

Multi-agent workflows with Fetch.ai x Langchain

Multi-agent workflows are at the forefront of modern agent development; the idea that individual [Agents](#) can be utilized to create larger more complex services for people has created lots of excitement in the AI space. At Fetch.ai, we're building the agent communication layer which perfectly compliments Langchain libraries.

Agents representing smaller parts of a service, allow for many agents to represent a whole. Agents reduce technical requirements in projects, for example you wouldn't need to write a function to calculate the historical index of a stock price, an agent will already return that data for you.

Before we go any further please read over our introduction guide to [Agents and Langchain](#)

The system

Three agents make up a simple agent workflow: RequestAgent, PDFQuestionAgent and PDFSplitAgent.

A variation of the following Model class is passed between each agent:

```
class
DocumentUnderstanding ( Model ): pdf_path :

str question :

str The flow is the following:

    • RequestAgent
    • provides the DocumentUnderstanding
    • object.
    • PDFQuestionAgent
    • upon receiving DocumentUnderstanding
    • sends a request to the third agent to validate and split the document.
    • PDFSplitAgent
    • upon receiving the request from PDFQuestionAgent, returns a list of pages to it.
    • PDFQuestionAgent
    • upto receiving the message from PDFSplitAgent processes the pages to answer the question from RequestAgent, then
    returns that answer to this latter one.
```

In this example, we are hard coding the agents addresses, meaning we know them. To search for agents dynamically, take a look at the [Almanac](#).

Installation

Run the following:

```
poetry
init poetry
add
uagents
requests
langchain
openai
langchain-openai
faiss-cpu
```

validators Versions used for this example are:

```
[tool.poetry.dependencies] python = ">=3.10,<3.12" uagents = "0.12.0" requests = "^2.31.0" langchain = "^0.1.7" openai =
"^1.12.0" langchain-openai = "^0.0.6" faiss-cpu = "^1.7.4"
```

Environment setup

```
export OPENAI_API_KEY="YOUR_OPENAI_API_KEY"
```

Agent 1 - RequestAgent: provides a question and a source

This is our simplest agent; this agent provides a link to a PDF and question to be answered from the document.

Self hosted request_agent.py from uagents import Agent , Context , Protocol , Model

```
class
```

```
DocumentUnderstanding ( Model ): pdf_path :
```

```
str question :
```

```
str
```

```
class
```

```
DocumentsResponse ( Model ): learnings :
```

```
str
```

agent

```
Agent ( name = "find_in_pdf" , seed = "" , port = 8001 , endpoint = [ "http://127.0.0.1:8001/submit" ], )
```

```
print ( "uAgent address: " , agent.address) summary_protocol =
```

```
Protocol ( "Text Summarizer" )
```

AGENT_2_FAISS

```
"""
```

```
@agent . on_event ( "startup" ) async
```

```
def
```

```
on_startup ( ctx : Context): await ctx . send ( AGENT_2_FAISS, DocumentUnderstanding (pdf_path = "/a.pdf" , question =  
"What is the story about?" ), )
```

```
@agent . on_message (model = DocumentsResponse) async
```

```
def
```

```
document_load ( ctx : Context ,
```

```
sender :
```

```
str ,
```

```
msg : DocumentsResponse): ctx . logger . info (msg.learnings)
```

```
agent . include (summary_protocol, publish_manifest = True ) agent . run ()
```

Agent 2 - PDFQuestionAgent: takes a request and returns a result

The PDFQuestionAgent gets the PDF and the request from the first agent, but is unable to split the PDF. This second agent sends a request to the third agent to split the PDF. Once the pages from the PDF are returned, a FAISS similarity search is ran on the pages by the second agent.

Self hosted pdf_question_agent.py from langchain_community . vectorstores import FAISS from langchain_community .
docstore . in_memory import InMemoryDocstore from langchain_openai import OpenAIEmbeddings from uagents import
Agent , Context , Protocol , Model from langchain_core . documents import Document from typing import List import os
import uuid import faiss

```
class
```

```
PDF_Request ( Model ): pdf_path :
```

```

str
class
DocumentUnderstanding ( Model ): pdf_path :
str question :
str
class
PagesResponse ( Model ): pages : List
class
DocumentsResponse ( Model ): learnings :
str

```

faiss_pdf_agent

```

Agent ( name = "faiss_pdf_agent" , seed = "" , port = 8002 , endpoint = [ "http://127.0.0.1:8002/submit" ], )
print ( "uAgent address: " , faiss_pdf_agent.address) faiss_protocol =
Protocol ( "FAISS" )

```

RequestAgent

```

""" PDF_splitter_address =
"""

```

openai_api_key

```

os . environ [ "OPENAI_API_KEY" ] embeddings =
OpenAIEmbeddings (model = "text-embedding-3-large" )
@faiss_pdf_agent . on_message (model = DocumentUnderstanding, replies = PDF_Request) async
def
document_load ( ctx : Context ,
sender :
str ,
msg : DocumentUnderstanding): ctx . logger . info (msg) ctx . storage . set ( str (ctx.session), { "question" : msg.question,
"sender" : sender}) await ctx . send ( PDF_splitter_address, PDF_Request (pdf_path = msg.pdf_path) )
@faiss_pdf_agent . on_message (model = PagesResponse, replies = DocumentsResponse) async
def
document_understand ( ctx : Context ,
sender :
str ,
msg : PagesResponse): index = faiss . IndexFlatL2 ( len (embeddings. embed_query ( "hello" )))

```

vector_store

```

FAISS ( embedding_function = embeddings, index = index, docstore = InMemoryDocstore (), index_to_docstore_id = {}, )

```

documents

```
[] for page in msg . pages : documents . append ( Document (page_content = page[ "page_content" ], metadata = page[ "metadata" ]) )
```

uuids

```
[ str (uuid. uuid4 ())  
for _ in  
range ( len (documents)) ]  
vector_store . add_documents (documents = documents, ids = uuids)
```

prev

```
ctx . storage . get ( str (ctx.session))
```

results

```
vector_store . similarity_search ( prev[ "question" ], k = 2 , )  
if  
len (results)  
0 : await ctx . send ( prev[ "sender" ], DocumentsResponse (learnings = results[ 0 ].page_content) )  
faiss_pdf_agent . include (faiss_protocol, publish_manifest = True ) faiss_pdf_agent . run () The core difference with agent  
two compared to other agents you have seen so far is that there are multiple message decorators. Your agent can have  
as many any number of message handlers as you want.
```

PDFQuestionAgent also has every request/response model to communicate with RequestAgent and PDFSplitAgent .

Agent 3 - PDFSplitAgent: validates the PDF, loads and returns the split

The PDFSplitAgent receives the PDF, and splits the document using the langchain_community document loader PyPDFLoader . It then returns an array of pages to the second agent.

Self hosted pdf_split_agent.py from langchain_community . document_loaders import PyPDFLoader from uagents import Agent , Context , Protocol , Model from typing import List

```
class  
PDF_Request ( Model ): pdf_path :  
str  
class  
PagesResponse ( Model ): pages : List
```

pdf_loader_agent

```
Agent ( name = "pdf_loader_agent" , seed = "" , port = 8003 , endpoint = [ "http://127.0.0.1:8003/submit" ], )  
print ( "uAgent address: " , pdf_loader_agent.address) pdf_loader_protocol =  
Protocol ( "Text Summarizer" )  
@pdf_loader_agent . on_message (model = PDF_Request, replies = PagesResponse) async  
def
```

```

document_load ( ctx : Context ,
sender :
str ,
msg : PDF_Request): loader =
PyPDFLoader (msg.pdf_path) pages = loader . load_and_split () await ctx . send (sender, PagesResponse (pages =
pages))
pdf_loader_agent . include (pdf_loader_protocol, publish_manifest = True ) pdf_loader_agent . run ()

```

Run the agents

We need to run the agents backwards so that we can generate their addresses and then update the other agents with their addresses respectively.

Let's run `PDFSplitAgent`, and update `PDFQuestionAgent` with its address:

```
Run: poetry run python pdf_split_agent.py
```

Update `pdf_question_agent.py` script by filling the `PDF_splitter_address` field with the address of the third agent.

```
Run: poetry run python pdf_question_agent.py
```

Update `request_agent.py` script by filling the `AGENT_2_FAISS` field with the address of the second agent.

```
Run: poetry run python request_agent.py
```

Add the address of the first agent in the dedicated field `RequestAgent` within the script for the second agent, `pdf_question_agent.py`.

Expected Output

- request_agent.py
- :
- uAgent address: agent1qf4au6rzaauxhy2jze6v85rspgvredx9m42p0e0cukz0hvh4dh2sqjuhuipp
- INFO: [find_in_pdf]: Manifest published successfully: Text Summarizer
- INFO: [find_in_pdf]: Registering on almanac contract...
- INFO: [find_in_pdf]: Registering on almanac contract...complete
- INFO: [find_in_pdf]: Starting server on http://0.0.0.0:8001 (Press CTRL+C to quit)
- INFO: [find_in_pdf]: GuidesAI AgentsGetting StartedWhat's an Agent?
- Beginner Python
- Agents - uAgents Framework
- Introduction
- Agents are autonomous software programs that can communicate with each other, they're
- designed to solve problems alone, or as part of a multi agent system. Standardised
- communication protocols allows agents to communicate efficiently, aiding negotiation and
- problem resolution. Unlike microservices or centralized systems, agents are reactive and self
- reasoning. Agents are built using the uAgents framework, and with additional libraries are
- party to the AI Engine.
- Agents offer interconnectivity for larger systems, where centralised systems may have some
- fault tolerance, multi agent systems allow for plug and play design; agents can be replaced
- or upgraded in a live system without downtime. Agents can be built to only care for one
- process or rule. In your systems, agents may have a totally independent network.
- Agents aren't just agents, you can wrap LLMs and other models to create self reasoning, self
- learning chat interfaces or services. This, is as simple as 12 lines of code link
- pdf_question_agent.py
- :
- INFO:faiss.loader:Loading faiss with AVX2 support.
- INFO:faiss.loader:Successfully loaded faiss with AVX2 support.
- uAgent address: agent1qt89fz44fp0nxvkpgfts4lm566lj8gs7qnlh7yz3lwz5f5scp7nrkcpt3qe
- INFO: [faiss_pdf_agent]: Manifest published successfully: FAISS
- INFO: [faiss_pdf_agent]: Registration on Almanac API successful
- INFO: [faiss_pdf_agent]: Registering on almanac contract...
- INFO: [faiss_pdf_agent]: Registering on almanac contract...complete
- INFO: [faiss_pdf_agent]: Starting server on http://0.0.0.0:8002 (Press CTRL+C to quit)
- INFO: [faiss_pdf_agent]: pdf_path='./a.pdf' question='What is the story about?'
- INFO:httpx:HTTP Request: POST https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"

- INFO:httpx:HTTP Request: POST https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
- INFO:httpx:HTTP Request: POST https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
- pdf_split_agent.py
- :
- uAgent address: agent1qf4au6rzaauxhy2jze6v85rspgvredx9m42p0e0cukz0hv4dh2sqjuhuipp
- INFO: [pdf_loader_agent]: Manifest published successfully: Text Summarizer
- INFO: [pdf_loader_agent]: Registering on almanac contract...
- INFO: [pdf_loader_agent]: Registering on almanac contract...complete
- INFO: [pdf_loader_agent]: Starting server on http://0.0.0.0:8003 (Press CTRL+C to quit)

Last updated on October 17, 2024

Was this page helpful?

You can also leave detailed feedback[on Github](#)

[Getting started with Fetch.ai x Langchain](#) [Getting started with Fetch.ai and Swarm](#)

On This Page

- [The system](#)
- [Installation](#)
- [Environment setup](#)
- [Agent 1 - RequestAgent: provides a question and a source](#)
- [Agent 2 - PDFQuestionAgent: takes a request and returns a result](#)
- [Agent 3 - PDFSplitAgent: validates the PDF, loads and returns the split](#)
- [Run the agents](#)
- [Expected Output](#)
- [Edit this page on github\(opens in a new tab\)](#)