



# Data Engineering 2: Big Data Architectures

## YouTube Data Pipeline

### Project Report

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# ABSTRACT

Many organizations rely on data insights to make better decisions and stay competitive. The challenge is not just storing and processing data but turning it into useful information. Analytics helps organizations figure out how to handle data and what to focus on. A well-designed data pipeline ensures data flows smoothly from where it is collected to where it can be used effectively.

To connect with Gen Z on YouTube, it's important to understand what keeps them interested. The Sidemen are a great example—they've built a huge following by looking at how people watch and interact with their videos, like tracking views, likes, and comments. This helps them see what content works best and make even better videos. For marketers, the Sidemen's engagement (how much people like and interact with their content) helps them plan better ads and sponsorships. The challenge is building an easy system to gather and show all this data so creators and businesses can learn from it and grow their own audiences.

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# 1. Introduction

This project report provides a clear explanation of how we built a system to analyse and improve the performance of the Sidemen's YouTube channels. It explains where the data comes from, the main problems we are solving, the steps taken to create the system, challenges we faced, and future possibilities.

The Sidemen, a well-known YouTube group, run multiple channels with different types of content, such as gaming videos, challenges, and collaborations. These channels produce a lot of data, including views, likes, comments, and how engaged viewers are. Collecting, organizing, and analysing this data is important for understanding which videos work best, increasing audience engagement, and creating better promotional strategies.

The main challenge is gathering data from multiple Sidemen channels, handling large amounts of information, and turning it into useful insights for content creators. This project focuses on building an efficient system that can automatically pull data from YouTube, process it, store it in a cloud platform, and display the results in an easy-to-understand dashboard.

## 2. Related Work

This section highlights existing tools, research, and systems related to YouTube data analytics and identifies how this project adds value.

Several platforms currently offer YouTube performance tracking and analytics. Third-party tools like **Social Blade**, **TubeBuddy**, and **VidIQ** offer additional insights, such as subscriber growth, keyword optimization, and surface-level trends. While useful, these tools lack the flexibility and scalability needed to process large datasets for deeper analysis and real-time insights.

Research in the field of YouTube analytics has focused on audience engagement. Studies have analysed metrics like views, likes, and comments to determine video popularity trends. Sentiment analysis techniques, using Natural Language Processing (NLP), have also been applied to YouTube comments to gauge audience reactions.

In the field of data engineering, cloud-based analytics frameworks like **Google Cloud Platform (GCP)**, **AWS**, and **Azure** have proven effective for building real-time pipelines. Tools like BigQuery and cloud storage solutions allow for large-scale data ingestion, processing, and visualization. While these frameworks are widely used, implementations specific to YouTube analytics remain limited.

This project addresses these gaps by combining real-time data extraction, processing, and visualization into a **custom, scalable pipeline**. By integrating YouTube video performance metrics this system provides actionable insights tailored for content creators, marketing teams.

### 3. Data Source

This project collects data using the **YouTube Data API v3** to analyse the performance of the Sidemen's YouTube channels. The API provides real-time access to key information, such as channel details and video performance metrics, which are crucial for understanding trends and engagement.

#### a. APIs Used

The following YouTube API endpoints are used to gather the data:

##### 1. Channel Statistics

- **Endpoint:** youtube/v3/channels
- **Purpose:** Fetch basic channel information like subscriber count, total views, video count, and description.

##### 2. Uploads Playlist

- **Endpoint:** youtube/v3/playlistItems
- **Purpose:** Retrieve all video IDs from the channel's uploads playlist.

##### 3. Video Details

- **Endpoint:** youtube/v3/videos
- **Purpose:** Collect video-specific metrics such as views, likes, comments, tags, and upload date.

#### b. Data Features

The collected data is divided into **channel-level data** and **video-level data**.

##### 1. Channel-Level Data:

- **Channel Name:** The name of the YouTube channel.
- **Subscriber Count:** Number of people subscribed to the channel.
- **Total Views:** Total views across all videos.
- **Total Videos:** Total number of uploaded videos.
- **Description:** Short description of the channel.
- **Country:** Country associated with the channel.
- **Thumbnail URL:** Link to the channel's logo or thumbnail image.

## 2. Video-Level Data:

- **Video ID:** Unique identifier for each video.
- **Title:** Title of the video.
- **Views:** Total number of views for the video.
- **Likes:** Number of likes.
- **Comments:** Number of comments.
- **Tags:** Keywords or hashtags associated with the video.
- **Description:** Video description.
- **Published Date:** Date and time the video was uploaded.
- **Duration:** The length of the video.

### c. Why This Data?

This data helps us:

- Track how each Sidemen channel is performing over time.
- Understand which videos are getting the most views, likes, and comments.
- Identify trends based on tags, video durations, and upload schedules.
- Compare the performance of Sidemen's channels to find what works best.

The data is saved in **Google Cloud Storage (GCS)** in JSON format, making it easy to access and analyse for reporting and visualization.

## 4. Methodology

The automated pipeline for analysing Sidemen's YouTube data uses Google Cloud services to manage everything smoothly.

### a. Automated Scheduling with Cloud Scheduler

Every Monday, a tool called Cloud Scheduler runs automatically. It triggers a program that fetches fresh data from YouTube for all Sidemen channels.

### b. Fetching Data Using Cloud Functions

A Python program hosted in Cloud Functions connects to YouTube's API to gather data about the channels (like subscribers, views, and video details). Once collected, this data is saved as files in Google Cloud Storage.

### c. Storing Data in Google Cloud Storage

The data files are stored in a Google Cloud Storage bucket. There are separate files for overall channel data and individual video details to keep things organized.

#### d. Starting Data Processing Automatically

When new files arrive in the storage, another program is triggered. This program triggers a pipeline that processes the raw data into a clean data.

#### e. Pipeline for Cleaning and Organizing

Pipelines takes the raw data and improves it:

- It calculates average views per video for each channel.
- For individual videos, it checks for trends, tags, and even figures out if a video is "viral" based on how many views it gets. The cleaned and processed data is then sent to a database for further use.

#### f. Storing Data in BigQuery

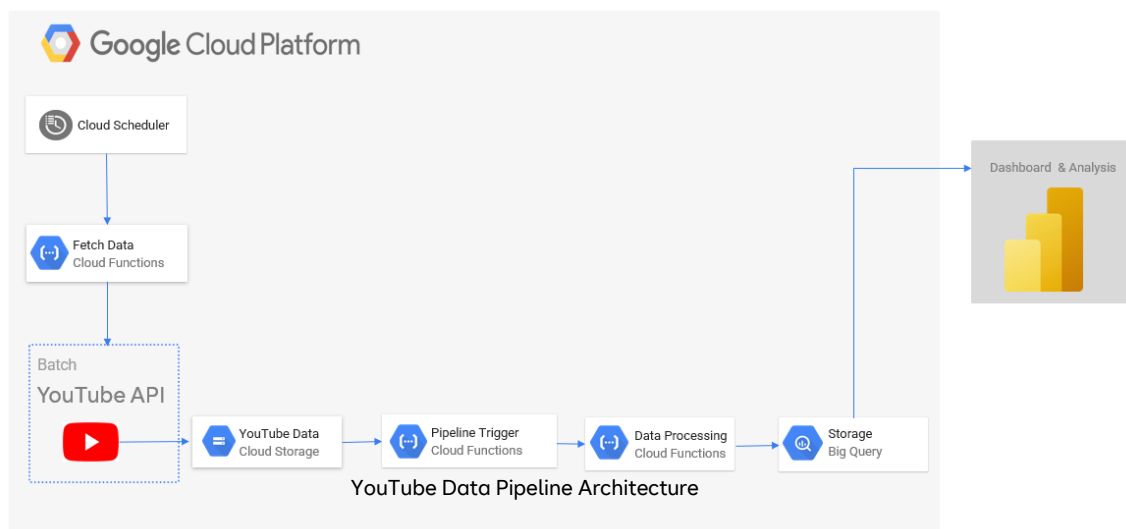
The processed data is stored in BigQuery. It organizes the information in a way that makes it easy to run calculations.

#### g. Creating Dashboards with PowerBi

Finally, the data in BigQuery is connected to PowerBi, a tool for making graphs and dashboards. This step creates visual reports to track Sidemen's subscriber growth, video performance, and engagement with their fans.

## 5. Data Pipeline

The architecture for the Sidemen YouTube data pipeline is designed to collect, process, store, and visualize batch data from YouTube's API. The pipeline involves several components and steps, ensuring an automated flow of data from YouTube to end-user dashboards for insights.



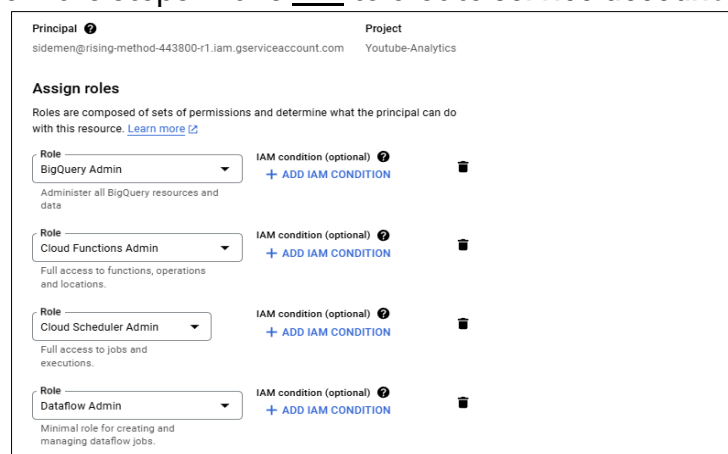
- **Cloud Scheduler** is configured to trigger the fetching of the data every Monday. It ensures that the data-fetching process starts consistently without manual intervention.
- **Google Cloud Functions(Api-Fetch)** executes a Python script when triggered by Cloud Scheduler. This Python script fetches channel-level and video-level data from the **YouTube Data API**.
  - The fetched data includes details like subscriber count, total views, video metrics (likes, comments, tags), and video durations.
  - Once fetched, the data is stored as JSON files in **Google Cloud Storage (GCS)**.
- **Google Cloud Storage (GCS)** acts as the storage layer for raw data. Two types of JSON files are saved:
  - `channel_stats.json` – containing channel-level statistics.
  - `video_analytics.json` – containing detailed video analytics.
- **Cloud Functions(Trigger)** automatically detect when new files arrive in the GCS bucket. Upon detecting these files, **Cloud Function(preprocessing)** starts executing data processing pipeline.
- The data is processed and transformed in two ways
  - **Channel Pipeline:** Processes channel statistics, calculates average views per video, and organizes the channel data.
  - **Video Pipeline:** Processes video-level metrics, calculates engagement rates, identifies viral videos, standardizes video durations, and cleans the tags and descriptions.
- The **processed data** is stored in **Google BigQuery**, a fully managed data warehouse.
  - BigQuery organizes the data into two tables: `channel_stats` and `video_stats`.
- **PowerBI** is used to create interactive dashboards that visualize the processed data.
  - Power BI's DirectQuery mode helps automate dashboard updates, ensuring that the latest data is reflected without manual intervention.
  - The dashboards provide insights into the Sidemen YouTube channels.



## 6. Implementation

### a. Create a Service Account with Required Permissions

1. Go to the **Google Cloud Console**.
2. Click on the **hamburger icon** on the left side.
3. Hover over **IAM & Admin**.
4. Click on **Service Accounts**.
5. Click on **CREATE SERVICE ACCOUNT**.
6. Add the following roles:
  - BigQuery Admin
  - Cloud Functions Admin
  - Cloud Scheduler Admin
  - Storage Admin
7. Click on **Done**.
8. Follow the steps in this [link](#) to create **service account keys**.



The screenshot shows the 'Assign roles' page in the Google Cloud IAM console. At the top, it identifies the 'Principal' as 'sidemen@rising-method-443800-r1.iam.gserviceaccount.com' and the 'Project' as 'Youtube-Analytics'. Below this, a section titled 'Assign roles' explains that roles are composed of sets of permissions. A table lists four roles: 'BigQuery Admin', 'Cloud Functions Admin', 'Cloud Scheduler Admin', and 'Dataflow Admin'. Each role entry includes a description of its permissions, an 'IAM condition (optional)' link, and a trash icon for removal.

Role	Description	IAM condition (optional)	Action
BigQuery Admin	Administer all BigQuery resources and data.	<a href="#">+ ADD IAM CONDITION</a>	
Cloud Functions Admin	Full access to functions, operations and locations.	<a href="#">+ ADD IAM CONDITION</a>	
Cloud Scheduler Admin	Full access to jobs and executions.	<a href="#">+ ADD IAM CONDITION</a>	
Dataflow Admin	Minimal role for creating and managing dataflow jobs.	<a href="#">+ ADD IAM CONDITION</a>	

Service Account creation page

### b. Create a Cloud Storage Bucket

1. Search for **Cloud Storage** in the Google Cloud Console.
2. Click on **CREATE**.
3. Provide a name for the bucket (e.g., sidemen-bucket-west1).
4. Choose a region and create the bucket.

Cloud Storage bucket creation page.

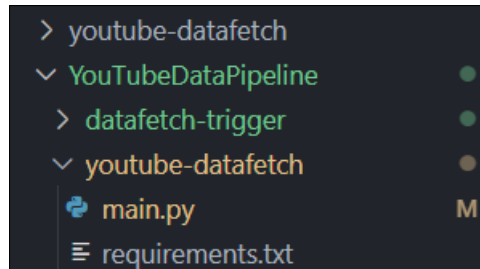
### c. Create Cloud Scheduler

1. Search for **Cloud Scheduler**.
2. Click on **CREATE JOB**.
3. Provide a name, select a region, and set the frequency to `0 0 * * 1` (every Monday).
4. Select time zone and click **Continue**.
5. Set target type as **HTTP**.
6. Provide the **HTTP URL** of the Cloud Function (generated later in step e).
7. Use HTTP Method **GET** or **POST** (depending on your function).
8. Save the job.

Cloud Scheduler setup with frequency

#### d. Prepare Python Code

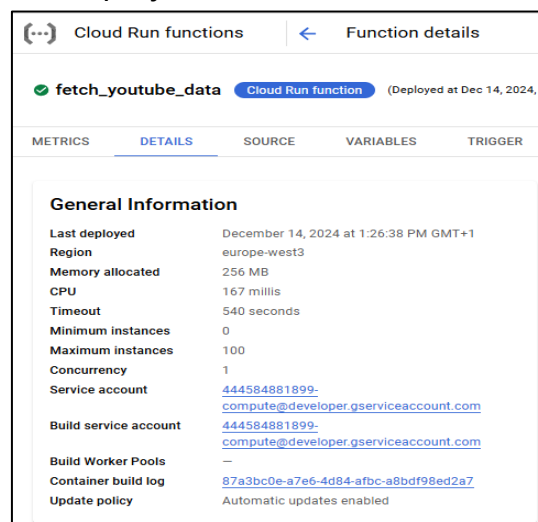
1. Replace the environment variable with the path of the generated service account key.
2. Create a requirements.txt file containing the dependencies
3. Zip the following files:
  - main.py (your Python script for fetching YouTube data).
  - requirements.txt.



Folder structure with main.py, requirements.txt

#### e. Create Cloud Functions

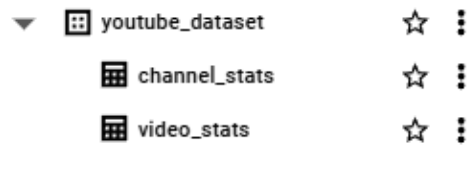
1. Search for **Cloud Functions**.
2. Click on **CREATE FUNCTION**.
3. Provide a name (e.g., fetch-youtube-data) and select a region.
4. Select trigger type as **HTTP**.
5. Click on **MORE OPTIONS** and select the created **service account**.
6. Click **NEXT**.
7. Select runtime: **Python 3.11**.
8. Select source code: **ZIP upload**.
9. Provide the **entry point** as `fetch_youtube_data`.
10. Upload the ZIP file and deploy.



Cloud Function details

## f. Create a BigQuery Dataset

1. Search for **BigQuery** in the Google Cloud Console.
2. Click on the **three dots** next to the project name.
3. Select **Create Dataset**.
4. Provide a name (e.g., youtube\_dataset), select a region, and click **Create**.



BigQuery dataset

## g. Create BigQuery Tables

1. In the **BigQuery console**, select the created dataset.
2. Click on the **three dots** next to the dataset and select **Create Table**.
3. Provide a name for the table:
  - channel\_stats for channel-level data.
  - video\_stats for video-level analytics.
4. Use the **schema** based on your pipeline:
  - Channel Table: channel\_id, title, subscribers, total\_views, etc.
  - Video Table: video\_id, title, views, likes, comments, isviral, etc.
5. Save the tables.

**Current schema**

ADD POLICY TAG

Filter Enter property name or value

<input type="checkbox"/>	Field name	Type	Mode	Collation	Default value	Policy tags	Description
<input type="checkbox"/>	channel_id	STRING	REQUIRED		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	title	STRING	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	subscribers	INTEGER	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	total_views	INTEGER	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	video_count	INTEGER	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	description	STRING	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	country	STRING	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	thumbnail	STRING	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	topics	STRING	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>
<input type="checkbox"/>	avg_views	FLOAT	NULLABLE		<input type="text" value="Default value"/>	-	<input type="text" value="Description"/>

**New fields**

☐ Edit as text

[+ ADD FIELD](#)

[SAVE](#) [CLOSE](#)

Schema of Channel information.

## Current schema

ADD POLICY TAG

Filter

Enter property name or value

<input type="checkbox"/>	Field name	Type	Mode	Collation	Default value	Policy tags	Description
<input type="checkbox"/>	channel_name	STRING	NULLABLE		Default value	-	Description
<input type="checkbox"/>	video_id	STRING	REQUIRED		Default value	-	Description
<input type="checkbox"/>	title	STRING	NULLABLE		Default value	-	Description
<input type="checkbox"/>	views	INTEGER	NULLABLE		Default value	-	Description
<input type="checkbox"/>	likes	INTEGER	NULLABLE		Default value	-	Description
<input type="checkbox"/>	comments	INTEGER	NULLABLE		Default value	-	Description
<input type="checkbox"/>	category_id	STRING	NULLABLE		Default value	-	Description
<input type="checkbox"/>	tags	STRING	NULLABLE		Default value	-	Description
<input type="checkbox"/>	description	STRING	NULLABLE		Default value	-	Description
<input type="checkbox"/>	published_date	DATE	NULLABLE		Default value	-	Description

Rows per page: 10
1 - 10 of 17

## New fields

Edit as text

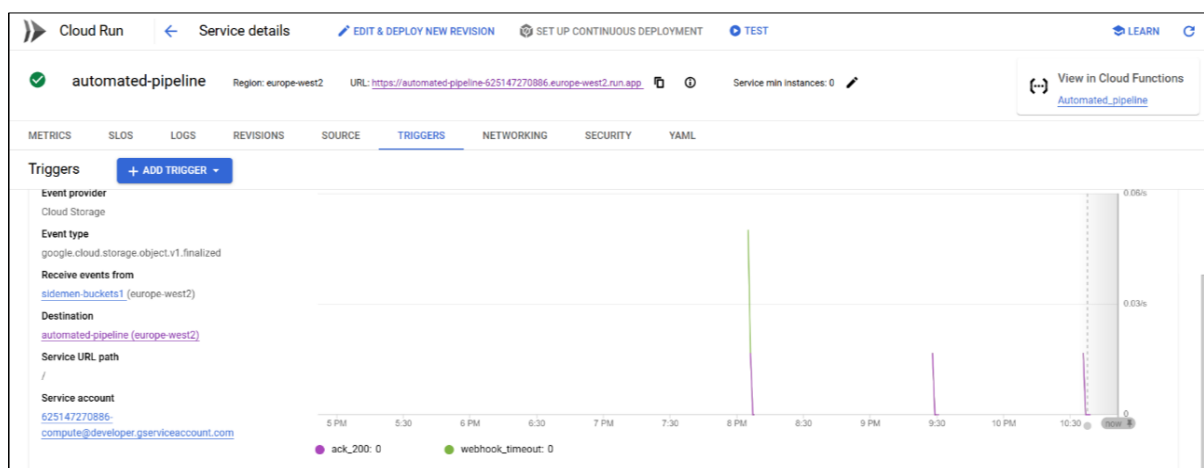
SAVE

CLOSE

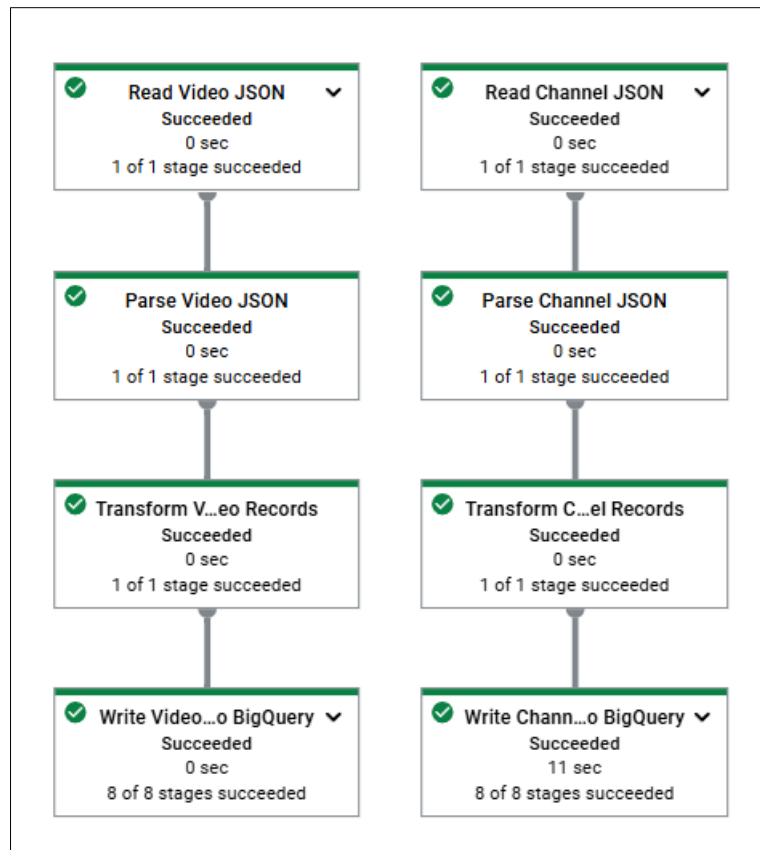
## Example Schema of Video information

## h. Create Cloud Functions to Trigger Pipelines

- Ensure the Cloud Function executes the logic to trigger two pipelines:
  - Channel Pipeline:** Processes channel\_stats.json.
  - Video Pipeline:** Processes video\_analytics.json.



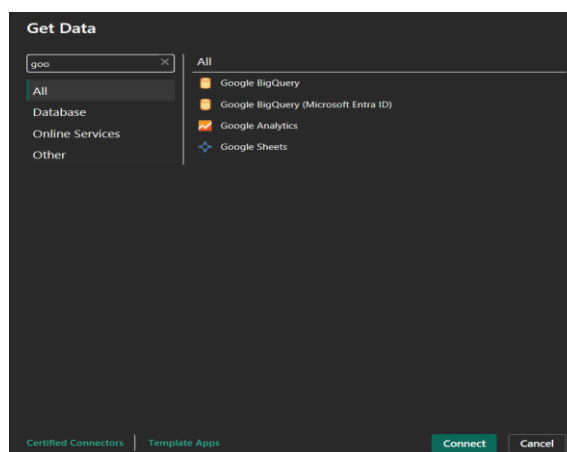
Graph of triggered functions



Triggered pipelines when ran individually

#### i. Connect Power BI to BigQuery in DirectQuery Mode

1. In Power BI Desktop, click **Get Data**.
2. Search for and select **Google BigQuery**.
3. Click **Connect**.
4. After authenticating, browse your BigQuery project
5. Select the desired dataset and table(s).
6. When prompted to choose the connection mode, select **DirectQuery**.
7. Click **Load** to establish the connection.



Connection to Google BigQuery

### Google BigQuery

Advanced options

Billing Project ID (optional)  
youtube-analytics-444717

Use Storage Api (optional)  
Example: true

Connection timeout duration  
Example: 0.00:01:02

Command timeout duration  
Example: 0.00:01:02

Project ID (optional)  
youtube-analytics-444717

SQL statement (optional, requires Project ID)  
Example: sele...

OK Cancel

Setting up same project id as your GCP project id

### Navigator

Display Options

- windy-planet-388515
- wired-sol-438907-s2
- wise-2024
- worker1-411919
- worker1-412215
- worker-1-429113
- woven-nash-384113
- yahoodm2
- yamini-hari-acs-2024
- yashwanth29
- youtube-421607
- youtube-440914
- youtube-analytics-444717 [4]
  - youtube\_dataset
  - youtube\_dataset1
  - youtube\_datasets
  - youtube\_datasets1 [2]
    - ☒ channel\_stats
    - ☐ video\_stats
- youtubedas1

Select Related Tables

#### channel\_stats

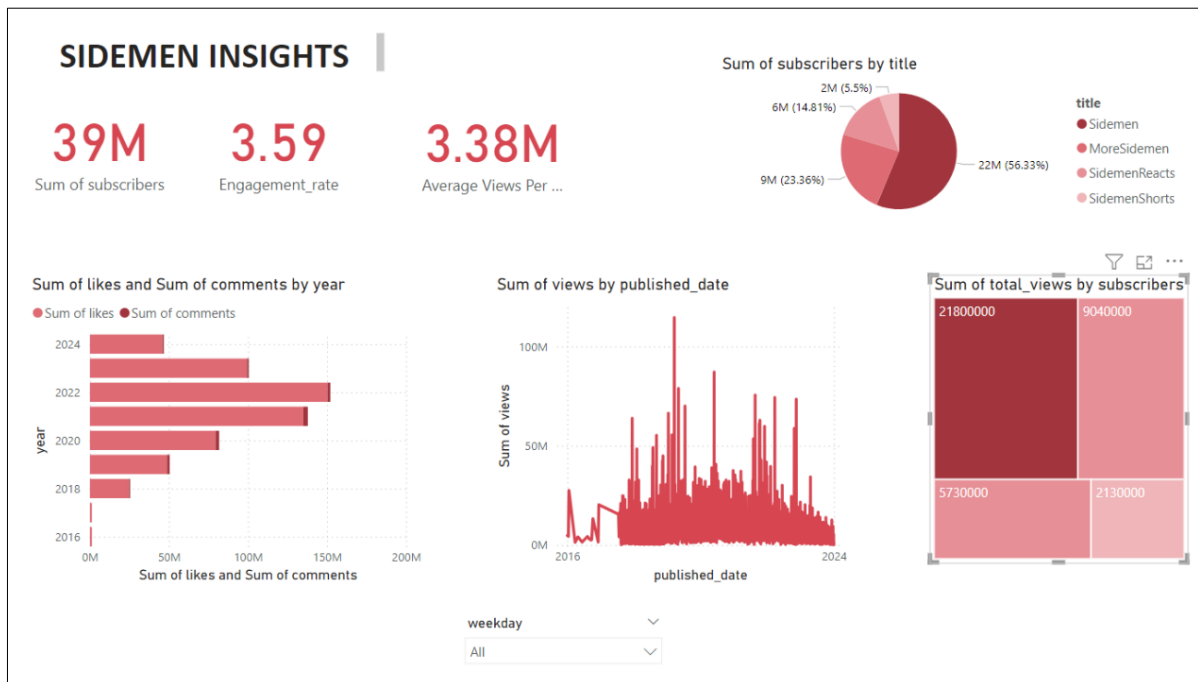
Preview downloaded on Tuesday

channel_id	title	subscribers	total_views	video
UCDogdU7Y7W4hQ9SaFwidMw	Sidemen	2180000	6633928497	
UCJRkTi_Hf4zOh3UFaThgRZw	SidemenReact	5730000	3169929687	
UCh5mLn90vUa81PbRfRy_AtAA	MoreSidemen	9030000	4894555002	
UCbA2H3nTuzylmehmTUhuUs4	SidemenShorts	2130000	1884572929	

Load Transform Data Cancel

Selecting table to be visualized

# Visualisation



## 7. Problems Faced

### 1. Data Fetching and API Usage

While fetching data from the YouTube API, "Too Many Requests" errors were encountered due to API rate limits, especially when handling multiple channels and videos.

**Solution:** Introduced delays between API requests to comply with rate-limiting restrictions. Optimized API usage by limiting data fetches to only required fields.

### 2. Runtime and Timeout Issues

The Cloud Function encountered timeout errors because processing large datasets (e.g., multiple videos and analytics) exceeded the default 60-second limit.

**Solution:** Extended the **Cloud Function timeout limit** to the maximum duration of **540 seconds (9 minutes)**.

### 3. Handling Large JSON Files in Apache Beam

Parsing and processing large JSON files in the pipeline posed challenges in maintaining performance.

**Solution:** Implemented custom pipeline options to efficiently handle file paths and JSON input formats.



#### 4. Connecting PowerBi to BigQuery

Errors occurred during the final stages of connecting PowerBi to BigQuery, especially related to schema mismatches and data compatibility.

**Solution:** Ensured proper datatype compatibility between PowerBi fields and BigQuery schemas.

#### 5. Designing Dashboards

Creating visually appealing and user-friendly dashboards was challenging due to resizing and layout issues in PowerBi.

**Solution:**

- Avoided overloading dashboards with unnecessary visuals

Adjusting to GCP services like Cloud Functions required additional time for debugging and learning. Certain services experienced provisioning delays in specific regions, requiring deployment in alternate regions (e.g., europe-west1)

## 8. Future Scope

Although we successfully achieved our project objectives, there are several opportunities for further enhancement and expansion. The potential areas for future development include:

- **Incorporate Real-Time Data Processing:** Currently, the pipeline operates in batch mode, using BigQuery for aggregations.
- **Integrating Data from Other Platforms:** Expand the project to include data from other social media platforms (e.g., Twitter, Instagram) to perform cross-platform analysis and gain a holistic view of the Sidemen's online presence.
- **Optimize API Usage and Performance:** Improve API request management further by implementing parallel data fetching and advanced error-handling strategies. This will allow the system to fetch large amounts of data faster while maintaining compliance with API rate limits.

## 9. Results and Conclusions

The project successfully automated the process of collecting and analyzing data from Sidemen's YouTube channels. Using tools like **Cloud Scheduler**, **Cloud Functions**, and **BigQuery**, the pipeline fetched data about subscribers, video views, likes, and other key metrics. This data was then cleaned, organized, and visualized in **Power BI dashboards** to provide clear insights into video performance, subscriber growth, and engagement trends.

In simple terms, this project made it easy to track how well Sidemen's videos and channels are doing. The results allow stakeholders to quickly see what works best, helping them make better decisions about future content and strategies.

Here you can view the entire pipeline : [Video link](#)

GitHub Link: <https://github.com/JamadadeHarshita/YouTubeDataPipeline>

## References

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7. Viralyft, "YouTube Engagement," [Online]. Available: <https://viralyft.com/blog/youtube-engagement>. [Accessed: 5-Dec-2024].
8. VidIQ, "YouTube Stats for Sidemen," [Online]. Available: <https://vidiq.com/youtube-stats/channel/UCDogdKI7t7NHZQ95aEwkdMw/>. [Accessed: 4-Dec-2024].

# Tasks Distribution

## Harshita

- 1) Setting up Cloud schedulers
- 2) Fetching YouTube Data Using cloud functions and Uploading in Cloud Storage Bucket
- 3) Designing The Architecture
- 4) Setting up PPT
- 5) Forming Detailed Report [Chapter 1,2,3,4,5,6]

## Navneeth

- 1) Creation of 2 pipelines: Channel and Video pipeline consisting of Preprocessing the Data
- 2) Setting up a trigger function for 2 pipelines.
- 3) Updating the table in BigQuery whenever a change occurs in bucket.
- 4) Formatting Detailed Report [Abstract, Chapter 4,5,6,7]

## Dharshan

- 1) Making BigQuery Schemas
- 2) Connecting to PowerBi for Visualisation
- 3) Writing Queries
- 4) Formatting detailed report[References, Chapter 4,6,8]