Building and deploying LLM applications with **Apache Airflow**

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Agenda

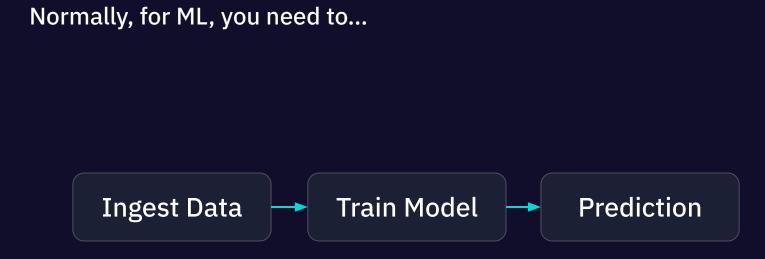
Why Airflow should be at the centre of LLMOps? Real Use-case & reference architecture Next Steps: Community collaboration



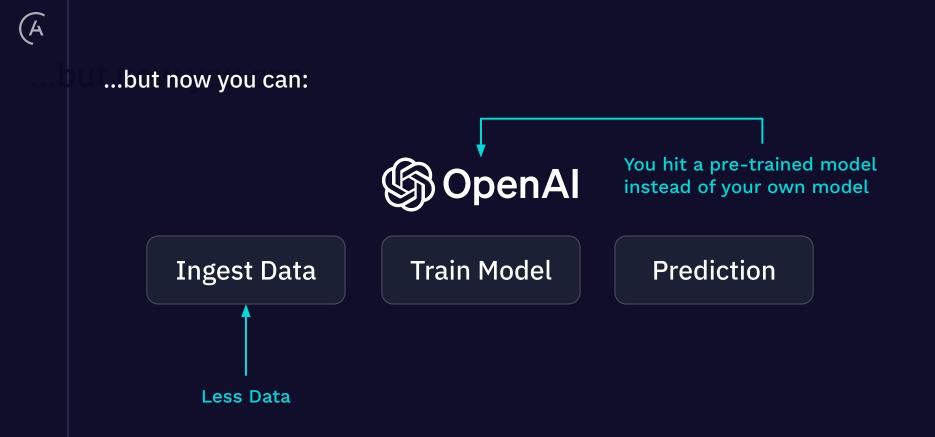
Generative AI: A Creative New World

A powerful new class of large language models is making it possible for machines to write, code, draw, and create with credible and sometimes superhuman results.











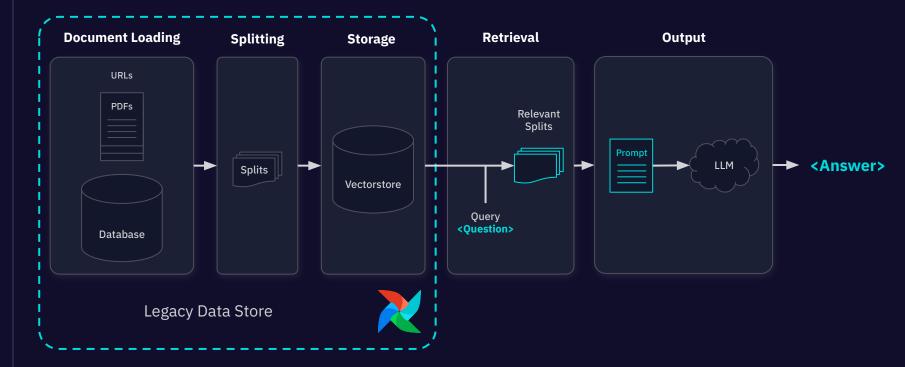


Going from "Idea to Production" with LLM Apps involves solving a lot of data engineering problems:

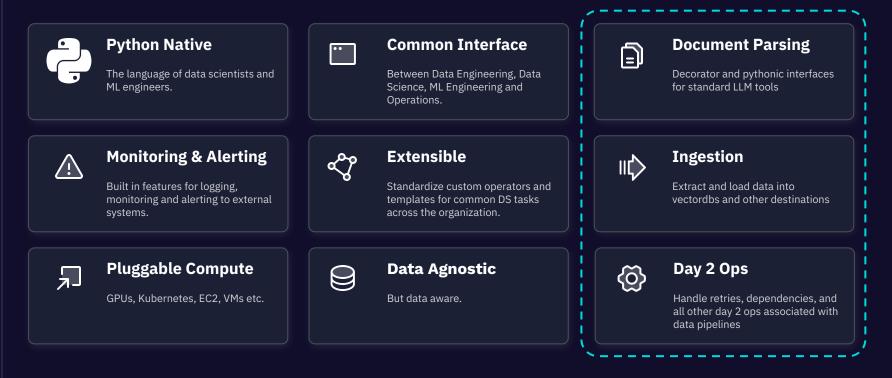
- Ingestion from several sources
- Day 2 operations on data pipelines
- Data preparation
- Data privacy
- Data freshness
- Model deployment & monitoring
- Scaling Models
- Experimentation & fine-tuning
- Feedback Loops



Typical Architecture for Q&A use-case using LLM



Airflow is a Natural Fit...





Let's Talk About a Real Use Case



Problem Statement:

We have customers, employees, and community members that ask questions about our product with answers that exist across several sources of documentation.

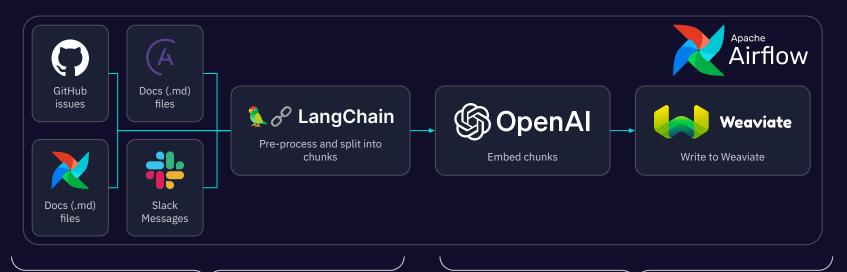
How do we provide an easy interface for folks to get their questions answered without adding further strain to the team?



Ask Astro

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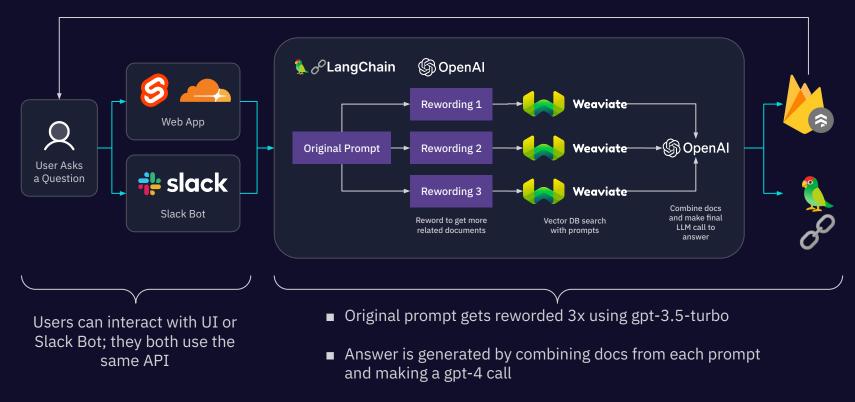
Data Ingestion, Processing, and Embedding



- Airflow gives a framework to load data from APIs & other sources into LangChain
- LangChain helps pre-process and split documents into smaller chunks depending on content type

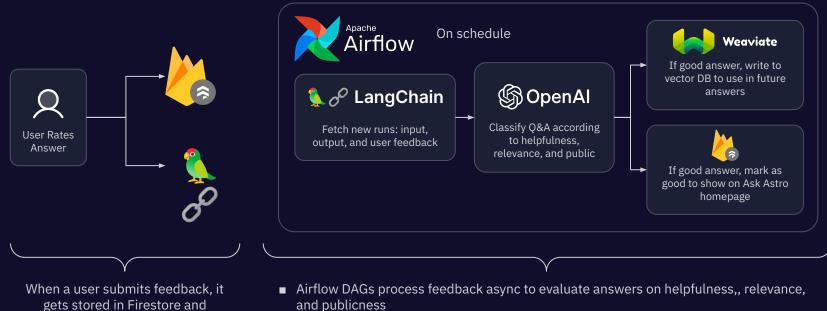
- After content is split into chunks, each chunk is **embedded into vectors** (semantic representations)
- Those vectors are written to Weaviate for later retrieval

Prompt Orchestration and Answering



 State is stored in Firestore and prompt tracing is done through LangSmith
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LLM & Product Feedback Loops



- LangSmith for later use
 - If answer is good, it gets stored in Weaviate and can be used as a source for future questions
 - UI also shows the most recent good prompts on the homepage

Running this in production meant:

- Experimenting with different sources of data to ingest
- Running the pipelines on *a schedule and ad-hoc*
- Running the same workloads with variable chunking strategies
- Needing to *retry tasks* due to finicky python libraries and unreliable external services
- Giving different parts of the workload *variable compute*
- Creating standard interfaces to *interact with external* systems



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Which is what Airflow's great at!



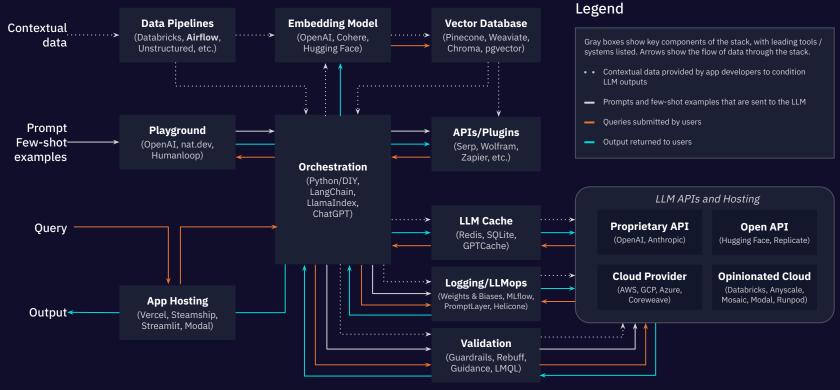
ask.astronomer.io

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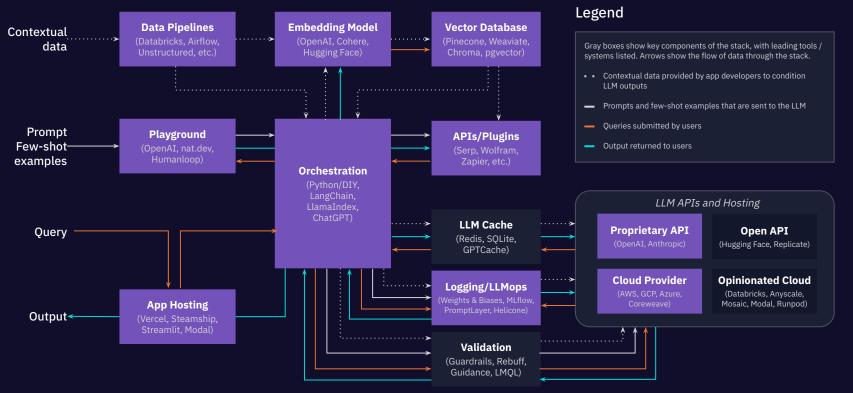
github.com/astronomer/ask-astro



a16z's Emerging LLM App Stack



AskAstro has a few parts of this...



...but there's even more to consider.

Data Governance

- How do you account for private data?
- How do you provide transparency into data lineage?

Fine Tuning

- Does it improve results?
- How much does it cost?

Feedback Loops

- Semantic cache for correct responses
- Ranking sources based on accuracy and ranking accordingly
- Prompt clustering what are people asking?

Airflow is foundational to best practices for all of this.



Thanks to the AskAstro Team:





Philippe Gagnon

Michael Gregory

Community Collaboration

Providers

Interfaces

Patterns and Use Cases



What are all the providers the ecosystem needs?





What's the interface that feels right for LLMOps?

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```
create_embeddings = OpenAIEmbeddingOperator(
   task_id="create_embeddings",
   conn_id="openai_prod",
   source_data="/usr/local/airflow/dags/github.pdf",
   output_file="/usr/local/airflow/data/embeddings.txt",
   model="text-embedding-ada-002",
   encoding="cl100k_base",
```

tore_embeddings = WeaviateOperator(task_id="check_schema", conn_id="weaviate_prod", embeddings="/usr/local/airflow/data/embeddings.txt",

•••

```
nOperator(
    task_id="openai_task",
    embedding="OpenAI",
    source_dataset=Dataset("/usr/local/airflow/dags/data/github.pdf"),
    target_dataset=Index(uri="pgvector://postgres", name="airflow_summit_test"),
    embedding_params={
        "embedder_model": "text-embedding-ada-002",
        "encoding_name": "cl100k_base"
},
```



What's the interface that feels right for LLMOps?

•••

```
sk
```

def generate_and_store_embedding(data_path):
 import os

from langchain.document_loaders import PyPDFLoader
from langchain.embeddings import OpenAIEmbeddings
from langchain.text_splitter import CharacterTextSplitter
from langchain.vectorstores import Chroma

```
documents = text_splitter.split_documents(pdf_docs)
Chroma.from_documents(documents=documents, embedding=OpenAIEmbeddings())
```

Patterns

What are the best practices for building pipelines for LLM Apps?

- Do you use one task to ingest and write?
- Can you use dynamic task mapping to break it out?
- Do you write to disk?
- Can you store embedding values in XCOMs?
- How do you reconcile Airflow orchestration with prompt orchestration?





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