

Boomm Post Recommendation System – Report

Objective

Build a hybrid post recommendation system for Boomm, a social platform where finance and community meet, to suggest relevant posts to users based on their interests, content similarity, engagement metrics and collaborative behavior.

Data

- **Users (users.csv):** user_id, interested_in (topics of interest)
- **Posts (posts.csv):** post_id, content, is_anonymous, created_at, likes, shares, reports, like_user_ids, topics

Cleaning & Preprocessing:

- Filled missing interests and post engagement metrics
 - Lowercased and stripped text fields
 - Converted created_at to datetime
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Methodology

1. Feature Engineering

1. **Popularity Score** – weighted combination of likes, shares, and reports:
$$popularity_raw = 0.7 \text{ likes} + 0.2 * \text{shares} - 0.1 * \text{reports}$$

Normalized between 0–1.
2. **Recency Score** – normalized post age:
$$recency = \frac{created_at - earliest_post}{latest_post - earliest_post}$$
3. **Semantic Score** – embedding-based similarity between user interests and post content using BERT:
 - Generated embeddings for user interests and post content
 - Computed cosine similarity for each user-post pair
4. **Collaborative Score** – recommendation based on similar users' interactions:
 - Built interaction matrix from like_user_ids
 - Computed user-user similarity and weighted post scores

2. Hybrid Scoring & Recommendations

- Final score for each user-post pair:
$$final_score = 0.5 * semantic + 0.25 * popularity + 0.15 * recency + 0.1 * collaborative$$
 - For users with no interests, higher weight given to popularity and recency.
 - Selected **top 10 posts per user** based on final_score.
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Results

- Produced a **CSV** with columns: user_id, recommended_post_ids (comma-separated top 10 posts)
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Insights

- Posts with **high semantic similarity and popularity** dominate top recommendations.
 - Users with no declared interests still receive meaningful recommendations via popularity and collaborative signals.
 - Combining **content-based and collaborative filtering** provides a balanced, human-like recommendation experience.
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Deliverables

- boommm_recommendations.csv – top 10 recommended posts per user
- Code implementation in Jupyter Notebook