## Experiments

## Overall Performance

Weibo							Twitter16										
Method	Class	Acc.	Prec.	Rec.	$ F_1 $		1	N	F	т	U			l NI	F	т	U
DTC	F T	0.831	0.847	0.815	0.831 0.819	Method	Acc.	$ F_1 $	$F_1$	$F_1$	$F_1$	Method	Acc.	$egin{array}{c c} N \ \hline F_1 \end{array}$	$F_1$	$F_1$	$F_1$
SVM-RBF	F	0.879	0.777	0.656		DTC	0.454	0.415	0.355	0.733	0.317	DTC	0.473	0.254	0.080	0.190	0.482
	T	1	0.579	0.708	0.615	SVM-RBF	0.318	0.225	0.082	0.455	0.218	SVM-RBF	0.553	0.670	0.085	0.117	0.361
SVM-TS	F T	0.885	0.950 0.124	0.932 0.047	0.938 0.059	SVM-TS	0.544	0.796	0.472	0.404	0.483	SVM-TS	0.574	0.755	0.420	0.571	0.526
RvNN	F	0.908	0.912	0.897		SVM-TK	0.750	0.804	0.698	0.765	0.733	SVM-TK	0.732	0.740	0.709	0.836	0.686
	l E		0.904	0.918	0.911	RvNN	0.723	0.682	0.758	0.821	0.654	RvNN	0.737	0.662	0.743	0.835	0.708
PPC_RNN+CNN	T	0.916	0.884 0.955	0.957 0.876	0.919 0.913	PPC_RNN+CNN	0.477	0.359	0.507	0.300	0.640	PPC_RNN+CNN	0.564	0.591	0.543	0.394	0.674
Bi-GCN	F T	0.961	0.961 0.962	0.964 0.962	0.961 0.960	Bi-GCN	0.886	0.891	0.860	0.930	0.864	Bi-GCN	0.880	0.847	0.869	0.937	0.865

- Bi-GCN outperforms PPC\_RNN+CNN in terms of all the performance measures, indicates the effectiveness of incorporating the dispersion structure for rumor detection.
- Since RNN & CNN can't process data with the graph structure, so ignore important structure features of dispersion.

## Experiments

## **Overall Performance**

Weibo						Twitter15						Twitter16						
Method	Class	Acc.	Prec.	Rec.	$ F_1 $		1	l NI	F	т			1	l NI	F	т		
DTC	F T	0.831	0.847	0.815	0.831 0.819	Method	Acc.	$egin{array}{c c} N \ \hline F_1 \end{array}$	$F_1$	$F_1$	$\dfrac{ ext{U}}{F_1}$	Method	Acc.	$egin{array}{c c} N \ \hline F_1 \end{array}$	$F_1$	$F_1$	$rac{ ext{U}}{F_1}$	
SVM-RBF	F	0.879	0.777	0.656		DTC	0.454	0.415	0.355	0.733	0.317	DTC	0.473	0.254	0.080	0.190	0.482	
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PPC_RNN+CNN	T		0.884			PPC_RNN+CNN	0.477	0.359	0.507	0.300	0.640	PPC_RNN+CNN	0.564	0.591	0.543	0.394	0.674	
Bi-GCN	F T	0.961	0.961 0.962	0.964 0.962	0.961 0.960	Bi-GCN	0.886	0.891	0.860	0.930	0.864	Bi-GCN	0.880	0.847	0.869	0.937	0.865	

- Bi-GCN is significantly superior to the RvNN method, RvNN only uses the hidden feature vector of all the leaf nodes so that it's heavily impacted by information of the latest post (lack of information such as comments, and just follow the former posts).
- Root feature enhancement of Bi-GCN to pay attention to the information of the source posts.