aiVOLUTION 2025 - Idea Submission Template

Title

MediaGraphAI: A Multimodal Knowledge Graph for Movies

Overview (max 300 words)

Today, entertainment discovery is limited to keywords, genres, or ratings (e.g., "action movies" or "movies by Tolkien"). But people often think in richer, story-driven terms:

"Show me movies where the hero wears green pants and is betrayed by their best friend."

MediaGraphAI solves this by building an AI-powered multimodal knowledge graph that ingests movies to extract fine-grained entities (characters, objects, themes, emotions, scenes) and their relationships. Using AI models for text, vision, and audio, we convert raw scripts, subtitles, and clips into structured graph data.

Core features:

- Natural language search across movies.
- Scene-level tagging (actions, moods, objects).
- Semantic similarity queries (e.g., "movies that feel like Spirited Away").
- Knowledge graph visualization for exploration.

The result is a smarter IMDb + Wikipedia mashup—a tool for discovery, analysis, and creative research.

AI/LLM Use Case

AI powers every layer of the system:

- Text (Scripts/Books): spaCy + GPT-4/gemini for entity/relation extraction.
- Vision: CLIP + YOLO for scene/object tagging.
- Audio: Whisper for ASR to extract dialogues.
- **LLM Querying:** GPT/gemini translates natural language into graph queries (Cypher/GraphQL).
- Embeddings: Vector search for "vibe" or thematic similarity.

LLMs act as the "semantic glue" connecting symbolic graph data with fuzzy, human-style queries.

Feasibility Plan

Defined Scope:

We will focus only on the smallest working slice of the project that still demonstrates the core idea:

1. Data Ingestion

○ Top 50 movies (via TMDB + OpenSubtitles).

2. Entity Extraction

 Use spaCy + GPT/Gemini API to extract characters, locations, events, add themes from subtitles and book text.

3. Graph Storage

- Store entities and relationships in Neo4j with a simple schema.
- Query via Cypher/GraphQl.

4. Minimal UI

 React frontend with a search box and graph visualization (Neo4j Browser integration or lightweight D3.js).

5. Demo Queries

- "Find movies about betrayal."
- "Books where Alice is the protagonist."

Outcome: A working prototype (text-first Knowledge Graph + query interface) that is demo-ready.

Why Achievable in 8-10 Hours:

- **Tightly Scoped**: We limit to *text-only pipeline* (no video/audio analysis in MVP).
- ullet Use of Existing APIs: TMDB, OpenSubtitles, Project Gutenberg \to no manual dataset prep.
- ullet **Pre-built NLP Models:** SpaCy + GPT/Gemini for NER and summaries \to no training required.
- Off-the-shelf Graph DB: Neo4j has quick setup and built-in visualization → avoids custom infra.
- Minimal Frontend: Only search + graph visualization, no complex UI features in MVP.

Team Allocation (3 Members):

- Frontend + Integration (1 member)
- Backend + AI/LLM (2 members)

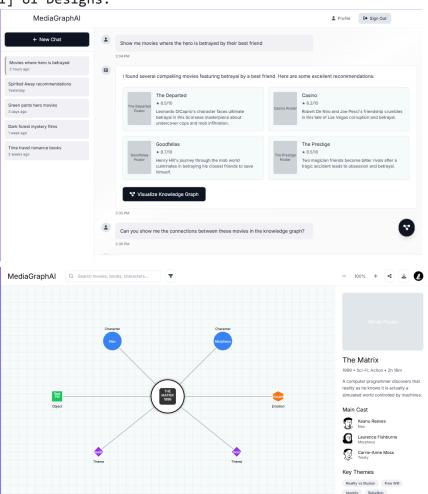
Constraints & Risk Management:

- ullet Time Constraint: 8-10 hours o only core features, leave advanced multimodal analysis for Iterations 2 & 3.
- API Dependence: Rely on public/free APIs (TMDB, OpenSubtitles, Gutenberg). If API limits hit → fall back to cached samples.
- LLM Cost/Latency: Keep GPT calls limited to short summaries/NER, not entire scripts.

Conclusion: With scoped-down objectives, proven APIs, and a clear team split, our MVP is **realistically achievable in 8-10 hours**, with optional Iterations 2 & 3 for polish and wow-factor if time allows.

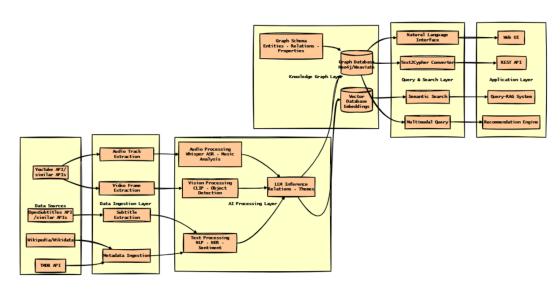
Final Product UX

1] UI Designs:

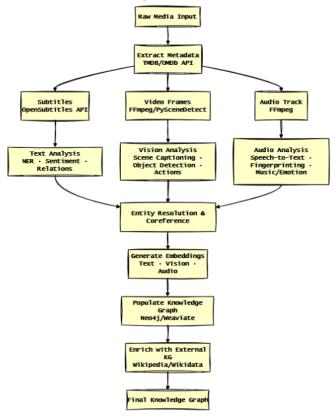


2]Do check our Full UI Collection (Clickable Prototype), made using UX Pilot: Complete Clickable Prototype
3]System Flowcharts:

I] System Architecture-



II] Data ingestion flowchart-



Minimum Viable Product (MVP) Definition

Core Functionalities (Hackathon Priority - 8-10 hrs)

- 1. Data Ingestion
 - Import top 50 movies (TMDB + OpenSubtitles).
- 2. Entity Extraction
 - Use spaCy + GPT/Gemini API to extract key entities: characters, locations, events, and themes.
- 3. Graph Storage & Query
 - Store extracted entities and relationships in Neo4j.
 - Enable querying through Cypher.
- 4. Minimal UI
 - A simple React search box for natural language queries.
 - o Basic graph visualization to display entity relationships.

Outcome: A functional knowledge graph where users can run queries like "Movies with betrayal theme" or "movies where Alice is the protagonist."

Additional Features (If Time Allows)

Iteration 2 - Enrichment & Polish (+2-3 hrs)

- Data: Add trailers for 1-2 movies via YouTube API.
- Extraction: Include emotional themes (sentiment analysis/GPT tags); run Whisper on a short clip to generate transcripts.
- Graph: Add scene-level nodes for 1-2 movies (chapters/scenes).
- **UI:** Pre-built query buttons (e.g., "Find betrayals," "Find sad scenes"); show results as cards with posters/snippets.
- **Demo Queries:** "Show me sad fight scenes" or "Which movies feel similar to Spirited Away?"

Outcome: A richer, more polished graph with emotional themes, scene-level detail, and improved usability.

Iteration 3 - Stretch / Wow Features (+2-3 hrs)

- **Vision:** Apply CLIP on frames to auto-tag objects/clothing; detect 1-2 iconic fight scenes via pretrained action classifiers.
- Audio: Prototype "Shazam for movies" by matching audio clips with Whisper transcripts.
- ullet UI Polish: Add scene timeline with clickable nodes; allow screenshot upload \to return closest scene.
- **Demo Queries:** "Find movies with rainy fight scenes" or "Upload an image to find the matching scene."

Outcome: A multimodal demo combining text, audio, and vision for a strong wow factor.