Table design

video_data

COLUMN_NAME	DATA_TYPE	PK	NULLABLE	DATA_DEFAULT	COLUMN_ID	COMMENTS
DATETIME	VARCHAR(30)	N	Yes	null	1	Data from
						raw file
VIDEOTITLE	VARCHAR(200)	N	Yes	null	2	Data from
						raw file
EVENTS	VARCHAR(150)	N	Yes	null	3	Data from
						raw file

staging

COLUMN_NAME	DATA_TYPE	PK	NULLABLE	DATA_DEFAULT	COLUMN_ID	COMMENTS
DATETIME	VARCHAR(20)	N	NO	null	1	Data
						reformatted
						from
						video_data.
						DATETIME
PLATFORM	VARCHAR(20)	Ν	NO	null	2	Data derived
						from
						video_data.
						VIDEOTITLE
SITE	VARCHAR(10)	Ν	NO	null	3	Data derived
						from
						video_data.
						VIDEOTITLE
TITLE	VARCHAR(2	Ν	NO	null	4	Data derived
	00)					from
						video_data.
						VIDEOTITLE

FACTVIDEOSTART

COLUMN_NAME	DATA_TYPE	PK	NULLA BLE	DATA_DEF AULT	COLUMN _ID	COMMENTS
DATETIME_SKEY	int	N	NO	null	1	Data derived from DIMDATE. DATETIME_SKEY
PLATFORM_SKEY	int	N	NO	null	2	Data derived from DIMPLATFORM. PLATFORM_SKEY
SITE_SKEY	int	N	NO	null	3	Data derived from DIMSITE. SITE_SKEY
TITLE_SKEY	int	N	NO	null	4	Data derived from DIMTITLE. TITLE_SKEY

dim_time

COLUMN_NAME	DATA_TYPE	PK	NULLA	DATA_DEF	COLUMN	COMMENTS
			BLE	AULT	_ID	
DATETIME	VARCHAR(2	N	NO	null	1	Data reformatted from
	0)					fact_dlt. DataTime
YEAR	smallint	N	NO	null	2	Data reformatted from
						fact_dlt. DataTime
MONTH	smallint	N	NO	null	3	Data reformatted from
						fact_dlt. DataTime
DAY	smallint	N	NO	null	4	Data reformatted from
						fact_dlt. DataTime
HOUR	smallint	N	NO	null	5	Data reformatted from
						fact_dlt. DataTime
MINU	smallint	N	NO	null	6	Data reformatted from
						fact_dlt. DataTime

dim_platform

COLUMN_NAME	DATA_TYPE	PK	NULLA	DATA_DEF	COLUMN	COMMENTS
			BLE	AULT	_ID	
PLATFORM	VARCHAR(2	N	NO	null	1	Data derived from
	0)					fact_dlt. platform

dim_site

COLUMN_NAME	DATA_TYPE	Р	NULLA	DATA_DEF	COLUMN	COMMENTS
		K	BLE	AULT	_ID	
SITE	VARCHAR(10)	Ν	NO	null	1	Data derived from
						fact_dlt. site

dim_title

COLUMN_NAM	DATA_TYPE	Р	NULLA	DATA_DEF	COLUMN	COMMENTS
E		K	BLE	AULT	_ID	
TITLE	VARCHAR(200)	N	NO	null	1	Data derived from
						fact_dlt. tilte

DIMDATE

COLUMN_NAME	DATA_TYPE	PK	NULLA	DATA_DEF	COLUMN	COMMENTS
			BLE	AULT	_ID	
DATETIME_SKEY	int	Υ	NO		1	
DATETIME	VARCHAR(12)	N	NO	null	2	Data derived from
						dim_time. DataTime

DIMPLATFORM

COLUMN_NAME	DATA_TYPE	PK	NULLA	DATA_DEF	COLUMN	COMMENTS
			BLE	AULT	_ID	
PLATFORM_SKEY	int	Υ	NO		1	
PLATFORM	VARCHAR(20)	N	NO	null	2	Data derived from staging. PLATFORM

DIMSITE

COLUMN_NAM	DATA_TYPE	PK	NULLA	DATA_DEF	COLUMN	COMMENTS
E			BLE	AULT	_ID	
SITE_SKEY	int	Υ	NO		1	
SITE	VARCHAR(10)	N	NO	null	2	Data derived from
						staging. SITE

DIMTITLE

COLUMN_NAME	DATA_TYPE	PK	NULLAB	DATA_D	COLUMN	COMMENTS
			LE	EFAULT	_ID	
TITLE_SKEY	int	Υ	NO		1	
TITLE	VARCHAR(200)	N	NO	null	2	Data derived from dim_title. TITLE

Process design

once upload raw data into S3, it will invoke lambda(python script) to extract data and transform data, generate dimension delta and fact delta csv file, store back to S3 and then invokes lambda(python script) to load processed data into database redshift.

Extraction and transformation(detailed code please find relative code)

- 1. Change encoding into UTF-8, otherwise the special characters will show.
- 2. Upload raw data video_data.csv into S3
- 3. Lambda will be invoked (Python script) to do data auditing

```
#data auditing: count total #records
print(f"total recieved records number is {len(raw)}")
```

CloudWatch Log Events:

```
▶ 2020-09-14T00:33:55.481+10:0... total recieved records number is 1333658
```

```
#data auditing: find max length of all field in fact
for col in fact.columns:
   print(f"{col} max len is {max([len(i) for i in fact[col] if i])}")
```

CloudWatch Log Events:

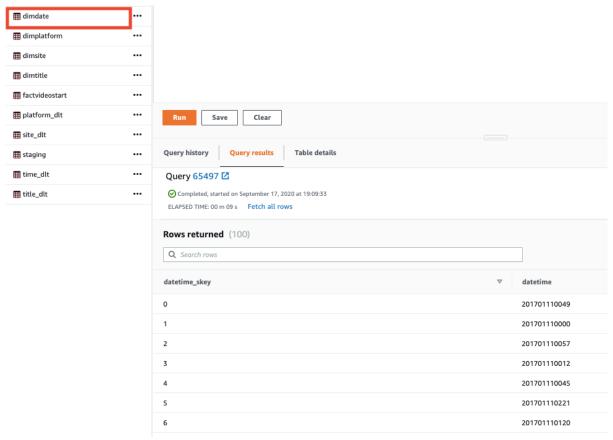
•	2020-09-14T00:56:29.870+10:0	DateTime max len is 16
•	2020-09-14T00:56:30.212+10:0	title max len is 106
•	2020-09-14T00:56:30.551+10:0	platform max len is 7
•	2020-09-14T00:56:30.851+10:0	site max len is 4

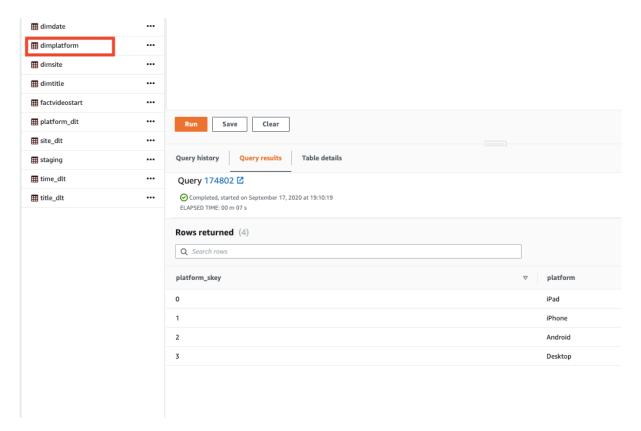
Use the result to adjust the length of columns in table in Redshift. Then wash raw data and load into fact delta and dimension delta tables.

4. The lambda will populate all dimension delta tables and fact delta table as following separately: dim_time.csv, dim_platform.csv, dim_site.csv, dim_title.csv, fact_dlt.csv.

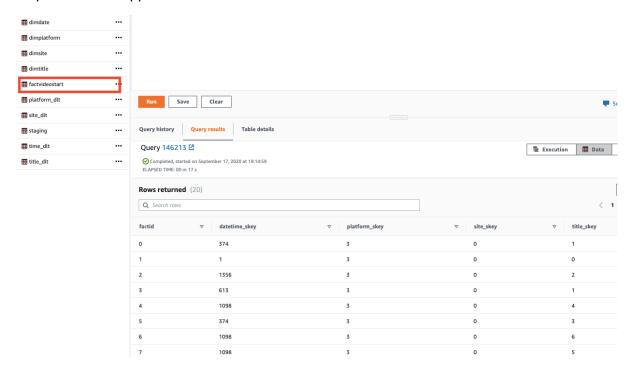
Load data from S3 to Redshift(detailed code please find relative code)

- 1. SQL create dimension tables in Redshift if those dimension tables not exist.
- 2. SQL create dimension delta tables and staging table if not exist in Redshift
- 3. Redshift SQL query to COPY data in processed folder in S3 into delta tables and staging table.
- 4. Insert and generate SKEY for DIM tables DIMDATE, DIMPLATFORM, DIMSITE and DIMTITLE.



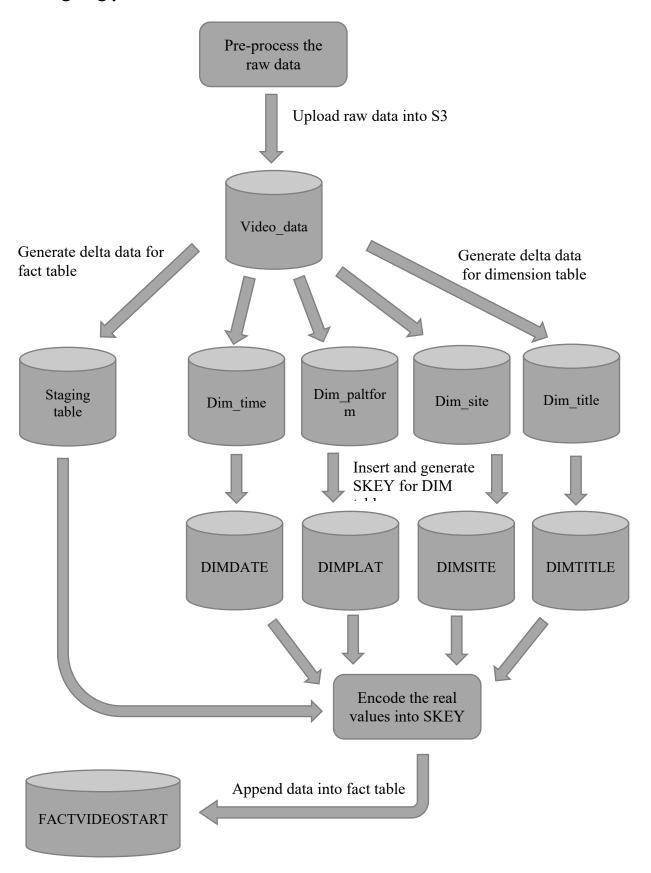


5. Use staging table, DIMDATE, DIMPLATFORM, DIMSITE and DIMTITLE to generate output data and append the data into fact table FACTVIDEOSTART.



6. Truncate staging, time_dlt, site_dlt, platform_dlt, title_dlt.

On-going process



NOTE:

- 1. SKEY stands for surrogate key.
- 2. The current design is Dimension Type One.
- 3. If the source dimension data contains other attributes other than just PK, and we want to track changes of those attributes, eg. sometimes we want to track the product's price changing, then we should use Dimension Type Two.

 One sample of Dimension Type Two

Delta Data from 06/04/2017:

Product_ID	Product_Name	Price	Location
P001	lphone6	750	Townhall Shop
P003	lphone7	1000	Townhall Shop

Data in dimension table:

Product_SKEY	Product_ID	Product_Name	Price	Location	Current_Flag	Start_Date	End_Date
111	P001	lphone6	800	Townhall	Υ	31/12/2016	31/12/9999
				Shop			
112	P002	Iphone6Plus	900	Townhall	Υ	20/01/2017	31/12/9999
		•		Shop			

Add new product (P003) and update product (P001) in dimension table:

7 tad new product (1 000) and apade product ((1 00 1) III dii iioliolioli tabio:			
Product_SKEY	Product_ID	Product_Name	Price	Location	Current_Flag	Start_Date	End_Date
111	P001	Iphone6	800	Townhall Shop	N	31/12/2016	05/04/2017
112	P002	Iphone6Plus	900	Townhall Shop	Υ	20/01/2017	31/12/9999
113	P003	Iphone7	1000	Townhall Shop	Y	06/04/2017	31/12/9999
114	P001	Iphone6	800	Townhall Shop	Y	06/04/2017	31/12/9999

Yellow part is updated, red part is insertion.

When there is a new record coming, we generate new record with new SKEY, Current_Flag='Y',Start_Date = Current_Date, End_Date=31/12/9999
When there is a updated record coming, we generate new record with new SKEY, Current_Flag='Y',Start_Date = Current_Date, End_Date=31/12/9999, and at the same time, we update the old record with Current_Flag='N', End_Date= Current_Date-1.

Therefore, when we populate new records into fact table, we need a filter eg. Current_Flag='Y' in order to get the correct SKEY;

If we want to track the history data for certain days, we use a time filter between Start_Date and End_Date.

For example, if in fact table we see a transaction like customer purchased product(P001) on 01/04/2017, by looking at product dimension table, we could find the price that customer paid at that moment was 800 not 750, although 750 is the current price of P001