

Revitalizing Rural Cinema: A Data-Driven Recommendation Engine

Market Analysis & Technical Strategy

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The Mission



Context



The Client:

A local cinema in La Creuse, Nouvelle-Aquitaine.



The Problem:

Declining attendance and a need for digital transformation.



The Goal:

1. Analyze the local market (Cold Start).
2. Build a "testable" movie recommendation engine.
3. Create a dashboard for decision-making.



Key Insight: We are not building Netflix; we are building a tool for a specific, aging, rural demographic.

Market Research - The “Creuse” Context

Demographics



Underpopulated region (115,529 inhabitants).



Aging population: Over 45s make up >50% of the populace.



Primary audience: Retirees and families living in couples.

Cinema Landscape



Small ecosystem: Only 9 cinemas in the region; absence of large franchises.



Current Status: Low attendance indices, though post-COVID numbers are rising.

User Preferences: The “Why” behind the Data

Content is King



French Films: Perform exceptionally well locally.



Language Barrier: 70% preference for “Version Française” (VF).



Top Genres: Strong preference for **Drama** and **Comedy** (specifically French humor).

Behavior



Cinema is viewed as a **social activity** “good for morale” (9 French out of 10), driven by trailers and internet searches.

From Insights to Strategy

The “Cold Start” Strategy



Since we have no user history, we use **Content-Based Filtering** tuned to our Market Research.

Data Cleaning Rules



Language Weight: Prioritize films with **original_language = 'fr'** or major blockbusters likely to be dubbed.



Genre Focus: Boost weights for '**Drama**' and '**Comedy**' in the vectorization process.



Era Sensitivity: Do not aggressively penalize older films (**1980-2000**) given the 45+ demographic.

Technical Architecture

Data Source:

 **IMDb**
(7M+ rows)

 **TMDB**
(Posters/Budgets)



ETL Pipeline:



Chunking
Chunking massive datasets to manage memory.



Filtering
Filtering non-relevant content (shorts, adult, unreleased).



The Engine:



Library: Scikit-Learn
(NearestNeighbors or CosineSimilarity).

Features: Genres, Actors, Directors, Keywords.

The App:



Streamlit
(Recommendation, Dashboard)

Roadmap & Next Steps



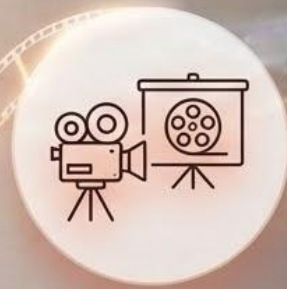
Phase 1 (Done): ✓
Market Research &
User Persona
Definition.



Phase 2 (Current): ((▶))
Data Cleaning &
Feature Engineering
(Pandas).



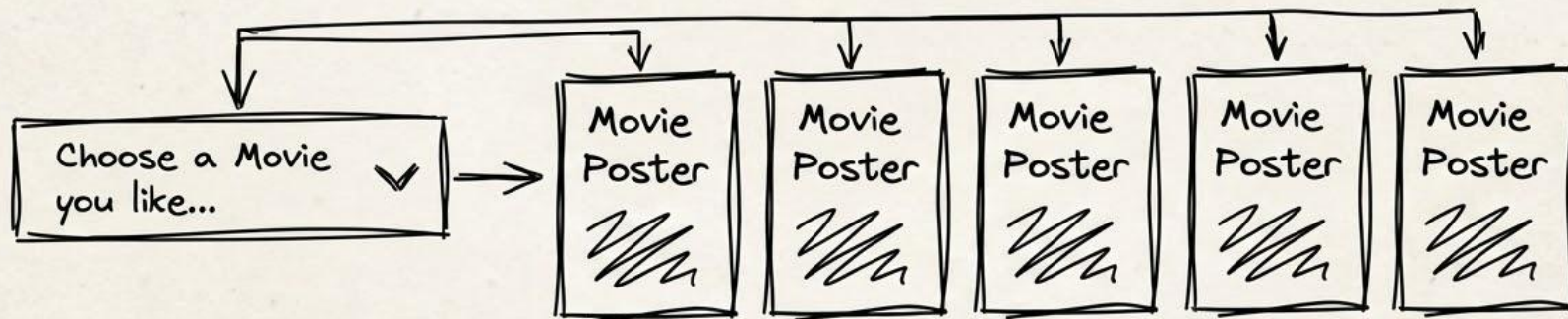
Phase 3:
Model Training &
Streamlit Interface
Development.



Phase 4:
Client Demo
(using *Anatomy of
a Fall* as the
calibration test).

The Interface Concept

70% French Content.



Thank you

