

Hands-on Lab: Prompt + Ontology Pattern with SPARQL + Internal API

This lab covers two parts:

PART A: Implement a Prompt + Ontology Pattern Using SPARQL Queries

We will:

- Create an ontology in Protégé
 - Store it in a triple store (e.g., Apache Jena Fuseki or Stardog)
 - Run SPARQL queries to fetch data
 - Dynamically generate a prompt using the query result
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◆ Step 1: Create an Ontology in Protégé

1. Install Protégé:

Download and install from <https://protege.stanford.edu>

2. Create a new ontology (e.g., HealthcareAssistant.owl)

3. Define Classes:

- Disease , Drug , Treatment , Symptom , MechanismOfAction

4. Define Properties:

- treats (Drug → Disease)
- hasMechanism (Drug → MechanismOfAction)
- hasSymptom (Disease → Symptom)

5. Add Individuals:

- Drug: Ibuprofen
- Disease: Arthritis
- Mechanism: COX Inhibitor
- Symptom: Joint Pain

6. Save as RDF/XML or TTL format

◆ Step 2: Load Ontology into a Triple Store

Use **Apache Jena Fuseki** (easy to run locally):

```
docker run -d -p 3030:3030 --name fuseki stain/jena-fuseki
```

1. Visit <http://localhost:3030>
 2. Create a dataset (e.g., healthcare)
 3. Upload your .ttl or .owl ontology
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◆ Step 3: Query Data Using SPARQL

Example SPARQL query:

```
PREFIX ex: <http://www.example.org/ontology#>
SELECT ?drug ?mechanism
WHERE {
  ?drug ex:treats ex:Arthritis .
  ?drug ex:hasMechanism ?mechanism .
}
```

Run this from the Fuseki UI or use Python (see next step).

◆ Step 4: Generate Prompt Dynamically in Python

Install required packages:

```
pip install rdflib requests openai
```

Python code:

```
import requests
import openai

SPARQL_ENDPOINT = "http://localhost:3030/healthcare/sparql"

def query_sparql():
    sparql_query = """
PREFIX ex: <http://www.example.org/ontology#>
SELECT ?drug ?mechanism
WHERE {
  ?drug ex:treats ex:Arthritis .
  ?drug ex:hasMechanism ?mechanism .
}
"""

    response = requests.post(SPARQL_ENDPOINT, data={'query': sparql_query},
                             headers={'Accept': 'application/sparql-
results+json'})
    data = response.json()
    bindings = data['results']['bindings']
    return [(b['drug']['value'].split('#')[-1], b['mechanism']
            ['value'].split('#')[-1]) for b in bindings]

def generate_prompt():
    pairs = query_sparql()
    for drug, mechanism in pairs:
        prompt = f"Explain how {drug} treats arthritis by acting as a
{mechanism}."
```

```
    return prompt

print(generate_prompt())
```

✓ PART B: Build an Internal API to Expose Graph-Based LLM Capabilities

We will now create a REST API using **FastAPI** to expose this functionality.

◆ Step 5: Build a FastAPI Server

Install FastAPI:

```
pip install fastapi uvicorn
```

Create `main.py`:

```
from fastapi import FastAPI
from fastapi.responses import JSONResponse
import requests

app = FastAPI()

@app.get("/generate-prompt")
def get_prompt():
    sparql_query = """
PREFIX ex: <http://www.example.org/ontology#>
SELECT ?drug ?mechanism
WHERE {
    ?drug ex:treats ex:Arthritis .
    ?drug ex:hasMechanism ?mechanism .
}
"""

    response = requests.post("http://localhost:3030/healthcare/sparql",
                            data={"query": sparql_query},
                            headers={"Accept": "application/sparql-
results+json"})

    data = response.json()
    pairs = [(b['drug']['value'].split('#')[-1], b['mechanism']
              ['value'].split('#')[-1])
              for b in data['results']['bindings']]
    prompt = f"Explain how {pairs[0][0]} treats arthritis by acting as a
{pairs[0][1]}."
    return JSONResponse(content={"prompt": prompt})
```

Run the API:

```
uvicorn main:app --reload
```

Test the endpoint:

```
curl http://localhost:8000/generate-prompt
```

Bonus: Call the API in ChatGPT or LLM Workflow

Now this API can be used in:

- LangChain tools
 - Agentic frameworks (CrewAI / AutoGen)
 - Internal chatbots
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What You Learned:

- How to link ontologies with SPARQL-driven prompt construction
 - How to serve that logic via a REST API
 - How to power GenAI apps with dynamic, explainable knowledge
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