



Recommendation system for cloud service on Trust management services

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Abstract: The trust management is challenging issue in cloud computing now a days. The main obstacle for growth and adoption of cloud computing is trust. Though many works has been proposed, the determination of credibility for trust feedbacks is ignored. In this paper, the Cloud Armor, a reputation and recommendation trust management is proposed in which set of functions are provided to delivery of Trust as a Service. Cloud administration faces significant challenge. The Service Level Agreements (SLAs) are facing difficulty in finding trust between cloud service customers and cloud service providers. Buyers' input can be sensible to the cloud administration characteristic to survey.

Keywords: Trust Management, Recommendation Trust, Reputation Trust, Cloud Service, Trust Results.

I. INTRODUCTION

A formal trust management model is proposed to managing the trust and the properties for cloud environments' SaaS. The time based experience, reputation and recommendation based concept is used for computation of trust. The consumers will provide feedback to relevant service providers with different level of access with respect to data security, service level agreements and performances etc. The trust life cycle consists of three activities with the basic steps: establishment of trust, updating the trust and revocation of trust. It consists of two entities: trust or (i.e., Service provider), trustee (i.e., Customer) and trust model. There two kind of trust: Direct trust and indirect trust. Direct trust is based on self experience and if they there is no direct interaction between two entities then trust is based on recommendation of other entities. Defining the concept of trust is one of the challenging tasks as it consists of many applications causing to divergence in terminology. Trust in simple words can be defined as "Generally, it is said that an entity trusts second entity when the second entity withstands with expectations of first entity".

Trust model is built on certain trust properties and they are as follows:

- Asymmetry: A trust relation is a symmetric a sit does not mean that Y must trust X as X trust Y.
- Reflexivity: the trust is considered to reflexive as each entity believes it self.
- Context Dependence: trust relation will not be generalized to other a sit regarded precise object.
- Scalability: Trust is also scalable as it may change while the period of communication. This change indicates the modification in the trust level which also indicates that change in entities reputation.
- Partial Transitivity: Trust follows transitivity property. Z get recommendation from Y to X only if Y is trust worthy to X and Z is trustworthy to Y. Otherwise the property is said not to be transitive.
- Subjective: Trust is also said to be level of subjective probability.
- Uncertainty: the trust's most important characteristic is uncertainty which indicates the relationship of trust between fuzzy and stochastic entities of strangers.

Time based: time based variant property is satisfied by direct trust.

Consumes usually do the verification of service providers' trust level of before accessing data and services from the cloud. Hence the trust model guarantees cloud service providers' trust worthiness then allow the consumer to access the services and data. The main proposition of the work is trust management modeling for recommendation trust based on space variant evaluation and direct trust based on time variant evaluation. Components of the trust model:



Trust:

Trust Degree:

Trust represents the trust degree which is evaluated using possible trust values from domain set trust or T_i views trustee T_j as and it is expressed as

$$Td_{ij}(T_i, T_j, S_k, t) \text{ where } i \neq j; \quad 0 \leq Td_{ij}(T_i, T_j, S_k, t) \leq 1 \tag{1}$$

Where k th service is represented as S_k and time defined by t . Trust degree value lies between 0 and 1. Direct trust represented as T_d divorce commendation trust represented evaluate the trust degree. In the case of new entity joining any cloud environment newly then ignorance value denoted by assigned.

$$\exists Td_{ij}(T_i, T_j, S_k, t) \tag{2}$$

$$Td_{ij}(T_i, T_j, S_k, t) \rightarrow \tag{3}$$

$$Td^{dir}(T_i, T_j, S_k, t) \oplus Td^{recom}(T_i, T_j, S_k, t) \oplus Td^{iv}(T_i, T_j, S_k, t) \tag{4}$$

$$Td_{ij}(T_i, T_j, S_k, t) = \begin{cases} 0, & \text{if } nd=0, nr=0 \\ Dt(T_i, T_j, S_k, t), & \text{if } nd \in \{1, 2, \dots\}, nr=0 \\ R(T_i, T_j, S_k, t), & \text{if } nr \in \{1, 2, \dots\}, nd=0 \end{cases} \tag{5}$$

The direct trust degree is represented by $Dt(T_i, T_j, S_k, t)$ and there commendation trust degree is represented by $Rt(T_i, T_j, S_k, t)$ of trust or T_i in the view of trustee T_j regarding the k th service S_k at time t . The values of nd and nr gives the direct trust degree number and nr gives there commendation trust degree number,

TrustRelation:

TABLE I. Satisfactory Level

Level	Label	Trust worthiness
1	No Opinion	$Td=0$
2	Low Distrust	$0 < Td < 0.5$
3	Medium Trust	$Td=0.5$
4	High Trust	$0.5 < Td < 1$
5	Complete Trust	$Td=1$



Trust chain In cloud computing system ,the trust chain is evaluated based on the partial transitive properties If

$$\exists T_p, T_q, T_r, T_s, T_i \in Q \tag{6}$$

and

$$Tr_{pq}(T_p, T_q), Tr_{qr}(T_q, T_r) \tag{7}$$

Direct trust:

Direct trust is the relationship of trust between the entities having direct communication. In this module, direct trust table is used to maintain the trust of each entity with other entity. The direct trust measures the level of subjective probability with respect to trustee T_j and trust or T_i based on satisfaction level for the service S_k at time of interaction t . The measurement of experience of direct interaction gives the trust degree.

Time based experience:

Direct trust always decays with time. The trust acquired by an entity at time t might get changed as trust attribute at the time $t + \Delta t$.

$$D(T_i, T_j, S_k, t + \Delta t) < D_i(T_i, T_j, S_k, t) \tag{8}$$

The time t_c represents current time and t_l represents last interaction time. The definition of decay function γ is given

$$\gamma(t_c, t_l) = e^{-(\Delta t)K} = e^{-(t_c - t_l)K} \tag{9}$$

Where

$$K \in \{1, 2, 3, \dots\}, \quad \gamma(t_c, t_l) \in [0, 1]$$

K helps in determining the decay rate of degree of direct trust with time Δt . If $R F_i$ is the trust or e 's reputation factor, then direct trust degree is calculated at current time t_c using

$$Dt_{t_c}(T_i, T_j, S_k, t_c) = \frac{\sum_{i=1}^N \gamma(t_c, t_l) * Dt_l(T_i, T_j, S_k, t_l) + R}{\sum_{i=1}^N \gamma(t_c, t_l)} \tag{10}$$

Recommended trust:

The two entities do not have direct interactions, and then their trust relation is based on recommendation by other entity. Recommended trust is used to measure the recommenders' subjective probability set regarding the trustee T_j to the trust or T_i for the service S_k by single or more trust chains and represented by

$$R(T_i, T_j, S_k, t) = \frac{\sum_{m=1}^{length} W_{m,m+1}^{-1} * Dt_{\gamma}(T_m, T_{m+1}, S_k, t)}{\sum_{m=1}^{length} W_{m,m+1}^{-1}} \tag{10}$$

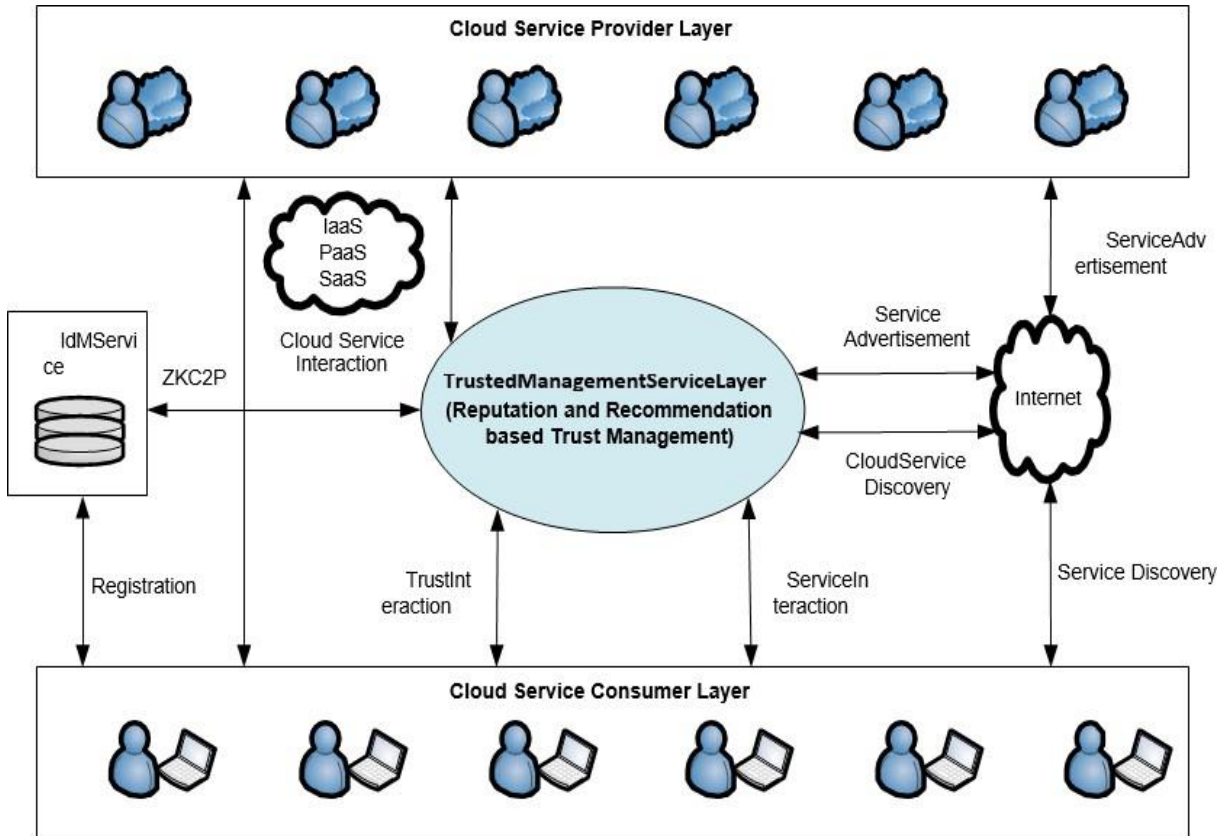


FIGURE1. System Architecture of Proposed Approach

II. METHODOLOGY

The figure shows the architecture of proposed system. The main layers of the proposed work are cloud service consumer, cloud service providers and trust management and are used for Trust as a Service (TaaS). Cloud Service Provider layer consists of cloud service providers offering one or more cloud services i.e., SaaS (Software as a Service), PaaS (Platform as a Service), and IaaS (Infrastructure as a Service) on the Web publicly. Trust Management consists of Trust manager and cloud services. The interfaces are exposed by the trust manager by which user can provide feedback and enquire about the trust result of cloud services. Interaction for cloud service layer includes the discovery to allow users for trust assessment through internet. The trust manager publishes the users' feedback and trust of cloud service. Cloud service consumer layer has variety of users using cloud services. The interactions involved in this are service discovery by users, service and trust interactions for providing feedback and obtaining the trust results of service provider by users. Identity management helps TM for detecting the Sybil attacks. First time users using TM need to register their credentials in IDM to identity to be established. The identity record is maintained for each user. This record consists of user name, credentials attributes like passwords and postal address at the time of registration. Trust manger will assess the trust of requested service provider by the user and gets feedback from user. The trust behavior of cloud service is computed by using users' feedback. This requires user identity, cloud service identity and QoS feedbacks.

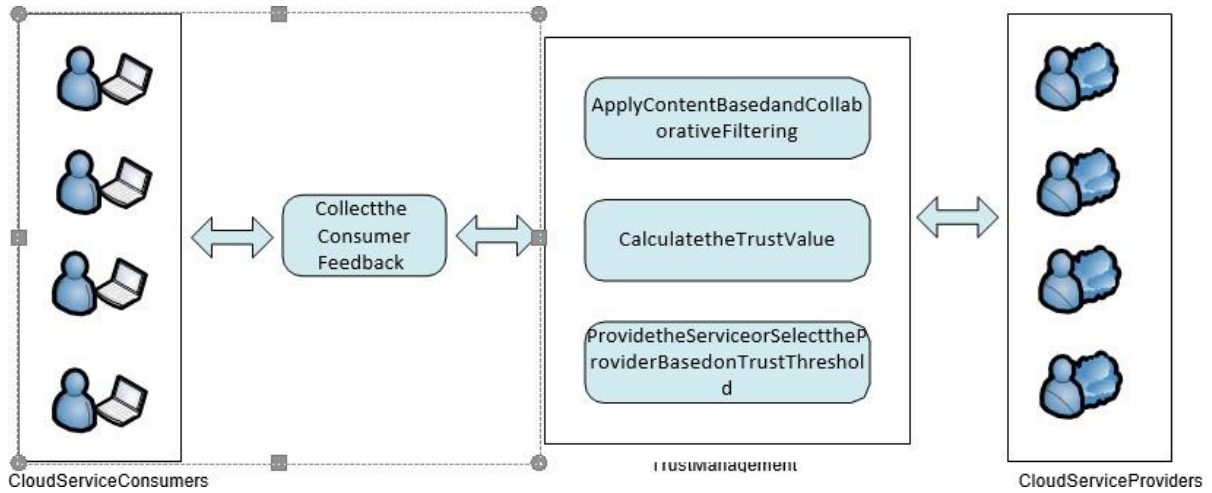


FIGURE2. System Architecture for Recommendation based Trust Management Approach

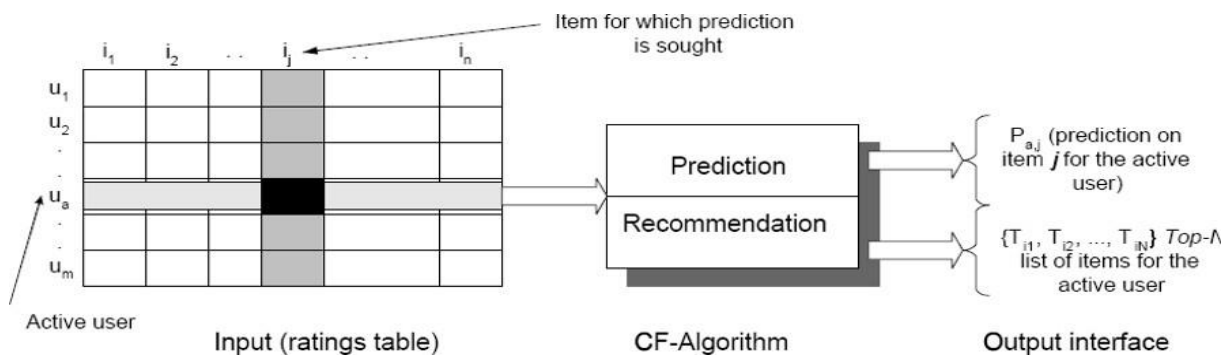


FIGURE3. The Collaborative Filtering Process

The similarity rating of the user is computed as

$$r = \frac{A-B}{\sqrt{\text{var}(x)\text{var}(x_{i-1})}} \tag{13}$$

In the above equations, x_i represents the rating of i^{th} service provider.

$$\beta = \frac{OF'_s}{OR'_s} \tag{11}$$

The similarity rating of the user is computed as

$$r = \frac{A-B}{\sqrt{\text{var}(x)\text{var}(x_{i-1})}} \tag{15}$$

x_i represents the rating of i^{th} service provider. If users' similarity rating is found to be greater than the mean of



$$\alpha = \frac{OF_s}{OR_s} \tag{12}$$

The ratio of failure requested service's from i^{th} user to overall services requested by i^{th} user is the requested service failure ratio. In similar fashion, the ratio of number of other requested service failures by i^{th} user to overall number of other services than the requested by i^{th} user is

$$\beta = \frac{OF'_s}{OR'_s} \tag{13}$$

If users' trust value is found to be higher than threshold then he is assigned to cloud service, if not the user gets terminated from procedure.

Trust Computation

At the initial step, history of services and user requests are collected. Then recommendation system based filter is applied to evaluate the users' similarity rating. The following equations are used for calculation:

$$A = \sum_{i=0}^n x_{i-1} \tag{14}$$

$$B = \sum_{i=0}^n x_i \tag{15}$$

$$va(x_i) = \frac{1}{n} \sum_{i=0}^n x_i^2 - \left(\frac{1}{n} \sum_{i=0}^n x_i \right)^2 \tag{16}$$

$$va(x_{i-1}) = \frac{1}{n} \sum_{i=0}^n x_{i-1}^2 - \left(\frac{1}{n} \sum_{i=0}^n x_{i-1} \right)^2 \tag{17}$$

The similarity rating of the user is computed as

$$U = \frac{A-B}{\sqrt{va(x)var(x_{i-1})}} \tag{22}$$

In the above equations x_i represents the rating of i^{th} service provider.

If users similarity rating is found to be greater than theme an of similarity rating, jump to next step otherwise terminate the user.

The failure ratio of requested service and others are calculated from below given equations

$$\alpha = \frac{OF_s}{OR_s} \tag{18}$$

The ratio of failure requested service's from i^{th} user to overall services requested by i^{th} user is the requested service failure ratio. In similar fashion, the ratio of number of other requested service failures by i^{th} user to overall number of other services than the requested by i^{th} user is



III. RESULTS

We mainly focus on trust value and credibility model to select the service provider and user. The two kinds of malicious behavior considered are Collusion and Sybil attacks. For proposed experiment, the data collected consists of five different cloud service and fifty feedbacks. The collusion attack and Sybil attack were considered and were resolved at time period 60 minutes. The trust result of all users' feedback and trust result of friends' feedback varies. The result is shown in below figure.

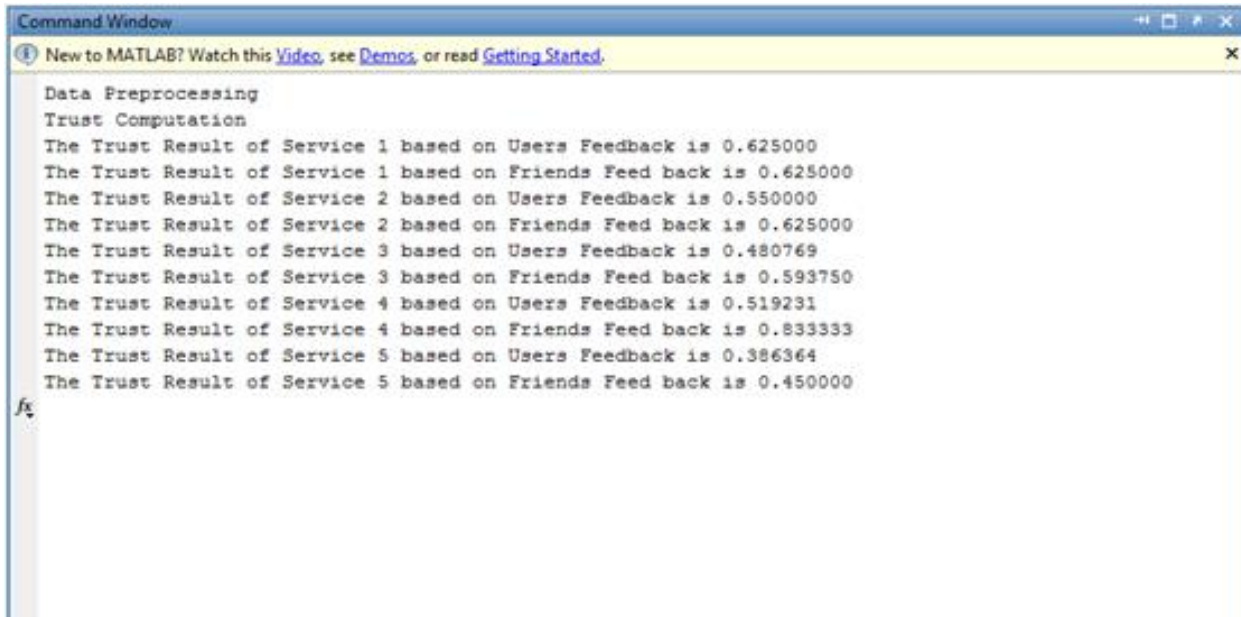


FIGURE4.TrustResultofServiceProviders

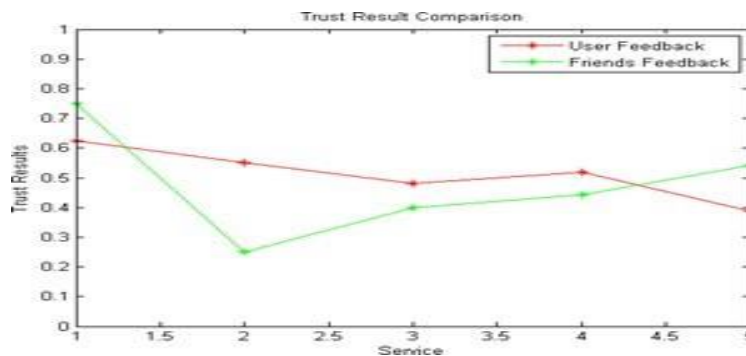


FIGURE5.Reputation and Recommendation based Trust Results

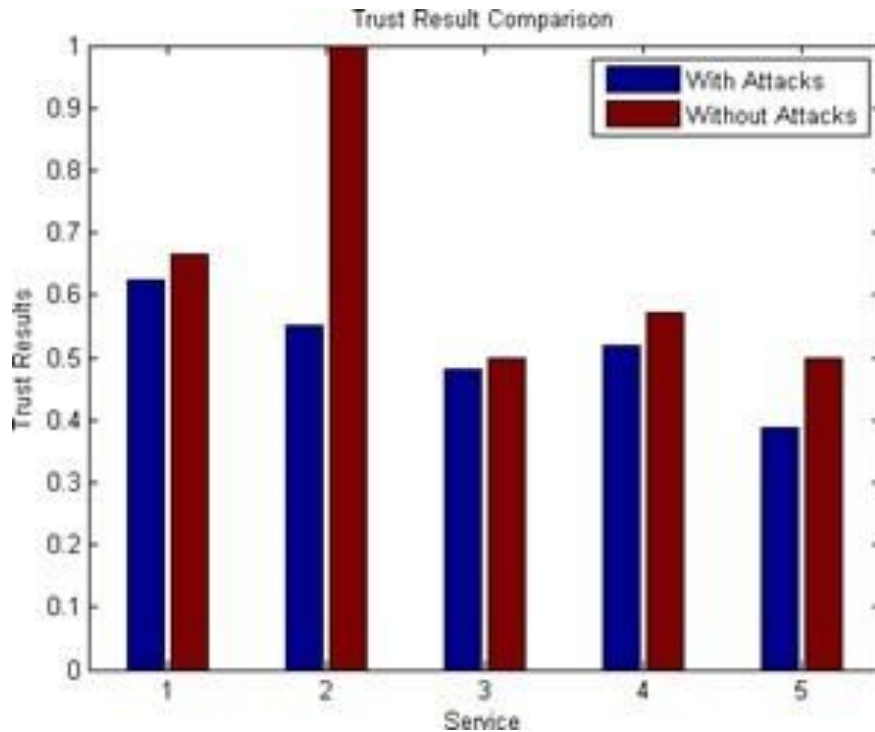


FIGURE6.TrustResultComparison

The result comparison of reputation and recommendation based trust result for service providers is shown in figure 5. The trust results in presence of attacks and in the absence of attacks in the figure 6. The analysis of trust management result of service provider is shown in figure 7.

Then the trust assessment for the user is done and result is given to the service providers and is as shown below in figure. The graph in figure shows the credibility of users. The user behavior in the form of trust value and credibility is shown in figure 9.



FIGURE7.Reputation and Recommendation based Result



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Command Window
New to MATLAB? Watch this Video, see Demos, or read Getting Started.
The Trust Result of Service 4 based on Friends Feed back is 0.833333
The Trust Result of Service 5 based on Users Feedback is 0.386364
The Trust Result of Service 5 based on Friends Feed back is 0.450000
Trust Management
The Service 1 has Trust Result 0.750000 and is suggested by Dariusz Mikulski.
The Service 2 has Trust Result 1.000000 and is suggested by Joyce Gilead-Andrew.
The Service 3 has Trust Result 0.750000 and is suggested by Erik Dodd.
The Service 4 has Trust Result 0.750000 and is suggested by Joyce Gilead-Andrew.
The Service 5 has Trust Result 0.750000 and is suggested by Dariusz Mikulski.
Server Side TMS
The customer with ID 1 is reliable with Trust Factor 0.540727 and Credibility 0.555556
The customer with ID 2 is not reliable as Trust Factor 0.411715 and Credibility 0.571429
The customer with ID 3 is reliable with Trust Factor 0.686049 and Credibility 1.000000
The customer with ID 4 is reliable with Trust Factor 0.617241 and Credibility 0.666667
The customer with ID 5 is not reliable as Trust Factor 0.388063 and Credibility 0.500000
The customer with ID 6 is not reliable as Trust Factor 0.416523 and Credibility 0.400000
The customer with ID 7 is reliable with Trust Factor 0.605465 and Credibility 0.750000
The customer with ID 8 is not reliable as Trust Factor 0.463096 and Credibility 0.250000
The customer with ID 9 is not reliable as Trust Factor 0.434669 and Credibility 0.500000
The customer with ID 10 is not reliable as Trust Factor 0.363330 and Credibility 0.250000
    
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FIGURE8.Users' Trust Result

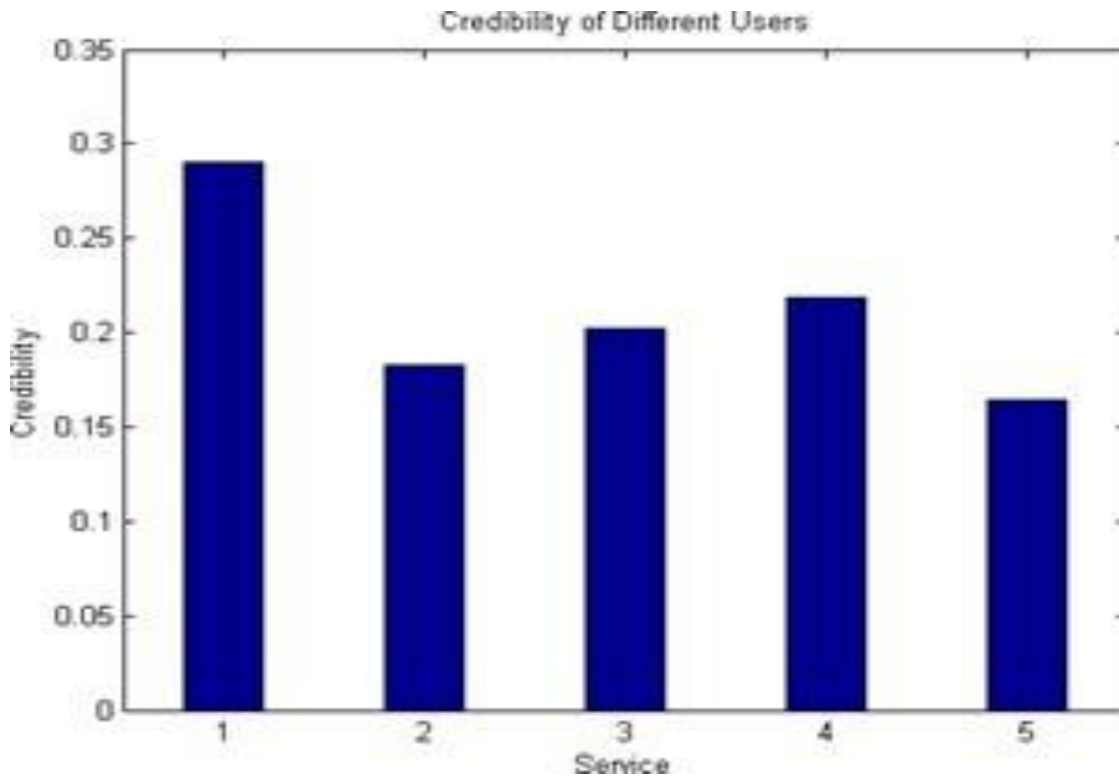


FIGURE9.Credibility of Different Users

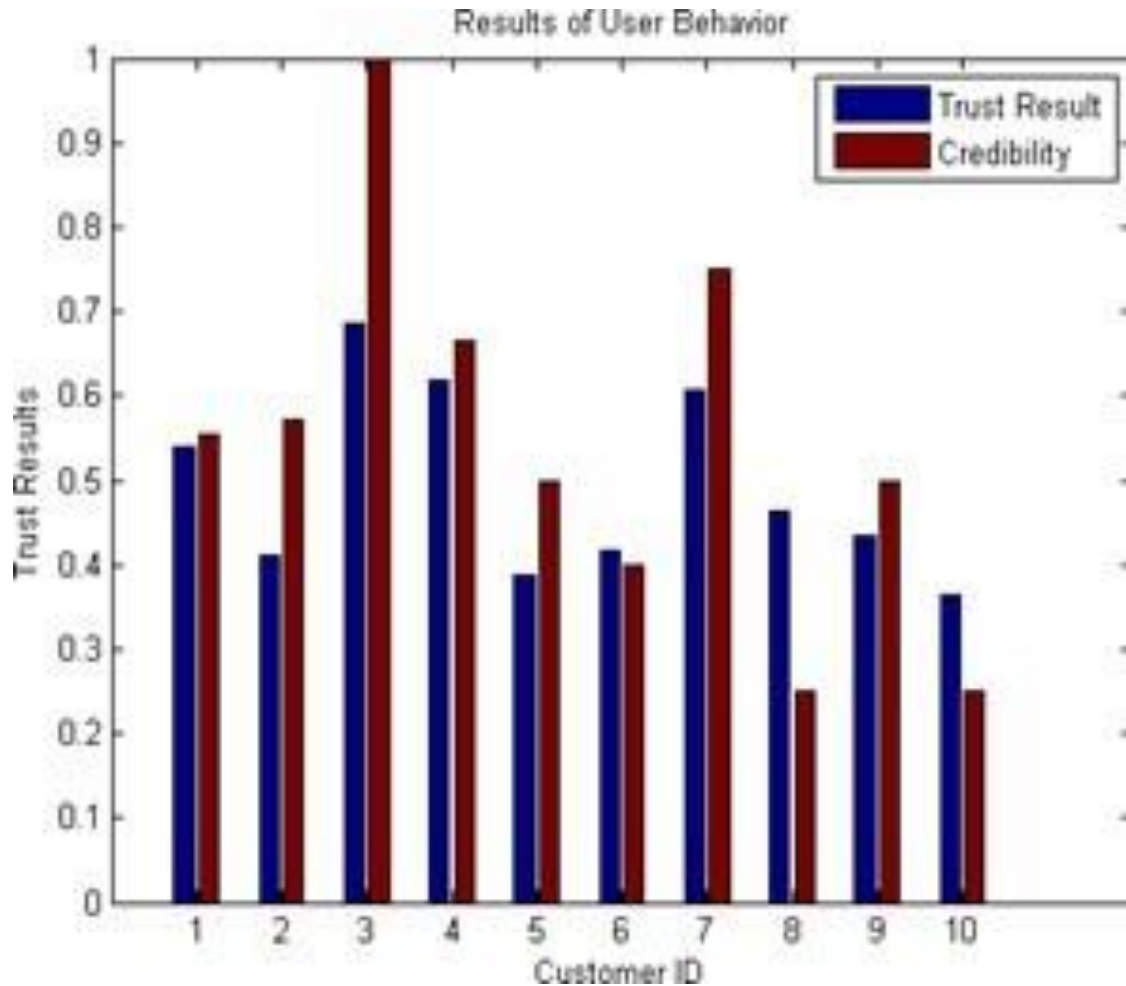


FIGURE10. User Behavior Result

IV. CONCLUSION

From various journal studies, it is concluded that trust-aware recommendation systems improve the accuracy, reliability, and transparency of cloud service selection. These systems help users identify dependable cloud providers while reducing risks associated with malicious feedback, fake ratings, and uncertain service performance. Techniques such as collaborative filtering, fuzzy logic, machine learning, block chain integration, and reputation-based models are widely used to enhance trust evaluation mechanisms.

Furthermore, trust management frameworks contribute to better decision-making by dynamically updating trust values based on real-time service performance and user experiences. This increases user satisfaction and strengthens confidence in cloud environments. Researchers also highlight that combining trust metrics with QoS parameters provides more effective and personalized cloud service recommendations compared to traditional methods.



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