admin
 Use log-centric design patterns
- Immutable logs (in a data lake), and materialized views
 Be cost-conscious
 - Big Data ≠ big cost



Get cracking with Gen Al



Get your Data Foundation Right



Empower your teams with tools and trainings



Get started on high value use cases







Thank you!

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