

# QueryTube: AI\_SemanticSearchTube: Building a Semantic Search App with YouTube Data

## Project statement:

The goal of this project is to build a YouTube semantic search system by extracting and transforming video data (title, date, transcript) via YouTube APIs. The system will allow users to input natural language queries and receive the top-5 most semantically relevant video titles or IDs. Interns will gain hands-on experience in NLP, data engineering, and information retrieval by building a complete semantic search engine.

## Outcomes:

- Extract and preprocess metadata and transcripts from YouTube videos using APIs.
- Understand and apply transformer-based text embedding models.
- Evaluate similarity and distance metrics for effective semantic retrieval.
- Build and deploy a complete semantic video search pipeline.

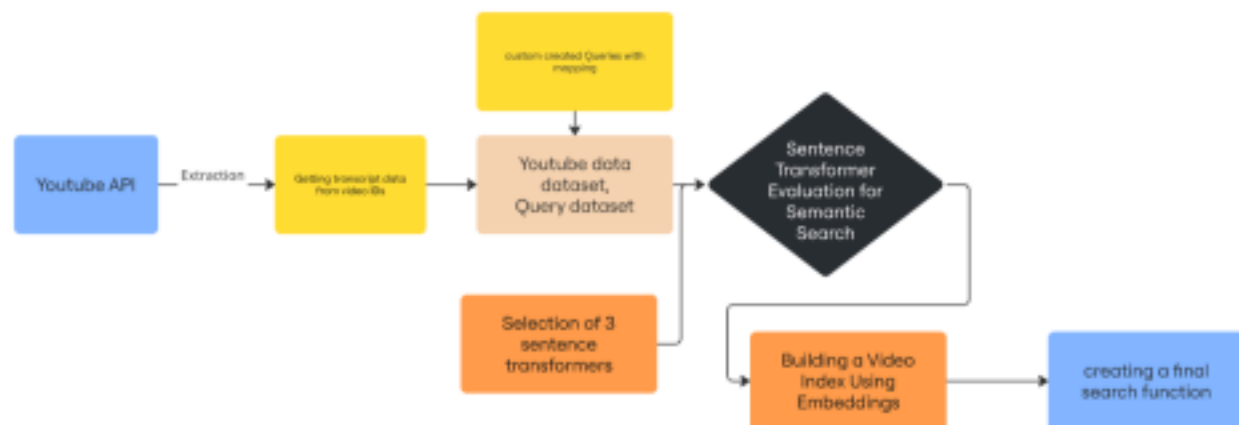
## Milestones:

Milestone 1: YouTube Data Collection and API Mastery (Weeks 1–2)

Milestone 2: Transcript Extraction and Data Cleaning (Weeks 3–4)

Milestone 3: Sentence Transformer Evaluation for Semantic Search (Week 5-6)

Milestone 4: Implementing and Tuning Semantic Search function (Week 7-8)



Milestone 1: YouTube Data Collection and API Mastery (Weeks 1–2)

## Module-1: YouTube API Fundamentals and Video Metadata Extraction

## Tasks:

- Set up a Google Cloud project and generate an API key
- Choose a YouTube channel (with 100–350 diverse videos; avoid repetitive content like stock updates)
- Learn how to use the endpoint: <https://www.googleapis.com/youtube/v3/search>

#### **Key Parameters:**

- o channelId: ID of the selected channel
- o maxResults: Up to 50 per request
- o part: ["snippet", "id"]
- o order: "date"
- o pageToken: for pagination

#### **Deliverables:**

- Python script to extract all videos from a selected channel using pagination logic
- Custom function to extract video ID, published date, and title.

#### **Learning Outcomes:**

- Understand the YouTube Data API v3
- Authenticate and interact with the API using an API key
- Fetch video metadata programmatically

## **Module 2: Exploratory Data Analysis (EDA) and Structuring Video Metadata**

#### **Tasks:**

- Store extracted data in Polars or Pandas dataframe
- Analyze:
  - o Row and column uniqueness
  - o Title distributions
  - o Publish frequency over time
- Perform EDA on metadata (date distribution, title uniqueness, missing values)

#### **Deliverables:**

- Notebook or script with exploratory charts/tables
- Cleaned video metadata dataset

## **Milestone 2: Transcript Extraction and Data Cleaning (Weeks 3–4)**

## **Module 3: Extracting Video Transcripts with YouTube Transcript API**

**Tasks:**

- Use the `youtube_transcript_api` Python package to fetch auto-generated captions or transcripts
- Use appropriate methods to collect and organize transcript text
- Identify and log videos without available transcripts

**Deliverables:**

- Dataframe enriched with a transcript column
- Transcript extraction logic and logs of failed extractions

**Module 4: Cleaning and Normalizing Transcripts****Tasks:**

- Standardize titles and transcripts by removing or replacing special characters
  - Handle null or missing transcript entries
  - Generate an initial list of ~70–80 search queries (topics, keywords, or phrases relevant to the selected videos) and map them to
    - Familiarize with at least three SentenceTransformer models (e.g., `all-MiniLM-L6-v2`) from `hugging face` or <https://sbert.net/>
  - Understand sentence-transformers and their role in semantic search. •
- Prepare evaluation setup to compare semantic similarity performance

**Deliverables:**

- Cleaned dataset with `video_id`, `title`, `datetime`, `transcript`
- Preliminary set of evaluation queries for Week 5
- Summary of semantic search understanding and model selection rationale

**Milestone 3: Sentence Transformer Evaluation for Semantic Search(Week 5-6)****Module 5: Cleaning and Normalizing Transcripts****Tasks:**

- Use SentenceTransformer to embed video titles and transcripts using three candidate models
- Use the pre-defined 70–80 search queries and embed them

- Compare distance-based (euclidean, manhattan, chebyshev) and similarity-based (cosine similarity, dot product) ranking methods
- Evaluate similarity between queries and video transcripts/titles and check how well each model ranks the correct video for each query

#### **Deliverables:**

- Evaluation summary: model performance across metrics (e.g., average rank, top-1, top-3 recall)
- Identify the best-performing model and method for semantic retrieval
- Visual or tabular comparison of results across models

### **Module 6: Building a Video Index Using Embeddings**

#### **Tasks:**

- Choose the best SentenceTransformer model from Week 5
- Embed the titles and transcripts of each video
- Concatenate and append these embeddings to the original dataset
- Save the final dataframe using Polars or pandas

#### **Deliverables:**

- Polars or pandas dataframe with ~768+ embedding features per video
- Parquet or csv file storing complete video index for search
- Sample visualization of embedding structure or dimensionality reduction (optional)

Milestone 4: Implementing and Tuning Semantic Search and creating a final search function (Week 7-8)

### **Module 7: Optimizing result quality**

#### **Tasks:**

- Load the video index and selected transformer model
- Encode incoming user query
- Use distance metrics (e.g., manhattan, euclidean, cosine) from sklearn to compute similarity between query embedding and stored vectors
- Rank the closest matches using title and transcript embeddings jointly
- Understand thresholds, top\_k rankings, and filtering logic
- Tune the threshold and top\_k to optimize result quality

#### **Deliverables:**

- Working search function returnSearchResults(query, df)
- Evaluation of different thresholds and distance types
- Sample query-to-result demo in notebook format

## **Module 8: Final Deployment & Search Interface**

### **Tasks:**

- Use the optimized model + distance metric pair
- Create a Python script or notebook that loads the index, runs query search, and returns top-5 matches
- Build an interactive Gradio interface with embedded video previews and markdown descriptions.
- Prepare summary presentation and code documentation.

### **Deliverables:**

- Final query-to-top-5 search engine function
- Gradio interface that takes a query and displays top-5 embedded YouTube videos
- Code submission on GitHub.