

# CrossFire: Cross Media Joint Friend and Item Recommendations

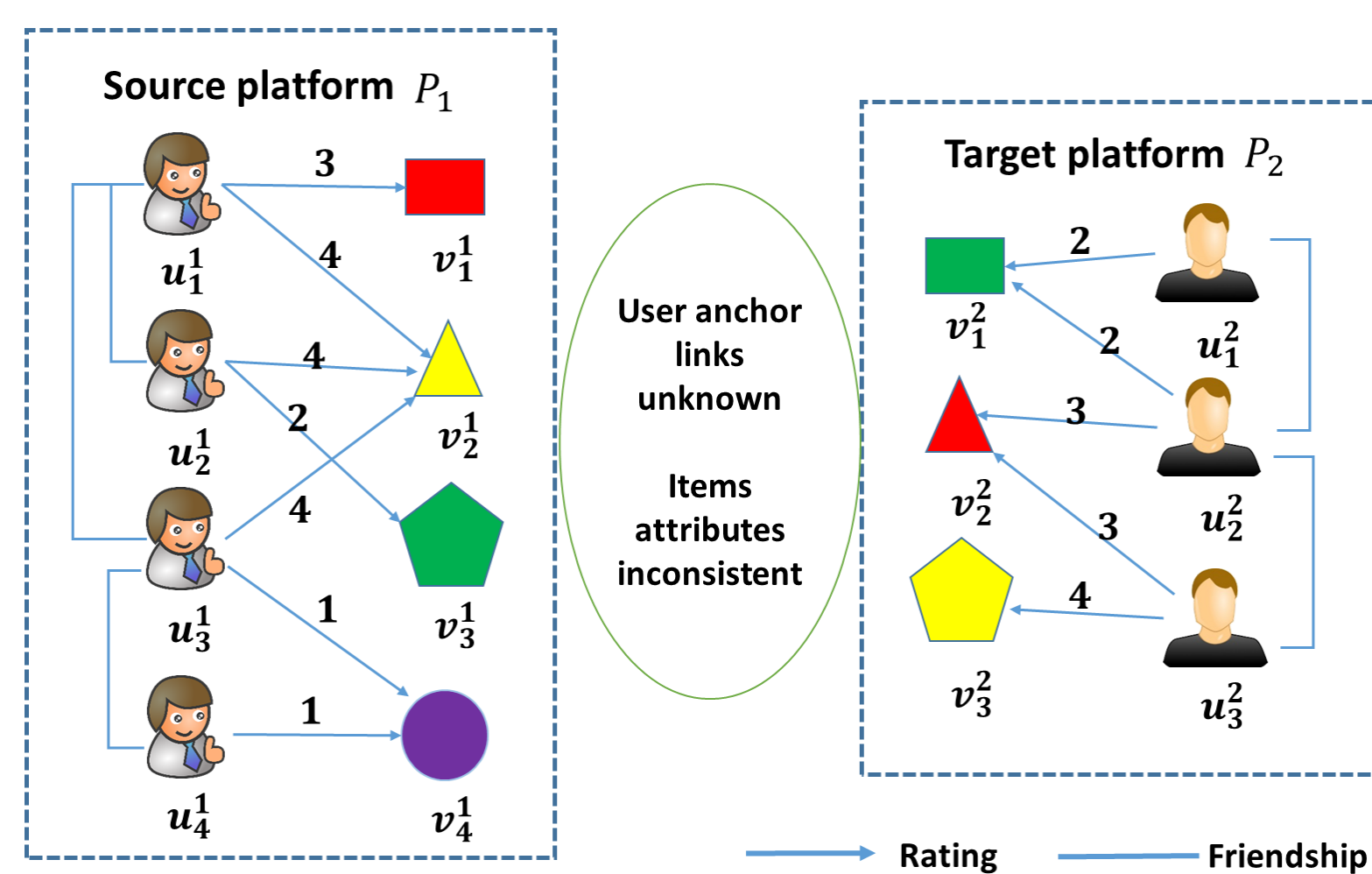
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## Introduction

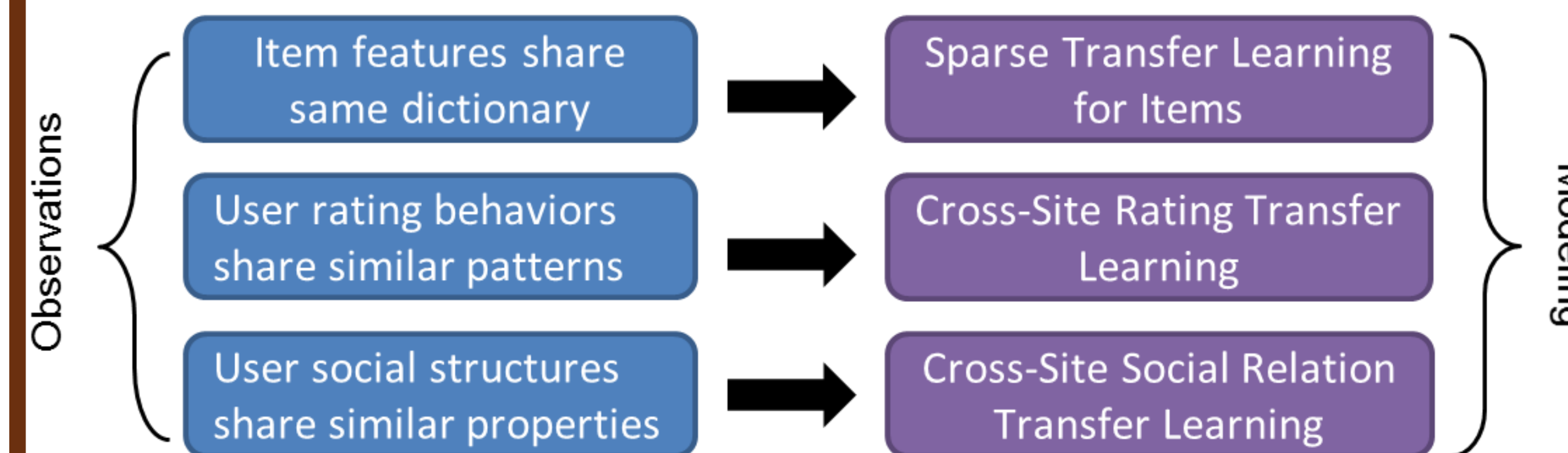
- Newly launched social media sites lack of user historical data to build effective recommender systems.
- Would auxiliary information be helpful in making friend and item recommendations?
- Challenges:**
  - Unknown user anchor links
  - Inconsistent item attributes



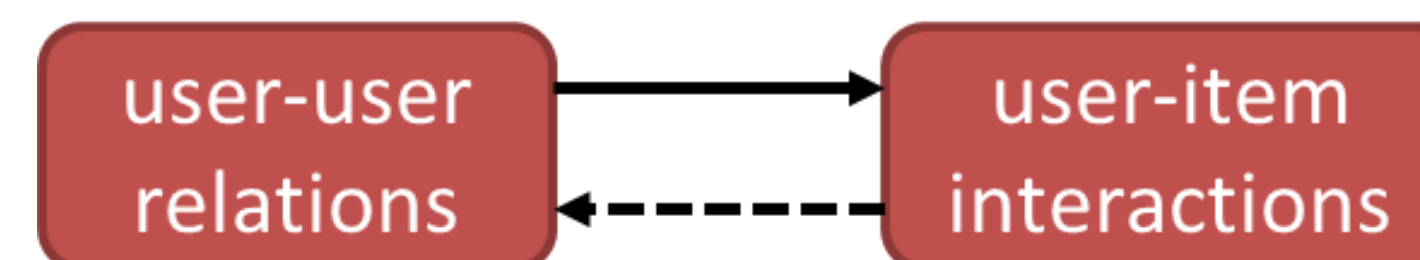
- Contributions:** (i) A novel problem of cross-media joint friend and item recommendations when no direct connections between users or items are given; (ii) A principled framework CrossFire, which integrates within-platform correlations and cross-platform information into a coherent model for joint friend and item recommendations.

## The Proposed Framework - CrossFire

- Cross media joint friend and item recommendations
  - Cross-platform knowledge transfer: build implicit bridge



- Within-platform joint learning: exploring mutual benefits



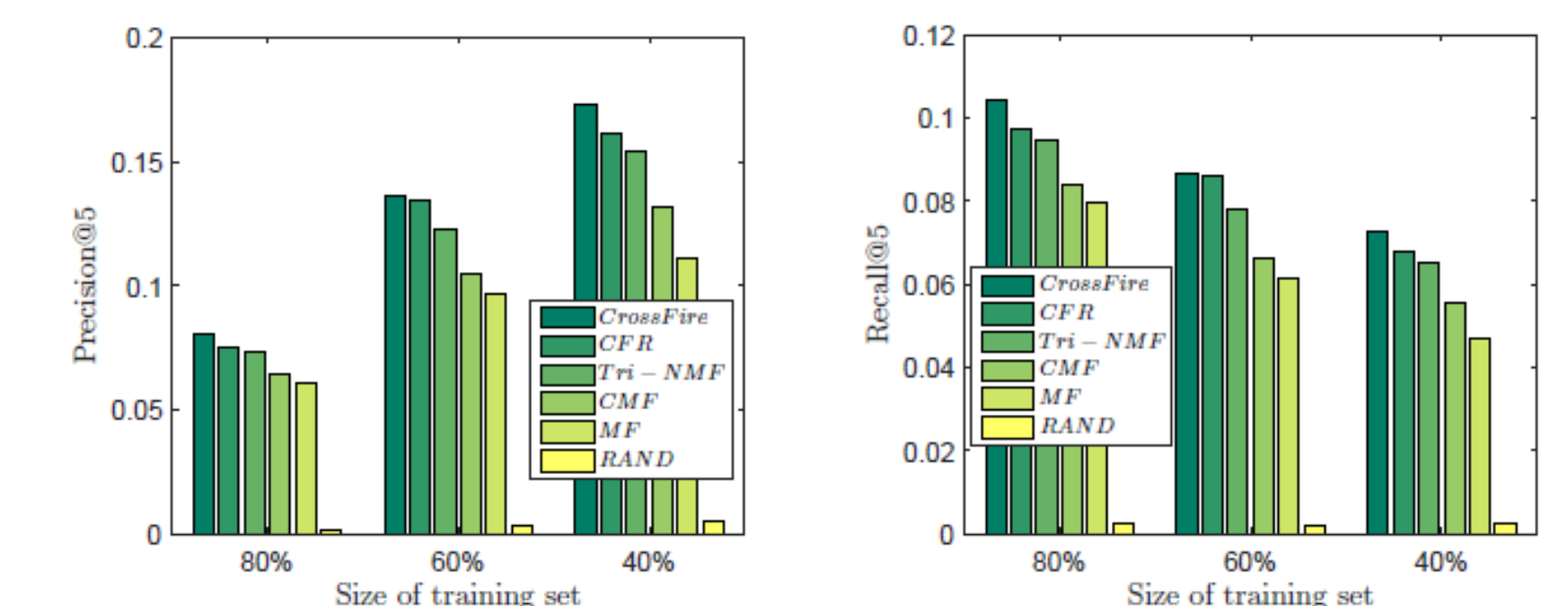
- Objective function of CrossFire

$$\begin{aligned} \min_{\theta} & \sum_{i=1}^2 \|X_i - DV_i\|_F^2 + \gamma \|V_i\|_1 + Tr(V(\mu L + \nu M)V^T) \\ & \underbrace{\hspace{10em}}_{\text{Item Sparse Transfer Learning}} \\ & + \alpha \sum_{i=1}^2 \|W_i \odot (R_i - U_i^T Q^T V_i)\|_F^2 + \lambda (\|P\|_F^2 + \|Q\|_F^2) \\ & \underbrace{\hspace{10em}}_{\text{Cross-Media Item Recommendation}} \\ & + \beta \sum_{i=1}^2 \|Y_i \odot (A_i - U_i^T P U_i)\|_F^2 + \lambda \sum_{i=1}^2 \|U_i\|_F^2 \\ & \underbrace{\hspace{10em}}_{\text{Cross-Media Friend Recommendation}} \\ \text{s.t.} & \|d_j\|_2^2 \leq 1, j = 1, \dots, p, Q^T Q = I \end{aligned}$$

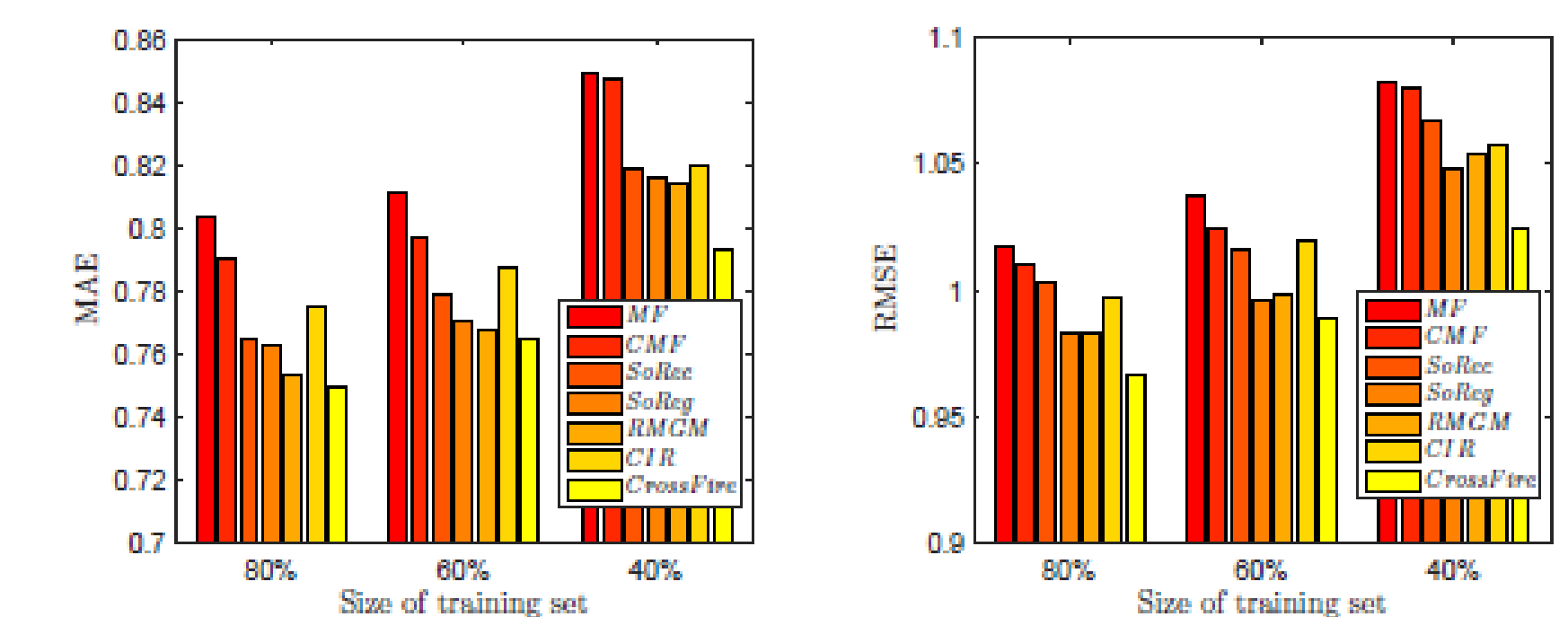
- Latent user and item feature matrix  $U_2$ ,  $V_2$ , projection matrix  $Q$ , and interaction matrix  $P$  are obtained to recommend friends and items.

## Experimental Results

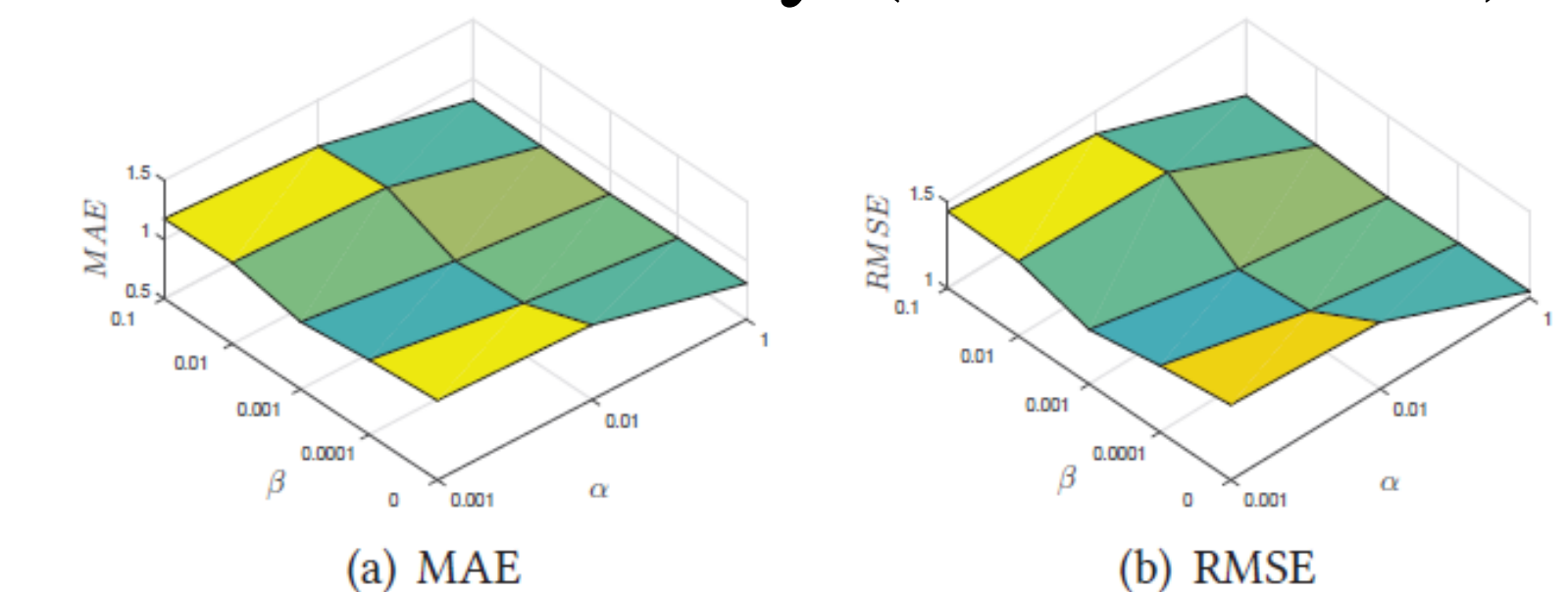
- Friend recommendation performance



- Item recommendation performance



- Parameter Sensitivity (on Ciao dataset)



## Future Work

- (i) Extend CrossFire to scenario with limited explicit user/item correspondences are given;
- (ii) Explore CrossFire in a streaming recommendation setting.

This material is based upon work supported by, or in part by, the ONR grant N00014-16-1-2257.