

# User-Based Collaborative Filtering

Understanding User Similarity for Personalized Recommendations

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# What is Collaborative Filtering?



Recommendation  
technique

Filters items based on preferences  
of similar users



User-Based approach

Groups similar users together



Key assumption

Similar people like similar things

# Basic Concept

## Find similar users

Group people with matching preferences

## Identify new items

Find what similar users liked but active user hasn't seen

## Generate recommendations

Suggest these new items to active user



# User Similarity Methods



## Cosine Similarity

Measures angle  
between user vectors



# Pearson Correlation

Considers rating scale differences



# Jaccard Similarity

Compares sets of items



# User-Item Matrix

	Movie 1	Movie 2	Movie 3	Movie 4
User A	5	3	-	4
User B	4	-	5	3
User C	5	2	-	-

# Step-by-Step Process

## Create user-item matrix

Map all user interactions with items

## Compute similarity

Calculate how similar each user is to others

## Find neighbors

Identify top-N most similar users

## Aggregate preferences

Collect items liked by similar users

## Rank and recommend

Score and suggest new items to user



## Example Walkthrough

### User A

- Liked: Item 1, Item 3
- Similar to: User B
- Recommendation: Item 5

### User B

- Liked: Item 1, Item 3, Item 5
- Similar to: User A

### User C

- Liked: Item 2, Item 4
- Not similar to others

# Pros and Cons



## Pros

- Simple implementation
- No content analysis needed
- Discovers unexpected items

## Cons

- Scalability issues
- Cold start problem
- Sparsity challenges



# Cold Start Challenge

## Problem

New users have no history

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## Solutions

Onboarding questions, demographics, hybrid approaches

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## Fallbacks

Popular items, trending content

# Evaluation Metrics



## Accuracy

Precision, Recall, F1-score

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## Error Rates

RMSE, MAE for rating predictions





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## User Experience

Coverage, Diversity, Novelty

# Implementation Challenges

-  **Data Sparsity**  
Most user-item matrices are >95% empty
-  **Computational Cost**  
 $O(n^2)$  complexity as users increase
-  **Real-time Updates**  
Recalculating similarities is expensive
-  **Privacy Concerns**  
User preferences reveal sensitive information

## Recommendation Algorithm

### Algorithms

Be combination of user  
to distance and item similarity

Additional Data

Performance Metrics

Performance API

### Datasets

### Performance metrics



## Unlock predictive power

Unlock predictive power with our recommendation engine. Our engine is designed to help you understand your users and their preferences, and to provide personalized recommendations that will keep them coming back for more.

Explore the API

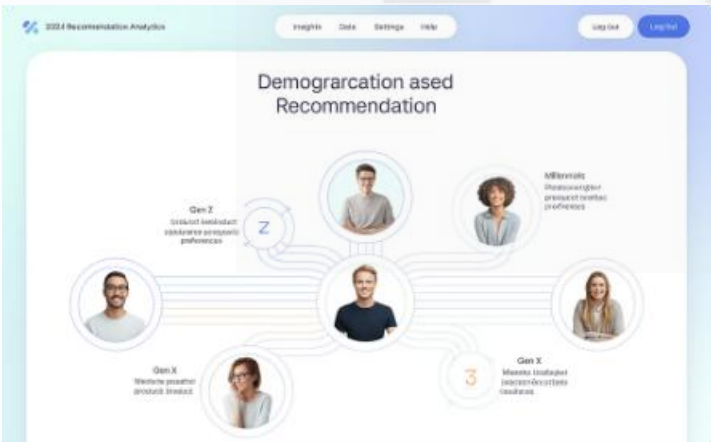
# Hybrid Approaches



**Content + Collaborative**  
Combines item features with user similarity



**Knowledge + Collaborative**  
Uses domain expertise with collaborative data



**Demographic + Collaborative**  
Groups similar user profiles together