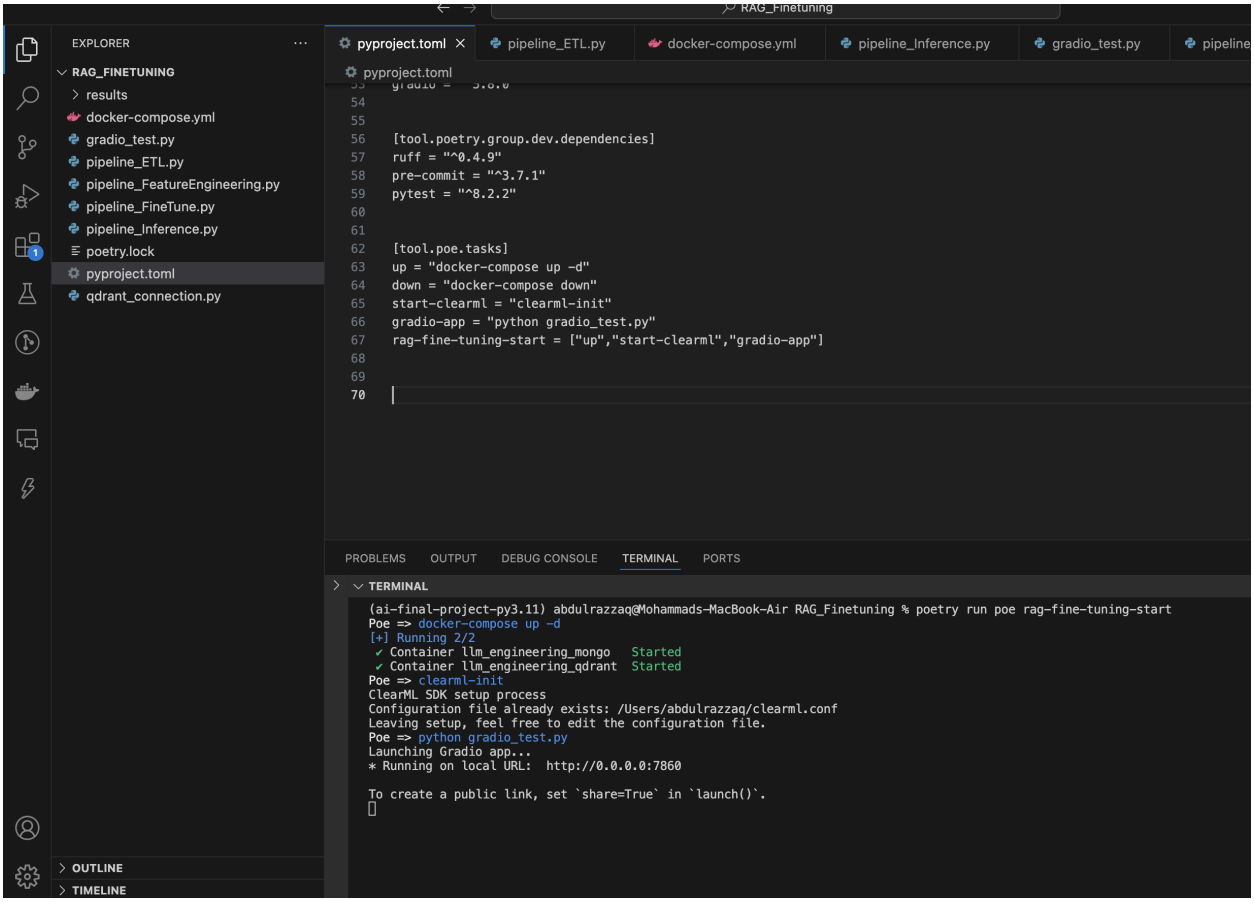
Finetuned RAG Systems Engineering Report

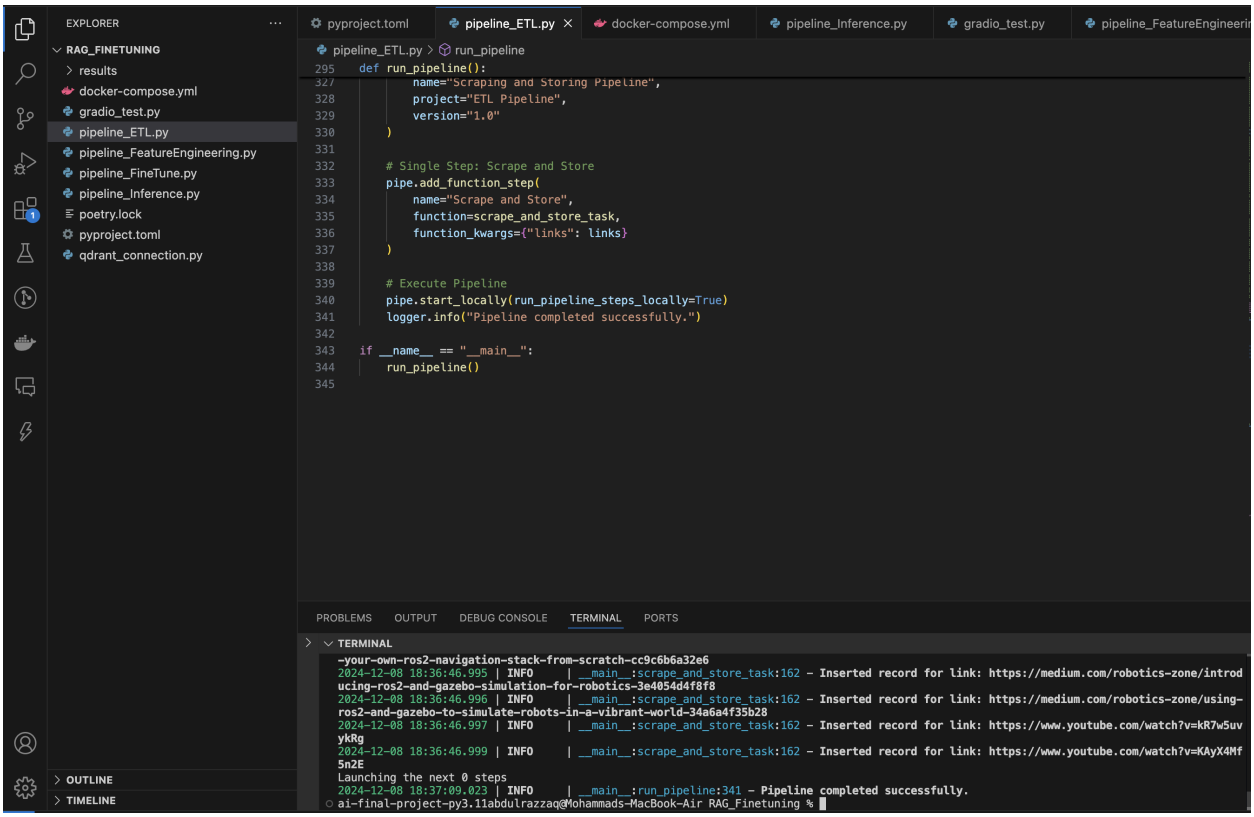
Step 1: Infrastucture

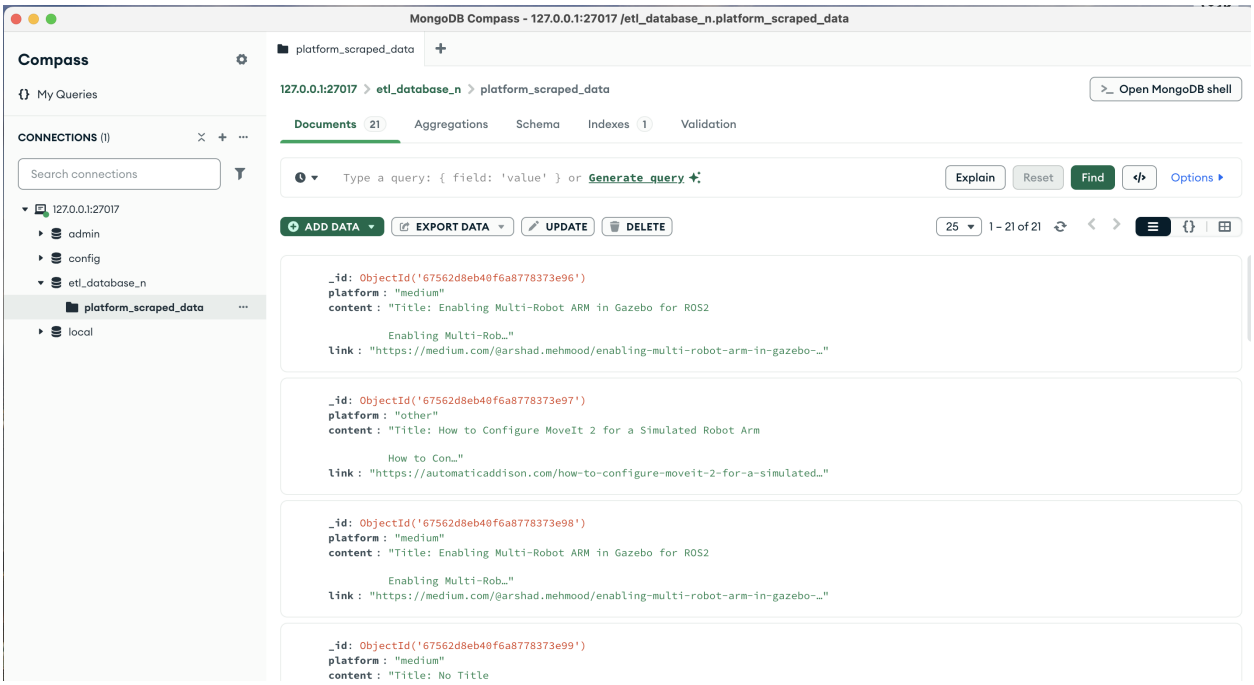
- **Goal**: Set up your environment, including Qdrant for vector storage, MongoDB for raw data storage, Gradio App , ClearML.
- **Completed Tasks**:
    - Infrastructure appears ready, including Qdrant and MongoDB.



Step 2: ETL Pipeline

- **Goal**: Collect high-quality Q&A data, store it in MongoDB, and prepare it for feature engineering.
- **Completed Tasks**:
  - Blogs and other data sources like Medium articles have been ingested.
  - Data is formatted into a Q&A structure.

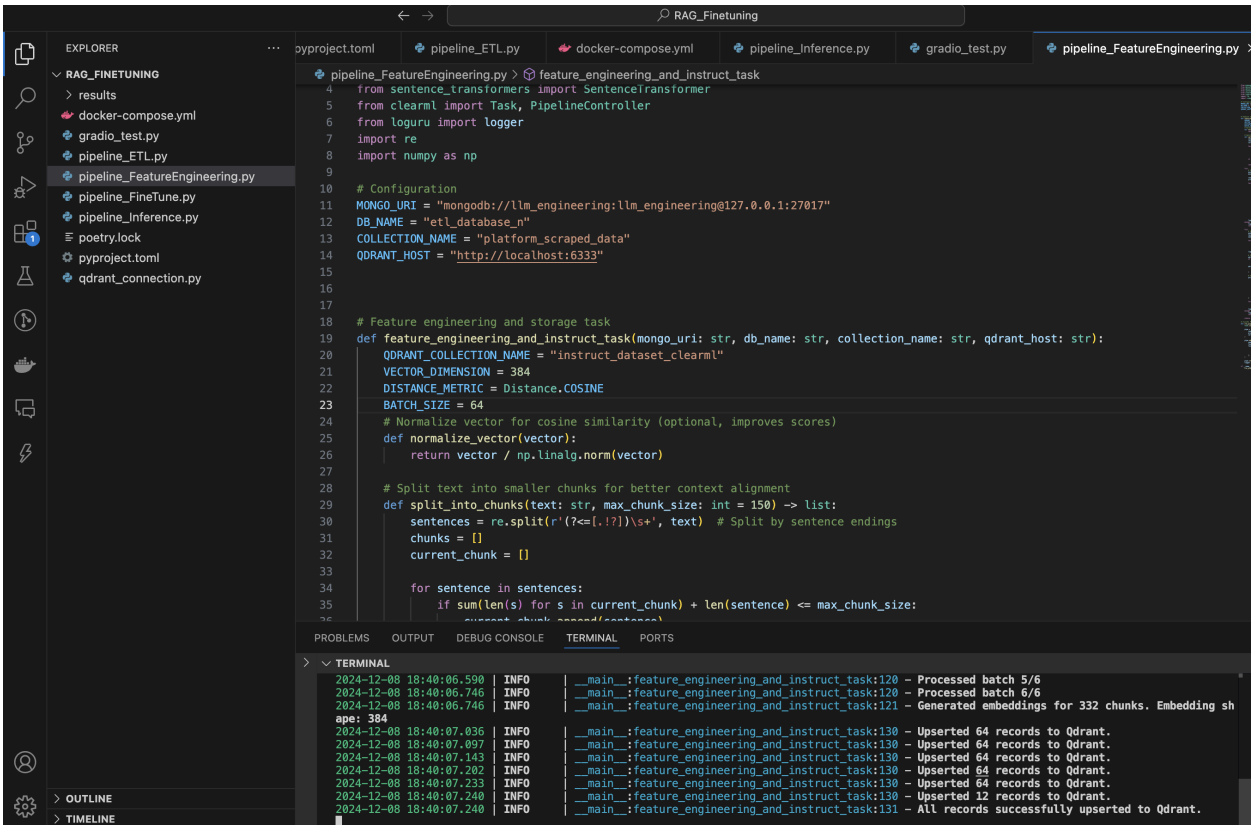


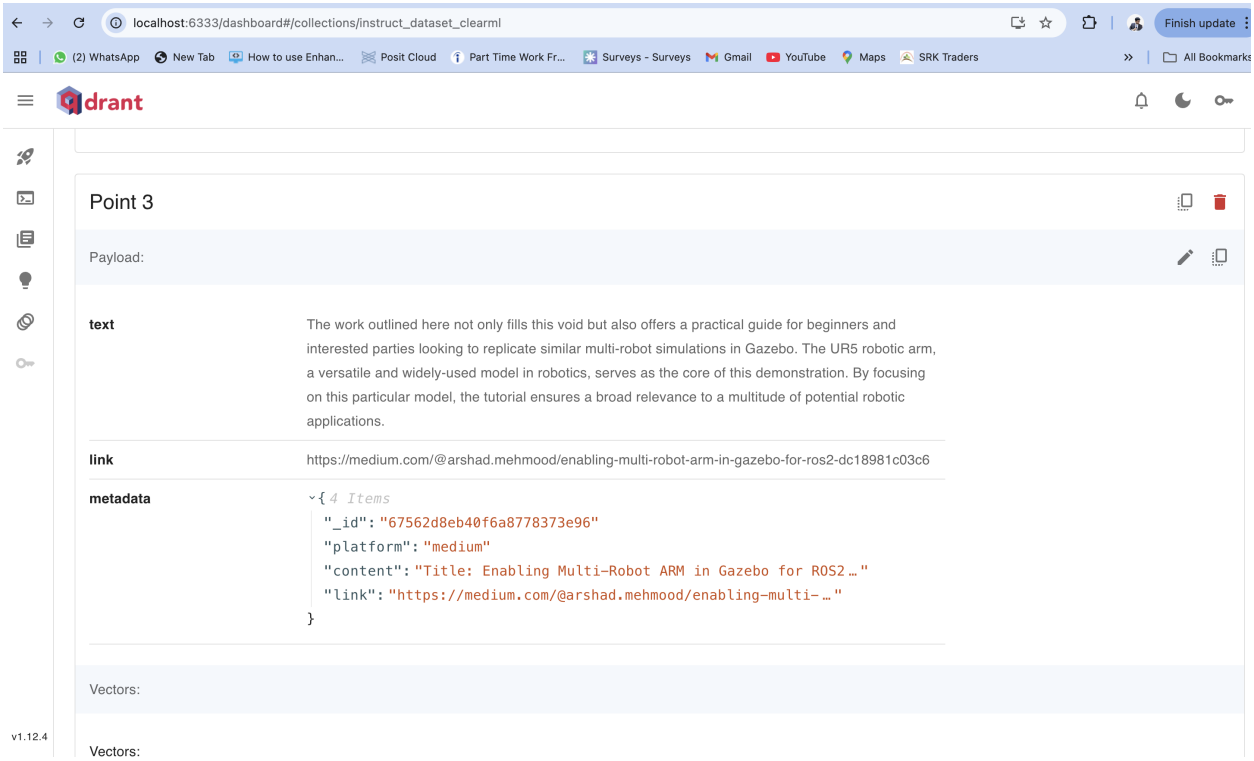


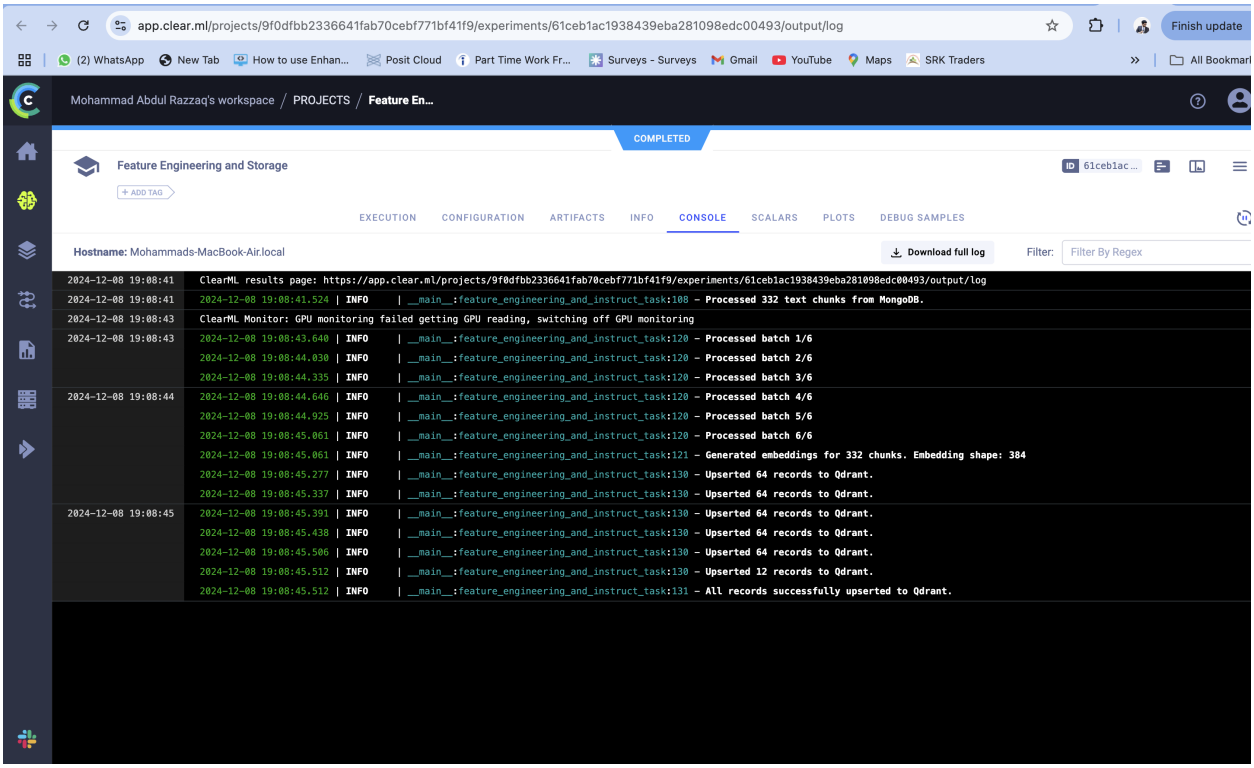
Added question-and-answer style blog posts to Medium to create a Q&A dataset

https://abdulrazzaq0902.medium.com/introduction-to-ros2-robot-operating-system-9fa9e9367263

Step 3: Feature Engineering

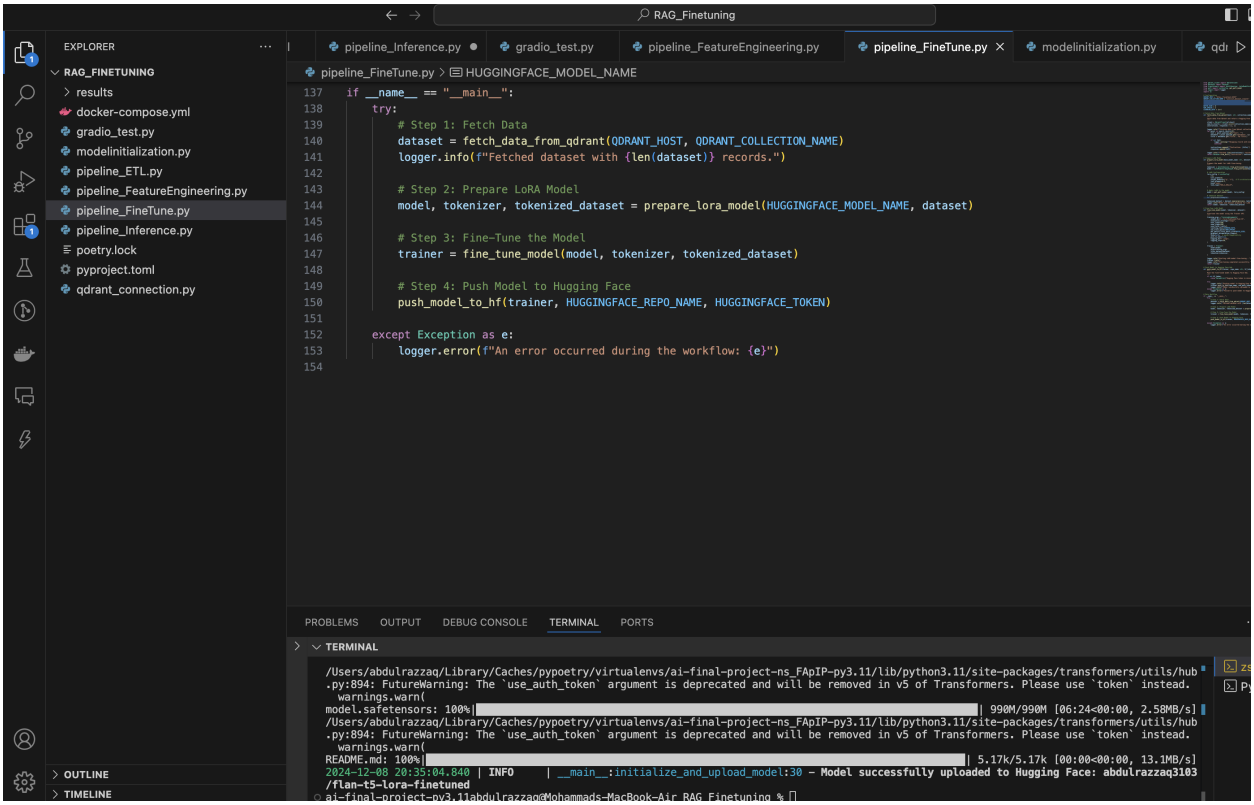

RAG_Finetuning
EXPLORER
RAG_FINETUNING
results
docker-compose.yml
gradio_test.py
pipeline_ETL.py
pipeline_FeatureEngineering.py
pipeline_FineTune.py
pipeline_Inference.py
poetry.lock
pyproject.toml
qdrant_connection.py
OUTLINE
TIMELINE
pyproject.toml
pipeline_ETL.py
docker-compose.yml
pipeline_Inference.py
gradio_test.py
pipeline_FeatureEngineering.py
pipeline_FeatureEngineering.py > feature_engineering_and_instruct_task
from sentence_transformers import SentenceTransformer
from clearml import Task, PipelineController
from loguru import logger
import re
import numpy as np
# Configuration
MONGO_URI = "mongodb://llm_engineering:llm_engineering@127.0.0.1:27017"
DB_NAME = "etl_database_n"
COLLECTION_NAME = "platform_scraped_data"
QDRANT_HOST = "http://localhost:6333"
# Feature engineering and storage task
def feature_engineering_and_instruct_task(mongo_uri: str, db_name: str, collection_name: str, qdrant_host: str):
QDRANT_COLLECTION_NAME = "instruct_dataset_clearml"
VECTOR_DIMENSION = 384
DISTANCE_METRIC = Distance.COSINE
BATCH_SIZE = 64
# Normalize vector for cosine similarity (optional, improves scores)
def normalize_vector(vector):
return vector / np.linalg.norm(vector)
# Split text into smaller chunks for better context alignment
def split_into_chunks(text: str, max_chunk_size: int = 150) -> list:
sentences = re.split(r'(?<=[.!?])\s+', text) # Split by sentence endings
chunks = []
current_chunk = []
for sentence in sentences:
if sum(len(s) for s in current_chunk) + len(sentence) <= max_chunk_size:
PROBLEMS
OUTPUT
DEBUG CONSOLE
TERMINAL
PORTS
TERMINAL
2024-12-08 18:40:06.590 | INFO | __main__:feature_engineering_and_instruct_task:120 - Processed batch 5/6
2024-12-08 18:40:06.746 | INFO | __main__:feature_engineering_and_instruct_task:120 - Processed batch 6/6
2024-12-08 18:40:06.746 | INFO | __main__:feature_engineering_and_instruct_task:121 - Generated embeddings for 332 chunks. Embedding shape: 384
2024-12-08 18:40:07.036 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 64 records to Qdrant.
2024-12-08 18:40:07.097 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 64 records to Qdrant.
2024-12-08 18:40:07.143 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 64 records to Qdrant.
2024-12-08 18:40:07.202 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 64 records to Qdrant.
2024-12-08 18:40:07.233 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 64 records to Qdrant.
2024-12-08 18:40:07.240 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 12 records to Qdrant.
2024-12-08 18:40:07.240 | INFO | __main__:feature_engineering_and_instruct_task:131 - All records successfully upserted to Qdrant.


localhost:6333/dashboard#/collections/instruct_dataset_clearml
Finish update
(2) WhatsApp
New Tab
How to use Enhan...
Posit Cloud
Part Time Work Fr...
Surveys - Surveys
Gmail
YouTube
Maps
SRK Traders
All Bookmarks
Qdrant
Point 3
Payload:
text
The work outlined here not only fills this void but also offers a practical guide for beginners and interested parties looking to replicate similar multi-robot simulations in Gazebo. The UR5 robotic arm, a versatile and widely-used model in robotics, serves as the core of this demonstration. By focusing on this particular model, the tutorial ensures a broad relevance to a multitude of potential robotic applications.
link
https://medium.com/@arshad.mehmood/enabling-multi-robot-arm-in-gazebo-for-ros2-dc18981c03c6
metadata
{ 4 Items
"_id": "67562d8eb40f6a8778373e96"
"platform": "medium"
"content": "Title: Enabling Multi-Robot ARM in Gazebo for ROS2 …"
"link": "https://medium.com/@arshad.mehmood/enabling-multi- …"
}
Vectors:
v1.12.4
Vectors:

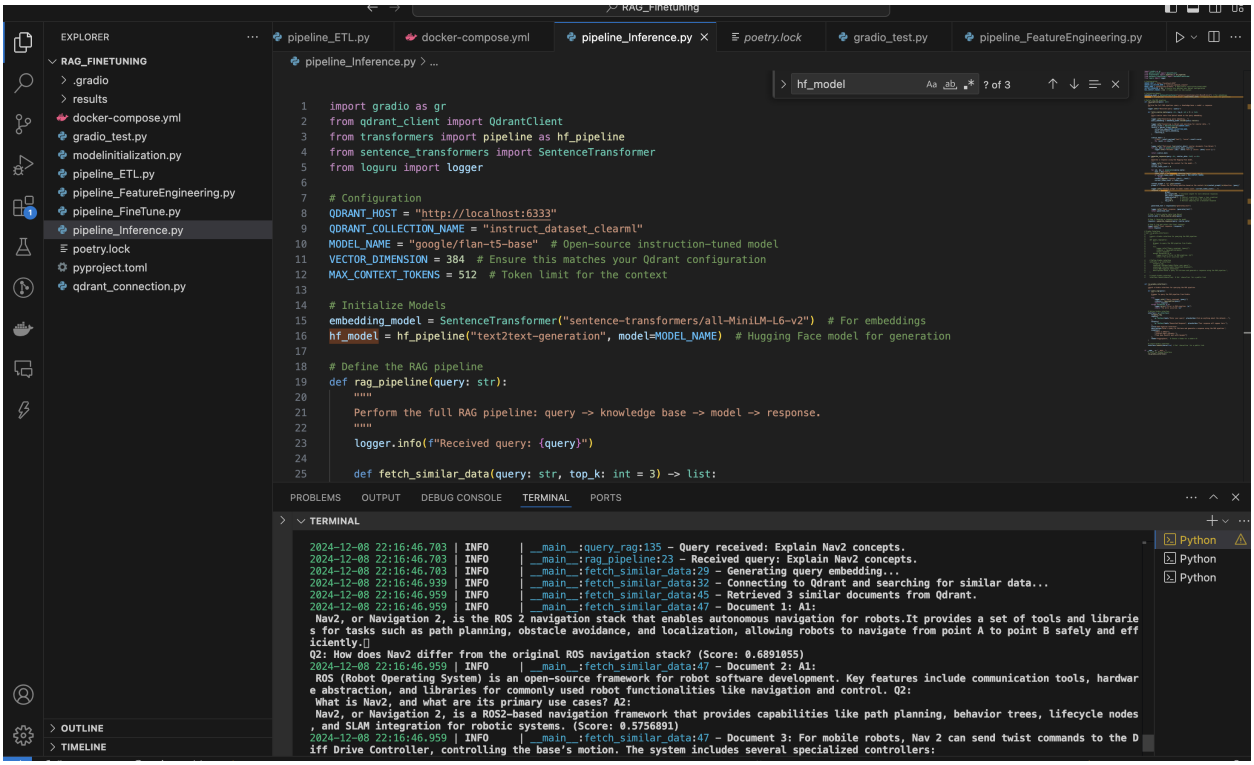


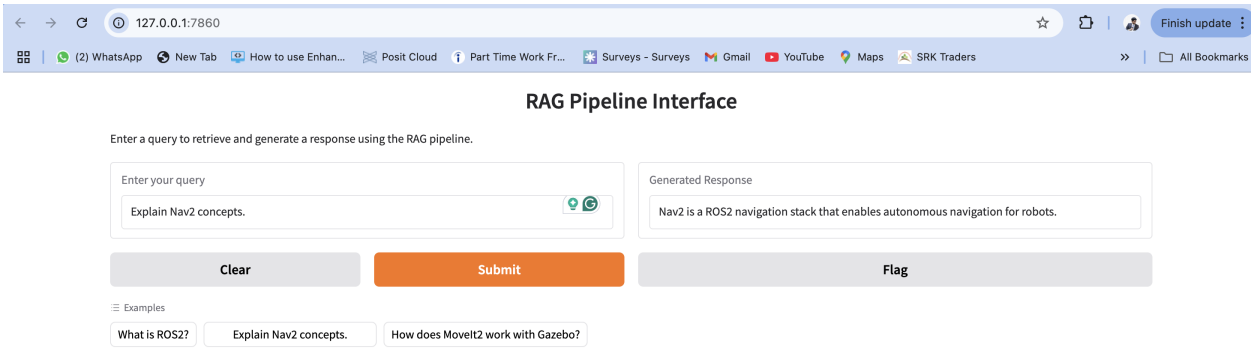
- **Goal**: Extract and preprocess features, such as embeddings, for use in Qdrant.
- **Completed Tasks**:
  - Sentence embeddings generated using `all-MiniLM-L6-v2`.
  - Data has been chunked and upserted into Qdrant.

Step 4: Fine-Tuning



- **Goal**: Fine-tune the `Flan-T5` model using LoRA to adapt it to your specific Q&A dataset.
- **Completed Tasks**:
  - Initial fine-tuning setup with LoRA and Hugging Face libraries.





Step 5: Contextual Retrieval and Inference

- **Goal**: Implement the end-to-end RAG pipeline for context-based response generation.
- **Completed Tasks**:
  - Query embeddings are generated, and Qdrant retrieves relevant context.

- Prompt structure integrates context with the user query.