

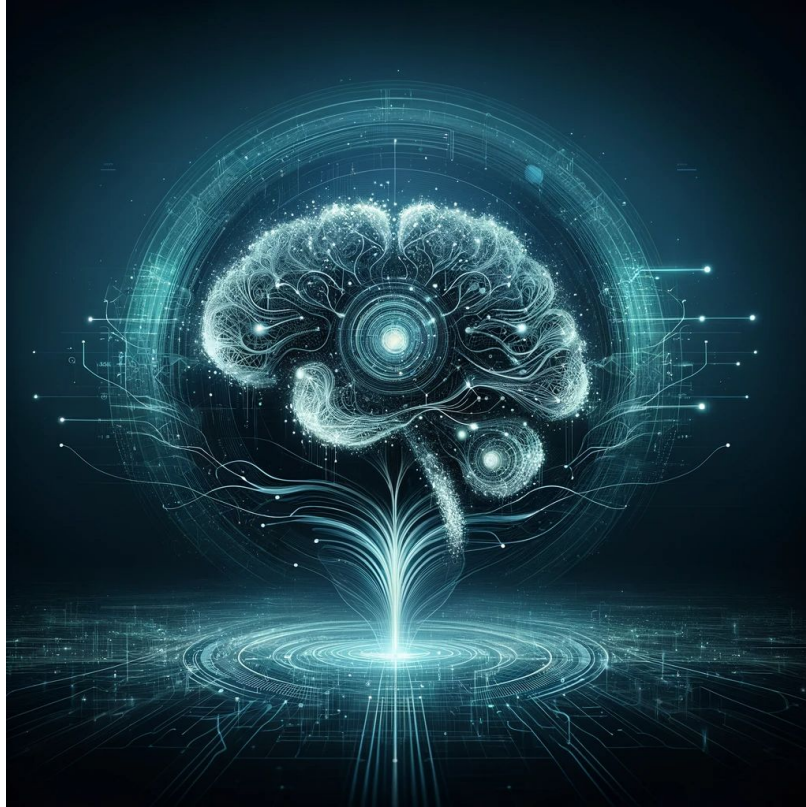


RECOMMENDER OF RELEVANT STORIES

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1- INTRODUCTION



2-LITERATURE REVIEW (SOTA)

- Historical context
- Advancements in NLP for Recommendation System
- Personalization Techniques
- Semantic Analysis and Relevance Matching
- Challenges and Considerations
- Future Directions



2-LITERATURE REVIEW (Relevant datasets)

NEWS
MIND



movielens

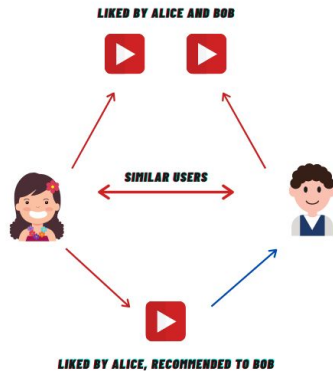
Non-commercial, personalized movie recommendations.

goodreads

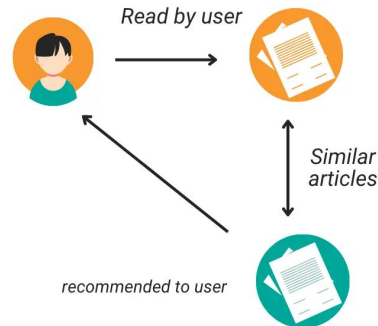
3-SOLUTION CONCEPT

How could we make a recommender of relevant stories?

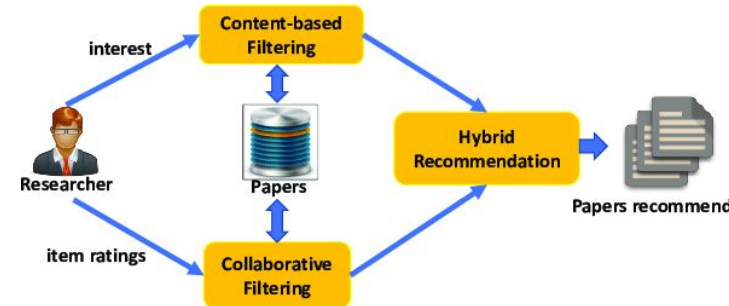
COLLABORATIVE FILTERING



CONTENT-BASED FILTERING

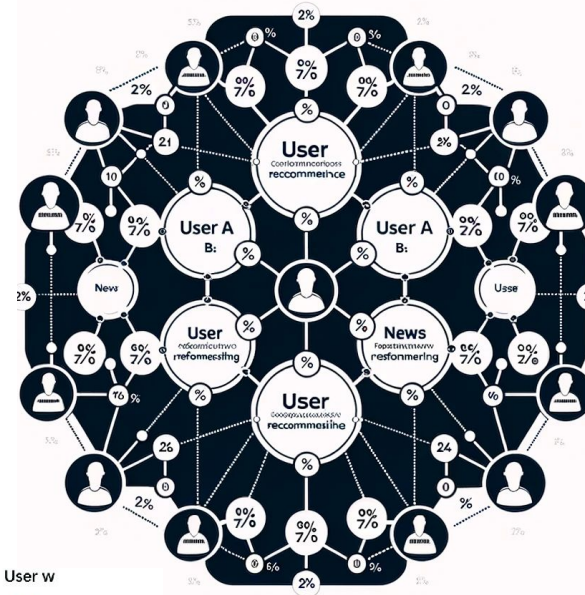
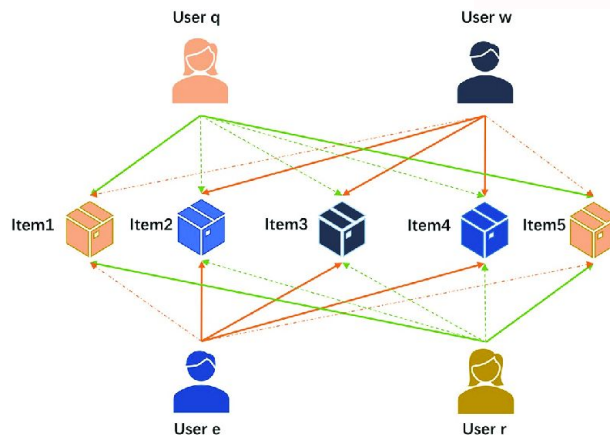
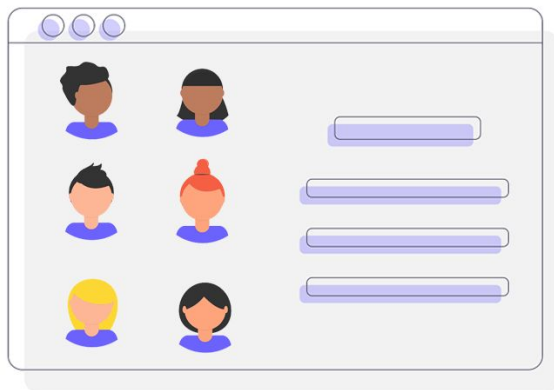


HYBRID FILTERING



3.1- CONTENT BASED FILTERING

- Content of items and aligning it with user preference
- Create profiles for users and items



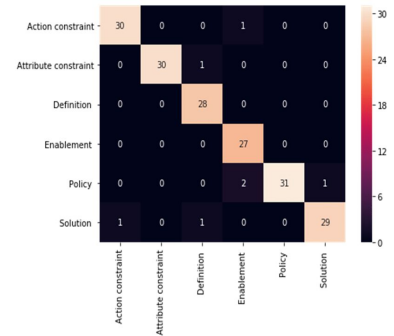
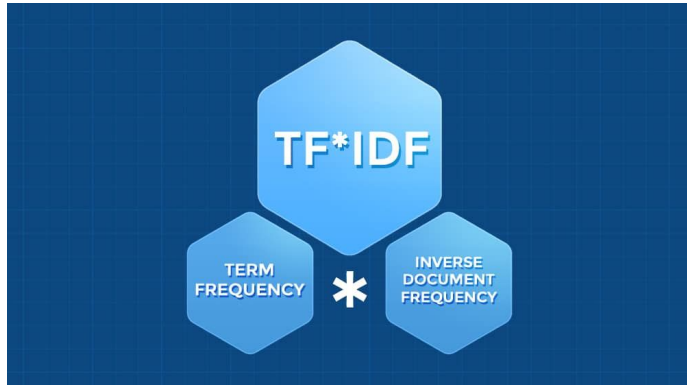
3.1- CONTENT BASED FILTERING

Used concepts:

1. Term Frequency (TF):
2. Inverse Document Frequency
3. TF-IDF Weight

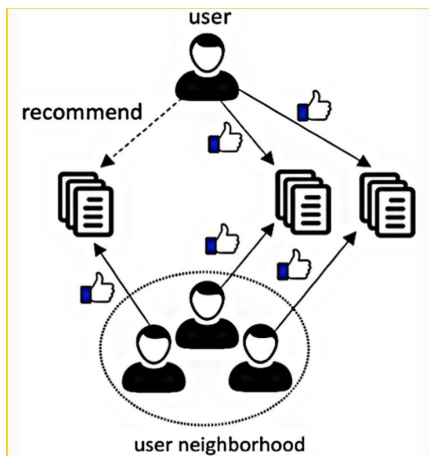
$$TF_{i,j} = \frac{n_{i,j}}{\sum_k n_{i,j}}$$

$$idf_j = \log \left[\frac{n}{df_j} \right]$$

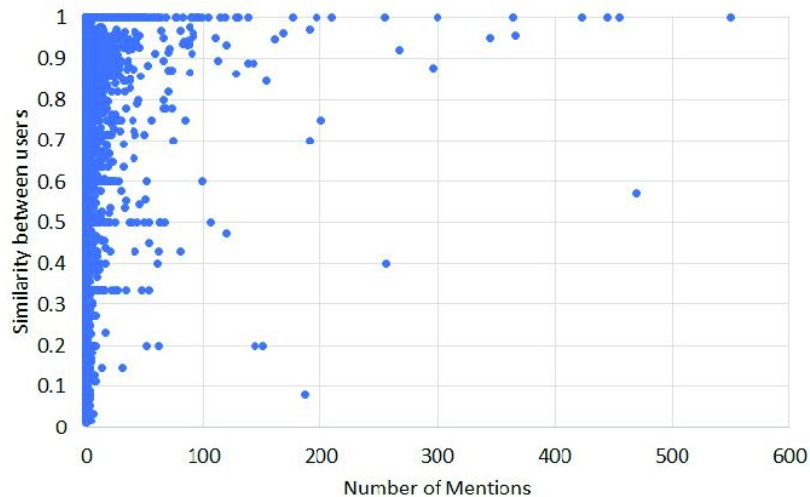
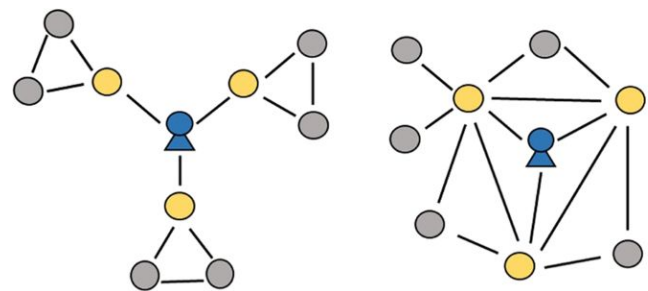


3.2-COLLABORATIVE FILTERING

- Similarity between users

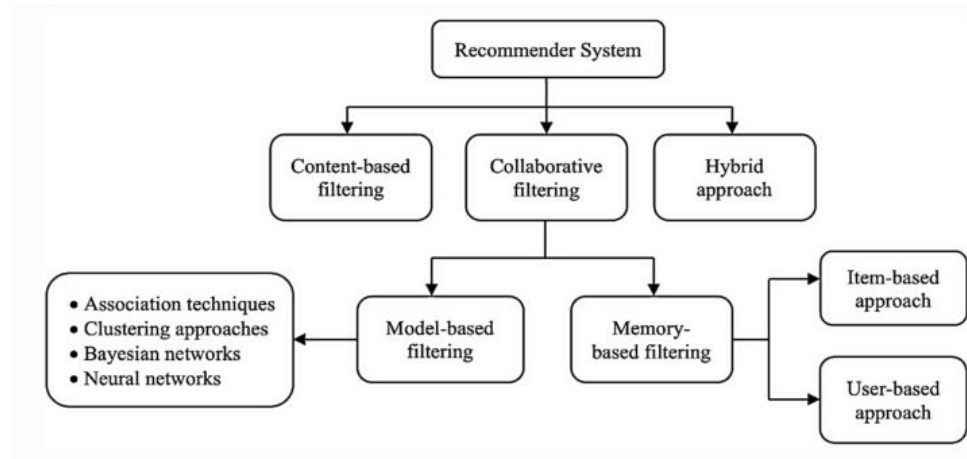


- Neighborhood X and user A



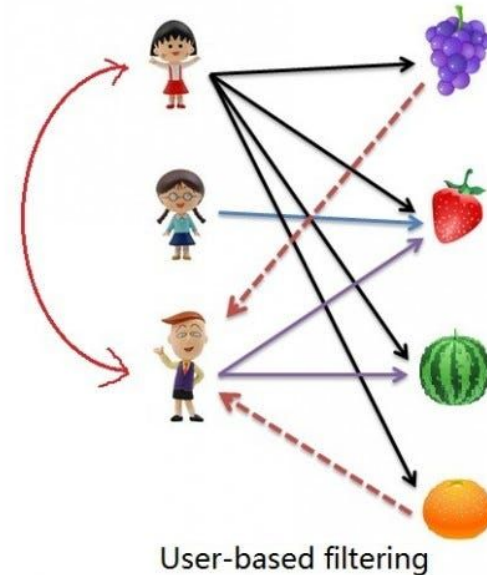
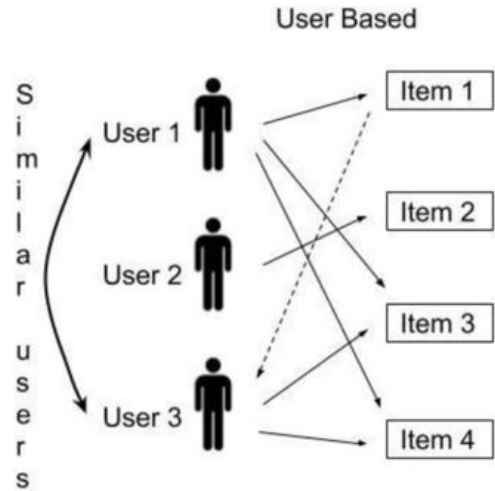
3.2- COLLABORATIVE FILTERING

- The recommendations are made based on a function that takes the model and user profile as input
- There are two types of collaborative filtering types:



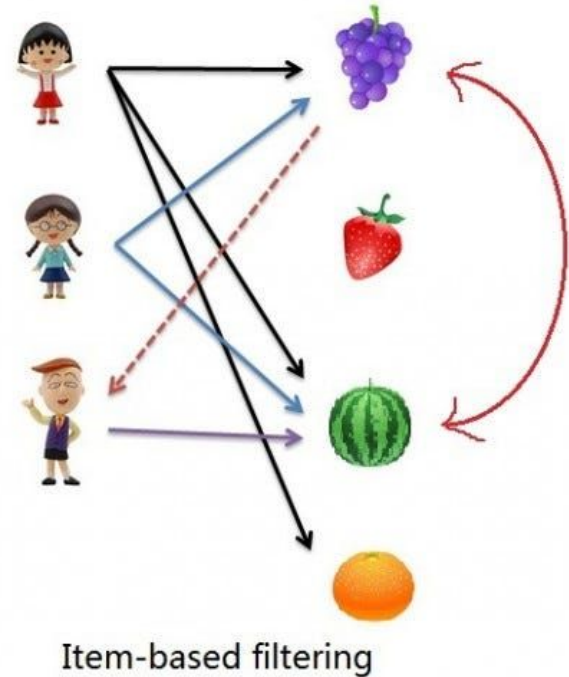
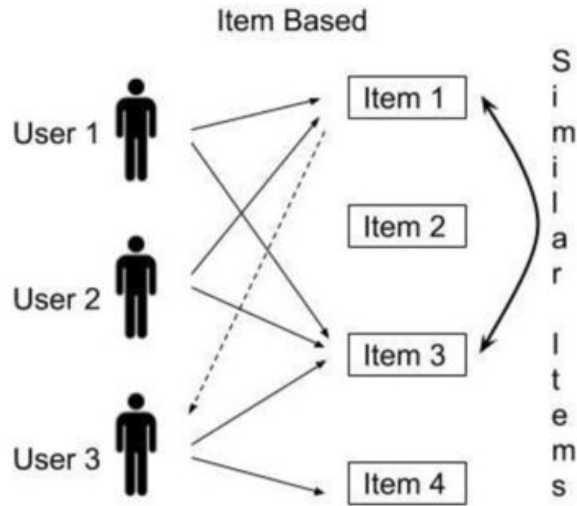
3.2.1- User-based collaborative filtering

- Neighbourhood users is needed



3.2.2- Item-based collaborative filtering

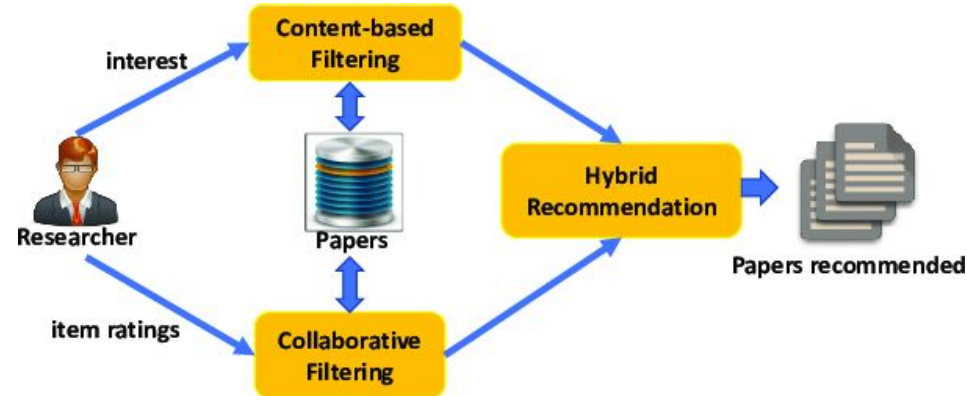
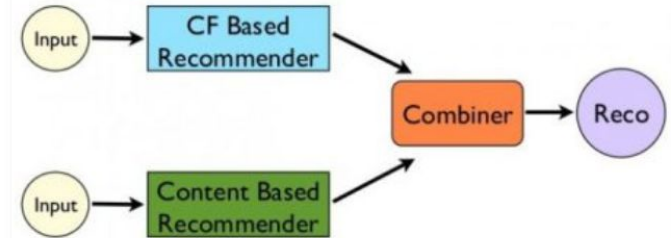
- A new item is predicted



3.3 HYBRID FILTERING

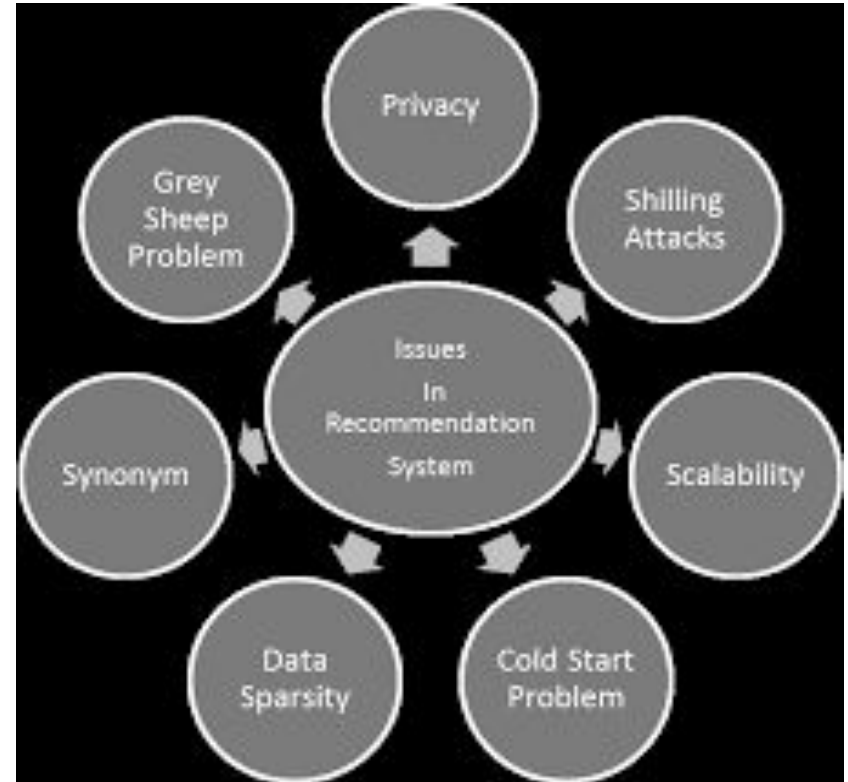
- Employed for addressing the limitations of individual recommender techniques
- Incorporate the results achieved from separate technique

Hybrid Recommendations

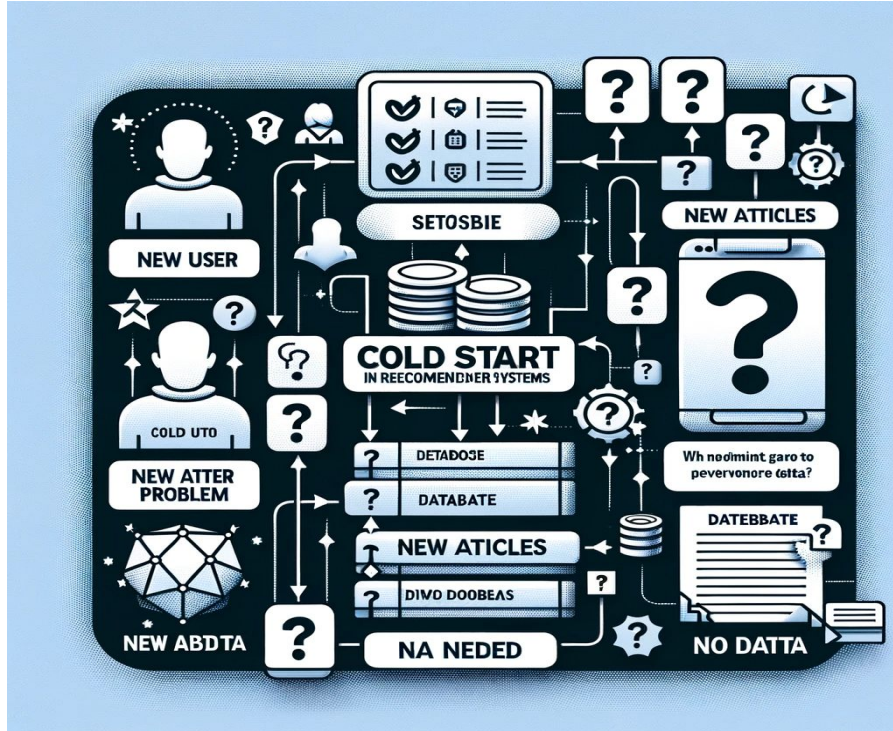


3.3-CHALLENGES

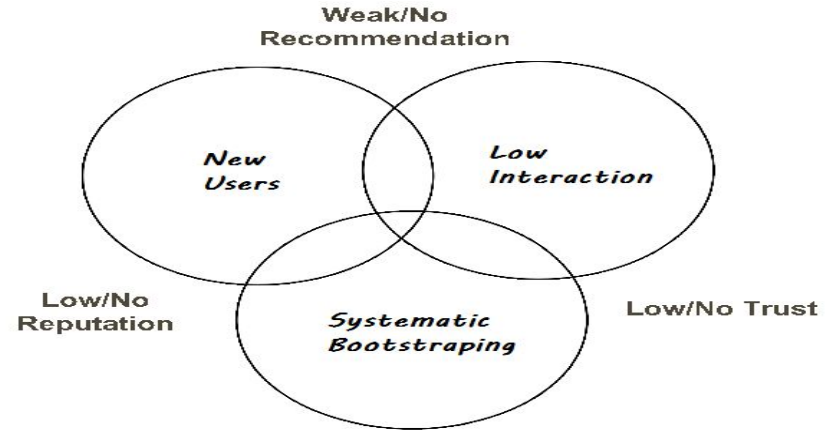
- **COLD START PROBLEM**
- **SPARSITY PROBLEM**
- **SHILLING ATTACK PROBLEM**
- **GREY SHEEP PROBLEM**



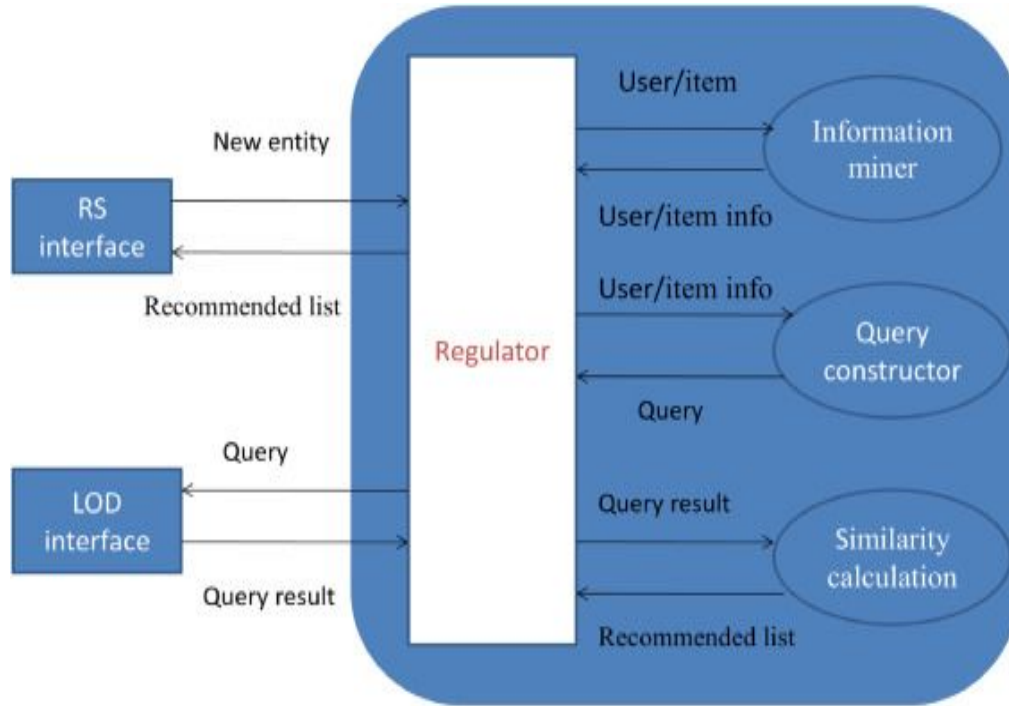
COLD START PROBLEM




























SOLUTION : COLLABORATIVE FILTERING



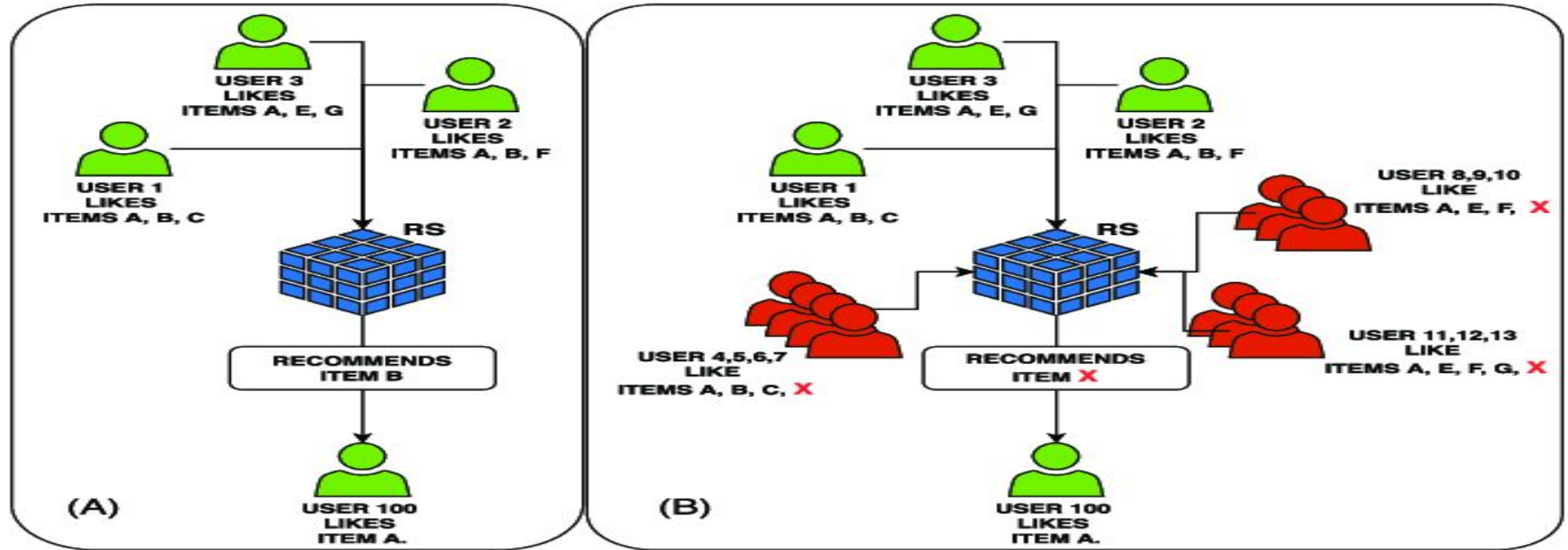
SPARSITY PROBLEM



SOLUTION : NEIGHBORHOOD-BASED METHODS

					
A					
B					
C					
D					
E					

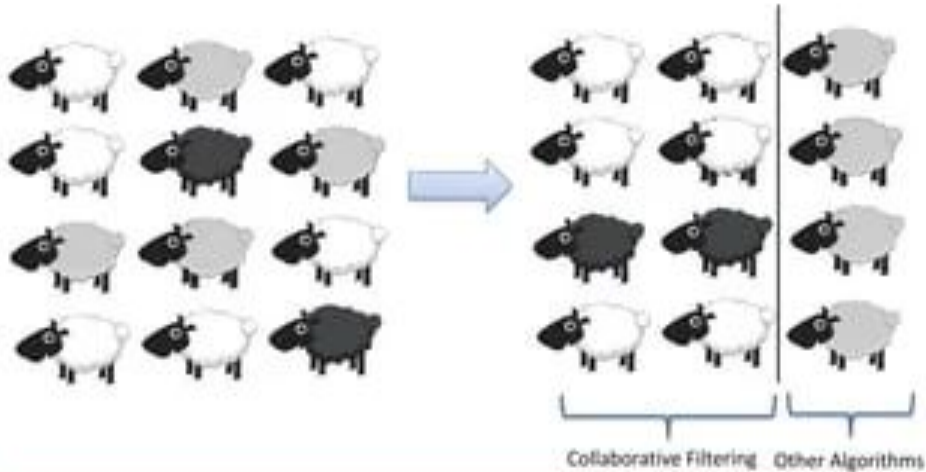
SHILLING ATTACK PROBLEM



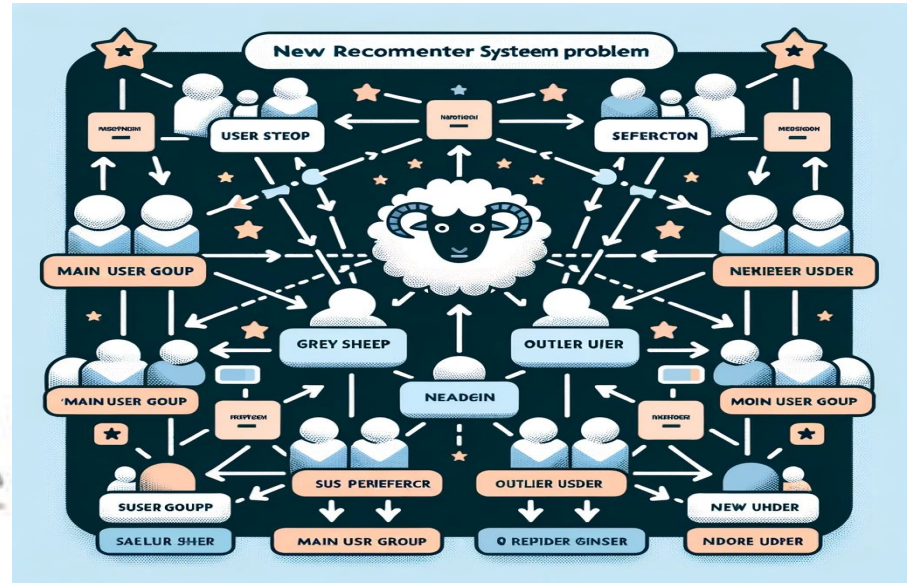
SOLUTION : DATA ANALYSIS AND ANOMALY DETECTION

GREY SHEEP PROBLEM

Research Problem: Identifying Grey Sheep Users



SOLUTION : ADAPTIVE ALGORITHMS



4-PROJECT PROPOSAL

- **OBJECTIVES**
- **METHODOLOGY**



4.1-OBJECTIVES

- Develop a robust recommendation engine that analyses user preferences and behaviour to suggest relevant stories.
- Build a scalable and efficient system capable of handling a large volume of users and stories.
- Evaluate the performance of the recommender system using appropriate metrics and user feedback.



4.2-METHODOLOGY

1. Data collection
2. Data preprocessing
3. Feature extraction
4. User Modelling
5. Recommendation Algorithms
6. Model Training and Validation
7. Evaluation



4.3-EVALUATION

- **PERFORMANCE METRICS**
- **A/B TESTING**
- **USER FEEDBACK**



THANK YOU

Questions