



# Assignment: Development of an Autonomous QA Agent for Test Case and Script Generation

## Objective

Build an intelligent, autonomous QA agent capable of constructing a “testing brain” from project documentation. The system will ingest support documents (e.g., product specifications, UI/UX guidelines, mock APIs) alongside the HTML structure of a target web project. Using these inputs, the agent should:

- **Generate Test Cases** — Produce comprehensive, documentation-grounded test plans and test viewpoints.
- **Generate Selenium Test Scripts** — Convert the generated test cases into executable Python Selenium scripts for automated testing.

The backend must be implemented using **FastAPI or Flask**, with a **Streamlit** user interface. All test reasoning must be grounded strictly in the provided documents—no hallucinated features.

## Submission Guidelines

Candidates must provide:

1. **Source Code Repository** (GitHub/GitLab/etc.)
2. **README.md** including:
  - Setup instructions (Python version, dependencies)
  - How to run FastAPI/Flask + Streamlit
  - Usage examples
  - Explanation of the included support documents
3. **Project Assets**
  - The checkout.html file
  - The 3–5 support documents used.
4. **Demo Video (5–10 minutes)** demonstrating:
  - Uploading documents + HTML
  - Building the knowledge base
  - Generating test cases
  - Selecting a test case
  - Generating Selenium scripts



## Evaluation Criteria

### 1. Functionality

- Does the system fulfill all phases (ingestion → test cases → script generation)?

### 2. Knowledge Grounding

- Are test cases based strictly on provided documents?
- No hallucinations or fabricated features.

### 3. Script Quality

- Are Selenium scripts clean, correct, and runnable?
- Do selectors match the actual HTML?

### 4. Code Quality

- Modular, readable, well-structured code
- Clean backend with FastAPI/Flask + Streamlit interface

### 5. User Experience

- Simple, intuitive UI
- Clear system feedback (e.g., “Knowledge Base Built,” “Generating Script...”)

### 6. Documentation

- Clear, detailed README.md
- Instructions for setup, dependencies, and usage



## Project Assets (To Be Created or Provided)

Your assignment will focus on a simple, single-page web application and a set of support documents.

### 1. Target Web Project (checkout.html)

A single-page “**E-Shop Checkout**” HTML file containing:

#### *Features (5–10 items)*

- 2–3 items with “**Add to Cart**” buttons
- A **cart summary** section with item quantity inputs and total price
- A **discount code** input field
- A **User Details form** (Name, Email, Address)
- **Form validation** with inline error messages (e.g., invalid email, required fields)
- **Shipping method** radio buttons (Standard, Express)
- **Payment method** radio buttons (Credit Card, PayPal)
- A “**Pay Now**” button that, if the form is valid, displays “Payment Successful!”

### 2. Support Documents (3–5 files)

Examples include:

#### 1. **product\_specs.md**

- Contains feature rules—for example:
  - “The discount code SAVE15 applies a 15% discount.”
  - “Express shipping costs \$10; Standard shipping is free.”

#### 2. **ui\_ux\_guide.txt**

- Contains UI/UX guidelines—for example:
  - “Form validation errors must be displayed in red text.”
  - “The ‘Pay Now’ button should be green.”

#### 3. **api\_endpoints.json (optional but recommended)**

- Example:

```
{  
  "POST /apply_coupon": {"code": "string"},  
  "POST /submit_order": {"name": "string", "email": "string"}  
}
```



## Functional Requirements

### Phase 1: Knowledge Base Ingestion & UI (Streamlit)

#### Required UI Features

The Streamlit UI must allow users to:

1. **Upload support documents** (MD, TXT, JSON, PDF, etc.)
2. **Upload or paste the checkout.html file**
3. Click “Build Knowledge Base”

#### Content Parsing

Use appropriate libraries to extract text from documents, such as:

- unstructured
- pymupdf (fitz)
- Custom parsers for JSON, HTML, etc.

#### Vector Database Ingestion

- Implement text chunking (e.g., **RecursiveCharacterTextSplitter**)
- Preserve metadata (e.g., "source\_document": "product\_specs.md")
- Generate embeddings using a model such as those from Hugging Face
- Store vectors + metadata in a vector DB (Chroma, FAISS, Qdrant, etc.)

### Phase 2: Test Case Generation Agent

#### UI

Provide an “Agent” section where the user can request test cases.

Example prompt:

“Generate all positive and negative test cases for the discount code feature.”

#### RAG Pipeline

1. Embed the user’s query
2. Retrieve relevant chunks from the vector database
3. Feed retrieved context + user query into an LLM (Ollama, Groq, local HF model, etc.)

#### LLM Output Requirements

The agent must respond with **structured test plans** in JSON or Markdown table format.



### *Example Test Case Output*

Test\_ID: TC-001

Feature: Discount Code

Test\_Scenario: Apply a valid discount code 'SAVE15'.

Expected\_Result: Total price is reduced by 15%.

Grounded\_In: product\_specs.md

*All test reasoning must reference the source document(s).*

## Phase 3: Selenium Script Generation Agent

Extend the UI to allow:

- Selecting one of the generated test cases
- Clicking “**Generate Selenium Script**”

### Agent Logic

The agent must:

1. Receive the selected test case
2. Retrieve the full content of **checkout.html**
3. Retrieve relevant documentation snippets from the vector DB
4. Use an LLM to generate a **runnable Selenium Python script**

### Prompt Requirements

The LLM must be instructed to:

- Act as a **Selenium (Python) expert**
- Use appropriate selectors (IDs, names, CSS selectors) based on the actual HTML
- Produce high-quality, fully executable code

### Output

Display the generated Python script in a code block for easy copying.