Developing a Replication Strategy Based on Fuzzy Logic to Enhance Data Availability in Cloud Data Center

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Abstract- In massive cloud systems, efficient data management is crucial. This element's management is greatly streamlined by file replication. The purpose of this project is to spread replicates over a variety of data centers in order to boost data availability in cloud data centers. How clones are created could depend on how useful and accessible the data is in the cloud data center. Throughout the replication process, high-access data is given precedence. The process of choosing the best copy to deploy across many data centers is known as replica selection.

Index Terms— Computing in the cloud, data centres, and replication are examples of index terms

I. INTRODUCTION

The users the flexibility of accessing a shared resource pool from any location globally, whenever needed, without themselves with concerning system installation maintenance. Provisioning hosted services for users is a key aspect for ensuring high data availability. The proposed approach encompasses three primary categories. The first section entails replica creation, contingent upon the data's accessibility. Subsequently, the second step involves selecting the most suitable copy based on network performance. The third section focuses on the replication installation process and replacement of older replicas when necessary. The final stage integrates the replica selection mechanism, utilizing fuzzy logic to determine the optimal quantity of replicas and enhance data availability through this replication approach.

II. REPLICATION OF DATA

Data replication stands as a viable approach to meet data integration needs. Through replicating data, synchronization is maintained, ensuring data availability, promoting efficient data expansion, and leveraging the latest data to enhance revenue. Real-time analytical data synchronization adds value to a diverse range of businesses.



Fig. 1 Statistics for Duplication

Data replication, owing to its robust capabilities in transferring vast data volumes swiftly with minimal latency, proves highly effective in distributing workloads across multiple sites and ensuring continuous data accessibility. 1111ustrated in Figure 1 is the fundamental methodology was employed for data replication.

The concept of data replication involves maintaining replicas, essentially multiple copies of the data. Replication enhances data availability, enabling data access even if some replicas are inaccessible. Moreover, it optimizes system performance by reducing latency for users, redirecting them to use local copies of data instead of accessing remote networks.

WORKS THAT ARE INTERCRETED

Cloud computing grapples with notable challenges concerning the effective implementation of data management strategies. Research has delved into the realm of replication, classifying the processes into two primary categories: While dynamic replication can dynamically generate or remove clones based on data accessibility. In 2012, Bakhta Meroufel and team introduced a dynamic replication strategy for hierarchical grids, considering system crashes and failures [1]. The bedrock of dynamic replication lies in data accessibility and its usage prevalence. Depending on data popularity, the proportion of available data might increase if the data is highly favored.

DHR deletes files within the local area network (LAN) or files with short transfer times. The algorithm stores replicas in the most accessed location, diverging from the approach of storing originals in various locations. In 2015, Reena S. More and team introduced the custom partitioning algorithm [7]. The custom partitioning method aims to break down extensive problems into more manageable subproblems. In this suggested approach, data would be replicated on a high-performance node, reducing energy consumption during data processing [8].

WORK TO BE PROPOSED

The main objectives of this endeavor include augmenting data availability through distributed replication across diverse Information hubs and curbing dynamism ingesting through regulating. Our strategy based on fuzzy logic led to notable reductions in energy consumption. We employ a technique known as MORM

A. System Architecture

Four modules make up the proposed architecture.

- 1. One of the modules is data centre configuration.
- 2. Making copies
- 3. Replica choice
- 4. Placing a duplicate

Configuration of the Data Center, to start



Fig. 2. Schematic of the proposed system

Replica Manager, Level Two: The replica manager oversees all replication phase functions, encompassing replica development, replica selection, and replica placement. This involves information on replica specifics such as their location, size, and quantity. If the primary data center experiences an outage, the replica manager redirects user requests to an operational zone. a) Replica Development: Replication occurs during the replica construction phase, with the timing influenced by user request numbers. The process involves generating the required additional copies and distributing them across different data centers.

Typically limit active replication copies to three. Once a replica is successfully installed in the primary data center, users gain immediate access to the file. Condition there's sufficient storage space in the primary data center, a replica is constructed and stored there. b) Replica Selection: Selecting a replica involves choosing the data center with the best replica from all available data centers (e.g., primary and secondary zones).

This optimizes provisioning efficiency when needed, reducing strain on the network bandwidth. Replica selection optimizes data processing in cloud data centers by providing the most suitable data set for user requests, resulting in minimal memory usage and swift data processing.

Two considerations for replica selection include: Opt for a nearby data center replica to reduce bandwidth requirements. Choose a network region with fewer user requests to minimize distributing replicas of the original resource across multiple

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data centers. The ongoing process of relocating replicas to new data centers without the resource reduces storage space requirements, and replica updates continue on a regular basis.

5. APPARATUS EXAMINATION

Mist Predictor: An imperative necessity to adopts a graphical user interface, simplifying the evaluation of social networking tools concerning the geographical dispersion. Dynamic re organization The design of the Cloud Analyst tool is depicted in Figure 3, with different colors representing diverse regions. In this representation, six geographical areas are encompassed.

Each region hosts one or more user bases linked to data centers situated in undisclosed regions.

TABLE I. Correlation of the three possibilities for 10 Under and 1 DC

Scenario 2: 50 User Bases & 1 Data Center

Data center region : 3 No.of data centers : 1 No.of userbase : 50 A) Assessment of Service Broker Policies: Comparative analysis across three scenarios.

Information centre area- 3 Data centre count: One User count: 10

Fig. 3. Cloud analyst screenshot



	Closest data center.	Optimize response time	Reconfigure dynamically
Overall response time	Avg (ms) - 571.23 Min (ms) - 38.86 Max (ms) -1245.12	Avg (ms) - 570.21 Min (ms) - 39.36 Max (ms) - 1260.12	Avg (ms) - 571.70 Min (ms) - 39.11 Max (ms) - 1245.61
Data <u>center</u> processing time	Avg (ms) - 0.27 Min (ms) - 0.02 Max (ms) - 0.86	Avg (ms) - 0.27 Min (ms) - 0.01 Max (ms) - 0.87	Avg (ms) - 0.75 Min (ms) - 0.02 Max (ms) - 30.01
Data <u>center</u> request servicing times	Avg (ms) - 0.27 Min (ms) - 0.02 Max (ms) - 0.86	Avg (ms) - 0.27 Min (ms) - 0.01 Max (ms) - 0.87	Avg (ms) - 0.75 Min (ms) - 0.02 Max (ms) - 30.01
Total virtual machine cost (\$)	0.50	0.50	3.26
Total data transfer cost (\$):	0.64	0.64	0.64
Grand total: (\$)	1.14	1.14	3.90

	Closest data <u>center</u>	Optimize response time	Reconfigure dynamically
Overall response time	Avg (ms) - 647.14 Min (ms) - 37.79 Max (ms) - 1300.12	Avg. (ms) - 647.48 Min (ms) - 38.61 Max (ms) - 1305.11	Avg (ms) - 647.78 Min (ms) - 38.41 Max (ms) - 2680.01
Data <u>center</u> processing time	Avg (ms) - 0.25 Min (ms) - 0.01 Max (ms) - 0.90	Avg (ms) - 0.25 Min (ms) - 0.01 Max (ms) - 0.90	Avg (ms) - 0.86 Min (ms) - 0.01 Max (ms) - 1635.01
Data <u>center</u> request servicing times	Avg (ms) - 0.25 Min (ms) - 0.01 Max (ms) - 0.90	Avg (ms) - 0.25 Min (ms) - 0.01 Max (ms) - 0.90	Avg (ms) - 0.86 Min (ms) - 0.01 Max (ms) – 1635.01
Total virtual machine cost	0.50	0.50	2.25
() Total data transfer cost (\$):	3.20	3.20	3.20
Grand total: (\$)	3.70	3.70	6.46

TABLE II. Correlation of 50 UB and 1 DC under three different circumstances

Considering response time, processing duration, data transfer expenditures, and related aspects, demonstrate nearly identical performance. In performance and cost, these two policies outshine the dynamically reorganize policy.

VI. ANALYSIS OF ALGORITHM

Algorithm- Placing replicas

```
Input: Integer r, the number of replicas
  Input: Integer dc, the number of data centers
  Output: Replicas [0...dc - 1][0...r -1]
  for all i such that i = 0 or i < 3 do
  /((i - maximum allowable count of replicas. Where i=3
  {
          Preload node dc's replicas with dc
  for all i such that i < 3 do
         repeat until i=3
         z = random node, s.t. 0 = z < dc
v = Rep[z]
         until z = i
         if v = i and Rep[i] = z then
                  valid rep = 1
                  if Rep[i]= v or Rep[i]= Rep[z] then
                            xalid rep = 0
                  if valid rep then
                            Rep[z]≡ Rep[i]
                            Rep[i] = v
```

The previously mentioned algorithm also includes an outline of the replication placement procedure. In this context, inputs encompass the replica count and the data center count, producing an output of successfully replicated replicas across the designated three for a specific resource, initiating the replication process requires preloading the original copy. The recently created replica must be situated on node Z, assumed to be a random node, and the replica count should align with the intended number of data centers containing the resource. This process continues until reaching the count of 3. If a replica does not precisely match the resource it was copied from, it is considered invalid (i.e., valid_rep=0). Conversely, if the replica perfectly matches the resource, it is considered valid (i.e., valid_rep=1).

VII. IMPLEMENTATION RESULTS



Figure 4 illustrates a screen capture displaying DC details.

The information that was retrieved about the chosen data center is shown in Figure 4. Since these features are only available to cloud service providers, only they have the power to add or alter data in the cloud.



Figure 5 provides a screen capture showcasing DC documents.

Figure 5 displays the documents that may be found in the main data center and are controlled by users under the supervision of the service provider.

Documents in Primary DC	
Primary DC	Replica Creation Check Availability
	cloud.img 👻
	Place File into
	⊯ dc1
_	₩ dc2
Message	
i	Inserted
	ок
	dc6
	PLACE

Figure 6 presents a screen capture of the replica creation process.

Figure 6 shows how to make copies in practice. The selected file ('cloud.img') is stored in two different data centers, DC1 and DC2, respectively.



Figure 7 displays a screen capture for availability verification.

The verification of Figure 7, where both Data Center 1 and Data Center 2 contain copies of the file named 'cloud.img.' The earlier described implementation integrates modules for responsibility of the service provider. The process of duplicating a file stored in the cloud, based on the file's access frequency, is denoted as replica creation.

VIII. IMPLEMENTATION IN TOOL

(CLOUD ANALYST)



Figure 8: Simulation employing a Single Data Center & 10 User Bases.

Picture 8 portrays simulation outcomes utilizing a single information Hubs and ten user bases. In this scenario, due to the presence of a solitary data center, the concept of replication is unachievable.

Overall Response T	lime Summary			
	Average (ms) Minimum (ms) Maxi	num (ms)	Farred Davalte	
Overall Response Time:	279.94 39.86 615.	1	Export nestris	
Rata Cantar Drocaccian Tima	2010 001 086			
Response Time By Regi	on Aq(ms)	Win (ms)	Max (ms)	
Response Time By Regi	on			
Response Time By Regi Userbase	on Ag(ms)	Win (ms)	Wax (ms)	
Response Time By Regi Userbase UB10	on Ag(ms) 498.713	Vin (ms) 412.111	Wax (ms) 615 11	
Response Time By Regi Userbase UG10 UG1	Ang (ms) 490.713 50.262	Min (ms) 410.111 40.609	Max (ms) 615.11 50.055	
Response Time By Regi Userbase UB10 UB1 UB2	Ang (ms) 498.713 50.262 200.172	Min (ms) 410.111 40.609 158.12	Max (ms) 615.11 59.855 252.118	
Response Time By Regi Userbase UB10 UB1 UB2 UB3	Aug (ms) 498.713 50.262 2001.172 300.019	Min (ms) 410.111 40.609 158.12 238.62	Max (ms) 615 11 50 855 252 118 361 613	
Response Time By Regi Userbase UB10 UB1 UB2 UB3 UB4	Ang (ms) 498,713 50,262 200,172 300,019 499,967 499,967	Min (ms) 410,111 41,609 158,02 238,62 377,631	Nax (ms) 815 11 58,855 252 118 361,613 585 119	
Userbase Userbase UB1 UB1 UB2 UB3 UB4 UB5	on Aug (ms) 60 260 200 172 300 113 499 977 499 977	Min (ms) 410.111 40.609 158.12 238.62 377.63 397.151	Max (ms) 615 11 58,855 252 118 361,813 565,119 615 114	
Response Time By Regil Userbase UE10 UE2 UE3 UE4 UE5 UE6	on Aug (ms) 50,262 2001,172 300,019 499,967 499,967 199,561	Win (ms) 410,111 44,009 168,12 228,62 377,631 300,116 105,127	Nat (ms) 615 11 58.855 252 118 361.813 565 119 615 114 245 114 245 114	
Response Time By Regi Userbase UE10 UE2 UE3 UE4 UE5 UE6 UE6 UE7	on Arg (ms) 488.713 200.172 200.017 300.019 489.907 499.903 199.963 50.051 50.051	Min (ms) 410,111 40,509 156 12 234,62 377,631 390,115 165,121 386	Nax (ms) 558,855 252,116 361,613 556,119 615,114 245,114 245,114	
Response Time By Regi Userbase UB10 UB1 UB2 UB3 UB4 UB5 UB6 UB7 UB8 UB7 UB8	en Arg (ms) 493 711 0 0 262 200 712 300 011 493 901 193 561 193 561	Win (ms) 410,011 40,000 158,12 230,62 377,631 390,119 165,121 39,811 39,8113	Max (ms) 515 11 58855 252 118 361 813 555 119 615 114 245 114 60 119 245 113	

Figure 9: Summary of Overall Response Time (For Figure 8.1).

The overall response time is shown in Figure 9 along with response times for various user bases. Response time is used in this situation to measure the algorithm.



Figure 10: Simulation utilizing 2 Data Centers & 10 User Bases.

Figure 10 showcases simulation results employing two data centers and 10 user bases, enabling the replication concept. As a result, the same resource can be held by two different data centers.

iverali Response i	ime sum	imary		
	Average (m	s) Minimum (ms) Maximu	m (ms)	Export Results
Overall Response Time:	190.82	38.36 610.11		
lata Contor Drocoreina Timo	. 0.33	39.0 10.0		
Licerhoos	ion	Aun (ma)	Nin (ms)	May (mo)
Userbase	ion	Avg (ms)	Min (ms)	Max (ms)
Userbase JB10	ion	Avg (ms) 50.081 50.145	Min (ms) 40.361 38.856	Max (ms) 61.115 60.365
Userbase JB10 JB1 JB2		Avg (ms) 50.081 50.145 200.513	Min (ms) 40.361 38.856 147.116	Max (ms) 61.115 60.358 240.113
Userbase JB10 JB1 JB2 JB3		Avg (ms) 50.081 50.145 200.513 300.569	Min (ms) 40.361 38.856 147.116 238.614	Max (ms) 61.115 60.358 240.113 367.615
Userbase UB10 UB1 UB2 JB3 JB4		Avg (ms) 50.081 50.145 200.513 300.559 49.808	Min (ms) 40.361 38.856 147.116 238.614 38.357	Max (ms) 61.115 60.358 240.113 367.615 60.358
Userbase JB10 JB2 JB2 JB3 JB4 JB5		Avg (ms) 50.081 50.145 200.513 300.563 49.808 500.305	Min (ms) 40.361 147.116 238.614 30.367 387.581	Max (ms) 61.115 60.358 240.113 367.615 60.358 610.13
Userbase JB10 JB1 JB2 JB3 JB4 JB5 JB6		Avg (ms) 50.081 50.145 200.513 300.569 49.808 500.306 199.879	Min (ms) 40.361 38.856 147 116 238.614 38.357 387.581 142.111	Max (ms) 61.115 60.358 240.113 367.615 60.358 610.113 249.114
Userbase UB10 UB2 UB3 UB4 UB5 UB5 UB5 UB7		Avg (ms) 50.081 50.145 200.513 300.569 48.808 500.306 199.879 50.278	Min (ms) 40.361 38.856 147.116 238.514 38.557 387.581 182.111 38.87 38.837	Max (ms) 61.115 60.358 240.113 3.367.515 60.358 610.13 249.114 61.622
Userbase UB10 UB2 UB2 UB3 UB3 UB4 UB6 UB6 UB6 UB7 UB8		Aug (ms) 50.081 200.513 300.569 48.808 500.306 198.879 50.278 200.758	Min (ms) 40.361 38.856 147.116 238.614 38.357 38.7591 182.111 38.87 182.111 38.87	Max (ms) 61.111 60.358 240.113 367.615 60.358 610.113 249.114 61.822 245.113

Figure 11: Summary of Overall Response Time (For Figure 8.3).

The response times for each user base are shown in Figure 11, along with the total response times. Response time is used as the measurement in this specific algorithm iteration. Utilizing a network of data centers dramatically decreases

response times overall and for each user base serviced when compared to using a single data center concept (Figures 9, 11).

IX.CONCLUSION

As a strategy to augment the accessibility of data within cloud data centers, an effective replica creation approach has been proposed. To ensure an optimum environment for customers, the data center's configuration is initially undertaken by an certified manipulator. Within the replica creation module, data replicas are produced based on their utilization frequency. The algorithm was formulated to streamline the deployment of replicas, consequently causing a deceleration in the reaction. Our upcoming endeavors drive focus happening refining an efficient method for replica management to reduce both costs and the requisite storage space.

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