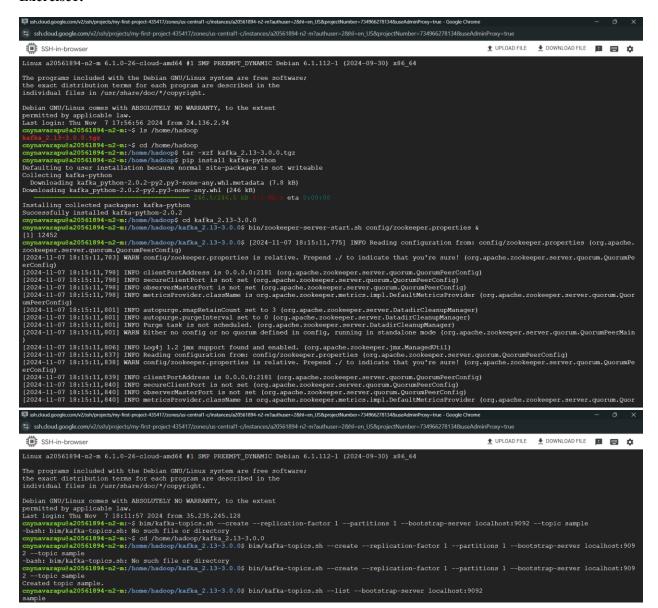
ASSIGNMENT - 9 BIG DATA (CSP 554)

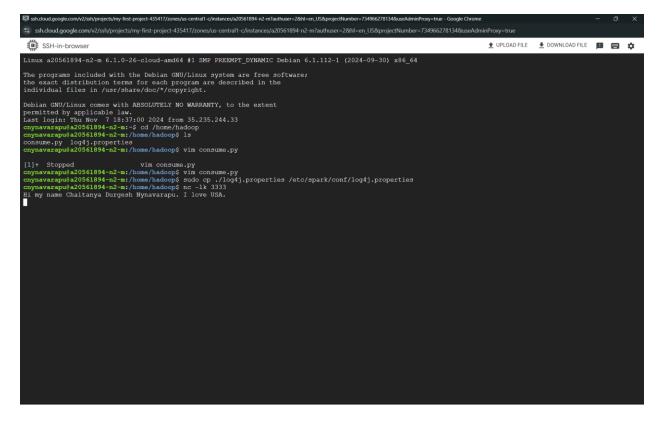
Exercise1:





```
| Part | Andropous | Part | Pa
```

Exercise2:



, ges_ap.lime*79, ges_ap.local_request_count*2, ges_count*2, ges_count*2, ges_count*2, ges_ap.local_request_count*2, ges_local_request_count*2, ges_local_request_ges

Exercise3:

"Feasibility Analysis of AsterixDB and Spark Streaming with Cassandra for Stream-Based Processing"

It explores two prominent technologies—AsterixDB and Spark (with Cassandra as a data store)—to determine their effectiveness in processing streaming data. This type of data processing, crucial for real-time insights, is particularly important for applications like social media analysis. In the study, the authors use a simulated Twitter environment to test how each technology performs with large-scale tweet ingestion, word count, and sentiment analysis tasks.

Research Focus

The authors address two main questions:

- 1. How do AsterixDB and Spark + Cassandra perform in terms of processing speed and responsiveness for real-time content and sentiment analysis?
- 2. How well do these systems scale when additional processing nodes are introduced?

These questions are especially relevant as organizations increasingly need data processing solutions that are not only fast but also capable of handling larger loads without sacrificing performance.

Methodology

In a thoughtfully designed experimental setup, the authors use Eucalyptus, a cloud platform, to simulate the process of streaming data into both AsterixDB and Spark with Cassandra. They implement word count and sentiment analysis algorithms, the latter using the SentiWordNet 3.0 lexicon. To replicate a high-throughput social media setting, the authors also test the scalability of each setup by incrementally adding nodes and observing the impact on performance.

Key Findings

- **Performance:** AsterixDB demonstrated higher throughput and lower latency than Spark + Cassandra in processing the simulated tweet data. This advantage is due, in part, to AsterixDB's data feed feature, which streamlined the data ingestion process. In contrast, Spark's integration with Cassandra introduced delays due to the need for data serialization.
- Scalability: Both technologies improved with additional nodes, but AsterixDB scaled more effectively, maintaining consistent performance gains. Spark + Cassandra, meanwhile, faced challenges with increasing node counts due to serialization and communication overhead between Spark and Cassandra.
- Sentiment Analysis: For sentiment analysis, AsterixDB's use of an inverted index significantly boosted processing speed, making it an efficient choice for tasks that rely on lexicon-based analysis like SentiWordNet. Spark required substantial resources to achieve similar performance, highlighting AsterixDB's efficiency in handling intensive data processing.

Conclusion and Implications

In closing, the authors find that AsterixDB's architecture, particularly its data feeds and flexible data model, makes it a superior choice for real-time, high-throughput applications. Its advantages in both latency and scalability suggest that AsterixDB may better support businesses and applications where low-latency and large-scale data processing are critical. Spark with Cassandra, while effective, faces limitations under high loads and may be more suited to applications with moderate real-time processing demands.

This research provides valuable insights for those looking to implement or improve stream processing capabilities. By highlighting the comparative strengths of AsterixDB and Spark + Cassandra, the authors offer a helpful framework for selecting a solution tailored to specific real-time data processing needs.