Activity Exemplar: Create your target table for Cyclistic

In this activity, you created target tables to consolidate and store the data you pulled from the Cyclistic datasets. These tables will allow you to develop a dashboard using Tableau in the upcoming end-of-course project activities in the next course. As a BI professional, you will need to be able to use programs such as BigQuery and Dataflow to move and analyze data with SQL. This end-of-course project showcases your ability to do just that.

The exemplar you are about to review will help you evaluate whether you completed the activity correctly. In this case, you might have discovered a solution that works just as well as the exemplar. That's great! This exemplar is an example of how a BI professional might have approached this challenge. As long as your process achieved the same results, you can move on to the next phase of the project.

If you find that the result you received is different from the exemplar provided, use the exemplar to iterate and adjust your own code

Exploring the exemplar code

For this activity, you could run the following SQL query to create a summary table for the entire year:

```
1
 2
     TRI.usertype,
 3
      ZIPSTART.zip_code AS zip_code_start,
 4
      ZIPSTARTNAME.borough borough_start,
 5
      ZIPSTARTNAME.neighborhood AS neighborhood start,
 6
       ZIPEND.zip_code AS zip_code_end,
 7
       ZIPENDNAME.borough borough_end,
 8
      ZIPENDNAME.neighborhood AS neighborhood end,
 9
       DATE_ADD(DATE(TRI.starttime), INTERVAL 5 YEAR) AS start_day,
10
       DATE_ADD(DATE(TRI.stoptime), INTERVAL 5 YEAR) AS stop_day,
11
       WEA.temp AS day_mean_temperature, -- Mean temp
12
      WEA.wdsp AS day_mean_wind_speed, -- Mean wind speed
13
       WEA.prcp day_total_precipitation, -- Total precipitation
14
      -- Group trips into 10 minute intervals to reduces the number of rows
       ROUND(CAST(TRI.tripduration / 60 AS INT64), -1) AS trip_minutes,
15
16
       COUNT(TRI.bikeid) AS trip count
17
18
      `bigquery-public-data.new_york_citibike.citibike_trips` AS TRI
19
     INNER JOIN
20
       `bigquery-public-data.geo_us_boundaries.zip_codes` ZIPSTART
21
      ON ST WITHIN(
22
      ST_GEOGPOINT(TRI.start_station_longitude, TRI.start_station_latitude),
23
     ZIPSTART.zip_code_geom)
24
     INNER JOIN
25
       `bigquery-public-data.geo_us_boundaries.zip_codes` ZIPEND
26
       ON ST WITHIN(
27
     ST_GEOGPOINT(TRI.end_station_longitude, TRI.end_station_latitude),
28
     ZIPEND.zip_code_geom)
29
     INNER JOIN
30
     `bigquery-public-data.noaa gsod.gsod20*` AS WEA
       ON PARSE_DATE("%Y%m%d", CONCAT(WEA.year, WEA.mo, WEA.da)) = DATE(TRI.starttime)
31
32
     INNER JOIN
33
       -- Note! Add your zip code table name, enclosed in backticks: `example_table`
34
       `(insert your table name) zipcodes` AS ZIPSTARTNAME
35
       ON ZIPSTART.zip code = CAST(ZIPSTARTNAME.zip AS STRING)
36
     INNER JOIN
37
       -- Note! Add your zipcode table name, enclosed in backticks: `example_table`
38
       `(insert your table name) zipcodes` AS ZIPENDNAME
39
       ON ZIPEND.zip_code = CAST(ZIPENDNAME.zip AS STRING)
40
```

The result of this query is a merged target table that JOINs the public datasets and the zip code table you uploaded.

Additionally, you needed to execute a query that captured data from just the summer season:

```
1    SELECT
2    TRI.usertype,
3    TRI.start_station_longitude,
4    TRI.start_station_latitude,
```

```
5
      TRI.end station longitude,
 6
      TRI.end station latitude,
 7
      ZIPSTART.zip code AS zip code start,
      ZIPSTARTNAME.borough borough_start,
 8
 9
      ZIPSTARTNAME.neighborhood AS neighborhood start,
10
      ZIPEND.zip code AS zip code end,
       ZIPENDNAME.borough borough_end,
11
12
       ZIPENDNAME.neighborhood AS neighborhood end,
      -- Since we're using trips from 2014 and 2015, we will add 5 years to make it look recent
13
14
       DATE_ADD(DATE(TRI.starttime), INTERVAL 5 YEAR) AS start_day,
15
      DATE_ADD(DATE(TRI.stoptime), INTERVAL 5 YEAR) AS stop_day,
16
       WEA.temp AS day_mean_temperature, -- Mean temp
      WEA.wdsp AS day mean wind speed, -- Mean wind speed
17
       WEA.prcp day total precipitation, -- Total precipitation
18
19
       -- We will group trips into 10 minute intervals, which also reduces the number of rows
20
     ROUND(CAST(TRI.tripduration / 60 AS INT64), -1) AS trip minutes,
      TRI.bikeid
21
22
     FROM
23
      `bigguery-public-data.new york citibike.citibike trips` AS TRI
24
25
     `bigquery-public-data.geo_us_boundaries.zip_codes` ZIPSTART
26
     ON ST WITHIN(
     ST GEOGPOINT(TRI.start station longitude, TRI.start station latitude),
27
      ZIPSTART.zip_code_geom)
28
29
     INNER JOIN
30
     `bigquery-public-data.geo_us_boundaries.zip_codes` ZIPEND
31
     ON ST WITHIN(
32
      ST_GEOGPOINT(TRI.end_station_longitude, TRI.end_station_latitude),
33
     ZIPEND.zip_code_geom)
34
     INNER JOIN
35
      -- https://pantheon.corp.google.com/bigquery?p=bigquery-public-data&d=noaa_gsod
      `bigquery-public-data.noaa_gsod.gsod20*` AS WEA
36
37
      ON PARSE DATE("%Y%m%d", CONCAT(WEA.year, WEA.mo, WEA.da)) = DATE(TRI.starttime)
38
     INNER JOIN
39
     -- Note! Add your zipcode table name, enclosed in backticks: `example_table`
     `legalbi.sandbox.zipcodes` AS ZIPSTARTNAME
```

This query results into a similar table as the previous query, except it focuses on trends from July through September.

Key takeaways

Storing data from multiple sources in target tables allows you to access and use consolidated data for reporting purposes. In the Course 3 end-of-course project, you will use the table you've created in this activity to design a dashboard and share insights with the Cyclistic product development team in order to help guide their process and make informed decisions