

Non-Redundant Indirect Trust Search Algorithm Based on a Cross-Domain Trust Model in Content Delivery Network

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Abstract— A content delivery network (CDN), as a distributed network architecture, enhances efficient delivery of content. The interconnection of different CDNs (CDNi) further improves efficiency and the experience of end users. As another distributed network with high availability and high performance, a peer-to-peer (P2P) network can provide efficient resource sharing. To combine the advantages of the two networks, we propose a hybrid CDNi-P2P architecture, along with trust management models to achieve more efficient content delivery. In CDNi-P2P architecture, end users can obtain the requested content from the nearest CDN edge server, and can also share these contents with other users in the same domain as a P2P network. After the transactions, users can rate each other based on the reputation evaluation method adopted in the system. For some mobile users, they can move among different domains and share the contents who have with the end users in different system. In general, different systems adopt different reputation evaluation standards. This leads to disparate trust values for mobile users in different systems. Based on the architecture, we propose two trust models to solve this problem: a local trust model and a cross-domain trust model. To evaluate reputation more effectively and accurately, we also propose a search algorithm for the trust model called the non-redundant indirect trust search algorithm (NRIT-SA). Using the proposed trust models, a mobile user can transform his/her local trust into mobile trust in a new domain. We thus avoid disparate trust values for a single user in different domains and improve the availability of the content possessed by mobile users as they move among different domains.

Keywords— CDNi, P2P, Trust, Reputation Evaluation, Mobile, Cross Domain

I. INTRODUCTION

Two major technologies provide large-scale video streaming over the Internet: content delivery networks (CDNs) and peer-to-peer (P2P) networks. Both of them can distribute content with high availability and high performance. With CDNs, an origin server distributes content to cache servers (edge servers) located close to end users, resulting in fast, reliable applications and Web services for users [1]. There are many commercial CDN companies: Akamai, AT&T, NTT Communication, Limelight, Mirror Image, Level 3, etc. In practice, it is extremely expensive to deploy and maintain CDN servers, so CDN architectures do not benefit from high scalability. On the other hand, P2P networks can be highly scalable because of

their low start-up costs and because they rely on peers instead of dedicated and expensive servers. The complementary advantages of CDN and P2P networks allow their combination into a hybrid CDN-P2P architecture that creates a distribution system with higher scalability and reliability than either kind of network alone [2], [3].

Content delivery network interconnection (CDNi) is a new interactive network infrastructure that allows information and content to be transmitted between different CDNs through specific interfaces. CDNi provides all of the benefits of CDN and also has some unique characteristics. In CDNi, end users do not need to register at all CDN providers to obtain content from different content service providers (CSPs). When requested content is not cached in any edge server of the registered CDN, the end-user's request will be redirected to other CDNs to capture the content through the interfaces among them. (CDNi requires the specification of interfaces and mechanisms to address issues such as request routing, distribution metadata exchange, and log in information exchange across CDNs [4].) As a result, content can be delivered via a CDN chain and transmitted to end users by the closest CDN. CDNi thus uses two categories of CDNs: the one that caches content from a CSP is called the upstream CDN (uCDN), and the one that delivers content directly to an end user is called the downstream CDN (dCDN).

To deliver content more efficiently, we combine CDNi and P2P architectures into a hybrid CDNi-P2P network that combines the advantages of both. Based on the proposed hybrid CDNi-P2P architecture, end-users could receive content from the closest edger server of the dCDN with which they are registered and from peers in the same domain.

In a hybrid CDNi-P2P network, all the end users, as peers, are both consumers and providers. When an end-user requests content from a peer, some users might be honest and provide accurate content received from the edge server of a dCDN, others might be self-serving and unwilling to provide content to peers, and still others might be malicious and provide false or harmful content [5]. A trust model provides a way to generate trust based on a peer's history of behavior [6]-[10]. A larger trust value indicates a higher probability of providing accurate content. Meanwhile, some mobile peers move among different systems or domains. From a reputation-estimation point of view,

they need to develop a trust value in each P2P system. In general, different P2P systems and domains adopt different trust models and reputation evaluation standards that lead to disparate trust values for a single peer in different domains, even if that user always has the same performance. Most research on existing reputation systems focuses on the trust model of a single system or domain. Creating a cross-domain trust model has never been considered. Therefore, in this paper, we propose a cross-domain trust model for a hybrid CDNi-P2P network.

The rest of this paper is organized as follows: we introduce our local trust model in Section 2 and propose our hybrid CDNi-P2P architecture and cross-domain trust model in Section 3. Section 4 concludes this paper.

II. PROPOSED LOCAL TRUST MODEL

A cross-domain trust model presupposes a relative reputation evaluation for each participant in a local system, which can indicate the reliability of each individual participant. In this section, we propose a local trust model to generate a relative reputation value for each participant according to the different reputation evaluation standards currently used by local systems.

A. Trust Value and Trust Degree

Existing online reputation systems and research use two main approaches to evaluate a participant's reputation within a specific network. In the first approach, the ratings for both the service receiver and provider are given via a bi-directional or one-directional rating after each transaction. The rating could take the form of reputation scores, feedback ratings, positive feedback rates, etc. The overall reputation of a participant is the sum of those ratings and is called that user's *trust value*, denoted by v . Online reputation systems that use this approach are the online auction system eBay, Amazon, and Taobao. Generally, the trust value is an integer equal to or greater than zero that is public to all system participants; other participants can decide whether to trust a participant based on this trust value. In the second approach, both the service receiver and provider can rate each other after a transaction, and they calculate the *trust degree* to the others [6-10]. The trust degree is a value between 0 and 1, denoted by d . For example, when there are two participants i and j , i can rate j after each transaction, and the trust degree of i trusting j (denoted by d_{ij}) is the ratio of positive ratings. The relations among those evaluations can be illustrated by a graph called a reputation evaluation diagram.

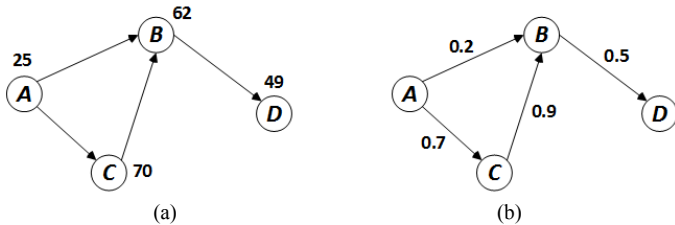


Figure 1. (a) Trust value reputation system. (b) Trust degree reputation system.

B. Local Trust Model

The objective of a local trust model is to calculate the relative reputation degree (*local trust degree*) of each participant in a local system using a value between 0 and 1. According to the two approaches explained above, we propose two methods to calculate the local trust degree of an individual participant. Fig. 1 illustrates the two types of reputation systems: the *trust value* reputation system and the *trust degree* reputation system.

1) Trust Value Reputation System

We assume n participants in the system, and each participant i , $i \in (1, n)$, has a trust value v_i given by other participants after a transaction in trust value reputation system. Thus, the local trust degree of i can be calculated as follows:

$$local_d_i = \frac{v_i}{\max_{j \in (1, n)} v_j} \quad (1)$$

Here, $local_d_i$ is a value between 0 and 1 that indicates the ranking of i 's trust value in the system. Thus, in Fig. 1 (a), the local trust degree of D $local_d_D$ is 0.7 according to equation (1).

2) Trust Degree Reputation System

To calculate the local trust degree of participant j in trust degree reputation system, we first need to obtain the trust degrees of all the other participants trusting in j , i.e., we need all d_{ij} , $i \in (1, n)$, which can be calculated as follows:

$$d_{ij} = \sum_{k=1, k \neq i, k \neq j}^n d_{ik} d_{kj} \quad (2)$$

Equation (2) is a recursive equation used in computing d_{ik} or d_{kj} . In Fig. 1 (b), d_{AD} can be calculated as $d_{AB} \times d_{BD} + d_{AC} \times d_{CD}$, and d_{CD} can be calculated as $d_{CB} \times d_{BD}$, i.e., $d_{AD} = d_{AB} \times d_{BD} + d_{AC} \times d_{CB} \times d_{BD} = d_{AB} \times d_{BD} + d_{AB} \times d_{BD}$. In other words, there are two ways to calculate d_{AB} : one results in the *direct trust degree*, and the other in the *indirect trust degree*. However, in a reputation evaluation system, direct trust will be more reliable than indirect trust. Thus, if a trust degree can be calculated both ways (direct and indirect), the direct way will be selected. In this case, we use only the direct trust degree to calculate d_{AD} , whose result is 0.1. Similarly, we can calculate d_{BD} and d_{CD} as 0.5 and 0.45 respectively.

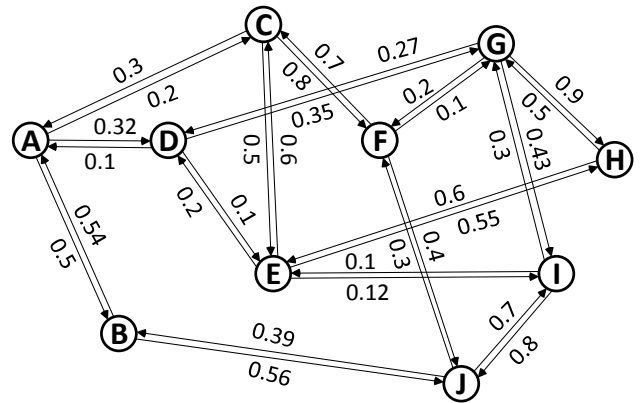


Figure 2. Reputation evaluation diagram.

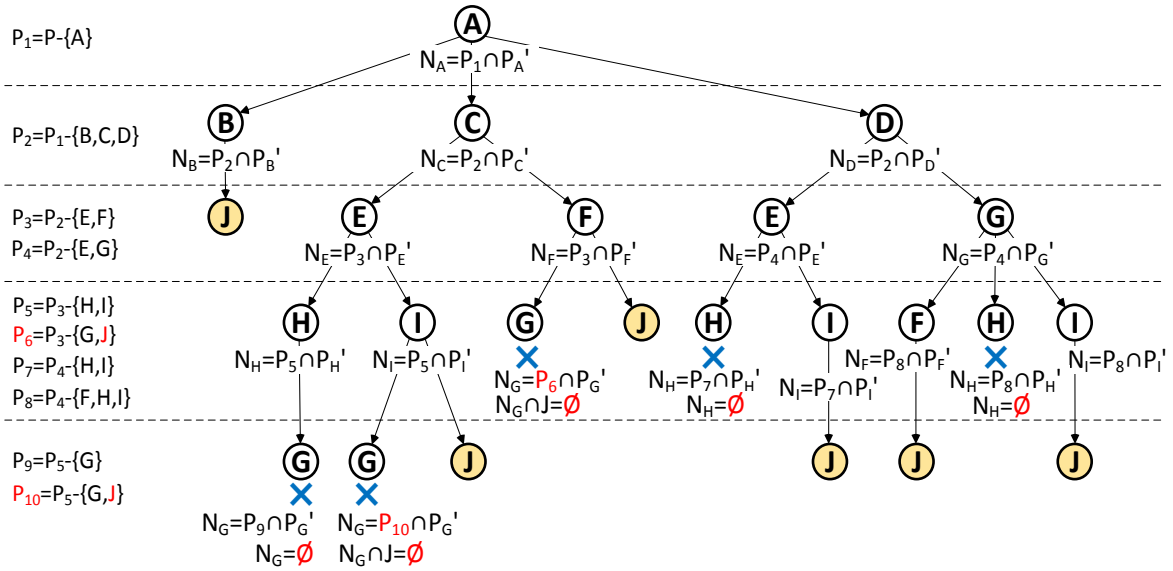


Figure 3. The calculation process of the trust degree from p_A to p_J using NRIT-SA.

However, in a realistic reputation system, a large number of peers can communicate and rate one another; thus, a realistic reputation evaluation diagram is huge and complicated. There are many different paths from one peer to another, which makes it difficult to select a direct trust way among them. Moreover, the reputation evaluation diagram will contain some cycles, as shown in Fig. 2, which is a difficult issue for the indirect trust degree calculation, and most research has excluded it from consideration. In this paper, we propose an algorithm to calculate the indirect trust degree by considering the cycles and excluding redundancies when there is direct trust. We call it the non-redundant indirect trust search algorithm (NRIT-SA).

We assume that a P2P evaluation system contains n peers, and each peer k can be considered as a point in the reputation evaluation diagram, which is denoted p_k , $k \in (1, n)$. If peer k evaluates peer l after a transaction between them, an arrow is generated from k to l in the diagram, denoted arr_{kj} . Thus, the reputation evaluation diagram composed of the point set $P = \{p_k\}$, $k \in (1, n)$, and the arrow set $ARR = \{arr_{kj} \mid k, j \in (1, n) \text{ and } k \neq j\}$. $\forall p_k \in P, \exists P_k'$, which is a set of points that p_k can indicate in the reputation evaluation diagram. As shown in Fig. 2, $P_C' = \{p_A, p_E, p_F\}$. Moreover, if $P_k' = \emptyset$, there is no point that p_k can indicate.

NRIT-SA is a breadth first search algorithm that does not search for the shortest path from source to destination. Its objective is to calculate the indirect trust degree using as much direct evaluation of others as possible. The reputation evaluation diagram shown in Fig. 2 is based on a trust degree reputation system. To calculate the indirect trust degree from p_A to p_J based on equation (2) and exclude redundant and cyclic evaluations, the NRIT-SA algorithm can search the indirect evaluation paths, as shown in Fig. 3.

Step 1: First, as a source point, p_A should select a set of points as its next step, denoted N_A . As shown in Fig. 2, a point set P_A' contains all the points that p_A can indicate, here $P_A' = \{p_B, p_C, p_D\}$. To avoid a redundant path, p_A should remove itself from

the available point set P and generate a new point set $P_1 = P - \{p_A\}$ that can be used instead of P in the next step. Then, the next step point set of p_A can be calculated as $N_A = P_1 \cap P_A' = \{p_B, p_C, p_D\}$. NRIT-SA will move to each of the next step points of p_A to continue. If there is a $p_i \in N_A$ as the destination point, then there is a direct evaluation from the source to the destination peer, and the algorithm is complete because there is no need to calculate the indirect trust degree.

Step 2: To avoid a redundant path, the available point set P_1 should remove each point p_i that belongs to set N_A (i.e., $p_i \in N_A$) to generate the newly available point set P_2 . In this case, $P_2 = P_1 - \{p_B, p_C, p_D\}$. $\forall p_i \in N_A$ will calculate its next step point set by $N_i = P_2 \cap P_i'$. If $N_i = \emptyset$, there is no point that p_i can indicate, and the search from p_i will be stopped. And if $N_i \cap p_j = \emptyset$ (here p_j indicates the destination point), a direct trust way has already been found in the previous step p_i to the destination point, so the search from p_i will be stopped. If $N_i = p_j$, then the path from the source to the destination should be recorded as one of the NRIT paths. For the other cases of N_i , NRIT-SA will move to each of the next step points of p_i and repeat the similar process in Step 2.

After finding all of the NRIT paths from a source point to a specific destination using NRIT-SA, the trust degree from the source point to the destination can be calculated using Equation (2). In the example shown in Fig. 3, the trust degree d_{AJ} can be calculated using all the NRIT paths: $d_{AB} \times d_{BJ} + d_{AC} \times d_{CE} \times d_{EI} \times d_{IJ} + d_{AC} \times d_{CF} \times d_{FJ} + d_{AD} \times d_{DE} \times d_{EI} \times d_{IJ} + d_{AD} \times d_{DG} \times d_{GF} \times d_{FJ} + d_{AD} \times d_{DG} \times d_{GI} \times d_{IJ}$. Similarly, we can also calculate the trust degrees from all the other points to this destination. According to the trust degrees of j given by all the other participants in the system, the local trust degree of participant j can be calculated as follows:

$$local_d_j = \frac{\sum_{i=1, i \neq j}^n d_{ij}}{n-1} \quad (3)$$

Here, n indicates the number of participant peers in the system, and $local_d_j$ is a value between 0 and 1 that represents the average trust ranking for peer j based on the ratings given by all the other participants in the system.

III. PROPOSED CROSS-DOMAIN TRUST MODEL FOR HYBRID CDNI-P2P NETWORK

From a reputation estimation point of view, mobile peers who move among different systems or domains need trust values in each P2P system. In general, different P2P systems or domains adopt different trust models and reputation evaluation standards. That leads to disparate trust values for a single peer in different domains, even if the user always offers the same performance. In this section, we propose a cross-domain trust model for a mobile peer in a hybrid CDNI-P2P network.

A. Hybrid CDNI-P2P Network Architecture

Our proposed hybrid CDNI-P2P network architecture is shown in Fig. 4. The architecture contains two types of CDNs: uCDNs and dCDNs. Content provided by CSPs is stored only in the edge servers of the uCDNs. If an end user u_i , who can only obtain service directly from dCDN-A, sends a content request to the origin server, the content will be delivered from the uCDN to dCDN-A and then transmitted to the end-user through the closet edge server of dCDN-A. If another end-user from the same domain also wants to obtain this content, s/he can get the content directly from u_i . In this situation, each CDN can act as an uCDN and dCDN simultaneously based on the content requested by the end-user.

B. Cross-Domain Trust Model

In Fig. 4, a mobile end user peer (the red solid point) can move between uCDN and dCDN domains. Some mobile peers with high trust values in one P2P domain could have their trust values initialized based on the trust model used in a new domain when they move because they are new-comers to the new domain. This will lead to a waste of resources, because other peers are unwilling to obtain content from a new peer. The mobile peer must wait a long time to accumulate trust in the new domain.

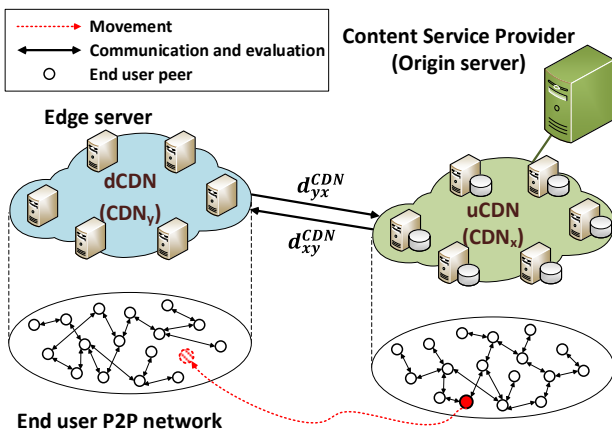


Figure 4. Hybrid CDNI-P2P network architecture.

We assume that the uCDN and dCDN can also rate each other after each transaction between them, and that their evaluation method is based on the trust degree reputation system described in Section 2. We denote the two CDNs as CDN_x and CDN_y respectively. The trust degree that CDN_x gives CDN_y is d_{xy}^{CDN} , and the trust degree that CDN_y gives CDN_x is d_{yx}^{CDN} . If a mobile end-user u_m in the CDN_x domain with a local trust degree of $local_d_{u_m}^{CDN_x}$ (as explained in Section 2) moves to the CDN_y domain for the first time, the *mobile trust degree* of u_m from CDN_x to CDN_y can be calculated as follows:

$$mobile_d_{u_m}^{CDN_x \rightarrow CDN_y} = local_d_{u_m}^{CDN_x} \times d_{yx}^{CDN} \quad (4)$$

Here, it should be noted that the trust degree between CDNs indicates how CDN_y rates CDN_x . If u_m already has a trust value (or trust degree) in CDN_y , it is used continuously.

If CDN_y is a trust value reputation system and the number of total participants is t , the *mobile trust degree* of u_m can be transformed to the trust value in CDN_y as follows.

$$v_{u_m}^{CDN_y} = mobile_d_{u_m}^{CDN_x \rightarrow CDN_y} \times \max_{j \in (1,t)} v_j^{CDN_y} \quad (5)$$

And if CDN_y is a trust degree reputation system, the *mobile trust degree* of u_m can be the trust degree in CDN_y for all other participants.

IV. PERFORMANCE ANALYSIS

Based on the proposed algorithm NRIT-SA, mobile users can calculate their local trust degree which is largely dependent on the direct evaluations among users in the local system. The proposed NRIT-SA can eliminate all the redundant indirect connections between two users, if there is the direct evaluation between them. Because we believe the fact that the direct evaluation is more trustworthy than the indirect one. Moreover, for users who cannot provide the direct evaluation to a specific user, i.e. these users have never conduct a transaction before, the indirect evaluation from others can also be used to generate the trust degree to the specific user. Thus, the connectivity rate among users become a significant determinant of the performance in NRIT-SA.

In this section, we will discuss the relationship between the connectivity rate of users and the calculation time of the proposed algorithm. First, we will give the definition of the connectivity rate, here it is called *connectivity degree*, as follows.

Definition 1 (Connectivity degree)

For a local system with n users, the connectivity degree can be calculated by

$$d_{connectivity} = \frac{\sum_{i=1}^n \sum_{j=1, j \neq i}^n [direct_d_{ij}]}{n(n-1)} \quad (6)$$

Here, $direct_d_{ij}$ indicates the direct trust degree between user u_i and u_j , which can be considered as the direct connection between node u_i and u_j in Fig. 2. From the definition, we know that the connectivity degree indicates the percentage of real connections among users in a local system. According to this

definition, we will analyse the relationship between the calculation time of proposed algorithm NRIT-SA and the connectivity degree of users in a local system.

The tool we use to implement NRIT-SA is JDK 1.7.0_80, and 3.60GHz processor with 64-bit OS. The connections among users are randomly selected based on the connectivity degree. For two specific user u_i and u_j , the time to calculate the trust degree d_{ij} based on NRIT-SA is shown in Fig. 5.

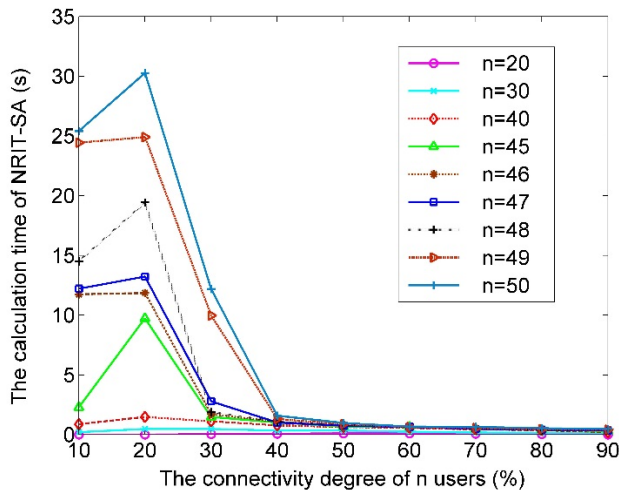


Figure 5. The calculation time of NRIT-SA based on different connectivity degree of the n users.

In Fig. 5, the calculation time indicates the average value of processing time based on different connection topologies which is randomly generated. From the figure, we can see that the processing time of NRIT-SA increases along with the growth of the number of users in the system. And when there are 50 users, the longest calculation time is around 30s which can be tolerant by an evaluation system, because the calculation of trust degree d_{ij} is used for generating the local trust degree of mobile user u_j which can be considered as an offline value during a short period of time. The reason is that, for user u_j , its local trust degree $local_d_j$ can be influenced by the change of other trust degree between any two users in the system, however, according to equation (2) and (3), this influence is negligible small when there are large numbers of users in the local system.

Moreover, from Fig. 5, we can see that, for different number of users, the peak values of calculation time typically appear around 20% of connectivity degree. And along with the increase of connectivity degree, NRIT-SA processing time decreases exponentially. When the connectivity degree increases to 40%, all the calculation time is around 1s. According to a well-known fact that the number of system users will tend to be stabilized, the connectivity degree will increase with time, thus the calculation time of proposed algorithm can also tend to a small and stable value.

V. CONCLUSIONS

In this paper, we propose a hybrid CDNi-P2P architecture, an NRIT search algorithm, and two trust models: a local trust model and a cross-domain trust model. Based on the proposed

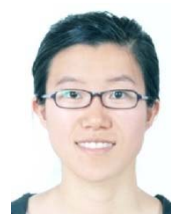
NRIT-SA and trust models, a user can calculate his/her local trust more effectively and accurately, and a mobile user can transform his/her local trust into mobile trust that can be taken to and used in a new domain. The proposed models can avoid disparate trust values for a single user in different domains and improve the availability of content possessed by mobile users as they move among different domains. And from the performance result, we know that the peak value of the calculation time appears around 20% of connectivity degree, and along with the increase of the connectivity degree, the calculation time will decreases exponentially. In the future, we will research more available cross-domain trust models for different network architectures.

ACKNOWLEDGMENT

This work was supported by the National Research Foundation of Korea (NRF) grant funded by the Korea government (MSIP) (2016R1A2B4015899). It is also partially supported by the State Scholarship Fund organized by the China Scholarship Council (CSC).

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