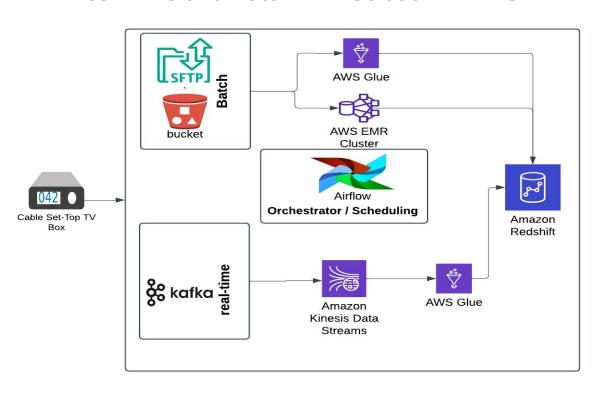
MBC CASE STUDY ASSESSMENT

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TV Viewership Data Processing Pipeline

Real-Time and Batch ETL Solution in AWS



Architecture Components

Components:

Data Sources:

TV viewership logs, metadata, and demographic data from sources like CSVs, JSON, or APIs.

2. ETL Layer:

- Extraction: PySpark handles data ingestion from source files.
- Transformation: Cleaning, joining, and enriching data.
- Loading: Load into a staging area and transform into warehouse tables.

3. Data Warehouse:

- Staging Layer: Raw ingested data.
- Core Layer: Transformed and business-ready data.
- Analytics Layer: Aggregations and summaries for business analysis.

4. Orchestration:

Apache Airflow: Workflow scheduling and monitoring.

5. **Downstream Integration**:

- Business Intelligence Tools: Dashboards and reports.
- Machine Learning Pipelines: Predictive analytics.

Design Decisions

- 1. Hybrid Data Processing (Batch and Real-Time):
 - Batch Processing: Data from an SFTP server is ingested into an S3 bucket for periodic processing. This approach supports the processing of large datasets at scheduled intervals.
 - Real-Time Processing: Kafka is used for streaming data, which is processed continuously, enabling real-time insights.
- AWS Glue:
 - Acts as an ETL service for data transformation and cataloging.
 - Simplifies schema management and ensures compatibility with downstream systems like Amazon Redshift.
- 3. AWS EMR (Elastic MapReduce):
 - Used for large-scale distributed data processing (e.g., Spark or Hadoop workloads).
 - Chosen for its cost-effectiveness and ability to handle complex data processing at scale.
- 4. Amazon Kinesis Data Streams:
 - Facilitates real-time data ingestion and processing from Kafka.
 - Integrates seamlessly with AWS Glue for transformation.
- Amazon Redshift:
 - Serves as the data warehouse for storing processed and transformed data.
 - Optimized for analytical gueries and reporting.
- 6. Apache Airflow:
 - Acts as the orchestrator, managing workflows and scheduling jobs.
 - Enables seamless coordination between batch and real-time processes.

Tools Used

- AWS S3: Storage for raw data from SFTP, ensuring durability and availability.
- 2. Apache Kafka: Industry-standard streaming platform for real-time data pipelines.
- 3. AWS Glue: Simplifies ETL processes with serverless scaling and integration with other AWS services.
- AWS EMR: Handles batch processing for large datasets.
- 5. **Amazon Redshift**: Centralized data warehouse for querying and analytics.
- Apache Airflow: Workflow orchestration for monitoring and scheduling.

Future Scalability Considerations

Data Volume:

S3 and Kinesis can handle growing data volumes. AWS services like Glue and Redshift support horizontal scaling to meet increasing demands.

Integration with Additional Data Sources:

- Kafka supports various connectors, enabling integration with new real-time sources.
- AWS Glue and S3 can accommodate diverse data formats and sources.

Real-Time Analytics:

- Use Amazon Redshift Spectrum or AWS Athena to query data directly in S3 for ad-hoc analytics.
- Add Amazon QuickSight for visualization and reporting.

Fault Tolerance and Availability:

- Leverage multi-AZ setups for critical services like Kafka and Redshift.
- Use AWS Step Functions for better error handling and state management in workflows.

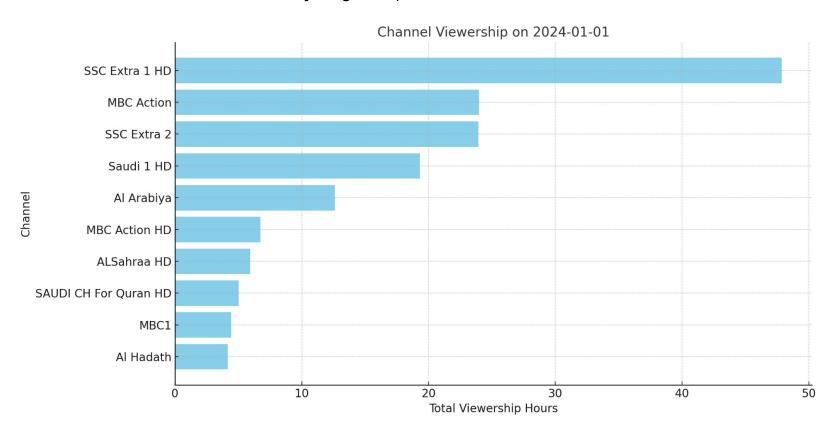
Cost Optimization:

- Utilize spot instances in EMR for batch jobs.
- Implement lifecycle policies for S3 to move less frequently accessed data to cheaper storage classes like Glacier.

Security Enhancements:

- Integrate AWS IAM roles and policies for secure access.
- Use encryption for data at rest (S3, Redshift) and in transit (SFTP, Kinesis)

Key Insights: Top 10 viewed channels on 2024-01-01



Key Insights: Average Viewing Duration Per User Over Time

