

# Translative Neural Team Recommendation

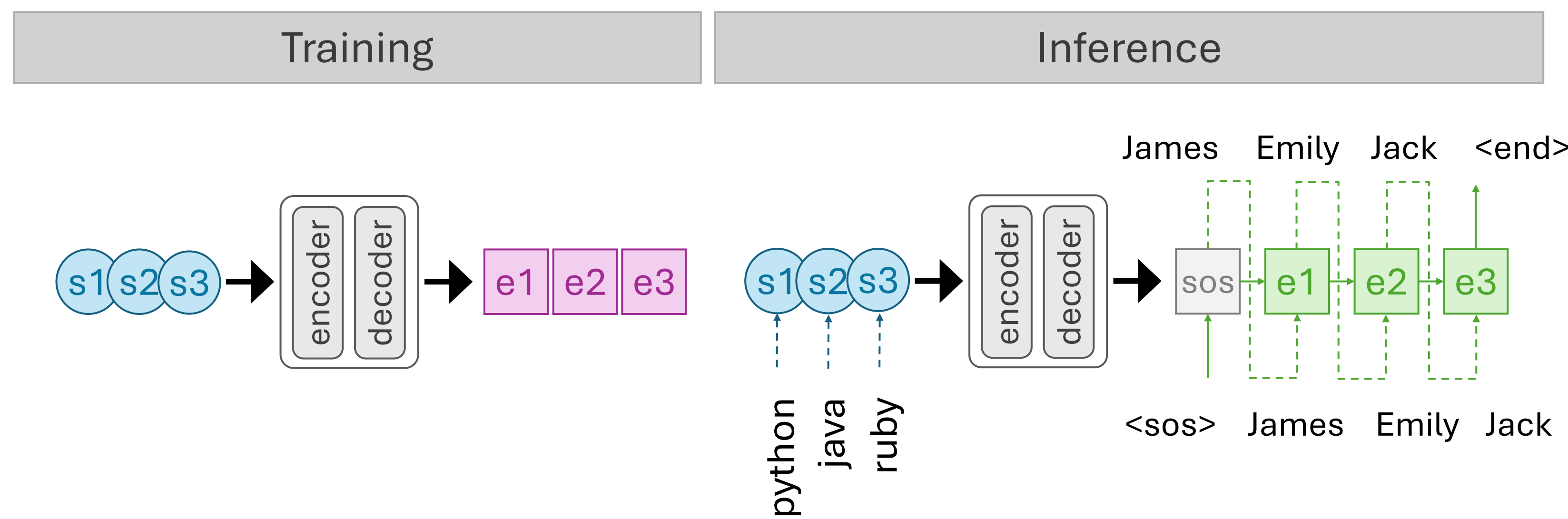
## From Multilabel Classification to Sequence Prediction

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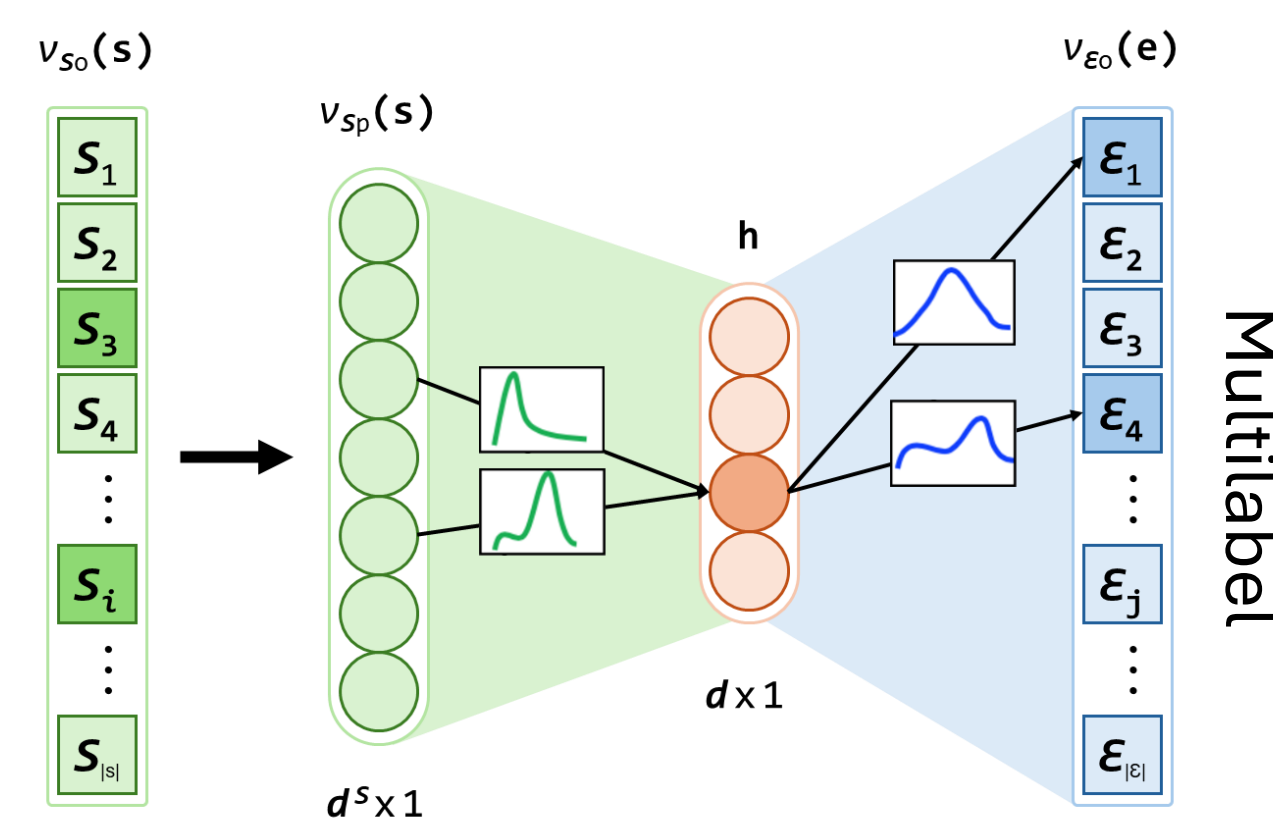


### Workflow



### Background

**Problem:** Current neural team recommendation uses multilabel classification (skills  $\rightarrow$  experts) but suffers from **sparsity curse** in high-dimensional expert outputs.



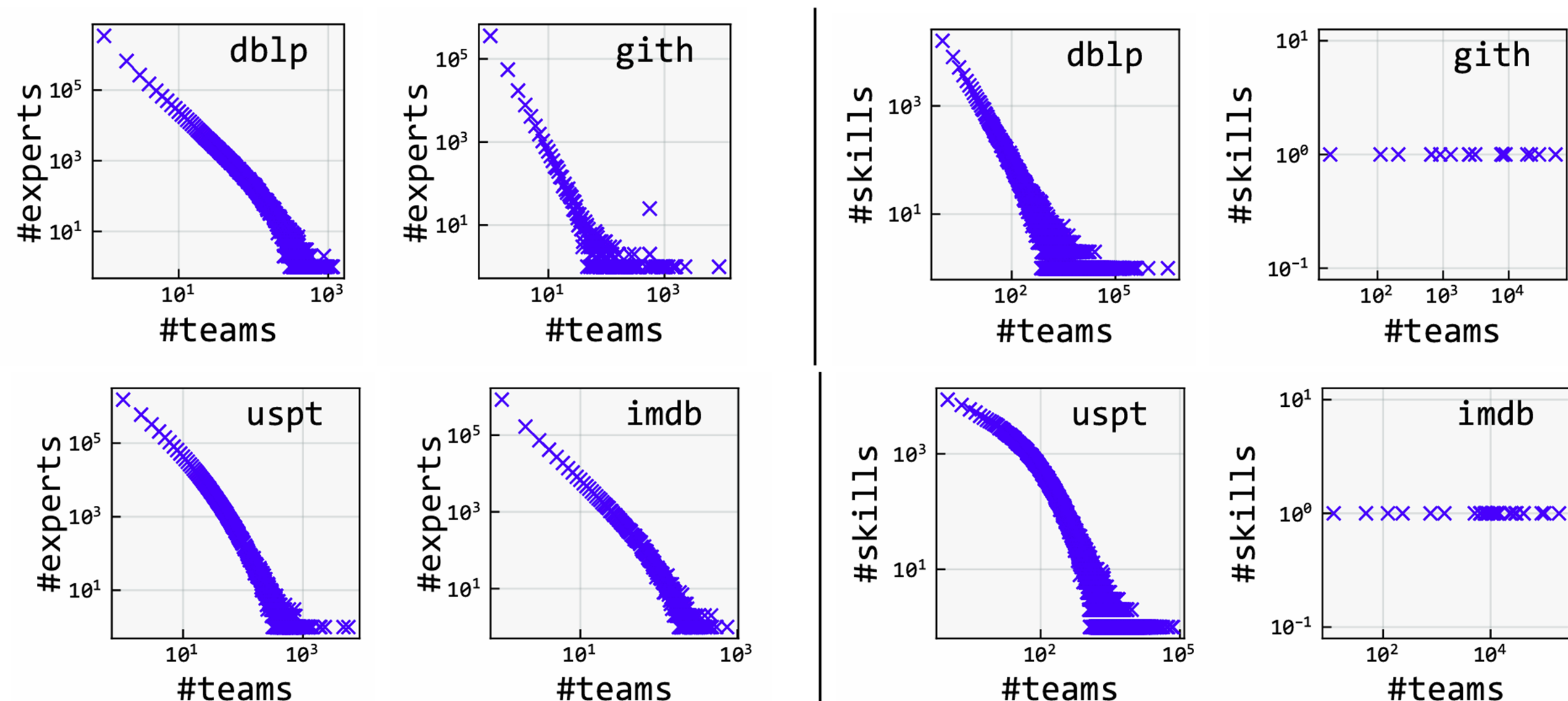
**Solution:** Reformulate as **sequence prediction** using seq-to-seq models

- Input: sequence of required skills
- Output: sequence of optimal experts
- Captures team dynamics through autoregressive modeling

**Advantage:** Sequential approach overcomes sparsity limitations and models real team formation patterns.

### Datasets

	dblp		uspt		imdb		gith	
teams $\mathcal{T}$	publications		patents		movies		software repos	
experts $\mathcal{E}$	authors		inventors		cast & crew		developers	
skills $\mathcal{S}$	keywords		subclasses		(sub) genres		prog. lang.	
success	published		issued		produced		released	
statistics	raw	prep.	raw	prep.	raw	prep.	raw	prep.
$ \mathcal{T} $	4.9M	99K	7.1M	152K	507K	32K	133K	46K
$ \mathcal{E} $	5.0M	14K	3.5M	13K	877K	2.0K	453K	1.2K
$ \mathcal{S} $	90K	30K	242K	67K	28	23	20	20
#teams w/ one expert	769K	0	2.6M	0	323K	0	0	0
avg. #experts/team	3.06	3.29	2.51	3.79	1.88	3.98	5.52	7.53
avg. #skills/team	8.57	9.71	6.29	9.97	1.54	1.76	1.37	1.57



### Results

	k	transformer	convs2s	rnn-att	rrn	bnn	bnn_emb
dblp							
%precision	2	<b>10.4119</b>	2.4998	<u>3.6176</u>	0.0570	0.0570	0.1124
	5	<b>7.0113</b>	1.6122	<u>2.3581</u>	0.0391	0.0663	0.1290
	10	<b>3.5392</b>	0.8242	<u>1.1992</u>	0.0472	0.0710	0.1251
%recall	2	<b>6.3457</b>	1.5071	<u>2.1698</u>	0.0380	0.0351	0.0668
	5	<b>10.5477</b>	2.4177	<u>3.5115</u>	0.0630	0.0993	0.1909
	10	<b>10.6397</b>	2.4760	<u>3.5753</u>	0.1552	0.2118	0.3699
%ndcg	2	<b>10.3611</b>	2.4770	<u>3.5822</u>	0.0478	0.0538	0.1083
	5	<b>10.4597</b>	2.4276	<u>3.5184</u>	0.0523	0.0806	0.1555
	10	<b>10.4824</b>	2.4487	<u>3.5391</u>	0.0959	0.1330	0.2397
%map	2	<b>5.9463</b>	1.3554	<u>1.9412</u>	0.0217	0.0242	0.0474
	5	<b>9.2909</b>	2.0008	<u>2.8791</u>	0.0281	0.0411	0.0792
	10	<b>9.3210</b>	2.0127	<u>2.8930</u>	0.0446	0.0558	0.1033
uspt							
%precision	2	<b>41.7289</b>	<u>28.5717</u>	23.9729	0.0239	0.0657	0.3663
	5	<b>31.0677</b>	<u>24.6530</u>	17.7873	0.0383	0.0769	0.4123
	10	<b>16.5169</b>	<u>15.2382</u>	9.4717	0.0654	0.0910	0.3748
%recall	2	<b>23.1038</b>	<u>13.9104</u>	12.9871	0.0140	0.0353	0.1608
	5	<b>41.1643</b>	<u>28.8167</u>	23.0358	0.0500	0.0976	0.4509
	10	<b>42.6086</b>	<u>33.7595</u>	23.8896	0.1370	0.2212	0.8141
%ndcg	2	<b>41.6095</b>	<u>28.3606</u>	23.9146	0.0221	0.0655	0.3652
	5	<b>42.0309</b>	<u>30.0325</u>	23.8227	0.0408	0.0883	0.4531
	10	<b>42.1435</b>	<u>31.4137</u>	23.8270	0.0868	0.1481	0.6094
%map	2	<b>22.4053</b>	<u>13.0305</u>	12.4784	0.0096	0.0266	0.1212
	5	<b>38.6272</b>	<u>24.4598</u>	21.3567	0.0186	0.0433	0.2027
	10	<b>39.7591</b>	<u>27.1994</u>	22.0112	0.0340	0.0592	0.2583
imdb							
%precision	2	1.5454	<u>1.6097</u>	<b>1.6985</b>	0.0000	0.2128	0.4255
	5	<u>1.4574</u>	1.4552	<b>1.4804</b>	0.8511	0.5106	0.5106
	10	<b>0.9035</b>	0.8998	<u>0.9027</u>	0.8511	0.4255	0.6383
%recall	2	0.7669	<u>0.7952</u>	<b>0.8193</b>	0.0000	0.1418	0.2837
	5	<b>1.8093</b>	1.8013	<u>1.8043</u>	1.4184	0.8511	0.8511
	10	<b>2.2085</b>	2.1926	<u>2.1792</u>	2.8369	1.3050	1.9574
%ndcg	2	1.5479	<u>1.6173</u>	<b>1.7003</b>	0.0000	0.1646	0.3292
	5	1.7364	<u>1.7595</u>	<b>1.7883</b>	0.8163	0.5699	0.5923
	10	1.9039	<u>1.9222</u>	<b>1.9333</b>	1.4606	0.7848	1.1358
%map	2	0.6172	<u>0.6506</u>	<b>0.6650</b>	0.0000	0.0709	0.1418
	5	1.0327	<b>1.0487</b>	<u>1.0450</u>	0.3191	0.2600	0.2813
	10	1.0914	<b>1.1041</b>	<u>1.0975</u>	0.6265	0.3148	0.4389
gith							
%precision	2	<b>32.1596</b>	25.0590	<u>29.7008</u>	0.0000	3.0693	7.3267
	5	<b>21.6055</b>	16.9509	<u>20.1806</u>	0.1980	2.8515	4.7129
	10	<b>12.7104</b>	9.9503	<u>12.0029</u>	0.0990	2.6931	3.3861
%recall	2	<b>13.8543</b>	11.0787	<u>12.8103</u>	0.0000	1.2164	3.5441
	5	<b>22.2914</b>	18.0735	<u>20.7963</u>	0.0619	2.8846	5.1580
	10	<b>24.0868</b>	19.4837	<u>22.6186</u>	0.0619	5.1174	6.1885
%ndcg	2	<b>32.4291</b>	25.3569	<u>29.7647</u>	0.0000	3.1365	6.4753
	5	<b>28.2538</b>	22.3664	<u>26.1975</u>	0.1679	3.2893	5.8418
	10	<b>26.9900</b>	21.5849	<u>25.1263</u>	0.1090	4.2340	6.2665
%map	2	<b>12.9552</b>	10.5164	<u>11.9759</u>	0.0000	1.0104	2.3424
	5	<b>19.3215</b>	15.5615	<u>17.9982</u>	0.0206	1.5706	3.0822
	10	<b>20.7984</b>	16.6428	<u>19.4844</u>	0.0206	2.1633	3.3837

	transformer	convs2s	rnn-att	rrn, bnn bnn_emb
batch size	128	8 <sup>+</sup> , 128	128	128
learning rate	-----	Vaswani et al. [56]	-----	0.1
epochs	20	1 <sup>+</sup> , 20	20	20
optimizer	-----	Adam	-----	-----
hidden layer size	512	128	128, 512*	128
hidden activation	relu, softmax	glu	tanh, sigmoid	relu
output layer	128	128	128	$ \mathcal{E} $
output activation	-----	softmax	-----	sigmoid

\*: convs2s model setting for uspt dataset.



Paper

Codebase