

# Design of Connected Data Lake System based on Micro Cloud Storage

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**Abstract**—The increasing use of ICBM (IoT & Cloud & Big Data & Mobile), which is the basis of ICT diversity, has led to an increase in various large amounts of data. This increase in data is transforming the society into a data-oriented society. Edge Cloud is increasing to manage increasing data efficiently. However, it is difficult to efficiently integrate distributed data between Edge Cloud and Cloud. This paper proposes a Connected DataLake system that can solve these problems.

**Keywords**—Cloud, micro cloud storage, data lake, message queue, stream processing

## I. INTRODUCTION

In a world where technology and life change differently from one day to another, which revolutionary change is named Industry 4.0, smart factory, and manufacturing innovation 3.0 strategy. On the other hand, the maturity of ICT (Information & Communication Technology) convergence has been diversified into autonomous vehicles, smart medical, health care, drones, robots, smart culture & life. With the increasing use of ICBM, a large variety of ICT data is increasing. The increase of this data is making society become a data-oriented society. Increasing volumes of data and a data-oriented society are increasingly important to storage.

An increase in IoT (Internet of Things) devices is a factor that explosively accelerates data growth. In other words, data generated from IoT and various things are increasing day by day, which Edge Cloud for efficient management of such data is increasing. This trend has increased the need to reliably and flexibly manage the massive data stored in the Edge Cloud storage. There is increasing interest in integrating between Edge Cloud and Cloud to efficiently manage growing data in the Edge Cloud. However, it is difficult to efficiently integrate large amounts of data between Edge Cloud and Cloud. In order to solve this problem, there is a need of research for scalability, high availability, fault-tolerant, high throughput, low latency, fast recovery, re-scaling, end-to-end consistency, and transactional integration are needed [1, 2].

In this paper, Connected DataLake is designed that a large amount of data stored in various edge clouds is integrated into a micro cloud storage for supporting scalability, high availability, fault-tolerant, high throughput, low latency and fast recovery.

## II. CONNECTED DATALAKE SYSTEM

A concept of Connected Data Architectures is an interconnected Data Pools which can connect to pools in cloud data centers. That can flow to the pools at the edge and everywhere in between. All data connected and flowing all the time [3]. The Connected DataLake in this paper limits the data connection between the edge cloud and the micro cloud storage. The micro cloud storage in this paper is constructed private cloud storage based on Ceph [4].

The Connected DataLake system proposed in this paper consists of MQT (Message Queue Transport) module and STR (Stream Processing Router) module as shown in Fig 1. The MQT module transfers a large amount of data from the distributed Edge Cloud to Micro Cloud storage. The SPR module manages data transmission securely and flexibly. The MQT module and SPR module are supported as containers in MCS (Management Center Server).

The MQT module transmits large amount of data securely and flexibly from distributed edge clouds to the SPR module. When failures or crashes occur in the Edge Cloud, it takes control of the SPR which it rolls back massive data from the micro cloud storage to the Edge Cloud. The MQT module is developed using the Kafka framework which it consists of Producer Module, Broker Cluster, and Consumer Module. The Producer module sends large amounts of data from the Edge Cloud to the Broker Cluster. the Producer module divides the data into sizes that can be transmitted. Then the divided data is serialized and posts it to the broker's topic. Broker Cluster distributes published data to cluster nodes. The Consumer module fetches the posted data in the Broker Cluster and sends it to the SPR module [5].

The SPR is a module for stream processing to manage a large amount of streaming data received from the MQT module securely and flexibly. In other words, this module safely transfers large amounts of data from multiple Edge Clouds to micro cloud storage for integration. It traces failure of a large amount of data transmitted from a variety of sources. It stores checkpoint settings and snapshots for rollback control of transmission fail or an error in the Edge cloud. It transmits and stores the object data for flexible management of large amount of transmission data of each Edge cloud. SPR module consists of Streaming Dataflow Module, Distributed Snapshots Module, Failure Tracing Module and Recovery Module. Streaming Dataflow module

The diagram illustrates the architecture of a stream processing framework. At the top, the **Management Center Server** (represented by a server rack icon) is connected to a **Data Source**. The **Data Source** feeds into a **Stream Processing Router**, which is part of a larger processing unit. This unit also contains an **MQ Transport** and a **Window Operator**. The **MQ Transport** is connected to three **Edge Cloud** components via **HTTP/RPC**. The **Stream Processing Router** is connected to a **Task Manager**, which is in turn connected to a **Job Manager**. The **Job Manager** triggers checkpoints and interacts with a **Task Manager** and a **Snapshot Store**. The **Stream Processing Router** also interacts with a **Window Operator**. The diagram includes various time-related components: **Event Time**, **Ingestion Time**, and **Window Processing Time**. A legend at the bottom identifies the components: **Producer**, **Broker Cluster**, **Consumer**, **Streaming Dataflow**, **Distributed Snapshot**, **Failure Tracing**, and **Recovery**.

### III. CONCLUSION

## ACKNOWLEDGMENT

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