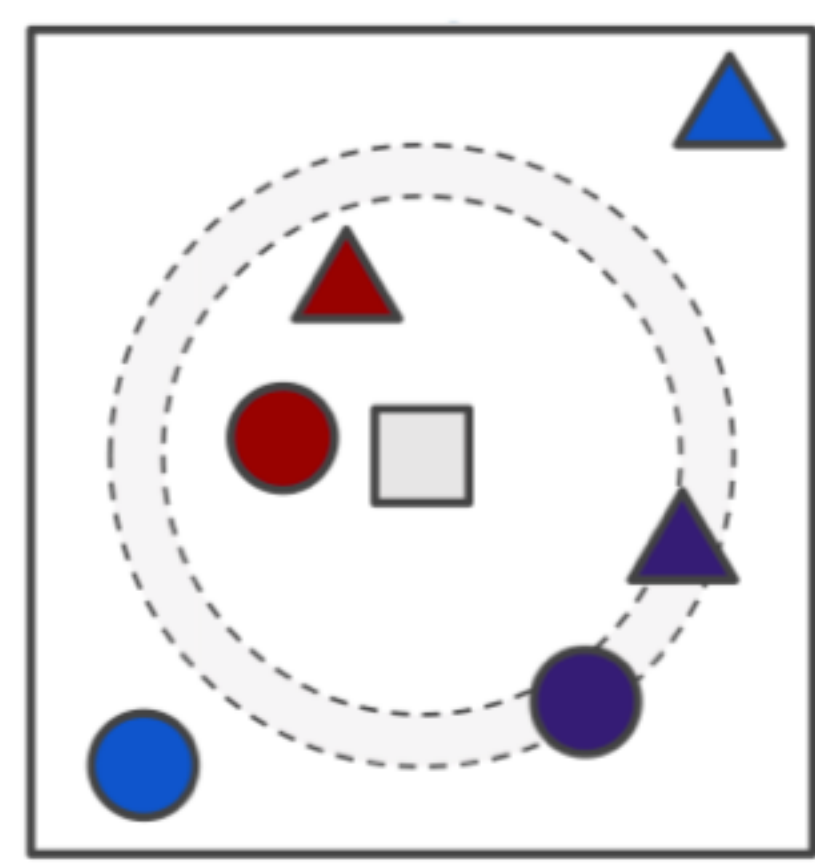


# Proposed Method

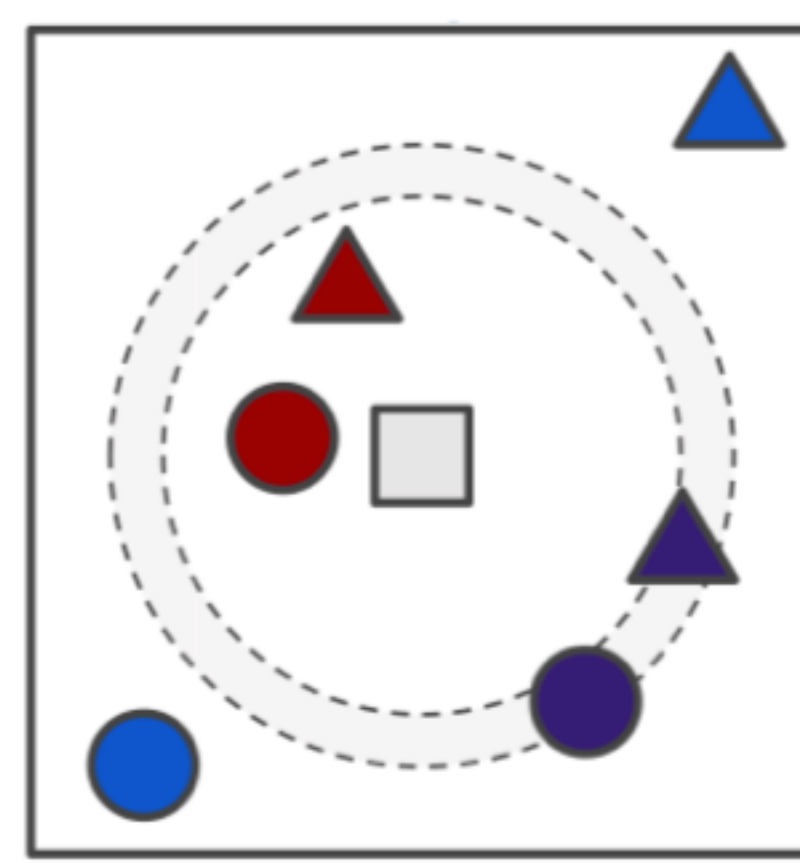
## Metric Learning with Interest Sustainability Score



- Incorporate the ISS in the above metric learning framework to consider how users' interest in each item will sustain in the future.
- The underlying idea is pull items with high ISS to users and to push items with low ISS from users.
- Design a ISS-based objective  $L_S$  with continuous labels ( $p_i$ ):
  - $L_S(u, i^+, i^-) = \{(d(\mathbf{u}, \mathbf{i}^+) - d(\mathbf{u}, \mathbf{i}^-)) - (p_{i^-} - p_{i^+})\}^2$
- The goal of  $L_S$  is to arrange item  $i^+$  and  $i^-$  by according to the difference of their ISSs ( $p_{i^-} - p_{i^+}$ ).
- For example, if  $p_{i^-} - p_{i^+} < 0$ , the objective makes the positive item will be closer to the user than the negative item by  $|p_{i^-} - p_{i^+}|$ .

# Proposed Method

## Metric Learning with Interest Sustainability Score



(b)  $\text{CRIS}^{\text{reg}}$

- The final loss is a linear combination of both objectives:

$$L = \sum_{(u, i^+) \in P} \sum_{(u, i^-) \notin P} L_C(u, i^+, i^-) + \lambda L_S(u, i^+, i^-)$$

- $P$ : set of user-item interactions,  $\lambda$ : balancing coefficient,  $L_S$ : regularization on metric learning framework.
- Given the combination of both objectives, the metric learning method can build a representation space with considering both whether users liked items (by  $L_C$ ) and how users' interest in the items sustain in the future (by  $L_S$ ), name this method as  $\text{CRIS}^{\text{reg}}$