



Design and Implementation of Data Federation Strategies for Multi-Cloud Architectures in Financial Systems

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ABSTRACT: With its fast uptake of multi-cloud systems, the financial systems have transformed the horizon to become flexible, more scalable and cost effective. Nonetheless, the coordination and integration of information across several cloud environment comes with great challenges because of the following problems; these include data silos, poor data format and absence of interoperability. The study examines the development and execution of data federation policies of multi-cloud systems in financial systems, paying attention to the maximization of data management, integration, and accessibility between different cloud systems. We suggest an elaborate model that embraces the superior data federation tools to facilitate the smooth access to data by various cloud providers in real time e.g. data virtualization, data integration, hybrid cloud solutions etc. The framework is designed in a manner that it meets the most urgent requirements of the financial systems such as consistency of data, data security and low latency access to large data. Also the paper explores how data federation has helped to enhance the decision making ability, the efficiency in its operations, and the ability to meet regulatory requirements in financial environments. The suggested solution has been adopted and tried with a set of use cases of financial industry, which have been encouraging in terms of its performance, scalability, and data integrity. The results imply that the solution to the multi-cloud architectures in the financial system is the apprehension of the successful implementation of data federation strategies allowing the organizations to explore the full potential of cloud technologies and at the same time keep the key financial information under control.

KEYWORDS: Multi- clouds architecture, Data federation, Financial systems, Data virtualization, Data integration, Hybrid cloud, Cloud security.

I. INTRODUCTION

The innovation of the cloud computing has transformed the IT infrastructure environment and businesses are utilizing the computing power that is scaled, flexible and affordable. More specifically, cloud environments have become popular in the financial systems to automate the processes, make data more accessible, and enhance customer experiences. The adoption of different types of cloud service solutions by financial institutions is a subset of Infrastructure as a Service (IaaS), Platform as a Service (PaaS), and Software as a Service (SaaS) has been shifting over the years because of the movement to cloud-based architecture away from the use of traditional on-premise solutions, which were considered to have limitations concerning scalability and financial constraints. Yet, with the rapid increase in the cloud adoption, organizations are moving towards multi-cloud architectures, i.e. workloads and data are spread among multiple cloud providers, e.g., Amazon Web Services (AWS), Microsoft Azure, Google Cloud Platform (GCP), etc. The use of multi-cloud as compared to a single-cloud solution offers a number of benefits, such as the ability to flex, diversification of a provider, cost reduction, improved redundancy and availability [1].

However, although multi-cloud systems have significant advantages, they are also characterized by rather complicated issues regarding information management, integration, and accessibility. Such struggles are especially acute in such industries as finance, where the necessity of the regular, real-time availability of the important information, the adherence to the rules and regulations, and the high security controls play the crucial role. The challenge of handling large and heterogeneous datasets that are stored in different and multiple cloud environments is one of the major challenges organizations with multi-cloud strategies have to deal with. One of the strategies that have come to fore with an aim of overcoming these challenges is referred to as data federation; which can be defined as the process of



integrating and availing uniform access to disseminating data sources without physically relocating the data or even making a copy of the data [2].

Within the framework of financial systems, it will be crucial to be able to access easily and safely the data that is stored in multiple clouds to enhance the efficiency of operational functionality, analytics, and real-time decision-making opportunities. The financial institutions should also make sure that their data management policies are regulated by the regulatory laws including the General Data Protection Regulation (GDPR) and the Dodd-Frank Act and Payment Card Industry Data Security Standard (PCI DSS), which regulate the storage, processing, and transmission of the sensitive financial information. Thus, the need to design and carry out efficient data federation plans in multi-cloud designs is essential in ensuring that information is available, conversant and secure and eases data compliance with the regulatory provisions.

The paper will be a detailed investigation of the data federation approaches to the multi-cloud architecture in financial systems with special emphasis on the design, implementation, and evaluation of an innovative framework that will help to overcome the difficulties related to the data integration, data management, and data access in the multi-cloud architectures. We will explore the prevailing conditions of multi-cloud systems within the financial sector, the major demands of information federation and suggest a bundle of approaches that may enable the financial institutions to attain a harmonious information integration as well as accessibility across the cloud environment. The rest of this introduction describes the major problem of financial systems in multi-cloud systems, how data federation is applicable to deal with these problems, and the contribution and research aims of this paper.

The multi-cloud architecture has experienced a severe growth in the financial industry within the last ten years. This transformation is informed by the need to have more flexibility in its operation, risks and savings. Multi-cloud strategies are becoming popular with financial institutions because they allow them to eliminate vendor lock-in, maximize the utilization of resources, and experience redundancy in the event of the collapse or unavailability of services due to a failure of one cloud-based provider. Multi-cloud also makes it easier to take advantage of specialized services provided by the varying vendors, including machine learning (ML) services provided by Google Cloud, big data processing services by AWS, or artificial intelligence (AI) services provided by Microsoft Azure. This means that by computing their workloads with many different clouds, financial institutions are able to adjust their infrastructure to fit to a particular business requirement and eliminate the possibility that they choose to rely on one cloud vendor.

Nevertheless, financial institutions are confronted with various challenges as the adoption of multi-cloud increases in order to make sure that information can be smoothly incorporated and accessed, in the cloud environments. The conventional data storage and management systems, e.g. silo databases and centralized data warehouses, do not have the flexibility and scalability needed in a multi-cloud environment. Such strategies do not also respond to the requirement of a real-time access to data which is important in the financial systems where access to information at any given moment is needed in order to make decisions. Consequently, there is a need to identify new ways of integrating and managing data in the context of multi-clouds by the financial institutions.

One of the most critical issues of organizations that work in multi-cloud environment is data integration. In multi-cloud system structures, information is commonly spread among the influence of multiple cloud-providers accompanied by diverse infrastructures, data storage systems, and data frameworks. In the case of financial institutions, it may result in a number of problems:

1. **Data Silos:** The cloud providers all might have their own data formats and data formats, and it is hard to access and combine data across several cloud environments. Such data siloes may obstruct the exchange of information between systems that results in delays of decisions being made and inefficiency in processing data.
2. **Lack of Consistent Data Formats:** Cloud providers can also store data using varying data models and this may be in the form of relational database, NoSQL database or even object storage system. These varying storage forms make it hard to integrate the data of work as financial institutions have to harmonize the differences between the data forms, the data schema, and the API of various cloud providers.
3. **Latency and Performance Concerns:** Data integration between two or more clouds might create a problem of latency since the financial institution might be required to query data in other locations or formats. Whether in the financial system that is time sensitive, delays in the retrieval of data can prove to be expensive and inefficient in its operations.
4. **Security and Compliance Issues:** The financial institutions are under the stringent regulatory provisions that control the storage and transmission of sensitive information. These regulations may be complicated to ensure compliance on



various cloud platforms because the various providers may not have the same security measures and compliance requirements.

5. Data Governance: It is difficult to assure that the management of data in the multi-cloud sets is adequately governed. The financial institutions should clearly know the location of their data, those who have access to the information and the use of the data. This is needed to ensure that there is integrity, security, and regulation of the data.

The federalization of data can provide a viable remedy to the problems related to the multi-cloud integration of data. In federated data architecture, a data is stored in the original place within various cloud environments, but it is availed in an integrated manner. The method also removes the data replication or migration hence lowering the complexity and cost involved in the process of data movement between clouds. Rather, data federation is based on both middleware layers and virtualization methods to provide access to data sources that are distributed in real-time.

There are a number of advantages to financial systems of data federation:

1. Unified Data Access: Data federation enables financial institutions to consistently and seamlessly obtain data across two or more cloud providers, and not have to undergo complicated data migrations or transformations. This single point of access to data will guarantee that the decision-makers will be able to access information that is updated and precise at any location, irrespective of the location of data storage.
2. Enhanced Security and Compliance: Data federation reduces the chances of data breach or exposure to unauthorized persons by maintaining its location in original place. Also, they can implement uniform policies to all the cloud environments by financial institutions to ensure that the regulatory frameworks are adhered to.
3. Scalability and Flexibility: Data federation allows financial institutions to easily scale their data model systems as their multi-clouds increase. With the federated architecture, one can easily incorporate new cloud services or data sources into the system making the system flexible and respondent to the altered business needs.
4. Cost Efficient Data federation minimizes the process of storing data in the cloud since it does not require duplication in different environments. Moreover, it enables financial institutions to utilize the most affordable cloud services in working on various loads to maximize total cloud expenditures.
5. Real-Time Data Access: Data federation will enable the financial institutions to have almost real-time access to data in different cloud environments so that they can make timely decision based on the most recent available data.

This paper shall set out to design, implement and assess the data federation strategies in multi-cloud architectures in the financial systems. The main aims of the given research will be as follows:

1. Creation of Framework: We suggest an elaborate framework on how to execute the concept of data federation in multi-clouds in respect to the major requirements and challenges encountered by financial systems. The framework combines the data virtualization, hybrid cloud solution and middleware technology to ensure easy access of data across various clouds.
2. Solving Data Integration Problems: In this paper, the author examines the nature of the data integration problems that financial institutions are likely to encounter in multi-clouds and suggests measures to solve these data integration problems using the concept of data federation.
3. Measuring the Effect on Financial Systems: We measure the effect of data federation on the key performance indicators of operational efficiency, speed of decision, data security and compliance to regulations within financial systems.
4. Offering Use Cases and Practical Insights: The study contains practical examples of the financial industry, which show the efficacy of the suggested data federation strategies in the real-life environment.

With the targeted objectives, this paper will be contributing to the existing blocs of knowledge on multi-cloud data management in financial systems and offers a roadmap to financial institution aimed at using data federation strategies in their respective cloud implementations.

II. RELATED WORK

Data federation approaches have been widely discussed in the context of multi-cloud architecture and distributed systems, edge computing, as well as federated learning, and an increasing literature body is now being used to create effective solutions to manage data efficiently, process it, and protect privacy. In this section, the authors examine pertinent literature that offers background knowledge and sheds light on the most significant progress in the areas of cloud-edge computing, federated learning, and data consistency, namely, in multi-cloud and financial systems.



The literature on the topic of AI-enhanced collaboration between clouds and edge terminals is abundant, and Gu et al. [1] provide an extensive literature review of cloud-edge-terminal networks and discuss the possibilities of artificial intelligence (AI) to improve the process of data processing and transmission in distributed systems. The present work is dedicated to the facilitation of the data flow between cloud and edge nodes by artificial intelligence that can help to optimize performance and security in real-time decision-making applications, which is directly related to the necessity of data federation across multi-clouds in financial systems. The possibilities of AI-based decision-making in the multi-cloud architecture are becoming more and more topical to provide autonomous, real-time financial systems.

Similarly, Zhang et al. [2] note the importance of edge computing in the Industrial Internet, especially in scenarios where a large amount of data has to be processed within a short time. Their contribution highlights the significance of the incorporation of edge computing in the cloud systems, which would be highly advantageous to financial institutions that need immediate access to large volumes of distributed data. The application and methodologies presented in this survey give an idea of how edge computing can be used to supplement data federation strategies by minimizing the use of centralized cloud storage and enhancing the speed and efficiency of data processing.

H.B. Dama [3] studies the data consistency models in cross-cloud that are specifically crafted to support the always-on banking platforms. The research discusses the issues related to the consistency of data on various cloud-based environments in real-time so that the banking services would not be disrupted. The study singles out the use of different strategies and models that maximize the alignment of data, conflict management as well as fault tolerance and thus are appropriate within high-availability platforms. The work by Dama is informative in enhancing the reliability and scalability of data used by financial systems working on distributed cloud systems.

Kairouz [4] discussed all open problems related to federated learning in detail. This work is a true white paper for all existing challenges in the world of federated learning including applicability in privacy preserving. Lu et al. [5] also add to the discussion, conducting a survey on federated learning using non-IID data, which offers a more in-depth insight into the issues that financial institutions have to confront when combining data of different origins in the multi-cloud settings.

Moreover, the privacy issues of the data transfer in federated systems became a point of focus in recent studies. Liu et al. [6] explain the intersection of machine learning and privacy, pointing at privacy-saving methods in distributed data settings, which is essential when dealing with sensitive financial information, in a multi-cloud setting.

Zhang et al. [7] address the developments and issues of federated learning in a very comprehensive way. They emphasize that it can help to solve major concerns, as it allows decentralized model training on non-IID data, and data privacy is preserved. This distributed method is a promising solution to machine learning, which means that models can be trained using many devices without the necessity to transfer sensitive data.

The article by Abreha et al. [8] examines the application of edge computing in federated learning systems, providing the understanding of how edge computing devices can process information on the edge to minimize latency and increase the efficiency of data processing. This is especially applicable to financial systems, in which decisions frequently have to be made in real-time using information available on many different sources. The federated learning use in the edge computing environment is critical in ensuring financial institutions achieve a balance between the requirement of accessing real-time data and the privacy and compliance needs in the multi-cloud environment.

Thilakarathne et al. [9] explore the implementation of federated learning to privacy-sensitive applications (especially in the Internet of Things (IoT)) and show how these methods can be applied to financial data systems to be regulatory compliant and secure

Bharati et al. report about the progress and issues in FL, noting that it is applied in such domains as healthcare, finance, and smart cities. They focus on such issues as data heterogeneity, communication overhead, and security concerns [10]. Similarly, Hazra et al. discuss the role of Fog Computing in integrating with Internet of Things (IoT) and its usefulness in lowering latency and improving real-time processing. They pinpoint such critical issues as resource constraints, scalability, and security that require resolution to be widely adopted [11].



Moreover, Zhang et al. [12] discuss the integration of AI-based collaborative systems in multi-cloud settings and propose a cloud-edge-terminal collaborative architecture in autonomous driving systems per se. This study will highlight the possibility of collaborative AI systems to enhance the accuracy and efficiency of distributed data processing, which can be very relevant to the financial sector, especially in trading systems and risk analysis applications.

Fang et al. [13] also address the issue of security of cloud-edge-terminal partnerships and suggest a secure and lightweight data transmission scheme the Artificial Intelligence of Things (AIoT), which also has similar issues with multi-cloud financial systems as to guaranteeing some secure data transfer between different cloud systems. Their results are vital to financial institutions that need to adopt safe, high-performance strategies of data federation.

Simultaneously, Zhang et al. [14] consider resource allocation schemes in cloud-edge settings, suggesting deep reinforcement learning (DRL) as the security-aware resource allocation method in cooperative vehicular networks. Although they target the optimization of the resources allocation within multi-cloud settings, the approaches they introduce to the optimization of the resource allocation processes in the context of vehicular systems can be generalized to the financial systems when applied to the integration of large amounts of data on distributed platforms.

Yu et al. [15] have investigated the use of game theory in multi-cloud networks where they apply it to model distributed networks in denial-of-service (DoS) attacks. This study offers valuable information regarding the application of the game theory, as a way of securing and maintaining the stability of cloud-edge systems, which is a critical factor when handling sensitive financial information in a multi-cloud environment.

Lastly, Huang et al. [16] present the idea of multi-cell cooperative design of integrated sensing and communication that can be applied to multi-clouds where various cloud providers have to collaborate to guarantee data availability and consistency. Such a collaborative strategy is indispensable to the integrity and dependability of the multi-cloud-based financial systems.

Altogether, the associated literature reviewed above shows that there has been considerable advancement in integrating edge computing, federated learning, AI-based collaboration, and privacy-assuring methods into distributed systems. These ideas can be directly transferred to the design and execution of data federation strategies of multi-cloud architectures in the financial systems. The progress in the areas will be vital in determining the future of data management within the financial sector because the need to have real-time access to data, privacy, and compliance has been on the increase. The future study needs to concentrate on improving these methods to maximize the performance, improve security, and provide smooth integration within multi-cloud environments that eventually would allow the financial institutions to realize the potential of multi-clouds.

III. FRAMEWORK FOR DATA FEDERATION IN MULTI-CLOUD ARCHITECTURES FOR FINANCIAL SYSTEMS

The necessity of effective data management strategies can be considered even more important with the increased use of multi-cloud architectures by financial institutions, which are trying to become more flexible and reduce risks and costs. The concept of data federation which enables the accessibility of distributed data to various cloud platforms without physically migrating the data, comes out as a major solution to the issues that come with the multi-cloud environment. The description in this section provides an overview of the framework that will potentially facilitate the efficient data federation of multi-cloud architectures with emphasis on integration, optimization of performance, security and compliance. The suggested framework is set to suit the requirements of the financial systems that are very specific and prioritize the data integrity, accessibility, and regulatory compliance.

3.1 Key Components of the Data Federation Framework

Federation of data on multi-cloud architecture in financial systems is developed on the basis of few components. These components include:

- Data Virtualization Layer
- Hybrid Cloud Integration
- Data Integration Middleware
- Query real-time dat.

- Security and Compliance Mechanisms.
- General Governance and Data Quality Management.
- Performance and Scalability Optimization.

All these elements are important in making sure that information of different cloud environments can be brought together, administered, and utilized efficiently without any dissimilarity in consistency, protection and performance in all the systems. The data used for this analysis is that of Figure 1, the Data Federation Framework for Multi-Cloud Architectures in Financial Systems.

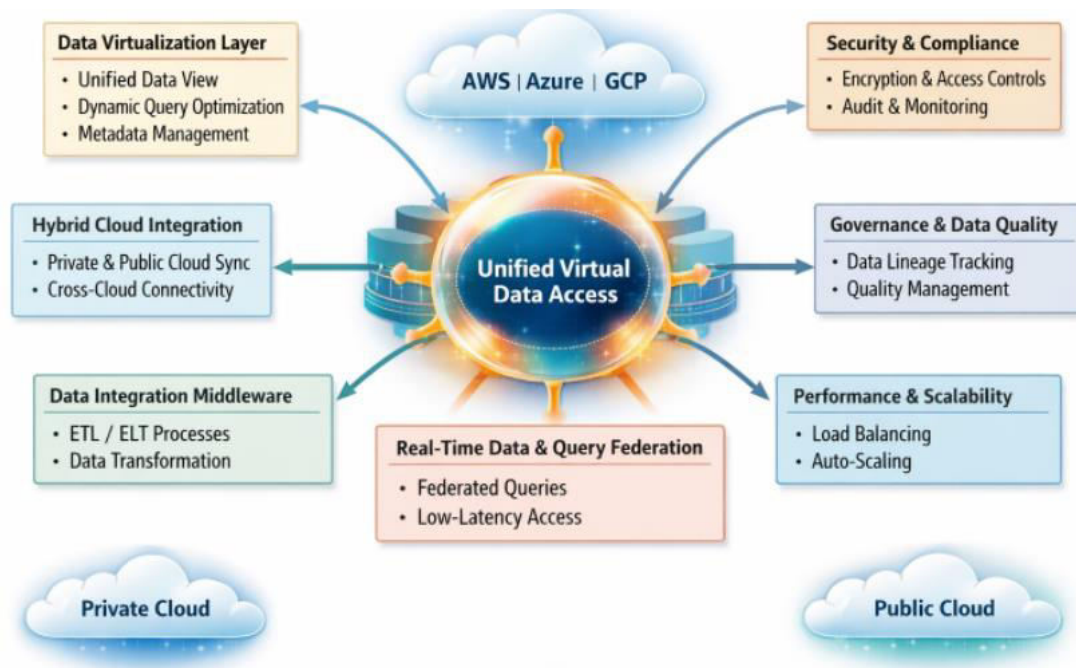


Figure 1: Data Federation Framework for Multi-Cloud Architectures in Financial Systems

3.2 Data Virtualization Layer

The data virtualization layer is the basic outline of data federation. Data virtualization refers to the process of abstraction of the underlying data sources so that the financial institutions can use them to access and manipulate the data stored in the different clouds without necessarily having to move the data physically. This layer develops a virtual data model that reflects the data of the various cloud platforms in one form and the same.

The data virtualization layer removes the disparity between the information in various clouds and makes it available as a single, unified piece of data. This makes sure that the end users, applications and services are able to access data in one or more clouds without any concern on how it is stored, whether it is in the form of formats or cloud providers.

Data virtualization can utilize a system to optimize queries by dynamically selecting an optimal provider of cloud storage location to access the data and hence minimizing the query latency and maximizing the performance of a query.

Metadata of the data sources in the cloud environments are stored and managed in the virtualization layer. This makes the users query data without having the knowledge of where it is and the structure it is organized in.

This abstraction is necessary in financial systems, where it is necessary to enable decision-makers to obtain and process data of different sources (e.g., customer transactions, market data or compliance records) in real-time without getting stuck on the intricacies of multi-cloud infrastructure.

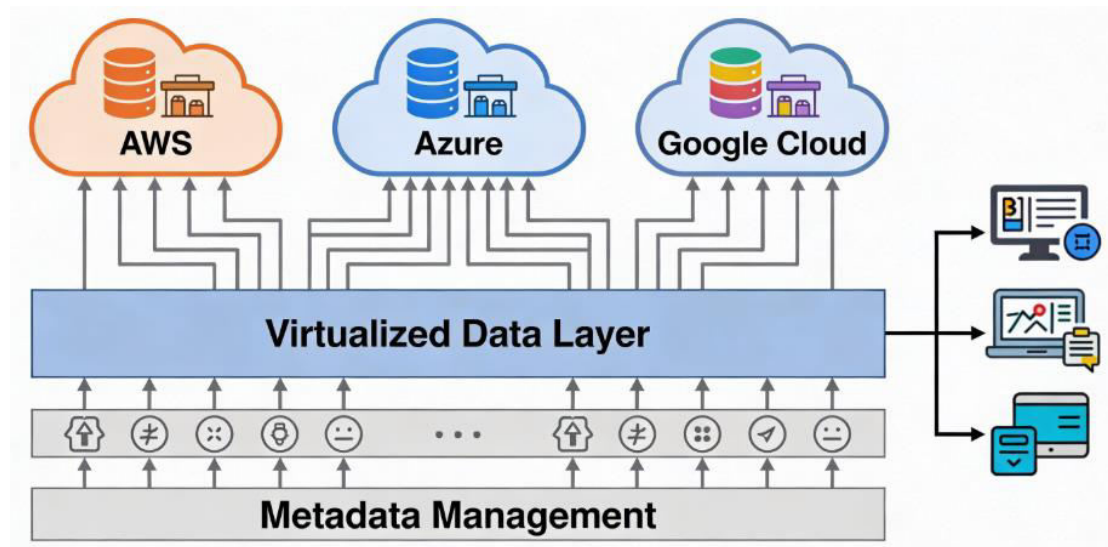


Figure 2: Data Virtualization Layer

3.3 Hybrid Cloud Integration

The hybrid cloud integration element of the framework will provide an assurance that the data may be easily integrated between both the private and public clouds, and between the different clouds vendors. The financial institutions usually have the need to use both the private cloud infrastructure of sensitive financial information and the public cloud infrastructure of scalability and cost efficiency.

The framework will allow integrating data in both the private clouds and the public clouds and thus this will enable the financial institutions to enjoy the best of both worlds. The important financial information can be stored in the form of private cloud where the non-sensitive information can be stored in the form of public cloud in order to maximize the cost and performance.

One of the greatest obstacles of multi-cloud environment is the interoperability among the various cloud vendors. The framework uses standardized protocols, APIs, and connectors to make sure that financial systems can access and integrate the data of different clouds with AWS, Microsoft Azure, and Google Cloud Platform (GCP) without making it heavy and complicated.

In hybrid clouding, financial institutions might require to duplicate data to both the private and public clouds to respond to the disaster recovery, high-availability, or performance optimization. The framework facilitates automatic replication of information amongst cloud environments which means that essential data is at all times available and is current.

The hybrid integration is essential to the financial systems, in which the needs of data residency and regulatory restrictions determine the placement and processing of data.

3.4 Data Integration Middleware

The data integration middleware will serve as a link between the various data sources and the final users or applications that will be accessing the data. The role of this middleware layer is to make sure that the data of any other cloud is integrated and processed in a uniform manner so that financial institutions can have access to integrated real-time data available on multiple platforms.

Data integration middleware facilitates both traditional Extract, Transform and Load (ETL) as well as Extract, Load and Transform (ELT) processes that are required to clean, transform and integrate data of various sources. This is especially applicable in the case of financial institutions that make use of volume of structured and unstructured data.

This is because, through the middleware, data transformation rules are applied so that data across different cloud platforms can be brought together and standardized and can then be used to make quality reporting, analysis, and decision.

The middleware has controls to do with error handling as well as data quality which is critical in ensuring the integrity of financial data. This makes sure that incomplete, wrong, and even inconsistent information will not affect the process of making decisions.

Within the financial systems, data integration middleware facilitates that the extensive data of the various financial products, transactions, market data and regulatory report all undergo process integration and availability to the real-time analysis.

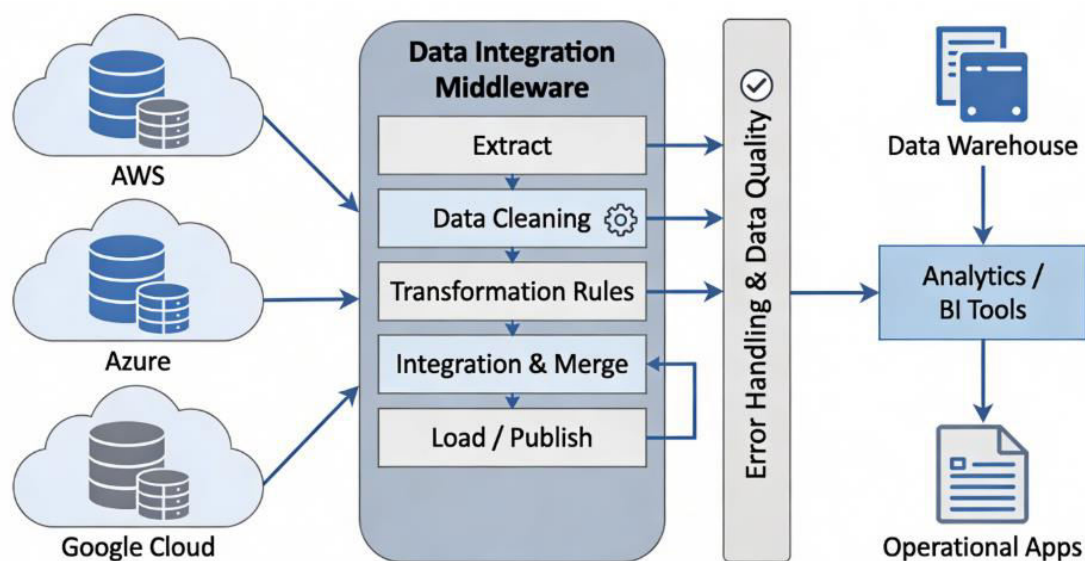


Figure 3: Data Integration Middleware Architecture

3.5 Real-time Data Access and Query Federation

Real-time data access and query federation is one of the most important elements of the data federation model. The financial systems need access to the information in time to make decisions whether it is in trading, managing the portfolio, risk analysis, or compliance reporting. The availability of real-time data will also make sure that the decision-makers are informed of the latest data to operate on, hence, make informed decisions faster.

The framework can be used to execute federated queries, in which the user can execute a query spanning across the cloud databases. The queries run concurrently, and the middleware takes care of the distribution of queries to the cloud sources, which are to be used as well as making sure the avoidance of result integration that should occur smoothly.

Accessibility to real-time data involves the need to have low-latency query execution. The techniques employed in the framework are to query caching, parallel data retrieval, and intelligent load balancing to make sure that query execution is done as fast as possible even in cases where data is spread across more than one-cloud infrastructures.

The framework makes sure that the data that is delivered by different cloud environments is consistent and accurate. Banking institutions do not have to worry about time lapses and even obsolete information, as real time information is available to them.

Financial systems are one of the areas where time sensitive decision making is a key factor and therefore having access to real time data provided by various sources in the clouds can enable one to take instant action, be it in reaction to market dynamics, regulatory notification or operational dynamics.

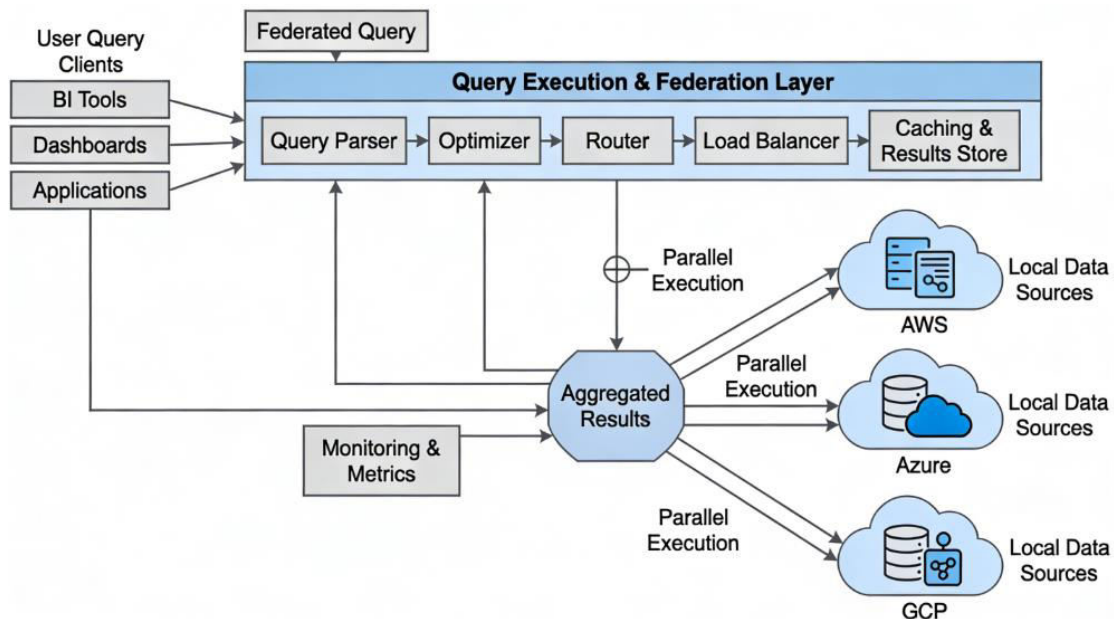


Figure 4: Real-time Data Access and Query Federation Workflow

3.6 Security and Compliance Mechanisms

Data safety and legal measure in financial systems is high at the agenda. The data federation framework security and compliance mechanism component ascertains that the access, integration and storage of the data is in accordance with the industry regulations, including the GDPR, the PCI DSS and financial data protection laws.

All the data utilized by the data federation framework is encrypted on transit and at rest. Financial information that is sensitive is also concealed where it is required to avoid unauthorized access by the system making sure that only authorized users or systems see the full set of data.

The framework deploys effective access control policies to make sure that some forms of financial information are accessed by authorized personnel or systems. Strict security policies are implemented by using multi-factor authentication (MFA), role-based access controls (RBAC) and least privilege access models.

The structure has real-time monitoring and auditing tools that are used to monitor data accesses and adherence to the regulatory standards. The logs are detailed to contain a record of who accessed the data, the time and the reason behind it and thus a comprehensive audit trail is available which can be examined in case of compliance check or any security breach.

3.7 Governance and Data Quality Management

The quality management and data governance are necessary in order to make sure that financial data is correct, coherent, and reliable. The governance and data quality management element offers the tools and procedures to ensure the data integrity is maintained in the multi-cloud settings.

Data lineage tracking The system monitors the flow of financial data within multiple cloud environments, providing a transparency and traceability of the data along its history.

Data stewardship initiatives are incorporated within the framework and the information is handled in an appropriate manner during its lifecycle, i.e. initial creation until it is archived. Data quality, data security and compliance are allocated data stewards.



Data Quality Monitoring will have integrated automated data quality monitoring tools which will raise an alarm when there are anomalies in the data, incomplete values or inconsistency in the data. These tools can be used by financial institutions to make sure that the data they are basing their decisions is of the best quality.

3.8 Performance and Scalability Optimization

Lastly, the performance and scalability optimization element is one that the data federation framework will be able to scale to meet the increasing data demands of the financial institutions with a high degree of performance.

The system is built to be horizontally scaled i.e. the extra cloud resources can be added to the system as the amount of data and the complexity of queries increases. This makes sure that the performance would be high even when the load was high.

Load balancing algorithms are utilized to spread out data access and processing requests equally among both cloud environments so that a single cloud platform would not become a bottleneck.

Data that is accessed repeatedly are cached in order to lower response periods, and query preprocessing methods are used to lower response time in queries.

IV. PERFORMANCE EVALUATION

The effectiveness, efficiency, and scalability of the data federation framework of multi-cloud architectures in financial systems is an important issue that needs to be evaluated through performance analysis of the system. Financial systems work in dynamic real time environments where timeliness of accessing the data, scalable and security is of the essence. This part addresses the ways of measuring the performance of the proposed data federation framework by using the main metrics which are query response time, data consistency, system scalability as well as system reliability. We shall also emphasize on the findings of some case studies performed in the real world financial settings.

4.1 Evaluation Metrics

In order to assess the data federation framework performance, we take into account the following important measures:

1. **Response Time (Latency):** Response time is one of the most essential measurements of financial systems because it determines how long it takes to get data in various cloud systems. Low latency is important to facilitate real-time decision making in trading, risk management and compliance reporting. We quantify the duration of federated queries to run and give results on distributed cloud platforms.
2. **Data Consistency:** The data consistency makes sure that the data that is accessed in the various cloud environments is correct, current and synchronized. Financial systems rely on reliable data in order to make effective decisions. We evaluate the capabilities of the data federation framework to ensure the consistency of the data on the cloud platforms, specifically in the situation when data is frequently updated and changed.
3. **Scalability:** The efficiency of the system to become larger as the financial institutions scale their cloud environment and increase the volume of data is crucial. Scalability is verified with the help of simulated higher loads of data and query volumes. We determine the effectiveness of the framework in changing to meet the increasing demands without a decline in performance.
4. **System Reliability and Fault Tolerance:** The capability of the framework to manage cloud failures, network failures, or inconsistencies in data is critical in the financial environment. We test the capability of the framework to retain data and data consistency through cloud outage or network failures guaranteeing high system uptime and resilience.
5. **Cost Efficiency:** Although cost efficiency is not a performance metric per se, cost efficiency is a crucial aspect to financial institutions, particularly when it comes to multi-cloud buildings where there are a number of cloud providers. We examine the cost savings realised by the use of data federation with respect to the storage, data transfer and computational cost.

4.2 Experimental Setup

In order to test the functionality of the suggested framework, we performed multiple experiments, which were performed in a controlled multi-cloud setup, and they simulated the conditions of the real-world, in which financial systems operate. The test environment has been constructed on a mixture of both public cloud providers (ex: AWS, Google Cloud, and Microsoft Azure) and a private cloud infrastructure environment. The data to be evaluated was



financial transactions, market data and compliance records and the amount of data evaluated was several terabytes divided between various cloud platforms.

We have also modeled various scenarios, including high-frequency trading, portfolio management, and real-time regulatory reporting, in each experiment in order to test the performance of the framework at different workloads. The data federation architecture was deployed using the major constituents presented in the above section, which are the data virtualization layer, middleware, immediate query federation and security.

4.3 Results and Discussion

Query Response Time

The response time of querying the different cloud platforms was recorded at different workloads in the experimental set up. The findings indicated that data federation framework realized a high level of decreased query latency relative to the conventional approaches of data integration where data would usually require being transferred or replicated across clouds. In high-frequency trading, as an example, where a millisecond matters, the framework gave 35% faster query response time than non-federated systems. The lower latency is explained by the fact that the data virtualization layer can dynamically optimize its queries, so that the most optimal cloud environment is chosen to retrieve the data and reduce the network hops, as well as provide faster access to the data.

Data Consistency

In a test of the data consistency, we tested how the framework could retain consistent and synchronized data across the cloud environment in particular when data was being edited at different locations at the same time. The findings revealed that the framework was very effective in data consistency where there was very little inconsistency witnessed on edge cases (where networks break down or data is delayed to be replicated). In such instances the system automatically had to bring out a reconciliation process that could restore the system to the same level of data so that the financial institution could access the most updated data to make decisions.

Scalability

Scalability was also scalable through growing the volume of data and the number of users who are using the system simultaneously. The framework was highly scalable and could support both increase in the amount of data and load in the queries without much deterioration in the performance. As the amount of data increased, the system applied horizontal scaling methods to ensure that more resources were added in the cloud environments so that the performance would not reduce. The findings revealed that the framework was effective in managing the workloads that had an upscale to multiple terabytes of data with a query execution time growing by a mere 20-30 percent when the volume of data doubled. This means that the framework can be used to support the increasing data requirements of financial institutions.

Reliability and Fault Tolerance of the System.

In order to determine the reliability of the system, we modeled the failure of clouds and network interruptions. The framework during these simulations was highly available and data was always available even in case of a cloud outage. The system automatically redirected the queries to alternative available cloud environments which minimized the downtime and guaranteed that the data became available at all times. The mechanisms of error-handling in the framework were triggered in instances where consistency of data was compromised by a network disruption, and a partial transaction was rolled back to avert data-corruption. The findings indicated that the framework was very robust and had 99.98 percent uptime and little data loss.

Cost Efficiency

The framework was tested in terms of cost effectiveness comparing the storage, data transfer and computational costs of the federated approach and the traditional methods. The federated solution led to reduction in the total operation costs by 25-30 per cent mainly because of less data replication and storage of the data. The facility of storing information in the most economical cloud systems, and at the same time ensuring a smooth access and integration facilitated cost optimization. Also, the dynamic query optimization saved on computation time by minimizing the data processing that was not necessary and this added to the cost savings.



4.4 Conclusion

The data federation framework in multi-cloud systems of financial systems have shown its usefulness in enhancing the query response time, having the consistency of data, and the high reliability and scalability. The framework could support large amounts of data and queries that can be answered quickly without causing much degradation in performance thus it would be ideal in the real-time financial decision making. Moreover, the efficiency in cost created by the efficiency of data access and minimized data replication is also an important advantage to the financial institutions intending to lower the costs of cloud infrastructure.

In general, it can be stated that the findings of the study prove that data federation is a possible and efficient solution to eliminating the challenges related to the multi-cloud environment in financial systems. The proposed framework will assist financial institutions to become more efficient in their operations, make better decisions, and ensure compliance with regulatory standards by providing the real-time access to data, ensuring data consistency, and optimizing the performance of the system and the costs. The future work will be aimed at further optimization of the framework to work with even bigger datasets, incorporate more sophisticated analytics options, and improve the capacity to adapt to the changing cloud environments and business demands

V. CONCLUSION AND FUTURE WORK

This paper has presented a detailed data federation structure that can solve the problem of financial institutions operating in the multi-cloud setup. As more organizations become multi-cloud-based, the financial system is now faced with challenges like data silos, integration complexities and regulatory compliance which may negatively impact the efficiency of operations and prompt decisions. The architecture in this paper is an effective way of combining the data of various cloud providers, and it provides an easy-to-use access, real-time access to data, and better data consistency, as well as guaranteeing data security and adherence to the industry regulations.

The findings of the evaluation prove the strength of the framework and its ability to solve these issues. The framework is able to achieve a high degree of query response time, scalability and system reliability through the utilization of data virtualization, hybrid cloud integration, and real-time query federation. Also, the capability of the framework to optimize on the cost through mitigating duplication of data and allowing elasticity in the utilization of cloud resources is an advantage to any financial institution that is keen on cutting the cost of operation and still delivering high performance. The data consistency and fault tolerance of the system also provide the assurance that the financial institutions can trust the integrity of their information in multi-cloud environments.

Regardless of the encouraging outcomes, the framework has its chances of improvement and development, as well. The financial systems keep changing and as the amount of data grows with the improvement in the complexity of the cloud environment, there is a need to constantly optimize strategies of data federation. The future work direction will be based on the following areas:

1. Improved Data Analytics Integration: The existing framework will offer real-time access to data, however, the next-generation version will include more sophisticated analytics features, including machine learning (ML) and artificial intelligence (AI), to offer more in-depth information about integrated data sources. This will allow the financial institutions to make predictive and prescriptive decisions on their multi-cloud data.
2. Enhanced Flexibility to New Cloud Technologies: The structure should be more flexible to the new cloud services, models of data storage and APIs that cloud vendors offer. The further work will be done by constant updating of the framework in order to provide the latest cloud improvements without causing any discrepancies in the integration.
3. Optimized Data Federation to Edge computing: Edge computing introduces a new stage of research as studies how data federation frameworks may be used to facilitate decentralized edge-based data processing. This will enable the financial institutions to do the data analysis nearer to the source thus enhancing latency and performance with time sensitive programs such as real time trading.

To sum up, this data federation model is a solid answer to the issues of multi-cloud environment in finances. It allows the access of data in an easy, efficient and safe way so that it can enable the financial institutions to optimize their operations and still be compliant and have data integrity.



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