



AI Box

Towards Universal Sequence Representation Learning for Recommender Systems

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Most Recommenders are not Transferable

* Cold-Start Items

ID-based Sequential Models



cannot recommend
New items

* New Domains / Platforms

Model on an Existing Domain

cannot help

Model on a New Domain

w/ different item IDs

How to develop transferable recommendation models? 🤔

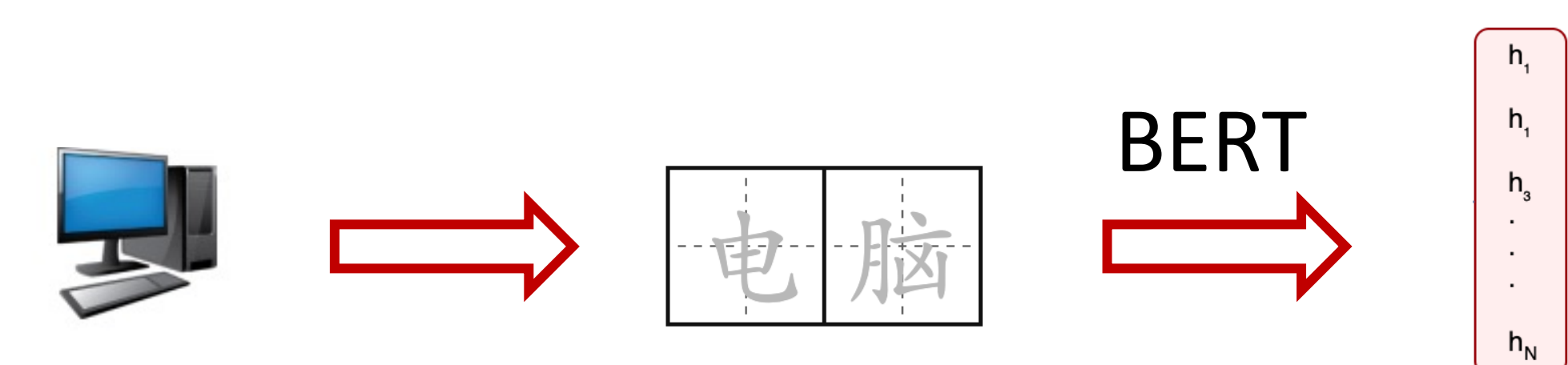
Inspired by Large Pre-trained Models

Pre-trained Sequential Recommender?

- ✓ Same data format;
- ✓ Large corpus;
- ✗ Different dictionaries (item IDs)
- ? ..., but what if we obtain universal item representations?



Describe Items via Natural Language



Item ID → Item Text (e.g. Title, Brand, Description) → Textual Representations (Universal)

Challenges

- * Textual representations are **not directly suitable** for recommendation tasks;

Pre-trained Language Model, e.g., BERT

Need fine-tuned Anisotropy

- * How to learn from multiple domains?

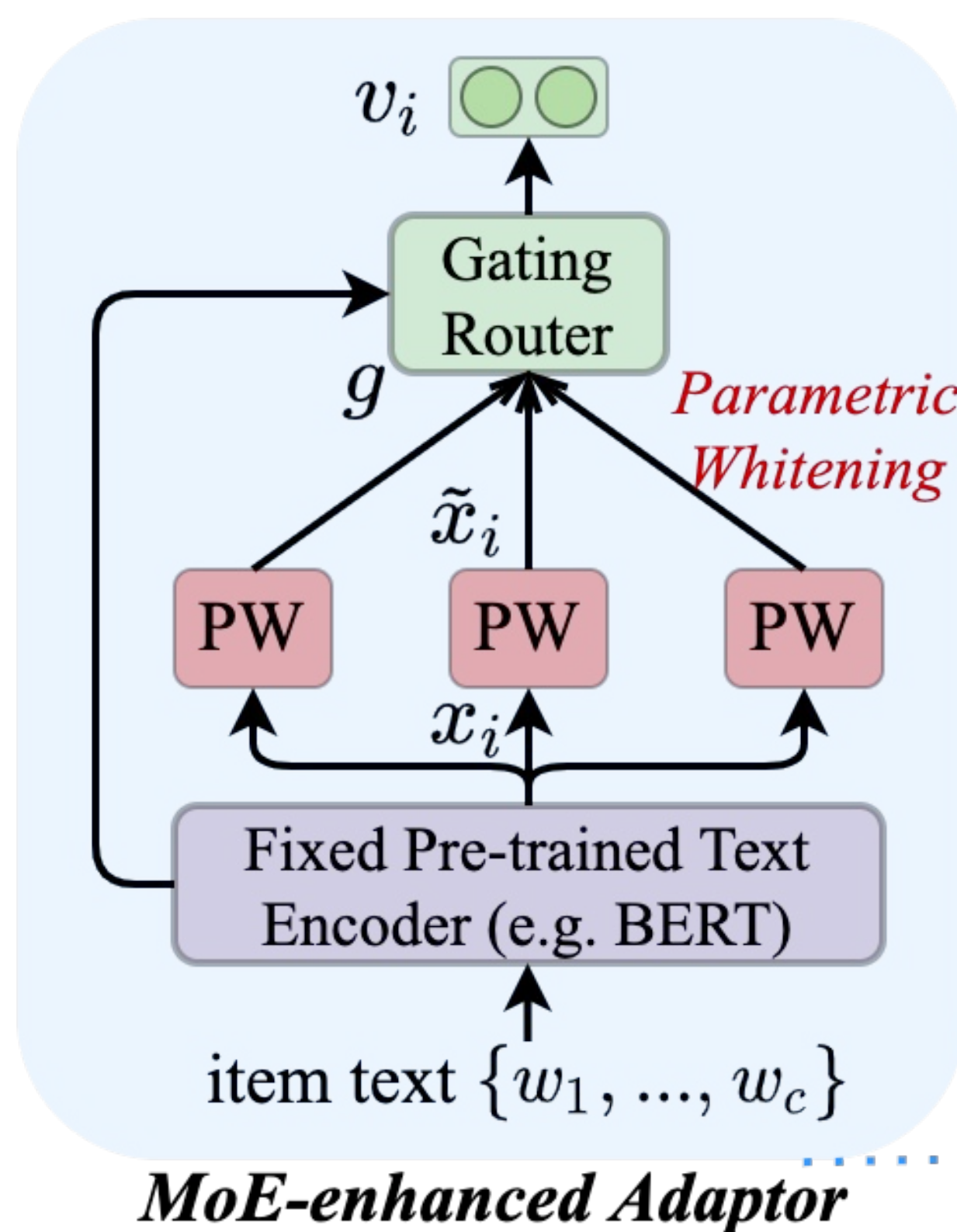


Overall Framework of UniSRec

1 Single Item Encoding

Universal Sequential Recommender

Universal Item Representation



1. PLM encoding
2. Fine-tune + reduce anisotropy
3. MoE for domain fusion & adaptation

2 Sequence Encoding & Multi-Domain Pre-training

T1: next-item prediction

T2: augmented CL

⚡ Negative instances are from multiple domains for fusion & adaptation.

Same as SASRec →

From 1 →

3 Fine-tuning on Target Domains / Platforms

Inductive:

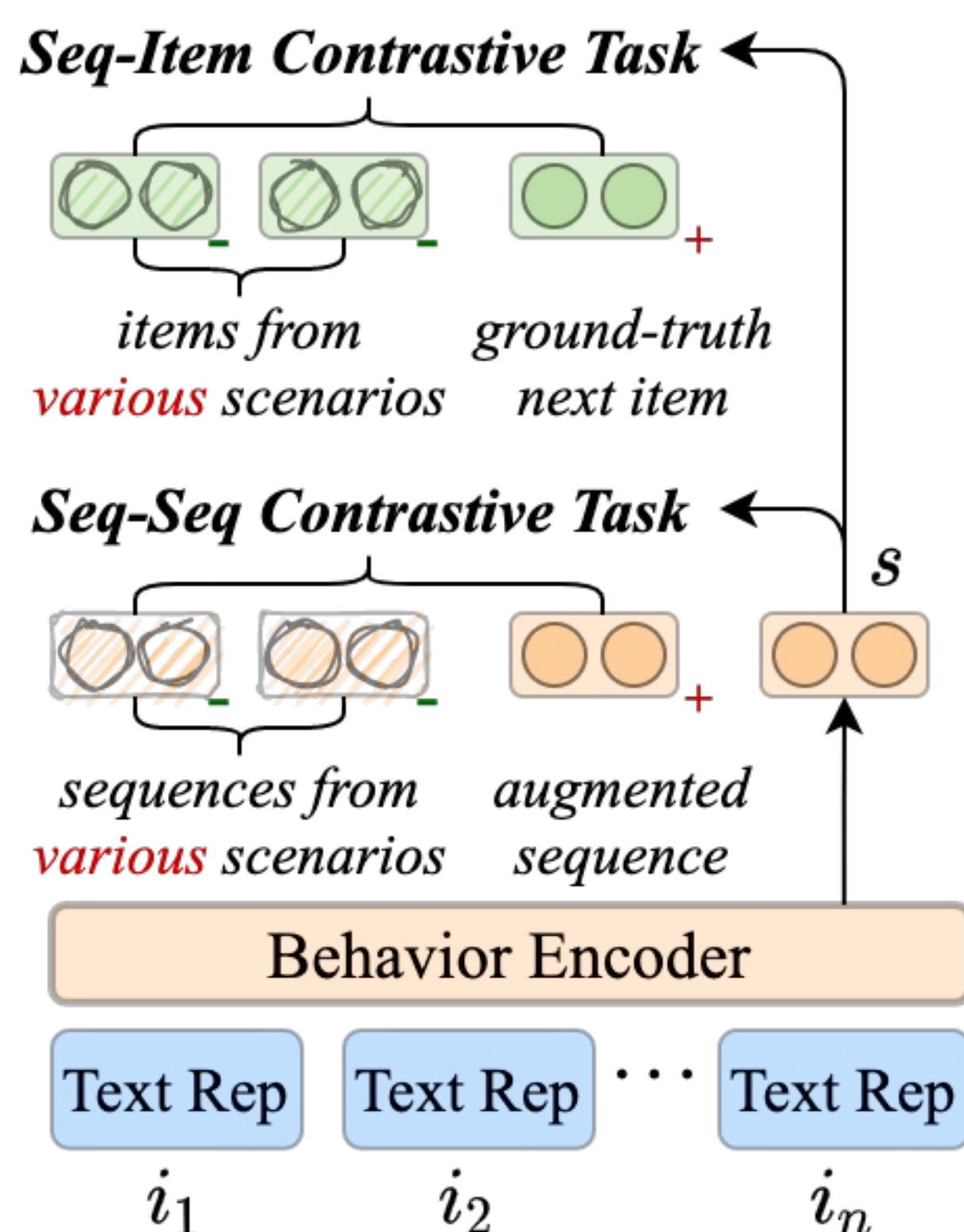
- Many new items;
- Do not use item IDs;

Transductive:

- Few new Items;
- Can use item IDs;

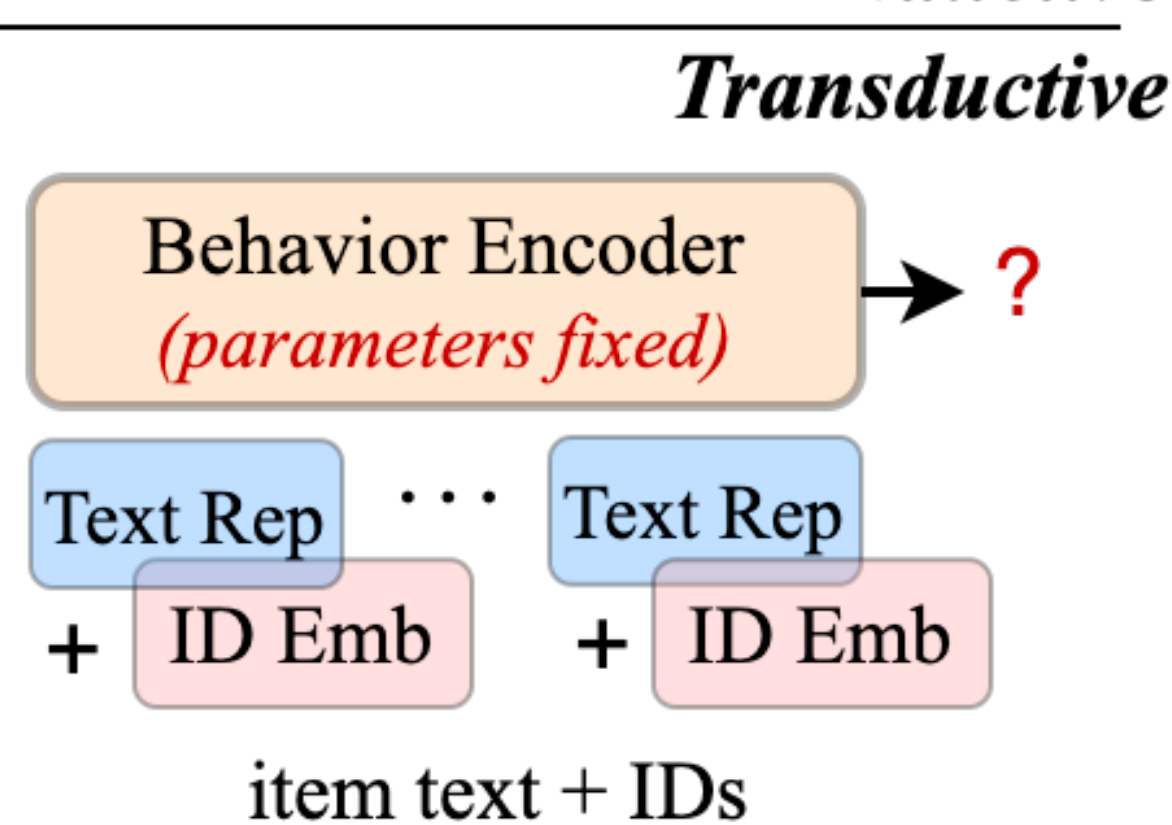
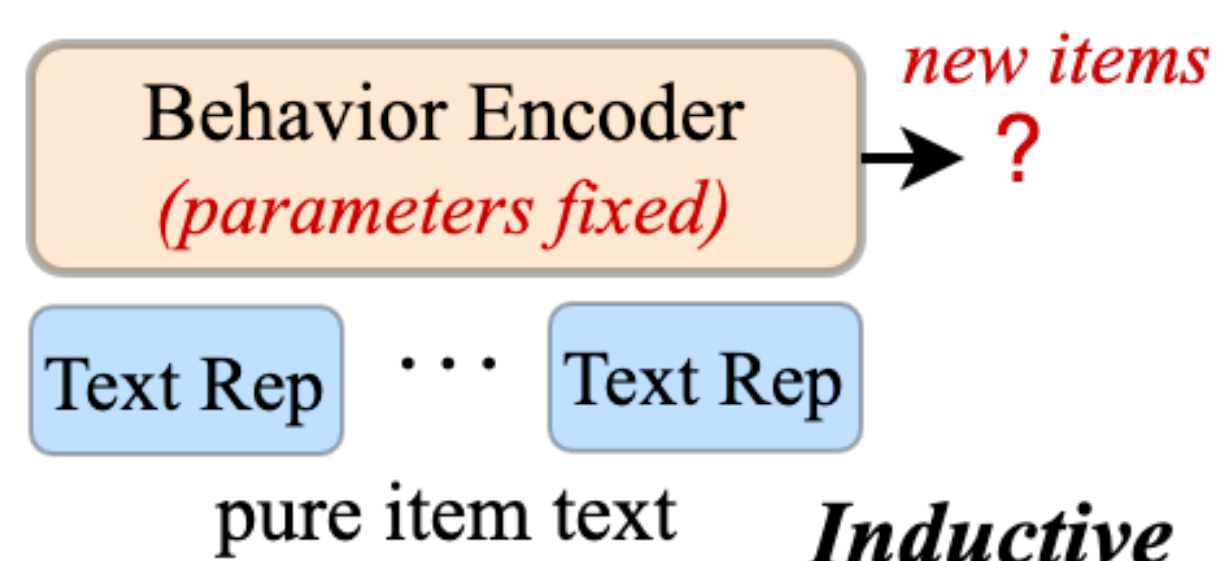
⚡ Only tune parameters in item encoding module.

Universal Sequence Representation Pre-training



Parameter-Efficient Fine-tuning

be able to recommend new items



Benchmark for Pre-trained Recommenders

One UniSRec model

Pre-trained Amazon: Food, CDs, Kindle, Movies, Home

Fine-tuned

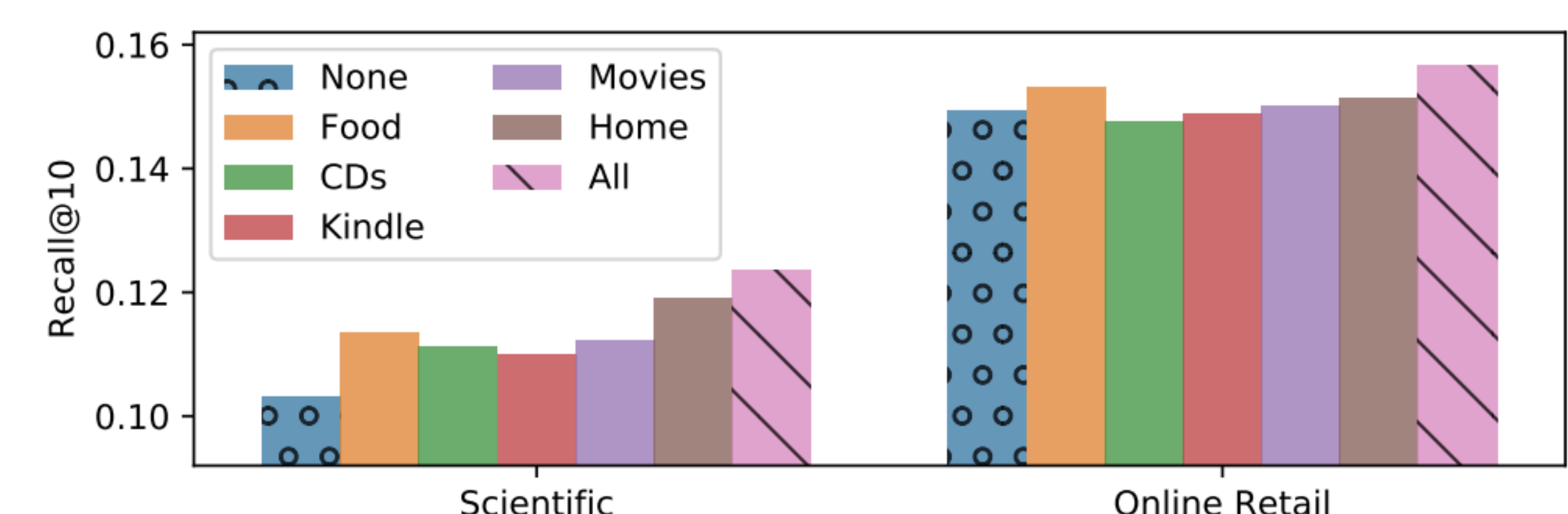
New domains Scientific, Pantry, Arts, Instruments, Office

New platform Online Retail in UK

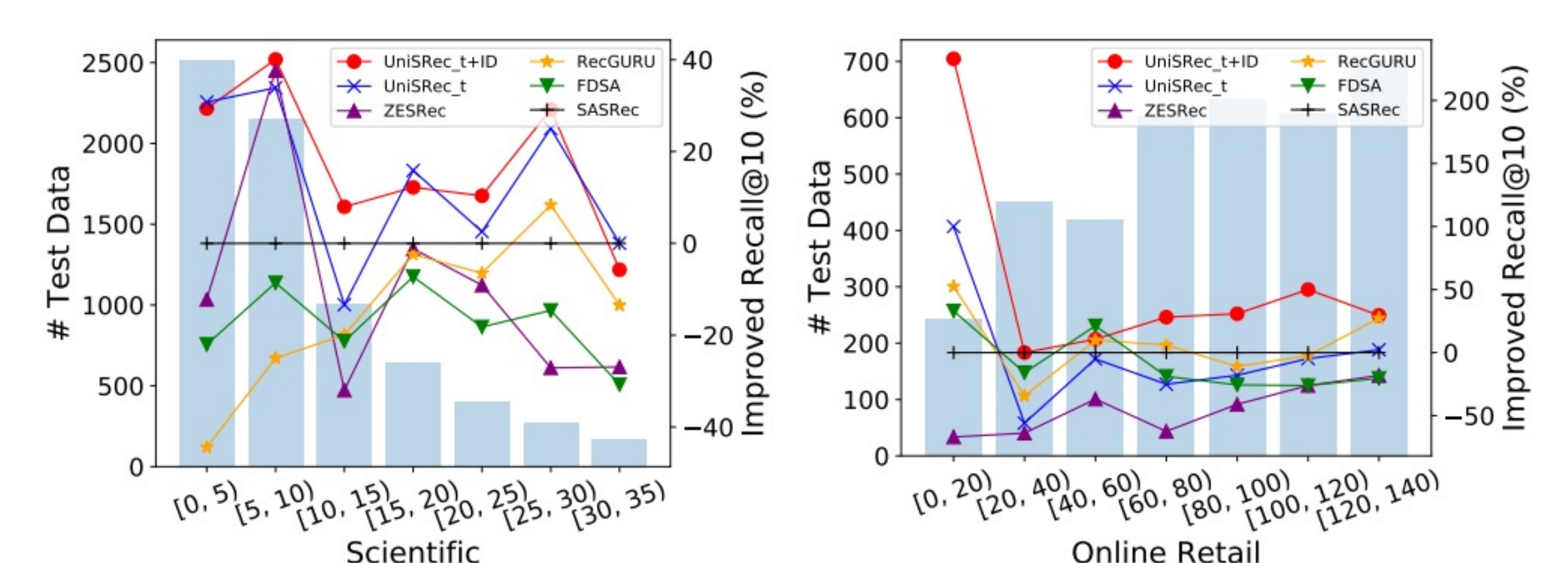
Experiments

Dataset	Metric	SASRec	FDSA	UniSRec _t	UniSRec _{t+ID}
Online Retail	Recall@10	0.1460	0.1490	0.1449	0.1537*
	NDCG@10	0.0675	0.0719	0.0677	0.0724
	Recall@50	0.3872	0.3802	0.3604	0.3885
	NDCG@50	0.1201	0.1223	0.1149	0.1239*

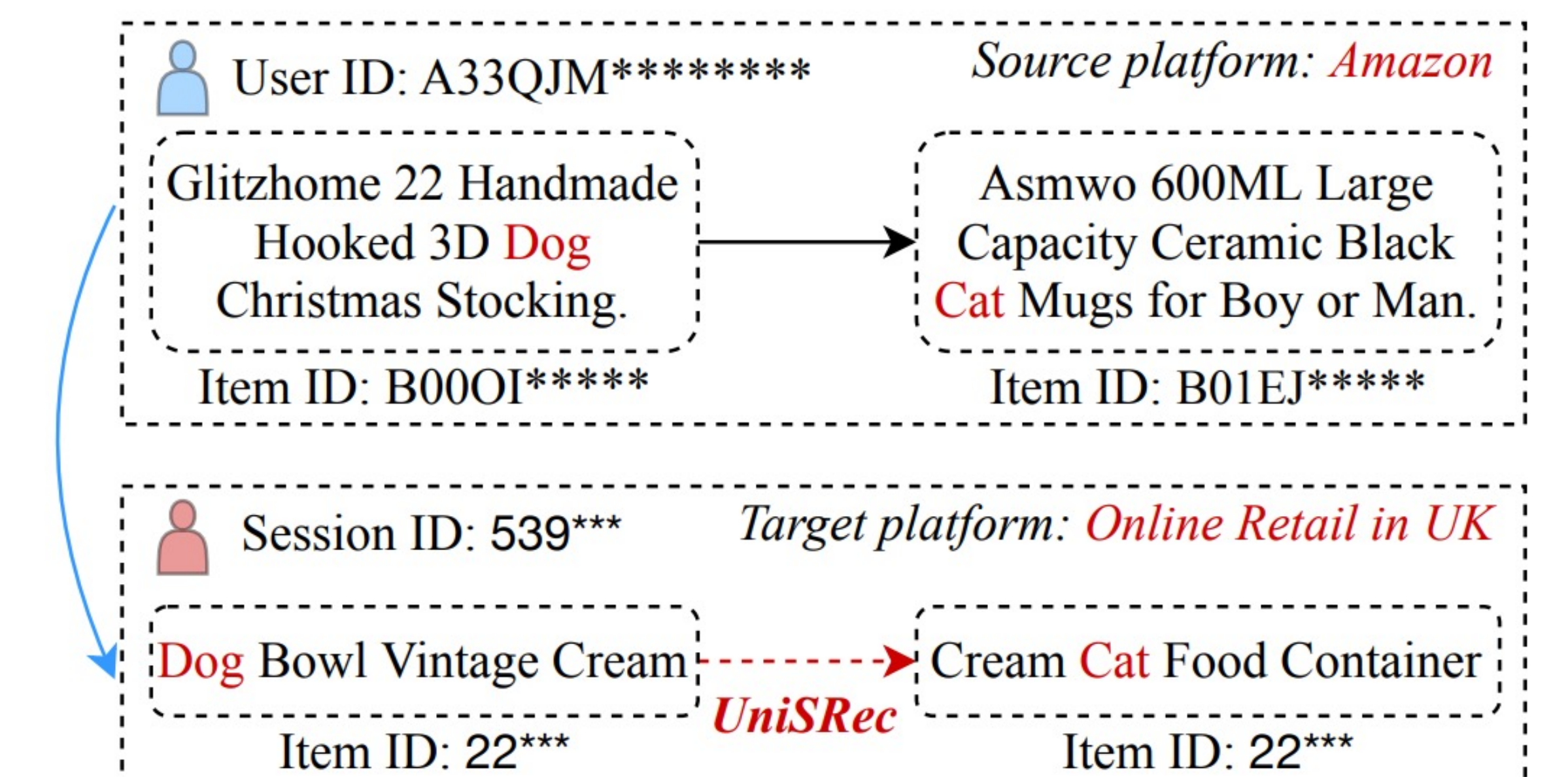
⚡ Performances improv. w/o any shared users or items between the pre-training (Amazon) & downstream (Online Retail in UK) platforms.



⚡ Pre-training on 5 domains > any 1 domain.



⚡ Significant improvements on cold-start items.



⚡ Capture and transfer semantic patterns.

Code implemented by RecBole are publicly available: github.com/RUCAIBox/UniSRec