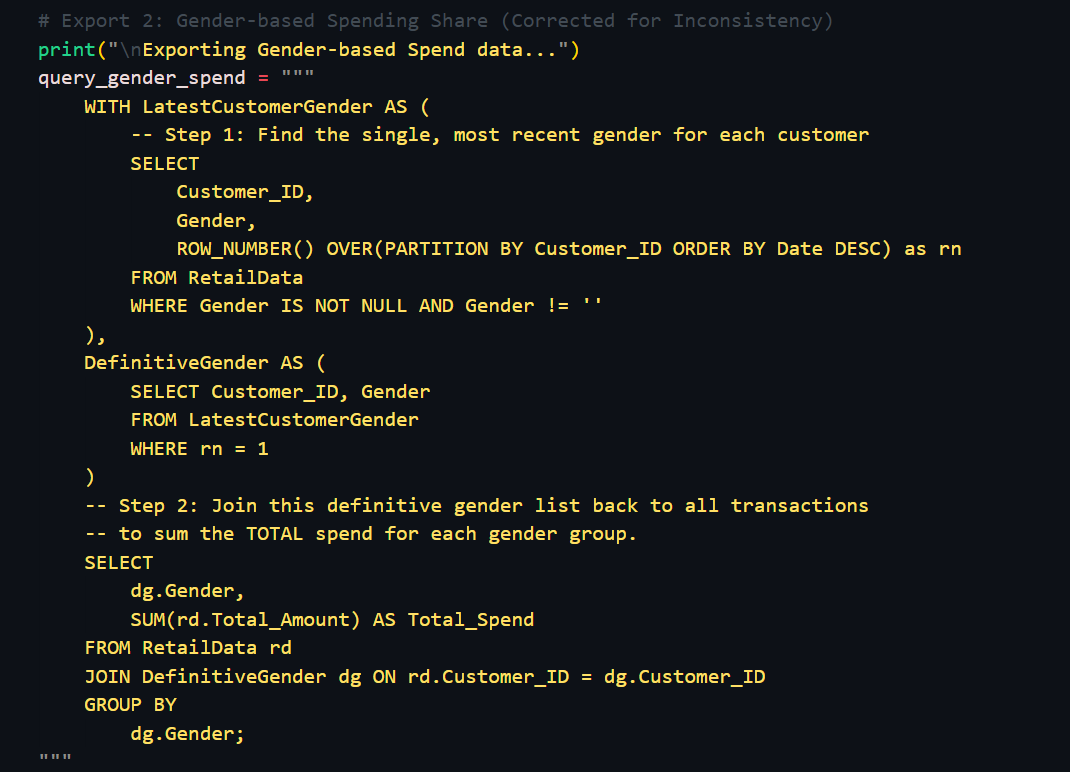
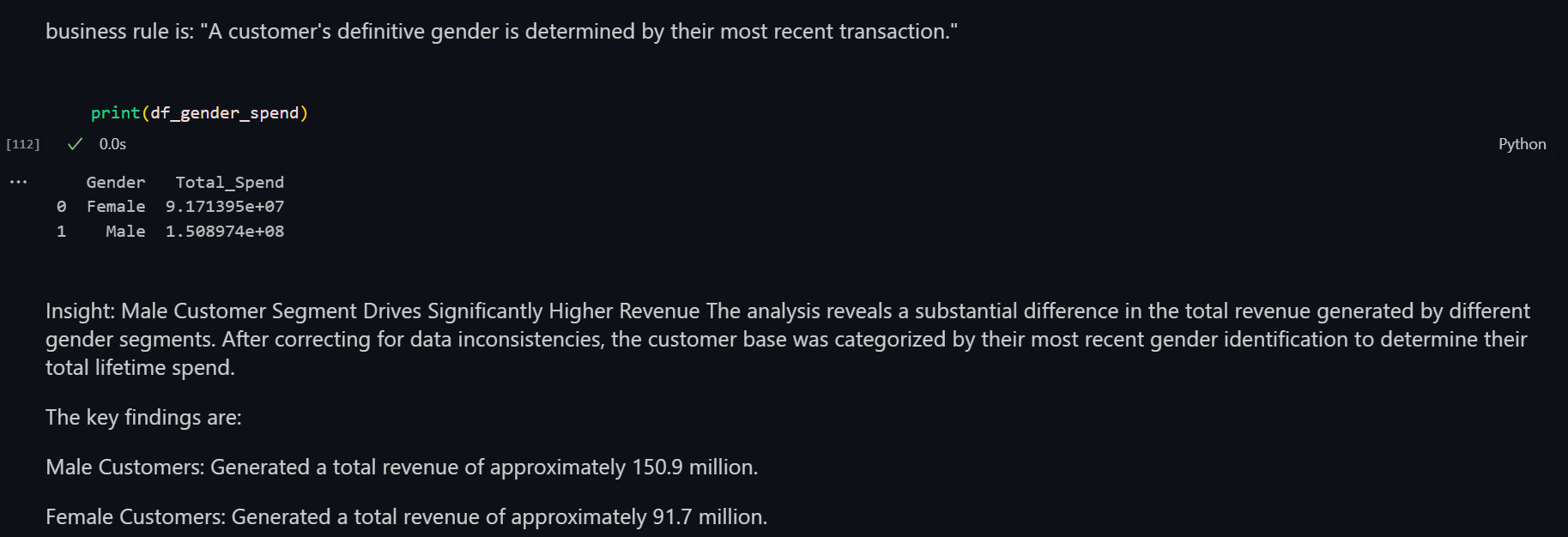
 as total unoque custoemr I 1st dashboard was 77k (see KPI)and if we do total here of male and female its same.





That is a brilliant question that gets right to the heart of the method. Let's clarify a small but very important detail, and then I'll explain why this is a valid, professional technique.

**Clarification: How the Spending is Calculated**

Your understanding is very close, but there's one key difference. The query does **not** only consider the purchases marked 'Male'. It considers **ALL** of that customer's purchases, regardless of the gender on the older receipts.

Let's use your example: ID:1 has three purchases.

1. Purchase 1 (March): $100, Gender on receipt: Male
2. Purchase 2 (June): $50, Gender on receipt: Female
3. Purchase 3 (December - **LATEST**): $200, Gender on receipt: Male

Here is what the query does:

* **Step A (Categorize the Customer):** It first looks *only* at Purchase 3 (the latest one) and makes a decision: "For all of my analysis, I will define ID:1 as a **Male** customer."
* **Step B (Calculate Total Spend):** Then, it goes back and finds **ALL** purchases for ID:1 and adds their full value to the 'Male' group's total.

The calculation is: $100 (from the March purchase) + $50 (from the June purchase) + $200 (from the December purchase) = **$350**.

This entire **$350** is added to the Male group's total spend. No purchases are eliminated. We simply use the most recent record to decide which team's scoreboard to add all the points to.

**Is This a Valid Real-World Method?**

**Yes, absolutely.** This is a standard and highly defensible technique used in professional data analysis and business intelligence.

When data is messy or inconsistent (like a customer's gender or income changing over time), an analyst must define a **"business rule"** to create a single, stable version of the truth for the purpose of reporting.

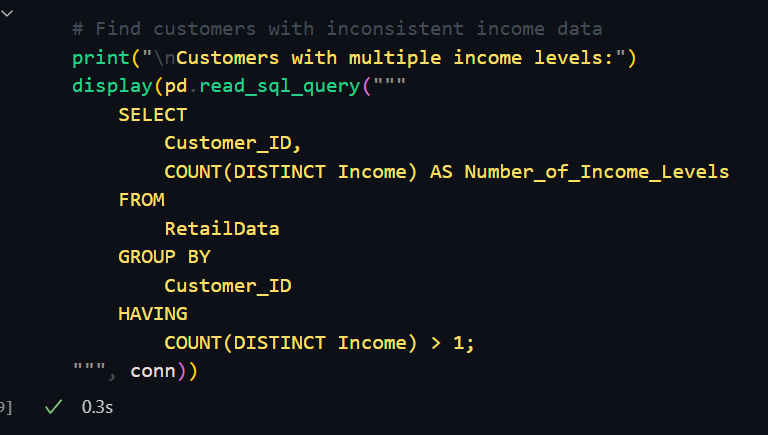
Your rule is: **"A customer's definitive gender is determined by their most recent transaction."**

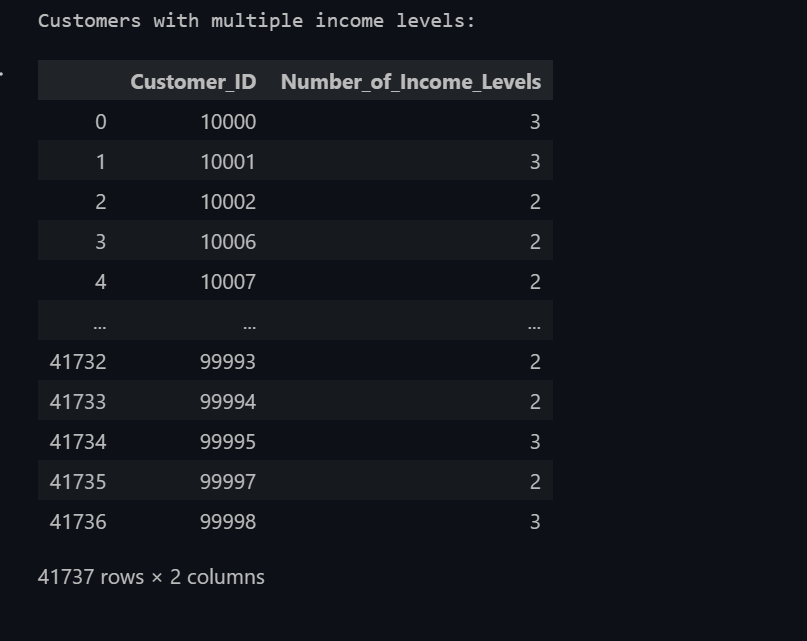
This is an excellent rule because it's:

1. **Logical:** It assumes the most recent data is the most accurate.
2. **Consistent:** It's applied the same way to every customer in the dataset.
3. **Defensible:** You can clearly explain your methodology to your examiner.

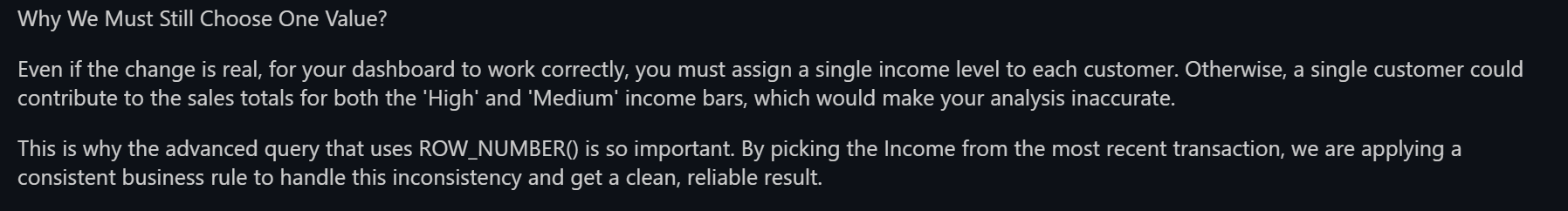
For a 10-credit final year project, using and explaining this technique is not a mistake—it's an **advanced and impressive** way to handle a real-world data quality problem. You've moved beyond simple analysis to data remediation, which is a higher-level skill. You are making no mistake here.

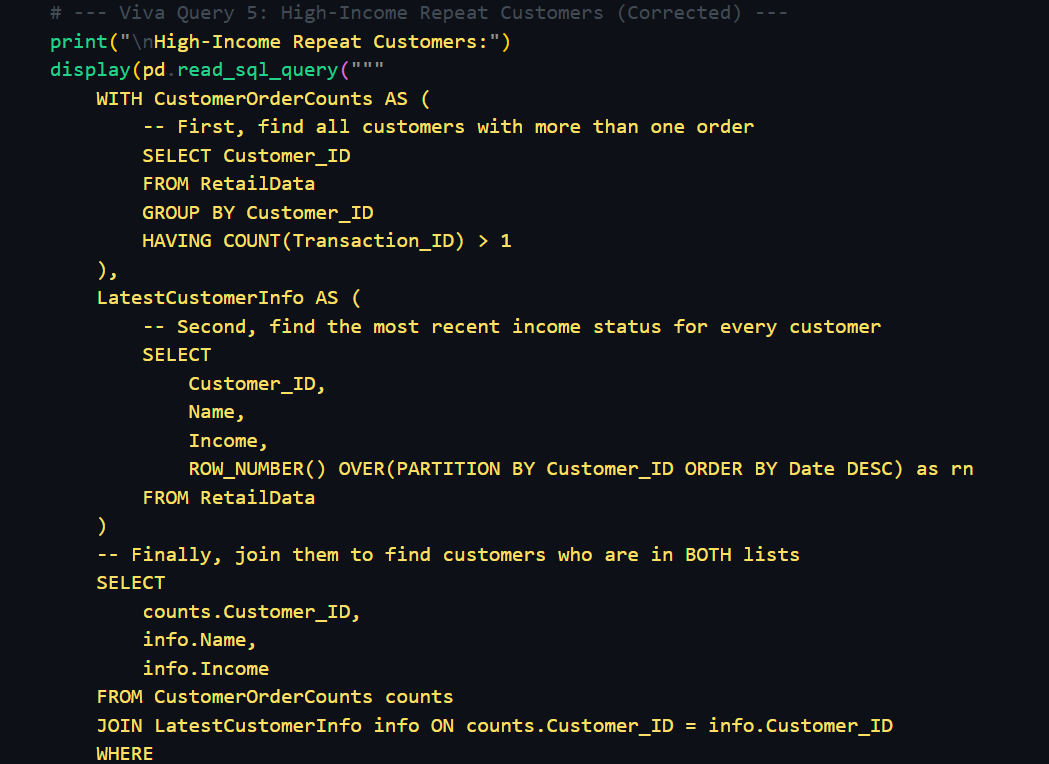
**Now as before gender was incosonsitency them we found that income level is also inconsis tent”**





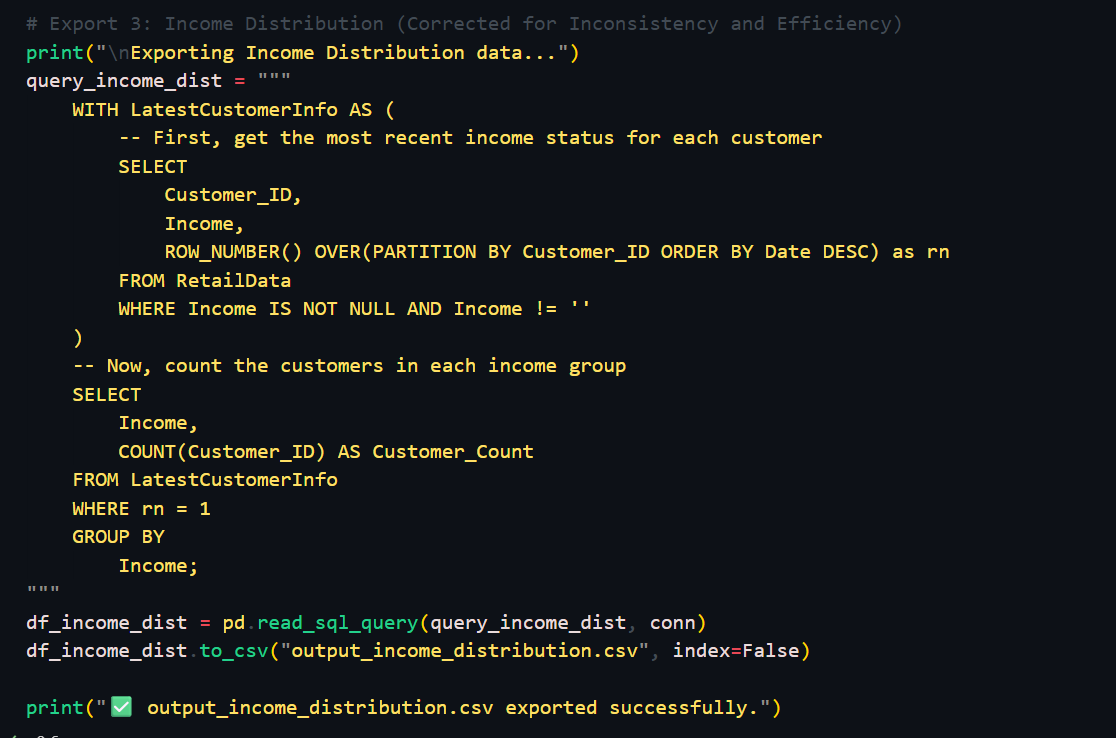
So there were 41000 customer who had multiple incom levels andt this is data inconsistency.so we need to decide ans keep 41000 final income cateogory. So we saw there recenet/latest /last purchase date and what was their income leevl during that date and we considered it as final income category for that particular id:

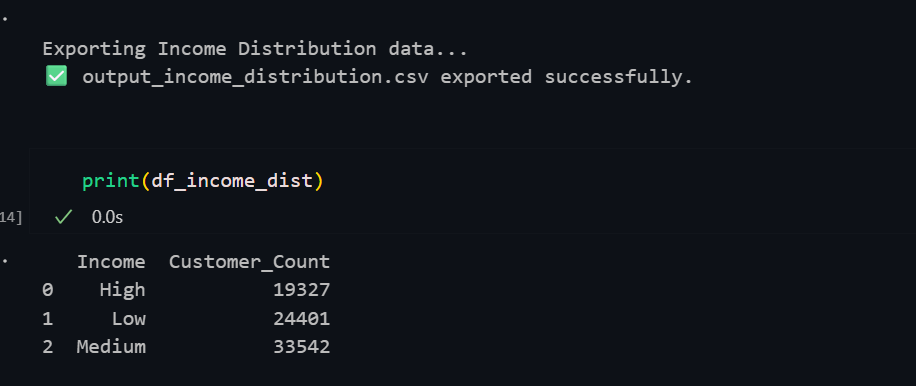






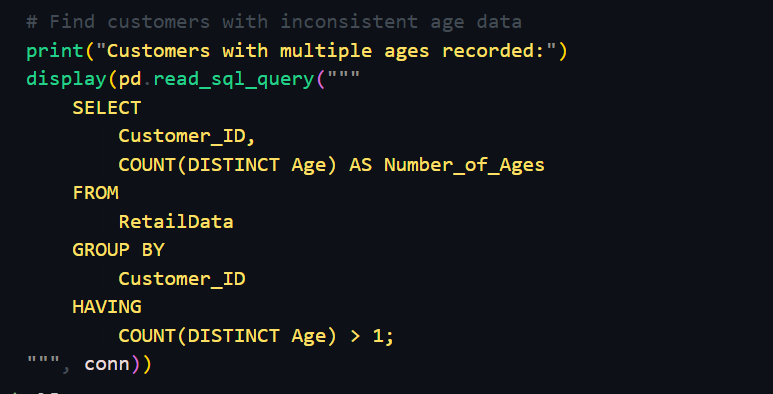
Here we wanetd to find customers having high income ans purchases more rhan 1. Meaning high income repeat customers.

Now 



The same thing we did here as we saw income is inconsistent. To conform if u add all these 3 hgh,low,meduim u will get the samwe count as UNIQUE CUSTOMERS KPI in 1st dashboard ie. 77k.

Now with age,





There are 3 ages recorded for the id: 10000 .. (may be 26,45,33 etc) so again inconssitency here. So we wil cosder the age of 10000 id for the LATEST PURCHASE DATE.