

Finetuned RAG Systems Engineering Report

Step 1: Infrastructure

- **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.

```
pyproject.toml
[tool.poetry.group.dev.dependencies]
ruff = "^0.4.9"
pre-commit = "^3.7.1"
pytest = "^8.2.2"

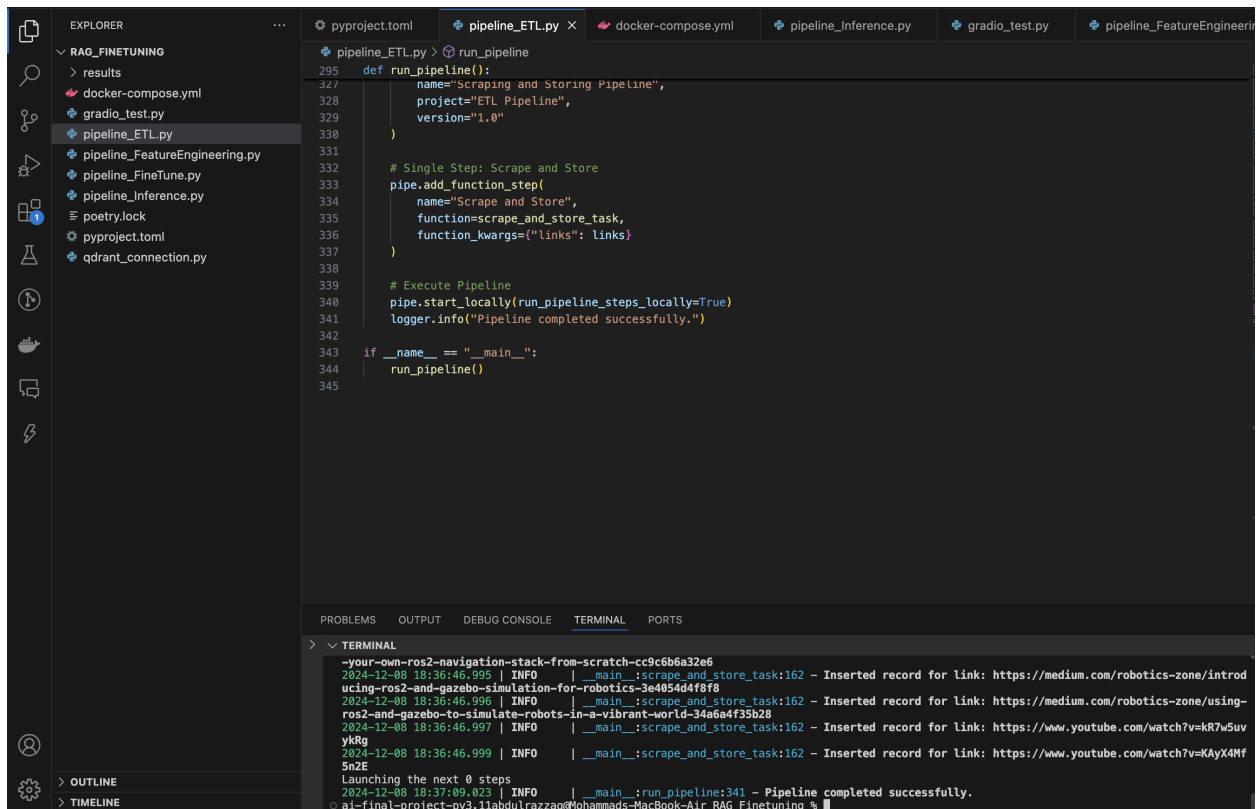
[tool.poe.tasks]
up = "docker-compose up -d"
down = "docker-compose down"
start-clearml = "clearml-init"
gradio-app = "python gradio_test.py"
rag-fine-tuning-start = ["up", "start-clearml", "gradio-app"]

TERMINAL
(ai-final-project-py3.11) abdulrazzaq@Mohammads-MacBook-Air RAG_Finetuning % poetry run poe rag-fine-tuning-start
Poe => docker-compose up -d
[+] Running 2/2
✓ Container llm_engineering_mongo Started
✓ Container llm_engineering_qdrant Started
Poe => clearml-init
ClearML SDK setup process
Configuration file already exists: /Users/abdulrazzaq/clearml.conf
Leaving setup, feel free to edit the configuration file.
Poe => python gradio_test.py
Launching Gradio app...
* Running on local URL: http://0.0.0.0:7860

To create a public link, set 'share=True' in 'launch()'.
```

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.



The screenshot displays a code editor with a file explorer on the left and a terminal at the bottom. The file explorer shows a project structure for 'RAQ_FINETUNING' with files like 'docker-compose.yml', 'gradio_test.py', 'pipeline_ETL.py', 'pipeline_FeatureEngineering.py', 'pipeline_FineTune.py', 'pipeline_Inference.py', 'poetry.lock', 'pyproject.toml', and 'qdrant_connection.py'. The 'pipeline_ETL.py' file is open in the editor, showing a function 'run_pipeline()' that defines a pipeline step for scraping and storing data. The terminal output shows the execution of this pipeline, with logs indicating the insertion of records for various links into a MongoDB database. The logs include timestamps, log levels (INFO), and the specific links being processed.

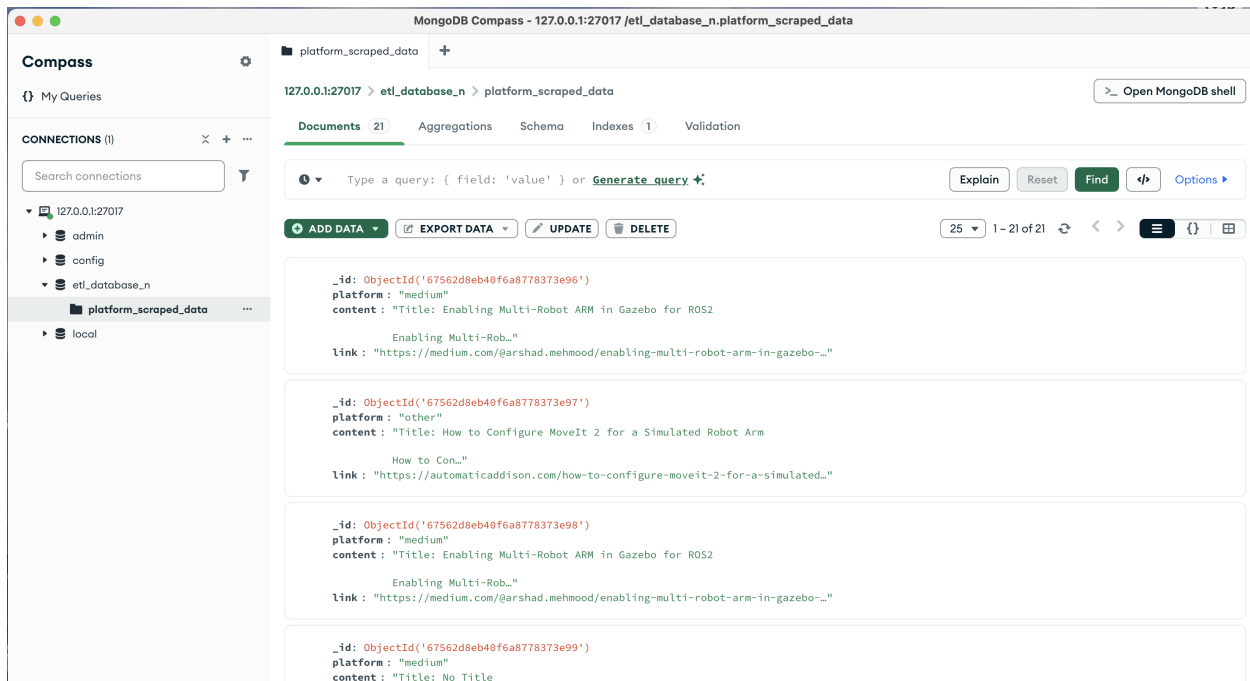
```
def run_pipeline():
    name="Scraping and Storing Pipeline",
    project="ETL Pipeline",
    version="1.0"

    # Single Step: Scrape and Store
    pipe.add_function_step(
        name="Scrape and Store",
        function=scrape_and_store_task,
        function_kwargs={"links": links}
    )

    # Execute Pipeline
    pipe.start_locally(run_pipeline_steps_locally=True)
    logger.info("Pipeline completed successfully.")

if __name__ == "__main__":
    run_pipeline()
```

```
2024-12-08 18:36:46.995 | INFO | __main__:scrape_and_store_task:162 - Inserted record for link: https://medium.com/robotics-zone/introducting-ros2-and-gazebo-simulation-for-robotics-3e4054d4f8f8
2024-12-08 18:36:46.996 | INFO | __main__:scrape_and_store_task:162 - Inserted record for link: https://medium.com/robotics-zone/using-ros2-and-gazebo-to-simulate-robots-in-a-vibrant-world-34a6a4f35b28
2024-12-08 18:36:46.997 | INFO | __main__:scrape_and_store_task:162 - Inserted record for link: https://www.youtube.com/watch?v=KR7wSuvykRg
2024-12-08 18:36:46.999 | INFO | __main__:scrape_and_store_task:162 - Inserted record for link: https://www.youtube.com/watch?v=KAYX4Mf5nZE
Launching the next 0 steps
2024-12-08 18:37:09.023 | INFO | __main__:run_pipeline:341 - Pipeline completed successfully.
ai-final-project-py3.11abduLraZZaQ@Mohammads-MacBook-Air RAQ_Finetuning %
```



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

```
4 from sentence_transformers import SentenceTransformer
5 from clearml import Task, PipelineController
6 from loguru import logger
7 import re
8 import numpy as np
9
10 # Configuration
11 MONGO_URI = "mongodb://llm_engineering:llm_engineering@127.0.0.1:27017"
12 DB_NAME = "etl_database_n"
13 COLLECTION_NAME = "platform_scraped_data"
14 QDRANT_HOST = "http://localhost:6333"
15
16
17
18 # Feature engineering and storage task
19 def feature_engineering_and_instruct_task(mongo_uri: str, db_name: str, collection_name: str, qdrant_host: str):
20     QDRANT_COLLECTION_NAME = "instruct_dataset_clearml"
21     VECTOR_DIMENSION = 384
22     DISTANCE_METRIC = Distance.COSINE
23     BATCH_SIZE = 64
24
25     # Normalize vector for cosine similarity (optional, improves scores)
26     def normalize_vector(vector):
27         return vector / np.linalg.norm(vector)
28
29     # Split text into smaller chunks for better context alignment
30     def split_into_chunks(text: str, max_chunk_size: int = 150) -> list:
31         sentences = re.split(r'(?<=[!?!])s+', text) # Split by sentence endings
32         chunks = []
33         current_chunk = []
34
35         for sentence in sentences:
36             if sum(len(s) for s in current_chunk) + len(sentence) <= max_chunk_size:
37                 current_chunk.append(sentence)
38             else:
39                 chunks.append(' '.join(current_chunk))
40                 current_chunk = [sentence]
41
42         if current_chunk:
43             chunks.append(' '.join(current_chunk))
44
45     return chunks
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```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

TERMINAL

```
2024-12-08 18:40:06.598 | 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 18:40:07.142 | 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 18:40:07.248 | INFO | __main__:feature_engineering_and_instruct_task:130 - Upserted 12 records to Qdrant.
2024-12-08 18:40:07.248 | INFO | __main__:feature_engineering_and_instruct_task:131 - All records successfully upserted to Qdrant.
```

localhost:6333/dashboard#/collections/instruct_dataset_clearml

Finish update

drant

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>

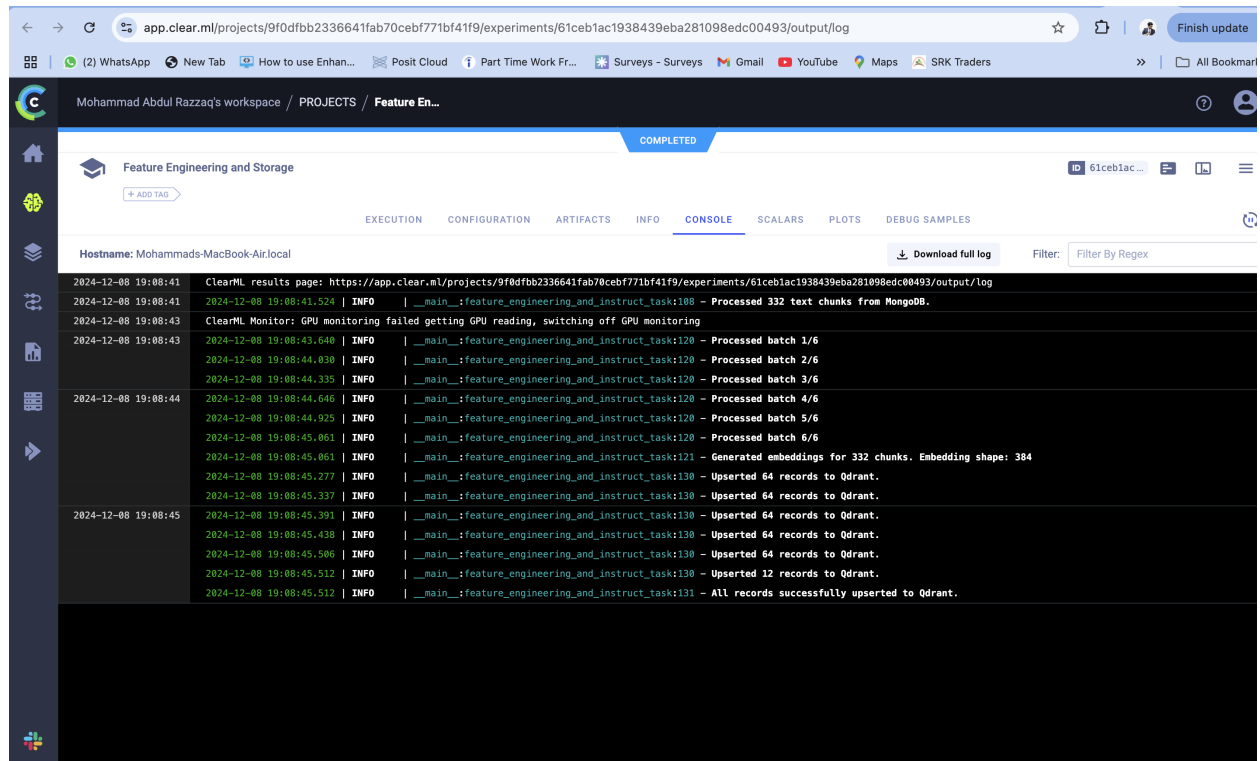
metadata

```
~{ 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:

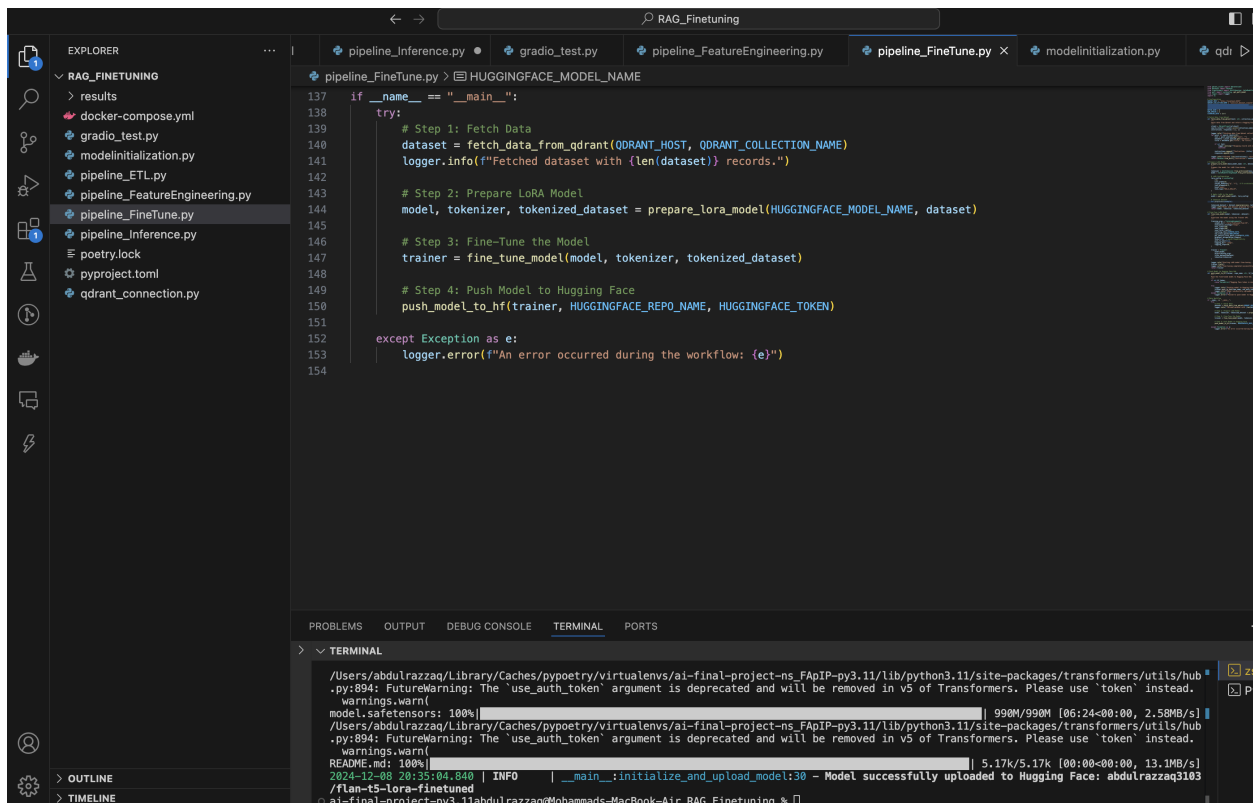
Vectors:

v1.12.4



- **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.

```

1 import gradio as gr
2 from qdrant_client import QdrantClient
3 from transformers import pipeline as hf_pipeline
4 from sentence_transformers import SentenceTransformer
5 from loguru import logger
6
7 # Configuration
8 QDRANT_HOST = "http://localhost:6333"
9 QDRANT_COLLECTION_NAME = "instruct_dataset_cleanml"
10 MODEL_NAME = "google/flan-t5-base" # Open-source instruction-tuned model
11 VECTOR_DIMENSION = 384 # Ensure this matches your Qdrant configuration
12 MAX_CONTEXT_TOKENS = 512 # Token limit for the context
13
14 # Initialize Models
15 embedding_model = SentenceTransformer("sentence-transformers/all-MiniLM-L6-v2") # For embeddings
16 hf_model = hf_pipeline("text2text-generation", model=MODEL_NAME) # Hugging Face model for generation
17
18 # Define the RAG pipeline
19 def rag_pipeline(query: str):
20     """
21     Perform the full RAG pipeline: query -> knowledge base -> model -> response.
22     """
23     logger.info(f"Received query: {query}")
24
25     def fetch_similar_data(query: str, top_k: int = 3) -> list:

```

TERMINAL

```

2024-12-08 22:16:46.703 | INFO | _main_:query_rag:135 - Query received: Explain Nav2 concepts.
2024-12-08 22:16:46.703 | INFO | _main_:rag_pipeline:23 - Received query: Explain Nav2 concepts.
2024-12-08 22:16:46.703 | INFO | _main_:fetch_similar_data:29 - Generating query embedding...
2024-12-08 22:16:46.939 | INFO | _main_:fetch_similar_data:32 - Connecting to Qdrant and searching for similar data...
2024-12-08 22:16:46.959 | INFO | _main_:fetch_similar_data:45 - Retrieved 3 similar documents from Qdrant.
2024-12-08 22:16:46.959 | INFO | _main_:fetch_similar_data:47 - Document 1: A1:
Nav2, or Navigation 2, is the ROS 2 navigation stack that enables autonomous navigation for robots. It provides a set of tools and libraries for tasks such as path planning, obstacle avoidance, and localization, allowing robots to navigate from point A to point B safely and efficiently.
Q2: How does Nav2 differ from the original ROS navigation stack? (Score: 0.6891055)
2024-12-08 22:16:46.959 | INFO | _main_:fetch_similar_data:47 - Document 2: A1:
ROS (Robot Operating System) is an open-source framework for robot software development. Key features include communication tools, hardware abstraction, and libraries for commonly used robot functionalities like navigation and control. Q2:
What is Nav2, and what are its primary use cases? A2:
Nav2, or Navigation 2, is a ROS2-based navigation framework that provides capabilities like path planning, behavior trees, lifecycle nodes, and SLAM integration for robotic systems. (Score: 0.5756691)
2024-12-08 22:16:46.959 | INFO | _main_:fetch_similar_data:47 - Document 3: For mobile robots, Nav 2 can send twist commands to the Diff Drive Controller, controlling the base's motion. The system includes several specialized controllers:

```

RAG Pipeline Interface

Enter a query to retrieve and generate a response using the RAG pipeline.

Enter your query

Explain Nav2 concepts.

Clear
Submit

≡ Examples

What is ROS2?
Explain Nav2 concepts.
How does MoveIt2 work with Gazebo?

Generated Response

Nav2 is a ROS2 navigation stack that enables autonomous navigation for robots.

Flag

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.