

Build a LangChain RAG Agent

Introduction

In this guide, we'll walk through how to create agents capable of answering questions based on any provided document using the Fetch.ai uAgents and ai_engine python libraries as well as openai and cohere. The aim is to assist you in building a LangChain Retrieval-Augmented Generation (RAG) Agent !

The process of creating RAG agents has been improved by offering a decentralized and modular framework through the uAgents and AI Engine libraries. These tools streamline the integration of AI models like OpenAI and Cohere, enabling more efficient and scalable development of intelligent agents. With enhanced interoperability, security, and resource management, this framework allows developers to quickly build and deploy sophisticated agents that can effectively answer questions based on any provided document, making the entire process faster and more robust.

Check out the [AI Engine package\(opens in a new tab\)](#) and [uAgents\(opens in a new tab\)](#) packages to download them and start integrating these tools within your Agents project!

Current version of the AI Engine package is 0.6.0 . Current version of the uAgents package is 0.17.1 . Let's dive into the LangChain RAG Agents development!

Prerequisites

Make sure you have read the following resources before going on with this guide:

- [Creating an agent](#)
- [Creating an interval task](#)
- [Communicating with other agents](#)
- [Agent Handlers](#)
- [Register in Almanac](#)
- [Almanac Contract](#)
- [Mailbox](#)
- [Utilising the Agentverse Mailroom service](#)
- [Protocols](#)
- [Agentverse Functions](#)
- [Register an Agent Function on the Agentverse](#)

API KEYS

Importantly, you will need two API keys to correctly go through this guide: one from OpenAI and one from Cohere . Follow the steps provided below to obtain these keys:

OpenAI API Key

1. Visit the [OpenAI website\(opens in a new tab\)](#)
2. .
3. Sign up or log in to your account.
4. Navigate to the API
5. section.
6. Generate or retrieve your API key.

Cohere API Key

1. Visit the [Cohere website\(opens in a new tab\)](#)
2. .
3. Sign up or log in to your account.
4. Go to the API Keys
5. section.
6. Copy an existing key or create a new one.: one from OpenAI and one from Cohere. Follow the steps below to obtain these keys.

Project structure and overview

Project structure

The project is structured as follows:

langchain-rag/ . ├── poetry.lock ├── pyproject.toml └── src ├── agents | ├── langchain_rag_agent.py | ├── langchain_rag_user.py ├── main.py ├── messages ├── requests.py The source code directory (src) contains the following directories and files:

- agents
 - : Contains scripts for the LangChain agents.* langchain_rag_agent.py
 - - : Script for the RAG agent (retrieves info, finds documents, generates answers).
 - - langchain_rag_user.py: Script for the User agent (asks questions, handles responses).
- main.py
 - : Starts both the RAG and user agents.
- messages
 - : Defines custom message models.* requests.py
 - - : Defines the RagRequest message model (question, URL, optional deep read).

Environment variables

You'll need to set up environment variables for your project to run correctly. These variables include your API keys, which should be stored in a .env file within the src directory.

To do this, navigate to the src directory. Here, create and source the .env file. Within this one, add the following:

```
export COHERE_API_KEY="YOUR_COHERE_API_KEY" export OPENAI_API_KEY="YOUR_OPENAI_API_KEY"
```

Langchain RAG setup

Project dependencies

You'll need several Python packages for the project. These can be managed efficiently using [Poetry\(opens in a new tab\)](#) . The following dependencies are needed for the correct development of the Langchain RAG Agent project:

```
[tool.poetry.dependencies] python = ">=3.10,<3.12" uagents = "*" requests = "^2.31.0" langchain = "^0.2.11" openai =
"^1.12.0" langchain-openai = "^0.1.19" tiktoken = "^0.7.0" cohere = "^4.47" faiss-cpu = "^1.7.4" validators = "^0.22.0"
uagents-ai-engine = "^0.4.0" unstructured = "^0.12.4"
```

Messages Data Model

RagRequest Model

We now need to define the requests.py file under the messages folder in the project.

windows echo .

requests . py The script look like the following:

Self hosted requests.py from typing import Optional from uagents import Model , Field

class

RagRequest (Model): question :

str

=

Field (description = "The question that the user wants to have an answer for.") url :

str

=

Field (description = "The url of the docs where the answer is.") deep_read : Optional [str]

=

Field (description = "Specifies weather all nested pages referenced from the starting URL should be read or not. The value should be yes or no." , default = "no" ,) Here's a breakdown of the RagRequest message data Model:

- question (str)
- : This is the user's question that needs to be answered based on the provided website URL.
- url (str)
- : This is the URL of the website where the answer should be found.
- deep_read (Optional[str], default="no")
- : This optional field allows you to specify whether the RAG agent should follow and read nested pages (i.e., pages linked from the starting URL). Valid values are "yes"
- or "no"
- . By default
- , the agent focuses only on the initial URL.

Agents

LangChain RAG Agent

This step involves setting up the LangChain Retrieval-Augmented Generation (RAG) Agent which is able to scrape web content, retrieve relevant documents, and generate answers to user queries based on that content. Let's create the file within the agents directory we created under src directory of our project and name it langchain_rag_agent.py by using the following command:

windows echo .

langchain_rag_agent . py The agent is defined as [a local agent with a Mailbox \(opens in a new tab\)](#) and it is able to answer questions by fetching and summarizing information from a given website URL.

The script for this agent looks as follows:

```
Self hosted langchain_rag_agent.py
import traceback
from uagents import Agent, Context, Protocol
import validators
from messages import requests
import RagRequest
import os
from langchain_openai import ChatOpenAI
from langchain import prompts
import ChatPromptTemplate
from langchain_community.document_loaders import UnstructuredURLLoader
import requests
from bs4 import BeautifulSoup
from urllib.parse import urlparse
from langchain_openai import OpenAIEmbeddings
from langchain_community.vectorstores import FAISS
from langchain import retrievers
import ContextualCompressionRetriever
from langchain import retrievers
import document_compressors
import CohereRerank
from ai_engine import UAgentResponse, UAgentResponseType
import nltk
from uagents import setup
import fund_agent_if_low
```

```
nltk.download("punkt")
nltk.download("averaged_perceptron_tagger")
```

LANGCHAIN_RAG_SEED

```
"YOUR_LANGCHAIN_RAG_SEED" AGENT_MAILBOX_KEY =
```

```
"YOUR_MAILBOX_KEY"
```

agent

```
Agent ( name = "langchain-rag-agent", seed = LANGCHAIN_RAG_SEED, mailbox = f "{ AGENT_MAILBOX_KEY } @https://agentverse.ai", )
```

```
fund_agent_if_low (agent.wallet.address ())
```

docs_bot_protocol

```
Protocol ( "DocsBot" )
```

PROMPT_TEMPLATE

```
""" Answer the question based only on the following context:
```

```
{context}
```

```
Answer the question based on the above context: {question} """
```

```
def
```

```

create_retriever ( ctx : Context ,
url :
str ,
deep_read :
bool ) -> ContextualCompressionRetriever: def
scrape ( site :
str ): if
not validators . url (site): ctx . logger . info ( f "Url { site } is not valid" ) return

```

r

```

requests . get (site) soup =
BeautifulSoup (r.text, "html.parser" )

```

parsed_url

```

urlparse (url) base_domain = parsed_url . scheme +
"://"
+ parsed_url . netloc

```

link_array

```

soup . find_all ( "a" ) for link in link_array : href :
str
= link . get ( "href" , "" ) if
len (href)
==
0 : continue current_site =
f " { base_domain } { href } "
if href . startswith ( "/" )
else href if ( ".php"
in current_site or
"#"
in current_site or
not current_site . startswith (url) or current_site in urls ) : continue urls . append (current_site) scrape (current_site)

```

urls

```

[url] if deep_read : scrape (url) ctx . logger . info ( f "After deep scraping - urls to parse: { urls } " )
try : loader =
UnstructuredURLLoader (urls = urls) docs = loader . load_and_split () db = FAISS . from_documents (docs,
OpenAIEmbeddings ()) compression_retriever =
ContextualCompressionRetriever ( base_compressor = CohereRerank (), base_retriever = db. as_retriever () ) return

```

compression_retriever except

Exception

```
as exc : ctx . logger . error ( f "Error happened: { exc } " ) traceback . format_exception (exc)
```

```
@docs_bot_protocol . on_message (model = RagRequest, replies = {UAgentResponse}) async
```

```
def
```

```
answer_question ( ctx : Context ,
```

```
sender :
```

```
str ,
```

```
msg : RagRequest): ctx . logger . info ( f "Received message from { sender } , session: { ctx.session } " ) ctx . logger . info ( f  
"input url: { msg.url } , question: { msg.question } , is deep scraping: { msg.deep_read } " )
```

parsed_url

```
urlparse (msg.url) if
```

```
not parsed_url . scheme or
```

```
not parsed_url . netloc : ctx . logger . error ( "invalid input url" ) await ctx . send ( sender, UAgentResponse ( message =  
"Input url is not valid" , type = UAgentResponseType.FINAL, ), ) return base_domain = parsed_url . scheme +
```

```
".:/"
```

```
+ parsed_url . netloc ctx . logger . info ( f "Base domain: { base_domain } " )
```

retriever

```
create_retriever (ctx, url = msg.url, deep_read = msg.deep_read ==
```

```
"yes" )
```

compressed_docs

```
retriever . get_relevant_documents (msg.question) context_text =
```

```
"\n\n---\n\n" . join ([doc.page_content for doc in compressed_docs]) prompt_template = ChatPromptTemplate .  
from_template (PROMPT_TEMPLATE) prompt = prompt_template . format (context = context_text, question =  
msg.question)
```

model

```
ChatOpenAI (model = "gpt-4o-mini" ) response = model . predict (prompt) ctx . logger . info ( f "Response: { response } " )  
await ctx . send ( sender, UAgentResponse (message = response, type = UAgentResponseType.FINAL) )
```

```
agent . include (docs_bot_protocol, publish_manifest = True )
```

```
if
```

```
name
```

```
==
```

```
"main" : agent . run () In order to correctly run this code, you need to provide the name ,seed ,mailbox  
,LANGCHAIN_RAG_SEED ,PROMPT_TEMPLATE andAGENT_MAILBOX_KEY parameters to correctly run this example.
```

The agent fetches and processes information from specified URLs to answer users' questions. Let's explore the script above in more detail. Initially, multiple modules are imported, including tools for making HTTP requests, parsing HTML, and handling Natural Language Processing (NLP). The agent uses OpenAI's GPT-4o-mini model for generating answers based on retrieved documents. We then up an agent called `langchain_rag_agent()` with a unique seed and mailbox address. Check out the [Mailbox](#) guide for additional info on Agent Mailboxes.

To ensure the agent has sufficient funds to operate, the `fund_agent_if_low()` function is invoked.

Next, the `DocsBot` [protocol](#) is defined using the `Protocol` class of the `uagents` library to handle message interactions. This protocol uses a predefined prompt template to structure the questions and context for the language model.

The `create_retriever()` function is responsible for fetching and parsing web pages. If necessary, it can perform deep scraping, which involves gathering all linked pages within the same domain. The function validates URLs, retrieves HTML content, and uses `BeautifulSoup` to parse the HTML and find links. The documents are then loaded and split using `LangChain's UnstructuredURLLoader`, indexed with `FAISS`, and compressed with `Cohere`, creating a retriever that can extract relevant information.

The `answer_question()` function, decorated with the `on_message()` [handler](#), is triggered when the agent receives a message matching the `RagRequest` message data model. The function validates the input URL and creates a retriever to fetch relevant documents based on the question. The context alongside with the question are then used to create a prompt for the language model (`ChatOpenAI`), which generates the final answer.

The generated answer is then sent back to the user.

Finally, the agent is set to include the `DocsBot` protocol and is then run, which starts the agent and allows it to receive and process incoming messages.

LangChain User Agent

We are now ready to define the `LangChain` User Agent which interacts with the `LangChain` RAG agent defined previously to ask a question based on documents found at specified URLs. The user agent sends a request to the RAG agent to retrieve and process information from the web page and then handle the response. The agent is defined as a [local agent with an endpoint \(opens in a new tab\)](#).

Create a file for this agent:

`windows echo .`

`langchain_rag_user . py` The script looks as follows:

Self hosted `langchain_rag_user.py` from `uagents` import `Agent`, `Context`, `Protocol` from `messages` . `requests` import `RagRequest` from `ai_engine` import `UAgentResponse` from `uagents` . `setup` import `fund_agent_if_low`

QUESTION

"How to install uagents using pip" URL =

"https://fetch.ai/docs/guides/agents/installing-uagent" DEEP_READ = ("no")

RAG_AGENT_ADDRESS

"YOUR_LANGCHAIN_RAG_AGENT_ADDRESS"

user

`Agent (name = "langchain_rag_user" , port = 8000 , endpoint = ["http://127.0.0.1:8000/submit"],)` `fund_agent_if_low`
(`user.wallet.address ()`) `rag_user =`

`Protocol ("LangChain RAG user")`

`@rag_user . on_interval (60 , messages = RagRequest) async`

`def`

`ask_question (ctx : Context):` `ctx . logger . info (f "Asking RAG agent to answer { QUESTION } based on document located at { URL } , reading nested pages too: { DEEP_READ } ")` `await ctx . send (RAG_AGENT_ADDRESS, RagRequest`
(`question = QUESTION, url = URL, deep_read = DEEP_READ`))

`@rag_user . on_message (model = UAgentResponse) async`

`def`

`handle_data (ctx : Context ,`

```

sender :

str ,

data : UAgentResponse): ctx . logger . info ( f "Got response from RAG agent: { data.message } " )

user . include (rag_user)

if

name

==

```

"main" : rag_user . run () Remember that you need to provide theQUESTION ,URL ,DEEP_READ ,RAG_AGENT_ADDRESS ,name ,seed andendpoint parameters to correctly run this code.

Here, we have created the User Agent which interacts with the LangChain RAG Agent we previously defined. The User Agent periodically asks the RAG Agent a predefined question and then handles the response.

After importing all required modules and classes, a specificQUESTION is provided:"How to install uagents using pip" . We also provide theURL for the webpage where the relevant information can be found to answer the question. TheDEEP_READ variable is set to 0 thus indicating that the agent should only read the main page and not follow and read any nested or linked pages.

TheRAG_AGENT_ADDRESS variable holds the address of the RAG agent, which is responsible for retrieving and processing the webpage content to answer the user's question.

We are now ready to define the User Agent. We create thelangchain_rag_user is created with a specificport andendpoint , which is where it will communicate with the RAG Agent. The agent's wallet is funded if needed using thefund_agent_if_low() function.

We then proceed and define theprotocol . A protocol namedLangChain RAG user is defined using theProtocol class of theuagents library to handle the agent's interactions. This protocol contains two important functions:

1. Theask_question()
2. function is decorated with the.on_interval()
3. [handler](#)
4. and it is set to run at 60-second intervals. This function sends aRagRequest
5. message to the LangChain RAG agent, asking it to answer theQUESTION
6. based on the specifiedURL
7. . It then logs theQUESTION
8. andURL
9. details for debugging purposes.
10. Thehandle_data()
11. function is decorated with the.on_message()
12. handler and it handles incoming messages from the LangChain RAG agent. When a response is received, it logs the response message.

Finally, the LangChain RAG user protocol is then included using the.include() method into the agent which is then run by callingrag_user.run() in the main block.

Main script

We are now ready to define the main script for our project. In thesrc folder of our project we create a Python script namedmain.py using the following command:

```

windows echo .

```

main . py We then define the code within this one which looks like the one provided here below:

```

Self hosted main.py from uagents import Bureau from agents . langchain_rag_agent import agent from agents .
langchain_rag_user import user

```

```

if

name

==

"main" : bureau =

```

Bureau (endpoint = "http://127.0.0.1:8000/submit" , port = 8000) print (f "Adding RAG agent to Bureau: { agent.address } ")
bureau . add (agent) print (f "Adding user agent to Bureau: { user.address } ") bureau . add (user) bureau . run () Now, both
of agents are set up, and you will need to connect your local agents to the Agentverse so for them to be retrievable for
communication and interaction with any other registered agent on the Fetch Network. This way, thelangchain_rag_agent will
be connected to the Agentverse using a [Mailbox](#) and afterward a [Agent Function \(opens in a new tab\)](#) will be created
and [registered \(opens in a new tab\)](#) for this agent.

The langchain_rag_user agent is used as a testing agent for the RAG agent being registered on the Agentverse and made
subsequently available on DeltaV for queries.

➤ Head over to the following [guide](#) for a better understanding of local agent registration on the Agentverse.

Expected output

The expected output for this example should be similar to the following one where we questioned the RAG Agent with the
question How to install uagents using pip from the RAG user agent:

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[Hosted agent REST endpoints](#)

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