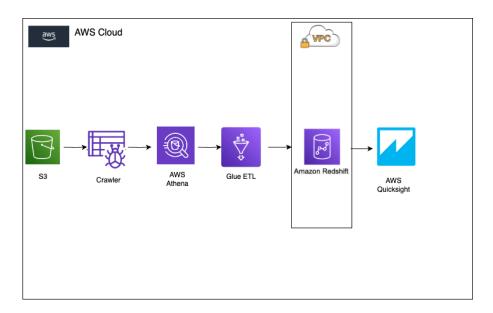
COVID - 19 End to End Data Engineering Project

- This study employs data mining techniques to analyze the COVID-19 project dataset, aiming to uncover patterns, identify trends, and determine risk factors associated with SARS-CoV-2 infection.
- By examining the provided data, the research seeks to extract valuable insights into the dynamics of COVID-19 transmission and susceptibility.
- Data set used: Available on AWS Open Dataset https://aws.amazon.com/covid-19-data-lake/

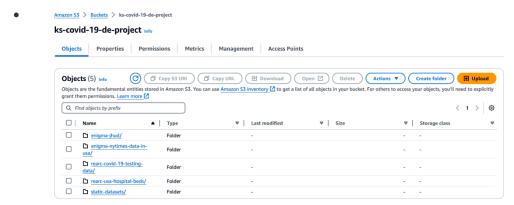
ARCHITECTURE:



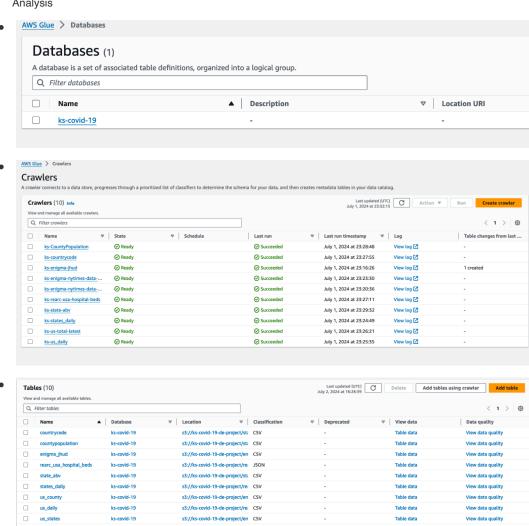
- This Architecture depicts a data processing and analytics pipeline in AWS Cloud.
- Here's an explanation of the flow:
 - o S3 (Simple Storage Service): The starting point, likely where raw data is initially stored.
 - o Crawler: A tool that scans the data in S3 to determine its structure and schema.
 - o AWS Athena: A query service that allows analysis of data directly in S3 using SQL.
 - $\circ~$ Glue ETL: Performs Extract, Transform, Load operations to prepare and structure the data.
 - Amazon Redshift: A data warehouse service where the processed data is stored for analytics.
 - AWS QuickSight: A business intelligence tool for creating visualizations and dashboards from the data in Redshift.
- All of these services (except QuickSight) are shown within a VPC (Virtual Private Cloud), indicating a secure, isolated network environment.
- This pipeline allows for ingesting raw data, processing it, storing it in a structured format, and then analyzing and visualizing it all within the AWS ecosystem.
- This kind of architecture is for big data analytics and business intelligence applications.

WORKFLOW:

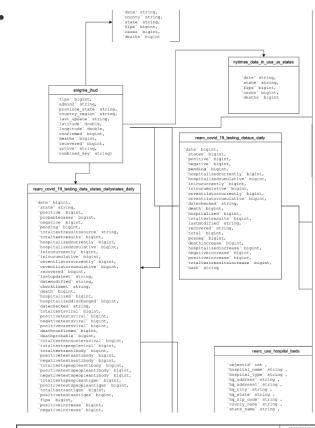
• Step 1: Manually cleaned the data and uploaded to a S3 bucket in folders.

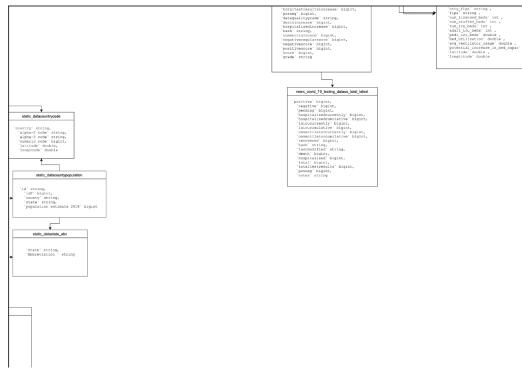


• Step 2: Ran Crawler on all the S3 folders to get Metadata information in the form of tables to be used with Athena for Analysis

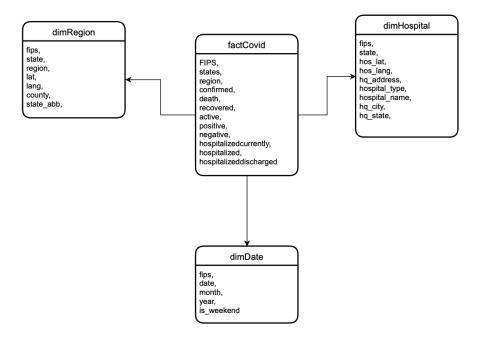


• Step 3: Once all the tables were created by Crawler, used Athena to preview the tables created and get the DDL scripts of those table to create Data Model.





• Step 4: With the help of the above Data Model, created a Dimension Model using Star Schema.



- Step 4: Using Python, Connected to Athena to query the data and storing the data in the tables created using Panda
- Step 5: With Pandas, transformed the data into the appropriate format: like replacing nulls, having the right column header, converted integer date into proper date format (yyyy-mm-dd), etc.

```
In [22]: new_header = static_datastate_abv.iloc[0] #grab the first row for the header
In [23]: new_header
Out[23]: col0
                         State
                 Abbreviation
         col1
         Name: 0, dtype: object
In [24]: static_datastate_abv = static_datastate_abv[1:] #take the data less the header row
In [26]: static_datastate_abv.columns = new_header #set the header row as the df header
In [27]: static_datastate_abv.head()
Out[27]:
               State Abbreviation
          1 Alabama
                           AL
                           AZ
             Arizona
                           AR
          5 California
                          CA
```

```
2 20210307
           0
           1
              1 20210307
           2 5 20210307
           3 60 20210307
              4 20210307
In [35]: dimDate['date'] = pd.to_datetime(dimDate['date'], format='%Y%m%d')
          /var/folders/0n/nb074wzd4kl1nmdvkbdnpgym0000gn/T/ipykernel_40148/572748324.py:1: SettingWithC
          A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide
            dimDate['date'] = pd.to_datetime(dimDate['date'], format='%Y%m%d')
In [36]: dimDate.head()
Out[36]:
             fips
           0
              2 2021-03-07
              1 2021-03-07
           2 5 2021-03-07
           3 60 2021-03-07
               4 2021-03-07
```

• Step 6: Using Pandas, created Fact and Dimension Tables from the above Dimension model and transformed the data.

Created Date dim table:

```
In [37]: dimDate['year'] = dimDate['date'].dt.year dimDate['month'] = dimDate['date'].dt.month dimDate['month'] = dimDate['date'].dt.dayofweek

/var/folders/0n/nb074wzd4kllnmdvkbdnpgym0000gn/T/ipykernel_40148/2445661104.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

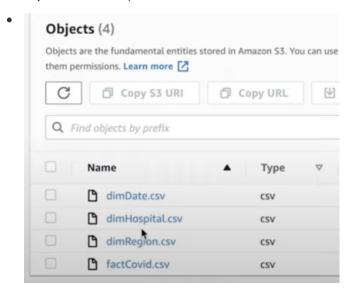
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returni ng-a-view-versus-a-copy
dimDate['year'] = dimDate['date'].dt.year
/var/folders/0n/nb074wzd4kllnmdvkbdnpgym0000gn/T/ipykernel_40148/2445661104.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returni ng-a-view-versus-a-copy
dimDate['month'] = dimDate['date'].dt.month
/var/folders/0n/nb074wzd4kllnmdvkbdnpgym000gn/T/ipykernel_40148/2445661104.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returni ng-a-view-versus-a-copy
dimDate["day_of_week"] = dimDate['date'].dt.dayofweek
```



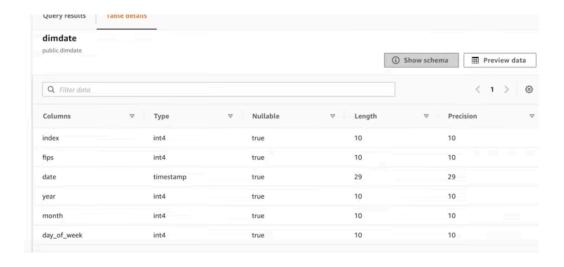
• Step 7: Stored the output of the Fact and Dimension tables in another S3 bucket.



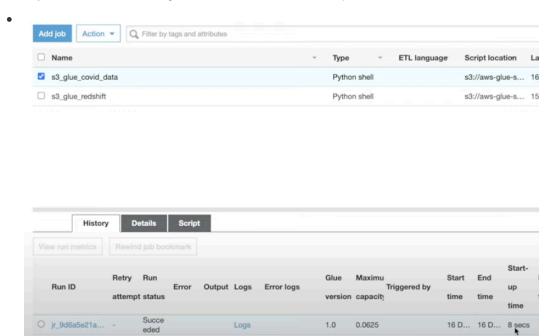
• Step 8: Create Schemas from the tables created above (Fact and Dimension) using pandas

```
Out[133]:
                         date year month day_of_week
                2 2021-03-07 2021
                                       3
                 1 2021-03-07 2021
                                       3
                                                    6
                5 2021-03-07 2021
                                       3
                                                    6
                60 2021-03-07 2021
                                                    6
                                       3
                4 2021-03-07 2021
                                       3
                                                   6
In [132]: dimDatesql = pd.io.sql.get_schema(dimDate.reset_index(), 'dimDate')
           print(''.join(dimDatesql))
            CREATE TABLE "dimDate" (
            "index" INTEGER,
              "fips" INTEGER,
"date" TIMESTAMP,
              "year" INTEGER,
              "month" INTEGER,
              "day_of_week" INTEGER
  In []: factCovidsql = pd.io.sql.get_schema(factCovid.reset_index(), 'factCovid') print(''_{3}join(factCovidsql))
  In []: dimRegionsql = pd.io.sql.get_schema(dimRegion.reset_index(), 'dimRegion')
           print(''.join(dimRegionsql))
  In [ ]: dimHospitalsql = pd.io.sql.get_schema(dimHospital.reset_index(), 'dimHospital')
print(''.join(dimHospitalsql))
```

• Step 9: Using Redshift Connector library , connect to Redshift programmatically and create these tables in Redshift.



- Step 10: Using Copy command copy the data from S3 to Redshift cluster.
- Step 11: Create Glue Job using Redshift connector and create script



• Step 12: Now once the data is in Redshift, you can use Quick-sight to visualize it.

SUMMARY:

- This document outlines an end-to-end data engineering project focused on analyzing COVID-19 data using AWS services.
- The project aims to uncover patterns, trends, and risk factors associated with SARS-CoV-2 infection using data mining

techniques.

• Key takeaways:

- o The project uses a dataset available on AWS Open Dataset.
- The architecture involves several AWS services: S3, Crawler, AWS Athena, Glue ETL, Amazon Redshift, and AWS QuickSight, mostly within a VPC for security.
- o The workflow consists of 12 steps, including:
 - Data cleaning and uploading to S3
 - Using Crawler to create metadata tables
 - Analyzing data with Athena
 - Creating a dimensional model using Star Schema
 - Data transformation using Python and Pandas
 - Creating fact and dimension tables
 - Storing processed data in S3
 - Creating schemas and tables in Redshift
 - Copying data from S3 to Redshift
 - Creating a Glue Job
 - Visualizing data with QuickSight
- This project demonstrates a comprehensive approach to handling big data, from ingestion and processing to analysis and visualization, all within the AWS ecosystem.
- It showcases the integration of various AWS services to create a robust data pipeline for COVID-19 data analysis.