

## EVALUATING THE MAXIMUM CONFIDENCE OF FACTS FOR INFORMATION PROVIDED ON THE WEB

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### ABSTRACT

*The World Wide Web is the important information source for us. But, there is no guarantee for the relevance of information retrieved from the Web. The information provided by one Website may be conflicting with the information on the property of an object from other website. In this paper, we study a concept, called Veracity, means, i.e. conformity to truth, which studies how to find true facts from a large amount of conflicting information of an object, provided by various websites. The TCRC (Trustworthiness and Confidence based on Ratio Contribution) algorithm, which consider the relationships between websites and their information and evaluate the confidence of facts and trustworthiness of websites from Ratio Contribution of the facts for that website.*

*Index Terms : WebMining, Graph Theory, Relevant Information.*

### 1. INTRODUCTION

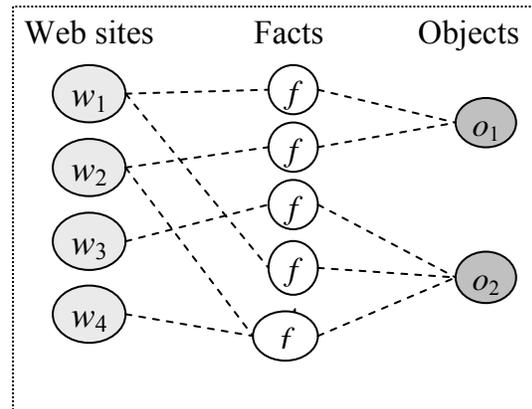
The appearance of the World Wide Web (WWW) at the end of the last century led to a rapid growth in the Internet and in the quantity of accessible information for users. The information that has accumulated on WWW represents an enormous knowledge base that may prove useful for numerous applications.

Everyday, people retrieve all kinds of information from the Web. For example, when they want to know the answer to a certain question, they go to Ask.com or Google.com.

“Is the World Wide Web always trustable?”

Example: (Height of Mount Everest). Suppose a user is interested in how high Mount Everest is and queries Ask.com with “What is the height of Mount Everest?” Among the top 20 results, he or she will find the following facts: four websites (including Ask.com itself) say 29,035 feet, five websites say 29,028 feet, one says 29,002 feet, and another one says 29,017 feet. Which answer should the user trust?

In this paper, we study a problem called the Veracity problem, and influence of one fact on other.



**Fig1: Input to TCRC**

Given the conflicting information about many objects, which is provided by multiple websites, how can we discover the true fact about each object?

Example: considering the names in the table1, resultant for the query “top ranking batsman in cricket ODI”.

The facts in table 1 are conflicting with each other, as some websites provide some players name while others provide some other players name including some names in common with other websites. Thus considering the facts and influence between the facts, trustworthiness of websites and confidence of a fact can be maximized.

**Table 1: Conflicting information about the players**

<i>websites</i>	<i>players</i>
Sheetudeep	R.T.Pointing, A.Symonds
Headlinesindia	M.S.Dhoni, M.E.K Hussey, Yuvraj Singh
Altiusdirectory	M.S.Dhoni, C.H.Gayle, M.E.K Hussey
Thatscricket.oneindia.in	M.S.Dhoni, M.E.K.Hussey Yuvraj Singh
All37	Ricky Pointing, Micheal Hussey, Gambir

**2. PROBLEM DEFINITIONS**

In this paper, we study the problem of finding true facts and trustworthiness for websites.

**2.1 Basic Definitions**

Trustworthiness and Confidence Definition1 (Confidence of facts). The confidence of a fact  $f$  is the sum of websites trustworthiness pointing towards it and is denoted by  $cf(f)$ .

Definition2 (Trustworthiness of websites). The trust worthiness of a website  $w$  is the ratio contribution of confidence of the facts which the websites point to and is denoted by  $tw(w)$ .

The Influence between the facts may exist and is indicated as, if first website indicates that the author of the book is “Jennifer Widom,” which is fact  $f_1$ . The second website says that there are two authors “Jennifer Widom and Stefano Ceri,” which is fact  $f_2$ . If  $f_2$  is correct, then  $f_1$  is incomplete and will have low confidence, and thus,  $Imp(f_2 \rightarrow f_1)$  is low. On the other hand, we know that it is very common for a website to provide only one of the authors for a book. Thus,  $f_1$  may only tell us that “Jennifer Widom” is one author of the book instead of the sole author. If we are confident about  $f_1$ , we should also be confident about  $f_2$  because  $f_2$  is consistent with  $f_1$ , and  $imp(f_1 \rightarrow f_2)$  should be high. So the value of  $imp(f_1 \rightarrow f_2)$  lies between 0 to 1 for symmetric facts and negative for conflicting information.

$$Imp(f_1 \rightarrow f_2) = sim(f_1; f_2) - base\_sim,$$

Where  $sim(f_1; f_2)$  is the similarity between  $f_1$  and  $f_2$ , and  $base\_sim$  is a threshold for similarity.

So the website providing facts may be having implication influence which may be positive or negative influence. So considering all the facts, influences and using Ratio Contribution we find a solution to solve conflict information,

**2.2 Basic Assumptions**

Assumption 1. Usually there is only one true fact for a property of an object.

Example: The captain of Indian cricket team is M.S.Dhoni.

Assumption 2. This true fact appears to be the same or similar on different websites.

Example : “Sachin Tendulkar” and “S.Tendulkar”.

Assumption 3. The false facts on different websites are less likely to be the same or similar.

Assumption 4. In a certain domain, a website that provides mostly true facts for many objects will likely provide true facts for other objects.

Example: Wiki

**3 COMPUTATIONAL MODEL**

If a fact is provided by many trustworthy websites, it is likely to be true; and the website is trustworthy if it provides facts with high confidence.

**Table 2 : Variables and parameters**

Name	Description
$tw(w)$	Trustworthiness of website $w$
$cf(f)$	Confidence of a fact $f$ .
$w$	Website
$F(w)$	Set of facts provided by $w$
$f$	Fact
$in(f)$	Influence score
$in^*(f)$	Adjusted influence score
$W(f)$	The set of websites providing $f$
$O(f)$	The object $f$ is about
$Imp(f_i \rightarrow f_j)$	Influence between the facts
$\rho$	Weight of object about same object
$d$	Max difference between two iterations
$cf^*(f)$	Adjusted confidence of fact

**3.1 Fact Confidence and Website Trustworthiness**

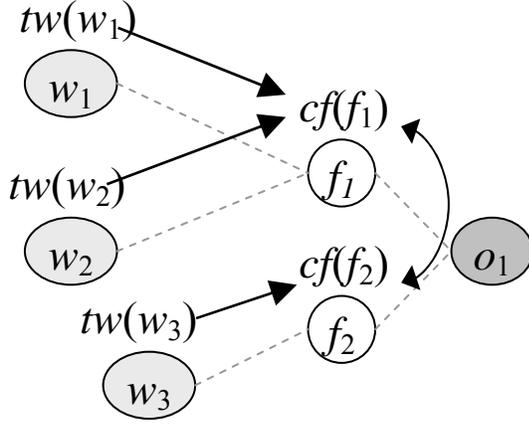
We first discuss how to infer website trustworthiness and fact confidence from each other.

**3.1.1 Basic Inference**

As defined in Definition 2, the trustworthiness of a website is the contribution of facts at instant pointed by the website. For website  $w$ , we compute its trustworthiness  $tw(w)$  by calculating the contribution of facts for that website:

$$tw(w) = \frac{\sum_{f \in F(w)} cf(f)^2}{\sum_{f \in F(w)} cf(f)} \quad (1)$$

where  $F(w)$  is the set of facts provided by  $w$ .



**Fig. 2. Computing confidence of a fact.**

The confidence of a fact can be calculated based on

$$cf(f) = \sum_{w \in W(f)} tw(w). \quad (2)$$

Where  $W(f)$  is the set of websites providing  $f$  and value of  $tw(w)$  in initial is taken as any positive value greater than one.

### 3.1.2 Influences between Facts

There are usually many different facts about an object (such as  $f_1$  and  $f_2$  in Fig. 2), and these facts influence each other.

We define the influence score of a fact as

$$in(f) = \ln(cf(f)) \quad (3)$$

Any value of  $cf(f)$  if less than 1, we discard the fact as it is having too less confidence value.

Suppose in Fig. 2 that the implication from  $f_2$  to  $f_1$  is very high (e.g., they are very similar). If  $f_2$  is provided by many trustworthy websites, then  $f_1$  is also somehow supported by these websites, and  $f_1$  should have reasonably high confidence. Therefore, we should increase the confidence score of  $f_1$  according to the confidence score of  $f_2$ , which is the sum of the trustworthiness of websites providing  $f_2$ . We define the adjusted influence score of a fact  $f$  as

$$in^*(f) = in(f) + \rho \cdot \sum_{o(f')=o(f)} in(f') \cdot imp(f' \rightarrow f) \quad (4)$$

$\rho$  is a parameter between zero and one, which controls the influence of related facts. We can see that  $in^*(f)$  is the sum of the influence scores of  $f$ , and a portion of the influence score of each related fact  $f'$  multiplies the implication from  $f'$  to  $f$ . Please notice that  $imp(f' \rightarrow f) < 0$  when  $f$  is conflicting with  $f'$ .

We use  $cf^*(f)$  to represent this confidence for  $in^*(f)$ :

$$cf^*(f) = e^{in^*(f)} \quad (5)$$

$cf^*(f)$  is the adjusted confidence of fact which is assigned back to  $cf(f)$ .

In each step of the iterative procedure, TCRC first uses the website trustworthiness to compute the fact confidence and then recomputes the website trustworthiness from the fact confidence.

TCRC stops iterating when it reaches a stable state. The stableness is measured by how much the trustworthiness of websites changes between iterations. If  $tw(w)$  vector only changes a little after an iteration (measured by cosine similarity between the old and the new  $tw(w)$  vector), then TCRC will stop.

### Algorithm:

**Input:** set of Websites, Facts and links between them.

**Output:** Trustworthiness and Confidence of websites and facts respectively.

**For each** websites

$tw(w) = x$ , // where  $x$  is default initial value;

**repeat**

$cf(f) = \sum_{w \in W(f)} tw(w)$  //confidence

**if** influence between facts exist then

**repeat**

$in(f) = \ln(cf(f))$

$in^*(f) = in(f) + \rho \cdot \sum_{o(f')=o(f)} in(f') \cdot imp(f' \rightarrow f)$

**until** (number of facts)

$cf^*(f) = e^{in^*(f)}$  //trustworthiness

**For each** fact

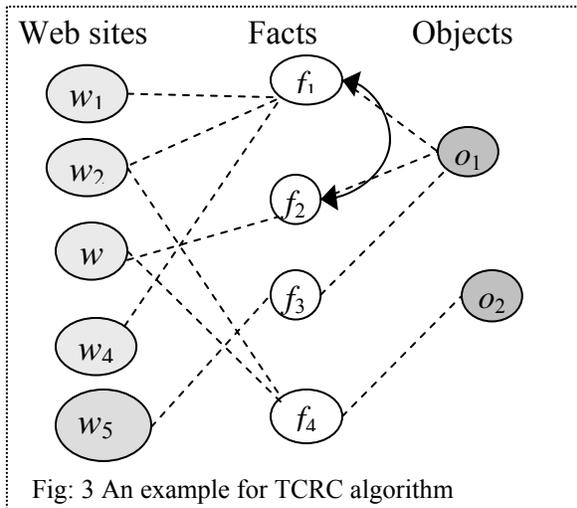
$cf(f) = cf^*(f)$  // adjusted confidence

$tw(w) = \frac{\sum_{f \in F(w)} cf(f)^2}{\sum_{f \in F(w)} cf(f)}$

**until** (cosine similarity of iterations greater than 1-d)

## 4. IMPLEMENTATION DETAILS

Let us consider an example as in fig 3, and assume the initial trustworthiness of every website equally likely to be 20, the value of  $imp(f_2 \rightarrow f_1)$  be 0.5, difference between two iterations be less than 0.05 and value of  $\rho = 0.5$ .



Then calculating the trustworthiness of websites gives the values as  $tw(w_1)=60$ ,  $tw(w_2)=52$ ,  $tw(w_3)=49.253$ ,  $tw(w_4)=60$ ,  $tw(w_5)=20$ , and confidence of facts as  $cf(f_1)=60$ ,  $cf(f_2)=55.877$ ,  $cf(f_3)=20$ ,  $cf(f_4)=40$ . where as if average is taken we may get  $tw(w_3)=47.938$ , using TCRC we can increase the trustworthiness and confidence of both websites and facts.

Example : Considering Table1, the websites.

Thatscricket.oneindia.com, Altiusdirectory.com, Headlinesindia, gives much relevant information for the facts they provide.

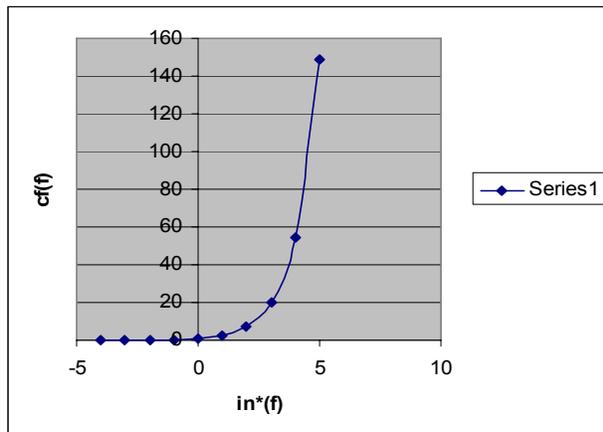


Fig4: Confidence vs. Adjusted Influence score.

For any value of influence score the confidence of a fact is greater than zero. So even for negative influence score the confidence score is positive and greater than but not zero.

## 5. CONCLUSION

In this paper, we study and formulate the Veracity problem, which aims at resolving conflicting facts from

multiple websites and finding the true facts among them. We propose TCRC, an approach that utilizes the interdependency between website trustworthiness and fact confidence to find trustable websites with ratio contribution and true facts.

## 6. REFERENCES

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