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DEPARTMENT OF ARTIFICIAL INTELLIGENCE & DATA SCIENCE

COURSE NAME: 19AD701 RECOMMENDER SYSTEMS

IV YEAR /VII SEMESTER

Unit 2- CONTENT-BASED FILTERING

Topic 2: High Level Architecture of Content based Systems





A content-based system is a type of recommendation system that suggests items to users based on the attributes and characteristics of the items themselves and the preferences of the users.

➤ Here's a high-level architecture of a content-based recommendation system:

➤ User Interface: This is the part of the system that interacts with users. It can be a website, mobile app, or any other user-facing application. Users provide input and receive recommendations through this interface.





➤ User Profile: The system maintains a user profile for each user. This profile contains information about the user's preferences, past interactions, and behavior within the system. It can be built over time as the user interacts with the system, providing explicit feedback (e.g., ratings) or implicit feedback (e.g., clicks, views).

➤ Item Database: The system has a database of items that it can recommend to users. These items can be products, articles, movies, or any other type of content. Each item is associated with a set of attributes or features that describe it, such as keywords, categories, tags, or metadata.





➤ Content Analyzer: This component is responsible for processing and analyzing the content of items in the database. It extracts relevant features from the items. For example, in a movie recommendation system, the content analyzer might extract attributes like genre, director, actors, and plot keywords.

➤ User Preference Model: Using the user's profile and the item features, this component builds a model of the user's preferences. It assigns weights or importance values to different attributes based on the user's historical interactions and preferences. For instance, if a user has shown a strong preference for action movies, the preference model might give a high weight to the "action" genre attribute.





➤ Recommendation Engine: The recommendation engine is responsible for generating personalized recommendations for each user. It uses the user's profile and the item features to calculate a relevance score for each item. Items with higher relevance scores are recommended to the user. Common recommendation algorithms include cosine similarity, TF-IDF (Term Frequency-Inverse Document Frequency), and machine learning models.

Filtering and Ranking: After generating a list of recommended items, the system may apply additional filters or ranking algorithms. This step helps ensure that the most relevant and appropriate items are presented to the user. Filters could include removing items the user has already interacted with or filtering out items that don't meet certain criteria.

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➤ Presentation Layer: The final list of recommended items is presented to the user through the user interface. The presentation layer formats the recommendations and displays them to the user, often with additional information about each item.

➤ Feedback Loop: To continually improve the system's performance, it's essential to collect user feedback. Users can provide explicit feedback by rating items or implicit feedback through their interactions. This feedback is used to update the user's profile and refine the recommendation algorithms.





➤ Monitoring and Evaluation: The system should be monitored for performance and evaluated regularly to ensure that it's providing accurate and relevant recommendations. A/B testing, user surveys, and metrics like click-through rate and conversion rate can be used for evaluation.

➤In summary, a content-based recommendation system leverages user profiles, item attributes, and content analysis to provide personalized recommendations. It involves components for user modeling, item analysis, recommendation generation, and user feedback. Regular monitoring and evaluation are critical to maintaining and improving system performance.