Catching Fraud

1. Analysis of the query

```
WITH processed users -- Creation of a subtable
    AS (SELECT LEFT(u.phone_country, 2) AS short_phone_country, --Selection
of the first two characters of the phone number
              u.id
       FROM users u)
SELECT t.user_id,
      t.merchant country,
       Sum(t.amount / fx.rate / Power(10, cd.exponent)) AS amount -- Convert
amount to Euro
FROM transactions t
      JOIN fx rates fx
       ON (fx.ccy = t.currency
              AND fx.base ccy = 'EUR' ) -- Get the exchange rate info to
convert all amounts in Euros
     JOIN currency details cd
        ON cd.currency = t.currency
      JOIN processed users pu
       ON pu.id = t.user id
WHERE t.source = 'GAIA'
        AND pu short_phone_country = t merchant_country -- Filter to get
transaction in the country of the phone of the user
GROUP BY t.user id,
       t.merchant country
ORDER BY amount DESC;
```

The goal of this query is to get the amount of money spent, in Euros, by users in the country of their phone. It sorts amounts from the highest to the lowest, filtering only expenses coming from source GAIA.

The query itself is working as it is correctly written. But it give us no results as the condition:

```
AND pu.short_phone_country = t.merchant_country
```

doesn't send us anything as short_phone_country has only two characters, while merchant country has three.

We fix the query:

```
ON cd.currency = t.currency

JOIN processed_users pu
ON pu.id = t.user_id

WHERE t.source = 'GAIA'

AND pu.short_phone_country = LEFT(t.merchant_country, 2)

GROUP BY t.user_id,
t.merchant_country

ORDER BY amount DESC;
```

USER_ID	MERCHANT_COUNTRY	amount
f4f81f33-7ae1-45f3-9011-3ef47ae51d38	HUN	299317598.1484371
c649559f-8f5e-4a3e-901f-46aeeccde74d	HUN	81117252.77073133
74fdc60d-ee12-47c1-85f0-1c7dd1dbbf16	HUN	76878611.20641503
33930839-d0f3-478c-a807-8b5c1de9f5dd	JPN	21755975.95664639
47a0a032-fd77-4161-aee7-51b0f804d4b2	HUN	16681469.761387784
3c1aa14d-818a-474f-847f-3d24907dd1c7	HUN	12865527.401201654
65815942-9d63-42d9-a64d-85f8b3bef819	HUN	9683314.26919845
9ff7ad28-2b96-4db2-9980-62d48d6d7b3a	HUN	8351035.744316829
b7496a8d-ffd1-4f56-864d-2ad25bafc243	HUN	5242486.349094559
dd672c5e-0cab-4a94-9a93-98334a6b915	HUN	5024811.514295287
9faf4b80-4720-49b1-b021-591f8d462591	HUN	4327769.661032245

2. Catching users with first succeed purchase over 10 USD

```
with completed transaction as (
  select t."USER ID"
,t."CREATED DATE"
,(t."AMOUNT" / fx."rate" / Power(10, cd.exponent)) AS amount
FROM transactions t
     JOIN fx rates fx
       ON (fx."ccy" = t."CURRENCY"
            AND fx. "base ccy" = 'USD' )
           JOIN currency_details cd
       ON cd."currency" = t."CURRENCY"
where t."STATE" = 'COMPLETED'
and (t."AMOUNT" / fx."rate" / Power(10, cd.exponent))>10)
first_date as (
  select "USER_ID"
  ,min("CREATED DATE") AS first purchase date
  from completed transaction ct
  GROUP BY "USER ID")
SELECT
fd."USER_ID"
,fd."first purchase date"
,ct."amount"
from first date fd
LEFT JOIN completed_transaction ct on (fd."first_purchase_date"=ct."CREATED_DATE"
and fd. "USER ID"=ct. "USER ID")
ORDER BY ct. "amount";
```

ı			
	USER_ID	first_purchase_date	amount
	330b0904-0271-43c1-80e6-e3395f6d888	2016-02-18	10.000711361205937
	93dd18e3-c631-4d29-bd41-bf67525bc037	2017-09-11	10.000711361205937
	0ad6e671-7346-44c1-8216-6a78eed73a5d	2018-07-17	10.018601078143435
	8b810569-016f-4a08-9094-a1d49068ae4e	2016-03-22	10.038592843634747
	8053054c-a220-49d2-8c80-6de5984eb26	2017-02-07	10.061052777627093
	d8760af4-7e45-4f72-9c6c-0964d0f33157	2018-06-14	10.07647432606356
	01589ac2-05bd-4414-b9fb-f9793e1171c2	2018-06-15	10.078033457420558
	397965b5-03c6-4627-afd8-faac3fb54202	2018-07-13	10.103504477110754
	9818f617-fb92-42f9-9862-e471ac890a17	2017-01-08	10.129508401463893
	714661e3-b09b-42ef-a7b8-14a90bd455f5	2016-08-26	10.15981358740694

3. Fraudster

I didn't have time to realize this exercise but I would apply the following strategy.

- We are missing labelled data of people that are not fraudsters, we only have Fraudster people. We cannot build a Supervised predictive model. If we consider not identified people as non fraudster we could catch false positive and give a wrong information to our sistem
- 2. I would transform the dataset to have a list of features that could describe each users, such as:
 - a. Prefered transaction type
 - b. Average transaction amount
 - c. Amount of first transaction
 - d. Percentage of transaction in the country
- 3. I would use all of these features to build a non supervised clusterization probably using the K-Means method.
- 4. I would see the percentage of Fraudsters in each group. Depending on the result I would be able to identify people with really similar behaviours than our identified fraudsters and probably catch them as Fraudsters.