

Table 3: Comparison of different methods: In all MF based methods, the number of latent factors  $K$  is set to 10.

Method	Precision	Recall	F1-score
SVM	0.781 $\pm$ 0.027	0.952 $\pm$ 0.030	0.857 $\pm$ 0.020
SF+SVM	0.810 $\pm$ 0.029	0.949 $\pm$ 0.021	0.874 $\pm$ 0.026
MF+SVM	0.807 $\pm$ 0.036	0.966 $\pm$ 0.038	0.879 $\pm$ 0.029
MF+SR+SVM	0.839 $\pm$ 0.026	0.978 $\pm$ 0.027	0.903 $\pm$ 0.013
SM+SR	0.851 $\pm$ 0.032	0.980 $\pm$ 0.037	0.911 $\pm$ 0.031

We plan to extend our work in the following directions. Firstly, we wish to use other matrix factorization methods to explore additional hidden factors to represent users. In particular, recent advances in tensor factorization could be used to factorize the performer-activity-receiver interaction tensor, which contains richer information than the user-activity matrix used in this work. However, the great sparseness in such representation poses challenges. Secondly, we will integrate our method as a new component into the existing immune system in Renren. Thirdly, we are particularly interested in identifying normal users whose accounts are hijacked by spammers to send spam to his social friends.

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