



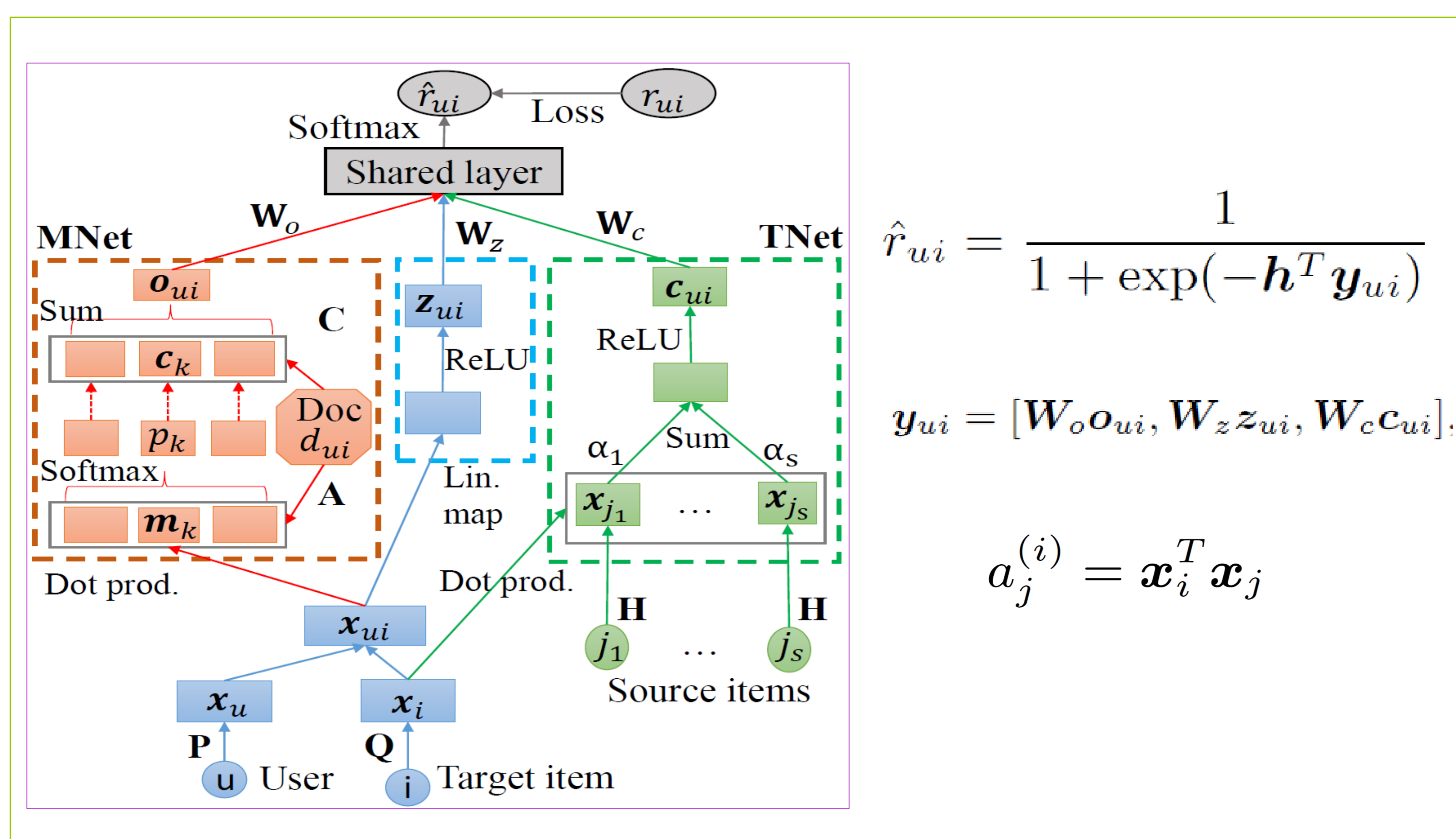
## 2 TMH: Transfer Meets Hybrid

TMH brings an effort on combining two threads to alleviate the data sparsity issues including cold-user and cold-item start.

### Contributions

- MNet (a memory network) can learn high-level representations of unstructured text w.r.t. the given user-item interaction.
- TNet (a transfer network) can selectively transfer knowledge from source items in a coarse-to-fine way.

### Architecture



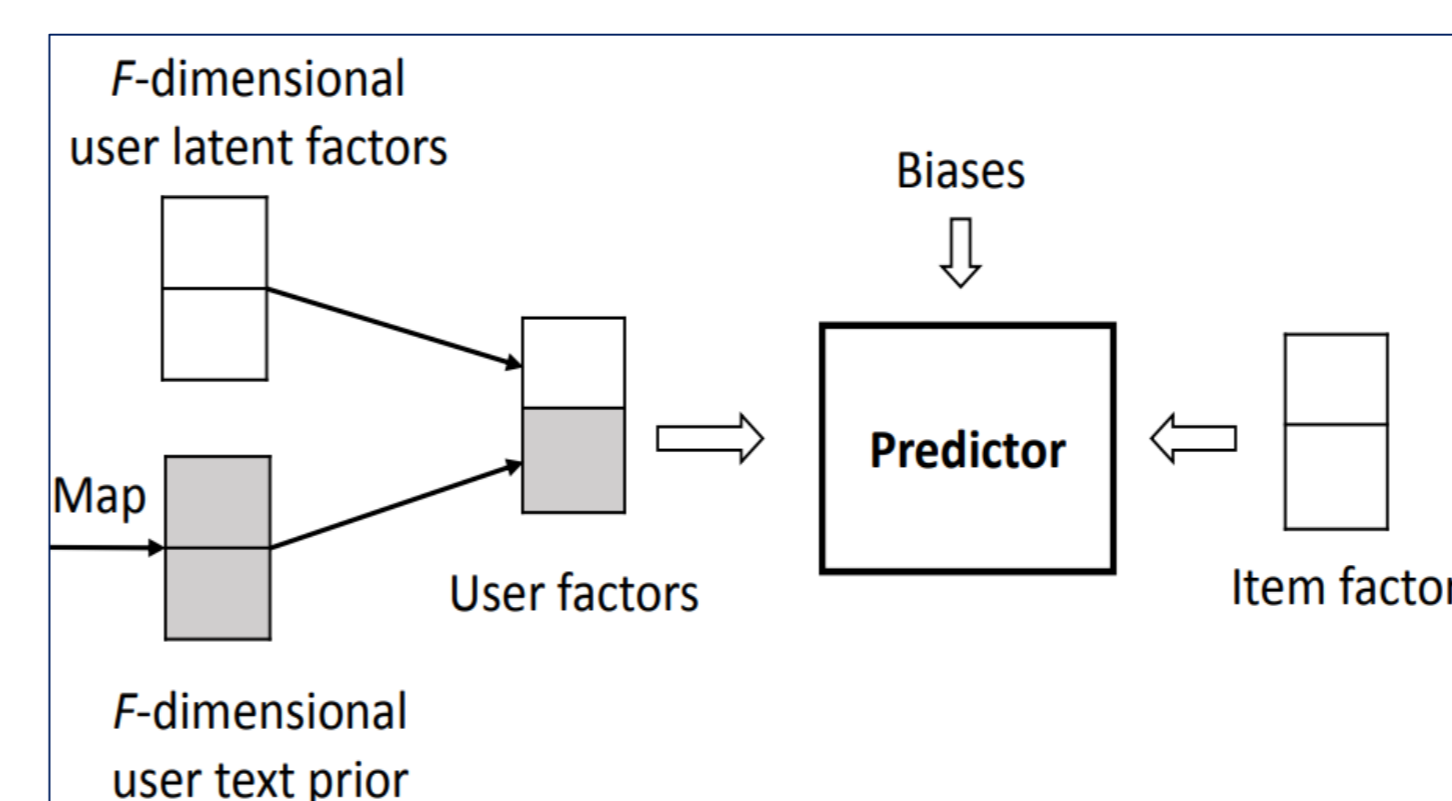
Quadrant II

Quadrant III

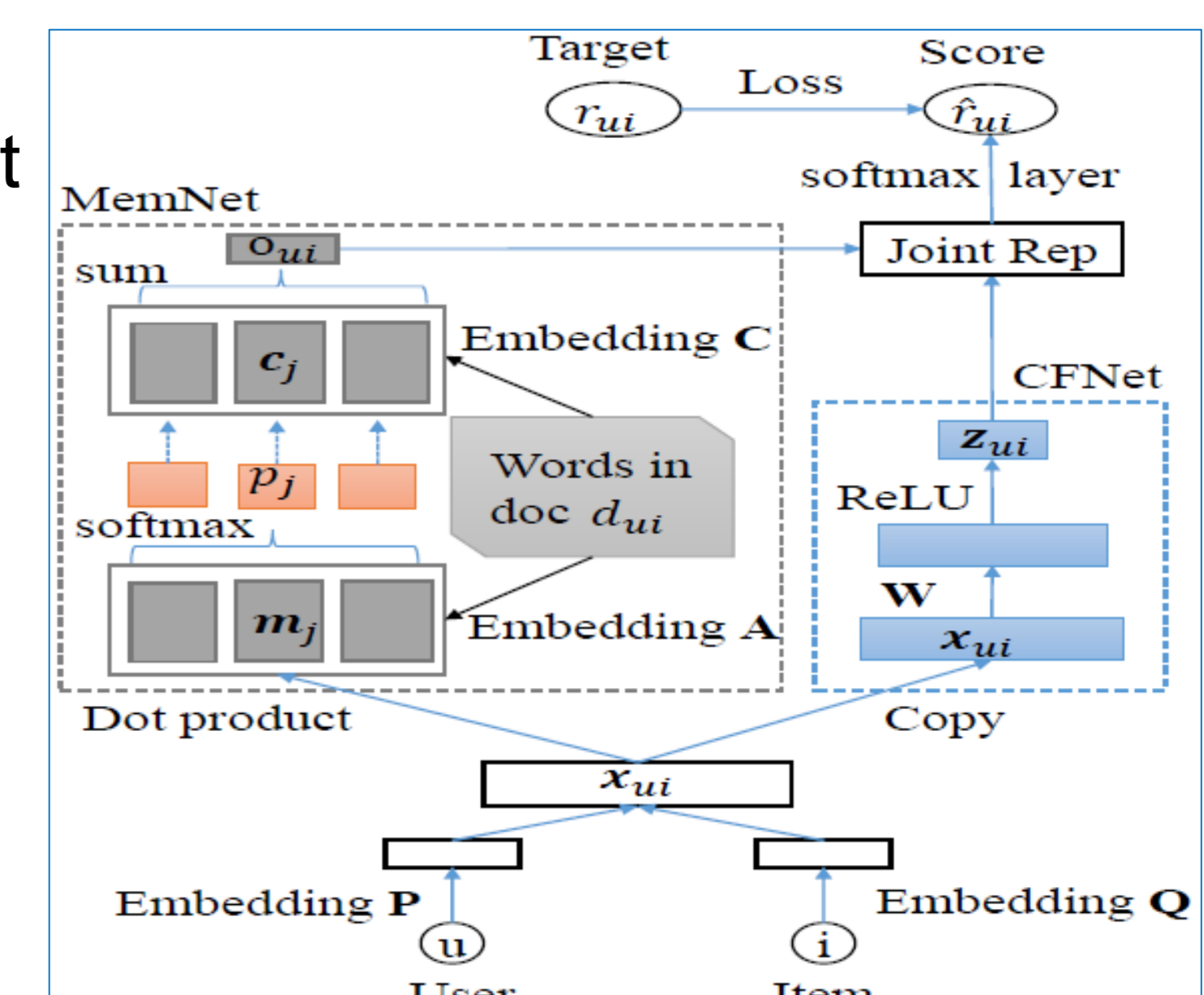
## 1 Motivation

Two threads to alleviate the data sparsity issue in collaborative filtering:

- Hybrid filtering methods integrate the content information, e.g. product reviews and news titles

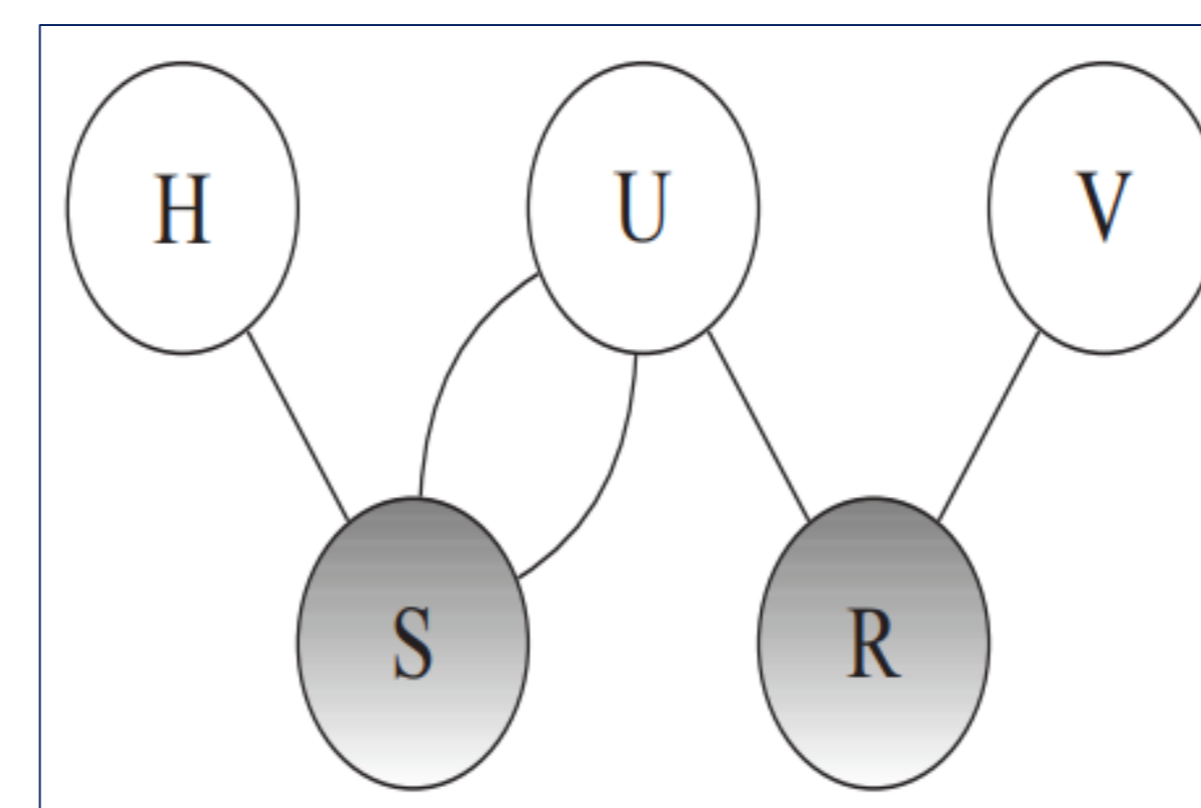


TBPR, Hu et al, PAKDD'17

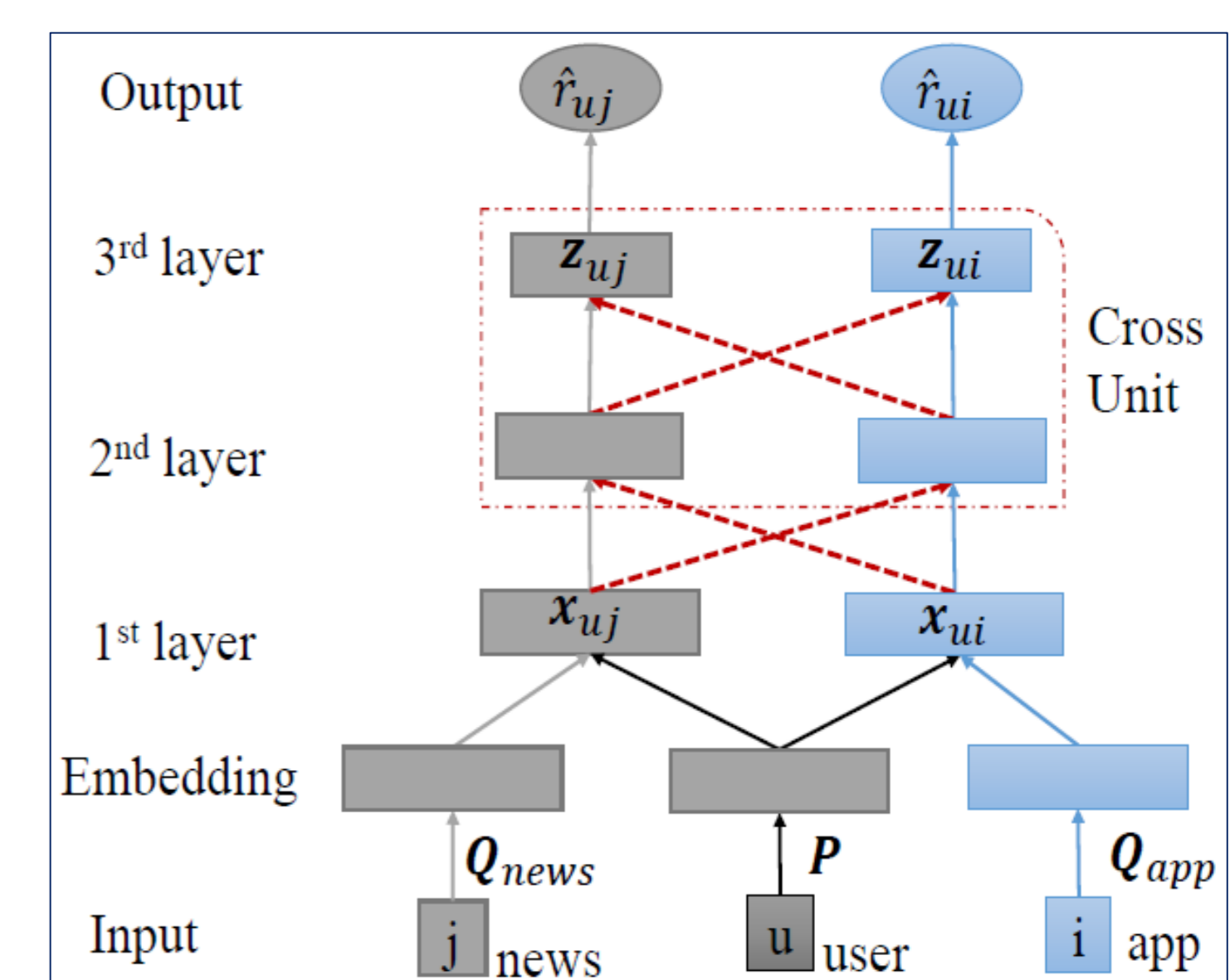


PNE, G. Hu, NAACL'19

- Cross-domain methods leverage the knowledge from a related domain, e.g. from App to News



eSMF, Hu et al, IJCAI'15



CoNet, Hu et al, CIKM'18

Quadrant I

Quadrant IV

## 3 Datasets and Baselines

Dataset	Domain	Statistics	Amount
Mobile News	Shared	#Users	15,890
		#News	84,802
		#Reads	477,685
	Target	Density	0.035%
		#Words	612,839
		Avg. Words Per News	7.2
Source	#Apps	14,340	
	#Installations	817,120	
	Density	0.359%	
Amazon Men	Shared	#Users	8,514
		#Clothes (Men)	28,262
		#Ratings/#Reviews	56,050
	Target	Density	0.023%
		#Words	1,845,387
		Avg. Words Per Review	32.9
Source	#Products (Sports)	41,317	
	#Ratings/#Reviews	81,924	
	Density	0.023%	

- Preprocessing:
  - Shared users between two domains
  - Filter users who have too few interactions
  - Filter items which have too few interactions
- Check Python script on my homepage  
<http://home.cse.ust.hk/~ghuac/>

Baselines	Shallow method	Deep method
Single-domain	BPRMF	MLP
Cross-domain	CDCF, CMF	MLP++, CSN
Hybrid	HFT, TextBPR	LCMR
Cross + Hybrid	CDCF++	TMH (ours)



## 4 Results

Table 2: Results on Amazon data.

Method	$topK = 5$			$topK = 10$			$topK = 20$		
	HR	NDCG	MRR	HR	NDCG	MRR	HR	NDCG	MRR
BPRMF	.0810	.0583	.0509	.1204	.0710	.0561	.1821	.0864	.0602
CDCF	.1295	.0920	.0797	.2070	.1167	.0897	.3841	.1609	.1015
CMF	.1498	.0950	.0771	.2224	.1182	.0863	.3573	.1521	.0957
HFT	.1077	.0815	.0729	.1360	.0907	.0767	.2782	.1252	.0854
TextBPR	.1517	.1208	.1104	.1777	.1291	.1138	.2268	.1414	.1171
CDCF++	.1314	.0926	.0800	.2102	.1177	.0901	.3822	.1605	.1016
MLP	.2100	.1486	.1283	.2836	.1697	.1371	.3820	.1899	.1426
MLP++	.2263	.1626	.1417	.2992	.1862	.1514	.3810	.2069	.1570
CSN	.2340*	.1680*	.1462*	.3018*	.1898*	.1552*	.3944*	.2091*	.1605*
LCMR	.2024	.1451	.1263	.2836	.1678	.1356	.3951	.1918	.1420
TMH	.2575	.1796	.1550	.3490	.2077	.1666	.4443	.2311	.1727
Our improve	10.04%	6.90%	6.01%	15.63%	9.43%	7.34%	12.65%	10.52%	7.60%

