Calculate the score of a social media profile and posts on twitter and other social media sites on the internet

Pooja Chaudhary^[0000-0003-4613-8921], Ravin Kumar^[0000-0002-3416-2679]

Department of Computer Science, Meerut Institute of Engineering and Technology,
meerut-250005, Uttar Pradesh, INDIA.

poojachaudhary831@gmail.com , ravin.kumar.cs.2013@miet.ac.in

Abstract. In the age of social networking, finding the worth of one's social presence and online interactions can have a significant impact on tackling the problem of fake news. We are proposing a scoring mechanism which can also be used to calculate the worth of one's online presence, and also to suggests numerical score to a particular thought posted on the internet.

Keywords: twitter, social score, influence score, worth of your post, fake news

1 Introduction

Internet has made humans highly interconnected and have fastened the process of sharing ideas, information, and thoughts like never before. This interconnected network can also be used to spread lies [1], spamming [2], sell a perception to the world, and make people believe in a specific type of ideology. In this research paper, we are proposing a method to evaluate the score of any social media interaction including scores for the social account, and posts. These scores can be used in identifying the accounts with the most probability of being spam. Although, these scores can also be found helpful in finding the most important post from the set of reviews posted online, and can also be used to judge the authenticity of the message posted is coming from actual people and not from spam or fake news spreading bots [3].

2 Proposed Mechanism

Our proposed mechanism utilizes a self-development algorithm which utilises features such as how many followers does an account has, how many other accounts does it follows, along with other features to calculate a social score. This score suggests the importance or influence ability of that account within its circle.

To better understand our proposed mechanism, let us first look at the description of the variables presented in Table 1.

Table 1 : Parameters Related Details

S.NO	Parameter	Details
1	user_id	Unique id for each person's account
2	FOLLOWERS(id),FRIENDS (id)	Return list of ids of followers and friends respectively.
3	zeros_list(Z)	Return a list of length Z containing 0 elements.
4	LENGTH(List)	Returns number of elements present in a list.

Our algorithm does require the degree_of_score to be a non-negative integer. It is used to set how deep one wants to go to calculate the score. Higher the value, more the number of relations (i.e. followers-friend, followers-followers) required to calculate the score for a “user_id”.

Function get_zeroscore(user_id)

START

return LENGTH(FOLLOWERS(user_id)) / (LENGTH(FRIENDS(user_id)) + 1)

END

Function get_multiscore(list_of_ids , degree_of_score)

START

IF degree_of_score == 1

net_score = 0

FOR element_id in list_of_ids

net_score = net_score + get_zeroscore(element_id)

net_score = net_score / LENGTH(list_of_ids)

return net_score

ELSE

degree_of_score = degree_of_score - 1

tmp_list = zeros_list(LENGTH (list_of_ids))

FOR elem_idx in RANGE(0 , LENGTH (list_of_ids))

new_list_ids = FOLLOWERS (list_of_ids [elem_idx])

tmp_list [elem_idx] = get_multiscore(new_list_ids, degree_of_score)

fin_val = 0

FOR elem in tmp_list

fin_val = fin_val + elem

fin_val = fin_val / LENGTH(tmp_list)

return fin_val

END

```

Function get_score( user_id , degree_of_score )
START
IF degree_of_score <= 0
    return get_zeroscore( user_id )
ELSE
    list_of_ids = FOLLOWERS ( user_id )
    return get_multiscore( list_of_ids , degree_of_score )
END

```

```

Function Social_Score ( user_id , degree_of_score )
START
    delta = get_zeroscore( user_id )
    score = get_score ( user_id , degree_of_score )
    return delta * score
END

```

To calculate the score for a person with id “user_id” at the degree of “degree_of_score” we simply pass these parameters to the function Social_Score.

$$\text{Score} = \text{Social\_Score}(\text{user\_id}, \text{degree\_of\_score})$$

3 Working Example

We have provided open-source code for calculating the social_score [4]. We have also tested it on the twitter platform for the better analysis of our proposed algorithm. Due to the security reasons, we have changed the usernames in the results shown in Table 2.

Higher the calculated score, higher the influence or importance of that user within its group or friend and follower circle. The proposed scoring mechanism can be also used to score a conversation by averaging the score of all the users who participated during the conversation. Similarly, the effect of the agreement or disagreement within the conversation can also be calculated using separately calculating scores for members with agreement and members with disagreement. Later, those two scores can be used to analyze the agreement and disagreement influence in that conversation.

Table 2 : Experimental Data with degree_of_score = 1

S.no	User_id	Followers	Friends	Calculated Score
1	ABC	9929	129	60.0663
2	DEF	0	1000	0
3	GHI	1000	0	11.83
4	JKL	55	21	0.96567
5	MNO	2	2	0.84313
6	PQR	4.2 Million	2228	317
7	STU	111	40	68.2
8	VWX	100	100	0.043
9	YZA	52	30	0.001
10	BCD	16000	96	185.72

One can also use parallel computing assisted with an effective scheduling mechanism [5] to simultaneously calculate the social score of large number of accounts and then analyze the obtained scores to compare the influence one a member within its friends and follower circle.

4 Conclusion

Our proposed scoring mechanism has a wide range of applicability. It can be used to provide the relevance of one's social profile. The scoring mechanism can also be used to find the importance of a post in a conversation, or even the importance of the conversation as a whole. This score can be used to easily eliminate or detect the spamming accounts, and even detect followers increasing bots.

References

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