## PGConf NYC

**Building GenAl Applications With Postgres and Gemini** 



### **Speakers**



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#### Where you can find me:

- LinkedIn: <a href="https://www.linkedin.com/in/cofin">https://www.linkedin.com/in/cofin</a>
- Github
  - https://github.com/cofin
  - <a href="https://github.com/litestar-org">https://github.com/litestar-org</a>



## Agenda

- Provide a introductory guide on using Gemini with Postgres
- How you can use agents today
  - Agents that enhance Postgres (Gemini CLI & MCP Toolbox)
  - Building agents from scratch with Postgres

The magic of GenAl captured the world's imagination

Today organizations are looking to turn that excitement into business results



Your Al needs to be fueled by your business data

Your business data needs to be unlocked by Al



85%

companies experimenting or in limited production with GenAl

5%

companies that have GenAl applications implemented at scale

However, existing data platforms don't meet the needs of the Al era

Source: Wavestone, 2024 Data and Al Leadership Executive Survey





## Gemini 2.5 Ecosystem



### What is Vertex AI?



#### Vertex Al

cloud.google.com/vertex-ai

- Managed, End-to-End AI & ML Platform on Google Cloud
- Model Garden Generative & Predictive Al
   Models from Google, Partners and Open Source
- Vertex Al Studio Experiment with Models
- Custom Models Training/Prediction Pipelines
- Vector Search for Embeddings
- Colab Enterprise for Jupyter Notebooks
- Enterprise-Grade Security/Reliability

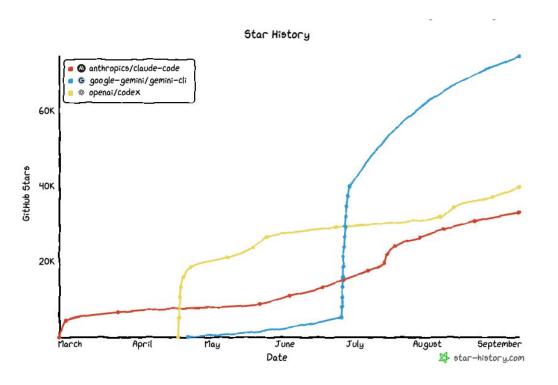
Gemini CLI is a Powerful Al agent that wraps your terminal

## What can I use it for?



- Coding
- Deep Research
- Task Automation
- Debugging
- Database Analysis
- Custom Workflows
- Extendable with slash commands and MCP integrations

## Gemini CLI is popular!

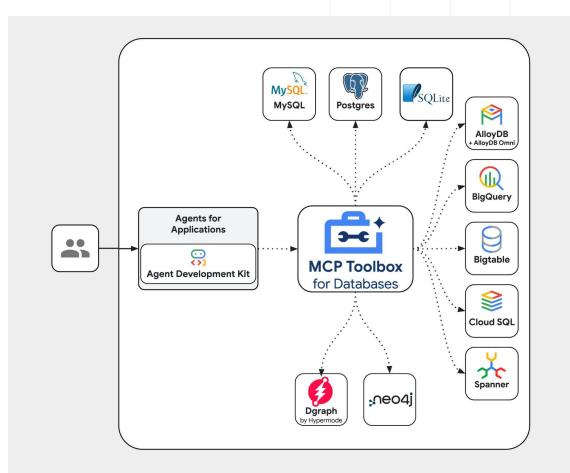


#### Open source Al Agent for

- Accelerated coding and debugging
- Streamlined development workflows
- Effortless data exploration and manipulation
- Content creation & summarization
- Gemini CLI is easily extensible!!

#### MCP Toolbox for Databases

- Standalone tooling to enable MCP integration for agentic services
- Supports multiple databases with a single tool
- Natively integrates as a Gemini CLI extension



**MCP Quickstart** 

## **Gemini CLI + PostgreSQL**

Goal: <u>Meet users where they are</u> and provide an easy & elegant agentic experience for <u>PostgreSQL</u> via <u>Gemini CLI</u>

#### **Postgres Extensions**

- postgres
- <u>cloud-sql-postgresql</u>
- <u>cloud-sql-postgresql-observability</u>
- alloydb
- <u>alloydb-observability</u>

# **Growing First Party Ecosystem**

- bigquery-conversational-analytics
- bigguery-data-analytics
- <u>cloud-sql-mysql</u>
- <u>cloud-sql-mysql-observability</u>
- <u>cloud-sal-salserver</u>
- <u>cloud-sql-sqlserver-observability</u>
- dataplex
- firestore-native
- looker
- Mysql
- spanner
- sql-server
- mcp-toolbox

## Third Party Utilities

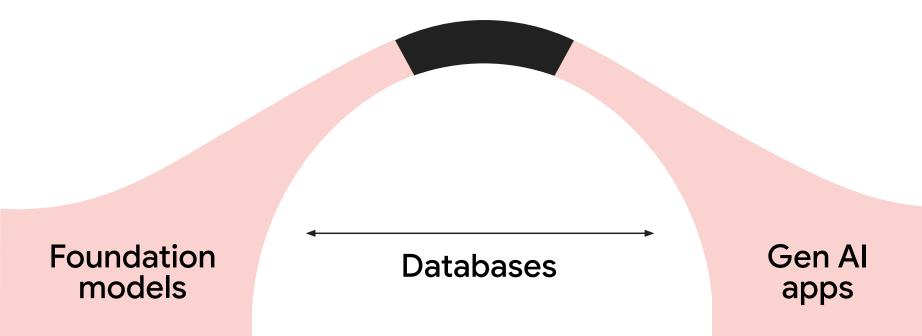
- Zen MCP Server
- Context7



## Let's see Gemini CLI in action

# Building an Application

# Operational data is key to exploiting the power of gen Al



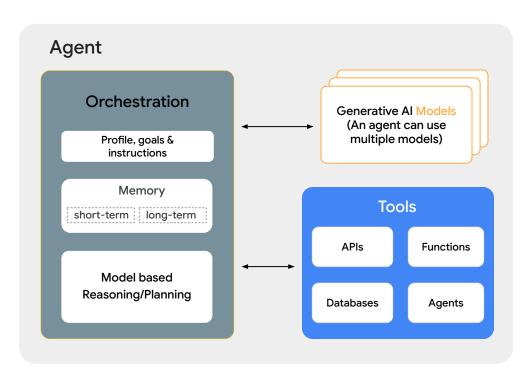
## Google Gen Al SDK

## pip install google-genai

```
from google import genai
# Gemini Developer API
client = genai.Client(api_key="YOUR_API_KEY")
# Vertex AI API
client = genai.Client(vertexai=True, project="your-project-id", location="us-central1")
response = client.models.generate_content(
   model="gemini-2.5-flash",
    prompt="How does AI work?"
print(response.text)
```

#### goo.gle/genai-sdk

### General agents architecture

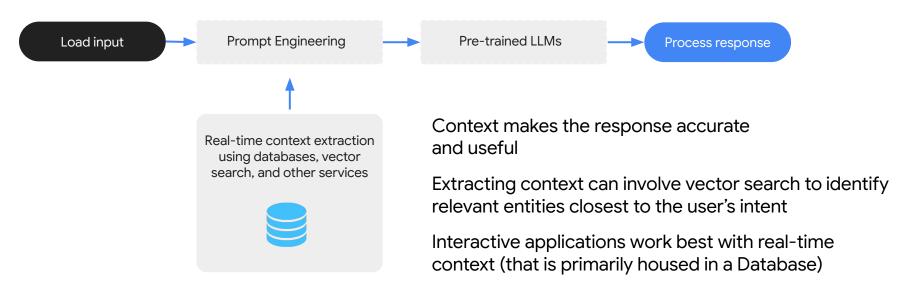


## Key components

- Model: Used to reason over goals, determine the plan and generate a response
- Tools: Fetch data, perform actions or transactions by calling other APIs or services
- Orchestration: Maintain memory and state (including the approach used to plan), tools, data provided/fetched, etc

## **Grounding an LLM with RAG**

Retrieval Augmented Generation





Numeric representations (vectors) of words and phrases that help define the way that the phrases relates to various concepts and other words.

# What are Embeddings?

#### **Traditional Search:**

"espresso" matches only "espresso"

#### **Embedding Search:**

"espresso" also finds:

- "strong coffee" (close in meaning)
- "concentrated brew" (similar concept)
- "intense flavor" (related characteristic)



#### async def create embedding(self, text: str) -> list[float]: """Create embeddings using Vertex AI.""" # Use the native Vertex AI embedding model model = TextEmbeddingModel.from pretrained(self.embedding model) embeddings = await model.get embeddings async([text]) if embeddings and len(embeddings) > 0: return cast("list[float]", embeddings[0].values) # Fallback to mock embedding for development except Exception: # Log the error and fallback to mock embedding logger.exception("Embedding generation failed, using fallback") return [0.0] \* 768 # Standard embedding dimension return [0.0] \* 768

## Populating Embeddings

```
- Yummy Waffles: [37.7749, -122.4194, ...]
- Pancakes: [0.2, -0.5, 0.8, ...]
```

```
1  UPDATE product
2  SET embedding = :embedding,
3     embedding_generated_on = SYSTIMESTAMP
4  WHERE id = :id
```



Try searching for coffee products using natural language:

Hi

**Search Products** 

407ms Total Time 375.5ms Embedding Generation 32.0ms Vector Search

#### Found 5 matching products:

## Vector **Similarity**

#### **Hazelnut Horizon**

A sweet and nutty latte with a hint of hazelnut.

35.5%

#### Java Journey

Java Journey - Medium

35.5%

#### **Hot Chocolate**

Rich chocolate drink made with premium cocoa and steamed milk

34.8%

#### **Midnight Espresso**

A shot of espresso with steamed milk, topped with a layer of foam.

34.4%

#### **Honey Bee Latte**

A sweet and creamy latte with a hint of honey.

34.3%





"I need something bold"
"Caffeine please"
"How's it going?"

# Intent **Exemplars**

```
of Intent Detection
DETECTION RESULTS
                                                       PRODUCT_SEARCH
Intent
Confidence
                                                                  79.3%
POSTGRESQL QUERY
  WITH query_embedding AS (
          SELECT intent, phrase,
              1 - (embedding <=> $1) AS similarity,
              confidence_threshold,
              usage_count
          FROM intent_exemplar)
  SELECT intent, phrase, similarity, confidence_threshold, usage_count
  FROM query_embedding
  WHERE similarity > $2
  ORDER BY similarity DESC
  LIMIT $3
```





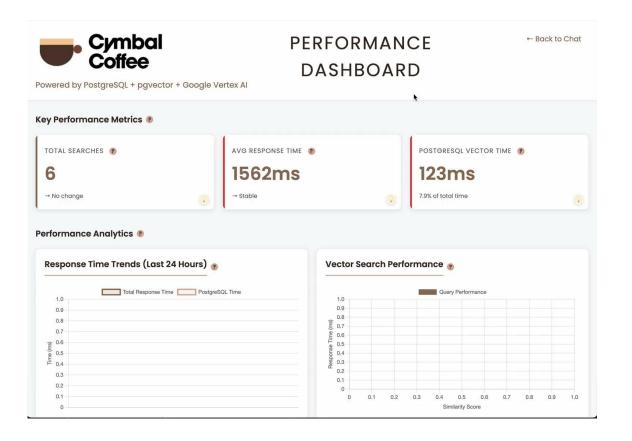
## Let's Demo!

## **Gemini CLI Demonstration**





## **Vector Search Demonstration**





## **Coffee Chat Demonstration**





#### Al Cookbook

#### Gemini Quickstarts



Intro to Gemini 2.0 Flash

Get started with Gemini 2.0 Flash in Vertex AI with the Gen AI Python SDK.

FUNCTION CALLING GEMINI
GROUNDING MULTIMODAL

PROMPTING

View on GitHub →



Get Started with the Multimodal Live API

Get started with Gemini 2.0 Multimodal Live API in Vertex AI using the Gen AI Python SDK

GEMINI LIVE API MULTIMODAL

View on GitHub →



Intro to Gemini 2.0 Pro

Get started with Gemini 2.0 Pro in Vertex AI with the Gen AI Pvthon SDK.

FUNCTION CALLING GEMINI

GROUNDING MULTIMODAL

PROMPTING

View on GitHub →



Get Started with Gemini 2.0 Flash Thinking

Get started with Gemini 2.0 Flash Thinking in Vertex AI using the Gen AI Python SDK to get more detailed reasoning and thinking steps.

GEMINI MULTIMODAL

View on GitHub →



(%)

View on GitHub →

Intro to Prompt Engineering

Intro to Gemini 2.0 Flash-Lite

Get started with Gemini 2.0

FUNCTION CALLING GEMINI

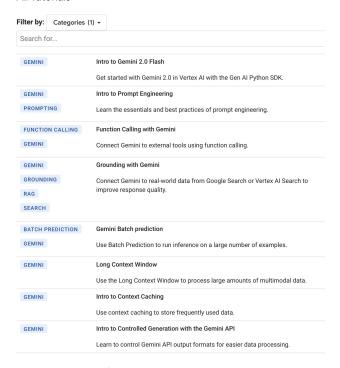
Flash-Lite in Vertex AI with the Gen AI Python SDK.

Learn the essentials and best practices of prompt engineering.

GEMINI PROMPTING

View on GitHub →

#### All Tutorials





## Google Cloud Generative Al GitHub

	Description
gemini/	Discover Gemini through starter notebooks, use cases, function calling, sample apps, and more.
search/	Use this folder if you're interested in using <u>Vertex Al Search</u> , a Google-managed solution to help you rapidly build search engines for websites and across enterprise data. (Formerly known as Enterprise Search on Generative Al App Builder)
rag- grounding/	Use this folder for information on Retrieval Augmented Generation (RAG) and Grounding with Vertex Al. This is an index of notebooks and samples across other directories focused on this topic.
conversation/	Use this folder if you're interested in using <u>Vertex Al Conversation</u> , a Google-managed solution to help you rapidly build chat bots for websites and across enterprise data. (Formerly known as Chat Apps on Generative Al App Builder)
language/	Use this folder if you're interested in building your own solutions from scratch using Google's language foundation models (Vertex AI PaLM API).
vision/	Use this folder if you're interested in building your own solutions from scratch using features from Imagen on Vertex AI (Vertex AI Imagen API). These are the features that Imagen on Vertex AI offers:  Image generation  Image editing  Visual captioning  Visual question answering
<b>Q</b> audio/	Use this folder if you're interested in building your own solutions from scratch using features from Chirp, a version of Google's Universal Speech Model (USM) on Vertex AI (Vertex AI Chirp API).



# Thank you

github.com/google-gemini/gemini-cli

github.com/cofin/postgres-vertexai-demo



