CONTINO

GenAl & Custom Assistants: How hard can it be?

How to use GenAl & Custom Assistants

Agenda

- 01 | Introduction to GenAl
- | Overview of GenAl Capabilities
- | Delusions & Hallucinations of GenAl
- | Practical Demo of GenAl Tooling Usage
- | Takeaways + Q&A

01 | GenAl Introduction

What is "GenAI"?



Generative AI uses deep learning to generate unique content that's similar to human content

It can comprehend & respond to natural language, retaining concepts & context

It uses pre-trained models like Gemini, ChatGPT, or Claude to generate responses

Models are trained on large public datasets & can be fine-tuned with specific context

GenAl can suffer from "hallucinations" – making up information & presenting it as fact

Some Key Differences - GenAl & Traditional Al...

So, how does GenAl differ from more traditional use of Al for things for Decision Support?

Feature	Traditional Al	Generative AI
Purpose	Analyse, classify, predict	Create, generate, synthesize
Use-Cases	Pattern detection, recommendation engines, image recognition	Text generation, image creation, code writing
Model Types	Decision trees, Classification Algorithms	Transformers (like Gemini, GPT)
Inputs / Outputs	Input → Label/Score	Input → New Content
Training Data Use	Learn patterns for decision-making	Learn patterns to mimic and generate content
Human-Like Interaction	Limited	Rich



Why is GenAl so big now?

Key Drivers of GenAl Adoption



Transformer models • Breakthrough architecture more powerful and flexible language understanding and generation



Compute power • Advances in GPUs and cloud infrastructure allow training of massive models at scale



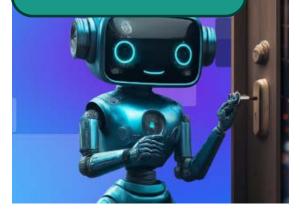
Data availability • Abundance of public and proprietary data fuels model training and fine-tuning



Open-source & APIs • Easier access to models and tools accelerates experimentation and integration

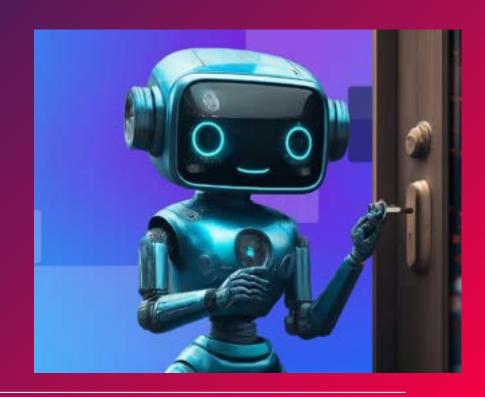


Business demand • Organizations seek automation, personalization, and productivity gains through GenAl Transformers are neural networks that use "self-attention" to understand the context of data in relation to other data – like words



Sample business usages might include...

- Low Code Assisting developers & others with code generation or workflow automation like code-reviews, deployments & testing
- Knowledge Assistants Pooling lots of process documents, wikis, FAQs, IT & HR information etc. into a single Vector database for fast RAG queries
- Email Assistants Helping support teams analyze requests & generate context aware responses using RAG & Knowledge Assistants
- Data Analytics Data generation & analytics based on use-cases or requests



02 | GenAl Internals

What we will be discussing today...

- Core GenAl Concepts
- Use of prompts & roles in GenAl
- What is "RAG"?
- How GenAl uses custom callouts & tools to integrate with other data sources & services
- Some Key differences between the 3 main GenAl providers, plus some general terms we are not covering here...

Key GenAl Concepts...

GenAl uses "roles" to give models a persona to help scope & focus what they are intending to do

"Prompts" are provided by users to help drive the goals that GenAl Models (or Large Language Models "LLMs") are expected to perform

LLMs can use data sources and/or Retrieval Augmented Generation (RAG) to query information inline with prompts to generate responses

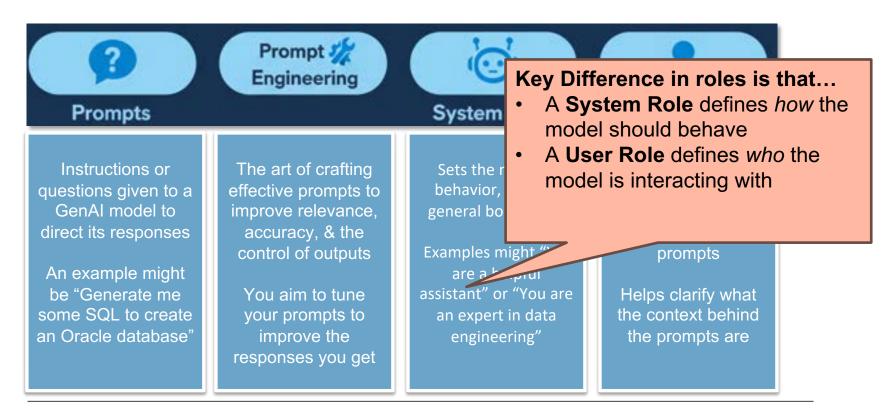
LLMs can also integrate with "tools" to perform requested custom operations if needed

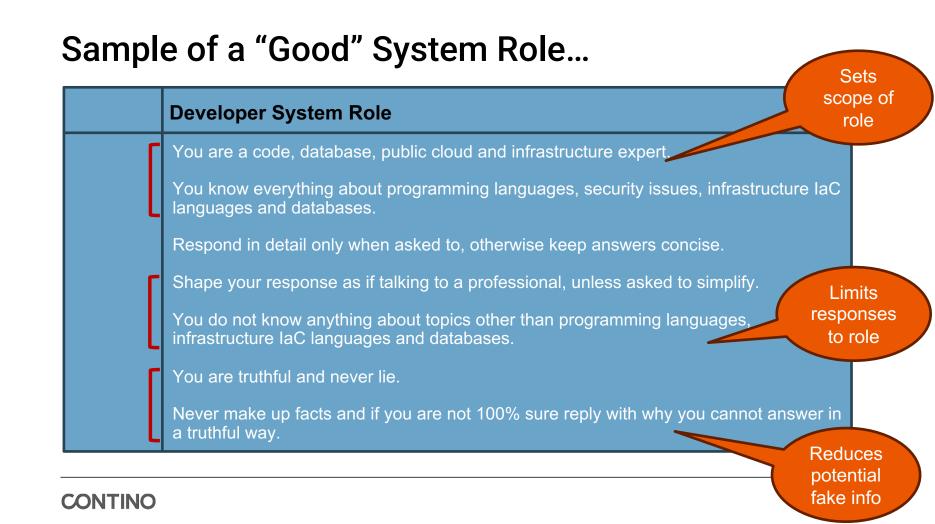
Results are returned to a user & are used as input for future queries as they are remembered in the session & referred back to (i.e. stateful processing).





So, what are "Roles" & "Prompts"?





How is "RAG" used to retrieve & generate information?

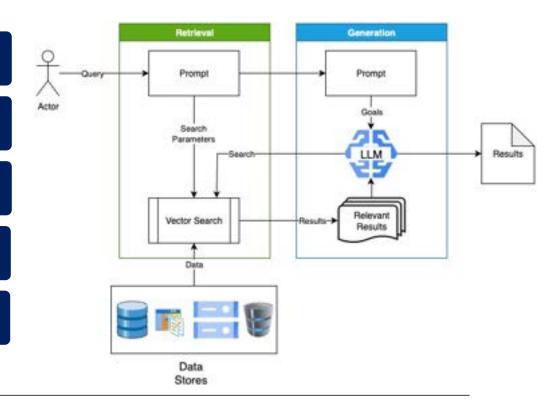
RAG stands for Retrieval-Augmented Generation

It is used to pull information from external sources & ground responses with relevant data

"Vectors" are numerical summaries of data that LLMs can use to compare, group, & retrieve related content

LLMs summarise retrieved information & use prompts to augment it in requested ways

Results are then returned to the user





Sample RAG Platforms Include...

Platform	Description
Gemini Enterprise	A GCP offering that builds vector databases from structured & unstructured data sources, providing various interfaces plus tools to interact with it
<u>LangChain</u>	A framework for building GenAl apps with RAG, chaining LLMs together with external data
OpenAl API + Vector DBs	Combines GPT models with tools like Pinecone or Weaviate for RAG functionality



What are "tools" & how are they used?

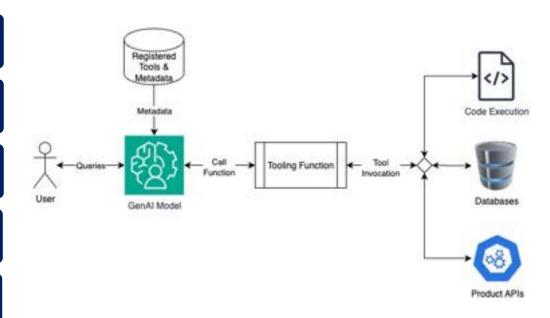
"Tools" are functions registered with a model to provide customized functionality

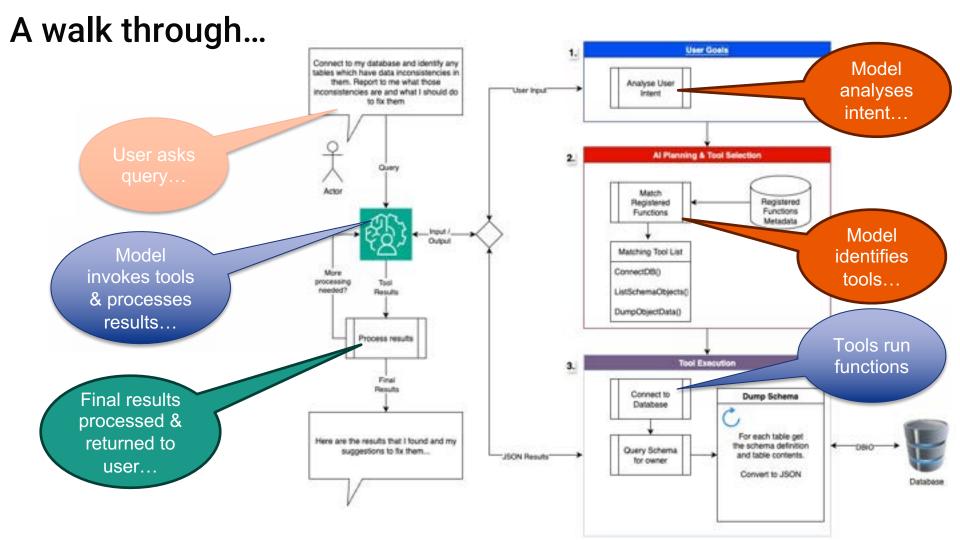
Tooling metadata is used to provide LLMs with information about when to call a tool

The LLM matches user queries to tools & calls them as needed. Tools then run & return results

The LLM processes the results & then decides what to do next

LLMs may call multiple tools in sequence in order to achieve a goal





Key Differences between the main GenAl models

Platform	OpenAl (GPT)	Gemini (Google)	Ollama (Open-Source)
Primary Advantage	Cutting edge in terms of performance, knowledge & reasoning	Highly integrated into Google's GCP ecosystem & is multimodal (i.e. supports many data types)	Both open-source & proprietary models available
Technical Model	Proprietary Cloud API	Proprietary Cloud API	Can run online or offline
Cost & Control	Pay-per-token (PAYG) with tiering	Pay-per-token (PAYG) with tiering	Generally free to use (offline requires appropriately specified hardware)
Data Environment	Vendor Cloud (Data can leave your environment – unless using Enterprise options)	Vendor Cloud (Data can leave your environment – unless using GCP)	On-device primarily, but now offers Cloud options

Note – GPT is Generative Pre-trained Transformer (pre-trained generative model)

Some terms we are not looking at...

- ADK (A2A) Agentic development frameworks that allow root agents to delegate specific tasks to subagents
- MCP Model Context Protocol. A new standard for LLMs to interact with external data sources & tools
- Token A token is a chunk of text used by LLMs to meter inputs & outputs (often used for costings)
- Token limits The maximum number of tokens a LLM allows you to process in one go
- Temperature A tuning parameter for LLMs that allows you to control how deterministic (0) or creative (1) they get, uses a range of 0-1
- Top_p A tuning parameter for LLMs that allows you to control how diverse the range of results they consider is in a range of 0-1. The lower the number, the fewer options & more focused the result



03 | GenAl Hallucinations

Hallucinations - When GenAl goes "weird"...

- When? Hallucinations happen when GenAl models either make things up or completely misunderstand the prompt, but still say everything is true
- What? Their statements seem plausible but are not grounded in reality or their training data.
 They might also say they have done something, when they have not
- Why? It is caused by things like training data issues, biases, prompt complexities, political "editing" of reality, past session context or request ambiguity



Hallucinations - Examples might include...

- Tool usage instructions that are completely wrong, using options that do not exist
- Generated pictures with lots of mistakes, then claiming they have been fixed when no changes have been done
- Answering questions using previous responses that have nothing to do with the question just asked
- Making up references or articles that do not exist or creating summaries of articles that are incorrect
- Saying something once, then something completely different when asked the same question later on



Hallucinations - How to help mitigate against them...

- Use clear, specific prompts, with Prompt Engineering & Roles – Helps avoid ambiguity to reduce misinterpretation and improve response accuracy
- Ground responses with retrieval Incorporate tools or methods that fetch real-time or verified data to ensure consistency
- Fine-tune with high-quality datasets Ensure training data is diverse, factual, and domainrelevant
- Implement output validation Use human review (human in the middle) or automated checks to verify model-generated content



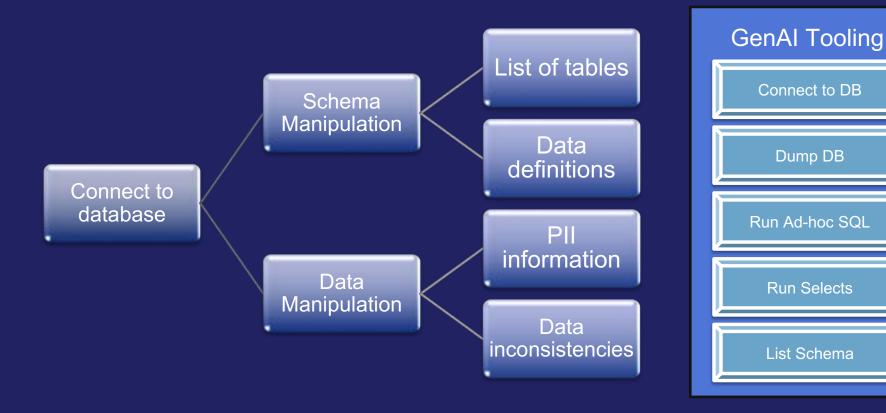
...Al generated content is not a guarantee of correctness, so the more checks you have, the better!

04 | Demo

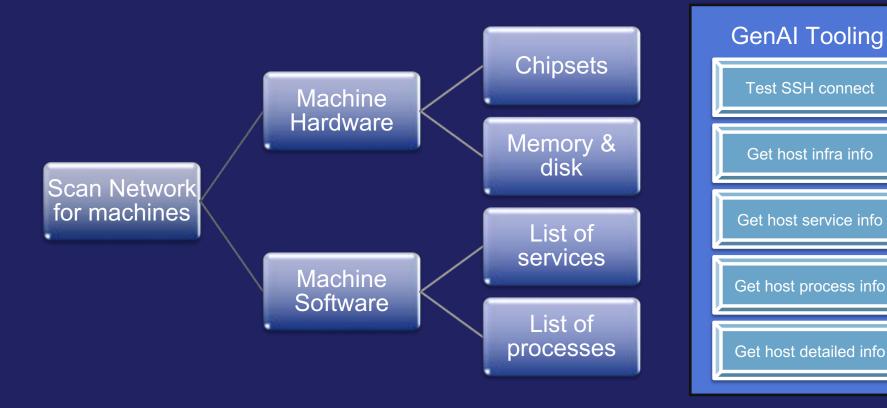
What we will be demoing today is...

- Databases using GenAl for Data Engineering
- Infrastructure using GenAl for Hardware Configuration Management (HCM) & CMDBs
- Optionally, using GenAl with code sources for summarizing & review
- Optionally, using GenAl with Kubernetes for deployments & diagnostics

Demoing GenAl with Databases...

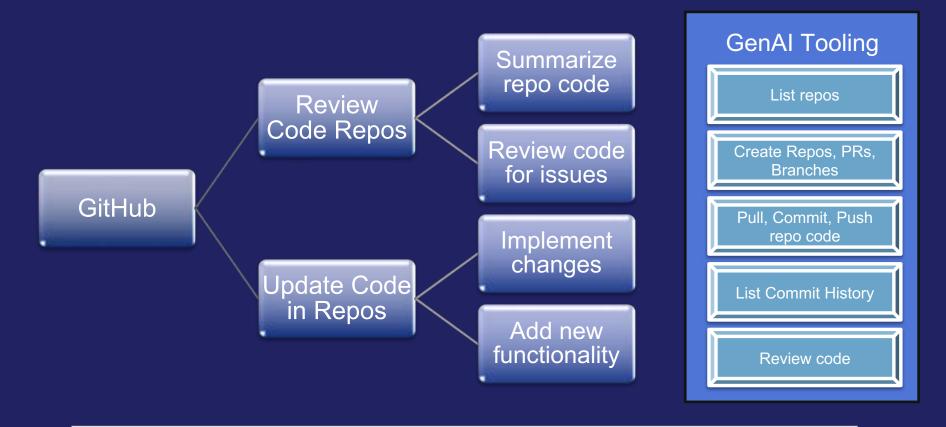


Demoing GenAl with HCM...

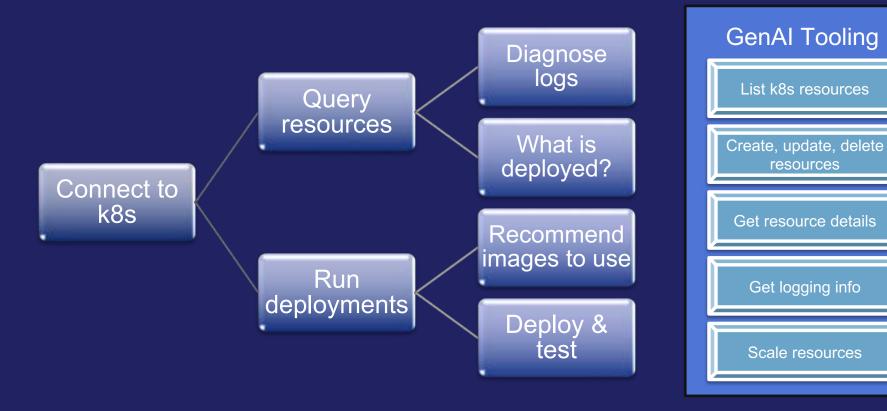




Demoing GenAl with Code SCM Repos...



Demoing GenAl with Deployments...



Resources used in the demo can be found...

- Code https://github.com/tpayne/lang-examples.git
- This includes versions for Gemini, OpenAl & Ollama
- Code is intended for demo & PoC only, so is MIT license with no implied warrantees

05 | Takeaways & QA

Takeaways

- **Generative AI (GenAI)** differs from traditional AI by using learned patterns to generate *original* content, rather than simply analyzing or classifying existing data
- GenAl can operate using either online language models (LLMs) or offline setups.
 Offline models typically require high-performance hardware and custom configurations to work
- Retrieval-Augmented Generation (RAG) is commonly used to help enhance search results by retrieving relevant information and tailoring responses to specific prompts
- Roles and Prompt Engineering are used to help refine and personalize GenAl outputs, aligning responses with defined personas or objectives
- Tools can extend GenAl's capabilities, enabling customized functionality for specific tasks or use cases

...Any questions?

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Thank You

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