

# SPOT-Trip: Dual-Preference Driven Out-of-Town Trip Recommendation

Yinghui Liu<sup>1\*</sup>, Hao Miao<sup>2\*</sup>, Guojiang Shen<sup>1</sup>, Yan Zhao<sup>3</sup>, Xiangjie Kong<sup>1†</sup> and Ivan Lee<sup>4</sup>

1 Zhejiang University of Technology, 2 Hong Kong Polytechnic University, 3 University of Electronic Science and Technology of China, 4 University of South Australia

\* The authors contributed equally, †Corresponding author: xjkong@ieee.org

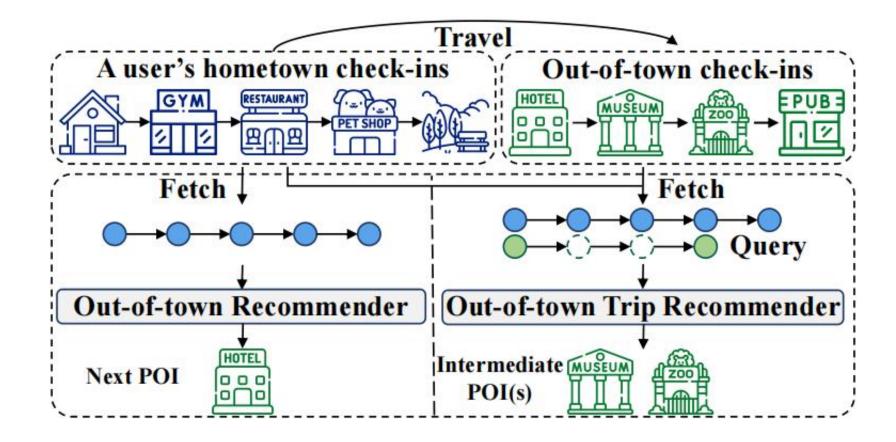








#### **Motivation**



**Out-of-town trip recommendation** aims to generate a sequence of Points of Interest (POIs) for users traveling from their hometowns to previously unvisited regions based on personalized itineraries, e.g., origin, destination, and trip duration.

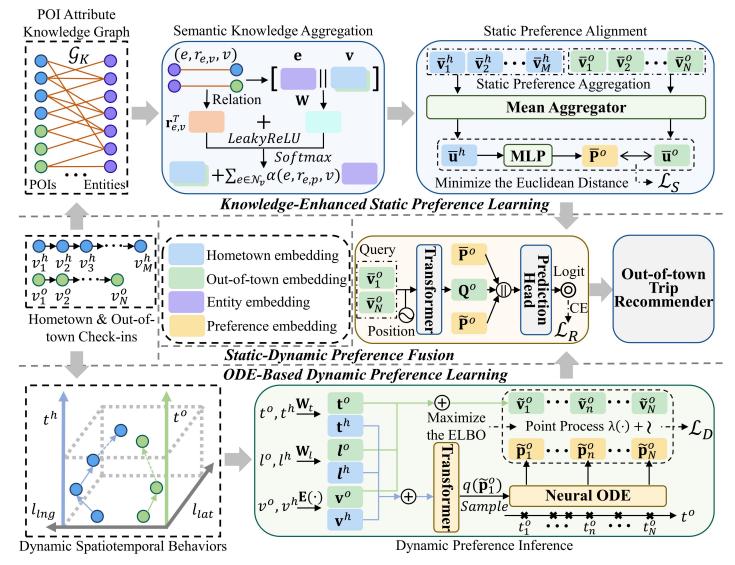
### Challenge

• Data Sparsity. Users often have few or even no historical check-in records in out-of-town regions.

• Complex User Preferences. Intuitively, the user preferences can be categorized into two complementary components: static (or invariant) and dynamic preferences.

#### Method

We propose a Static-dynamic Preference aware Out-of-Town Trip recommendation framework, SPOT-Trip.



## **Experiment**

Table 1: The overall comparison between SPOT-Trip and baselines, where the best performance is marked in bold while the second-best results are underlined. \* denotes improvements that are statistically significant, where we use two-sided t-test with p-value < 0.05 [31].

Method	Foursquare					Yelp				
Wethod	$F_1(\uparrow)$	$PairsF_1(\uparrow)$	$Full-F_1(\uparrow)$	$Full-PairsF_1(\uparrow)$	$F_1(\uparrow)$	$PairsF_1(\uparrow)$	$Full-F_1(\uparrow)$	$Full$ - $PairsF_1(\uparrow$		
PersTour [26]	0.0258	0.0016	0.4421	0.1572	0.0251	0.0066	0.5059	0.2074		
Popularity [6]	0.0261	0.0013	0.4423	0.1565	0.0257	0.0056	0.5065	0.2058		
POIRank [6]	0.0253	0.0019	0.4416	0.1582	0.0264	0.0079	0.5068	0.2093		
GraphTrip [14]	0.0295	0.0048	0.4498	0.1620	0.0289	0.0126	0.5107	0.2184		
MatTrip [54]	0.0311	0.0037	0.4530	0.1656	0.0301	0.0119	0.5117	0.2191		
AR-Trip [39]	0.0304	0.0045	0.4512	0.1673	0.0307	0.0153	0.5115	0.2204		
Base	0.0339	0.0069	0.4571	0.1698	0.0315	0.0149	0.5097	0.2215		
Base + KDDC [32]	0.0375	0.0079	0.4606	0.1822	0.0341	0.0156	0.5126	0.2256		
Base + CNN-ODE [21]	0.0367	0.0094	0.4578	0.1843	0.0326	0.0168	0.5124	0.2237		
Base + PPROC [21]	0.0330	0.0071	0.4550	0.1687	0.0334	0.0159	0.5110	0.2218		
SPOT-Trip	0.0400*	0.0109*	0.4723*	0.1960*	0.0399*	0.0190*	0.5261*	0.2347*		
Improvement	+6.67%	+15.96%	+2.54%	+6.34%	+17.01%	+13.90%	+2.63%	+4.03%		

SPOT-Trip w/o KS w/o (OD+KA) w/o KA w/o PA w/o SI w/o (OD+PA) w/o AT w/o OD 4.00 2-0 3.75 × 3.50  $\overset{-}{\underset{\times}{\boxtimes}} 3.75$ € 0.90 × 0.75 ≘ 1.60 × 1.40 3.25 3.00 2 3.25 3.00 0.60 0.45 i 1.20 Foursquare Yelp Yelp Foursquare (a) F<sub>1</sub> (b) PairsF<sub>1</sub>

Figure 6: Performance of SPOT-Trip and its variants on two datasets.

Table 2: The effect of data sparsity. We sample different fractions of the training data.

Dataset	Method	Hometown Data during Training								
		40%		60%		80%		100%		
		$F_1(\uparrow)$	$PairsF_1(\uparrow)$	$F_1(\uparrow)$	$PairsF_1(\uparrow)$	$F_1(\uparrow)$	$PairsF_1(\uparrow)$	$F_1(\uparrow)$	$PairsF_1(\uparrow)$	
Foursquare	Base + KDDC Base + CNN-ODE SPOT-Trip	0.0325 0.0314 0.0347	0.0038 0.0046 <b>0.0073</b>	0.0348 0.0336 0.0369	0.0058 0.0052 <b>0.0085</b>	0.0364 0.0352 0.0401	0.0069 0.0074 <b>0.0091</b>	0.0375 0.0367 0.0400	0.0079 0.0094 0.0109	
Yelp	Base + KDDC Base + CNN-ODE SPOT-Trip	$\begin{array}{c} 0.0321 \\ \hline 0.0313 \\ \textbf{0.0327} \end{array}$	0.0140 0.0154 <b>0.0162</b>	$\begin{array}{ c c }\hline 0.0331\\\hline 0.0327\\ \textbf{0.0343}\\ \end{array}$	0.0149 0.0156 <b>0.0178</b>	0.0338 0.0333 0.0384	0.0151 0.0161 <b>0.0216</b>	0.0341 0.0326 0.0399	0.0156 0.0168 <b>0.0190</b>	

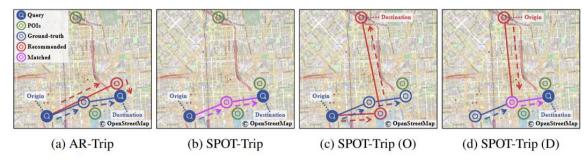


Figure 5: Visualizations of recommendation results for the user 2964 on Foursquare. (O) denotes a query with a single origin, while (D) denotes a query with a single destination.

# Thank you for listening!

