

Consumers' Privacy. The adoption of cloud computing raise privacy concerns .Consumers can have dynamic interactions with cloud providers, which may involve sensitive information. There are several cases of privacy breaches such as leaks of sensitive information (e.g., date of birth and address) or behavioral information (e.g., with whom the consumer interacted, the kind of cloud services the consumer showed interest, etc.) . Undoubtedly, services which involve consumers' data (e.g., interaction histories) should preserve their privacy. Cloud Services Protection. It is not unusual that a cloud service experiences attacks from its users. Attackers can disadvantage a cloud service by giving multiple misleading feedbacks (i.e., collusion attacks) or by creating several accounts (i.e., Sybil attacks). Indeed, the detection of such malicious behaviors poses several challenges. Firstly, new users join the cloud environment and old users leave around the clock. This consumer dynamism makes the detection of malicious behaviors (e.g., feedback collusion) a significant challenge. Secondly, users may have multiple accounts for a particular cloud service, which makes it difficult to detect Sybil attacks. Finally, it is difficult to predict when malicious behaviors occur (i.e., strategic VS. occasional behaviors). Trust Management Service's Availability. A trust management service (TMS) provides an interface between users and cloud services for

effective trust management. However, guaranteeing the availability of TMS is a difficult problem due to the unpredictable number of users and the highly dynamic nature of the cloud environment. Approaches that require understanding of users interests and capabilities through similarity measurements or operational availability measurements (i.e., uptime to the total time) are inappropriate in cloud environments. TMS should be adaptive and highly scalable to be functional in cloud environments.

2.THE FRAMEWORK

We propose a framework using the Service Oriented Architecture (SOA) to deliver trust as a service. SOA and Web services are one of the most important enabling technologies for cloud computing in the sense that resources (e.g., soft-ware, infrastructures, and platforms) are exposed in clouds as services [6,16]. In particular, our framework uses Web services to span several distributed TMS nodes that expose interfaces so that trust participants (i.e., the cloud service consumers) can give their trust feedbacks or inquire about the trust results based on SOAP or REST [15] messages. Figure 1 depicts the framework, which consists of three different layers, namely the Cloud Service Provider Layer,

the TrustManagement Service Layer, and the Cloud Service Consumer Layer. The Cloud Service Provider Layer. This layer consists of different cloud service providers who provide cloud services. The minimum indicative feature that every cloud service provider should have is to provide the infrastructure as a service (i.e., the cloud provider should have a data center that provides the storage, the process, and the communication).

feedbacks or inquire about the trust results represents. Our framework also contains a Registry Service (see Figure 1) that has several responsibilities including i) Service Advertisement: both cloud service providers and the TMS are able to advertise their services through the Service Registry; ii) Service Discovery: the TMS and cloud service consumers are able to access the Service Registry to discover services.

A. TRUST FEEDBACK COLLECTION AND ASSESSMENT

In our framework, the cloud service trust behavior is represented by a collection of invocation history records denoted as H . Each cloud service consumer c holds her point of view regarding the trustworthiness of a specific cloud service s which is managed by the assigned TMS. H is represented in a tuple that consists of the cloud consumer primary identity C , the cloud service identity S , a set of trust feedbacks F and the aggregated trust feedbacks weighted by the credibility F_c , i.e., $H = (C, S, F, F_c)$. Each trust feedback in F is represented in numerical form in which the range of the normalized feedback is $[0, 1]$, where 0, +1, and 0.5 means negative, positive, and neutral respectively. Whenever a cloud consumer inquires the TMS about the trustworthiness of a cloud service s , the trust result ($Tr(s)$), is calculated using: where $V(s)$ is all trust feedbacks given to the cloud service s and $|V(s)|$ represents the length of the $V(s)$. $F_c(l, s)$ are trust feedbacks from the l^{th} cloud consumer weighted by the credibility.

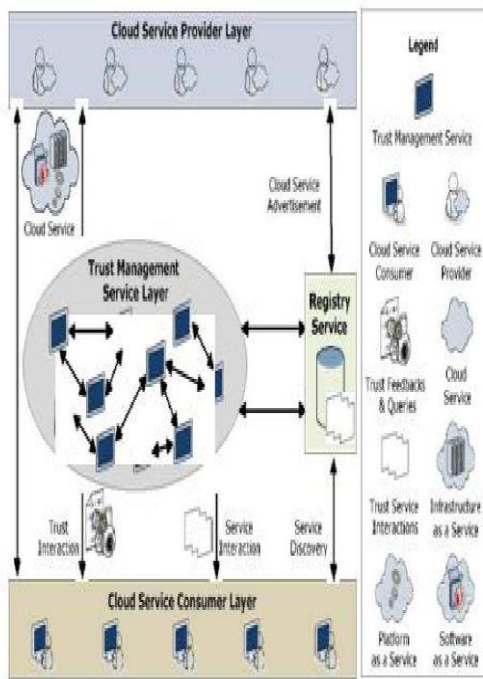


Fig. 1. Architecture of the Trust as a Service Framework

– The Trust Management Service Layer. This layer consists of several distributed TMS nodes that expose interfaces so that cloud service consumers can give their trust

The TMS distinguishes between credible trust feedbacks and malicious trust feedbacks through assigning the Cloud Consumer's Experience aggregated weights $Exp(l)$ to trust feedbacks $F(l, s)$ as shown in Equation 2, where the result $F_c(l, s)$ is held in the invocation history record h and updated in the assigned TMS.

$$F_c(l, s) = F(l, s) * Exp(l) \quad (2)$$

3. CREDIBILITY MODEL

There is a considerable possibility that the TMS receives inaccurate or even malicious trust feedbacks from amateur cloud service consumers (e.g., who lack experience) or vicious cloud service consumers (e.g., who submit lots of negative feedbacks to disadvantage a particular cloud service). To overcome these issues, we propose a credibility model, which is centered on the cloud consumer's experience. To differentiate between expert and amateur cloud service consumers, we consider the Majority Consensus and the Cloud Consumer's Capability.

Majority Consensus. It is well-known that the majority of people usually agree with experts' judgments about what is good [4]. Similarly, we believe that the majority of cloud consumers agree with Expert cloud service consumers' judgments. In other words, any cloud service consumer whose trust feedback is close to the majority of

trust feedbacks is considered an Expert Cloud Service Consumer (ECSC), or an Amateur Cloud Service Consumer (ACSC) otherwise. In order to measure how close the cloud service consumer's trust feedbacks to the majority (i.e., the Majority Consensus (J(c))) which is calculated as follows: the numerator represents the mean of the majority trust feedbacks given by other cloud service consumers ($F(l, k)$) (i.e., the l^{th} cloud service consumer, except the cloud service consumer c) to the k^{th} cloud service.

Cloud Service Consumer's Capability. It is a common sense that older people are likely to be more experienced in judging things than younger people [14]. However, this is only true if the older people have experienced considerable number of judging practices. As a result, we believe that "older" cloud service consumers who have many judging practices are likely to be more experienced and capable. A cloud service consumer's capability (B) is measured as follows: where $V_c(c)$ represents all good feedbacks (i.e., feedbacks which are close to the majority) given by the cloud service consumer c . $Ag(c)$ denotes the virtual Age of a certain cloud service consumer, measured in days since the registration in the TMS. The idea behind adding the number 1 to this ratio is to increase the value of a cloud service consumer experience based on B(c) result. In other words, we use B(c) as a reward factor. The higher B(c) is, the more experienced a cloud service consumer is. It should be noted that even if a malicious cloud service consumer attempts

to manipulate the capability result, the capability result will not exceed 2.

Based on the specified cloud service consumer's experience factors (i.e., $B(c)$ and $J(c)$), the TMS distinguishes between ECSC and ACSC through assigning the cloud service consumer's Experience aggregated weights $Exp(c)$ to each of the cloud consumers' trust feedbacks as shown in Equation 2. $Exp(c)$ is calculated as follows:

$$Exp(c) = \frac{\beta * B(c) + \mu * J(c)}{\Lambda} \quad (5)$$

where β and $B(c)$ denote the cloud service consumer's Capability factor's normal-ized weight and the factor's value respectively. The second part of the equation represents the Majority Consensus factor where μ denotes the factor's normal-ized weight and $J(c)$ denotes the factor's value. λ represents the number of factors used to calculate $Exp(c)$ (e.g., if we only consider cloud service consumer's capability, $\lambda = 1$; if we consider both cloud service consumer's capability and majority consensus, $\lambda = 2$). We use $J(c)$ as a penalty factor (i.e., because $J(c)$ ranges $[0,1]$ as described in equation 3). The lower $J(c)$ is, the lower the experience of the cloud service consumer c is. However, $B(c)$ is used as a reward factor (i.e., because $B(c)$ ranges $[1, 2]$ as described in equation 4). Higher $B(c)$ means more experienced of a cloud service consumer. It is worth mentioning that our credibility is dynamic and is able to detect behavior

changes. For example, if a cloud service consumer behaves good for a period of time (e.g., to gain credibility) and then starts misbehaving, $J(c)$ can detect such behavior through applying the standard deviation.

4.DISCUSSIONS

Trust management is one of the critical issues in cloud computing and a very active research area [10,12,8,2]. For instance, Hwang et al. [8] proposed a security-aware cloud architecture where trust negotiation and data coloring techniques are used to support the cloud service provider perspective. The cloud service consumer's perspective is supported using the trust-overlay networks to deploy a reputation-based trust management. Brandic et al. [2] proposed a central-ized approach for compliance management in cloud environments that supports the cloud service consumer's perspective using compliant management to help the cloud service consumers in selecting proper cloud services. Unlike previous works that use centralized architecture, we present a credibility model support-ing distributed trust feedback assessment and storage. This credibility model also distinguishes between trustworthy and malicious trust feedbacks. the feedback [18]. However, this ap-proach is inappropriate in cloud environments because peers give and receive services and they are evaluated on that base. In other words trust results are used to distinguish between credible and malicious feedbacks.

5.Conclusion

Given the highly dynamic, distributed, and nontransparent nature of cloud services, managing and establishing trust between cloud service users and cloud services remains a significant challenge. Cloud service users' feedback is a good source to assess the overall trustworthiness of cloud services. However, malicious users may collaborate together to i) disadvantage a cloud service by giving multiple misleading trust feedbacks (i.e., collusion attacks) or ii) trick users into trusting cloud services that are not trustworthy by creating several accounts and giving misleading trust feedbacks (i.e., Sybil attacks). In this paper, we have presented novel techniques that help in detecting reputationbased attacks and allowing users to effectively identify trustworthy cloud services. In particular, we introduce a credibility model that not only identifies misleading trust feedbacks from collusion attacks but also detects Sybil attacks no matter these attacks take place in a long or short period of time (i.e., strategic or occasional attacks respectively). We also develop an availability model that maintains the trust management service at a desired level. We have collected a large number of consumer's trust feedbacks given on real-world cloud services (i.e., over 10,000 records) to evaluate our proposed techniques. The experimental results demonstrate the applicability of our approach and show the capability of detecting such malicious behaviors. There are a few directions for our future work. We plan to combine different trust management techniques such as reputation and recommendation to increase the trust results

accuracy. Performance optimization of the trust management service is another focus of our future research work.

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VIMALA THULLURI received B.Tech in Information Technology from Malineni Lakshmaiah Engineering College affiliated to the Jawaharlal Nehru technological university Kakinada in 2014, and pursuing M. Tech in Computer Science and Engineering from Malineni Lakshmaiah Engineering College affiliated to the Jawaharlal Nehru technological university Kakinada in



SIVA KRISHNA is working as asst professor in CSE department in Malineni Lakshmaiah Engineering College, Singarayakonda, Prakasam(Dt), AP,