Modules

Data connection

Retrievers

Self-querying

Chroma self-querying

Chroma self-querying

Chroma is a database for building AI applications with embeddings.

In the notebook we'll demo the SelfQueryRetriever wrapped around a Chroma vector store.

Creating a Chroma vectorstore

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

NOTE: The self-query retriever requires you to have lark installed (pip install lark). We also need the chromadb package.

```
#!pip install lark

#!pip install chromadb
```

We want to use OpenAIEmbeddings so we have to get the OpenAI API Key.

```
import os
import getpass

os.environ["OPENAI_API_KEY"] = getpass.getpass("OpenAI API Key:")
```

```
OpenAI API Key: ······
```

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



```
embeddings = OpenAIEmbeddings()
```

API Reference:

- Document from langchain.schema
- OpenAlEmbeddings from [langchain.embeddings.openai]
- Chroma 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": "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",
    "rating": 9.9,
    },
    ),
]
vectorstore = Chroma.from_documents(docs, embeddings)
```

Using embedded DuckDB without persistence: data will be transient

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 guery.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={'year': 1993, 'rating': 7.7, 'genre': 'science fiction'}),

Document(page_content='Toys come alive and have a blast doing so',
metadata={'year': 1995, 'genre': 'animated'}),

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='Leo DiCaprio gets lost in a dream within a

```
dream within a dream within a ...', metadata={'year': 2010, 'director':
   'Christopher Nolan', 'rating': 8.2})]

# 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={'year': 2006, 'director': 'Satoshi Kon', 'rating': 8.6}),

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'})]

```
# 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={'year': 2019, 'director': 'Greta Gerwig', 'rating': 8.3})]

```
# 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={'year': 1979, 'rating': 9.9, 'director':
'Andrei Tarkovsky', 'genre': 'science fiction'})]
```

```
# 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), Comparison(comparator=<Comparator.LT: 'lt'>,
attribute='year', value=2005), Comparison(comparator=<Comparator.EQ:
'eq'>, attribute='genre', value='animated')])
```

[Document(page_content='Toys come alive and have a blast doing so', metadata={'year': 1995, 'genre': 'animated'})]

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.

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

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

[Document(page_content='A bunch of scientists bring back dinosaurs and mayhem breaks loose', metadata={'year': 1993, 'rating': 7.7, 'genre': 'science fiction'}),

Document(page_content='Toys come alive and have a blast doing so',
metadata={'year': 1995, 'genre': 'animated'}),

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='Leo DiCaprio gets lost in a dream within a dream within a ...', metadata={'year': 2010, 'director': 'Christopher Nolan', 'rating': 8.2})]