

Detecting Opinion Spammer Groups

Lars Stelzer
lars.stelzer@studium.uni-hamburg.de
7346178

Niklas von Boguszewski
nvboguszewski@googlemail.com
6790872

Knowledge Processing in Intelligent Systems: Practical Seminar

Knowledge Technology, WTM, Department of Informatics, University of Hamburg

Abstract—This paper aims to detect opinion spammer groups on the Yelp dataset.
Methode nennen
results zeigen (conclusion)

freitext
yelp hat 20 fake reviews
yelp filterd selber

I. Introduction

This chapter shows the motivation of why this topic is important.

Wir haben user mit unserer methode gefunden, die highly suspicious sind und dann ein beispiel nennen warum sind fake reviews doof was ist der impact

leute treffen wirtschaftliche entscheidungen vermehrt anhand der reviews

reviews tragen zur entscheidungsfindung bei und dabei ist ein verlässliches raiting essentiell

viele fake reviews zersören das vertrauen in Yelp und ähnliche Plattformen wie bspw. Amazon.

yelp hat 20 prozent fake reviews und filtert selber es sind noch fake reviews da weitere fake reviews reduzieren gruppenansatz von paper CPM XY Twist nennen story telling eine suspicious gruppe nennen (wo jeder sagen würde es ist klar spam) dann glaubt uns jeder idee warum gruppen -> Venn Diagramm zeigen

II. Related Work

Was haben andere gemacht
single fake review erkenne (3)
gruppen fake reviews erkennen
ergebnisse ...

gruppenansatz von paper CPM XY 2 andere gruppenansätze (man kann Twist nennen)

— In this section we show important work which has been made by other researchers like [2] and [1].

In this section we also want to highlight why this approach can be beneficial for this research area.

III. Yelps Review System

In this section we shortly describe how Yelps review system works.

you can give stars to business
free for all
yelp filters reviews on its own

IV. Dataset

The dataset we use is publically available on Yelp.com and consists of real-world reviews about businesses. Yelp is a recommendations platform based on user-generated reviews. The dataset contains information about reviews, businesses, user data, and geographical data. More than 8 million reviews and more than 200.000 businesses are included in the dataset. For this paper, we look at reviews of 2016 that appeared in the area of Charlotte in North Carolina USA. In total, we examine 51261 reviews and 5954 businesses.

V. Feature

BST

t=28

Review Count for each user

Cosine Similarity

Duplicates/Near Duplicates (beta>0.7)

In this section we describe approximately 3 features we want to experiment with.

Extreme Rating ():

VI. Model

In this section we describe our model/graph which aims to find opinion spamming groups.

1. undirected Graph (G) 2. We look at the reviews of each business 3. Then, if 2 reviewer commented on the same business within in specific time window (alpha=6, in days) and the same star rating, then we add them to the Graph G 4.

Twist nennen (+ Grenzwertvariablen)

Image of Groups

VII. Results/Analysis

In this section we show our plots and describe the results/interpretation.

look at arteicial groups

- wie gut finden wir unseren suspicious score - reasoning über correlation - reasoning über vermutlichen prozentualen anteil -

VIII. Conclusion

Twist nennen und sagen ob es eine gute Idee war

Finale interpretations der Ergebnisse - Was haben wir eigentlich gemacht - was war das ziel - gruppen erkennen, die fake reviews schreiben - haben wir das erreicht

Was machen wir nächstes mal anders + Empfehlungen

IX. Experimental plots

200.000 rows of the review json

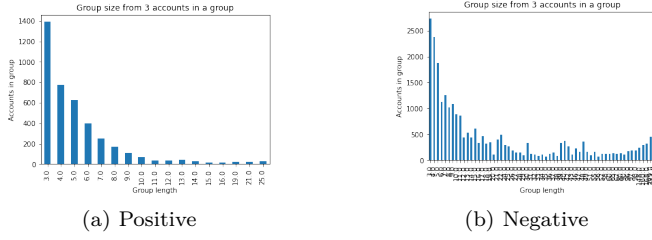


Figure 1: Groups

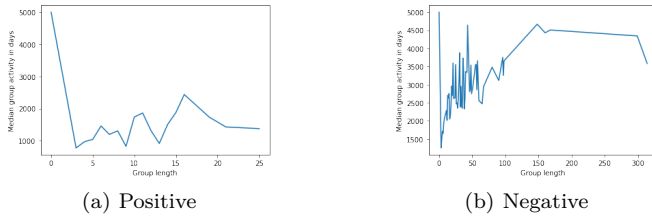


Figure 2: Activity

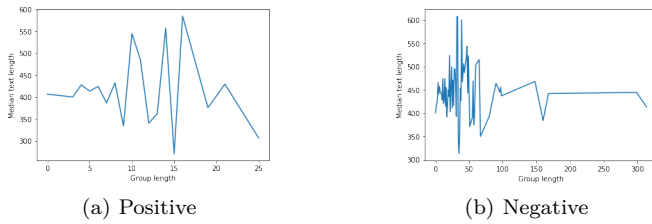


Figure 3: Length

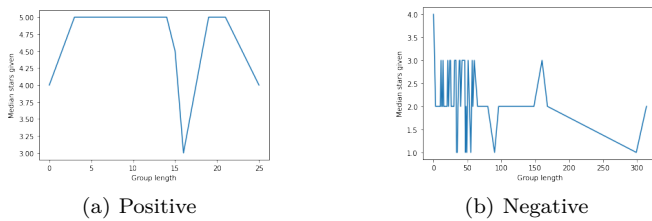


Figure 4: Stars

References

- [1] Euijin Choo, Ting Yu, and Min Chi. Detecting opinion spammer groups through community discovery and sentiment analysis. In IFIP annual conference on data and applications security and privacy, pages 170–187. Springer, 2015.
- [2] Arjun Mukherjee, Abhinav Kumar, Bing Liu, Junhui Wang, Meichun Hsu, Malu Castellanos, and Riddhiman Ghosh. Spotting opinion spammers using behavioral footprints. In Proceedings of the 19th ACM SIGKDD international conference on Knowledge discovery and data mining, pages 632–640, 2013.