Troubleshooting and Solving Data Join Pitfalls

GSP412



Google Cloud Self-Paced Labs

Overview

<u>BigQuery</u> is Google's fully managed, NoOps, low cost analytics database. With BigQuery you can query terabytes and terabytes of data without having any infrastructure to manage or needing a database administrator. BigQuery uses SQL and can take advantage of the pay-as-you-go model. BigQuery allows you to focus on analyzing data to find meaningful insights.

Joining data tables can provide meaningful insight into your dataset. However when you join your data, there are common pitfalls that could corrupt your results. This lab focuses on avoiding those pitfalls. Types of joins:

- *Cross join*: combines each row of the first dataset with each row of the second dataset, where every combination is represented in the output.
- *Inner join*: requires that key values exist in both tables for the records to appear in the results table. Records appear in the merge only if there are matches in both tables for the key values.
- Left join: Each row in the left table appears in the results, regardless of whether there are matches in the right table.
- Right join: the reverse of a left join. Each row in the right table appears in the results, regardless of whether there are matches in the left table.
 For more information about joins, see <u>Join Page</u>.
 The dataset you'll use is an <u>ecommerce dataset</u> that has millions of Google Analytics records for the <u>Google Merchandise Store</u> loaded into BigQuery. You have a copy of that dataset for this lab and will explore the available fields and row for insights.
 For syntax information to help you follow and update the queries, see <u>Standard SQL Query Syntax</u>.

What you'll do

In this lab, you perform these tasks:

- Use BigQuery to explore a dataset
- Troubleshoot duplicate rows in a dataset
- Create joins between data tables
- Understand each join type

Setup and requirements

Before you click the Start Lab button

Read these instructions. Labs are timed and you cannot pause them. The timer, which starts when you click **Start Lab**, shows how long Google Cloud resources will be made available to you.

This Qwiklabs hands-on lab lets you do the lab activities yourself in a real cloud environment, not in a simulation or demo environment. It does so by giving you new, temporary credentials that you use to sign in and access Google Cloud for the duration of the lab.

What you need

To complete this lab, you need:

- Access to a standard internet browser (Chrome browser recommended).
- Time to complete the lab.

Note: If you already have your own personal Google Cloud account or project, do not use it for this lab.

Note: If you are using a Pixelbook, open an Incognito window to run this lab.

How to start your lab and sign in to the Google Cloud Console

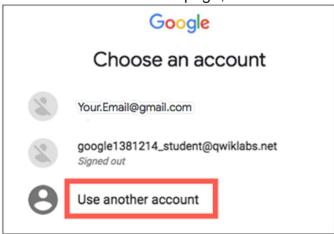
1. Click the **Start Lab** button. If you need to pay for the lab, a pop-up opens for you to select your payment method. On the left is a panel populated with the temporary credentials that you must use for this lab.



2. Copy the username, and then click **Open Google Console**. The lab spins up resources, and then opens another tab that shows the **Sign in** page.

Tip: Open the tabs in separate windows, side-by-side.

If you see the Choose an account page, click Use Another



Account.

3. In the **Sign in** page, paste the username that you copied from the Connection Details panel. Then copy and paste the password.

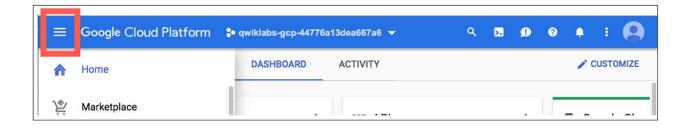
Important: You must use the credentials from the Connection Details panel. Do not use your Qwiklabs credentials. If you have your own Google Cloud account, do not use it for this lab (avoids incurring charges).

- 4. Click through the subsequent pages:
 - · Accept the terms and conditions.
 - Do not add recovery options or two-factor authentication (because this is a temporary account).
 - Do not sign up for free trials.

After a few moments, the Cloud Console opens in this tab.

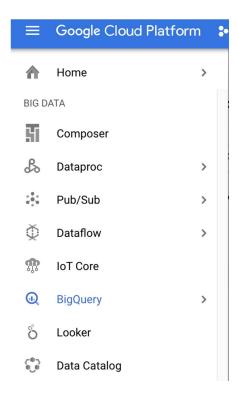
Note: You can view the menu with a list of Google Cloud Products and Services by clicking the **Navigation menu** at the top-

left.



Open BigQuery Console

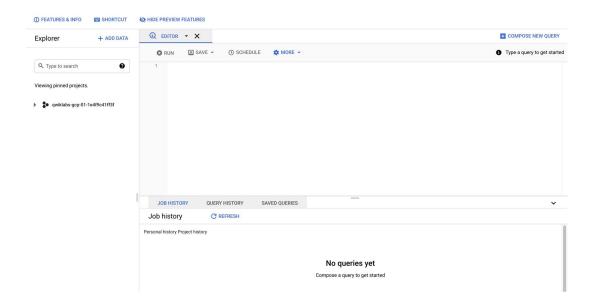
In the Google Cloud Console, select **Navigation menu > BigQuery**:



The **Welcome to BigQuery in the Cloud Console** message box opens. This message box provides a link to the quickstart guide and the release notes.

Click Done.

The BigQuery console opens.

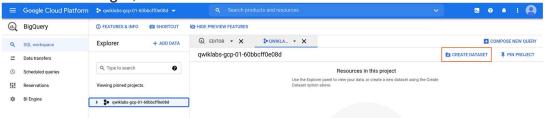


Create a new dataset to store your tables

In your BigQuery project, create a new dataset titled ecommerce.

 In the left pane in the Resources section, click your project (this starts with qwicklabs-gcp-xxx).

2. On the right, click **CREATE DATASET**.



The **Create dataset** dialog opens.

3. Set the *dataset ID* to ecommerce. Leave the other options at their default values, and click **Create dataset**.

In the left pane, you see an ecommerce table listed under your project.

Click **Check my progress** to verify the objective.

Create a new dataset
Check my progress

Pin the Lab Project in BigQuery

Scenario: Your team provides you with a new dataset on the inventory stock levels for each of your products for sale on your ecommerce website. You want to become familiar with the products on the website and the fields you could use to potentially join on to other datasets.

The project with the new dataset is **data-to-insights**.

BigQuery public datasets are not displayed by default in the BigQuery web UI. Since **data-to-insights** is a public dataset project, you have to pin it to your Resource Tree:

- Click on HIDE PREVIEW FEATURES. In a new browser window, open the public datasets project, https://console.cloud.google.com/bigquery?p=data-to-insights&page=ecommerce.
- 2. In the left pane, in the Resource section, click **data-to-insights**. In the right pane, click **Pin Project**.



- 3. Close this browser window.
- 4. Return to and refresh the first BigQuery browser window to refresh the BigQuery web UI. Click on **SHOW PREVIEW FEATURES**.

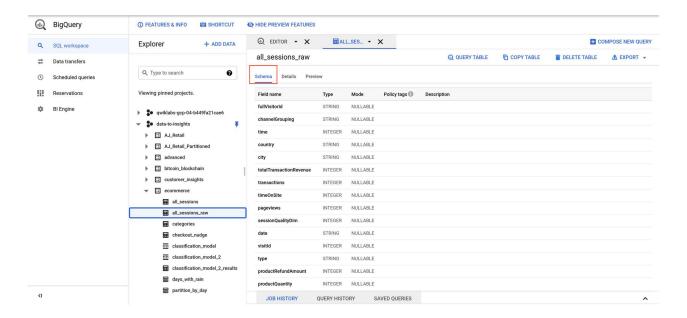
The data-to-insights project is listed in the Explorer section.

Examine the fields

You want to become familiar with the products on the website and the fields you could use to create queries to analyze the dataset.

In the left pane in the Resources section, navigate to data-to-insights > ecommerce > all sessions raw.

On the right, under the Query editor, click the **Schema** tab to see the Fields and information about each field.



Identify a key field in your ecommerce dataset

Examine the products and fields further. You want to become familiar with the products on the website and the fields you could use to potentially join on to other datasets.

Examine the Records

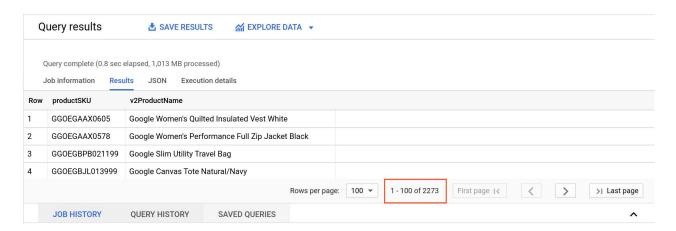
In this section you find how many product names and product SKUs are on your website and whether either one of those fields is unique.

Find how many product names and product SKUs are on the website. **Copy and Paste** the below query in bigquery **EDITOR**.

```
#standardSQL
# how many products are on the website?
SELECT DISTINCT
productSKU,
v2ProductName
FROM `data-to-insights.ecommerce.all_sessions_raw`
```

Click Run.

Look at the pagination results in the console for the total number of records returned.



How many rows of product data are returned?

1,925 products and SKUs

2,205 products and SKUs
check2,273 products and SKUs
Submit

But...do the results mean that there are that many unique product SKUs? One of the first queries you will run as a data analyst is looking at the uniqueness of your data values.

Clear the previous query and run the below query to list the number of distinct SKUs are listed using <code>DISTINCT</code>.

```
# find the count of unique SKUs
SELECT
DISTINCT
productSKU
FROM data-to-insights.ecommerce.all sessions raw`
How many DISTINCT SKUs are returned?
C
119 distinct SKUs
C
2,273 distinct SKUs
check1,909 distinct SKUs
Submit
We have fewer DISTINCT SKUs than the SKU & Product Name query we had before. Why do you think that is?
checkOur first query also returned Product Name. It appears multiple Product Names can have the same SKU.
C
Our first query was excluding some Product Names.
C
Our first query showed that only one Product Name can belong to a SKU.
Submit
```

Examine relationship between SKU & Name

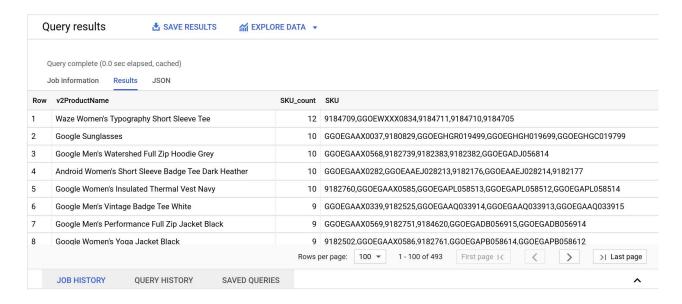
Let's determine which products have more than one SKU and which SKUs have more than one Product Name.

Clear the previous query and run the below query to determine if some product names have more than one SKU. Notice we use the STRING_AGG() function to aggregate all the product SKUs that are associated with one product name into comma separated values.

```
SELECT
   v2ProductName,
   COUNT(DISTINCT productSKU) AS SKU_count,
   STRING_AGG(DISTINCT productSKU LIMIT 5) AS SKU
FROM `data-to-insights.ecommerce.all_sessions_raw`
   WHERE productSKU IS NOT NULL
   GROUP BY v2ProductName
   HAVING SKU_count > 1
   ORDER BY SKU_count_DESC
```

Click Run.

Results:



Do some product names have more than one SKU? Look at the query results to confirm check Yes

O

No

Submit

Which product has the most SKUs associated? checkWaze Womens Typography Short Sleeve Tee

C

Android Womens Short Sleeve Badge Tee Dark Heather

C

Google Sunglasses

Submit

You can see on the <u>ecommerce website catalog</u> that each product name may have multiple options (size, color) -- which are sold as separate SKUs. So we have seen that 1 Product can have 12 SKUs. What about 1 SKU? Should it be allowed to belong to more than 1 product?

Clear the previous query and run the below query to find out.

```
SELECT

productSKU,

COUNT(DISTINCT v2ProductName) AS product_count,

STRING_AGG(DISTINCT v2ProductName LIMIT 5) AS product_name

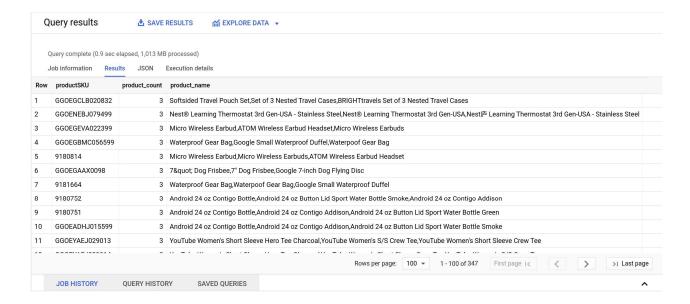
FROM `data-to-insights.ecommerce.all_sessions_raw`

WHERE v2ProductName IS NOT NULL

GROUP BY productSKU

HAVING product_count > 1

ORDER BY product_count DESC
```



Note: Try replacing STRING_AGG() with ARRAY_AGG() instead. Pretty cool, right? BigQuery natively supports nested array values. You can learn more <u>here</u>.

When you look at the query results, are there single SKU values with more than one product name associated? What do you notice about those product names?

O

No, the Product SKUs match the Product Names one-for-one. checkYes, most of the product names are similar but not exactly the same. Submit

You will see why this many-to-many data relationship will be an issue in the next section.

Click **Check my progress** to verify the objective.

Identify a key field in your ecommerce dataset Check my progress

Pitfall: non-unique key

In inventory tracking, a SKU is designed to uniquely identify one and only one product. For us, it will be the basis of your JOIN condition when you lookup information from other tables. Having a non-unique key can cause serious data issues as you will see.

Write a query to identify all the product names for the SKU 'GGOEGPJC019099'.

Possible Solution:

```
SELECT DISTINCT
  v2ProductName,
 productSKU
FROM `data-to-insights.ecommerce.all sessions raw
WHERE productSKU = 'GGOEGPJC019099'
```

Click Run.

v2ProductName	productSKU
7" Dog Frisbee	GGOEGPJC019099
7" Dog Frisbee	GGOEGPJC019099
Google 7-inch Dog Flying Disc Blue	GGOEGPJC019099

What do you notice about the product names?

close They are exactly the same.

checkThey are mostly the same except for a few characters.

Submit

From the guery results, it looks like there are three different names for the same product. In this example, there is a special character in one name and a slightly different name for another:

Joining website data against your product inventory list

Let's see the impact of joining on a dataset with multiple products for a single SKU. First explore the product inventory dataset (the products table) to see if this SKU is unique there.

Clear the previous query and run the below query.

```
SELECT
  name,
  stockLevel
FROM `data-to-insights.ecommerce.products`
WHERE SKU = 'GGOEGPJC019099'
Is the SKU unique in the product inventory dataset?
```

No, there are duplicate SKUs in the inventory dataset.

```
checkYes, just one record is returned.
Submit
How many dog frisbees do you have in inventory?

C
0
check154
C
10,540
Submit
```

Join pitfall: Unintentional many-to-one SKU relationship

We now have two datasets: one for inventory stock level and the other for our website analytics. Let's JOIN the inventory dataset against your website product names and SKUs so you can have the inventory stock level associated with each product for sale on the website.

Clear the previous query and run the below query.

```
SELECT DISTINCT
  website.v2ProductName,
  website.productSKU,
  inventory.stockLevel
FROM `data-to-insights.ecommerce.all sessions raw` AS website
JOIN `data-to-insights.ecommerce.products` AS inventory
  ON website.productSKU = inventory.SKU
   WHERE productSKU = 'GGOEGPJC019099'
What happens when you join the website table and the product inventory table on SKU? Do you now have inventory
stock levels for the product?
closeYes, we have inventory data and everything looks fine.
checkYes, we have inventory levels but the stockLevel is showing three times (one for each record).

Onumber of the product inventory data and everything looks fine.
CheckYes, we have inventory data, the join did not work.
Submit
```

Next, let's expand our previous query to simply SUM the inventory available by product.

Clear the previous query and run the below query.

```
WITH inventory_per_sku AS (
    SELECT DISTINCT
    website.v2ProductName,
    website.productSKU,
    inventory.stockLevel
FROM `data-to-insights.ecommerce.all sessions raw` AS website
    JOIN `data-to-insights.ecommerce.products` AS inventory
     ON website.productSKU = inventory.SKU
    WHERE productSKU = 'GGOEGPJC019099'
)

SELECT
    productSKU,
    SUM(stockLevel) AS total_inventory
FROM inventory_per_sku
GROUP BY productSKU
```

Is the dog Frisbee properly showing a stock level of 154? checkNo, it is now at 462 showing three times (one for each record!) close Yes, it is at 154 Submit

Oh no! It is $154 \times 3 = 462$ or triple counting the inventory! This is called an unintentional cross join (a topic we'll revisit later).

Click **Check my progress** to verify the objective.

Pitfall: non-unique key

Check my progress

Join pitfall solution: use distinct SKUs before joining

What are the options to solve your triple counting dilemma? First you need to only select distinct SKUs from the website before joining on other datasets.

We know that there can be more than one product name (like 7" Dog Frisbee) that can share a single SKU.

Let's gather all the possible names into an array:

```
SELECT

productSKU,

ARRAY_AGG(DISTINCT v2ProductName) AS push_all_names_into_array

FROM `data-to-insights.ecommerce.all_sessions_raw`

WHERE productSKU = 'GGOEGAAX0098'

GROUP BY productSKU
```

Now instead of having a row for every Product Name, we only have a row for each unique SKU.

If you wanted to deduplicate the product names, you could even LIMIT the array like so:

```
SELECT

productSKU,

ARRAY_AGG(DISTINCT v2ProductName LIMIT 1) AS push_all_names_into_array

FROM `data-to-insights.ecommerce.all_sessions_raw`

WHERE productSKU = 'GGOEGAAX0098'

GROUP BY productSKU
```

Join pitfall: Losing data records after a join

Now you're ready to join against your product inventory dataset again.

Clear the previous query and run the below query.

```
#standardSQL
SELECT DISTINCT
website.productSKU
FROM `data-to-insights.ecommerce.all sessions raw` AS website
JOIN `data-to-insights.ecommerce.products` AS inventory
ON website.productSKU = inventory.SKU
How many records were returned? All 1,909 distinct SKUs?
C
Yes, all 1,909 records
checkNo, just 1,090 records
Submit
```

It seems we lost 819 SKUs after joining the datasets Let's investigate by adding more specificity in your fields (one SKU column from each dataset):

Clear the previous query and run the below query.

```
#standardSQL
# pull ID fields from both tables
SELECT DISTINCT
website.productSKU AS website_SKU,
inventory.SKU AS inventory_SKU
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website
JOIN `data-to-insights.ecommerce.products` AS inventory
ON website.productSKU = inventory.SKU
# IDs are present in both tables, how can we dig deeper?
```

It appears the SKUs are present in both of those datasets after the join for these 1,090 records. How can we find the missing records?

Join pitfall solution: Selecting the correct join type and filtering for NULL

The default JOIN type is an INNER JOIN which returns records only if there is a SKU match on both the left and the right tables that are joined.

Rewrite the previous query to use a different join type to include all records from the website table, regardless of whether there is a match on a product inventory SKU record. Join type options: INNER JOIN, LEFT JOIN, RIGHT JOIN, FULL JOIN, CROSS JOIN

Possible Solution:

```
#standardSQL
# the secret is in the JOIN type
# pull ID fields from both tables
SELECT DISTINCT
website.productSKU AS website SKU,
inventory.SKU AS inventory_SKU
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website
LEFT JOIN `data-to-insights.ecommerce.products` AS inventory
ON website.productSKU = inventory.SKU
```

Click Run.

You have successfully used a LEFT JOIN to return all of the original 1,909 website SKUs in your results.

True or False: Many inventory SKU values are NULL. checkTrue

C

False

Submit
How many SKUs are missing from your product inventory set?

Write a query to filter on NULL values from the inventory table.

Possible Solution:

```
#standardSQL
# find product SKUs in website table but not in product inventory table
SELECT DISTINCT
website.productSKU AS website_SKU,
inventory.SKU AS inventory_SKU
```

```
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website

LEFT JOIN `data-to-insights.ecommerce.products` AS inventory

ON website.productSKU = inventory.SKU

WHERE inventory.SKU IS NULL
```

Click Run.

Question: How many products are missing?

Answer: 819 products are missing (SKU IS NULL) from your product inventory dataset.

Clear the previous query and run the below query to confirm using one of the specific SKUs from the website dataset.

```
#standardSQL
# you can even pick one and confirm

SELECT * FROM `data-to-insights.ecommerce.products`
WHERE SKU = 'GGOEGATJ060517'
# query returns zero results

Why might the product inventory dataset be missing SKUs?
checkSome SKUs could be digital products that you dont store in warehouse inventory
closeOld products you sold in past website orders are no longer offered in current inventory
closeThere is legitimate missing data from inventory and should be tracked
closeAll of the above
Submit
```

Now, what about the reverse situation? Are there any products are in the product inventory dataset but missing from the website?

Write a query using a different join type to investigate.

Possible Solution:

```
#standardSQL
# reverse the join
# find records in website but not in inventory
SELECT DISTINCT
website.productSKU AS website_SKU,
inventory.SKU AS inventory_SKU
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website
RIGHT JOIN `data-to-insights.ecommerce.products` AS inventory
ON website.productSKU = inventory.SKU
WHERE website.productSKU IS NULL
```

Click Run.

Answer: Yes. There are two product SKUs missing from the website dataset

Next, add more fields from the product inventory dataset for more details.

Clear the previous query and run the below query.

```
#standardSQL
# what are these products?
# add more fields in the SELECT STATEMENT
SELECT DISTINCT
website.productSKU AS website_SKU,
inventory.*
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website
```

Why would the below products be missing from the ecommerce website dataset?

website_SKU	SKU	name	orderedQuantity	stockLevel	restockingLeadTime	sentimen
null	GGOBJGOWUSG69402	USB wired soundbar - in store only	10	15	2	1.0
null	GGADFBSBKS42347	PC gaming speakers	0	100	1	null

Possible answers:

- One new product (no orders, no sentimentScore) and one product that is "in store only"
- Another is a new product with 0 orders
 Why would the new product not show up on your website dataset?
- The website dataset is past order transactions by customers brand new products which
 have never been sold won't show up in web analytics until they're viewed or purchased.
 You typically will not see RIGHT JOINs in production queries. You would simply just do a LEFT JOIN and switch the
 ordering of the tables.

What if you wanted one query that listed all products missing from either the website or inventory?

Write a query using a different join type.

Possible Solution:

```
#standardSQL
SELECT DISTINCT
website.productSKU AS website_SKU,
inventory.SKU AS inventory_SKU
FROM `data-to-insights.ecommerce.all sessions raw` AS website
FULL JOIN `data-to-insights.ecommerce.products` AS inventory
ON website.productSKU = inventory.SKU
WHERE website.productSKU IS NULL OR inventory.SKU IS NULL
```

Click Run.

You have your 819 + 2 = 821 product SKUs

LEFT JOIN + RIGHT JOIN = FULL JOIN which returns all records from both tables regardless of matching join keys. You then filter out where you have mismatches on either side

Join pitfall: Unintentional Cross Join

Not knowing the relationship between data table keys (1:1, 1:N, N:N) can return unexpected results and also significantly reduce query performance.

The last join type is the CROSS JOIN.

Create a new table with a site-wide discount percent that you want applied across products in the Clearance category.

Clear the previous query and run the below query.

```
#standardSQL
CREATE OR REPLACE TABLE ecommerce.site_wide_promotion AS
SELECT .05 AS discount;
```

In the left pane, site_wide_promotion is now listed in the Resource section under your project and dataset.

Clear the previous query and run the below query to find out how many products are in clearance.

```
SELECT DISTINCT
productSKU,
v2ProductCategory,
discount
FROM `data-to-insights.ecommerce.all sessions raw` AS website
CROSS JOIN ecommerce.site wide promotion
WHERE v2ProductCategory LIKE '%Clearance
How many products are in clearance?
check82
0
0
\Box
52
\Box
91
Submit
```

Note: For a CROSS JOIN you will notice there is no join condition (e.g. ON or USING). The field is simply multiplied against the first dataset or .05 discount across all items. Let's see the impact of unintentionally adding more than one record in the discount table.

Clear the previous query and run the below query to insert two more records into the promotion table.

Next let's view the data values in the promotion table.

Clear the previous query and run the below query.

How many records were returned?

Answer: 3

What happens when you apply the discount again across all 82 clearance products?

Clear the previous query and run the below query.

```
SELECT DISTINCT

productSKU,

v2ProductCategory,

discount

FROM `data-to-insights.ecommerce.all_sessions_raw` AS website

CROSS JOIN ecommerce.site_wide_promotion

WHERE v2ProductCategory LIKE '%Clearance%'
```

How many products are returned?

Answer: Instead of 82, you now have 246 returned which is more records than your original table started with.

Let's investigate the underlying cause by examining one product SKU.

Clear the previous query and run the below query.

```
#standardSQL
SELECT DISTINCT
productSKU,
v2ProductCategory,
discount
FROM `data-to-insights.ecommerce.all_sessions_raw` AS website
CROSS JOIN ecommerce.site_wide_promotion
WHERE v2ProductCategory LIKE '%Clearance%'
AND productSKU = 'GGOEGOLC013299'
```

What was the impact of the CROSS JOIN?

Answer:

Since there are 3 discount codes to cross join on, you are multiplying the original dataset by 3

Note: this behavior isn't limited to cross joins, with a normal join you can unintentionally cross join when the data relationships are many-to-many this can easily result in returning millions or even billions of records unintentionally

The solution is to know your data relationships before you join and don't assume keys are unique.

```
Click all that apply: What are the pitfalls of joining data?

kUnintentional cross join

Permanently deleting the dataset checkCounting some data more than once checkMissing data that should have been included Submit
```

LL JOIN returns what?				
ords from one table minus records in both tables				
ords that are not in both tables				
Records only in both tables checkAll records from both tables Submit				
Click Check my progress to verify the objective.				
Join pitfall solution				
Check my progress				

Congratulations!

You've concluded this lab and worked through some serious SQL join pitfalls by identifying duplicate records and knowing when to use each type of JOIN. Nice work!



Finish your Quest

This self-paced lab is part of the Qwiklabs Quest, <u>BigQuery for Data Warehousing</u> Quest. A Quest is a series of related labs that form a learning path. Completing this Quest earns you the badge above, to recognize your achievement. You can make your badge (or badges) public and link to them in your online resume or social media account. <u>Enroll in this Quest</u> and get immediate completion credit if you've taken this lab. See other available Qwiklabs Quests.

Take your next lab

Continue your Quest with another lab, for example Working with JSON, Arrays, and Structs in BigQuery, or check out these labs:

- Introduction to SQL for BigQuery and Cloud SQL
- Predict Taxi Fare with a BigQuery ML Forecasting Model

Next steps/learn more

Already have a Google Analytics account and want to query your own datasets in BigQuery? Follow this export guide.

Lean more about best practices that provide guidance on <u>Optimizing Query Computation</u>. If you want to practice with more SQL syntax for JOINS, check out the <u>BigQuery JOIN</u> <u>documentation</u>.

Google Cloud Training & Certification

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