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**DEPARTMENT OF ARTIFICIAL INTELLIGENCE AND DATA
SCIENCE**

Recommender System



User Profiles in Recommender Systems



- User profiles play a pivotal role in personalized recommendations. A user profile is a representation of an individual's preferences, interests, and behavior within a recommender system.
- By capturing and analyzing user profiles, recommender systems can provide tailored recommendations that align with each user's unique tastes.



User Profiles in Recommender Systems



- Here's how user profiles are typically used:
 - 1.Data Collection
 - 2.Profile Representation
 - 3.Updating user profiles
 - 4.Content based Profiles
 - 5.Collaborative Profiles
 - 6.Hybrid Profiles
 - 7.Implicit User Profiles
 - 8.Cold Start Problem



User Profiles in Recommender Systems



1.Data Collection:

- User profiles are built by collecting data about user interactions with items.
- These interactions can include ratings, clicks, views, purchases, likes, and more.
- The data is then used to infer the user's preferences and behaviors.



User Profiles in Recommender Systems



2. Profile Representation:

- User profiles are often represented as vector or feature sets.
- Each feature represents a specific aspect of user behavior, such as preferred genres, historical ratings, recently viewed items, and more.



User Profiles in Recommender Systems



3.Updating User Profile:

- User profiles evolve over time as users interact with the system. New interactions are used to update and refine the user's profile.
- For example, if a user starts watching documentaries, their profile will adjust to reflect this new interest.



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4. Content Based Profiles:

- In content-based recommender systems, user profiles are built based on the content attributes of items the user has interacted with.
- The system compares these attributes to the user's profile to recommend items with similar attributes.



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5. Collaborative Profiles:

- Collaborative filtering systems build user profiles by analyzing the behaviors of similar users.
- If User A and User B have similar interaction patterns, User A's profile might incorporate preferences of User B to improve recommendations.



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6.Hybrid Profiles:

- Hybrid recommender systems combine multiple sources of information to build comprehensive user profiles.
- This might involve merging content-based and collaborative profiles to capture both item attributes and user similarities.



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7.Implicit User Profiles:

- Sometimes, user profiles are built using implicit interactions, such as clicks or view durations.
- These interactions might not explicitly reflect preferences but can still offer insights into user interests.



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8. Cold Start Problem :

- A challenge with user profiles arises when dealing with new users who have limited interaction history.
- The cold start problem involves recommending items to these users based on their limited profile information.

