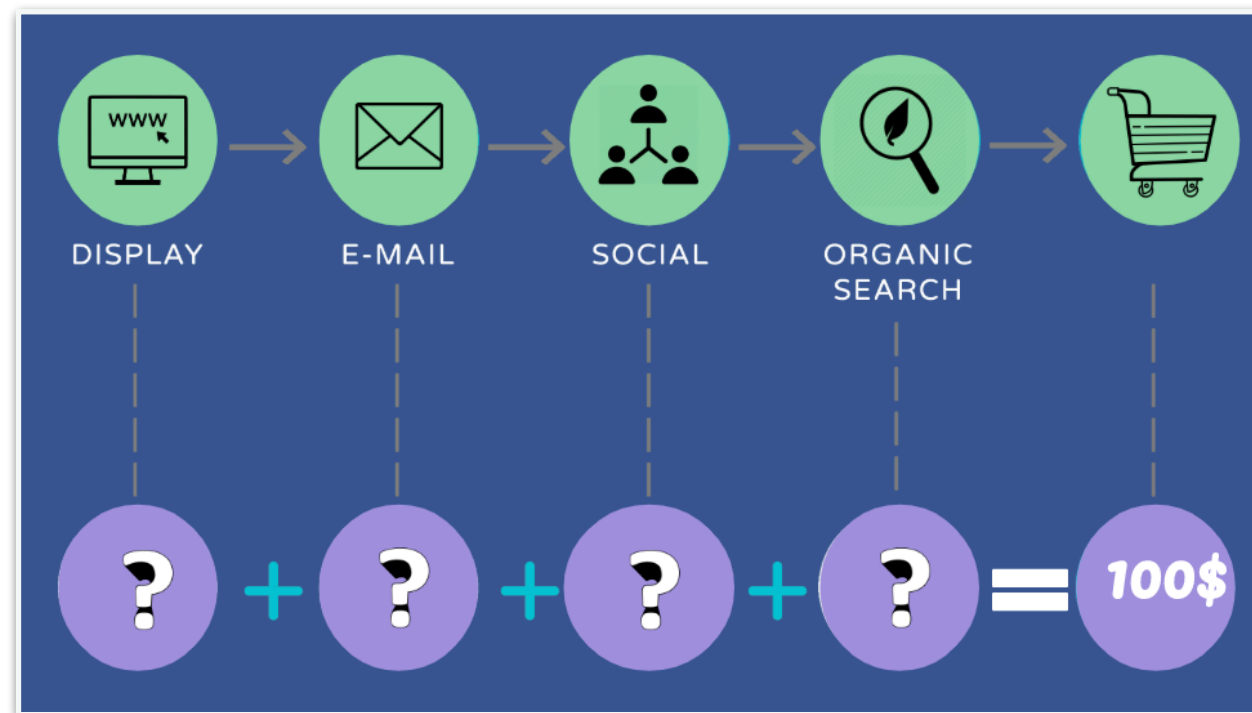


# Probability theory & Marketing Attribution Model

The marketing-attribution problem

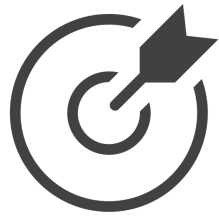
Game theory and Shapley value

# The Marketing-Attribution Problem



The problem: Assign the credit of a purchase (a conversion) to the correct marketing channel(s).

Assumption: We know which channels the customer interacted with prior to their purchase.



## Optimising spending on best channels

- ROI
- Understanding customer journeys
- Strategic investment





# GAME THEORY & SHAPLEY VALUE

DEFINITION



FORMULA

$fx$

APPLICATION





- Game theory: A branch of applied mathematics that provides tools for analysing situations in which parties, called players, make decisions that are interdependent.
- Game theory aims to understand situations where rational decision-makers interact (e.g., take actions, threaten each other and possibly form coalitions)



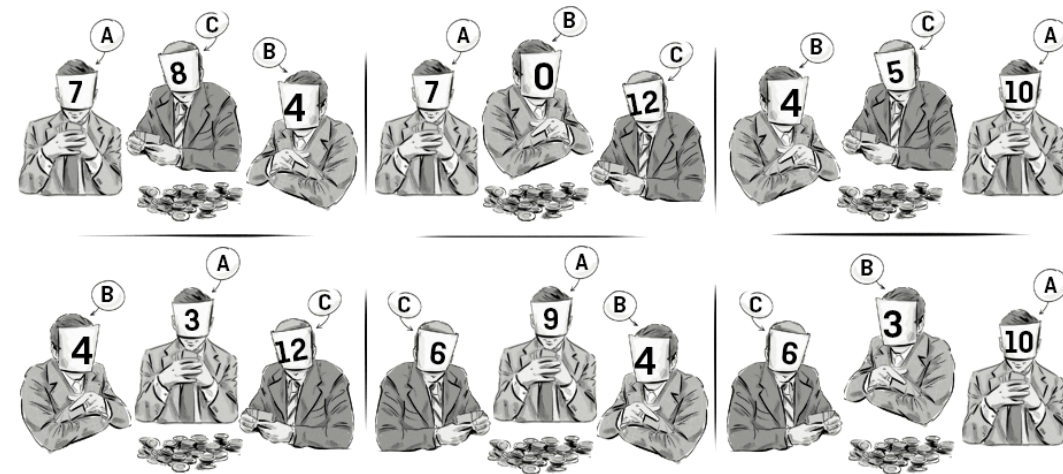
*fx*

- 1st: identify each player's contribution when they play individually, when 2 play together and when 3 play together ...




# $f_x$

- 2nd: calculate average weighted contribution (marginal contribution) of each channel



$$\phi_i(v) = \sum_{S \subseteq N \setminus \{i\}} \frac{|S|! (n - |S| - 1)!}{n!} (v(S \cup \{i\}) - v(S))$$

$$[7+7+10+3+9+10] / 6 =$$




Customer journey = cooperative game  
Each channel = one player  
Together driving conversions

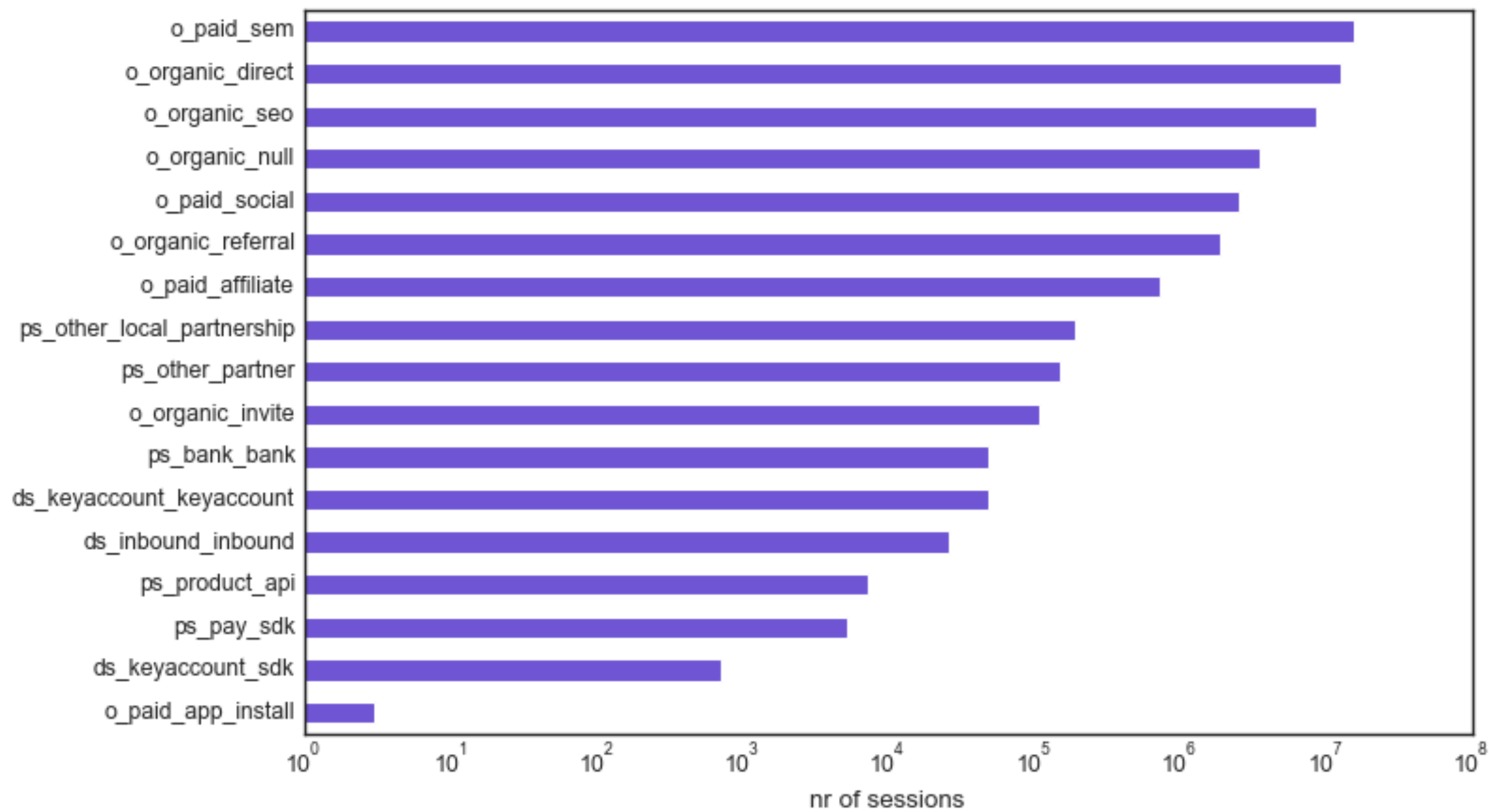
# Part 2: Data challenge

# Attribution in real life

- Last touch point model - one data point in customer journeys
- Internal datasets:
  1. App info (app\_installs)
  2. Own tracking (user\_campaign\_data)
  3. (merchant\_created)
- Solution: Shapley value model - all data points in customer journeys
  - + GA data
  - +...

# GA challenges

1. Cookies dependency
2. Unable to track minimal sign-up via app
3. Unable to track across devices & browsers
4. One visitor\_id: many organization\_uuid
5. One organization\_uuid: many visitor\_id
6. Some cases can't record organization\_uuid at minimal sign-up

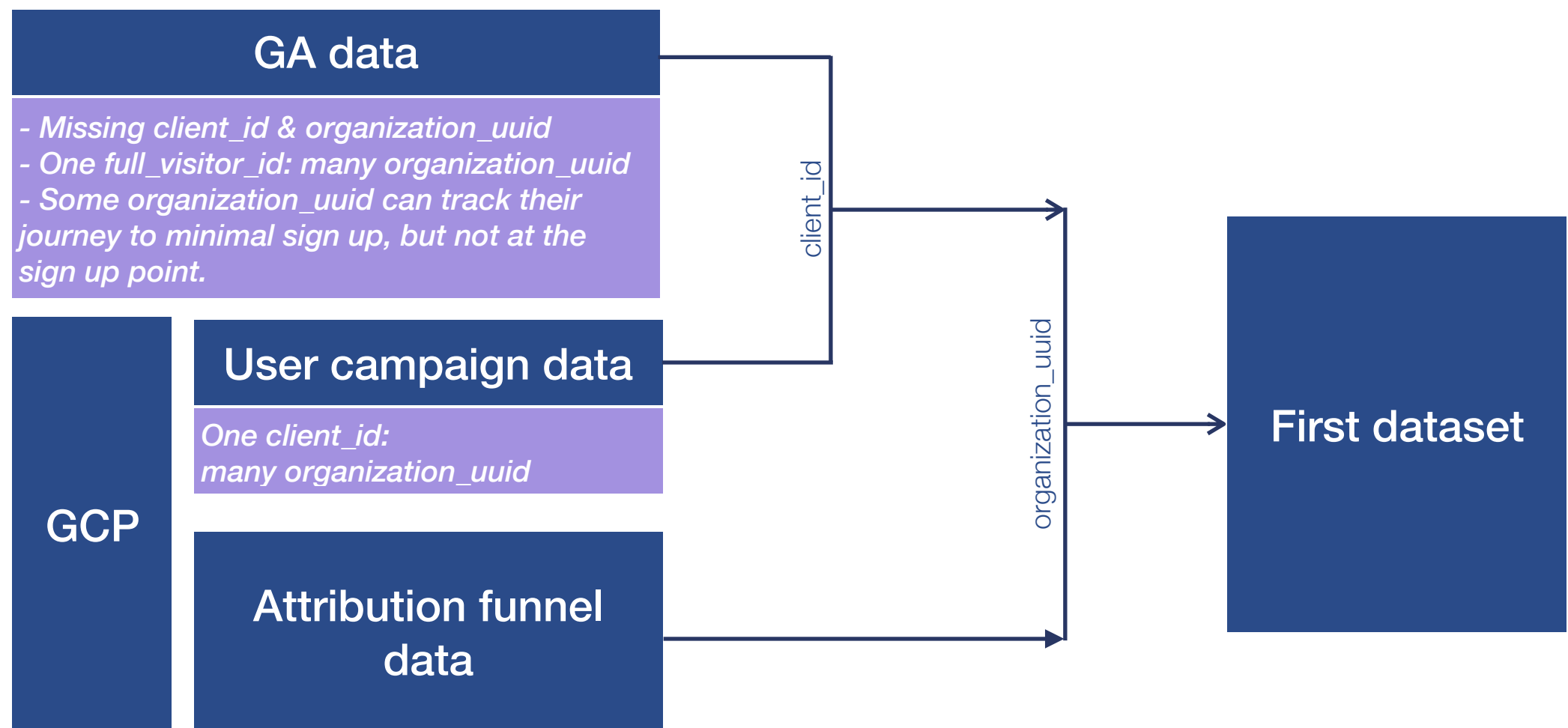


# Create first attribution dataset

- Dataset = GA data + GCP
- GCP = user\_campaign\_data + attribution\_funnel



# First dataset's challenges



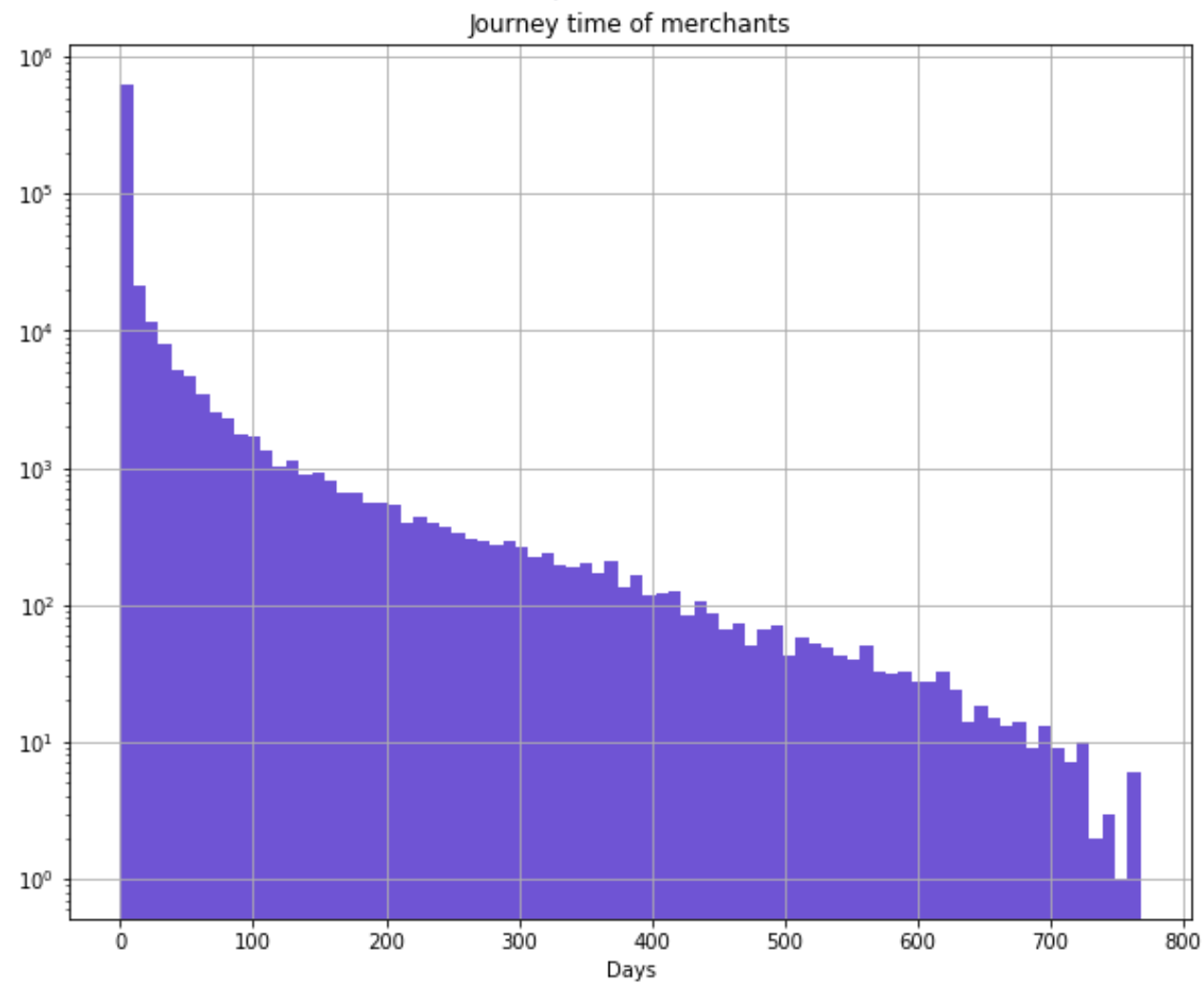
# Customer journey

	journey	nr_merchants	length
0	[o_paid_sem]	278539	1
1	[o_organic_null]	156363	1
2	[o_organic_direct]	142474	1
3	[o_organic_app]	99365	1
4	[o_organic_seo]	92065	1
...	...	...	...
481	[o_organic_app, o_paid_social, partnerships]	1	3
482	[o_organic_direct, o_organic_invite, o_organic...	1	4
483	[direct, o_organic_direct, o_organic_referral,...	1	5
484	[direct, o_paid_affiliate, partnerships]	1	3
577	[o_organic_null, o_organic_referral, o_paid_af...	1	5

- REG  
nr\_session = 1: 76.98%  
nr\_session <= 4: 95.87%  
nr\_session <= 40: 99.99%

- KYC  
nr\_session = 1: 68.44%  
nr\_session <= 4: 93.74%  
nr\_session <= 40: 99.98%

# Journey length



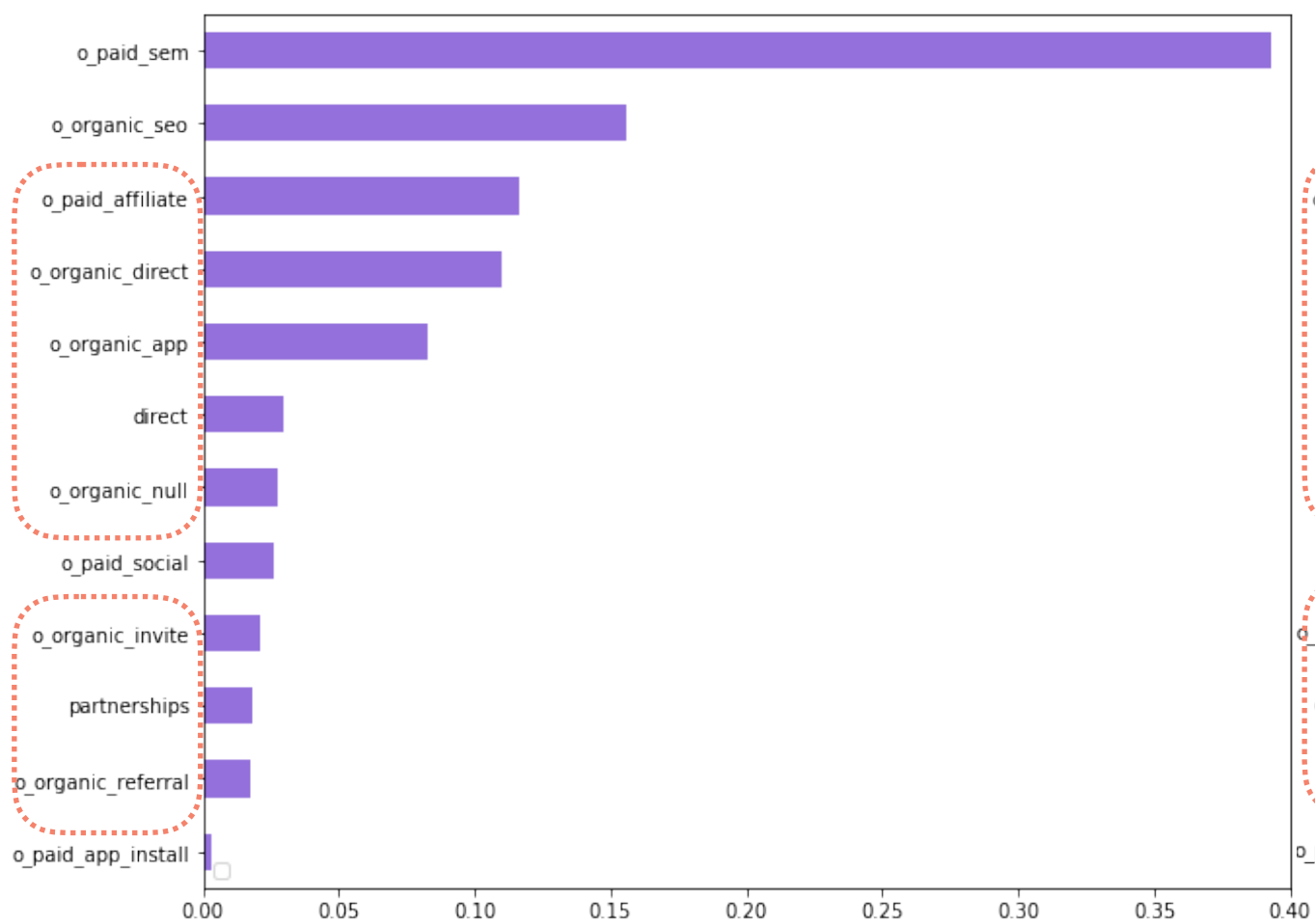
- ~50% of merchants register during the day they searched.
- ~60% of merchants register within 3 months of their search (11%  $\geq 1$  days)
- The rest: 38% of merchants we don't know their journey — we only know their last touch.

# Attribution models

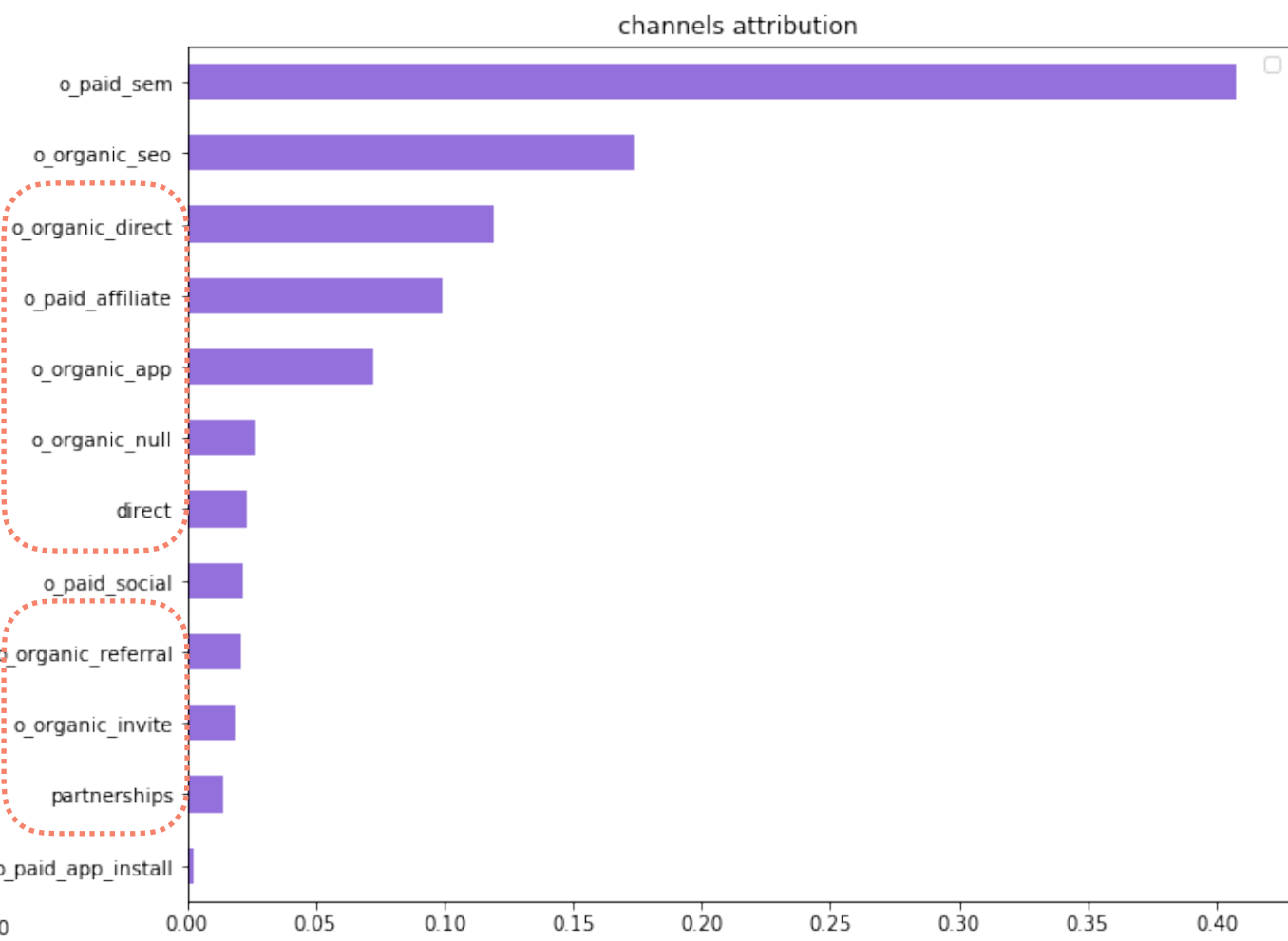
- Base model: last touch point
- Data driven model: Shapley value
- Conversion: REG, KYC, ATC10

Europe

Last touch: ATC10



Shapley model: ATC10



# Create second attribution dataset

Dataset =

Ga + GCP+ Fb+ Youtube + Google Display + expert's intuition

## Second dataset's challenges

- Collect data manually from # sources: Fb, Google Display and Youtube
- Fb, Google Display and Youtube: aggregated data
- Assumptions



# Thank you!

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