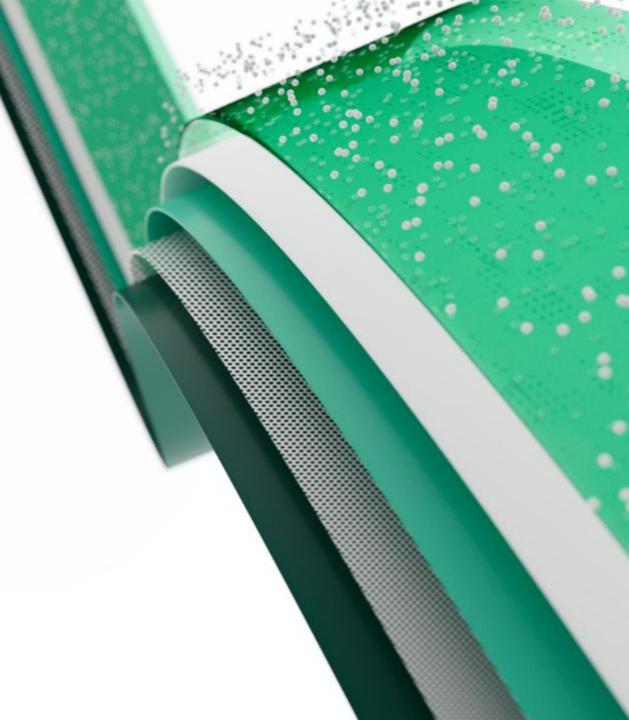


### Using Generative AI on Structured Data

(to query or modify data)

#### James Serra

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1/15/25

# About Me

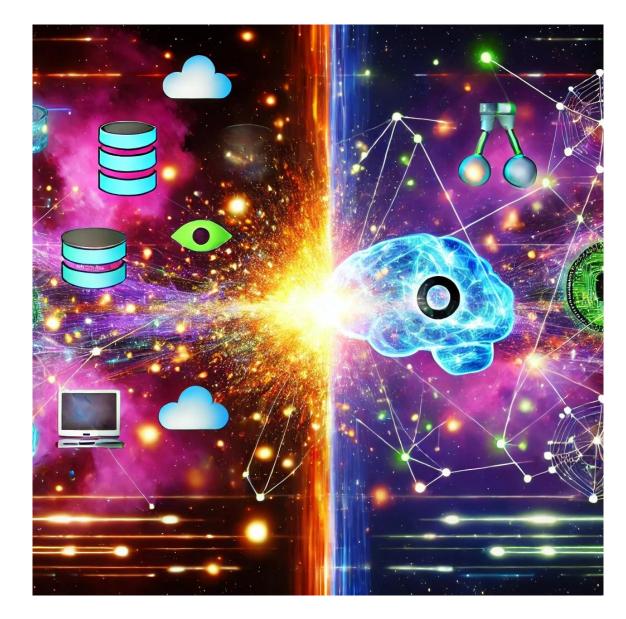


- Microsoft, Data & Al Solution Architect in Microsoft Federal Civilian
- At Microsoft for most of the last ten years as a Data & Al Architect, with a brief stop at EY
- In IT for 40 years, worked on many BI and DW projects
- Worked as desktop/web/database developer, DBA, BI and DW architect and developer, MDM architect, PDW/APS developer
- Been perm employee, contractor, consultant, business owner
- Presenter at PASS Summit, SQLBits, Data Summit, SQLDay, Enterprise Data World conference, Big Data Conference Europe, SQL Saturdays, Informatica World (sessionize top 3% most active)
- Blog at JamesSerra.com
- Former SQL Server MVP
- Author of the book "Deciphering Data Architectures: Choosing Between a Modern Data Warehouse, Data Fabric, Data Lakehouse, and Data Mesh"

# Agenda

- GenAl Definitions
- ChatGPT on semi-structured data
- Microsoft Fabric AI Skill on structured data
- Industry use cases

### Data platform and AI worlds are colliding...



### **GenAl Definitions**

# Is GenAl magic?



# Key definitions

- Al: Computational systems and models capable of performing tasks that typically require human intelligence. GenAl and ML are subsets of Al.
- **Generative AI (GenAI):** AI systems capable of *generating* new content, such as text, images, or audio. It does this by employing neural networks, a type of machine learning process that is loosely inspired by the way the human brain processes, interprets and learns from information over time.
- Large Language Models (LLMs): A type of GenAI designed for natural language (text) understanding and generation, trained on diverse datasets. Think of it like a super-smart auto-complete on steroids. GenAI uses other specialized models for image or video generation.
- **Machine learning (ML):** A broad subset of AI that encompasses algorithms and models capable of learning from data to make *predictions* or decisions. LLMs are a type of deep learning within ML, focused on language understanding and generation, whereas ML includes techniques for diverse tasks like image recognition, data analysis, and predictive modeling.
- **OpenAI**: A leading organization specializing in AI research and development, known for creating Generative AI models such as GPT (used in ChatGPT) and other AI technologies.
- **ChatGPT, Copilot**: Applications ("bots") built on Generative AI models (e.g., GPT). These tools allow users to interact with GenAI via natural language prompts, generating responses tailored to various contexts, such as conversations (ChatGPT) or coding (Copilot).
- **Prompt engineering**: The practice of designing input prompts to guide LLMs in generating accurate and relevant responses, optimizing the model's performance for specific tasks.

### GPT-3 was trained on multiple data sets

- Training Data was taken from the web with some processing
- Some biases inherent in the training data (as with any model)
- 530GB of text data, including conversational dialogues

Dataset	Quantity (tokens)	Weight in training mix	Epochs elapsed when training for 300B tokens
Common Crawl (filtered)	410 billion	60%	0.44
WebText2	19 billion	22%	2.9
Books1	12 billion	8%	1.9
Books2	55 billion	8%	0.43
Wikipedia	3 billion	3%	3.4

GTP-4 was trained using both text and image data and is significantly larger

• Details of training data are not public

Source: "Language Models are few shot learners", <u>https://doi.org/10.48550/arXiv.2005.14165</u> (Section 2.2)

# Data definitions

- Unstructured data: emails and documents such as .pdf, docx or .txt files (think text) -> old school
- Semi-structured data: files and logs in CSV, Parquet, XML, or JSON formats (often in table format, meaning rows and columns); Excel files -> new school
- Structured data: relational databases (SQL tables with rows and columns) -> new school
- **Binary data**: images, audio, video

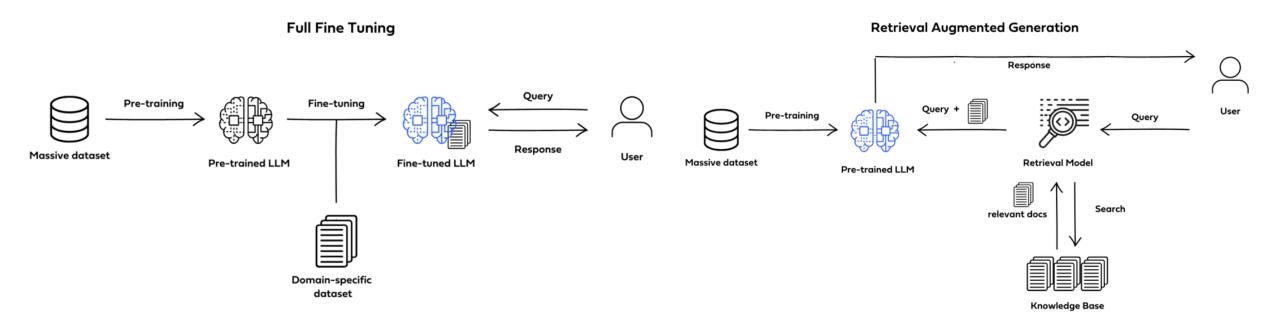
### **Model definitions**

**<u>Prompt Engineering</u>** is a technique that involves designing prompts for natural language processing models. This process improves accuracy and relevancy in responses, optimizing the performance of the model.

**Retrieval Augmented Generation (RAG)** improves LLM performance by retrieving data from external sources and incorporating it into a prompt. RAG allows businesses to achieve customized solutions while maintaining data relevance and optimizing costs.

**Fine-tuning** adapts an existing LLM using example data, resulting in a new "custom" LLM that has been optimized for the provided examples.

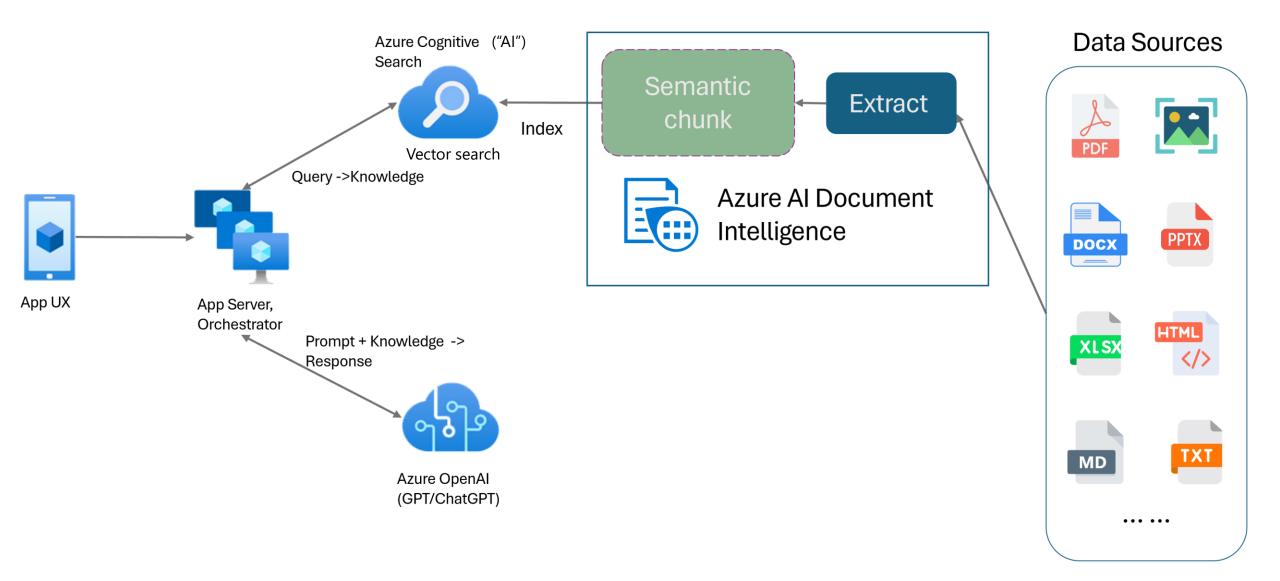
#### Fine-tuning vs RAG: A comparison of two techniques for enhancing LLMs



#### When to use Azure OpenAl fine-tuning

- Fine-Tuning creates a specialized model with fast, accurate responses for specific tasks but is costly in time and resources. It requires retraining to stay updated, making it static and harder to maintain over time.
- RAG provides real-time, dynamic information, ensuring more accurate answers for constantly changing data. It's cost-effective for scalability, but response times can be slower, and accuracy depends on the quality of external data sources.

#### Retrieval-Augmented Generation (RAG) with Azure AI Document Intelligence



# GenAl use cases on semi-structured and structured data

- Conversational querying
- Data enrichment and cleaning
- Sample data creation
- Data summarization
- Trend identification
- Forecasting and predictions
- What-if analysis
- Anomaly detection and correction
- Product/service recommendations
- Mapping fields, MDM, creating semantic models...

### **ChatGPT on semi-structured data**

### **ChatGPT definition**

The "GPT" in ChatGPT stands for generative pre-trained transformer. The acronym covers the key aspects of the magic behind the bot, as follows:

**Generative**: The model creates new content instead of merely classifying or labeling data.

**Pre-trained**: Before engaging in conversation or writing code, the model learns from massive datasets to understand language, grammar, and structure. ChatGPT is a master of pattern recognition and repetition, and it turns out that this is a nearly miraculous capability.

**Transformer**: Introduced in Google's 2017 paper <u>Attention Is All You Need</u>, this architecture excels at recognizing relationships between words, code, and data, making it adept at complex problem-solving.

### Bing Copilot (formerly Bing Chat) vs. ChatGPT

Copilot - here is a detailed list of the file types you can upload:

- Documents: DOCX, PDF, TXT, MD
- Images: JPG, JPEG, PNG, GIF
- Spreadsheets: XLSX, CSV
- **Presentations**: PPTX
- **Code files**: PY, JS, HTML, CSS, JAVA, C, CPP, SQL, XML, JSON, YAML, YML
- Compressed files: ZIP, RAR, TAR, GZ
- Audio files: MP3, WAV
- Video files: MP4, AVI, MOV Max file upload size: 50MB

Note: only the enterprise version of Copilot supports file uploads

	Copilot	ChatGPT
Model	OpenAI's GPT-4	OpenAl's GPT-40 mini (available for free); GPT-4, GPT-40, 01-preview, and o1 mini (paid tiers only)
Platform	Integrated with Microsoft's search engine; Google Chrome and Safari	Standalone website or API; iOS and Android apps
Internet access	Can perform web searches and offer links and recommendations	Web-browsing feature powered by Microsoft Bing
Image generation	Can generate images using DALL·E 3	Can generate images using DALL·E 3 (ChatGPT Plus only)
Voice capabilities	Speech-to-text only	Handles audio input and output (using GPT-40)
Conversation sharing	Can copy or export blocks of text	Can share links to entire conversations; anyone with the link can continue the conversation
Usage limits	Users get to ask 30 chats per session and 300 total chats per day	Unlimited conversations per day; ChatGPT Plus users get 50 GPT-4 messages every three hours
Pricing	Free	Free; ChatGPT Plus is available for \$20/month

ChatGPT - You can upload a variety of file types, including:

- Text and Document Files: PDF, DOC, DOCX, TXT, RTF
- Spreadsheet Files: XLS, XLSX, CSV
- Image Files: PNG, JPG, JPEG, BMP, GIF, and TIFF

• **Data Files**: JSON, XML, SQL Max file upload size: 50MB

Also: Google Gemini, Claude by Anthropic, Perplexity AI, Meta AI

#### ChatGPT versions

Plan	Cost	Features
Free	\$0/month	- Access to GPT-40 mini - Standard voice mode - Limited access to GPT-40 - <i>Limited access to file uploads, advanced data analysis, web browsing, and image generation</i> - Use custom GPTs
Plus	\$20/month	<ul> <li>Everything in Free</li> <li>Extended limits on messaging, file uploads, advanced data analysis, and image generation</li> <li>Standard and advanced voice mode</li> <li>Limited access to o1 and o1-mini</li> <li>Opportunities to test new features</li> <li>Create and use custom GPTs</li> </ul>
Pro	\$200/month	- Everything in Plus - Unlimited* access to GPT-4o and o1 - Unlimited* access to advanced voice - Access to o1 pro mode, which uses more compute for the best answers to the hardest questions
Team	\$25/user/month (annual) or \$30/user/month (monthly)	<ul> <li>Higher message limits than Plus on GPT-4, GPT-4o, and tools like DALL·E, web browsing, data analysis, and more</li> <li>Limited access to o1 and o1-mini</li> <li>Standard and advanced voice mode</li> <li>Create and share GPTs with your workspace</li> <li>Admin console for workspace management</li> <li>Team data excluded from training by default</li> </ul>
Enterprise	Custom Pricing	<ul> <li>Everything in Team</li> <li>High-speed access to GPT-4, GPT-4o, GPT-4o mini, and tools like DALL·E, web browsing, data analysis, and more</li> <li>Expanded context window for longer inputs</li> <li>Enterprise data excluded from training by default &amp; custom data retention windows</li> <li>Admin controls, domain verification, and analytics</li> <li>Enhanced support &amp; ongoing account management</li> </ul>

#### Microsoft 365 Copilot Microsoft Copilot

Feature	copilot.cloud.microsoft or m365.cloud.microsoft	copilot.microsoft.com
Enterprise Data Protection (EDP)	Yes	No
Microsoft Entra (Enterprise) Account Support	Yes	No
Access to Organizational Data (e.g., files, emails, chats)	Yes	No
Public Web Data & Bing Integration	No	Yes
AI-Powered Chat for Enterprise	Yes	No
AI-Powered Chat for General Use	No	Yes
Ad-free User Interface	Yes	No
Use in Microsoft 365 Apps	Yes (Word, Excel, PowerPoint, etc.)	No (Primarily browser- based)
File Upload Support	Yes (Various file types: .csv, .pdf, etc.)	Yes (Image uploads supported)
Image Upload Support	Yes	Yes
Customization (e.g., Copilot Studio)	Yes	No
Security, Privacy, Compliance	Enterprise-grade (high security)	Basic consumer-level security
Create Different Tabs for Separate Chats	Yes	No

Microsoft Copilot Free version

Microsoft Copilot Pro Subscription \$20.00 user/month

Microsoft 365 Copilot Subscription \$30.00 user/month

#### ChatGPT demo on semi-structured data

Also: askondata.com

#### **Microsoft Fabric AI Skill on structured data**

#### **Public Preview**

### **Generative AI on your data**

#### Deliver custom generative AI experiences for *your* data



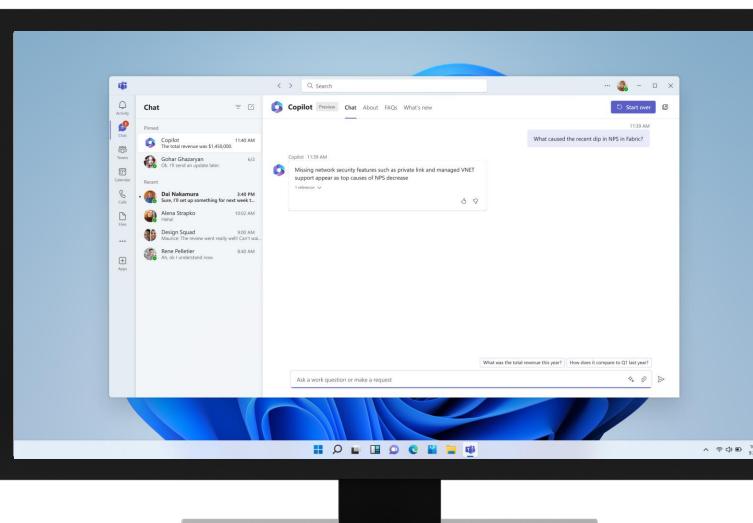
Enable custom Q&A on your data in Fabric



Define custom business semantics and grounding **unique to your org** 



Scale the custom experiences to M365 Chat, Copilot Studio, and Azure Al Studio



### AI Skills are evolving

A Data Agent for analyzing your data



The **conversational AI skill** allows users to interact with data naturally, enhancing accessibility and usability.



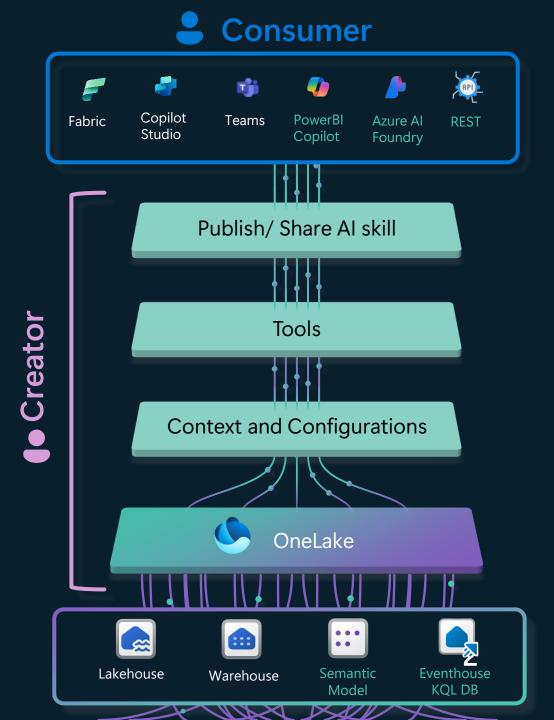
Improved chat canvas for creators with **new debugging capabilities**, making it easier to understand and refine responses.

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Seamlessly reason over multiple data sources, including **semantic models and Eventhouse KQL databases**, to create a powerful Data Expert tailored to your data domain.



Your Data Agent can be consumed inside and outside of Fabric. Stay tuned for **upcoming integrations** with Copilot Studio, Teams, Azure AI Foundry and your own custom applications.



### AI skill end-to-end flow

End-to-end scenario

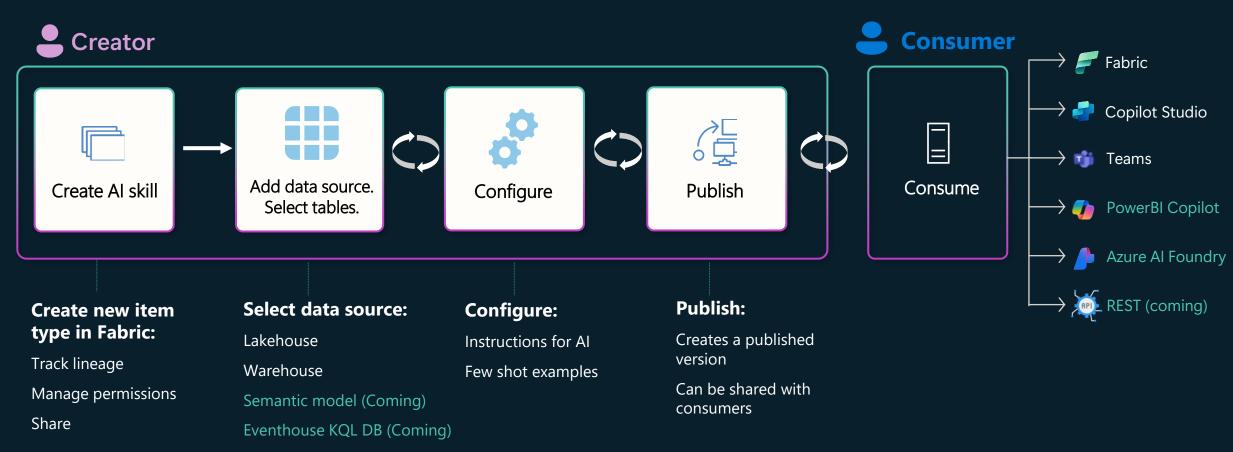
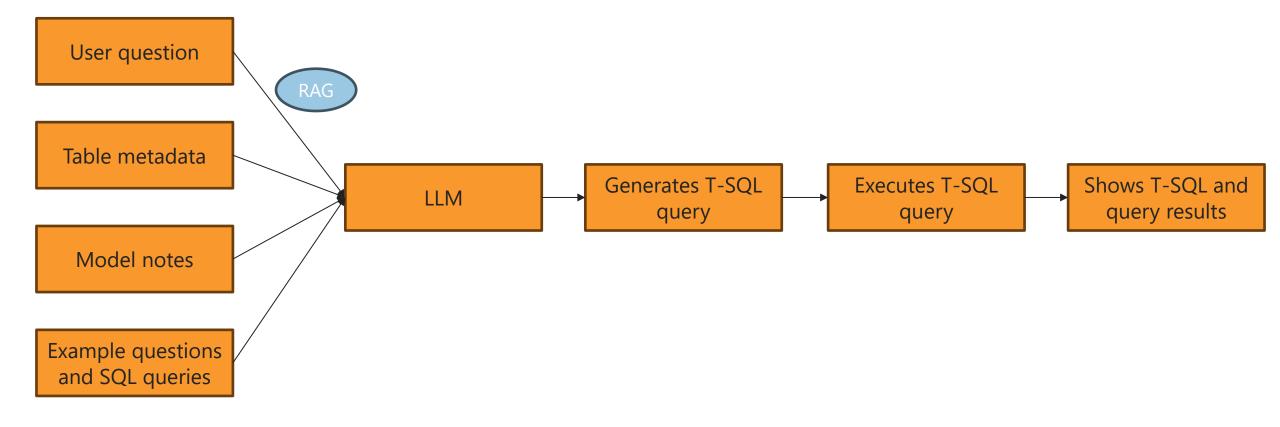
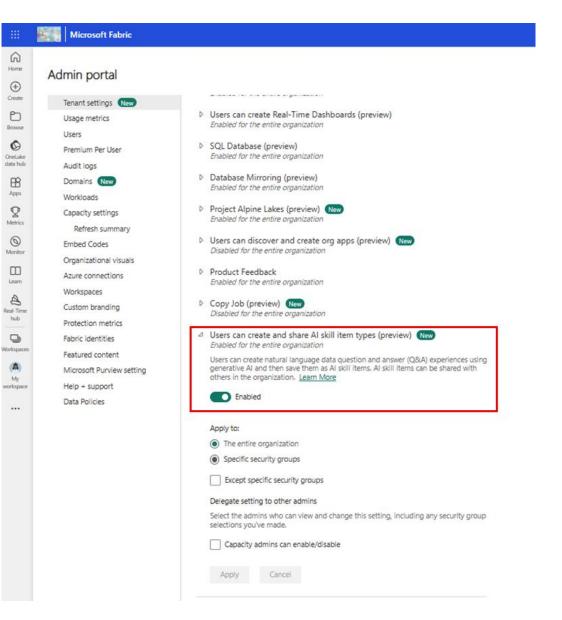


Table selection

## Microsoft Fabric AI Skill – model flow



# Al skill tenant setting



#### Microsoft Fabric AI Skill demo on structured data

https://learn.microsoft.com/en-us/fabric/data-science/ai-skill-scenario

#### Integration with Microsoft Copilot Studio

#### **Private Preview**

- Create custom agent in MCS with AI Skill as Knowledge Source
- Enhance your custom agent with different Knowledge Sources and Actions
- Publish your custom agent to Microsoft Teams

	ar Copilot Studio		Environment Fabric V2 & AU
Home (+) Create	+	Describe your copilot to create in Helpdesk Expense tracking HR and benefits	t
opilots	Use everyday w	ords to describe what your copilot should do This Al-powered feature above is currently in preview. <u>See terms</u> and <u>supplemental terms</u>	
	Start with a template Templates are currently a preview feature. See <u>supplemental terms</u>		See more
	Safe Travels Provides answers to common travel questions and related health and safety guidelines	Store Operations Improve the efficiency of retail frontline workers by enabling easy access to store procedures and policies	Sustainability Insights Enables users to easily get insights and data about a company's sustainability goals and progress
ħ	Team Navigator           Assists employees in finding colleagues and their hierarchy within the organization using Microsoft Teams	Voice           A Copilot with voice capabilities.	Weather Your go-to assistant for getting weather forecast
	Learning resources		
	X Quick start: Create and deploy a copilot	Documentation	2 Responsible AI FAQs
	Extend your copilots with plugins and extensions	Quick start: Use Generative Al in a copilot	Support community
	X Try Copilot Agents		

### What is the difference between AI Skill and Copilot?

The technology behind the AI skill and Fabric Copilot is similar. They both use Generative AI to reason over data. However, they have some key differences:

- **Configuration:** With an AI skill, you can configure the AI to behave the way you need. You can provide it with instructions and examples that tune it to your specific use case. A Fabric Copilot doesn't offer this configuration flexibility.
- Use Case: A Copilot can help you do your work in Fabric. It can help you generate Notebook code or Data Warehouse queries. In contrast, the AI skill operates independently, and you can eventually connect it to Microsoft Teams and other areas outside of Fabric.
- Purpose: Copilot on Power BI works against a report/model and updates the report, while AI skill is for ad-hoc queries that work against a lakehouse or warehouse and returns T-SQL and query results.

### Industry use cases

for using both structured and unstructured data

### Healthcare: Patient Treatment Optimization

- Scenario: A healthcare organization wants to improve patient outcomes by combining unstructured patient records, clinical notes, and structured data such as diagnostic codes and treatment history.
- **Approach**: Generative AI can analyze doctor notes, lab reports (unstructured data), and link them to patient demographics, medications, and treatments stored in relational databases (structured data) to suggest the most effective treatments.
- **Business Decision**: Optimizes treatment plans, improves patient outcomes, and reduces the time needed to identify suitable care paths.

### **Retail: Personalized Marketing**

- Scenario: A retail company wants to enhance its customer targeting for upcoming campaigns.
- **Approach**: The company combines historical purchase data (structured data) from SQL databases with customer feedback, product reviews, and social media sentiment (unstructured data) to generate personalized product recommendations.
- **Business Decision**: Increases customer engagement and drives revenue through targeted promotions.

#### **Finance: Fraud Detection**

- **Scenario**: A bank wants to improve its ability to detect fraudulent transactions by combining transactional data with suspicious customer communications.
- **Approach**: Generative AI processes emails, call transcripts, and legal documents (unstructured) alongside transactional records, financial logs, and account details (structured data) to identify patterns of potential fraud.
- Business Decision: Enhances fraud detection accuracy, reduces false positives, and minimizes financial risk.

### NFL Football: Game Strategy and Injury Risk Assessment

- Scenario: An NFL coaching staff wants to improve game strategy and reduce injury risks by analyzing both game data and medical reports.
- **Approach**: Structured data could include player stats (e.g., passing yards, rushing attempts, completion percentages), play-by-play game data, and historical win/loss records stored in relational databases. Unstructured data would include medical reports, injury rehabilitation progress notes, player interviews, and social media activity that may indicate player morale or readiness to play. Generative AI can combine both data types to generate recommendations for game-day decisions, such as which players to rest, which plays to run based on the opponent's weaknesses, and which players are at risk of re-injury.
- **Business Decision**: Al-generated insights can suggest optimal game strategies by evaluating player conditions, predicting fatigue or injury risk, and highlighting how certain players perform against specific types of defenses. This helps the coaching staff make more precise decisions regarding lineups, in-game play calling, and player usage, potentially leading to better team performance and fewer injuries.

### NFL Football: Play Calling and Defensive Adjustments

- **Scenario**: An NFL team wants to improve play-calling decisions during the game by dynamically adjusting based on the opposing team's formation and tendencies.
- **Approach**: Structured data includes real-time stats (down, distance, time left, score), formation data, historical play success rates, and player matchups, which are typically stored in databases. Unstructured data includes video feeds, playbook annotations, defensive alignments, and real-time comments from coaches and analysts about player performance or situational cues. Generative AI combines real-time game data with historical trends to suggest optimal play calls based on the current game scenario. For example, it might recommend a specific blitz when the opponent is in a known pass-heavy formation on third down, or suggest a specific offensive play based on the defensive alignment.
- **Business Decision**: The AI can help the coaching staff decide in real-time whether to run, pass, or kick, and what kind of formation or play to use based on factors like down and distance, current score, and opposing defensive tendencies. It can also suggest defensive adjustments, such as double-teaming a star receiver or adjusting coverage based on how an opposing quarterback is performing under pressure that day.

### Data query/analysis architecture approaches

- Traditional: Pulling numbers and text from documents (unstructured data) and putting it into a database (structured data) that includes other structured data and using SQL to query the data
  - Best when know what questions to ask
  - Returns consistent and accurate results
  - Use Azure AI Document Intelligence to pull numbers and text from documents and then use LLMs to convert from JSON to CSV to more easily add to a database
  - Can use LLMs to take English questions and convert to SQL queries on data (Microsoft Fabric AI Skill)
  - Cons: not using LLM's to improve answers to questions, just SQL queries; not accurately pulling data from documents
  - Ideal for financial reporting or operational dashboards where accuracy and consistency are critical
  - Document  $\rightarrow$  Database  $\rightarrow$  SQL Query  $\rightarrow$  Answer
  - Summarize: Best for accuracy and predefined queries
- GenAI: Feeding all types of data (structured, semi-structured, and unstructured) into an LLM and using a bot to ask questions on the data
  - Best when not sure what questions to ask
  - Can return inconsistent and inaccurate results
  - Bot can allow unstructured data (documents) and semi-structured data (csv files) to be uploaded
  - Structured data can be exported to a csv and uploaded, or bot can be made to pull structured data from database automatically
  - Cons: need to make sure csv data has a lot of metadata; not accurately pulling data from documents if using RAG
  - Useful for customer support bots or summarizing and analyzing large document repositories
  - Document/CSV/Database  $\rightarrow$  LLM  $\rightarrow$  Bot Query  $\rightarrow$  Answer
  - Summarize: Best for exploration and leveraging many types of data

# Helpful blogs

Microsoft Fabric Al Skill | James Serra's Blog

Introduction to OpenAI and LLMs | James Serra's Blog

Introduction to OpenAI and LLMs – Part 2 James Serra's Blog

Introduction to OpenAI and LLMs – Part 3 James Serra's Blog

Copilot in Microsoft Fabric | James Serra's Blog

# My book

	Deciphering Data Architectures 1st Edition, H
O'REILLY"	by James Serra (Author) Format: Kindle Edition
	5.0 $\bigstar \bigstar \bigstar \bigstar \bigstar \bigstar \bigstar \bigstar$ 3 ratings
Deciphering	#1 New Release ( in Data Modeling & Design
Data	Book description
Architectures Choosing Between a Modern Data Warehouse, Data Fabric, Data Lakehouse, and Data Mesh	Data fabric, data lakehouse, and data mesh have recently appeared as viable alternat These new architectures have solid benefits, but they're also surrounded by a lot of h book provides a guided tour of each architecture to help data professionals understa In the process, James Serra, big data and data warehousing solution architect at Micr
	architecture concepts, including how data warehouses have had to evolve to work wi data lakehouses can help you achieve, and how to distinguish data mesh hype from r determine the most appropriate data architecture for your needs. By reading this boo
	<ul> <li>Gain a working understanding of several data architectures</li> <li>Know the pros and cons of each approach</li> </ul>
James Serra	Distinguish data architecture theory from the reality
	Learn to pick the best architecture for your use case
	Inderstand the differences between data warehouses and data lakes

Roll over image to zoom in

#### **Kindle Edition** Û See all formats and editions Editorial reviews atives to the modern data warehouse. hyperbole and confusion. This practical tand its pros and cons. crosoft, examines common data vith data lake features. You'll learn what reality. Best of all, you'll be able to

ook, you'll:

- Understand the differences between data warehouses and data lakes
- Learn common data architecture concepts to help you build better solutions
- Alleviate confusion by clearly defining each data architecture
- Know what architectures to use for each cloud provider

#### Read entire book now with an O'Reilly subscription: Deciphering Data Architectures (O'Reilly.com)

- Foundation
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  - 2. Types of Data Architectures
  - 3. The Architecture Design Session
- Common Data Architecture Concepts
  - 4. The Relational Data Warehouse
  - 5. Data Lake
  - 6. Data Storage Solutions and Processes
  - 7. Approaches to Design
  - 8. Approaches to Data Modeling
  - 9. Approaches to Data Ingestion
- Data Architectures
  - 10. The Modern Data Warehouse
  - 11. Data Fabric
  - 12. Data Lakehouse
  - 13. Data Mesh Foundation
  - 14. Should You Adopt Data Mesh? Myths, Concerns, And The Future
- People, Process, and Technology
  - 15. People And Processes
  - 16. Technologies

#### More details



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