Distributed HDMR (Team name)
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# Part 2: Fragmentation and Replication Techniques

#### • Introduction

The second part of this project, "EcoSphere: Distributed Commerce Gateway", looks into the implementation of advanced database techniques—specifically, fragmentation and replication—to optimize the performance of our distributed database system. We will explore both horizontal and vertical fragmentation, aiming to tailor our database tables into more manageable and efficient subsets. Horizontal fragmentation will involve splitting tables into subsets based on specific criteria, making data queries and updates more efficient. Vertical fragmentation, on the other hand, will focus on dividing tables into smaller columns to optimize data retrieval and reduce unnecessary data transmission. Additionally, we will configure a master-slave replication model.

#### • Implementation Tools and Languages

<u>Tools used</u>: VS code, pgAdmin, Git <u>Database systems used</u>: PostgreSQL

Languages: Python, SQL

#### • Fragmentation

#### •.1. Horizontal Fragmentation:

We have implemented horizontal partitioning on the <u>Users</u> table based on the "registration\_date" column. This partitioning has led to the creation of two distinct tables: <u>users\_new</u> and <u>users\_old</u>. The users\_new table has recent registrants (from year 2016 to 2023), ensuring quick access and efficient handling of current user data, which is likely to be accessed more frequently. Conversely, the users\_old table contains older registration records (from 2010 to 2015). This separation aligns with our data access patterns, allowing for more efficient query processing. We can see the code for creating the partition tables below -

```
USERS_QUERY = f"""

CREATE TABLE IF NOT EXISTS {TABLE_NAMES[0]} (

user_id SERIAL,

username TEXT NOT NULL,

first_name TEXT NOT NULL,

last_name TEXT NOT NULL,

email TEXT NOT NULL,

password TEXT NOT NULL,

address TEXT,

phone_number TEXT,

registration_date TIMESTAMP DEFAULT CURRENT_TIMESTAMP,

PRIMARY KEY (user_id, registration_date)

PARTITION BY RANGE (registration_date);

"""
```

Fig 1: Code showing Users SQL query (above) and creation of horizontal partitions (below)

This can also be seen in pgAdmin, with Users table marked with a "P" superscript and the partitions demonstrated as below -

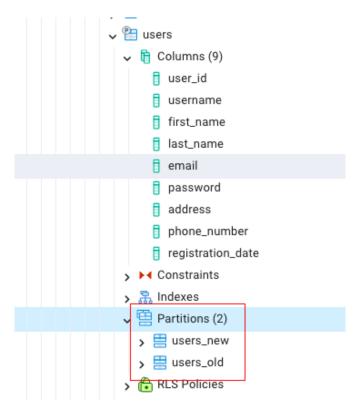


Fig 2: pgAdmin screenshot showing horizontal partitions of Users table

The initial mock data that was inserted into the Users table in Part 1, consisted of user registration date from the current decade (*registration\_date = fake.date\_time\_this\_decade()*). Hence, the Users\_old partition table should have been empty and the Users\_new partition table should contain all the 50 rows of data inserted, as can be confirmed from the images below-

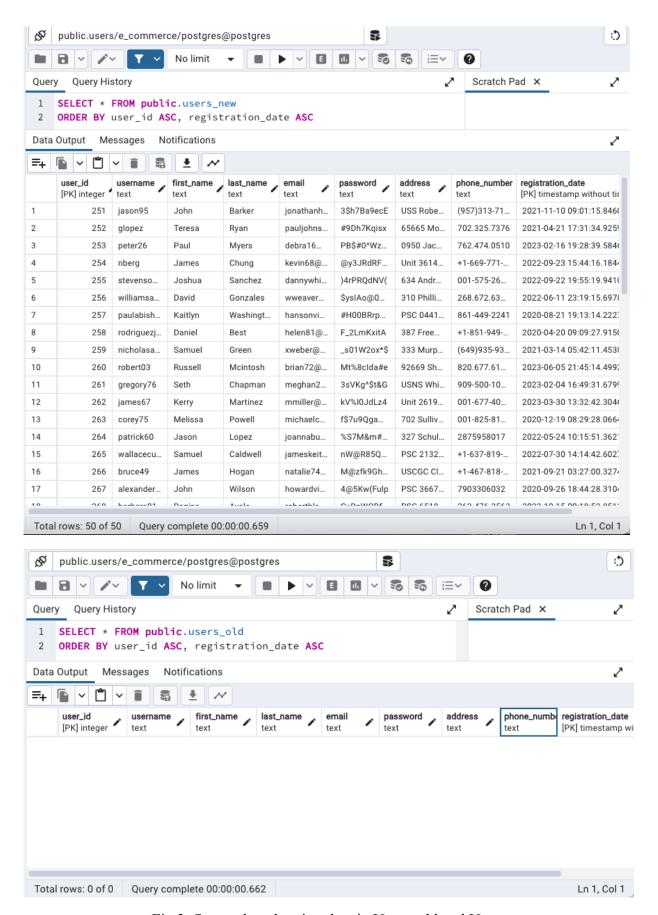


Fig 3: Screenshot showing data in Users old and Users new

Further, we are inserting 10 rows of data into the users table with registration year before 2015, and 5 rows of data with registration date in the current decade as shown below -

```
ightharpoonup fragmentation.py > \bigcirc demonstrate_H_partition
def demonstrate_H_partition(conn):
                      -HORIZONTAL PARTITIONING -----\n")
    print("\n--
    fake = Faker()
        cursor = conn.cursor()
        if conn and cursor:
            for _ in range(10):
    username = fake.user_name()
                first_name = fake.first_name()
                last_name = fake.last_name()
                email = fake.email()
                password = fake.password()
                address = fake.address()
                phone_number = fake.phone_number()
                registration_date = fake.date_time_between_dates(datetime_start=datetime(2010,1,1), datetime_end=datetime(2015,12,31))
                    INSERT INTO {table_names[0]} (username, first_name, last_name, email, password, address, phone_number, registration_date)
                VALUES (%s, %s, %s, %s, %s, %s, %s, %s, %s)
""", (username, first_name, last_name, email, password, address, phone_number, registration_date))
                username = fake.user name()
                first_name = fake.first_name()
                last_name = fake.last_name()
                email = fake.email()
                password = fake.password()
                address = fake.address()
                phone_number = fake.phone_number()
                registration_date = registration_date = fake.date_time_this_decade()
                cursor.execute(f"""
                    INSERT INTO {table_names[0]} (username, first_name, last_name, email, password, address, phone_number, registration_date)
                 """, (username, first_name, last_name, email, password, address, phone_number, registration_date))
            conn.commit()
```

Fig 4: Code showing insertion of data into Users to demonstrate partiotioning

It is expected that the data should automatically be partitioned into the Users\_old table and

Users\_new table, making it easier to handle and query for large databases. This can be verified from
the below image -

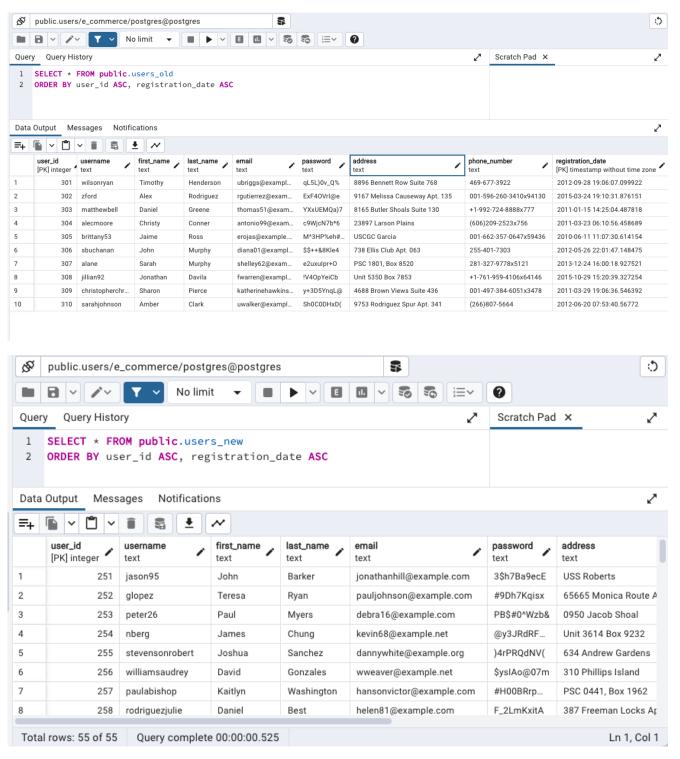


Fig 5: Data retrieval on pgAdmin showing successful horizontal partitioning As we can see in the above figures, the 15 rows of new that were inserted into the Users table got partitioned into the users\_old and users\_new tables, which are partitions of the main table. Thus we have 10 rows of old data in users\_old and 55 rows of new data in users\_new.

#### •.2. Vertical fragmentation:

We have implemented vertical fragmentation on the Products table, which resulted in the creation of two specialized tables: Products\_Info and Products\_Details. The Products\_Info table now contains key product attributes such as product\_id, product\_name, description, product\_price, category\_id, and product\_quantity. This table is tailored to serve queries primarily focused on product information and pricing, streamlining data retrieval for these common requests. On the other hand, the Products\_Details table has product\_id, created\_at, and updated\_at, for tracking and managing the product lifecycle and update history. Since postgres does not directly support vertical fragmentation, we have inserted the data into the fragments using the data from the main Products table. Following is the code for showing vertical fragmentation -

```
def demonstrate_V_partition(conn):
                          ---VERTICAL PARTITIONING -----\n")
          print("\n----
          fake = Faker()
              cursor = conn.cursor()
              if conn and cursor:
                  cat_ids = []
                  cursor.execute("SELECT category_id FROM Categories")
                  cat_ids = [row[0] for row in cursor.fetchall()]
                  for _ in range(5):
                      product_name = fake.word().capitalize()
                      description = fake.text(max_nb_chars=100)
                      product_price = round(random.uniform(5, 200), 2)
                      category id = random.choice(cat ids)
                      product_quantity = random.randint(0, 100)
                      cursor.execute(f"""
                          INSERT INTO {table_names[2]} (product_name, description, product_price, category_id,
                          product quantity)
                      """, (product_name, description, product_price, category_id, product_quantity))
122
                  conn.commit()
```

Fig 6: Code screenshot showing vertical fragmentation

Following is the data from the main Products table as well from the two vertical fragments Products Info and Products Details -

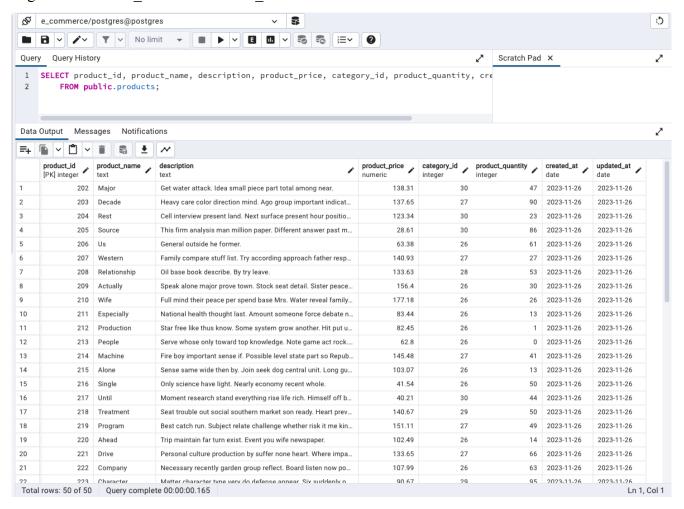


Fig 7: Screenshot of data from Products table

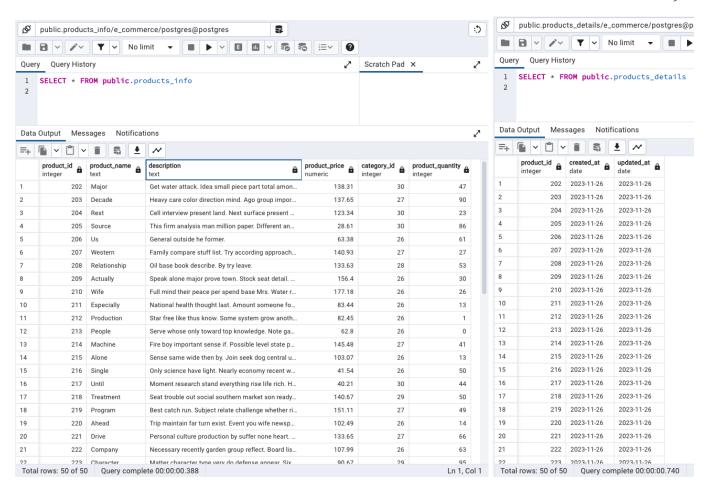


Fig 8: Screenshot of data from the two fragments Product Details and Product Info table

By dividing the Users table into users\_new and users\_old based on registration dates, and segmenting the Products table into Products\_Info and Products\_Details, we have effectively tailored our database structure to align with specific data access patterns and usage scenarios. This optimization not only enhances query performance in case of large databases by reducing data load and improving retrieval times but also helps in efficient resource management.

# Replication

The goal of this project was to establish a robust, real-time replication mechanism that ensures high data availability and disaster recovery. Streaming replication in PostgreSQL was chosen for its efficiency, low latency, and the ability to provide a near real-time copy of data from the primary to the replica node.

# **System Architecture:**

The system consists of two nodes: the **primary node**, which handles all the database write operations, and the **replica node**, which mirrors the primary node's data. The replica node operates in a read-only mode until a failover is triggered.

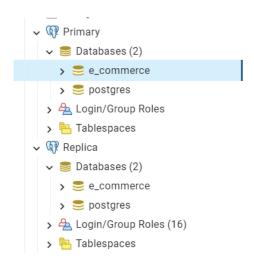
# **Configuration of PostgreSQL Streaming Replication:**

- Configuration involved setting up the primary node with the necessary parameters in **postgresql.conf and pg hba.conf** files to enable replication.
- On the replica node, similar configuration changes were made, and the node was set up to stream data from the primary node, using PostgreSQL's built-in replication functionality.

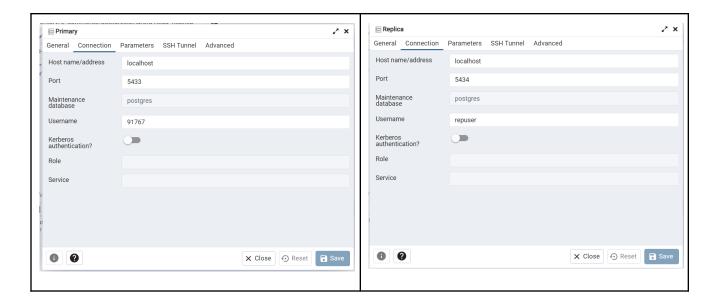
# **Testing and Validation:**

- Methods used to test the replication process, including scenarios such as creating, updating, and deleting records in the primary node and verifying these changes in the replica node in real-time.
- Testing failover scenarios to ensure the replica node can seamlessly take over as the primary node in case of a failure.

# Screenshot of Primary and Replica DB in PGAdmin



### Screenshots of Primary node running on port 5433 and Replica node running on port 5434



#### Conclusion

In this project, we have successfully implemented PostgreSQL streaming replication within a two-node architecture and fragmentation, comprising a primary node and a replica node. This setup has proven to be an efficient and robust solution for achieving real-time data replication, thereby significantly enhancing data availability and providing a solid foundation for disaster recovery. The meticulous configuration and validation of the replication process ensured that the system adhered to the key principles of data integrity and consistency. Our choice of streaming replication was instrumental in maintaining near real-time synchronization between the primary and replica nodes, with minimal replication lag and high fidelity of data.

The testing phase of the project, which included various scenarios of data manipulation and failover, further underscored the resilience and reliability of our system. The replica node's seamless takeover capabilities in failover situations ensured uninterrupted database service, thus maintaining continuous data accessibility. This successful deployment of streaming replication in a PostgreSQL environment showcases the importance of strategic planning and execution in database management, particularly in scenarios that demand high availability and data consistency. T

# References

- [1] *Streaming replication*. Streaming Replication PostgreSQL wiki. (n.d.). https://wiki.postgresql.org/wiki/Streaming\_Replication
- [2] "55.4. Streaming Replication Protocol." *PostgreSQL Documentation*, 9 Nov. 2023, www.postgresql.org/docs/current/protocol-replication.html.