# monopolio

by



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Introduction to DD360/Monopolio

Our problem: Who is using the website?

Our data: User click-data & active users

Our strategic approach: User segmentation through click-data

Key Discoveries, **user personas: Millennials, Dreamers, Clickers, Buyers** 



# About monopolio

### Your Digital Real Estate Partner

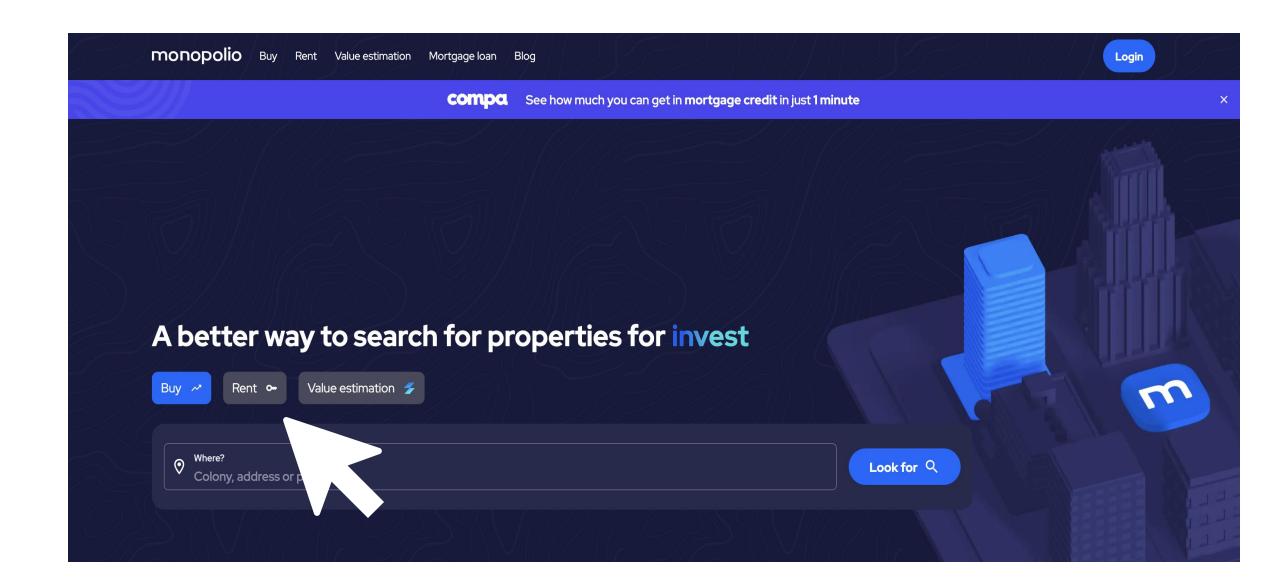
Monopolio is a real estate platform that allows users to buy and rent properties, connect with sellers, purchase appraisals, and estimate potential mortgage options

The Vision: Create a proptech ecosystem that enhances transparency, efficiency, and user experience through datadriven insights and seamless technology.

Mexico

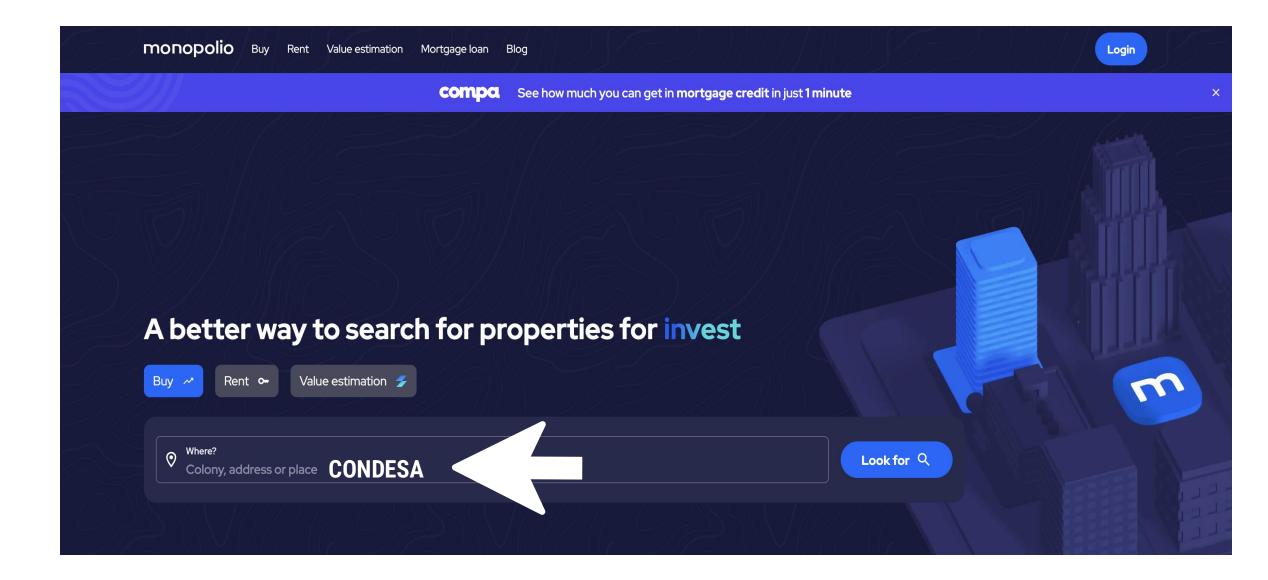


1.- Looking for a place to rent



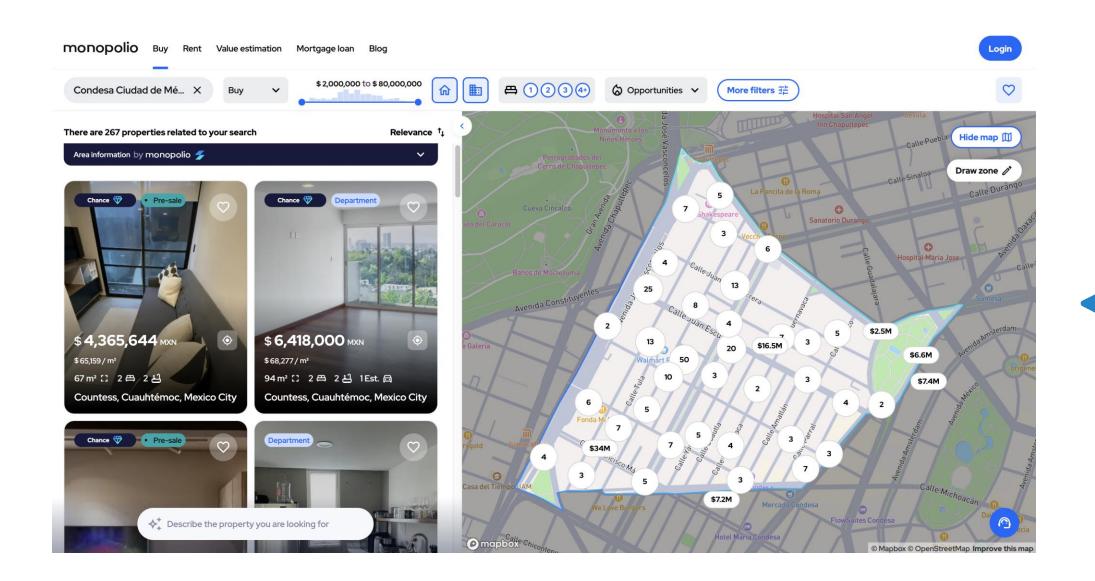
Navigating Monopolio monopolio

- 1.- Looking for a place to rent
- 2.- Indicating the neighborhood where they want to live

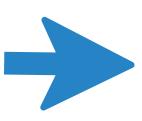


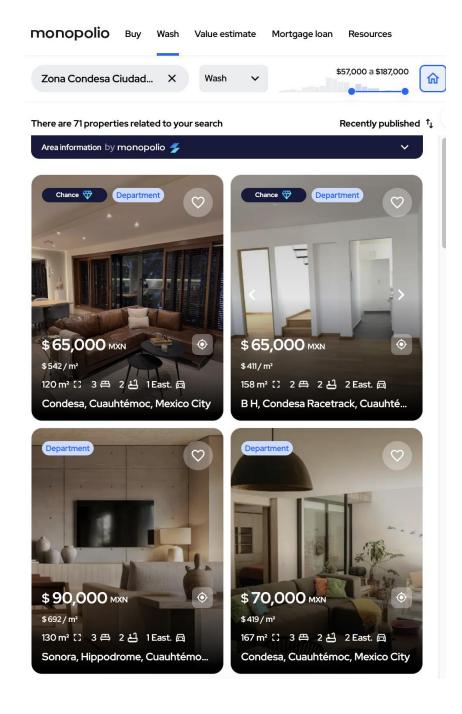
### monopolio

- 1.- Looking for a place to rent
- 2.- Indicating the neighborhood where they want to live
- 3.- Exploring the available rental properties

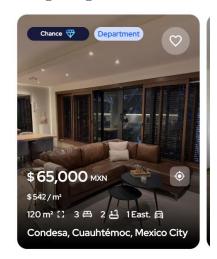


- 1.- Looking for a place to rent
- 2.- Indicating the neighborhood where they want to live
- 3.- Exploring the available rental properties
- 4.- Selecting a listing they like
- 5.- Contacting an agent

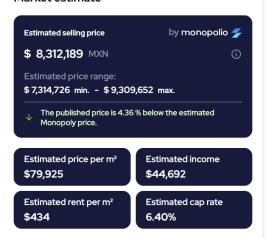




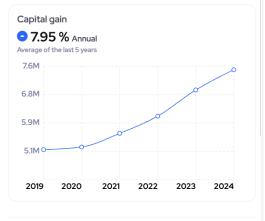
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#### Market estimate



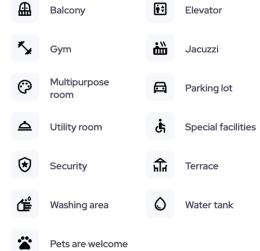
#### Price history



#### **Property features**



#### Amenities



Enjoy this spectacular apartment with a panoramic view of almost 300 meters from its 10th floor, you can see Chapultepec Castle, among others, and with the best amenities in an impeccable Element Condesa building. Ideal for investors, it is sold rented with a tenant, rent \$ 35,000. Great location. Luxury finishes, terraced balcony with a beautiful view, controlled access with 24-hour surveillance. 2 spacious bedrooms with closets, one of

See more

#### Area metrics



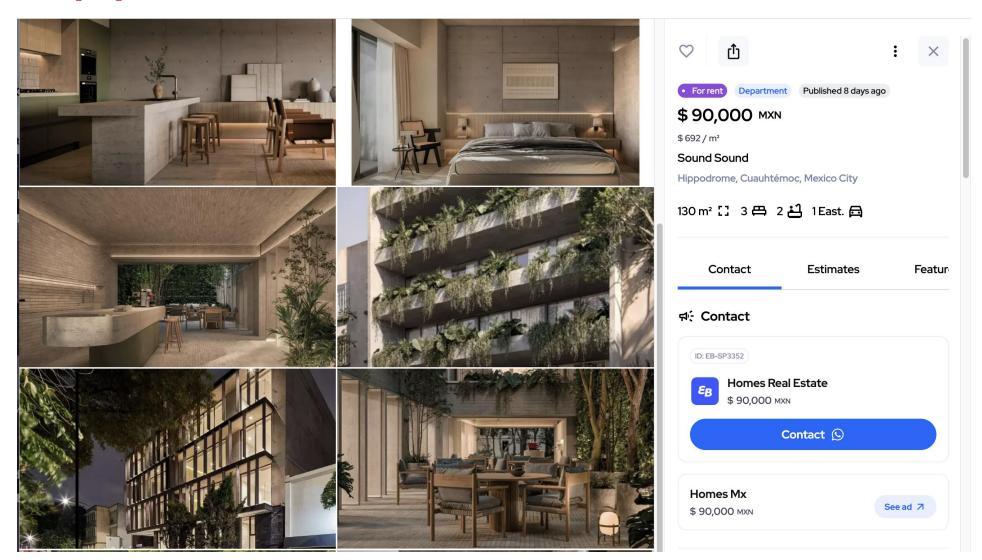
monopolio







- 1.- Looking for a place to rent
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# **Our Challenge**

### Understanding how users interact with monopolio

Monopolio seeks to **understand its user base and how visitors interact with the platform** to optimize the user experience, improve engagement, and drive conversions.

# **Our Proposal**

User segmentation through **unsupervised learning** based on click behavior, **combined with explainable ML** techniques to interpret and characterize the clusters.



# **Data overview**

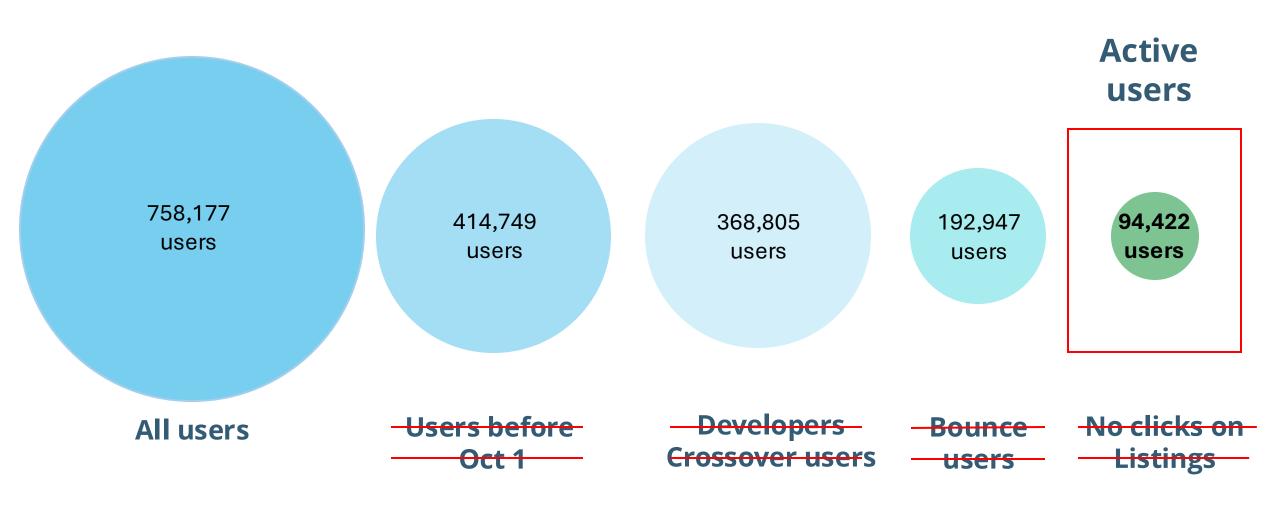
### User Data: ~10 million clicks

Userid	Type of click	<b>Event time</b>		
	autocomplete_selected	2024-10-21 22:21:54		
	search_selected	2024-10-21 22:21:57		
	gridView_viewed	2024-10-21 22:22:00		
	filters_applied	2024-10-21 22:22:04		
2161	search_results	2024-10-21 22:22:04		
2101	listing_selected	2024-10-21 22:23:58		
	gridView_viewed	2024-10-21 22:25:09		
	filters_applied	2024-10-21 22:25:13		
	search_results	2024-10-21 22:25:13		
	listing_selected	2024-10-21 22:30:12		

### Properties Data: ~1 million properties

Property id	Latitude	Longitude	Property type	Price sale
0	20.52	-103.48	house	\$2,558,162
1	21.03	-89.56	house	\$4,140,462
2	20.69	-100.45	apartment	\$4,100,188
3	25.73	-100.38	house	\$3,888,696
4	25.64	-100.38	apartment	\$11,482,160

# **Final Dataset**



# **Methodology Overview**



**Preprocessing**Feature Engineering



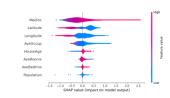


**Model** K-Means Clustering



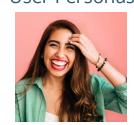


**Explainability** SHAP Values





Customer Insights
User Personas



# Preprocessing





**Preprocessing**Feature Engineering



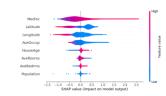


**Model** K-Means Clustering





**Explainability**SHAP Values







User Personas

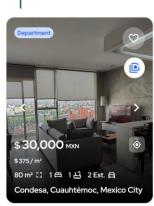


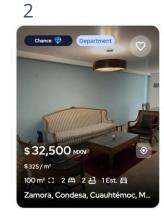


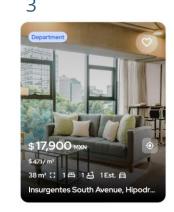


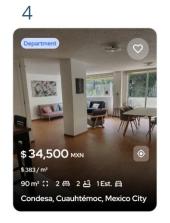
# Feature Engineering for Ana













Average Price

\$28,725 MXN



Standard deviation 7447

Radius of listings

3.5 km

Weekdays

2



# 1

# Preprocessing: Feature engineering

# **Engagement** metrics



Browsing time

Browsing sessions

Weekdays vs weekends

# **Property** interactions



Property Type

Location

Average distance

Rent or Buy properties

# Price metrics



Average price

Standard Deviation

Luxury score

# Filter applications

Essential characteristics

Extra characteristics

**Amenities** 

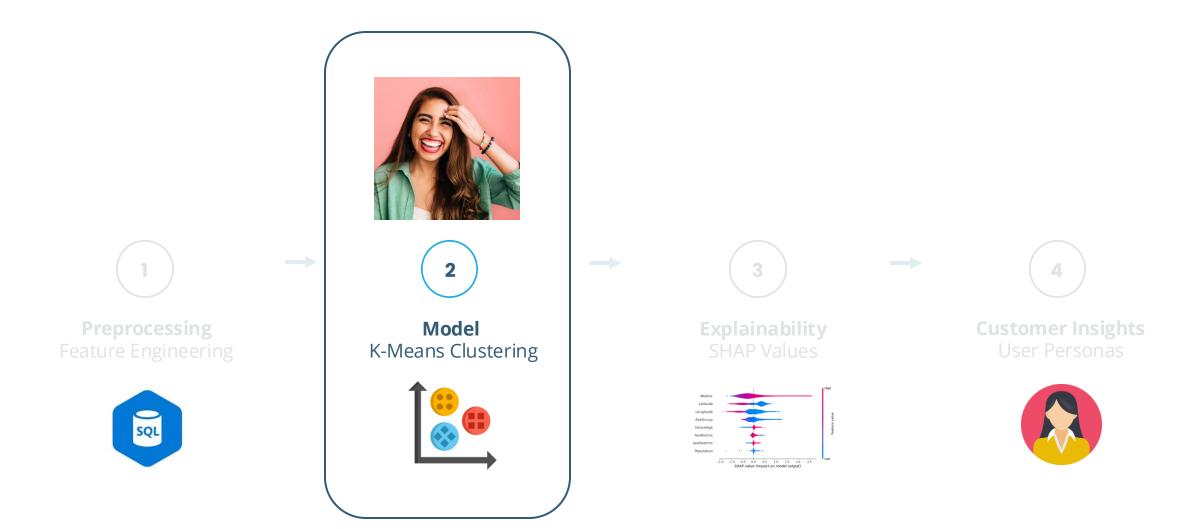
Opportunities



# Model input data: 94,422 users

Userid	Average price	Standard dev. price	# Visits	Unique properties	Dummy apartment	Dummy house	Filters on # rooms/baths	Filters on property type/ surface	Filters on amenities	Filters on investment value	% weekdays		
1	\$7,400 K	0	17	1	1	0	1	0	0	0	100	•••	
2	\$1,946 K	107K	20	2	1	0	0	0	0	0	67%	•••	
3	\$2,960 K	0	15	14	1	0	2	1	1	1	0%	•••	
5	\$6,800 K	0	8	1	0	1	0	0	0	0	45%	•••	
7	\$2,564 K	0	40	1	1	0	3	1	1	1	0%	•••	
8	\$3,980 K	0	1	54	0	1	0	0	0	0	100%	•••	

# K-Means Clustering







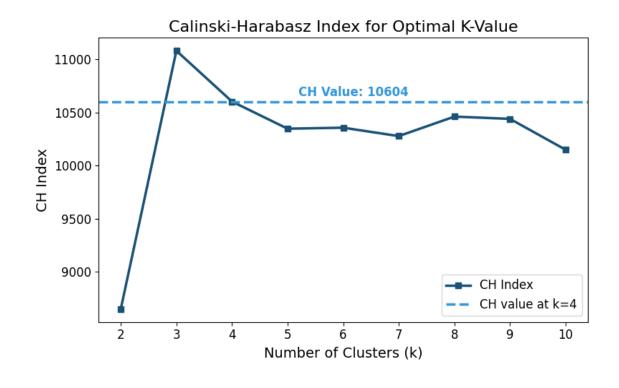
# K-Means Clustering

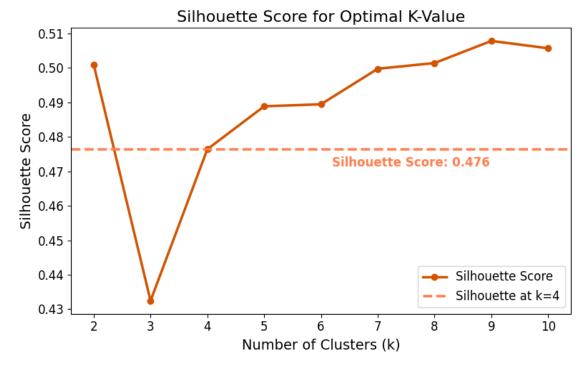
**Goal**: Applied K-means clustering to classify users based on behavioral patterns

**Tuning**: Determined optimal number of clusters using:

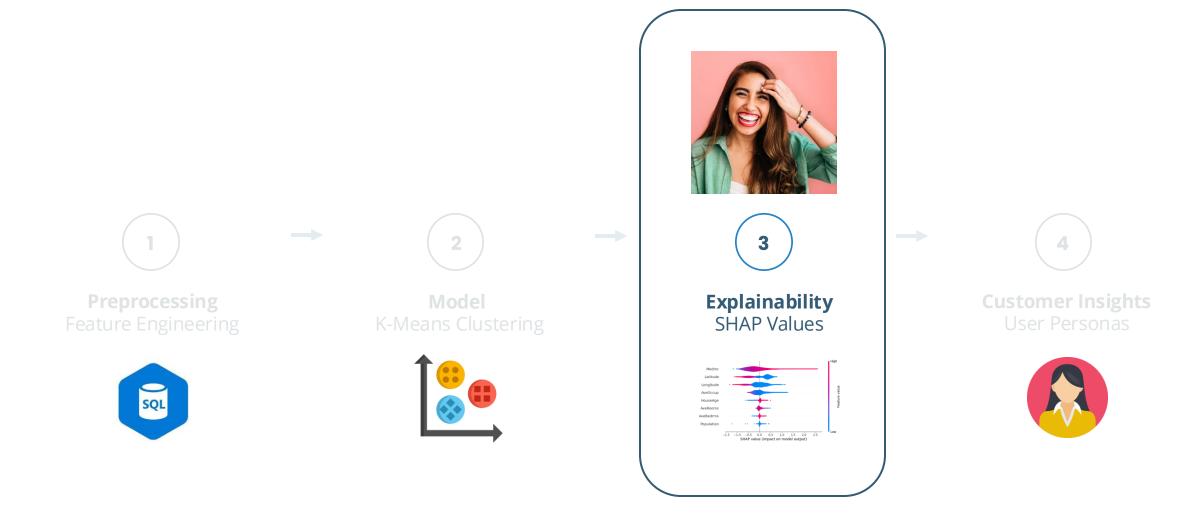
- Calinski-Harabasz Index to assess between vs. within cluster spread
- *Silhouette* score analysis to evaluate cohesion vs. separation (per point)

**Output**: Grouped users according to web browsing behaviors, creating user clusters



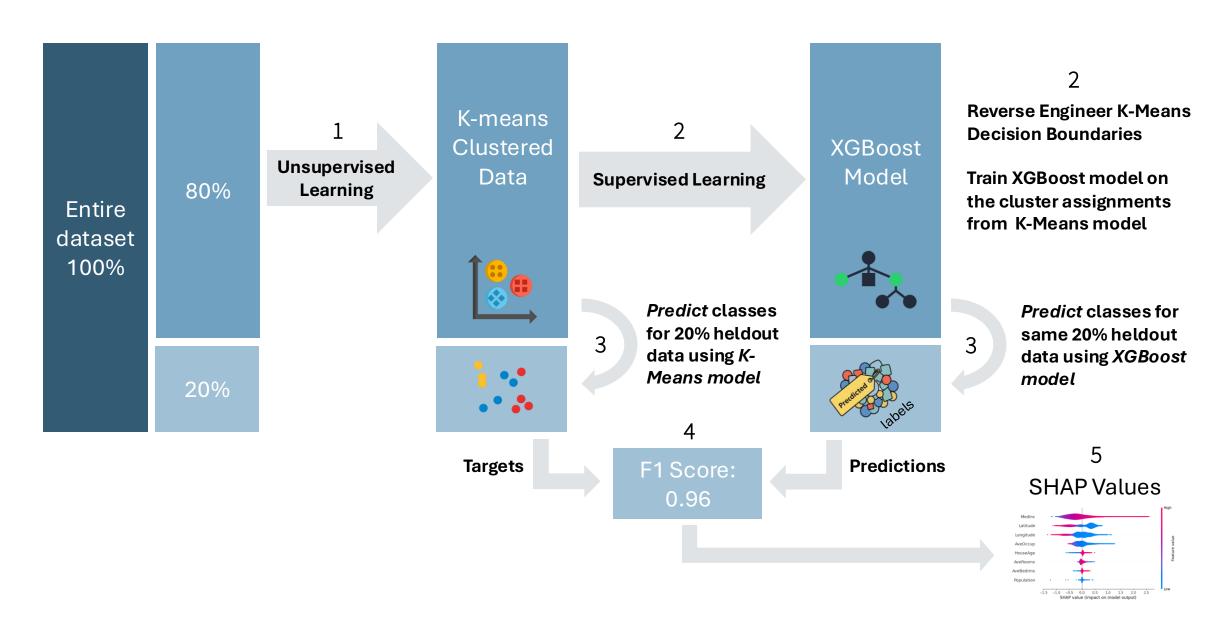


# **Explainability**



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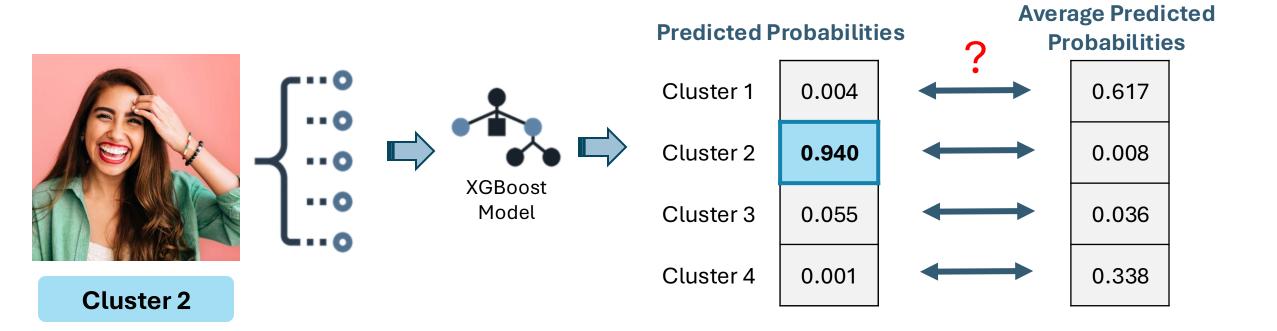
### K-Means and XGBoost Model Training Pipeline







### Local Interpretability: Explaining Ana's Predictions

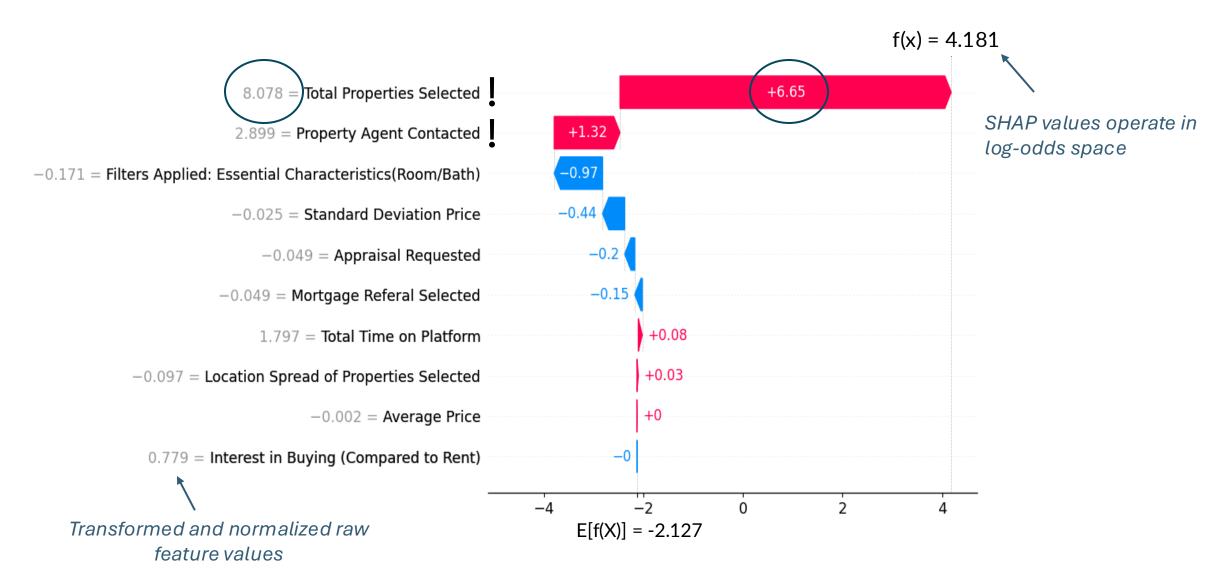


SHAP values describe each feature's contribution towards shifting individual predicted probabilities away from expected predicted probabilities.

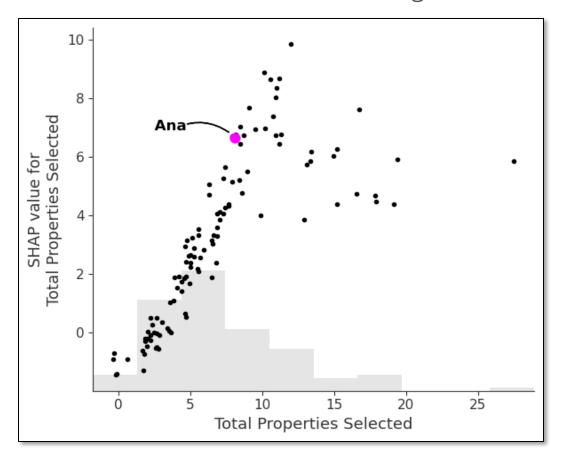


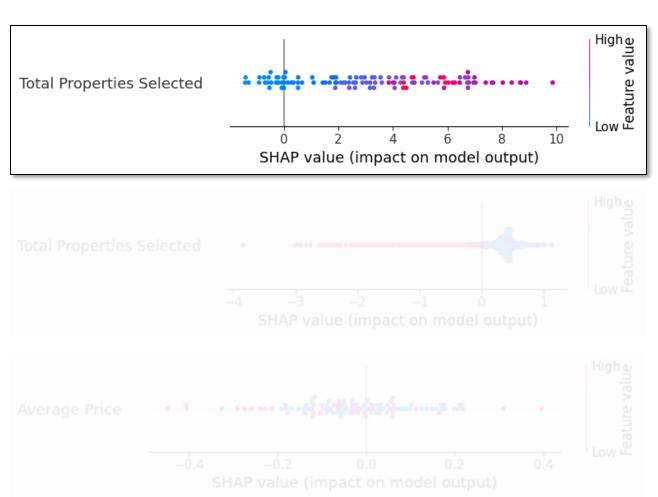


# Local interpretability: Explaining Ana's Cluster 2 SHAP Values.

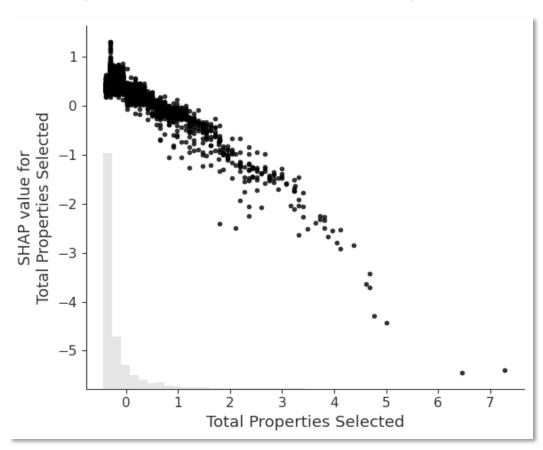


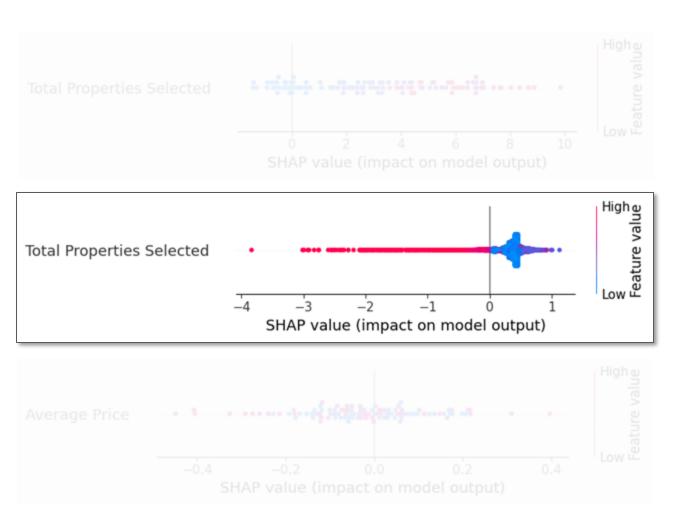
### Positive Effect on Cluster 2 Assignment



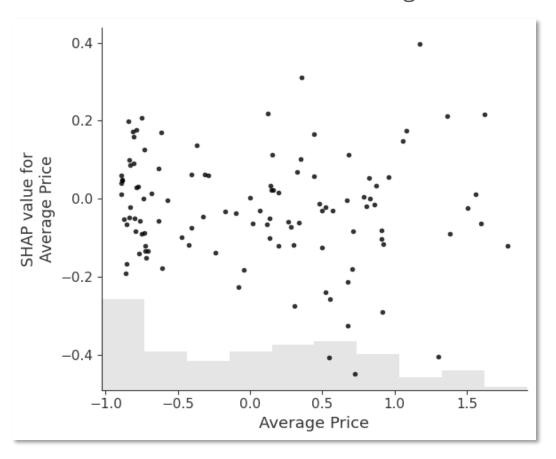


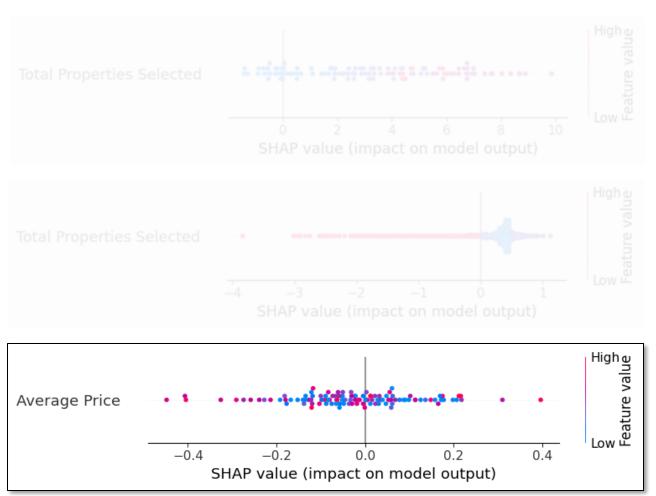


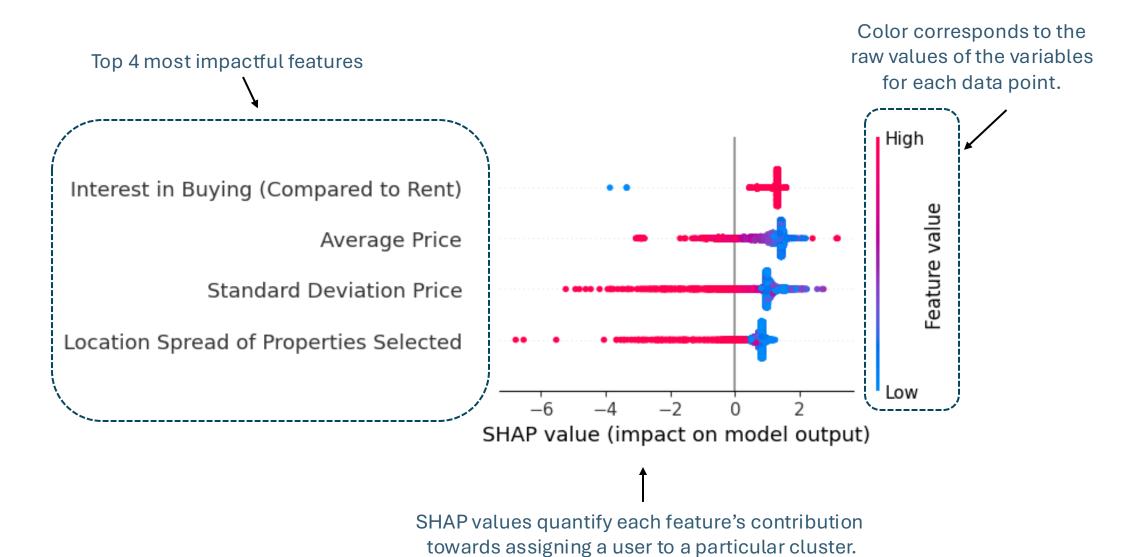




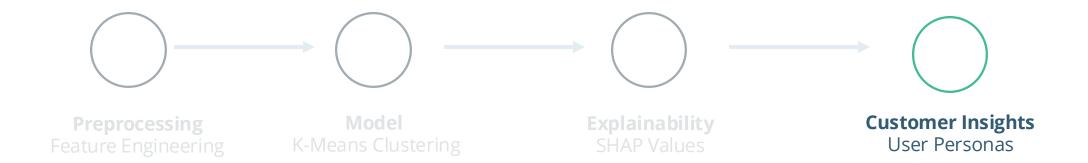
### Neutral Effect on Cluster 2 Assignment







# **Methodology Overview**





**Jose: ONE TIME BUYER** 



**Ana: CLICKER** 



**Juan: DREAMER** 



**Sofia: MILLENNIALS** 

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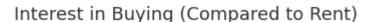
# Cluster 1: 53.4% of users: One Time Buyers



**Jose: ONE TIME BUYER** 

### Who is José?

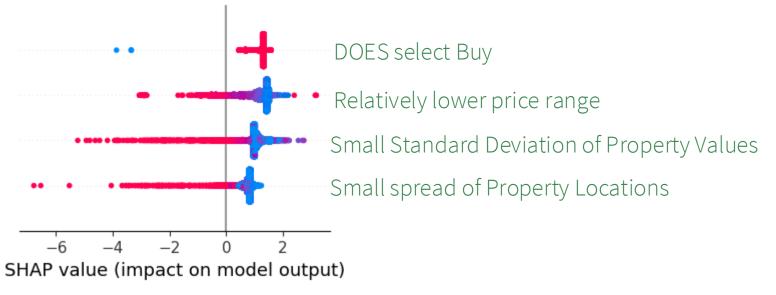
Jose is ready to buy! He has a specific location in mind and a specific price range he's looking for. The average price has higher variability, but be assured, Jose is NOT renting.



Average Price

Standard Deviation Price

Location Spread of Properties Selected



High

Feature value

Low

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High

Feature value

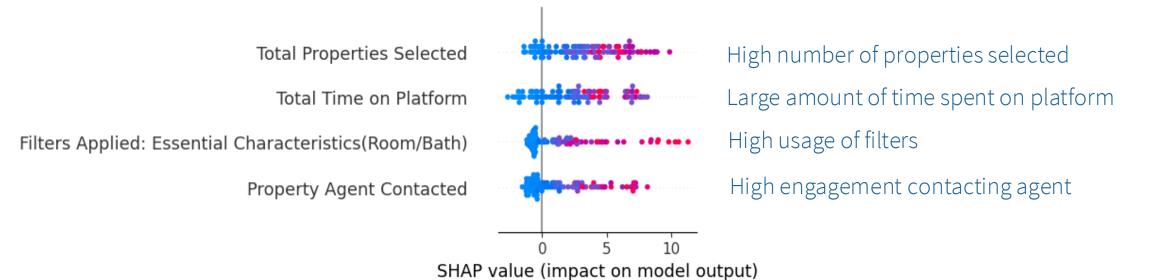
### Cluster 2: 0.7% of users: Clickers



**Ana: CLICKER** 

### Who is Ana?

Ana is an active Monopolio user! She uses all the filters and clicks on a wide range of property types. The properties are not limited to a particular neighborhood. She probably contacts a realtor in the platform.



### Cluster 3: 8.3% of users: The Dreamers



**Juan: DREAMER** 

### Who is Juan?

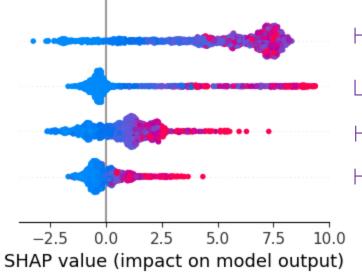
Juan is looking for properties all across Mexico and he doesn't care what the price is.

Standard Deviation Price

Location Spread of Properties Selected

Average Price

Total Properties Selected



High Standard Deviation of Prices

Large Range of Property Locations

High Average Price

High Number of selected properties

High

Feature value

Low

### Cluster 4: 37.5% of users: Millennials



**Sofia: MILLENNIAL** 

### Who is Sofia?

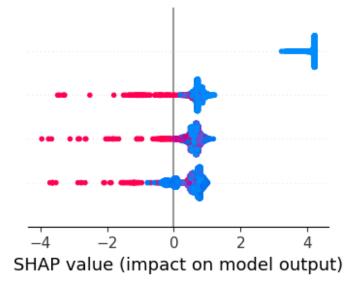
Sofia is exclusively a renter looking primarily in a small centralized area. This area has a relatively lower average price. Ana is NOT looking to buy.

Interest in Buying (Compared to Rent)

Total Time on Platform

Location Spread of Properties Selected

Average Price



Sole interest in renting

Low amount of time spent on website

Low range of Property Locations

Low average prices



# From user personas to actionable insights





Targeted marketing campaigns



Customer Retention & Personalization

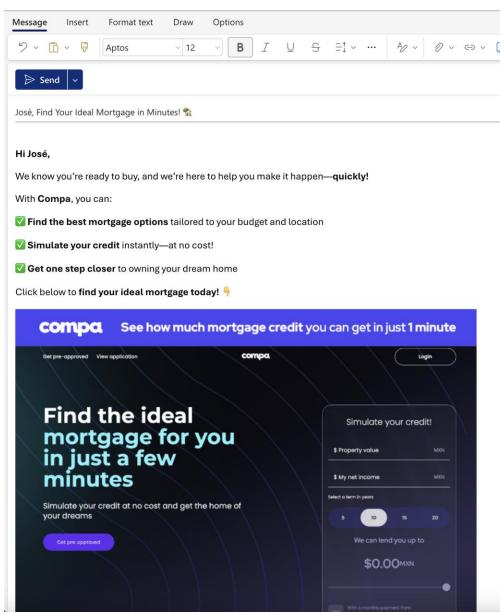


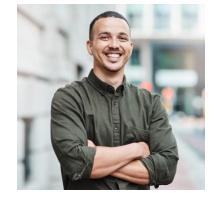
Pricing & Monetization Strategies



A/B Testing & Experimentation

# Targeted marketing campaigns







### Who is José?

Jose is ready to buy! He has a specific location in mind and a specific price range he's looking for. The average price has higher variability, but be assured, Jose is NOT renting.

(Existing DD360 product)

# Thank you! monopolio

