Modules

Data connection

Retrievers

Self-querying

Self-querying with Pinecone

Self-querying with Pinecone

In the walkthrough we'll demo the SelfQueryRetriever with a Pinecone vector store.

Creating a Pinecone index

First we'll want to create a Pinecone VectorStore and seed it with some data. We've created a small demo set of documents that contain summaries of movies.

To use Pinecone, you have to have pinecone package installed and you must have an API key and an Environment. Here are the installation instructions.

NOTE: The self-query retriever requires you to have lark package installed.

```
# !pip install lark

#!pip install pinecone-client

import os
import pinecone

pinecone.init(
    api_key=os.environ["PINECONE_API_KEY"],
environment=os.environ["PINECONE_ENV"]
)
```

```
/Users/harrisonchase/.pyenv/versions/3.9.1/envs/langchain/lib/python3.9/sipackages/pinecone/index.py:4: TqdmExperimentalWarning: Using `tqdm.autonotebook.tqdm` in notebook mode. Use `tqdm.tqdm` instead to foreconsole mode (e.g. in jupyter console) from tqdm.autonotebook import tqdm
```

```
from langchain.schema import Document
from langchain.embeddings.openai import OpenAIEmbeddings
from langchain.vectorstores import Pinecone

embeddings = OpenAIEmbeddings()
# create new index
pinecone.create_index("langchain-self-retriever-demo", dimension=1536)
```

API Reference:

- Document from langchain.schema
- OpenAIEmbeddings from langchain.embeddings.openai
- Pinecone from [langchain.vectorstores]

```
docs = [
    Document(
        page_content="A bunch of scientists bring back dinosaurs and
mayhem breaks loose",
        metadata={"year": 1993, "rating": 7.7, "genre": ["action",
"science fiction"]},
    ),
    Document(
        page_content="Leo DiCaprio gets lost in a dream within a dream
within a dream within a ...",
        metadata={"year": 2010, "director": "Christopher Nolan",
"rating": 8.2},
    ),
    Document(
        page_content="A psychologist / detective gets lost in a series
of dreams within dreams within dreams and Inception reused the idea",
        metadata={"year": 2006, "director": "Satoshi Kon", "rating":
8.6},
    ),
    Document(
        page_content="A bunch of normal-sized women are supremely
wholesome and some men pine after them",
        metadata={"year": 2019, "director": "Greta Gerwig", "rating":
8.3},
    ),
    Document(
        page_content="Toys come alive and have a blast doing so",
        metadata={"year": 1995, "genre": "animated"},
    ),
    Document(
```

```
page_content="Three men walk into the Zone, three men walk out
of the Zone",
    metadata={
        "year": 1979,
        "rating": 9.9,
        "director": "Andrei Tarkovsky",
        "genre": ["science fiction", "thriller"],
        "rating": 9.9,
      },
    ),
]
vectorstore = Pinecone.from_documents(
    docs, embeddings, index_name="langchain-self-retriever-demo"
)
```

Creating our self-querying retriever

Now we can instantiate our retriever. To do this we'll need to provide some information upfront about the metadata fields that our documents support and a short description of the document contents.

```
from langchain.llms import OpenAI
from langchain.retrievers.self_query.base import SelfQueryRetriever
from langchain.chains.query_constructor.base import AttributeInfo
metadata_field_info = [
    AttributeInfo(
        name="genre",
        description="The genre of the movie",
        type="string or list[string]",
    ),
    AttributeInfo(
        name="year",
        description="The year the movie was released",
        type="integer",
    ),
    AttributeInfo(
        name="director",
        description="The name of the movie director",
        type="string",
    ),
    AttributeInfo(
        name="rating", description="A 1-10 rating for the movie",
type="float"
```

```
),
]
document_content_description = "Brief summary of a movie"
llm = OpenAI(temperature=0)
retriever = SelfQueryRetriever.from_llm(
    llm, vectorstore, document_content_description,
metadata_field_info, verbose=True
)
```

API Reference:

- OpenAl from (langchain.llms)
- SelfQueryRetriever from langchain.retrievers.self_query.base
- AttributeInfo from langchain.chains.query_constructor.base

Testing it out

And now we can try actually using our retriever!

```
# This example only specifies a relevant query
retriever.get_relevant_documents("What are some movies about
dinosaurs")
```

```
query='dinosaur' filter=None
```

[Document(page_content='A bunch of scientists bring back dinosaurs and mayhem breaks loose', metadata={'genre': ['action', 'science fiction'], 'rating': 7.7, 'year': 1993.0}),

Document(page_content='Toys come alive and have a blast doing so',
metadata={'genre': 'animated', 'year': 1995.0}),

Document(page_content='A psychologist / detective gets lost in a series of dreams within dreams within dreams and Inception reused the idea', metadata={'director': 'Satoshi Kon', 'rating': 8.6, 'year': 2006.0}),

Document(page_content='Leo DiCaprio gets lost in a dream within a dream within a ...', metadata={'director': 'Christopher Nolan', 'rating': 8.2, 'year': 2010.0})]

This example only specifies a filter
retriever.get_relevant_documents("I want to watch a movie rated higher
than 8.5")

```
query=' ' filter=Comparison(comparator=<Comparator.GT: 'gt'>,
attribute='rating', value=8.5)
```

[Document(page_content='A psychologist / detective gets lost in a series of dreams within dreams within dreams and Inception reused the idea', metadata={'director': 'Satoshi Kon', 'rating': 8.6, 'year': 2006.0}),

Document(page_content='Three men walk into the Zone, three men walk out of the Zone', metadata={'director': 'Andrei Tarkovsky', 'genre': ['science fiction', 'thriller'], 'rating': 9.9, 'year': 1979.0})]

```
# This example specifies a query and a filter
retriever.get_relevant_documents("Has Greta Gerwig directed any movies
about women")
```

```
query='women' filter=Comparison(comparator=<Comparator.EQ: 'eq'>,
attribute='director', value='Greta Gerwig')
```

[Document(page_content='A bunch of normal-sized women are supremely wholesome and some men pine after them', metadata={'director': 'Greta Gerwig', 'rating': 8.3, 'year': 2019.0})]

```
# This example specifies a composite filter
retriever.get_relevant_documents(
    "What's a highly rated (above 8.5) science fiction film?"
)
```

```
query=' ' filter=Operation(operator=<Operator.AND: 'and'>,
arguments=[Comparison(comparator=<Comparator.EQ: 'eq'>,
attribute='genre', value='science fiction'), Comparison(comparator=
<Comparator.GT: 'gt'>, attribute='rating', value=8.5)])
    [Document(page_content='Three men walk into the Zone, three men
walk out of the Zone', metadata={'director': 'Andrei Tarkovsky',
'genre': ['science fiction', 'thriller'], 'rating': 9.9, 'year':
1979.0})]
# This example specifies a query and composite filter
retriever.get_relevant_documents(
    "What's a movie after 1990 but before 2005 that's all about toys,
and preferably is animated"
    query='toys' filter=Operation(operator=<Operator.AND: 'and'>,
arguments=[Comparison(comparator=<Comparator.GT: 'gt'>,
attribute='year', value=1990.0), Comparison(comparator=<Comparator.LT:
```

```
'lt'>, attribute='year', value=2005.0), Comparison(comparator=
<Comparator.EQ: 'eq'>, attribute='genre', value='animated')])
```

[Document(page_content='Toys come alive and have a blast doing so', metadata={'genre': 'animated', 'year': 1995.0})]

Filter k

We can also use the self query retriever to specify k: the number of documents to fetch.

We can do this by passing enable_limit=True to the constructor.

```
retriever = SelfQueryRetriever.from_llm(
    llm.
```

```
vectorstore,
document_content_description,
metadata_field_info,
enable_limit=True,
verbose=True,
)
```

```
# This example only specifies a relevant query
retriever.get_relevant_documents("What are two movies about dinosaurs")
```