

# Agenda



coursera

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- DATA PREPROCESSING
- MODELING AND RESULTS
- DEPLOYMENT

## INTRODUCTION



- - By leveraging data analysis, machine learning and Natural language processing algorithms, we streamline course discovery, empowering users to engage in lifelong learning and skill development by introducing recommendation based on precedence.

Our Course Recommendation System delivers

personalized course suggestions to users on

experience and maximizing learning outcomes

platforms like Coursera, enhancing user

through targeted recommendations.



## ABOUT DATA



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- Data source: Kaggle & Cousera
   <a href="https://www.kaggle.com/datasets/elvinrustam/c">https://www.kaggle.com/datasets/elvinrustam/c</a>
   oursera-dataset
- Data size: 52.8 MB
- Data shape: 8370 rows x 13 columns
- Columns Used: Course Title, What will you learn, Skill gained, Keyword, level, Rating.
- Data Types: Object, Float64, Int64
- **Definition:** The above cloumns will be used to find the similarity tags which helps the user to find the similar courses for his field in a priority list.

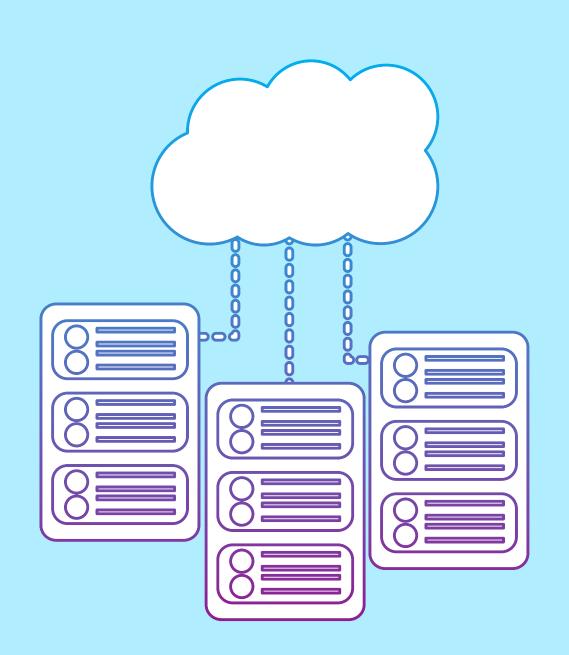
## PROJECT WORKFLOW



- The steps involved in building the recommendation system
  - Importing Libraries
  - Loading and Analyzing Data
  - Data Pre-Processing
  - Text Vectorization
  - Stemming Process
  - Similarity Measure
  - Recommendation Function
  - GUI Deployment



# DATA PREPROCESSING



- Importance of data preprocessing: Essential for improving the accuracy and efficiency of the recommendation model by ensuring that the input data is clean, relevant, and standardized.
- Steps involved in data preprocessing:
  - Cleaning data
  - Handling missing values
  - Feature selection
- Text preprocessing (Text vectorization, stemming, etc.)



## MODELING

- 1. **Model Overview**: Uses a content-based filtering approach, combining course attributes into a single "tags" feature for recommendation.
- 2.**TF-IDF Vectorization**: Converts text data into numerical vectors based on term importance, balancing term frequency and document rarity.
- 3. Cosine Similarity: Measures similarity as the cosine of the angle between course vectors, where values closer to 1 indicate higher similarity.
- 4. Recommendation Function: Accepts a course title, finds its vector, computes similarity scores with all courses, and outputs the most similar courses.



#### RESULTS

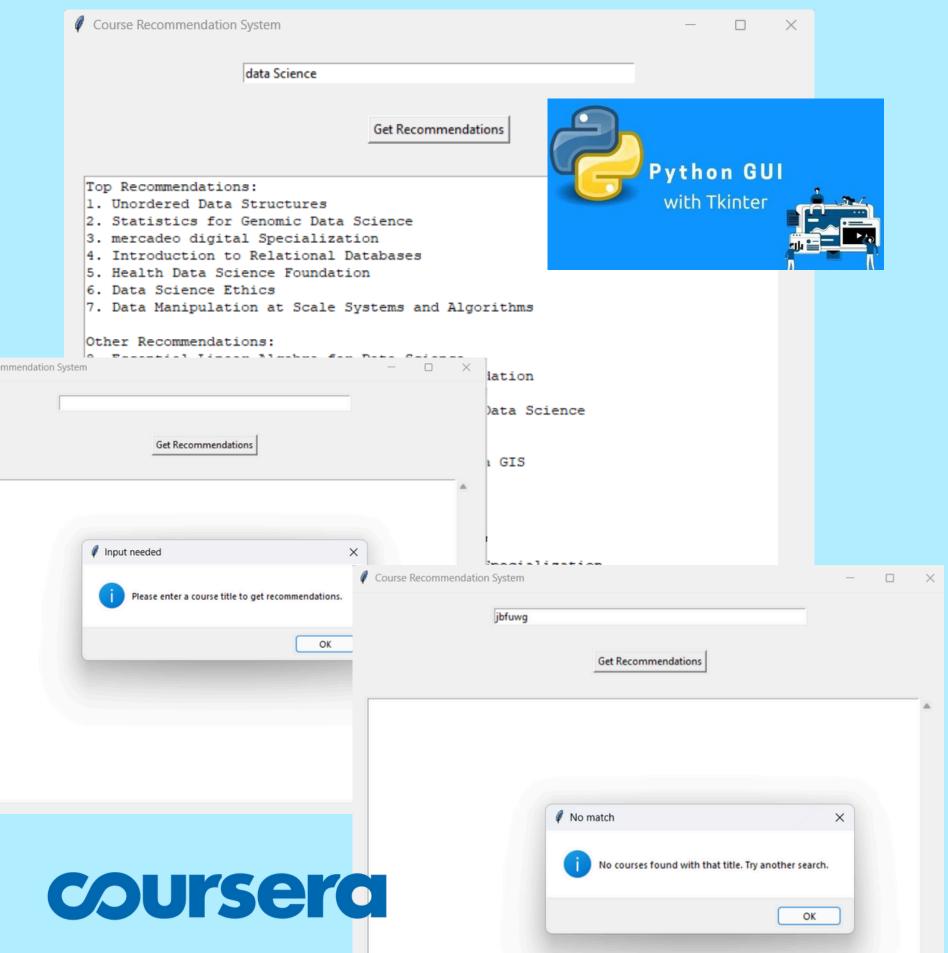
 Cosine Similarity Scores for the recommendation function to measure how similar the recommended courses are to the query.

1 recommend('Pixel Art for Video Games')

Concept Art for Video Games
Low Poly Art For Video Games
Character Design for Video Games
Art for Games Specialization
World Design for Video Games
Art and the Digital Transformation



## DEPLOYMENT



- Used Tkinter GUI library as it is widely used for developing desktop applications that are platformindependent and come with a rich set of widgets for creating interactive user interfaces and can run in every OS
- used (tk.Entry) for search and (tk.button) to trigger the framework taking entry as processing and matching logic
- We used scrolled.text logic to get the **Display** recommendation where it displays top 7
   recommendations followed by next in a priority
   using similarity matrix

