# MCDAN: A Multi-Scale Context-Enhanced Dynamic Attention Network for Diffusion Prediction

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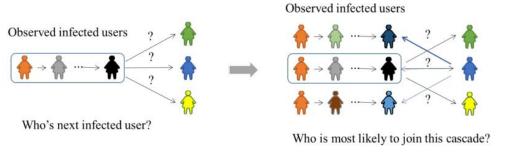
- 1.Introduction
- 2.Method
- 3. Experiments





#### Introduction

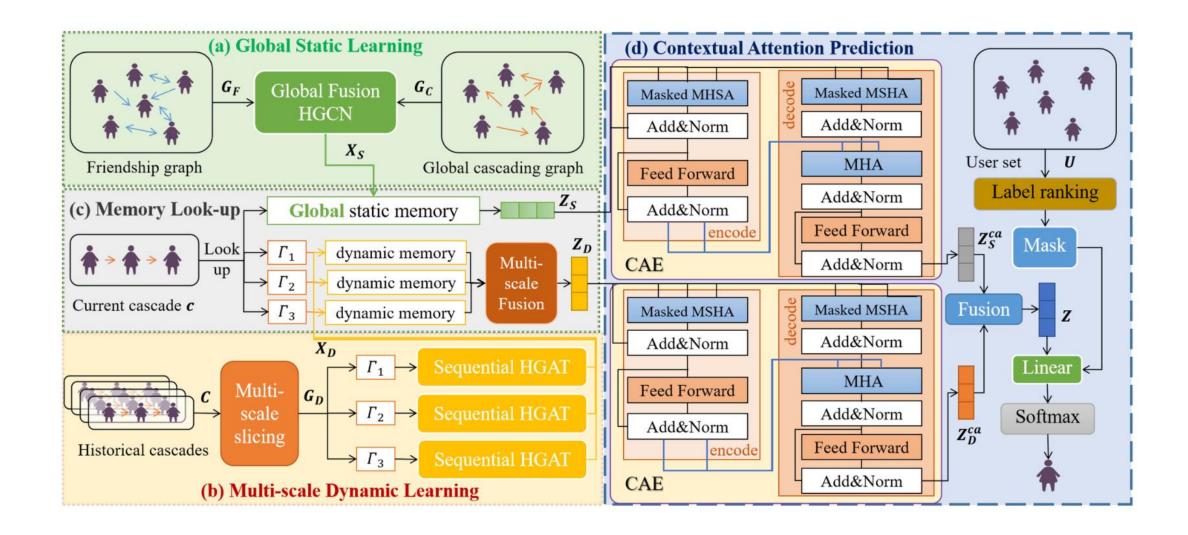
Extracting global interactive relationship among users from the social network and historical cascades.



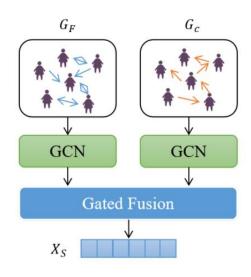
Proposing a multi-scale sequential hypergraph attention module to capture the dynamic preference of users at different time scales.

Designing a contextual attention enhancement module to strengthen the interaction of user representations within the current cascade.

Constructing a susceptibility label for each user based on user susceptibility analysis and use the rank of this label for auxiliary prediction.



#### Method



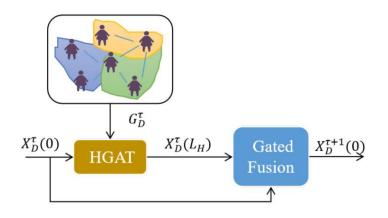
$$X_F(l+1) = ReLU(\widetilde{D}_F^{-\frac{1}{2}}\widetilde{A}_F\widetilde{D}_F^{-\frac{1}{2}}X_F(l)W_F)$$
 (1)

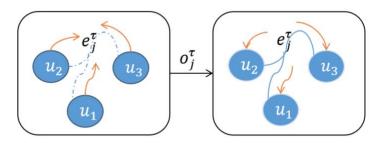
$$X_C(l+1) = ReLU(\widetilde{D}_C^{-\frac{1}{2}}\widetilde{A}_C\widetilde{D}_C^{-\frac{1}{2}}X_C(l)W_C)$$
 (2)

$$X_S = \alpha X_F + (1 - \alpha) X_C \tag{3}$$

$$\alpha = \frac{exp(W_S^T \sigma(W_1 X_F))}{exp(W_S^T \sigma(W_1 X_F)) + exp(W_S^T \sigma(W_1 X_C))}$$
(4)

#### Method





Nodes to hyperedge

Hyperedge to nodes

$$Z_{D} = \sum_{\Gamma \in \{\Gamma_{1}, \Gamma_{2}, \dots, \Gamma_{M}\}} m_{\Gamma} Z_{D}^{\Gamma}$$

$$m_{\Gamma} = \frac{exp(W_{D}^{T} \sigma(W_{m} Z_{D}^{\Gamma}))}{\sum_{\Gamma} exp(W_{D}^{T} \sigma(W_{m} Z_{D}^{\Gamma}))}$$
(12)

$$G_{D} = \{G_{D}^{\Gamma} | \Gamma = \Gamma_{1}, \Gamma_{2}, \dots, \Gamma_{M}\}$$

$$G_{D}^{\Gamma} = \{G_{D}^{\tau} = (U^{\tau}, E_{D}^{\tau}) | \tau = 1, 2, 3, \dots, \Gamma\}$$
(5)

$$X_D = \{X_D^{\Gamma} | \Gamma = \Gamma_1, \Gamma_2, ..., \Gamma_M \}$$

$$X_D^{\Gamma} = Sequential HGAT(G_D^{\Gamma})$$
(6)

$$X_D^{\tau}(L_H) = HGAT(X_D^{\tau}(0), G_D^{\tau}) \quad \tau = 1, 2, 3, ..., \Gamma$$
 (7)

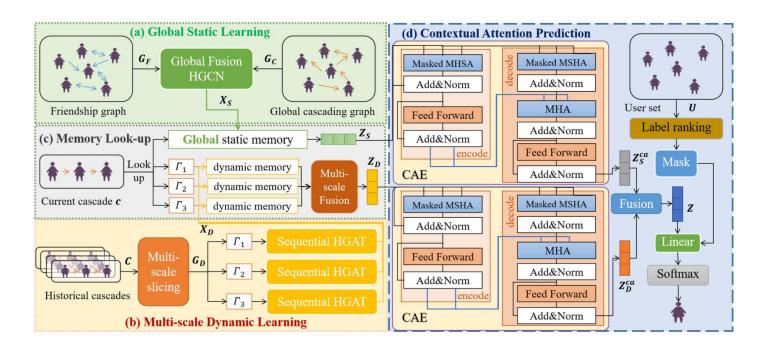
$$X_D^{\tau+1}(0) = g_f X_D^{\tau}(0) + (1 - g_f) X_D^{\tau}(L_H)$$
(8)

$$g_f = \frac{exp(W_{DH}^T \sigma(W_g X_D^{\tau}(0)))}{exp(W_{DH}^T \sigma(W_g X_D^{\tau}(0))) + exp(W_{DH}^T \sigma(W_g X_D^{\tau}(L_H)))}$$
(9)

$$o_j^{\tau}(l+1) = ReLU\left(\sum_{u_i^{\tau} \in e_j^{\tau}} W_{h_1} x_i^{\tau}(l)\right)$$
 (10)

$$x_i^{\tau}(l+1) = ReLU\left(\sum_{e_j^{\tau} \in E_{D,i}^{\tau}} W_{h_2} o_j^{\tau}(l+1)\right)$$
 (11)

#### Method



$$MaskedAtt(Q, K, V) = softmax \left(\frac{QK^{T}}{\sqrt{d_{\Omega}}} + \mathbb{M}\right) V,$$

$$h_{\omega} = MaskedAtt(Z_{S}W_{\omega}^{Q}, Z_{S}W_{\omega}^{K}, Z_{S}W_{\omega}^{V}), \qquad Z_{S}^{c} = ReLU(h_{S}W_{E_{1}} + b_{1})W_{E_{2}} + b_{2} \qquad (14)$$

$$h_{S} = [h_{1}; h_{2}; \dots; h_{\Omega}]W^{O}$$

$$Att(Q, K, V) = softmax \left(\frac{QK^{T}}{\sqrt{d_{\Omega}}}\right) V,$$

$$h_{\omega}^{ca} = Att(h_{S}^{c}W_{\omega}^{Q^{ca}}, Z_{S}^{c}W_{\omega}^{K^{ca}}, Z_{S}^{c}W_{\omega}^{V^{ca}}),$$

$$h_{S}^{ca} = [h_{1}^{ca}; h_{2}^{ca}; \dots; h_{\Omega}^{ca}]W^{O^{ca}}$$

$$(15)$$

$$Z_S^{ca} = ReLU(h_S^{ca}W_{E_3} + b_3)W_{E_4} + b_4$$
(16)

$$Z = \beta Z_S^{ca} + (1 - \beta) Z_D^{ca} \tag{17}$$

$$\beta = \frac{exp(W_Z^T \sigma(W_2 Z_S^{ca}))}{exp(W_Z^T \sigma(W_2 Z_S^{ca})) + exp(W_Z^T \sigma(W_2 Z_D^{ca}))}$$
(18)

$$\hat{y} = softmax(W_n Z + Mask) \tag{19}$$

$$Loss(\theta) = -\sum_{t=2}^{|c|} \sum_{i=1}^{N} y_{ti} log(\hat{y}_{ti})$$
 (20)

TABLE II OVERALL RESULTS WITH HITS@K SCORES FOR K  $=10,\,50,\,100$  on Four Public Datasets (%)

model		Twitter			Douban			Android		(	Christianit	ty
model	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100
DeepDiffuse	5.79	10.80	18.39	9.02	14.93	19.13	4.13	10.58	17.21	10.27	21.83	30.74
Topo-LSTM	8.45	15.80	25.42	8.57	16.53	21.47	4.56	12.63	16.53	12.28	22.63	31.52
NDM	15.21	28.23	32.30	10.00	21.13	30.14	4.85	14.24	18.97	15.41	31.36	45.86
SNIDSA	25.37	36.64	42.89	16.23	27.24	35.59	5.63	15.22	20.93	17.74	34.58	48.76
FOREST	28.67	42.07	49.75	19.50	32.03	39.08	9.68	17.73	24.08	24.85	42.01	51.28
Inf-VAE	14.85	32.72	45.72	8.94	22.02	35.72	5.98	14.70	20.91	18.38	38.50	51.05
DyHGCN	31.88	45.05	52.19	18.71	32.33	39.71	9.10	16.38	23.09	26.62	42.80	52.47
MS-HGAT	33.50	49.59	58.91	21.33	35.25	42.75	10.41	20.31	27.55	28.80	47.14	55.62
Topic-HGAT	35.12	51.41	61.15	23.50	37.58	45.66	11.76	21.72	29.39	30.02	48.73	57.80
RotDiff	35.90	52.46	61.21	22.16	38.23	46.37	11.44	23.04	31.30	32.37	56.25	66.74
MCDAN(ours)	38.45	55.78	64.25	49.39	58.58	62.81	11.89	25.10	32.79	35.49	56.92	67.41

TABLE III OVERALL RESULTS WITH MAP@k Scores for K = 10, 50, 100 on Four Public Datasets (%)

model		Twitter			Douban			Androic	ł	(	Christianit	ty
moder	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100
DeepDiffuse	5.87	6.80	6.39	6.02	6.93	7.13	2.30	2.53	2.56	7.27	7.83	7.84
Topo-LSTM	8.51	12.68	13.68	6.57	7.53	7.78	3.60	4.05	4.06	7.93	8.67	9.86
NDM	12.41	13.23	14.30	8.24	8.73	9.14	2.01	2.22	2.93	7.41	7.68	7.86
SNIDSA	15.34	16.64	16.89	10.02	11.24	11.59	2.98	3.24	3.97	8.69	8.94	9.72
FOREST	19.60	20.21	21.75	11.26	11.84	11.94	5.83	6.17	6.26	14.64	15.45	15.58
Inf-VAE	19.80	20.66	21.32	11.02	11.28	12.28	4.82	4.86	5.27	9.25	11.96	12.45
DyHGCN	20.87	21.48	21.58	10.61	11.26	11.36	6.09	6.40	6.50	15.64	16.30	16.44
MS-HGAT	22.49	23.17	23.30	11.72	12.52	12.60	6.39	6.87	6.96	17.44	18.27	18.40
Topic-HGAT	23.71	24.53	24.66	12.70	13.61	13.72	6.80	7.53	7.68	18.98	19.85	19.99
RotDiff	24.06	24.82	24.95	11.70	12.54	12.66	6.96	7.45	7.56	19.81	20.91	21.05
MCDAN(ours)	25.89	26.69	26.81	40.70	41.13	41.19	7.47	8.04	8.15	22.88	23.78	23.94

#### TABLE I STATISTICS OF THE PREPROCESSED DATASETS IN OUR EXPERIMENTS

Datasets	Twitter	Douban	Android	Christianity
# Users	12,627	12,232	2,927	1,651
# Fri. Links	309,631	198,496	24,459	21,955
# Cas. Links	73,036	51,797	23,958	11,328
# Cascades	3,442	3,475	678	589
Avg. Length	32.60	21.76	42.05	26.02

TABLE IV Ablation Study With Hits@k Scores for K = 10, 50, 100 on Four Public Datasets (%)

model	model Twitter				Douban			Android		(	Christianity		
moder	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100	
MCDAN	38.45	55.78	64.25	49.39	58.58	62.81	11.89	25.10	32.79	35.49	56.92	67.41	
w/o G w/o M w/o C	32.02 38.16 35.41	49.94 54.96 50.49	60.27 63.17 58.55	28.97 39.65 20.50	42.72 52.43 34.86	49.11 57.58 41.97	10.88 11.58 11.11	22.61 22.92 21.45	29.99 30.61 28.67	32.59 31.47 31.92	51.12 52.46 52.46	64.96 62.05 61.38	
w/o L	34.94	51.64	60.58	45.77	55.64	60.05	11.34	21.06	28.44	32.14	52.90	66.29	

Note that we use underlining to mark the results of the most effective component.

TABLE V Ablation Study With MAP@k Scores for K=10,50,100 on Four Public Datasets (%)

model	Twitter				Douban			Android	i	Christianity		
moder	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100
MCDAN	25.89	26.69	26.81	40.70	41.13	41.19	7.47	8.04	8.15	22.88	23.78	23.94
w/o G w/o M w/o C w/o L	19.90 25.18 23.59 22.73	20.72 25.96 24.29 23.49	20.87 26.08 24.40 23.62	19.68 29.89 11.17 37.36	20.30 30.50 <u>11.87</u> 37.81	20.39 30.57 11.97 37.87	6.63 6.99 6.85 7.09	7.17 7.50 7.29 7.52	7.27 7.61 7.39 7.62	20.34 20.20 19.26 19.64	21.23 21.14 20.16 20.52	21.42 21.28 20.28 20.71

Note that we use underlining to mark the results of the most effective component.

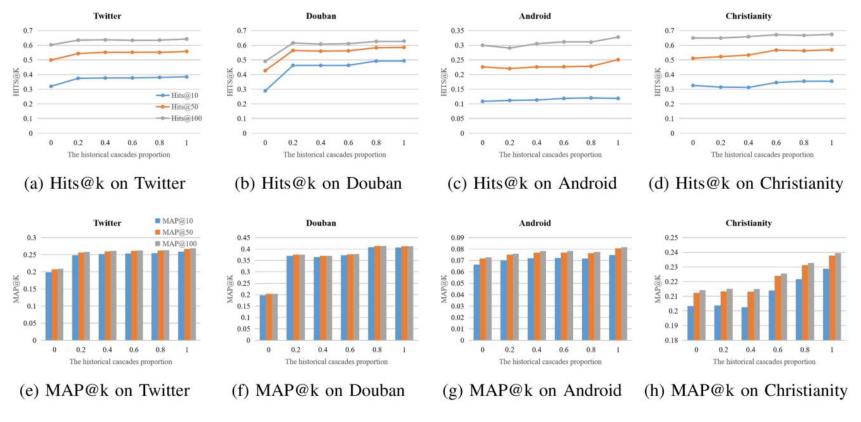


Fig. 6. Results of the impact of the historical cascades proportion on the four public datasets.

TABLE VI RESULTS OF THE IMPACT OF THE NUMBER OF TIME SCALES M on Four Public Datasets (%)

M	Г		Twitter		Douban				Android		Christianity		
1V1	i.	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100
1	8	38.16	54.96	63.17	39.65	52.43	57.58	11.58	22.92	30.61	31.47	52.46	62.05
3	4,8,16	38.45	55.78	64.25	49.39	58.58	62.81	11.89	25.10	32.79	35.49	56.92	67.41
5	2,4,8,16,32	34.86	52.04	61.25	45.11	54.96	59.88	10.65	22.92	31.24	32.37	50.89	65.63

Hits@k scores for k = 10, 50, 100.

TABLE VII RESULTS OF THE IMPACT OF THE NUMBER OF TIME SCALES  ${\cal M}$  on Four Public Datasets (%)

M	Г		Twitter			Douban			Android	I	(	Christianit	ty
171	1	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100
1	8	25.18	25.96	26.08	29.89	30.50	30.57	6.99	7.50	7.61	20.20	21.14	21.28
3	4,8,16	25.89	26.69	26.81	40.70	41.13	41.19	7.47	8.04	8.15	22.88	23.78	23.94
5	2,4,8,16,32	22.90	23.68	23.82	34.79	35.26	35.33	6.88	7.44	7.56	20.27	21.22	21.43

MAP@k scores for k = 10, 50, 100.



#### TABLE VIII INSUSCEPTIBLE LABEL THRESHOLD SETTING

tratio	Twitter				Douban				Android				Christianity			
$t_{ratio}$	0.02	0.04	0.06	0.08	0.02	0.04	0.06	0.08	0.02	0.04	0.06	0.08	0.02	0.04	0.06	0.08
# insusceptible users	252	505	757	1010	244	489	734	978	58	117	175	234	33	66	99	132

TABLE IX
RESULTS OF THE IMPACT OF THE CASCADE LENGTH ON FOUR PUBLIC DATASETS (%)

$Max_{len}$		Twitter			Douban			Android		Christianity			
Maxlen	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100	
200	38.45	55.78	64.25	49.39	58.58	62.81	11.89	25.10	32.79	35.49	56.92	67.41	
300	37.91	54.92	63.21	49.09	58.12	62.74	11.57	25.21	32.72	34.15	56.47	67.63	
400	37.78	54.58	62.75	49.06	57.93	62.67	11.42	25.44	32.87	33.48	55.58	68.30	
500	37.69	54.31	62.63	48.82	57.72	62.61	11.72	25.44	32.64	33.04	55.13	68.08	

Hits@k scores for k = 10, 50, 100. Note that  $Max_{len}$  denotes the maximum length of the cascades.

TABLE X RESULTS OF THE IMPACT OF THE CASCADE LENGTH ON FOUR PUBLIC DATASETS (%)

Max		Twitter			Douban			Android	l	Christianity			
$Max_{len}$	@10	@50	@100	@10	@50	@100	@10	@50	@100	@10	@50	@100	
200	25.89	26.69	26.81	40.70	41.13	41.19	7.47	8.04	8.15	22.88	23.78	23.94	
300	25.37	26.16	26.27	40.45	40.87	40.94	7.43	8.03	8.14	22.46	23.44	23.60	
400	24.76	25.54	25.65	40.26	40.67	40.73	7.37	7.99	8.10	22.32	23.31	23.50	
500	24.20	24.97	25.09	40.04	40.46	40.53	7.37	7.96	8.06	22.06	23.05	23.24	

MAP@k scores for k = 10, 50, 100. Note that  $Max_{len}$  denotes the maximum length of the cascades.

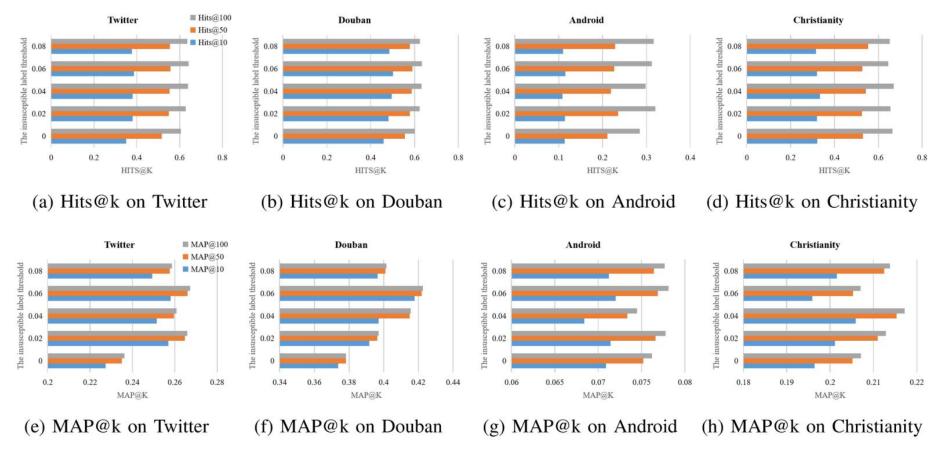


Fig. 7. Results of the impact of the insusceptible label threshold on the four public datasets.

# Thank you!