

# Implementation Plan: Brain Sparks Educational Recommender

Rugogamu Noela S23B8/016 B22775

Institution: Uganda Christian University (UCU)

December 2025

## 1. Project Overview

### 1.1 Project Name

**Brain Sparks** - Personalized Educational Recommender for Ugandan Learners

### 1.3 Project Scope

- Natural language query processing (Understand)
- Knowledge graph construction and traversal (Reason)
- Content based recommendation engine (Reason)
- User feedback integration (Learn)
- Web based interactive interface (Interact)

## 2. Work Breakdown Structure (WBS)

### Level 1: Project Phases

Brain Sparks Project

- 1.0 Problem Analysis & Design
- 2.0 Cognitive System Implementation
- 3.0 Evaluation & Documentation
- 4.0 Deployment & Presentation

### Level 2: Work Packages

#### 1.0 Problem Analysis & Design (Part A - Week 1)

Table 1: WBS for Part A

WBS Code	Task	Deliverable	Days
1.1	Problem Analysis	<code>problem_analysis.pdf</code>	2
1.2	Data Pipeline & Content Curation	<code>01_data_pipeline.ipynb</code> , Dataset	2
1.3	NLP Engine Implementation (Understand)	<code>nlp_utils.py</code> (Topic, Intent, Context)	3

#### 2.0 Cognitive System Implementation (Part B - Week 2)

Table 2: WBS for Part B and Evaluation

WBS Code	Task	Deliverable	Duration (Days)
2.1	Knowledge Graph (Reason)	<code>kg_utils.py</code> , KG	2
2.2	Recommender & Feedback (Learn)	<code>recommender.py</code> , <code>feedback.json</code>	2

Table 2 – continued

WBS Code	Task	Deliverable	Duration (Days)
2.3	Interactive UI (Interact)	app.py (Streamlit app)	2
3.1	Evaluation & Testing	evaluation_report.pdf, 20 Test Queries	2
3.2	Final Documentation	ethical_analysis.pdf, GitHub Upload	1

### 3. Project Timeline (Actual Execution)

**Actual Execution Dates: November 18 – December 1, 2025**

The project was executed in a two week sprint, structured around the four cognitive pillars.

#### 0.1 Week 1: Foundations and The Understand Pillar (Approx. Nov 18-24)

- **Days 1-2 (Nov 18-19):** Problem Analysis (`problem_analysis.pdf`) and defining the Cognitive Pipeline.
- **Days 3-4 (Nov 20-21):** Dataset acquisition, cleaning, and transformation using `01.data_pipeline.ipynb`. This included curating 87 educational resources across 20 topics.
- **Days 5-7 (Nov 22-24):** Implementation of the Understand pillar in `nlp_utils.py`. This module handles Natural Language Processing (NLP) to extract Topic, User Intent (such as 'learn', 'apply'), and Uganda Context.
- **Visualization:** The initial data distribution was visualized:

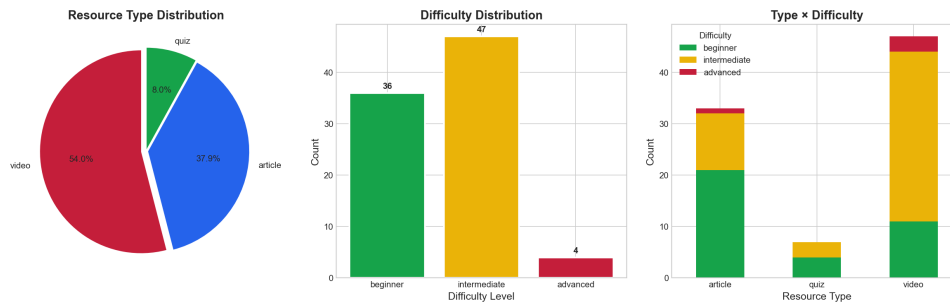


Figure 1: Educational Resources by Topic (Horizontal bar chart ranking topics by count, with Machine Learning at 9, Web Development at 8, Artificial Intelligence at 7, and others down to 2).

#### 0.2 Week 2: Reasoning, Learning, Interaction, and Evaluation (Nov 25-Dec 1)

- **Day 8-9 (November 25-26): Reasoning Pillar and Knowledge Graph (Milestone 3)**
  - Implementation of the **Reason** pillar in `kg_utils.py` using `NetworkX`. The `EducationalKnowledgeGraph` class was created, adding nodes for topics, resources, and applications, and edges like 'is\_about', 'applies\_to', and prerequisites.
  - The final graph size included **335 Total Nodes** and **1028 Total Edges**.

```

...
Cybersecurity --[has_subtopic]--> Auditing
Cybersecurity --[has_subtopic]--> Security
Cybersecurity --[has_subtopic]--> Basics
and 23 more

```

Figure 2: Relationship extraction: A snippet illustrating the semantic relationships extracted and stored in the graph, specifically showing 'Cybersecurity' connected to subtopics like 'Auditing', 'Security', and 'Basics' via the `--[has_subtopic]-->` edge.

```

Knowledge Graph Built with Entity Extraction!
=====

Graph Statistics:
  • Total Nodes: 335
  • Total Edges: 1028
  • Topics: 240
  • Resources: 87
  • Applications: 8
  • Density: 0.009187594959335061

Sample Topics in Graph:
  • Accessibility
  • Africa
  • Agile
  • Agriculture
  • Ai Diagnosis
  • Algorithms
  • Api
  • Applications
  • Architecture

```

Figure 3: A console output detailing the structure and size of the Knowledge Graph built via entity extraction. Key metrics include 335 Total Nodes, 1028 Total Edges, and a list of sample topics like 'Agriculture', 'Algorithms', and 'Accessibility'.

- **Day 10-11 (November 27-28): Recommendation and Learning Pillar**

- Creation of `recommender.py` using **\*\*TF-IDF\*\*** for content-based similarity scoring, integrated with knowledge graph traversal for context.

- Implementation of the **\*\*Learn\*\*** pillar using **FeedbackManager**, which stores user ratings in **feedback.json** and adjusts resource scores (boosting popular resources by up to 0.3).
- **Day 12-13 (November 29-30): Interaction Pillar and System Integration (Milestone 4)**
  - Development of the user interface in **app.py** using **\*\*Streamlit\*\***. The UI includes: home (query input), results (structured learning paths with Explanation and Uganda relevance tabs), feedback submission, knowledge graph visualization (**PyVis**), and a statistics dashboard.
- **Day 14 (December 1): Evaluation, Refinement, and Documentation (Milestone 5)**
  - Generation of 20 test queries to measure performance metrics like **precision@3** using a **compare\_with\_baseline** function.
  - Visualization of results using **Plotly** charts.
  - Finalization of all reports (**evaluation\_report.md**, **ethical\_analysis.md**) and code upload to GitHub.

```

Testing Multiple Queries
=====

Query: "I want to learn machine learning for agriculture"
  → Topic: educational_technology (67%)
  → Uganda: Yes
  → Intent: learn

Query: "How can cybersecurity help protect mobile money in Uganda?"
  → Topic: fintech (67%)
  → Uganda: Yes
  → Intent: solve

Query: "Teach me about blockchain applications"
  → Topic: mobile_development (67%)
  → Uganda: No
  → Intent: learn

Query: "What is artificial intelligence?"
  → Topic: artificial_intelligence (33%)
  → Uganda: No

```

Figure 4: Multiple Query Analysis Results. Examples demonstrating the system’s ability to analyze and categorize multiple user queries. Each query is mapped to a Topic, checked for Uganda Context, and assigned an Intent (such as ‘learn’ or ‘solve’).

## 4. Ethical Considerations

Throughout the implementation, ethics were prioritized by ensuring no personal data collection, maintaining a balanced dataset for fairness, and utilizing simple, accessible tools to keep the system available to all Ugandan learners.