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APPENDIX

Due to the page limit, the current submission of Paper 73 does not include the dataset statistics of Foursquare-LA, and its experiment results. They are all listed in this supplementary material.

Experiments

Datasets

Besides the two datasets Foursquare(NYC) and Gowalla which are provided by [Bao *et al.*, 2012] and [Cheng *et al.*, 2012] respectively, we also evaluate the models on Foursquare(LA) which is provided by [Bao *et al.*, 2012]. The statistics of the three datasets are listed in Table 1. Each dataset is split into two non-overlapping subsets to evaluate the model performance (for each user, the earliest 80% of check-ins as training set, and the remaining 20% check-ins as test set).

Table 1. Dataset Statistics

	#User	#POI	#Check-in
Foursquare-LA	2470	81361	123782
Foursquare-NYC	3401	106974	178143
Gowalla	1488	92679	226116

Performance Comparison on Next POI Recommendation (Foursquare-LA)

Fig.1(a) and Fig.1(b) show the experimental results for the “exact next POI recommendation and the “exact next new POI recommendation on Foursquare-LA dataset. The observations are consistent with that of the other two datasets described in the current submission (See Sec.4.3).

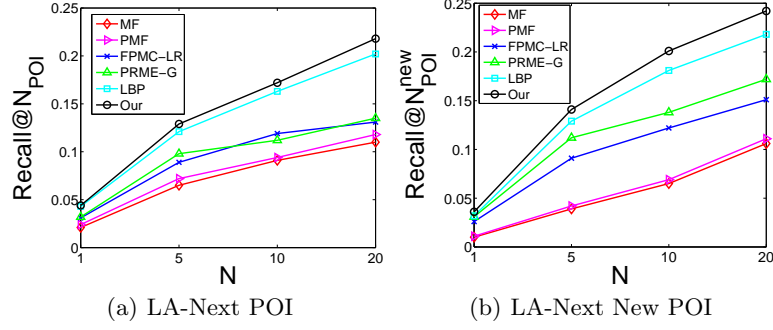


Fig. 1. Performance Comparison on Next POI Recommendation

Performance Comparison on γ -hour Next POI Recommendation (Foursquare-LA, $\gamma = 6$)

To make a fair comparison with the existing work, we further evaluate the performance of next POI recommendation by considering consecutive next check-ins within γ hours as the next location set (γ is set to 6 following [Cheng *et al.*, 2013] and [Feng *et al.*, 2015]).

Fig.2 depicts the comparison on Foursquare-LA dataset when considering the next POI as a set of locations. Similar outcomes to that in paper for the other two datasets can be observed. It is evident that our proposed model consistently outperforms other baselines.

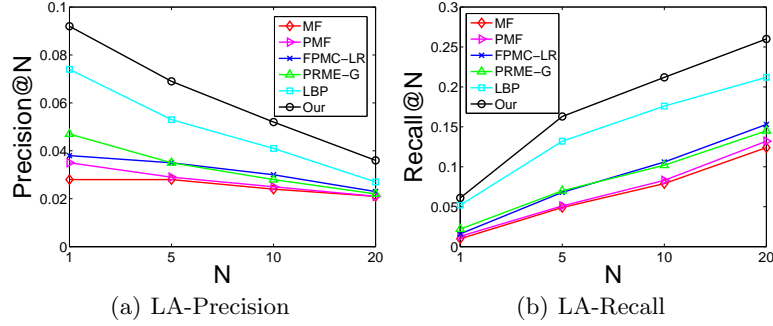


Fig. 2. Performance Comparison on γ -hour Next POI Recommendation($\gamma = 6$)

Performance Comparison on Transition Interval (Foursquare-LA)

Fig.3 shows the performance comparison on transition interval predictions. The results show that the proposed model always achieves the highest precision over

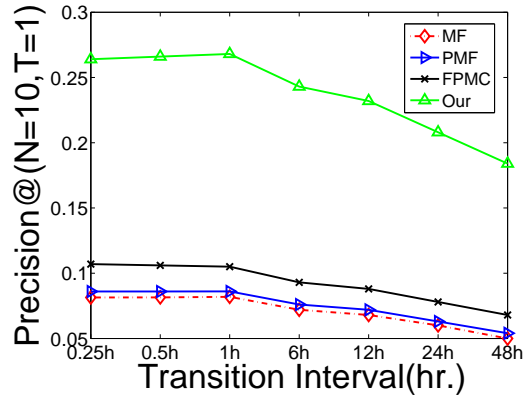


Fig. 3. Performance Comparison for Transition Interval Prediction (Foursquare-LA)

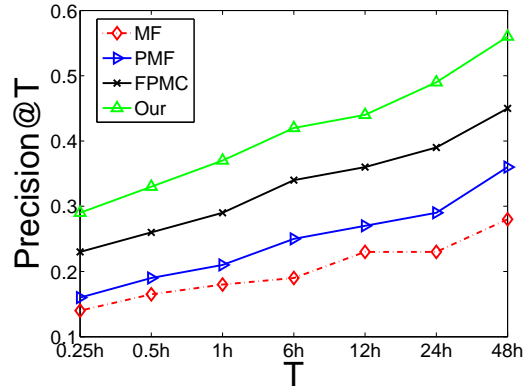


Fig. 4. Transition interval prediction v.s. T (Foursquare-LA)

baselines, which proves that our model is capable of providing effective POI recommendations to users as well as predicting when it will happen. We also compute MAPE between the predicted transition intervals and the ground truth of the test set (See Table 2). Lower values indicate more accurate predictions. It is evident that the proposed model outperforms the baselines by a significant margin. Fig.4 shows the performance comparison by relaxing the threshold T , and our method outperforms all the baselines again. The observations are consistent with that in the submitted paper.

Table 2 tabulates the MAPE between the predicted transition intervals and the ground truth of the test set. Our proposed model achieves more accurate predictions than other baselines on the three datasets.

Table 2. MAPE for our model and baselines on three datasets

	MF	PMF	FPMC	Our
Foursquare-LA	13.79	11.45	5.68	1.75
Foursquare-NYC	14.87	12.64	6.72	1.84
Gowalla	16.95	14.12	7.89	2.15

Detailed Performance Comparisons on Specific Time Period for Next POI Recommendation

In this section, we report the detailed performance comparisons for different settings of future time period. For example, the setting of $TP = [0.5, 1]$ denotes that we consider the next check-ins between 0.5 to 1 hour since the current time point as the next location set. In contrast to MF and PFM, FPMC-LR, PRME-G and LBP are more competitive baselines. To present a more clear comparison, we only report the performance comparisons on recall for the three datasets. The detailed performance comparisons are tabulated in Table 3- Table 8.

Please note that, in general, with the setting of time interval increasing, the number of check-in within the time interval increases which lead the upper bound of recall@N to decrease. Again, our proposed method consistently outperforms the competitive baselines by a large margin for all three datasets.

Table 3. Performance Comparison for Next POI Recommendation, $TP = [0.5, 1]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.030\0.042\0.011	0.027\0.043\0.011	0.045\0.062\0.013	.059\ .075\ .019
Imp.	96.7%\78.6%\72.7%	119%\74.4%\72.7%	31.1%\21.0%\46.2%	
top5	0.099\0.143\0.146	0.098\0.152\0.164	0.197\0.187\0.179	.255\ .233\ .269
Imp.	158%\62.9%\84.2%	160%\53.3%\64.0%	29.4%\24.6%\50.3%	
top10	0.142\0.181\0.223	0.145\0.178\0.237	0.265\0.235\0.276	.327\ .304\ .429
Imp.	130%\68.0%\92.4%	126%\70.8%\81.0%	23.4%\29.4%\55.4%	
top20	0.201\0.205\0.304	0.226\0.209\0.318	0.316\0.306\0.368	.399\ .383\ .554
Imp.	98.5%\86.8%\82.2%	76.5%\83.3%\74.2%	26.3%\25.2%\50.5%	

Table 4. Performance Comparison for Next POI Recommendation, $TP = [1, 2]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.021\0.027\0.009	0.022\0.026\0.010	0.042\0.037\0.012	.051\ .045\ .017
Imp.	143%\66.7%\88.9%	132%\73.1%\70.0%	21.4%\21.6%\41.7%	
top5	0.085\0.126\0.083	0.091\0.130\0.092	0.158\0.168\0.104	.198\ .207\ .157
Imp.	133%\64.3%\89.2%	118%\59.2%\70.7%	25.3%\23.2%\51.0%	
top10	0.133\0.167\0.161	0.141\0.171\0.159	0.246\0.243\0.203	.312\ .299\ .301
Imp.	135%\79.0%\87.0%	121%\74.9%\89.3%	26.8%\23.0%\48.3%	
top20	0.202\0.219\0.231	0.216\0.225\0.228	0.298\0.301\0.301	0.383\ .387\ .433
Imp.	89.6%\76.7%\87.4%	77.3%\72.0%\89.9%	28.5%\28.6%\43.9%	

Table 5. Performance Comparison for Next POI Recommendation, $TP = [2, 3]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.024\0.031\0.011	0.022\0.032\0.012	0.026\0.039\0.013	.034\ .047\ .018
Imp.	41.7%\51.6%\63.6%	54.5%\46.9%\50.0%	30.8%\20.5%\38.5%	
top5	0.132\0.138\0.069	0.126\0.141\0.072	0.173\0.179\0.082	.216\ .222\ .112
Imp.	63.6%\60.9%\62.3%	71.4%\57.4%\55.6%	24.9%\24.0%\36.6%	
top10	0.187\0.189\0.132	0.171\0.194\0.143	0.213\0.231\0.163	.275\ .293\ .218
Imp.	47.1%\55.0%\65.2%	60.8%\51.0%\52.4%	29.1%\26.8%\33.7%	
top20	0.234\0.221\0.193	0.216\0.226\0.192	0.265\0.291\0.243	.339\ .368\ .326
Imp.	44.9%\66.5%\68.9%	56.9%\62.8%\69.8%	27.9%\26.5%\34.2%	

Table 6. Performance Comparison for Next POI Recommendation, $TP = [3, 6]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.023\0.015\0.011	0.025\0.014\0.012	0.031\0.018\0.015	.038\ .022\ .018
Imp.	65.2%\46.7%\63.6%	52.0%\57.1%\50.0%	22.6%\22.2%\20.0%	
top5	0.081\0.071\0.062	0.078\0.072\0.068	0.101\0.085\0.081	.122\ .105\ .105
Imp.	50.6%\47.9%\69.4%	56.4%\45.8%\54.4%	20.8%\23.5%\29.6%	
top10	0.122\0.115\0.116	0.119\0.112\0.124	0.142\0.131\0.148	.173\ .166\ .191
Imp.	41.8%\44.3%\64.7%	45.4%\48.2%\54.0%	21.8%\26.7%\29.1%	
top20	0.168\0.143\0.169	0.171\0.141\0.176	0.200\0.173\0.224	.241\ .212\ .285
Imp.	43.5%\48.3%\68.6%	40.9%\50.4%\61.9%	20.5%\22.5%\27.2%	

Table 7. Performance Comparison for Next POI Recommendation, $TP = [6, 12]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.021\0.018\0.010	0.025\0.017\0.011	0.027\0.021\0.013	.033\ .026\ .016
Imp.	57.1%\44.4%\60.0%	32.0%\52.9%\45.5%	22.2%\23.8%\23.1%	
top5	0.086\0.073\0.054	0.078\0.072\0.061	0.113\0.087\0.069	.142\ .106\ .088
Imp.	65.1%\45.2%\63.0%	82.1%\47.2%\44.3%	25.7%\21.8%\27.5%	
top10	0.125\0.99\0.106	0.119\0.101\0.105	0.176\0.124\0.134	.217\ .151\ .170
Imp.	73.6%\52.5%\60.4%	82.4%\49.5%\61.9%	23.3%\21.8%\26.9%	
top20	0.165\0.145\0.164	0.171\0.151\0.162	0.221\0.178\0.216	.276\ .219\ .267
Imp.	67.3%\51.0%\62.8%	61.4%\45.0%\64.8%	24.9%\23.0%\23.6%	

Table 8. Performance Comparison for Next POI Recommendation, $TP = [12, 24]$

Met.	Foursquare-LA\Foursquare-NYC\Gowalla			
	F-LR	P-G	LBP	Our
top1	0.019\0.017\0.013	0.018\0.018\0.014	0.024\0.021\0.017	.031\ .026\ .021
Imp.	63.2%\52.9%\61.5%	72.2%\44.4%\50.0%	29.2%\23.8%\23.5%	
top5	0.072\0.068\0.068	0.073\0.067\0.073	0.098\0.079\0.089	.129\ .100\ .115
Imp.	79.2%\47.1%\69.1%	76.7%\49.3%\57.5%	31.6%\26.6%\29.2%	
top10	0.107\0.104\0.113	0.099\0.109\0.118	0.146\0.131\0.151	.184\ .162\ .188
Imp.	72.0%\55.8%\66.4%	85.9%\48.6%\59.3%	26.0%\23.7%\24.5%	
top20	0.151\0.147\0.168	0.144\0.143\0.172	0.201\0.186\0.224	.244\ .226\ .278
Imp.	61.6%\53.7%\65.5%	69.4%\58.0%\61.6%	21.4%\21.5%\24.1%	