

About me

You can find me as @mdelapenya everywhere



Manuel de la Peña

Staff Software Engineer @ Docker Computer Science degree, Master in Software Engineering

- Testcontainers Go maintainer since 2020
- Engineering Productivity at Elastic Observability
- QA Tech lead at Liferay Cloud
- Core Engineer at Liferay
- In OSS since 2011
- Hitting keyboards since 1994
- First time in Singapore!!! 🥌



What we are going to see today:

- 01. GenAl in today's software
- 02. A mental model for testing: Cloud apps
- 03. Gen Al Tooling in Go
- 04. Let's talk about tests, b-Al-by!
- 05. Conclusions





1. GenAl in today's software

GenAl in Today's software

FOMO: Fear Of Missing Out!

Every day there is a new company offering AI services, exposing their models for you to consume them, and new papers are published every day.

- → OpenAl
- → Google
- → Anthropic
- → Meta
- → ...



The M/L + AI + Data (MAD) landscape

Sources:

https://www.linkedin.com/pulse/ai-landscape-2024-trends-top-startups-leta-capital-orape

https://mad.firstmark.com

2011 logos

In 2024

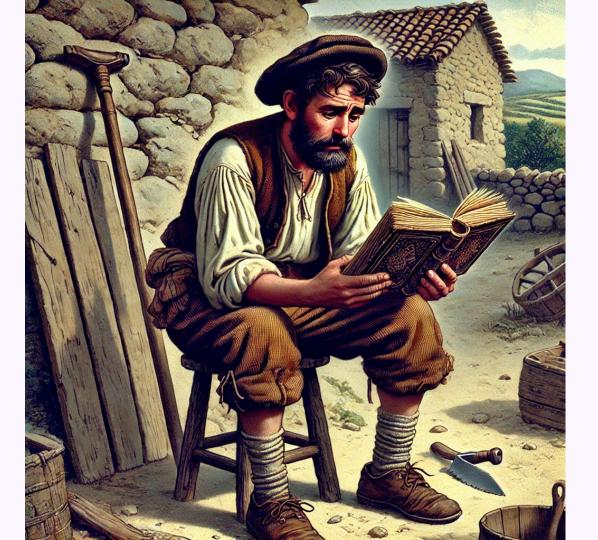
1416

In 2023

578

Newcomers in 2024









Develop with LLMs

Langchain (Python, Node), Langchain4j, SpringAl (Java), **Liamaindex**, and many more tools:

- Allow you to talk to LLMs
- Design prompts
- Create chats, tools and agents
- Talk to Vector databases

Depending on the model you talk to, you can use it for:

- Image recognition
- Text to text generation
- Text to image/video/audio
- Multimodal generation
- ...more in 3,2,1



LLMs



LLMs SDKs







A GenAl application





2. A mental model for testing: Cloud apps

Develop with the Cloud

How it works

- Our company uses a given Cloud provider
- We setup that Cloud's SDKs into your project
- We configure the credentials
- We start coding...

Seems pretty similar to the LLM approach, doesn't it?

But how do you test these applications?



GCloud, AWS, Azure



Cloud SDKs in Go







Your application



Testing Cloud applications

- No tests, my code is perfect!
- Local Service emulating a given Cloud service
- Test environment in the Cloud provider
 - Per team?
 - Per developer?
 - Shared across the company?
 - How long does it take to have them?
 - Do you prune outdated resources?
 - Do you measure costs?

- Do you know Localstack, Google Cloud and Azurite emulators?
 - Standalone applications, or
 - Docker containers.



Test Environments



Cloud SDKs in Go



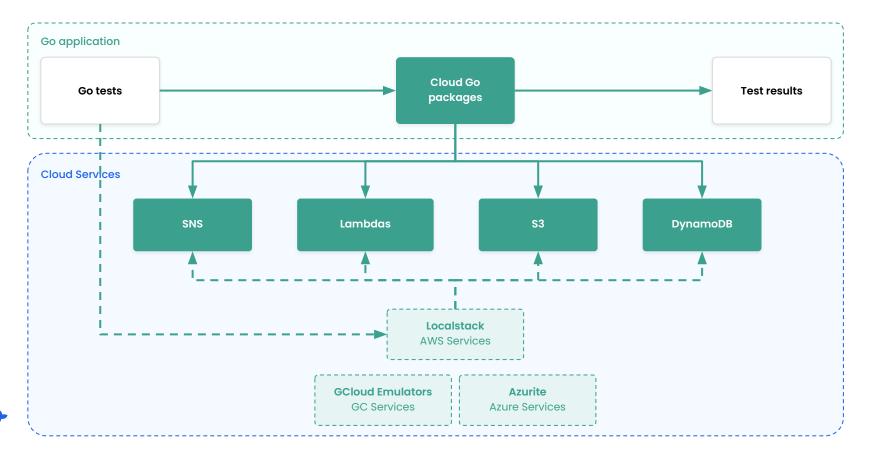








Testing Cloud applications with emulators







3. Gen Al Tooling in Go

Langchaingo

Go implementation for Langchain: https://github.com/tmc/langchaingo

Community driven project, led by Travis Cline.

- → Generate completions from an LLM (OpenAl, Anthropic, Google...)
- → Calculate embeddings for words, texts, images...
- → Talk to Vector databases to look for similar/relevant documents to augment LLM responses (Retrieval Augmented Generation)
 - Chroma, Milvus, pgVector, Pinecone, Qdrant, Weaviate...



langchaingo: completions

Create a completion from an LLM, using a streaming function so that the answer is printed at the moment it's produced by the LLM.

It comes with APIs to abstract the LLM creation, obtaining it from multiple providers: Google, OpenAI, Mistral, LlamaFile, Ollama:

 The completion code would be exactly the same.

https://github.com/mdelapenya/ge nerative-ai-with-testcontainers/tre e/main/02-streamin/main.go



```
// llm is llama3.2:3b
ctx := context.Background()
completion, err := llms.GenerateFromSinglePrompt(
  ctx, 11m, "Give me a detailed and long explanation of why
Testcontainers for Go is great",
  llms.WithTemperature(0.8),
  11ms.WithStreamingFunc(func(ctx context.Context, chunk []byte)
error {
   fmt.Print(string(chunk))
    return nil
 }),
if err != nil {
  log.Fatal(err)
```



langchaingo: embeddings

Using the right model, you can generate the embeddings for a text.

Embeddings are dense numerical representation (n-dimensional vectors) of texts, that can be used to calculate similarity between them.

https://github.com/mdelapenya/generative-ai-with-testcontainers/tree/main/06-embeddings/main.go

```
// llm is all-minilm:22m
embedder, err := embeddings.NewEmbedder(11m)
if err != nil {
  return fmt.Errorf("embedder new: %w", err)
docs := []string{
  "Testcontainers is a Go package that provides lightweight,
throwaway instances of common databases, web browsers, or anything
else that can run in a Docker container".
  "Docker is a platform designed to help developers build, share,
and run container applications.",
vecs, err := embedder.EmbedDocuments(context.Background(), docs)
if err != nil {
  log.Fatal("embed query", err)
```



langchaingo: RAG

Retrieval and Augmented Generation.

It's possible to pass a vector of embeddings to a vector database, and leverage the power of these systems to obtain relevant documents to enrich the response from the LLM.

https://github.com/mdelapenya/ge nerative-ai-with-testcontainers/tre e/main/07-rag/main.go



// llm is all-minilm:22m embedder, err := embeddings.NewEmbedder(11m) if err != nil { log.Fatalf("embedder new: %w", err) store, err := weaviate.NewStore(context.Background(), embedder) if err != nil { return fmt.Errorf("weaviate new store: %w", err) // ingest relevant documents in the store if err := ingestion(store); err != nil { log.Fatalf("ingestion: %w", err) // similarity search relevantDocs, err := store.SimilaritySearch(context.Background(), "What is my favorite sport?", 1, optionsVector...) if err != nil { log.Fatalf("similarity search: %w", err)

Ollama

Inference Engine: https://github.com/ollama/ollama

Like Docker, but for running models! Wrapper of Llama.cpp written in Go

- → ollama pull \$MODEL
- → ollama run \$MODEL
- → It has Modelfiles for customising models (temperature, system messages...)
- → It has a native app for Linux, Mac and Windows...
- → ... or you can run it as a Docker container
 - used in the demos, thanks to https://github.com/mdelapenya/dockerize-ollama-models/
 - https://hub.docker.com/u/mdelapenya



Ollama has native access to the GPUs of the host, which is key for speed.

Testcontainers Go

An Open Source Go package (MIT license) providing developer-friendly API's on top of the Docker engine.

https://github.com/testcontainers/testcontainers <u>-qo</u>

- Start, stop, terminate containers and networks
- Wait for containers on custom conditions
- Lifecycle hooks to inject custom code (Pre/Post)
- Copy files to/from containers
- Garbage collection of Docker resources



Go package: docker/docker



Go package: testcontainers-go











Testcontainers Go: modules

Go packages providing access to the most used technologies:

- Relational DBs: Mysql, Postgres, ...
- Vector DBs: Weaviate, Chroma, Qdrant,
 Milvus...
- Non Relational DBs: Elasticsearch, Redis, MongoDB, Neo4j, Opensearch...
- Cloud Emulators: Localstack, Google Cloud, Azurite
- Inference Engines: Ollama
- Keycloak, OpenFGA, Vault...
- 60 different Go modules!
- Convenient API specific to each module.

https://www.testcontainers.com/modules



Go package: testcontainers-go



Go packages: testcontainers-go/modules



Your Go app



Remember Ollama?

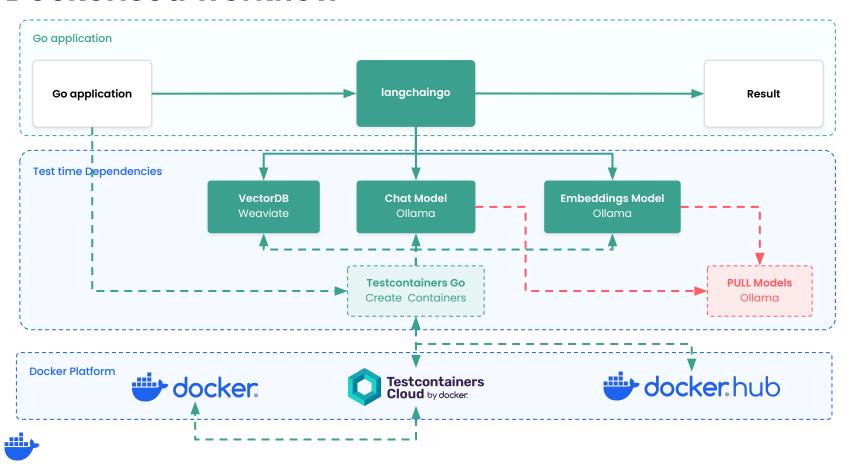
A module exists! https://testcontainers.com/modules/ollama/?language=go

Since v0.35.0, it's possible to interact with the local Ollama process as it was a container, honoring the create/start/stop/terminate container lifecycle.

```
ollamaContainer, err := ollama.Run(ctx, "ollama/ollama:0.5.4")
if err != nil {
    log.Printf("failed to start container: %s", err)
    return
}
code, reader, err := ollamaContainer.Exec(ctx, []string{"ollama", "pull", "all-minilm"})
```



Dockerised workflow





4. Let's talk about tests, b-Al-by!

The application

- → An application talking to two models:
 - Raw calls to the model
 - Calls to the same model using RAG
- → It uses langchaingo, ollama and testcontainers-go
- → How can we verify that non-deterministic LLM responses are correct?

https://github.com/mdelapenya/generative-ai-with-testcontainers/blob/main/08-testing



Round 1: strings COMPCIFISON https://github.com/mdelapenya/generative-ai-with-testcontainers/blob/main/08-testing



Round 2: cosine similarity

https://github.com/mdelapenya/generative-ai-with-testcontainers/blob/main/08-testing

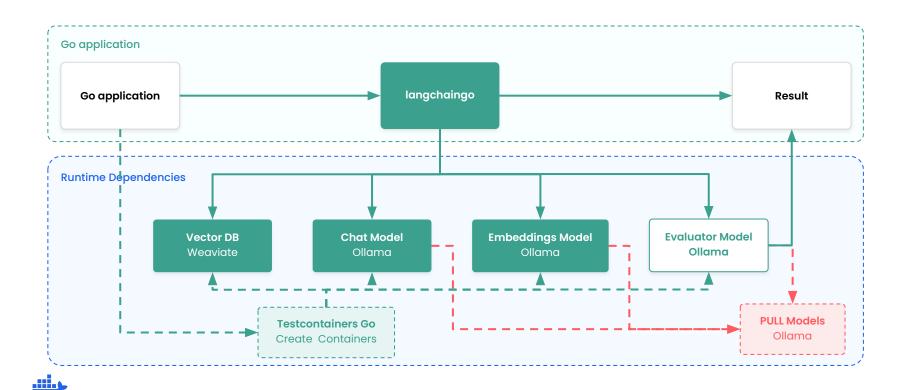


Enter Evaluators

- → AKA "LLM-as-a-Judge" (https://eugeneyan.com/writing/llm-evaluators/).
- → Evaluate the quality of another LLM's response to an instruction or query.
- → Define a very strict **System** Prompt:
 - Provide Instructions: e.g. response format
 - Provide reference examples
- → Define a very strict User Prompt:
 - Provide a detailed format: ### question ### answer ### reference ###.
 - Provide a reference (e.g. in the test as an expectation)
 - Structured output, semantic/style constraints
 - Respond with "yes" or "no" including the reasoning.



Adding an Evaluator



Final Round: using an Evaluator

https://github.com/mdelapenya/generative-ai-with-testcontainers/blob/main/08-testing





5. Conclusions

Conclusions

- → Langchaingo: contribute!
- → **Testcontainers Go** + **Ollama**: a really powerful and easy-to-use combo for local development experience.
- → Using **Evaluators** (models with a very strict system prompt) helps us in identifying if the model our application is using responds correctly:
 - Helps us tuning up our application: e.g. choosing a different model, a different vector store, or even modifying the metrics used to classify/correlate the responses at test time.
- → Different models can produce different responses:
 - E.g. Ollama + Llama3.2:3b can excel in one task, but its response could be different than using OpenAI + o4 (220b???).
- → Integration tests will give you <u>enough confidence so you can make progress</u> with speed, but you still need to test against the real thing, e.g. with OpenAI.
 - Run lots of integration tests but you still need to run some E2E tests against the real thing!









Resources

https://github.com/mdelapenya/generative-ai-with-testcontainers https://github.com/testcontainers/testcontainers-go https://github.com/testcontainers/workshop-go

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