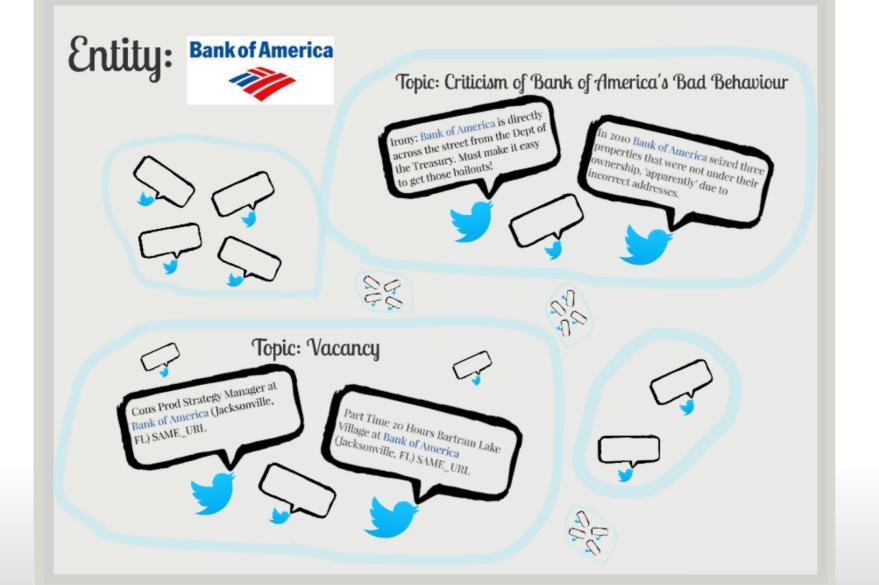


Topic Detection as a Clustering Task

What Are People Saying About a Given Entity Right Now?

> Entity = Brand, Company, Organization, Public Figure

Early Detection of topics that may damage the reputation of the entity



RepLab@CLEF 2013

Evaluation of Online Reputation Monitoring Systems http://nlp.uned.es/replab2013

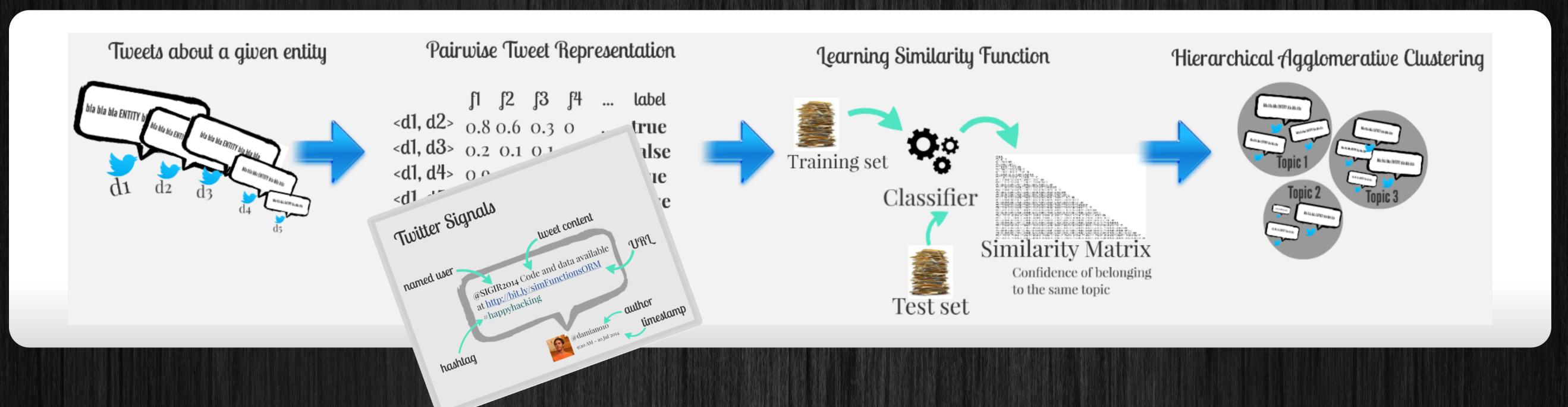
➤Twitter's Long Tail: Sparsity

Dataset for Topic Detection Task

61 entities 100,869 tweets annotated with 8,765 different topics

Training ≠ Test Topics, no overlap in time 544 (1,109) tweets and 57 (87) topics for training (testing) per entity

Learning Similarity Functions



Effect of Twitter Signals

Effect of Training Process

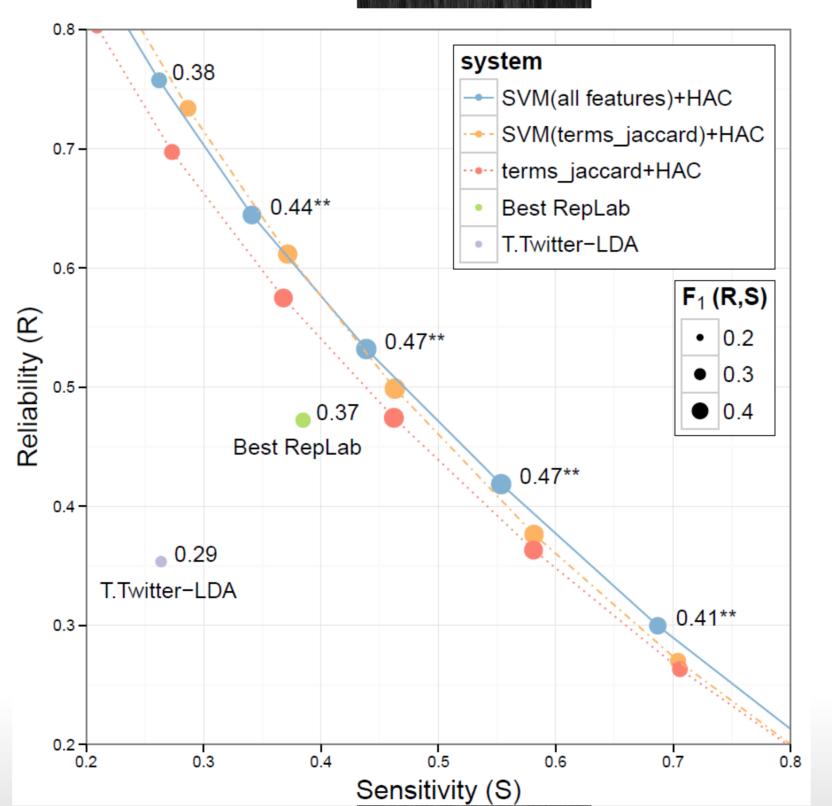
Linear Support Vector Machine (SVM)

➢ High correlation with Maximal Pairwise Accuracy (theoretical upper bound)

➢ Results can be extrapolated to other Machine Learning algorithms

Most signals are able to improve term-based features

System	AUC	MAR
Text only SVM(terms_jaccard)+HAC	0.40	0.59
All Twitter Signals SVM(all features)+HAC	0.41	0.61*

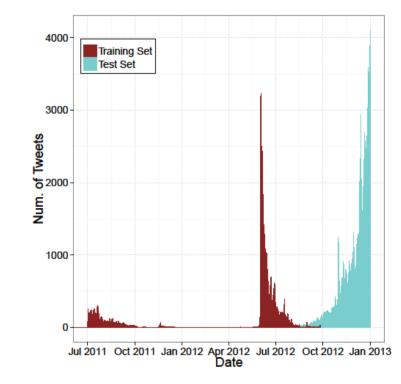


Conclusions

Supervised system consistently improves the performance of its unsupervised counterpart

Regardless of how the similarity threshold is set

Previous annotations can be used to learn better topic models, although differences are not large



System	AUC	MAR
Unsupervised terms_jaccard+HAC	0.38	0.57
Supervised SVM(terms_jaccard)+HAC	0.40	0.59*

➢Twitter signals (authors, timestamps, hashtags, etc.) can indeed be used to improve topic detection w.r.t. textual content only

It is possible to learn effectively from manually annotated topics (pairwise tweet similarity)

➢Organizational ("Hate Opinions") vs. Event-like Topics ("Director of the Bank Accused of Evading Taxes") Learning Similarity Functions + Hierarchical Agglomerative Clustering (HAC) significantly outperforms the state-of-the-art (RepLab Topic Detection systems)

It gets close to the inter-annotator agreement rate: 0.48 F₁(R,S)

When data is sparse (as in the ORM scenario) it can be more effective than using probabilistic generative models such as Temporal Twitter-LDA

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