

Marketing Analytics

May 25, 2019



- I used to work in fashion in New York
- Now I work at a local bank, managing a 38 people team of digital marketers, developers, and product folks
- We spend ~PHP200MM/yr on marketing media with 80% of allocated for performance
- I make excellent pancakes

About Me / Digital



About Security Bank

AGENDA

Discussion Points

What is marketing?

What is analysis?

What is Marketing Analytics?

- Measurement
- Analysis
- Management

Marketing Analytics in practice

Exercise



Let's start by asking...

**What is the purpose
of marketing?**



WHAT'S THE
R.O.I. OF OUR
MARKETING?



THERE'S AN
INCREASE IN
"BRAND
AWARENESS"



AND
"BRAND
AFFINITY"



AND
"BRAND
SENTIMENT"



AND
"BRAND
ENGAGEMENT"




HOW ABOUT IN TERMS
I CAN UNDERSTAND,
LIKE "E.B.I.T.D.A."?



IT SOUNDS
LIKE YOU
JUST MADE
THAT UP.



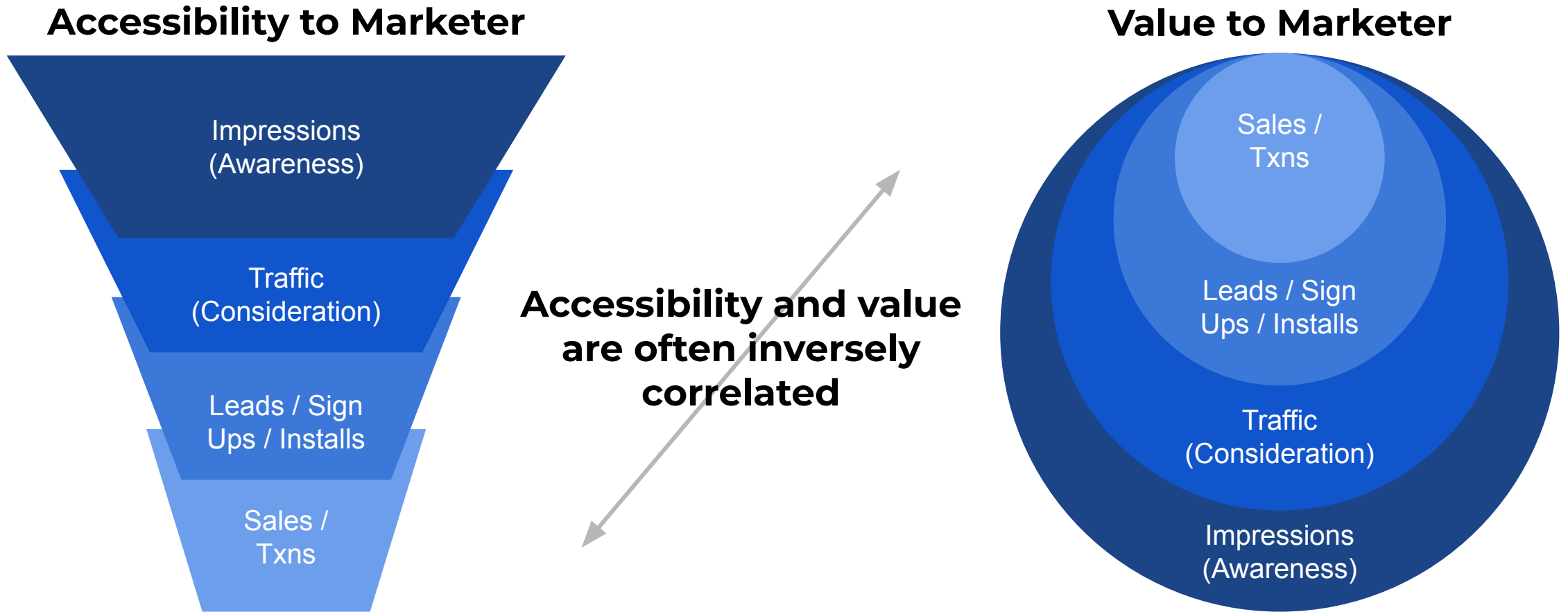
TOM
FISH
BURNE

 **Ultimately, the
purpose of
marketing is to drive
business growth

(usually sales)**



But sales can be ambiguous so marketers use proxy metrics for a customer's value



What is analysis?





a·nal·y·sis

/əˈnæləsəs/

noun

1. detailed examination of the elements or structure of something.

"statistical analysis"

synonyms: examination, investigation, inspection, survey, scanning, study, scrutiny, perusal;

[More](#)





INGREDIENTS

- 2 cups all-purpose flour**
- 2 teaspoons baking powder**
- ¼ teaspoon salt**
- 1 tablespoon sugar, optional**
- 2 eggs**
- 1 ½ to 2 cups milk**
- 2 tablespoons melted and cooled butter (optional), plus unmelted butter for cooking, or use neutral oil**



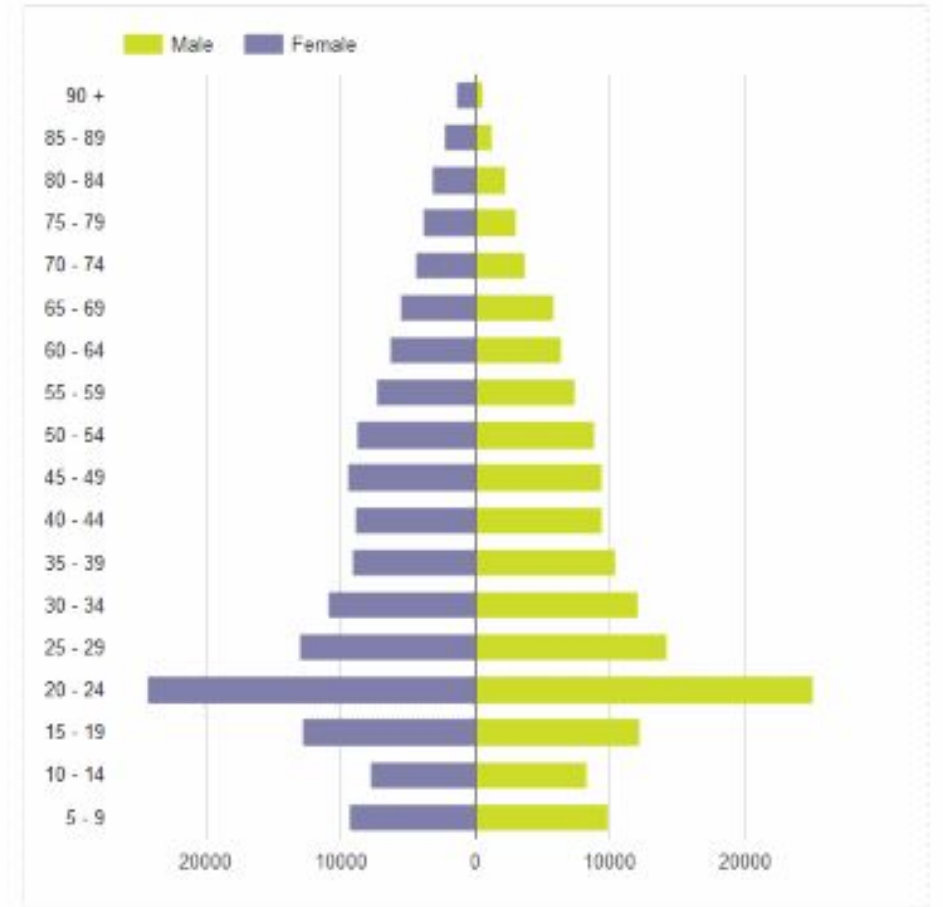
RATIO - doughs & batters

	Flour	Fat	Liquid	Egg	Sugar	Notes
Bread	5		3			Rule of thumb: Salt : 2% of flour's weight. Yeast : 1tsp for 1 pound (454gr) of flour. Bread's temperature: 180-210F (82-98C)
Pasta	3			2		1 egg per person. 1 egg = 2oz (56gr). If scale less: 2 eggs per 1 cup of flour
Pie dough	3	2	1			Butter (fat) must be as cold as possible. Dough must be worked as little as possible
Biscuit	3	1	2			1 tsp baking powder per 4oz (225 gr) of flour. Use butter as your fat
Cookie	3	2			1	Flavor will vary depending on additions. Use butter as your fat
Pound / sponge cake	1	1		1	1	Combining order gives different results. Pound cake order: butter, sugar, egg and flour. Sponge cake (foaming method): whip eggs and sugar first. Creaming method : paddling sugar into batter, add eggs then dry ingredients
Pate a' choux	1	1	2	2		Savory (Parisien gnocchi) & sweet (cream puff dough) preparations (depends on salt/sugar)
Muffin	2	1	2	1		Straight mixing method. Baking powder
Fritter	2		2	1		Straight mixing method. Baking powder
Pancake	2	1/2	2	1		Straight mixing method
Crepe	1/2		1	1		Works also with volumes ratio, if scale less
Popover	1		2	1		Straight mixing method. No baking powder

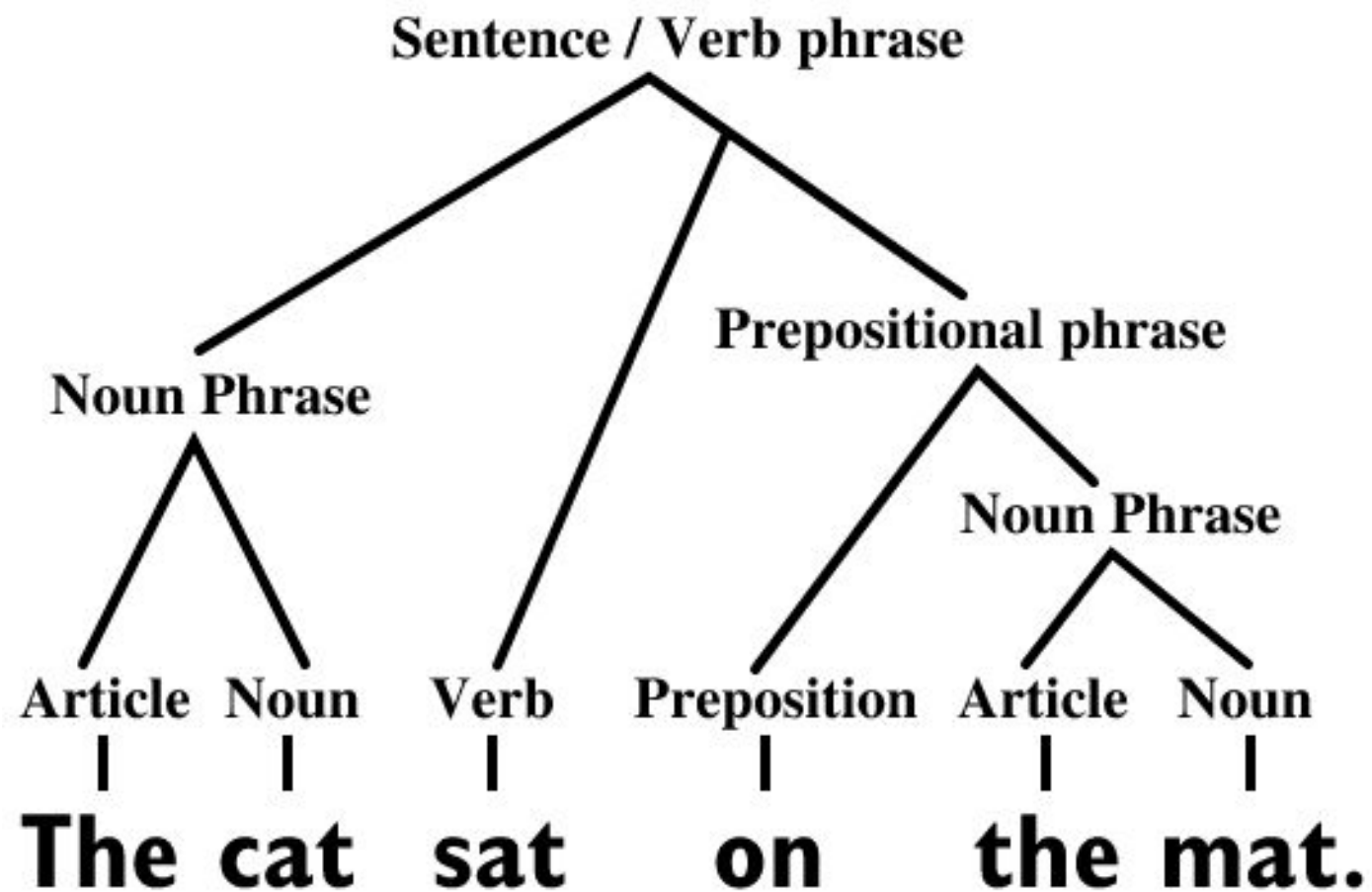




Total population: population pyramid



Basic constituent structure analysis of a sentence:



**So why analyze things?
Joke time!**

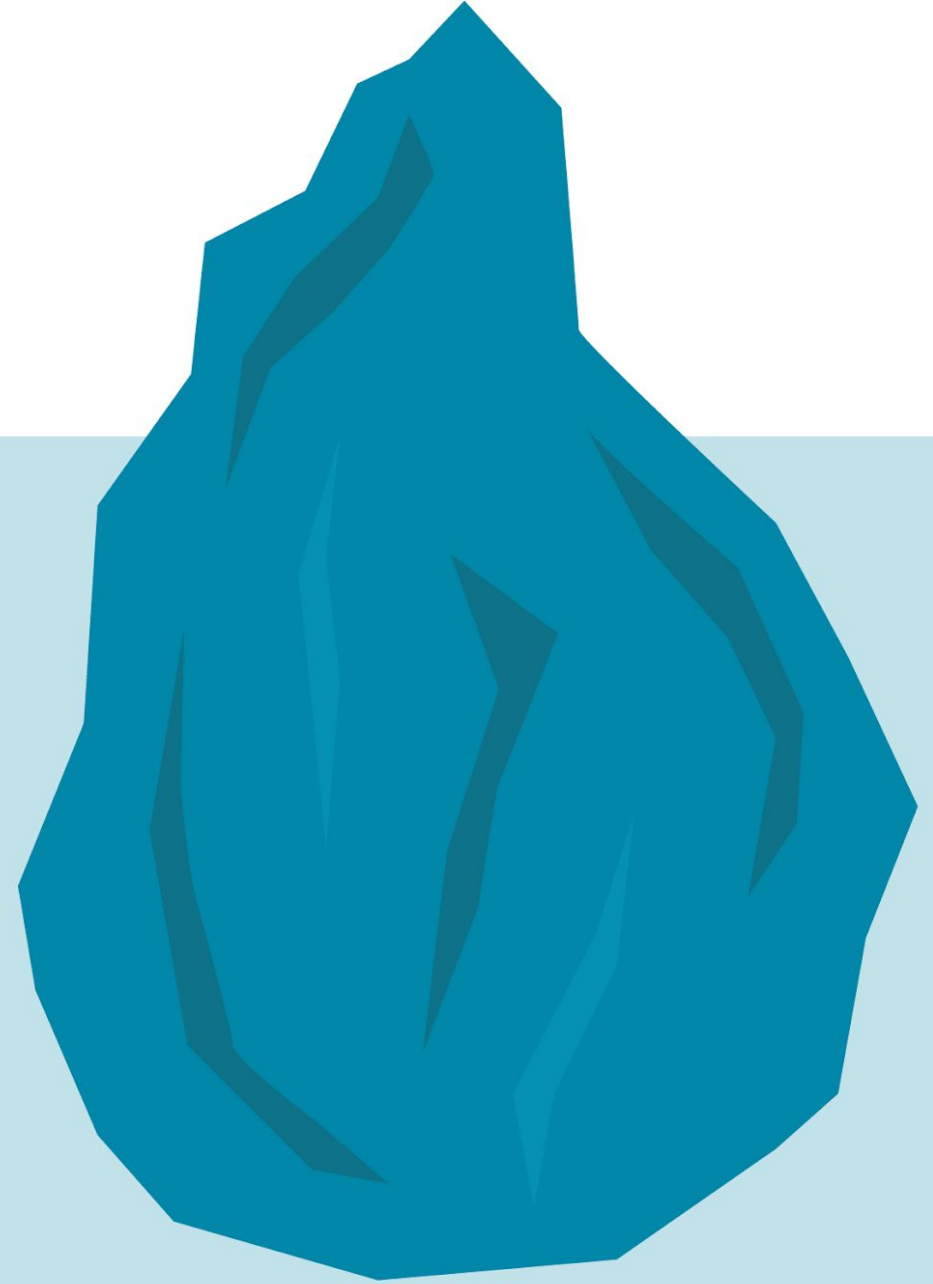


Bill Gates walks into a bar...
and everyone inside becomes
a millionaire
On average

20%
CAUSES

80%
EFFECTS

PARETO PRINCIPLE



What is marketing analytics?



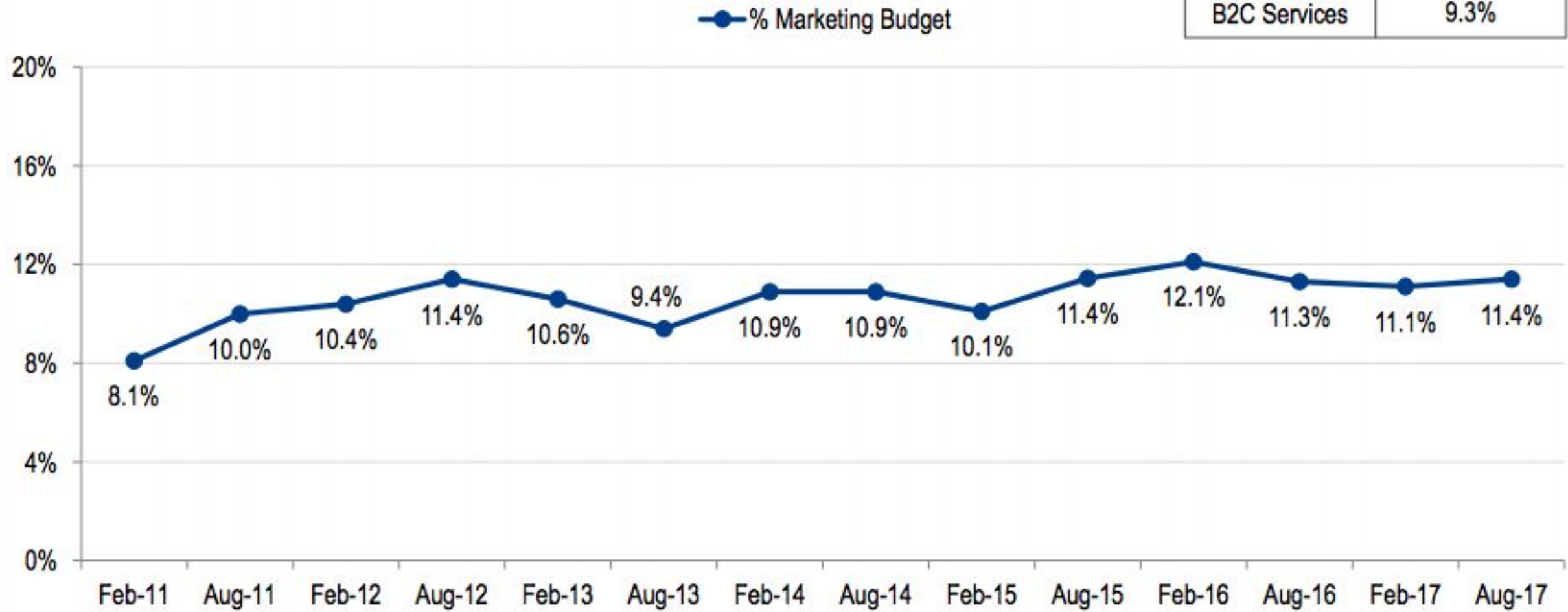
**Marketing effectiveness is measured
through incremental contribution to
business goals**

**(in relation to
used resources)**



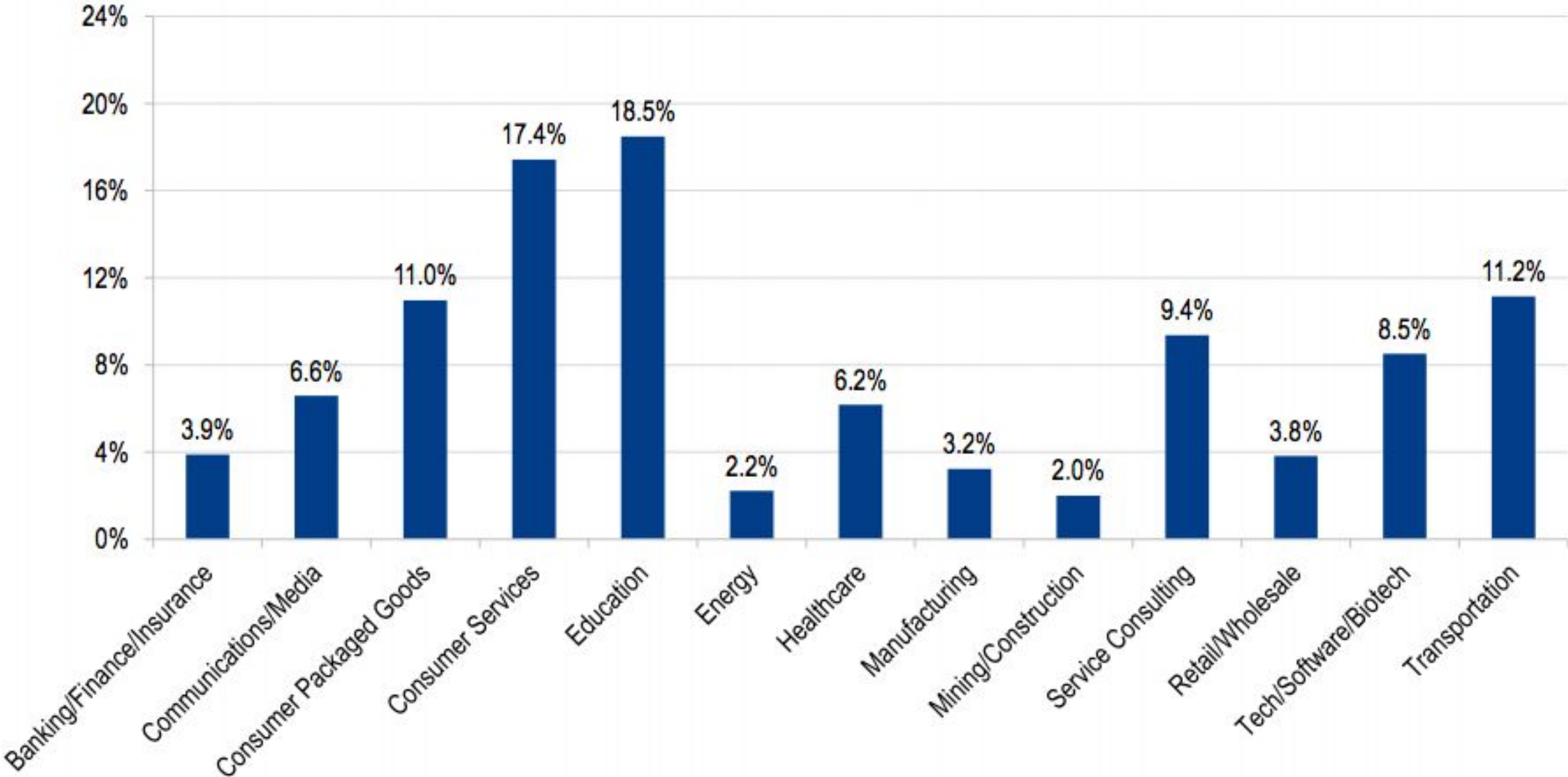
Figure 3.8. Marketing budget as percent of firm budget*

Overall	11.4%
B2B Product	10.4%
B2B Services	12.6%
B2C Product	13.4%
B2C Services	9.3%



Source: https://cmosurvey.org/wp-content/uploads/sites/15/2017/08/The_CMO_Survey-Highlights_and_Insights-Aug-2017.pdf

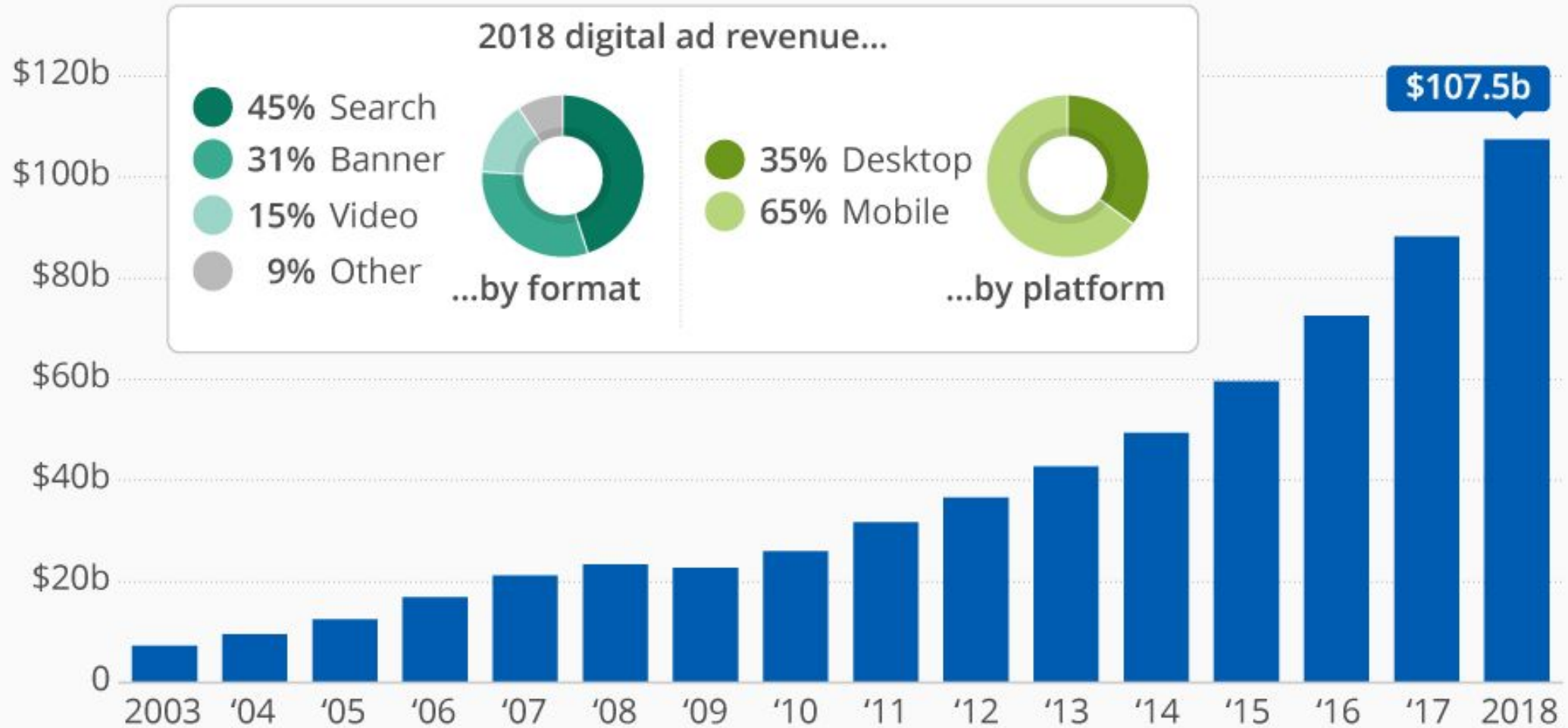
Figure 3.9. Marketing spending as percent of company revenues*



Source: https://cmosurvey.org/wp-content/uploads/sites/15/2017/08/The_CMO_Survey-Highlights_and_Insights-Aug-2017.pdf

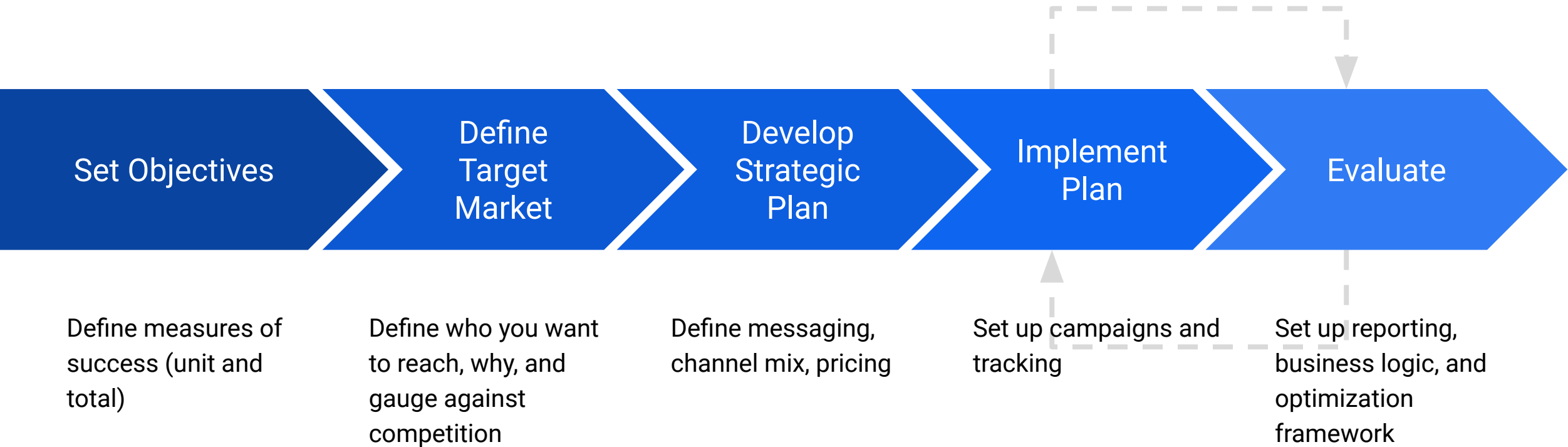
U.S. Online Ad Market Surpasses \$100 Billion

Internet advertising revenue in the United States*



* incl. desktop and mobile online advertising revenues from websites, commercial online services, ad networks and exchanges, mobile devices, and email providers, as well as other companies selling online advertising

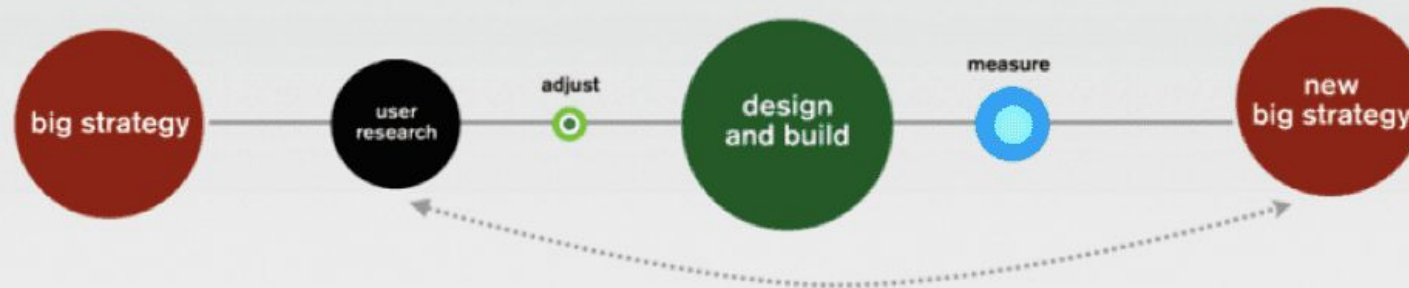
Marketing activities seem to be the same as they were in the past



But what has changed is that the marketing process has sped up and become much more iterative

Conventional, linear development process

Big ideas, big bang launch, big budgets



Iterative, Agile, emergent development process

Micro-strategies, big insights, rapid iterations



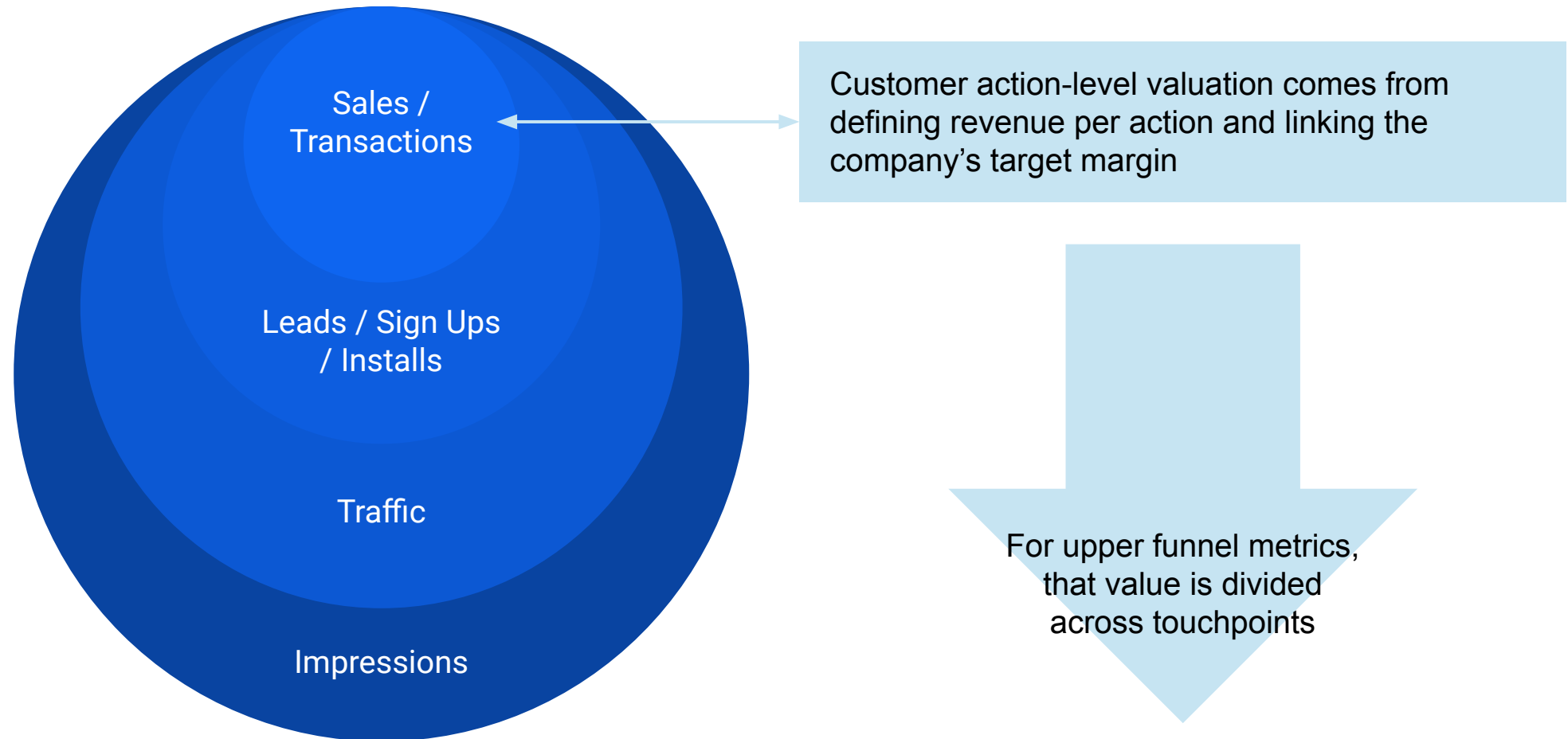
Nicked and adapted for Web from David Armano + MADE BY MANY



Data-driven marketing relies on 3 core analyses

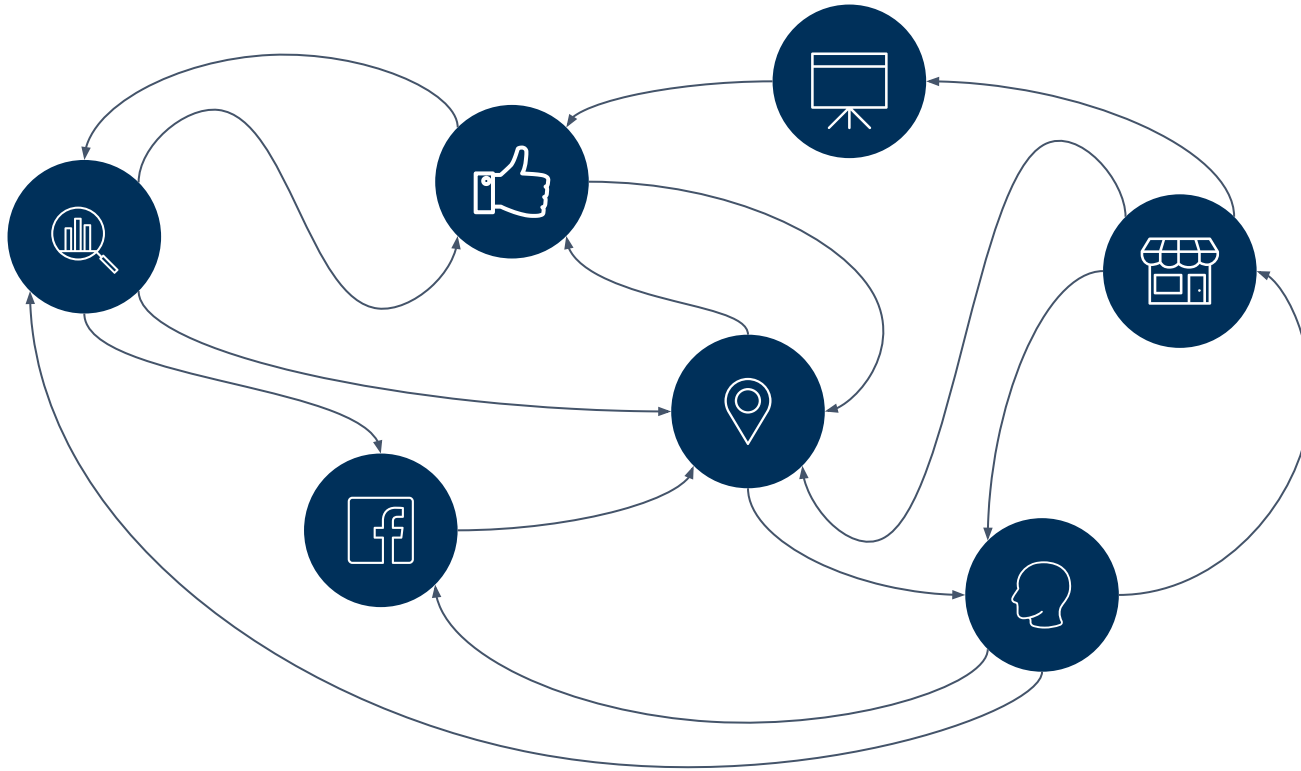


Valuation defines a “target” metric and a target cost for that metric (typically an action/value)

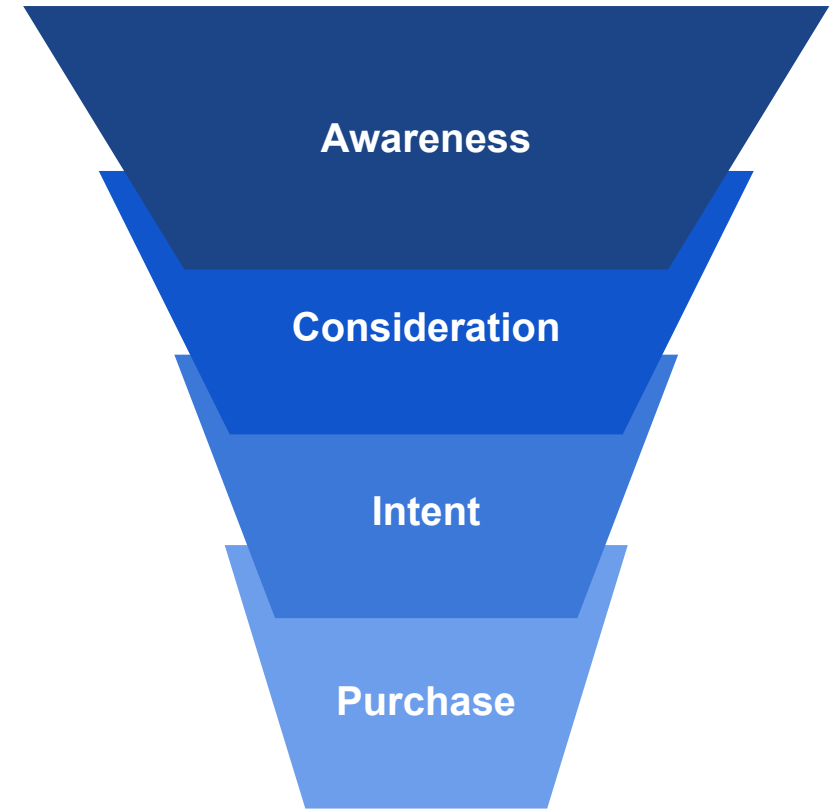


A customer's path to purchase is complex... so often abstracted by marketers to a purchase funnel

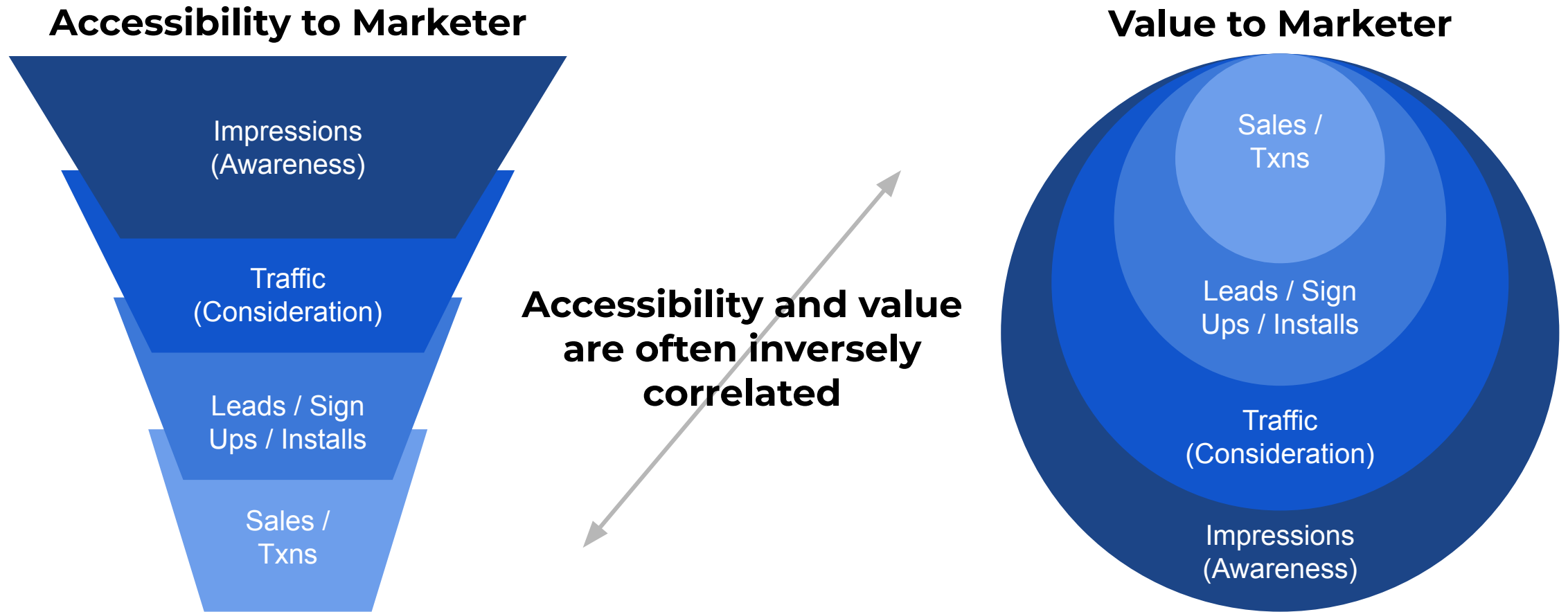
Customer Decision Journey



Purchase Funnel

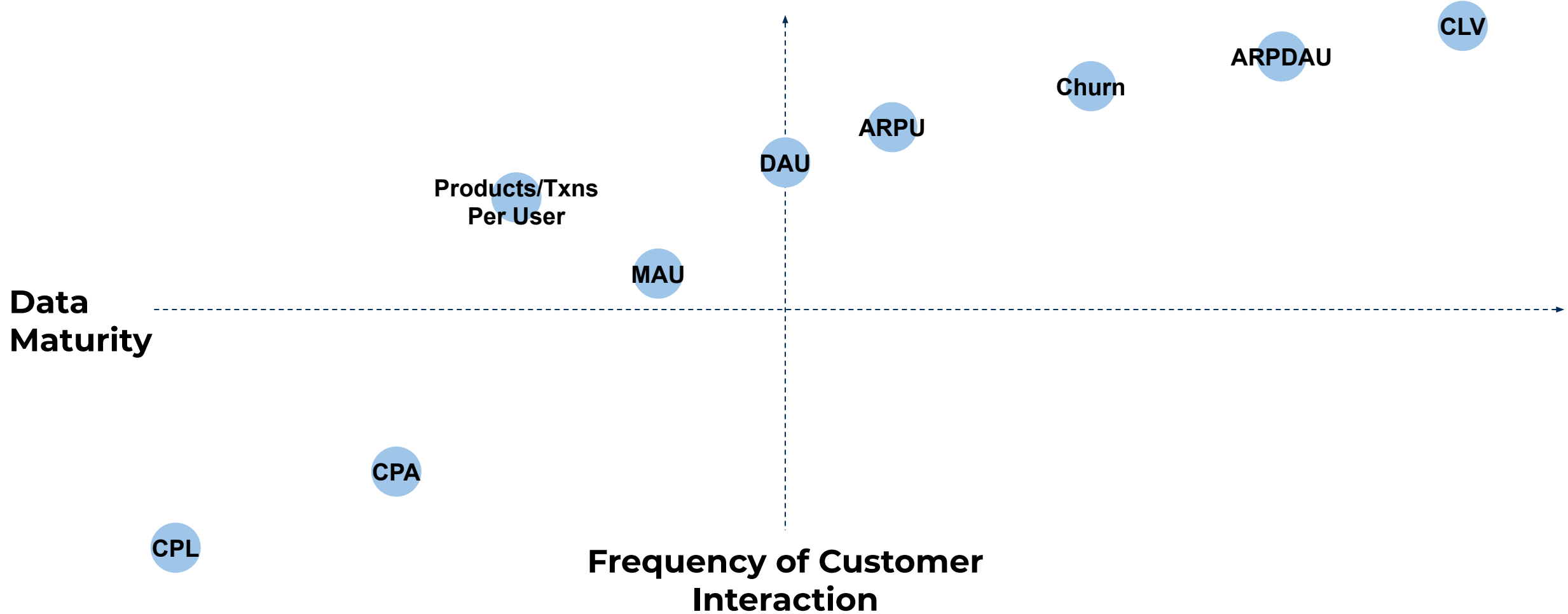


Proxy metrics at each step of the funnel stand for a customer's likelihood to purchase (aka value)



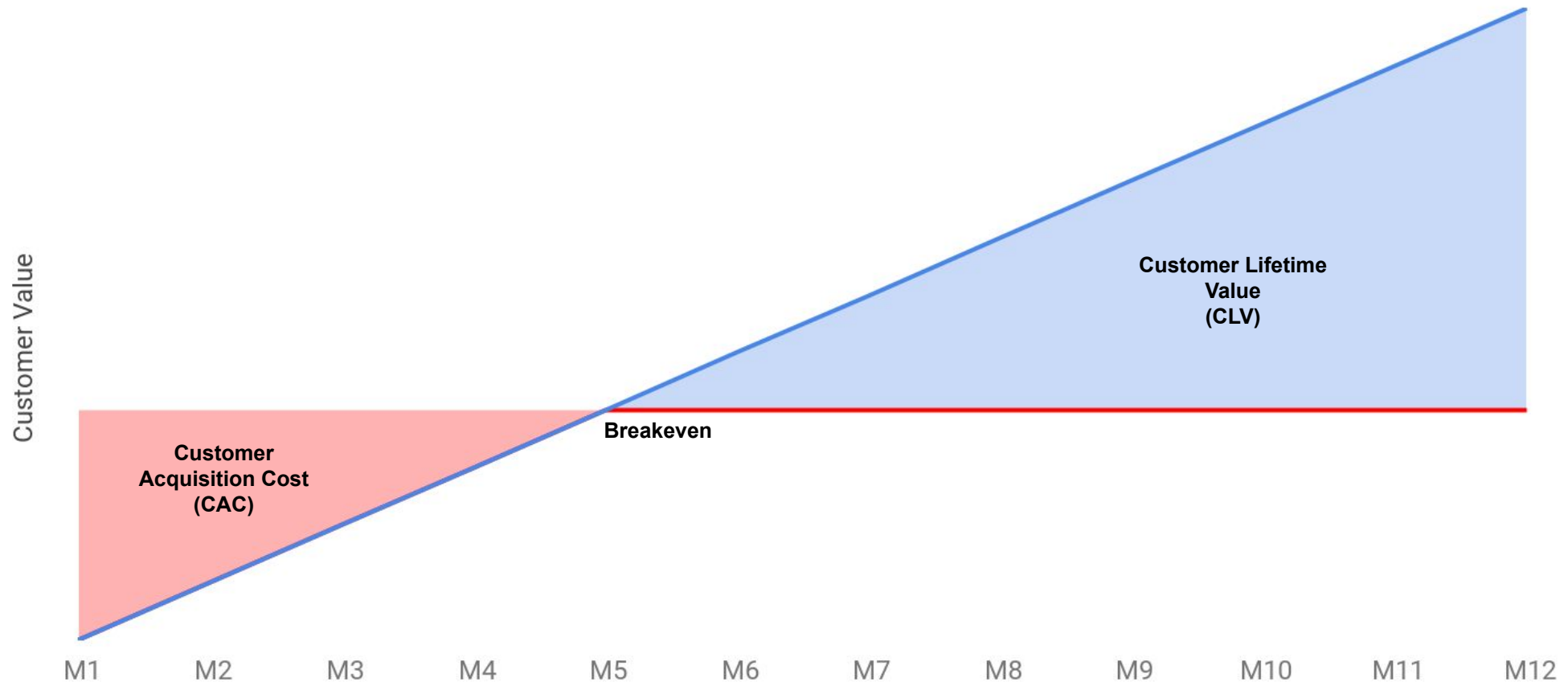
Terms	Meaning
CPL	Cost Per Lead
CPA	Cost Per Action
MAU	Monthly Active Users
Products/User	Products Per User
Transactions/User	Transactions Per User
DAU	Daily Active User
ARPU	Average Revenue Per User
Churn	How many users are leaving
ARPPDAU	Average Revenue Per Daily Active User
CLV	Customer Lifetime Value

The “right” target metrics depend on a company’s maturity and frequency of customer interaction

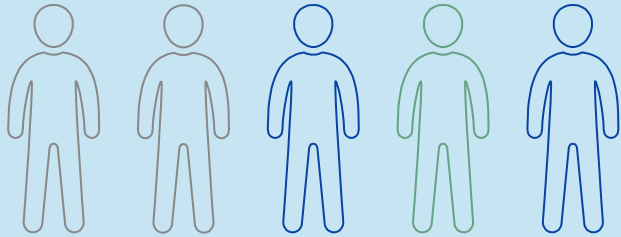


Terms	What's it good for
CPL: Cost per Lead	When you have a complex or undefined sales process
CPA: Cost per Action	When you have a specific action you want a user to do that results in something beneficial to your business
MAU: Monthly Active User	When you want to see a trend of your users over a long period of time
Products/User	When you want to optimize user product purchase behavior
Transactions/User	When you want to optimize user transaction numbers
DAU: Daily Active User	When you want to see a trend of your users over a short period of time
ARPU: Average Revenue Per User	When you want to know the effectiveness of project monetization for entire user base (not just paying users)
Churn	When you want to know the percentage of customers that stopped using your company's product or service during a certain time frame
ARPPDAU	When you want to measure the effectiveness of finding paying users for new marketing campaigns
CLV: Customer Lifetime Value	When you want a metric that indicates the total revenue a business can reasonably expect from a single customer account

For retail banks, a key metric should be CLV: a combination of transaction/product-level revenue



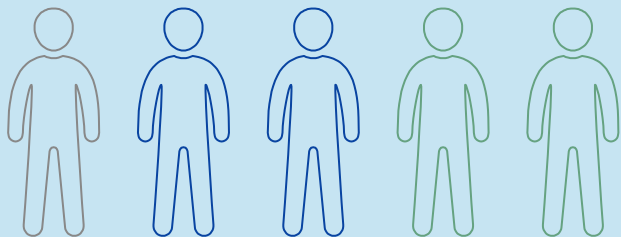
Attribution links target metrics (customers, transactions, etc) to marketing activities



Target metric:
Conference Tickets Sold

Total Tickets Sold: 548

Avg Cost per Sale: \$9.12

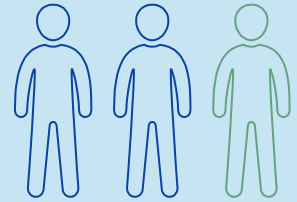


Campaign 1: Mall Event

Tickets Sold: 23

Event Cost: \$1,000

Cost per Sale: \$43.48

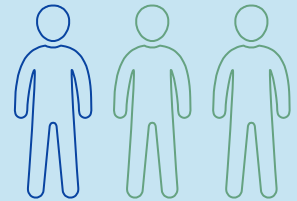


Campaign 2: Facebook Campaign

Tickets Sold: 225

Campaign Cost: \$2,500

Cost per Sale: \$11.11

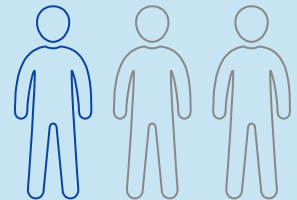


Campaign 3: Referral Program

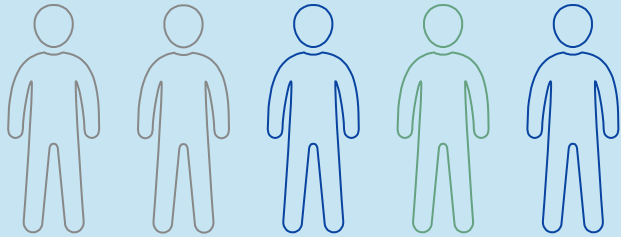
Tickets Sold: 300

Campaign Cost: \$1,500

Cost per Sale: \$5.45



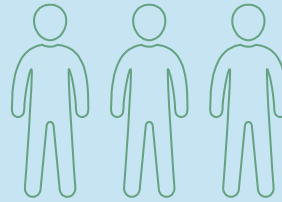
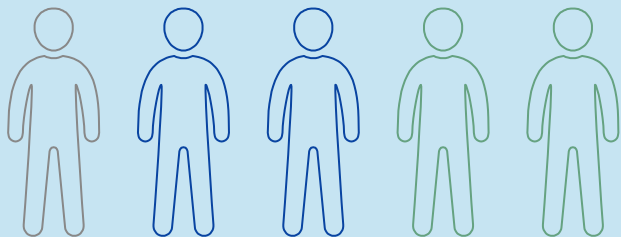
The same principle applies to Segmentation, which (should) links metrics to customer characteristics



Target metric:
Conference Tickets Sold

Total Tickets Sold: 548

Avg Cost per Sale: \$9.12

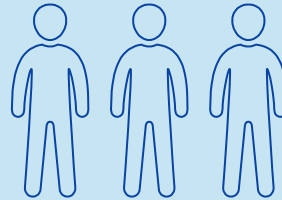


Age Between 18-24

Tickets Sold: 219

Attributed Cost: \$2,302

Cost per Sale: \$52.94

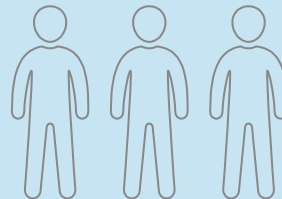


Age Between 25-34

Tickets Sold: 149

Attributed Cost: \$1,112

Cost per Sale: \$7.46



Age Between 35-44

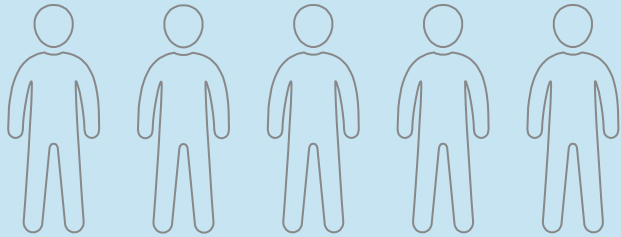
Tickets Sold: 140

Attributed Cost: \$1,586

Cost per Sale: \$11.32

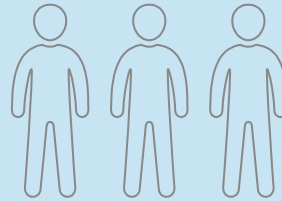
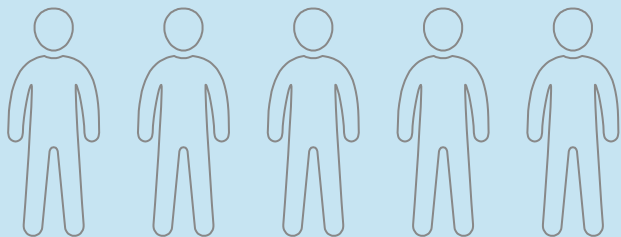


You can and should layer characteristics/activities to better understand marketing activity



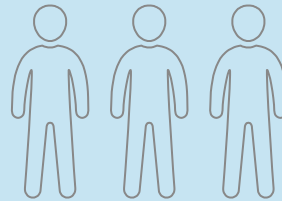
**Total Sign Ups (Ages 35-44):
140**

Avg Cost per Sale: \$11.32



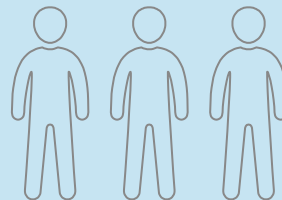
Campaign 1: Mall Event

Tickets Sold: 20
Attributed Cost: \$812
Cost per Sale: \$40.60



Campaign 2: Facebook Campaign

Tickets Sold: 20
Attributed Cost: \$372
Cost per Sale: \$18.60



Campaign 3: Referral Program

Tickets Sold: 100
Attributed Cost: \$500
Cost per Sale: \$5.00



Segmentation is just a way to classify data — it isn't limited to explicit characteristics like demographics

Demographic

Age
Income
Education
Job
Employer
Tenure
Ownership
Marital Status
with Children

Income/Age Ratio
of Degrees
Worked at X Companies

Geographic

Continent
Country
Province
State
City
Village
Climate

X Miles away from...
Population Density
Urban/Rural
City Size
Terrain

Behavioral

(Usage patterns,
propensity)

User Status
Depth of Engagement
Frequency
Likelihood to

Time spent per day
Total orders per month
Average order amount

Likelihood to Return
Price/Conv Sensitivity

Psychographic

(Activities, attitudes,
personality, etc)

Interested in
Believes in
Agrees with
Disagrees with
Motivated by

Family Oriented
Likes Dogs
Disagrees with RH law
Does things last minute
Utility-driven



USE YOUR

IMAGINATION



Allocation amounts to shifting resources to higher-efficiency activities/segments

Before
(\$9.12/sale)

Campaign 1: Mall Event

Tickets Sold: 23
Event Cost: \$1,000
Cost per Sale: \$43.48

Campaign 2: Facebook Campaign

Tickets Sold: 225
Campaign Cost: \$2,500
Cost per Sale: \$11.11

Campaign 3: Referral Program

Tickets Sold: 300
Campaign Cost: \$1,500
Cost per Sale: \$5.00

Worst performing
activity, stopped

Lowered budget
and bid cap

Increased
incentive amount
to \$7/ticket

After
(\$7.30/sale)

Campaign 2: Facebook Campaign

Tickets Sold: 184
Campaign Cost: \$1,500
Cost per Sale: \$8.14

Campaign 3: Referral Program

Tickets Sold: 500
Campaign Cost: \$3,500
Cost per Sale: \$7.00



 **Marketing analytics measures, analyzes, and manages the effectiveness of marketing activities**

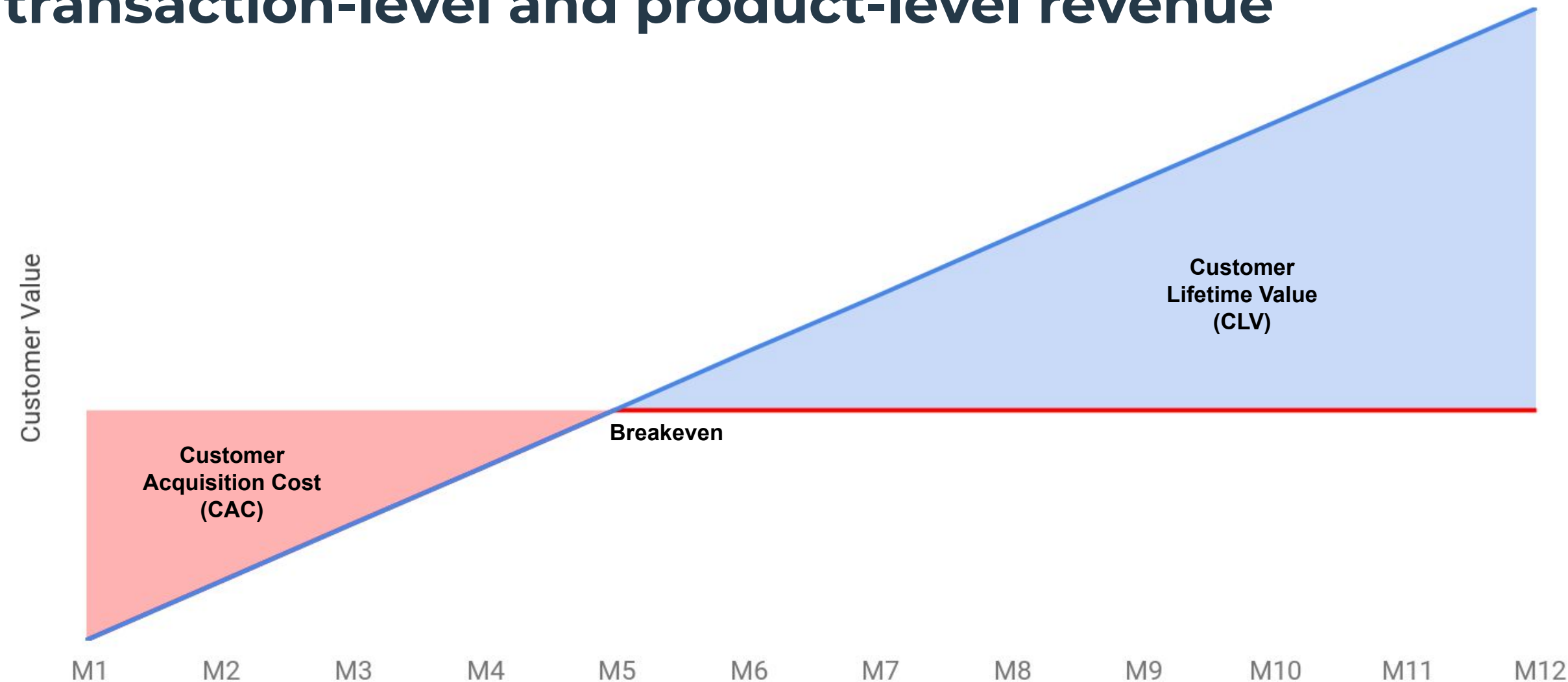


Marketing Analytics in Practice

Measurement: Valuation & CLV



For retail banks, the key metric must become CLV, which is determined from a combination of transaction-level and product-level revenue



CLV is important as the majority of banking customers are technically unprofitable from single product-level view (product-level CLV-CAC)

Customer Behavior / Transactions for Savings Accounts

Savings Account Balance

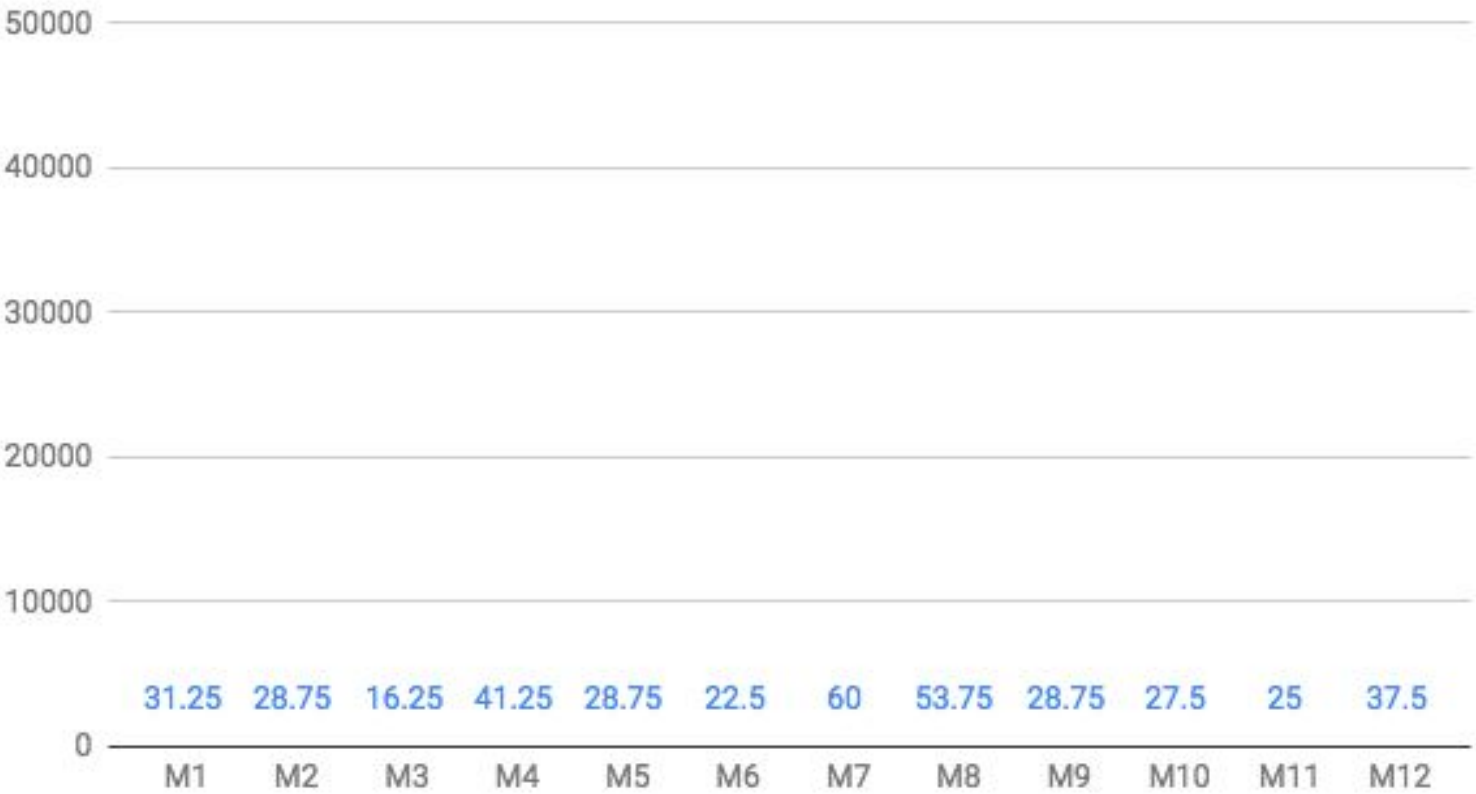


Example

CLV is important as the majority of banking customers are technically unprofitable from single product-level view (product-level CLV-CAC)

Revenue from Said Product

Savings Account Revenue

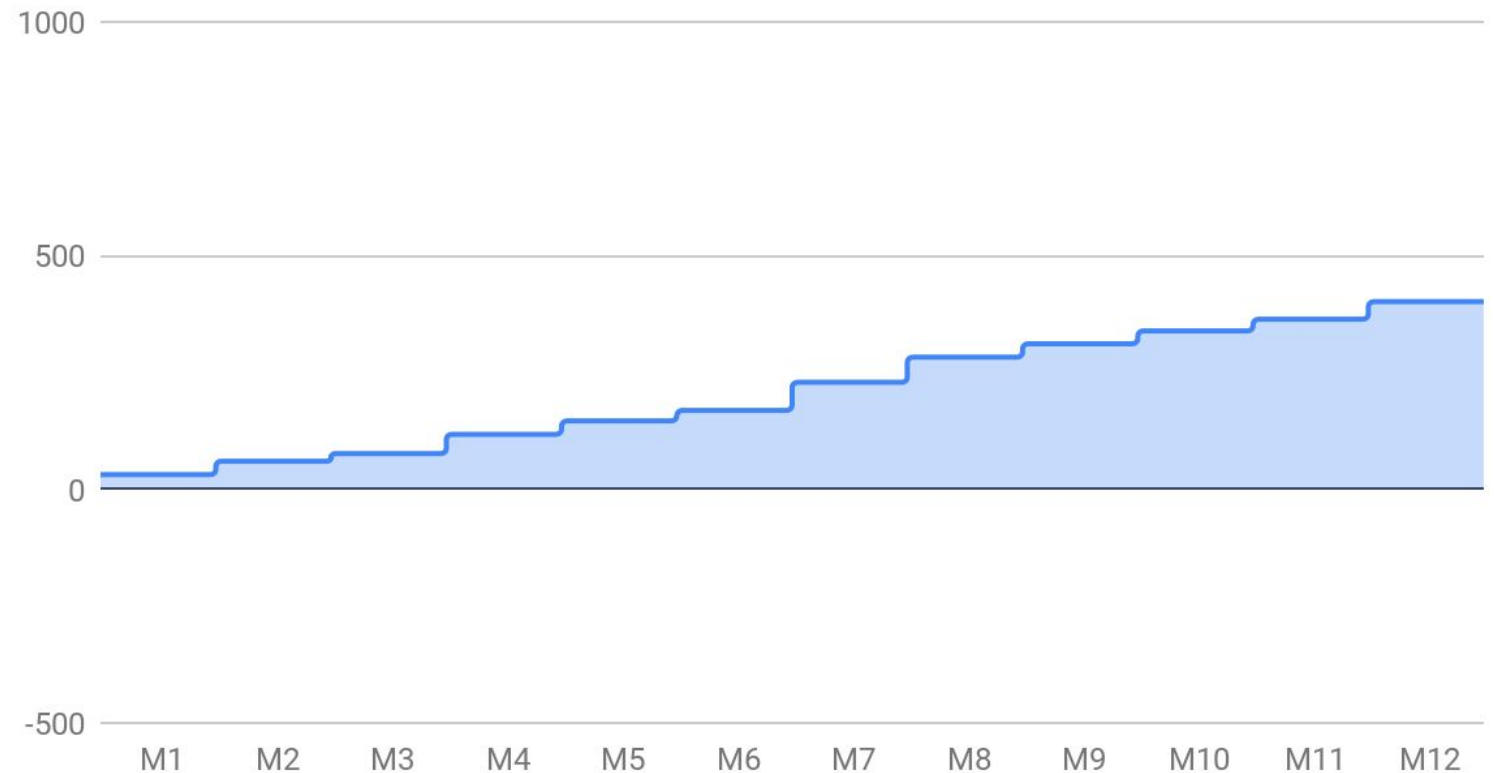


Example

CLV is important as the majority of banking customers are technically unprofitable from single product-level view (product-level CLV-CAC)

Compounded Revenue from Said Product

Savings Account Revenue (Compounded)

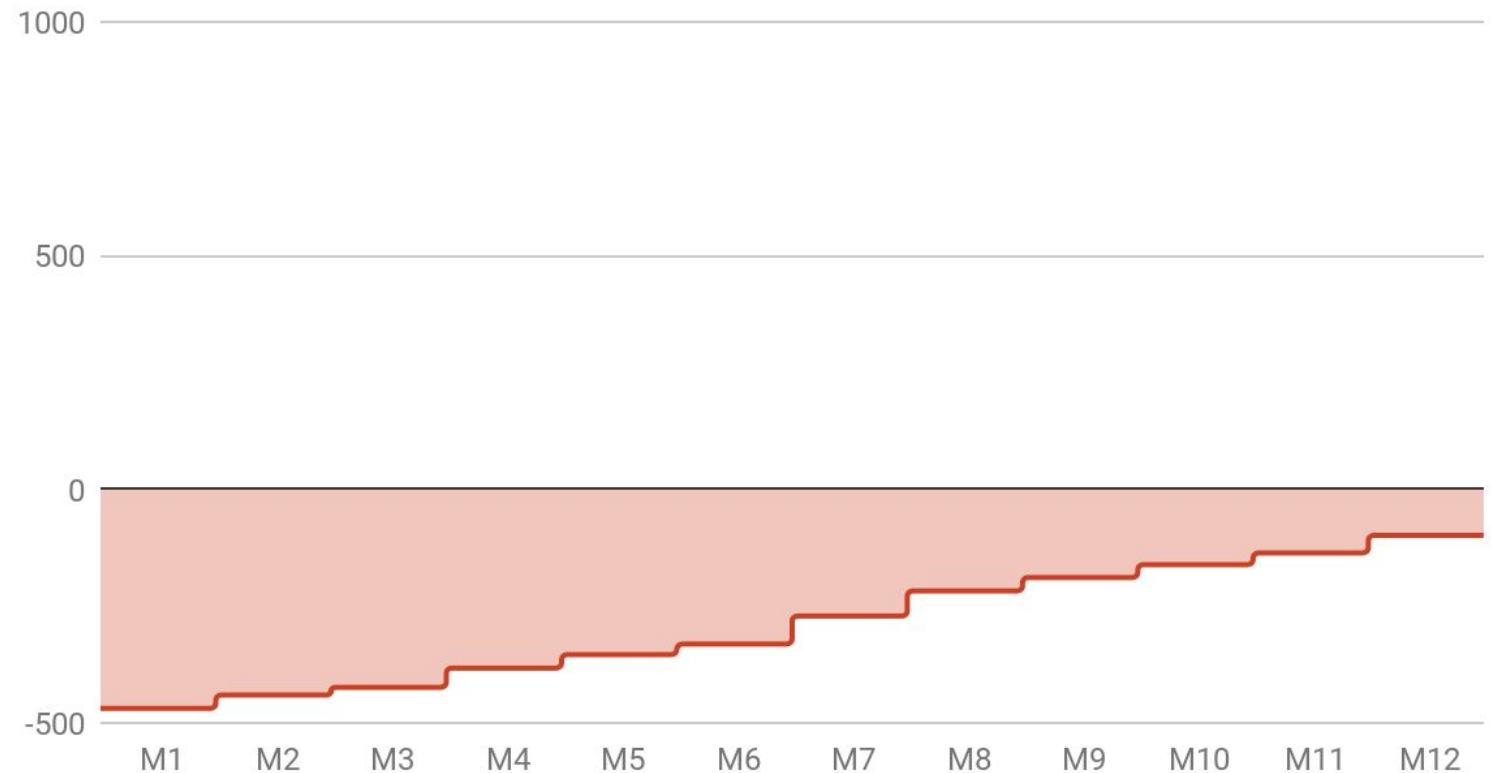


Example

CLV is important as the majority of banking customers are technically unprofitable from single product-level view (product-level CLV-CAC)

But what happens when you add in an acquisition cost of PHP500?

Savings Account Revenue (Compounded) - CAC



Example

CLV in retail banking has one a fundamental difference: the necessary inclusion of risk cost



To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #1

Customer who uses his credit card regularly and pays his bills in full and on time

Customer 1 CLV Items (Cards)

12mo Product-Level CLV



Example

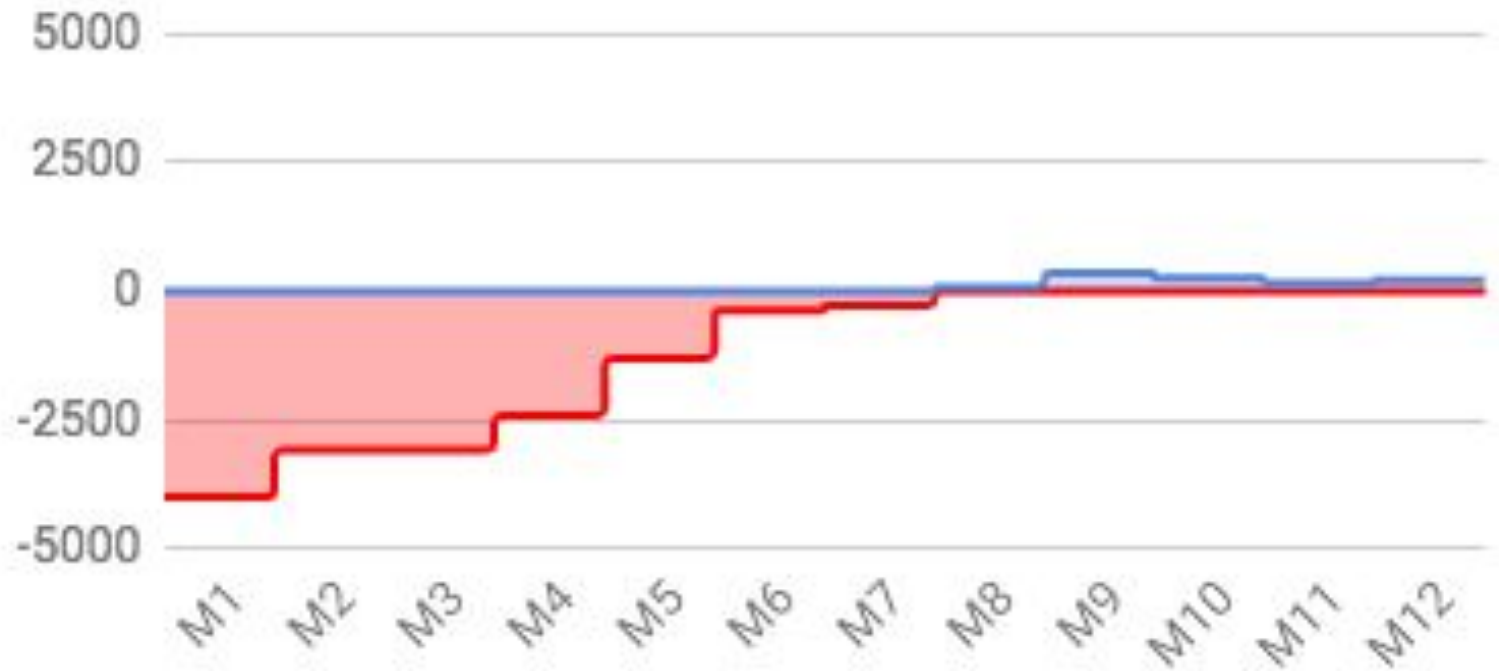
To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #1

Customer who uses his credit card regularly and pays his bills in full and on time

Customer 1 CLV (Cards)

12mo Product-Level CLV



Example

To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #2

Customer who uses his credit card regularly and carries a balance

Customer 2 CLV Items (Cards)

12mo Product-Level CLV



Example

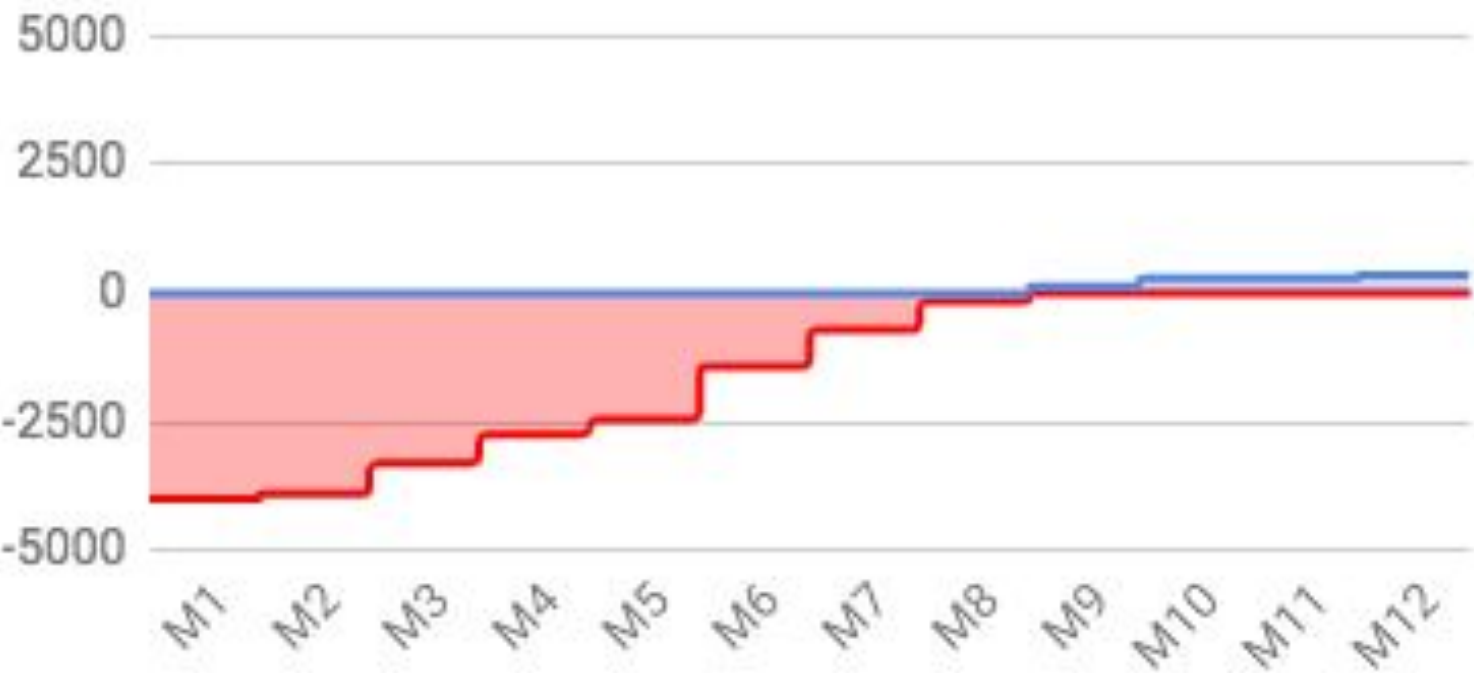
To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #2

Customer who uses his credit card regularly and carries a balance

Customer 2 CLV (Cards)

12mo Product-Level CLV



Example

To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #3

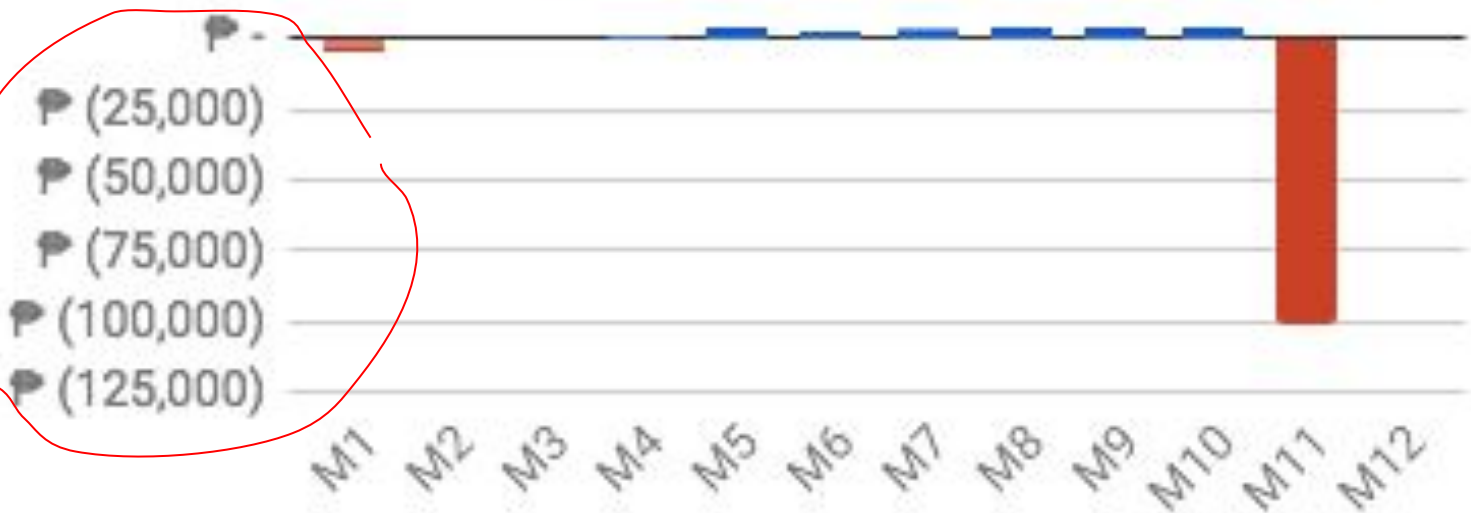
Customer who uses his credit card regularly and carries interest but stops paying his balance in the 11th month

Customer 3 CLV Items (Cards)

12mo Product-Level CLV



Note change in scale



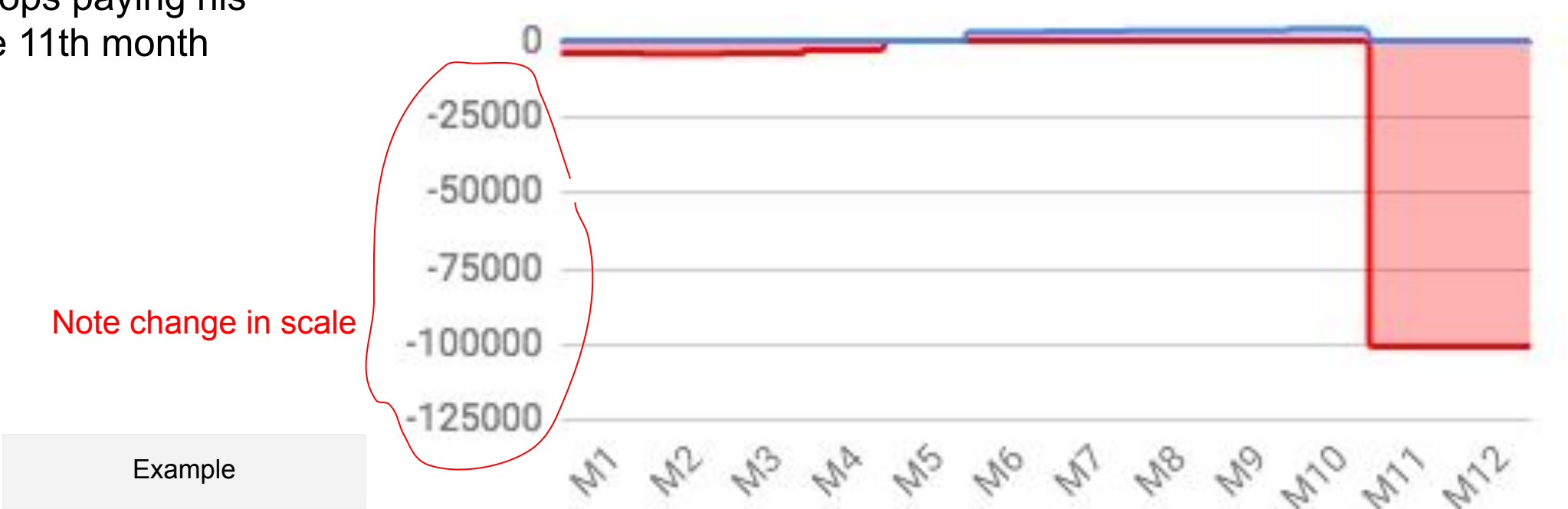
Example

To understand “risk cost”, let’s take a look at the profitability of 3 hypothetical credit card customers over 12mo

Customer #3

Customer who uses his credit card regularly and carries interest but stops paying his balance in the 11th month

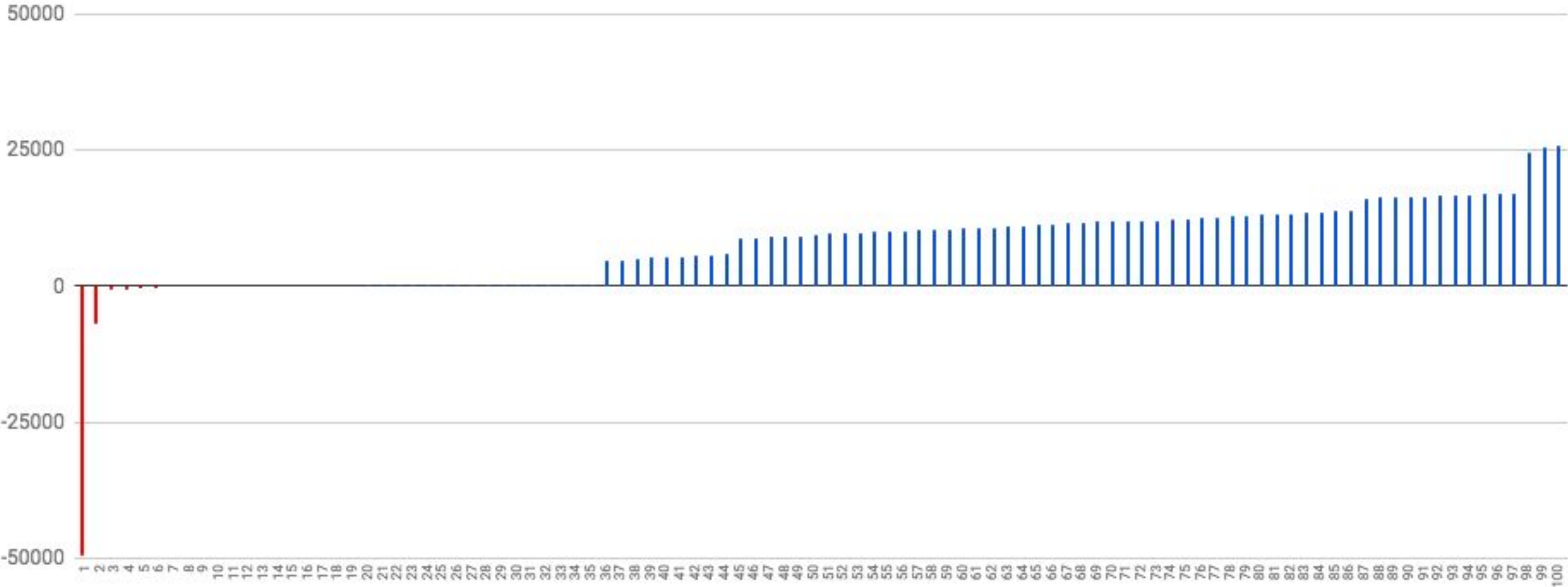
Customer 3 CLV (Cards)
12mo Product-Level CLV



Example

Segmentation is key to driving value: CLV can vary widely across a customer base

1st Year CLV Distribution (Cards)

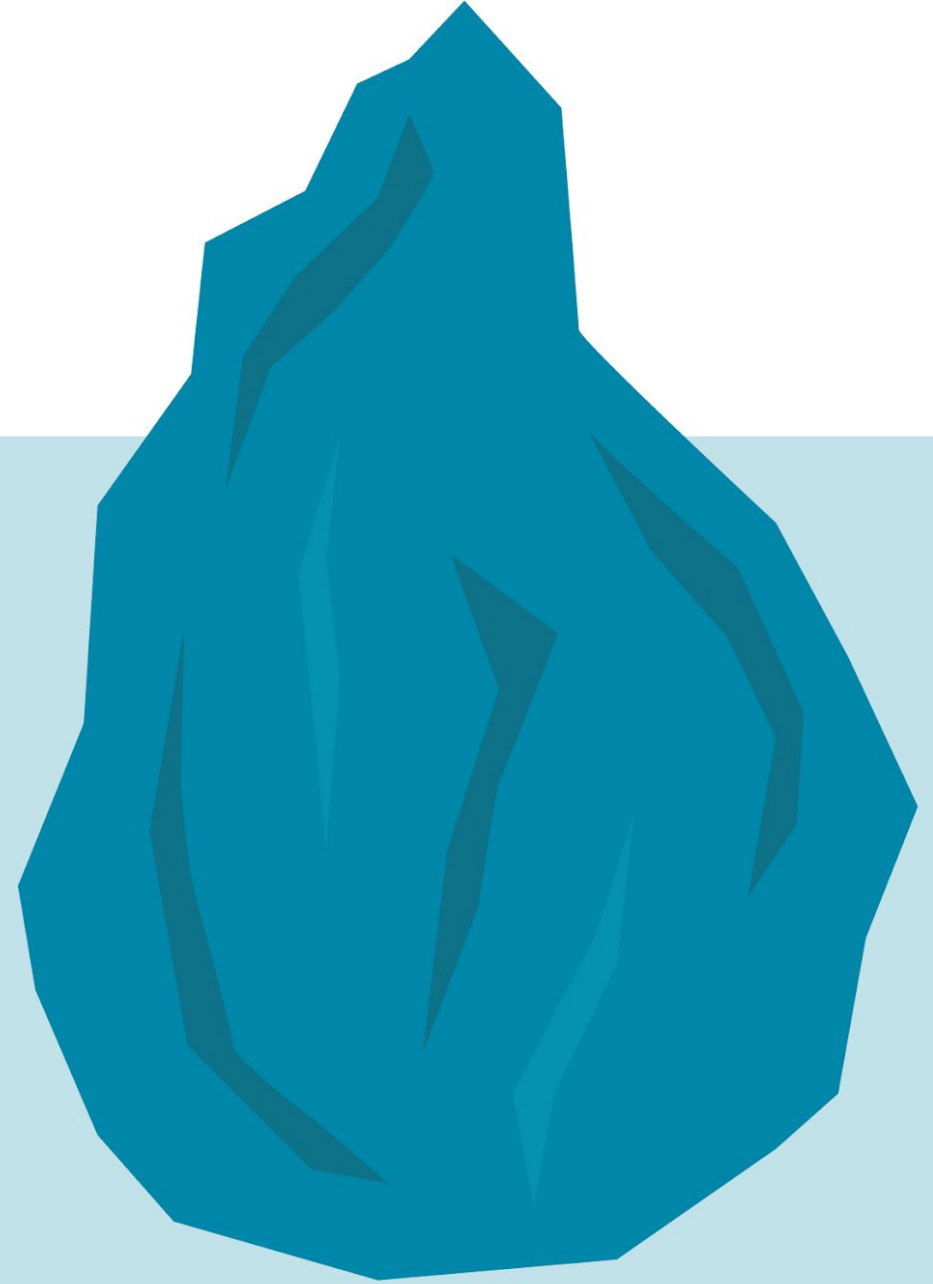


Example

20%
CAUSES

80%
EFFECTS

PARETO PRINCIPLE

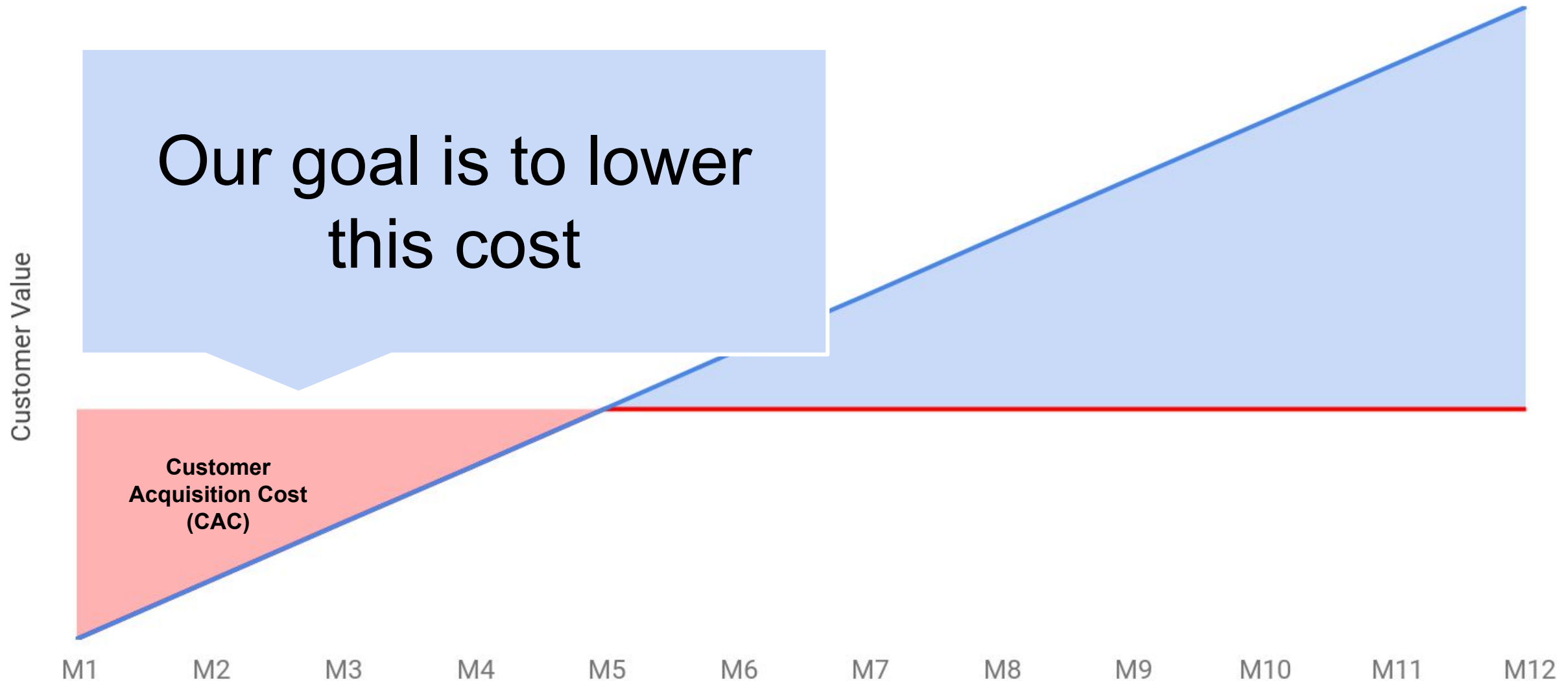


Marketing Analytics in Practice

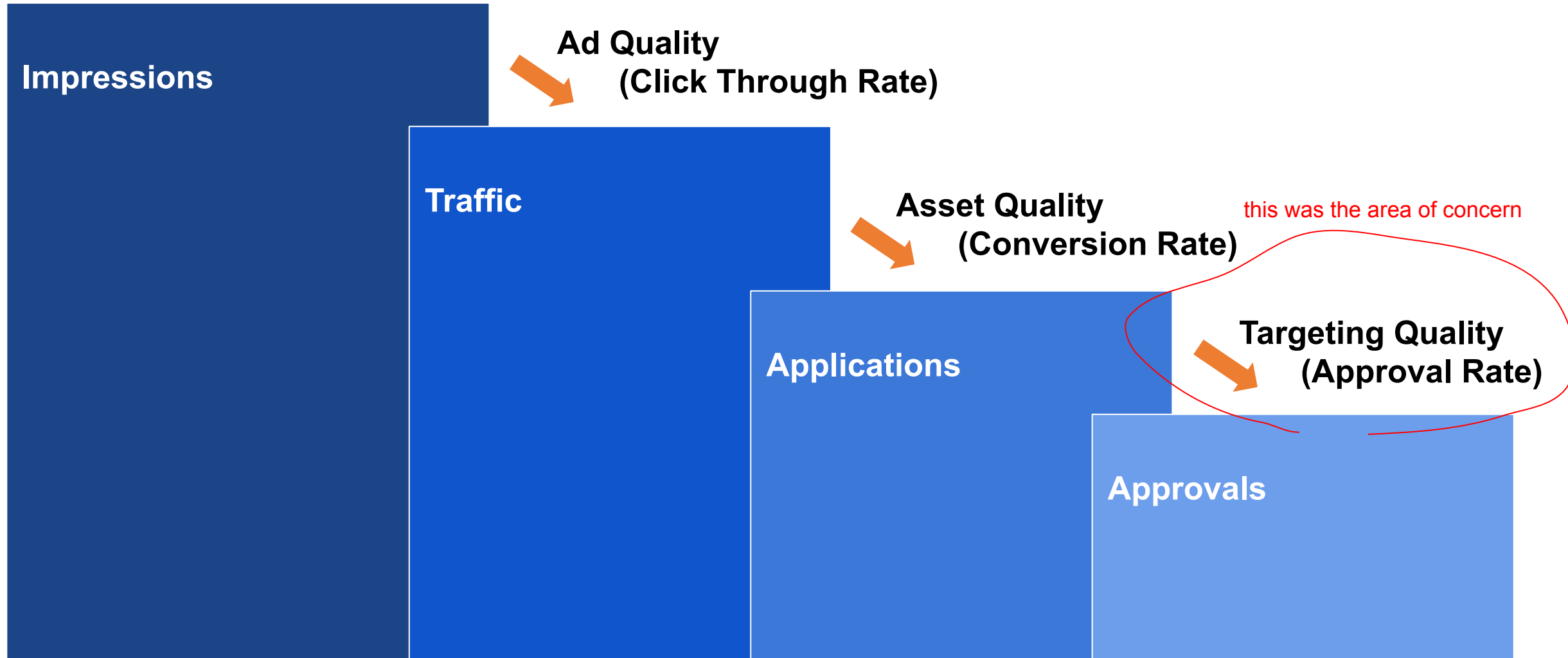
Analysis: Segmentation



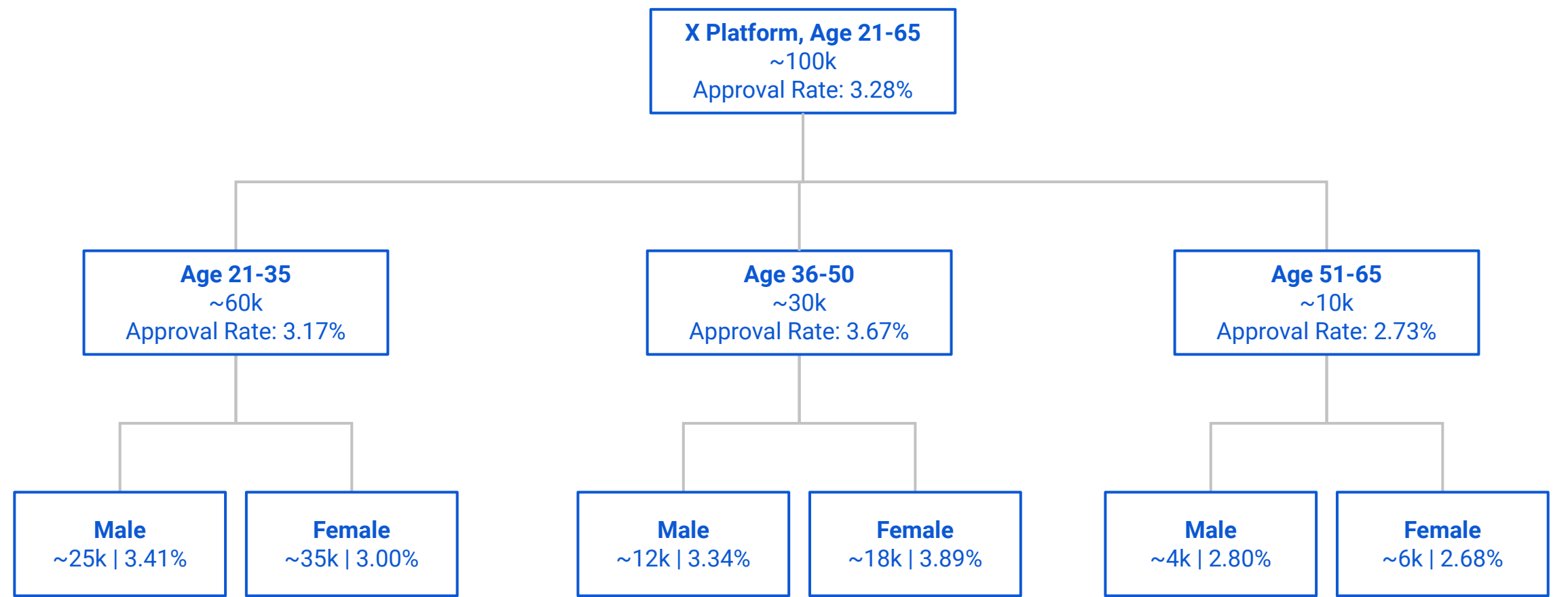
How did we use segmentation to lower our CAC? (and hypothetically increase customer profitability)



Customer acquisition process flow / funnel for consumer finance (loans, etc)

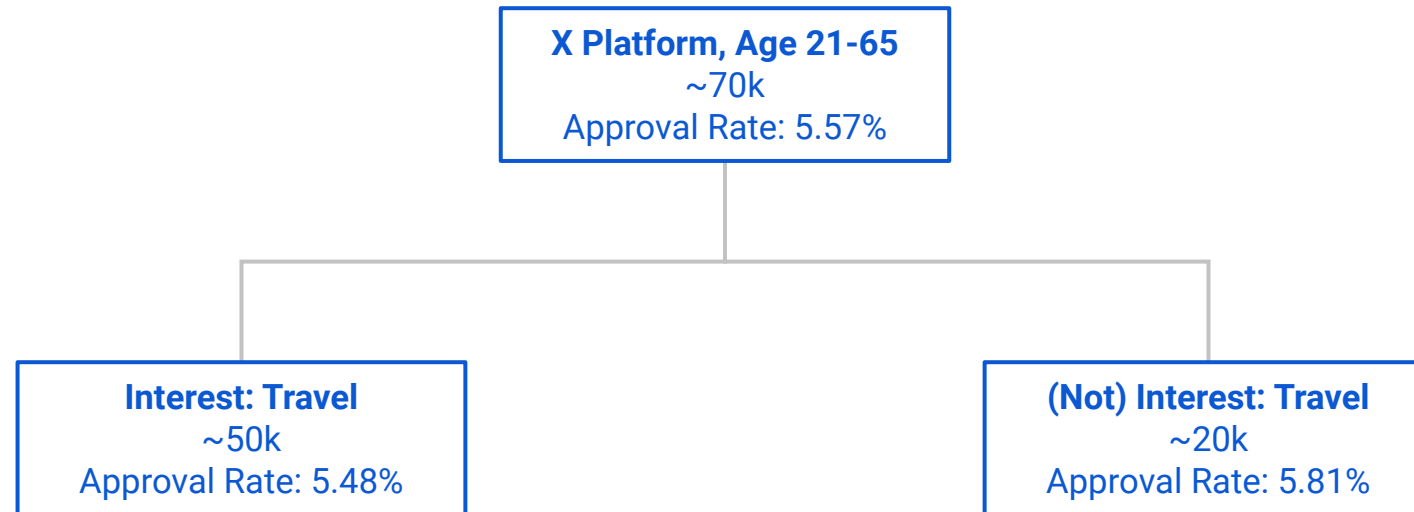


Traditional Demographic segments had low variance in approval rates



Example

The same was true of some natively available Psychographic Segments available for targeting on Facebook/Google

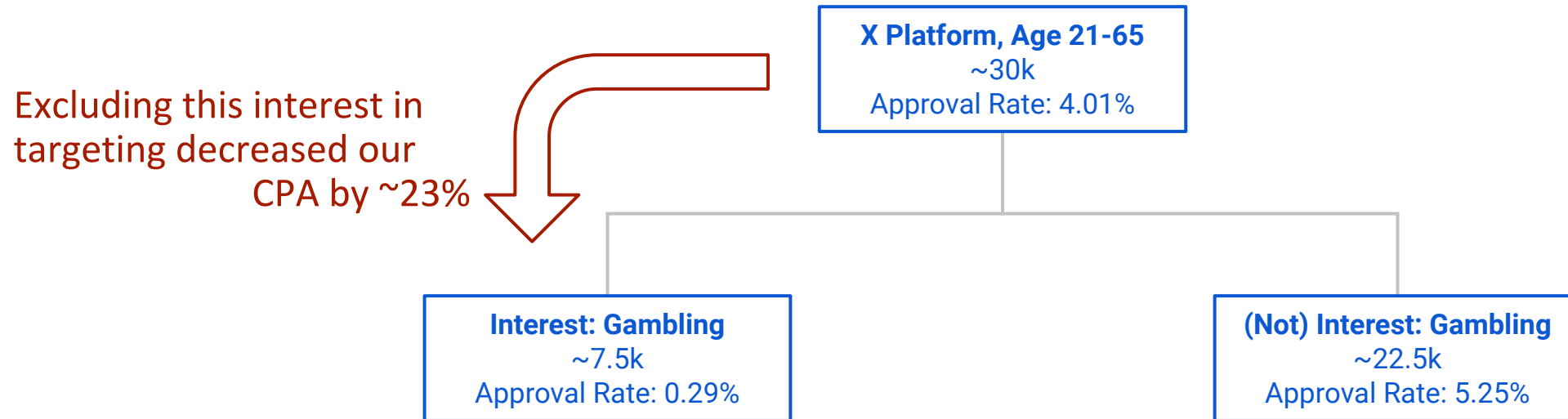


Example

TL;DR

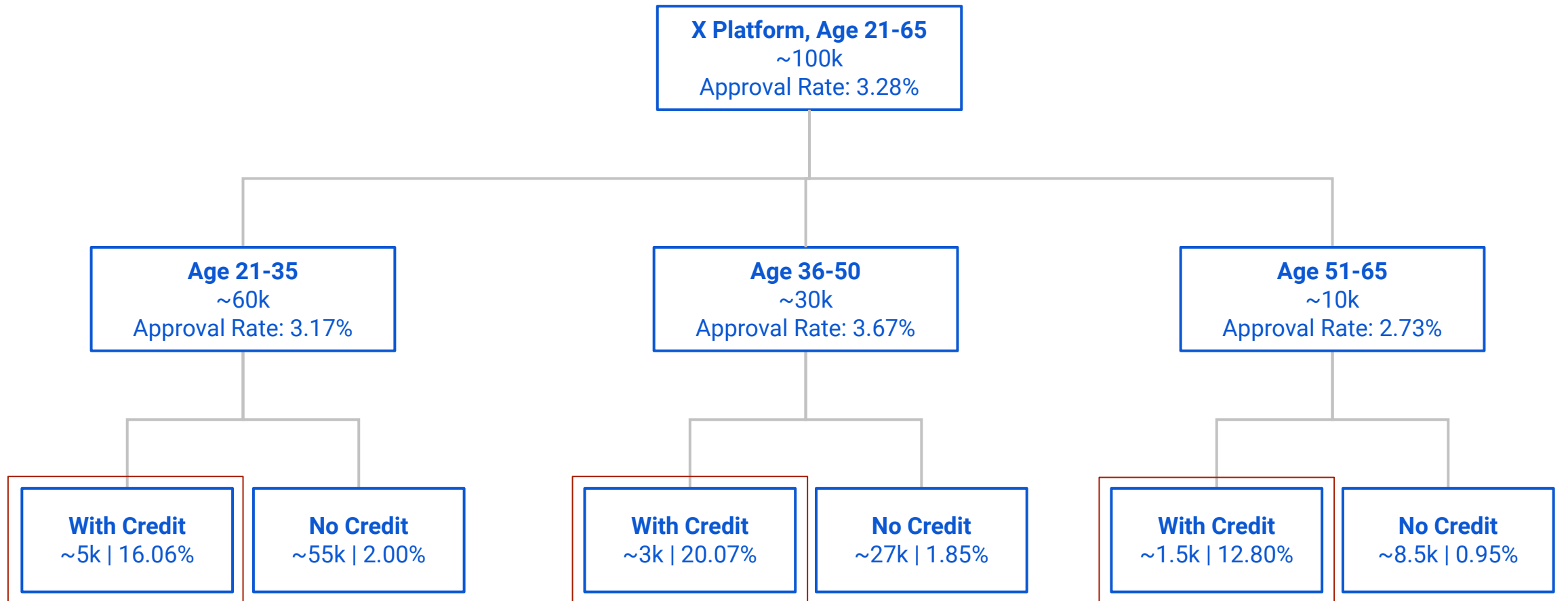
A lot of the marketing
decisions we were making
had ~zero impact

We looked for high variance segments using available Facebook targeting options; in this case, the action was exclusion leading to a decrease in CAC



Example

We also created lookalike audiences based on other identified high value segments



Creating a lookalike audience of “with credit” users created a new targeting segment with a CPA that was ~73% lower than average

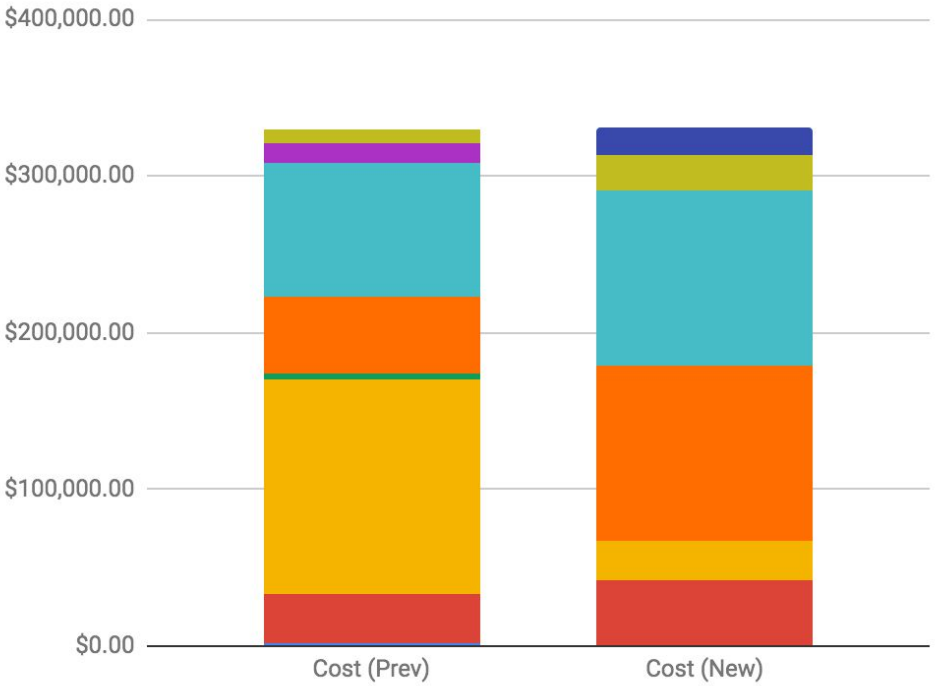
Example

Marketing Analytics in Practice

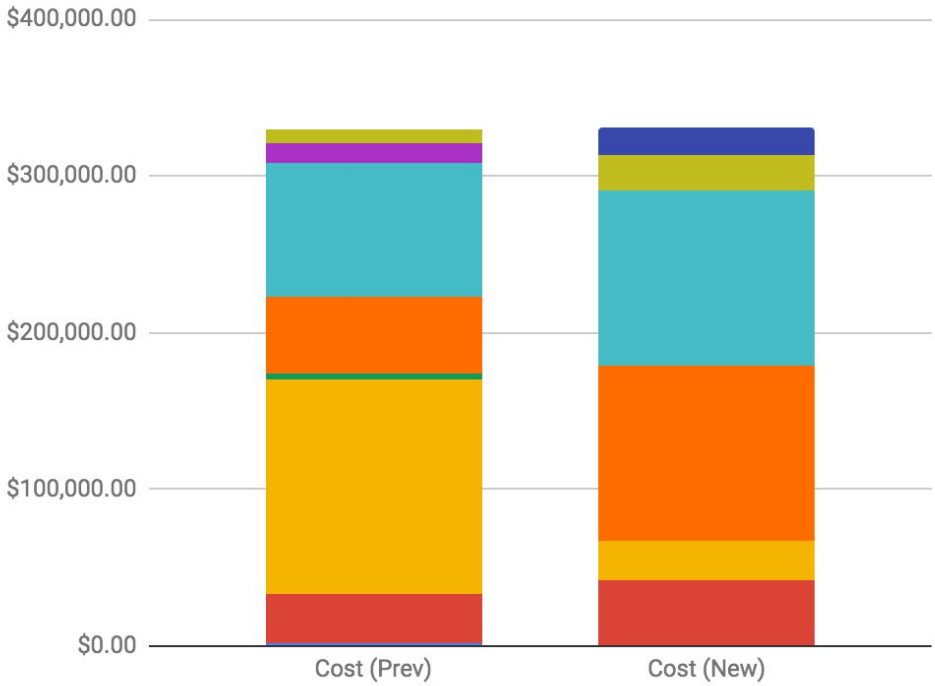
Management: Allocation



Then we allocated marketing spend to target high value customers to decrease overall CAC



Same spend



But 59% more customers
37% decrease in average CAC

Q&A



Exercise Time!



Spreadsheet Link: <https://urlzs.com/GZxyb>

Go to File > Make a Copy



Scenario: Creative Education Platform

Your company launched in 2016, producing learning content geared for creatives.

The business is a subscription (6 or 12 mo) where users access unlimited content. After registration, users access content free for the first 30 days

Content produced includes film, design, art, etc education.



Your company wants to understand whether the current marketing strategy is bringing in the right customers

Exercise 1: Calculate CLV and CAC



Your company wants to understand whether the current marketing strategy is working

Exercise 1: Calculate CLV and CAC



Your company wants to understand whether the current marketing strategy is bringing in the right customers

Exercise 2: Identify / create potential segments



Your company wants to know which customer types it should be going after next based on value

Exercise 3: Calculate CLV / CAC per segment



Your company wants to know which customer types it should be going after next based on value

Exercise 4: Make recommendations

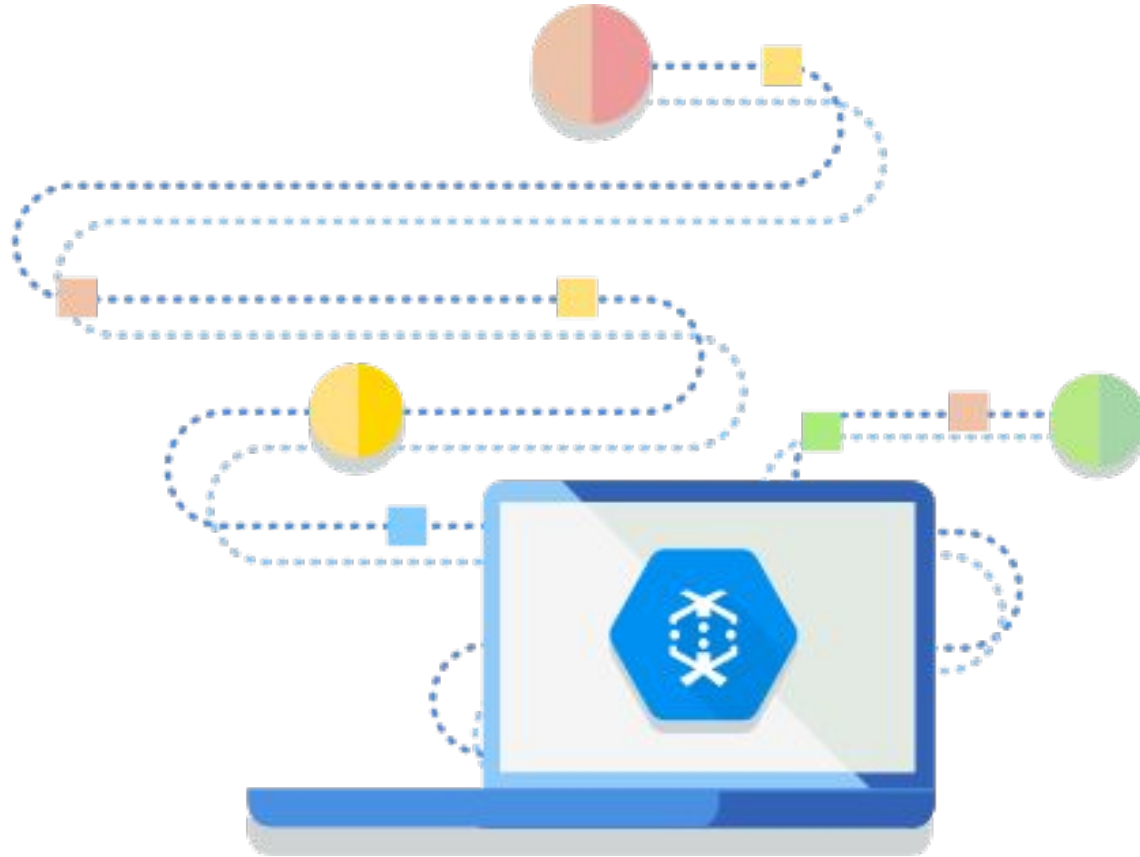


Thanks!

gdy@securitybank.com.ph
+63 917 548 0396



Attribution comes down to linking revenue generating actions (or customers) to costs and allocating them

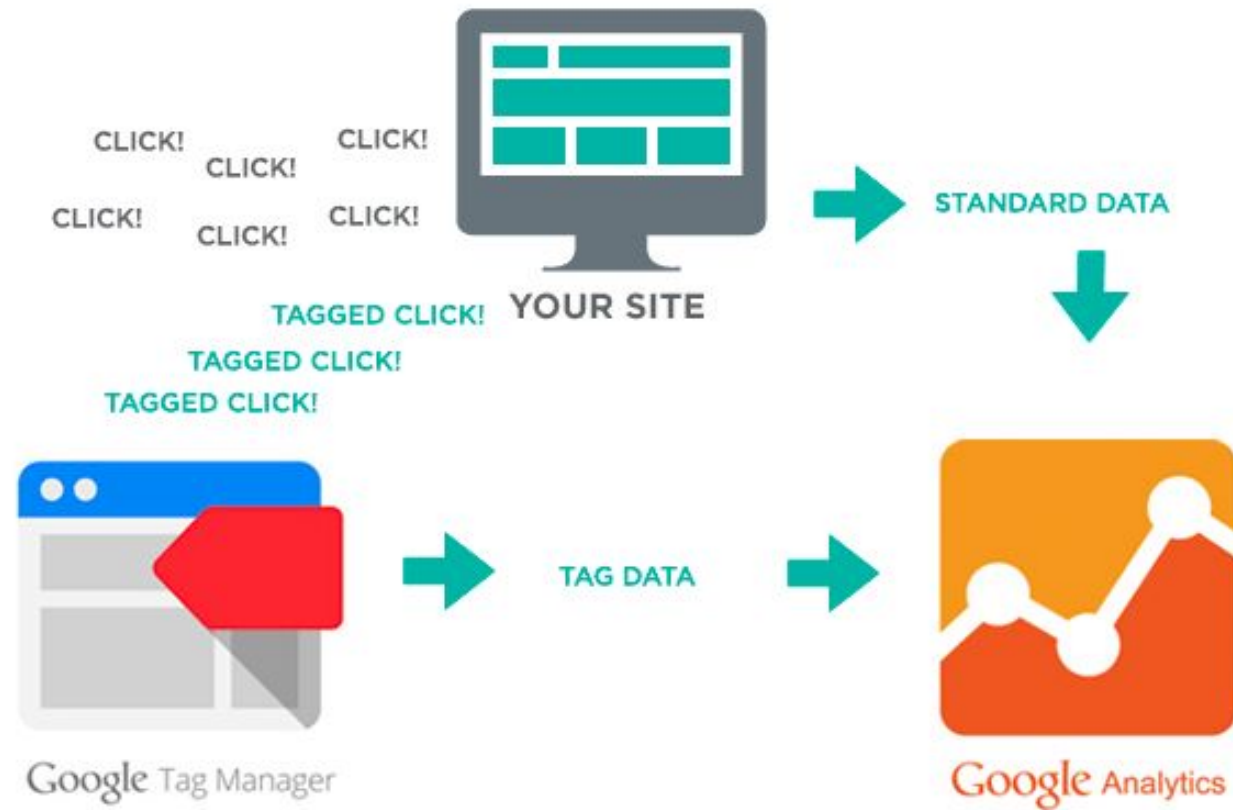


Attribution is basically

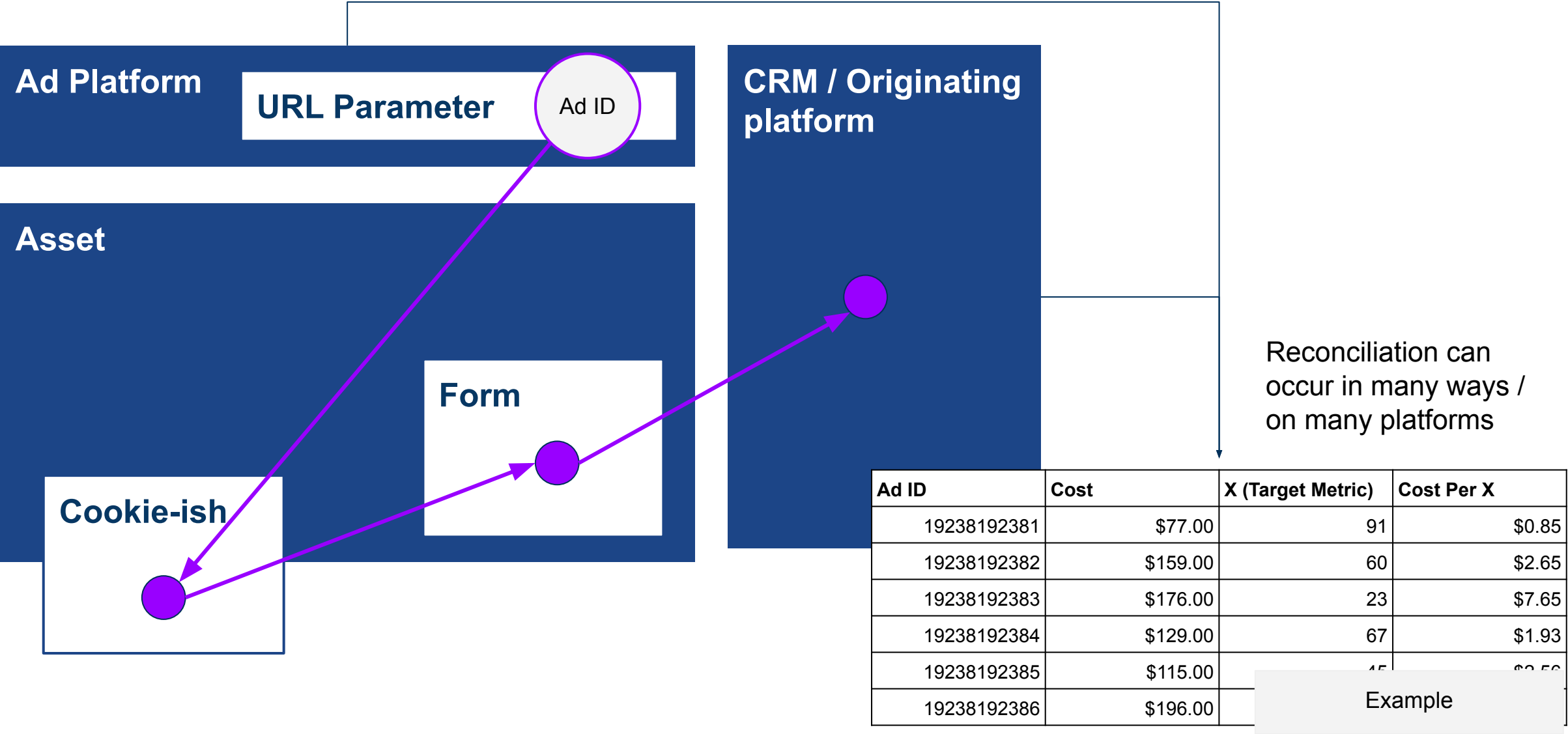
- 1) Setting up your data pipeline
- 2) Defining your match keys
- 3) Defining your target metrics
- 4) Linking your target metrics to costs

If your product's revenue-generating activity (or the activity you'd like to track) takes place completely online, it's (relatively) easy

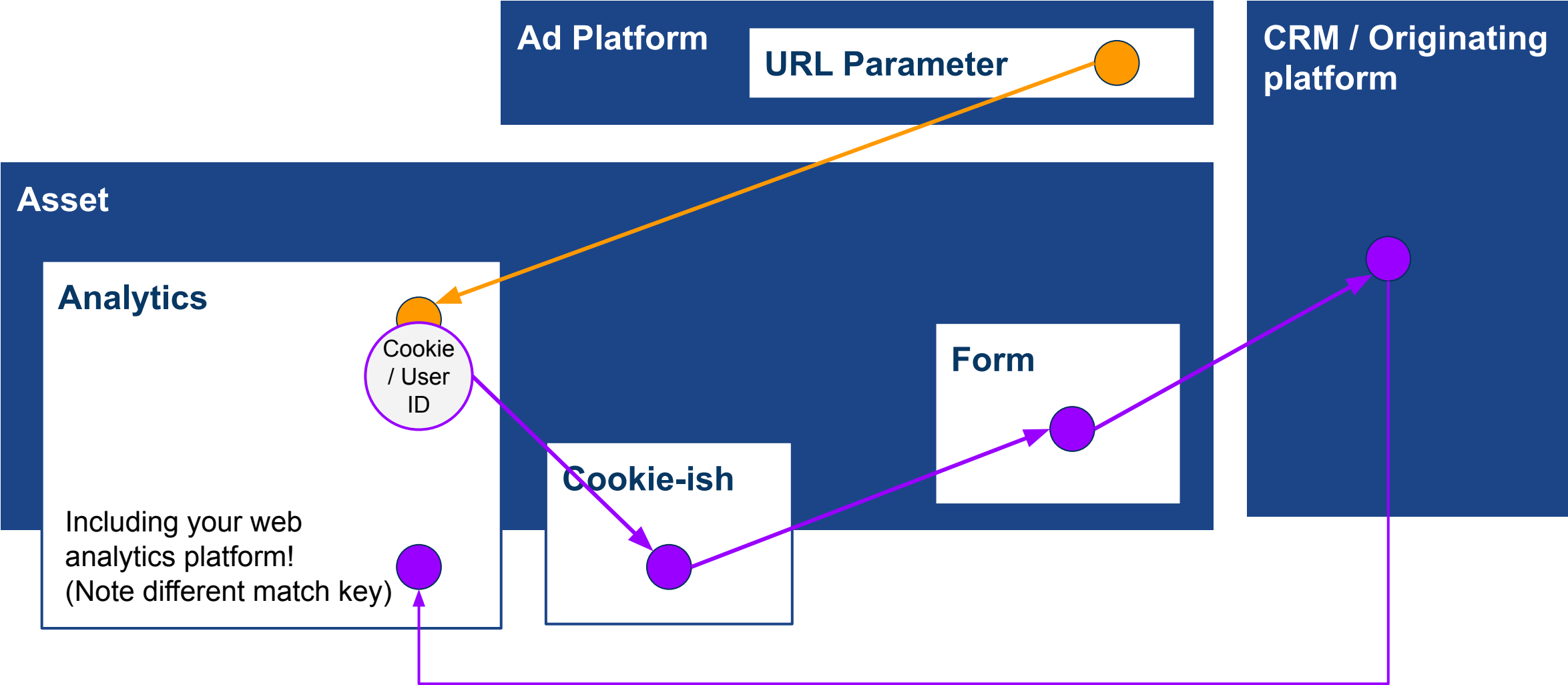
TAGGED EVENTS COLLECTED + SENT



But if your product's revenue-generating activity is offline, it amounts to linking cost and sales/revenue data using some type of match key

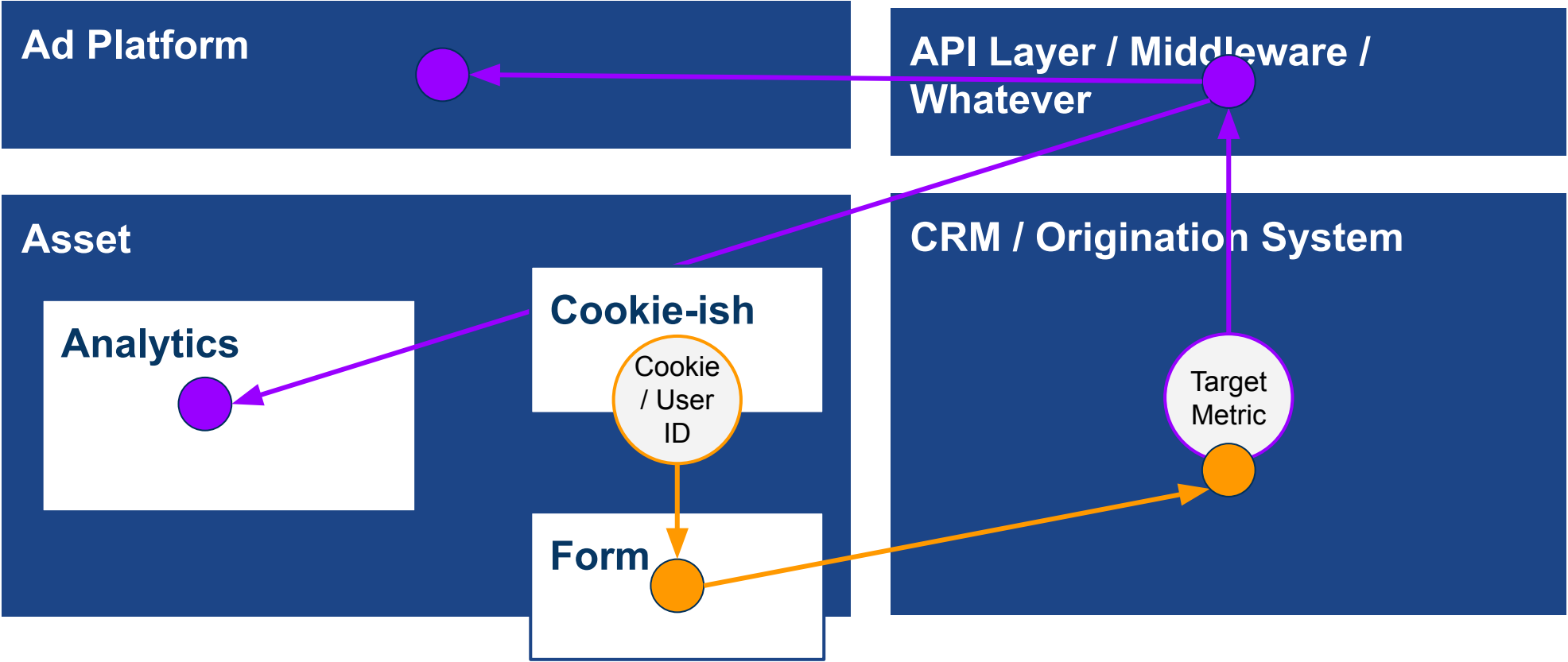


But if your product's revenue-generating activity is offline, it amounts to linking cost and sales/revenue data using some type of match key


































You can also do it the other way around (for some products) to answer the question, “How do we know what customers (who did X) this ad touched?”

Match keys include click IDs, cookie IDs, hashed customer identifiers, etc



Target Customers

	TV	Display Ad	Video Ad	Social	Email	Search	Brand web site	E-commerce Store	Physical Store	Purchase
										\$176
										\$268
										\$96
										\$64
										\$0
										\$128



Attribution Analytics

Simple models

Last Touch



First Touch



First and Last



Linear



Time Decay



Position Based (fixed)



Complex models

Custom or Rule-Based



Statistical Model



Note about attribution:
All views are technically imperfect
but provide valuable data.

Just remember to calibrate for lift.