You Shall Know a User by the Company It Keeps: Dynamic Representations for Social Media Users in NLP

Marco Del Tredici

University of Amsterdam m.deltredici@uva.nl

Diego Marcheggiani* Amazon

marchegg@amazon.es

Sabine Schulte im Walde

University of Stuttgart

schulte@ims.uni-stuttgart.de

Raquel Fernández

University of Amsterdam

raquel.fernandez@uva.nl

1 Extended Results Table

Tables 1, 2 and 3 include, for each task, the results for LING, LING+PV, LING+N2V and LING+GAT as reported in the main paper, together with the Standard Deviation values computed on the ten runs of each model. Additionally, the precision, recall and F-score for each class are reported.

^{*}Research conducted when the author was at the University of Amsterdam.

Model	Av. Rec.	Negative			Neutral			Positive		
		P	R	F1	P	R	F1	P	R	F1
LING	$\underset{\pm 0.005}{0.676}$	0.585	0.656	0.618	0.684	0.678	0.680	0.737	0.694	0.712
LING+PV	$\underset{\pm 0.004}{0.671}$	0.584	0.639	0.609	0.677	0.679	0.678	0.734	0.693	0.713
LING+N2V	$\underset{\pm 0.004}{0.672}$	0.584	0.639	0.609	0.681	0.679	0.680	0.734	0.699	0.715
LING+GAT	$\underset{\pm 0.01}{0.666}$	0.599	0.597	0.596	0.666	0.691	0.677	0.730	0.691	0.709

Table 1: Sentiment Analysis: Average Recall across the three classes, plus precision, recall and F1 per class.

Model	Av. Ag./Fa.	Against			Neutral			Favor		
		P	R	F1	P	R	F1	P	R	F1
LING	$\underset{\pm 0.01}{0.569}$	0.730	0.625	0.672	0.355	0.462	0.399	0.446	0.490	0.466
LING+PV	$\underset{\pm 0.02}{0.601}^*$	0.739	0.673	0.701	0.353	0.380	0.362	0.479	0.536	0.501
LING+N2V	$\underset{\pm 0.01}{0.629^{*\diamondsuit}}$	0.761	0.697	0.727	0.380	0.369	0.370	0.488	0.588	0.531
LING+GAT	$\begin{array}{c} 0.640^{*\diamondsuit^{\dagger}} \\ {\scriptstyle \pm 0.01} \end{array}$	0.749	0.725	0.734	0.380	0.316	0.330	0.507	0.600	0.545

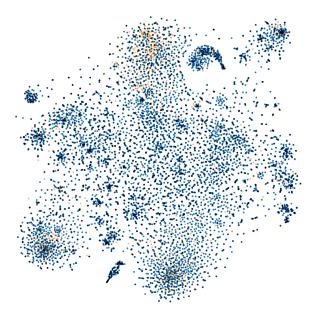
Table 2: Stance Detection: Average F1 of the Against and Favor classes, plus precision, recall and F1 per class.

Model	F1 Hateful		Normal			Hateful	
		P	R	F1	P	R	F1
LING	$\underset{\pm 0.01}{0.624}$	0.968	0.989	0.978	0.773	0.526	0.624
LING+PV	$\underset{\pm 0.02}{0.667^*}$	0.974	0.983	0.979	0.730	0.621	0.667
LING+N2V	$\begin{array}{c} 0.656^* \\ \scriptstyle{\pm 0.008} \end{array}$	0.972	0.986	0.979	0.742	0.589	0.656
LING+GAT	$\begin{array}{c} 0.674^{*\diamondsuit\dagger} \\ {\scriptstyle \pm 0.005} \end{array}$	0.973	0.989	0.980	0.765	0.605	0.674

Table 3: Hate Speech Detection: F1 for the Hateful class, plus precision, recall and F1 per class.

2 Additional Plots

The plot representing the N2V representations for the Hate Speech dataset. Orange dots are authors of tweets labelled as HATEFUL, blue dots of NORMAL tweets.



Below, the same plots in Figure (2) in the main paper showed in a larger size.

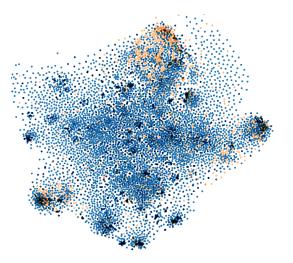


Figure 1: PV user representations for the Hate Speech dataset. Orange dots are authors of HATEFUL tweets, blue dots of NORMAL tweets.

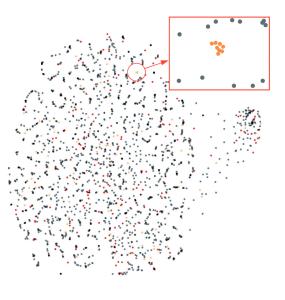


Figure 2: PV user representations for the Stance dataset. Orange dots are authors of tweets AGAINST atheism, red dots authors in FAVOR of 'climate change is a real concern'. All other users are represented as grey dots.

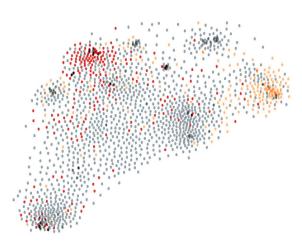


Figure 3: N2V user representations for the Stance dataset. Orange dots are authors of tweets AGAINST atheism, red dots authors in FAVOR of 'climate change is a real concern'. All other users are represented as grey dots.