

# Airbnb Dynamic Pricing Recommendation Engine

## Objective

This project analyzes historical Airbnb data to recommend optimal listing prices by considering city, room type, guest reviews, superhost status, and seasonal demand patterns. The goal is to help hosts maximize revenue while ensuring competitive pricing and guest satisfaction.

## Tools & Technologies

- Python (Pandas, Scikit-learn, Matplotlib/Seaborn)
- Power BI for dashboards & visualization
- Excel for initial exploration

## Approach

1. Data Preparation – Cleaned and structured Airbnb dataset.
2. Model Building – Developed a regression-based pricing prediction model.
3. Predicted Pricing – Generated Predicted Prices and calculated Price Gaps against actual values.
4. Visualization – Designed interactive Power BI dashboards for pricing trends, city-wise comparison, and host performance insights.

## Key Insights

- City-Wise: London and Paris have the highest predicted prices, while Budapest and Berlin remain lower-cost markets.
- Room Type: Entire home/apartment listings command premium pricing compared to private/shared rooms.
- Superhost Impact: Superhost status shows a moderate positive impact on pricing and guest trust.
- Mispricing Alert: Cities like Barcelona reveal large price gaps (predicted > actual), suggesting underpriced listings.
- Guest Satisfaction: Higher cleanliness and review scores are correlated with higher achievable prices.

## Recommendations for Airbnb Hosts & Platform

### ◆ Optimized Pricing by Location

- Adjust prices upwards in high-demand cities (London, Paris).
- Correct underpricing in cities with large negative price gaps (e.g., Barcelona).
- Keep Berlin and Budapest competitively priced to attract demand.

### ◆ Room Type Strategy

- Position Entire Homes/Apartments as premium offerings.
- Keep Private/Shared Rooms at competitive entry-level prices to capture budget travelers.

### ◆ Superhost Advantage

- Encourage hosts to achieve and maintain Superhost status, since these listings show higher trust and willingness-to-pay.

### ◆ Dynamic Adjustments

- Apply weekend premiums and seasonal adjustments to better capture demand spikes.
- Use dynamic pricing engines instead of static pricing.

### ◆ Data-Driven Alerts

- Introduce alert systems to notify hosts when their listing is significantly underpriced compared to predicted values.

## Recommended Price Example

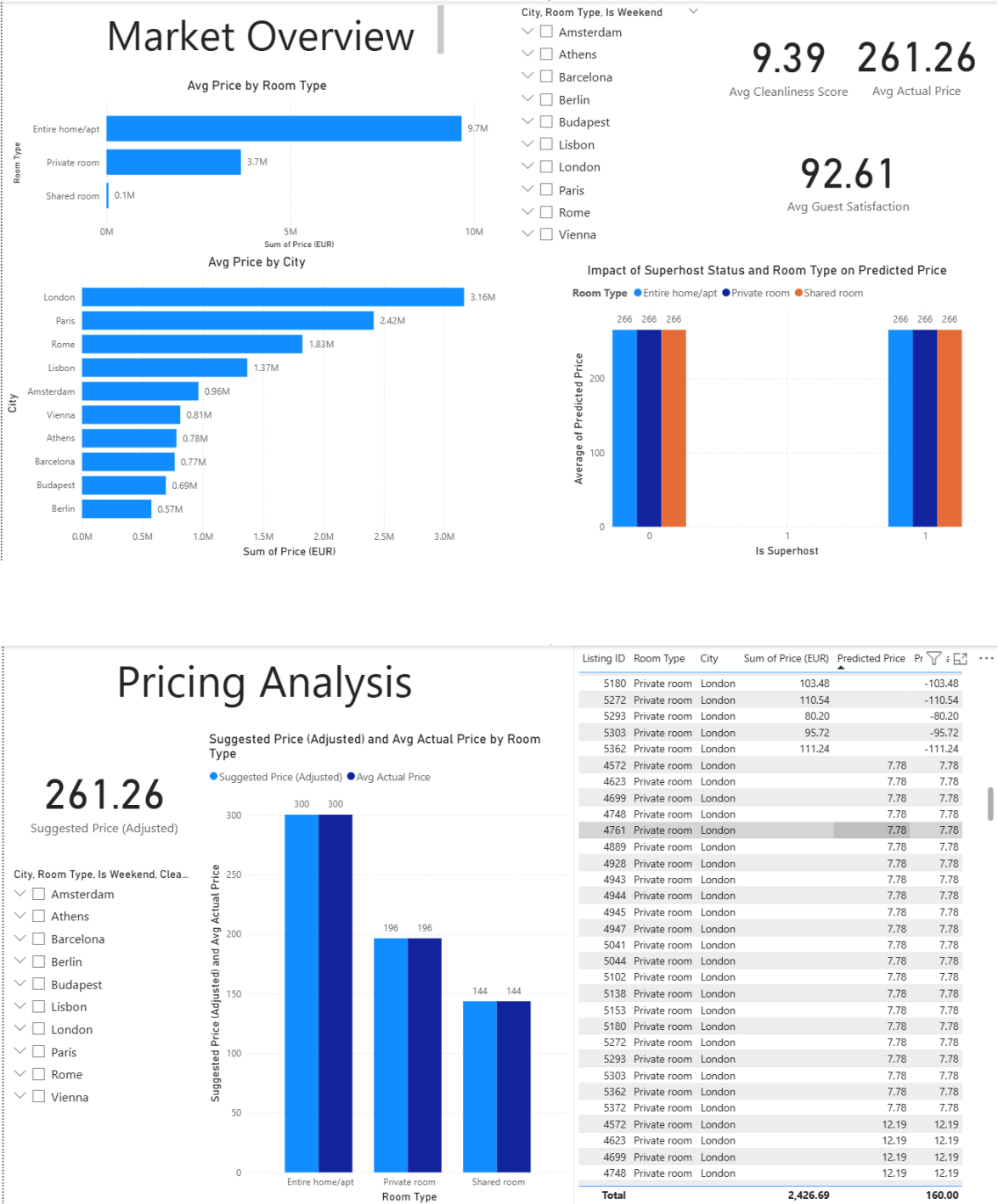
For one analyzed listing, the model suggested an optimal price of €687.42.

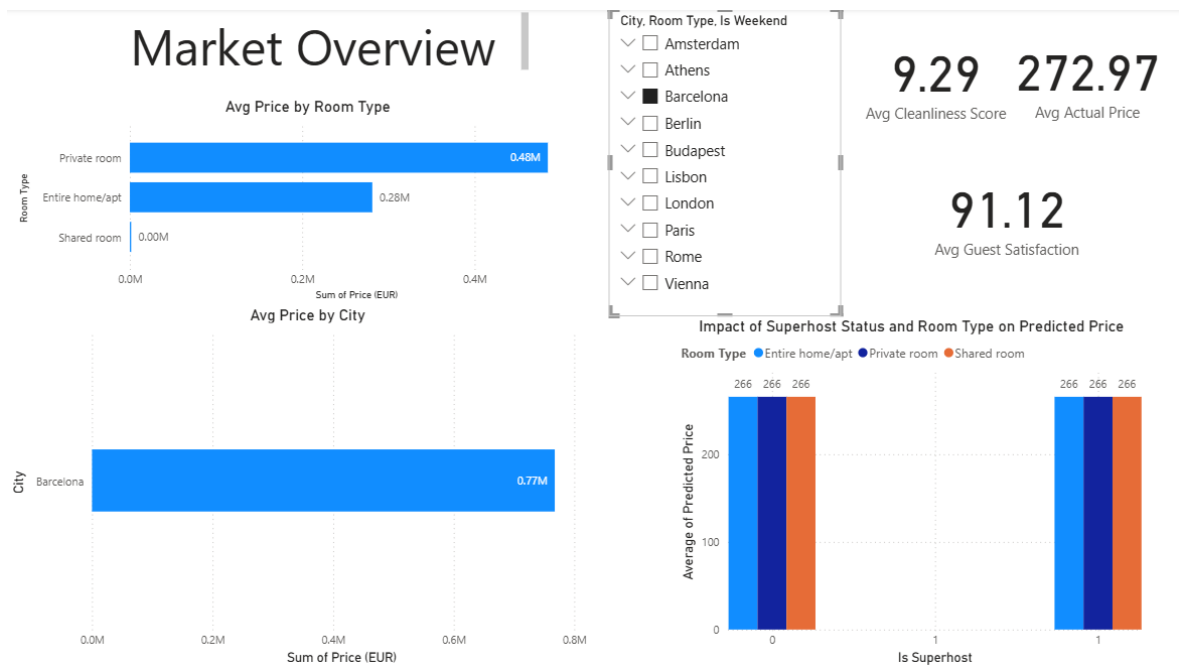
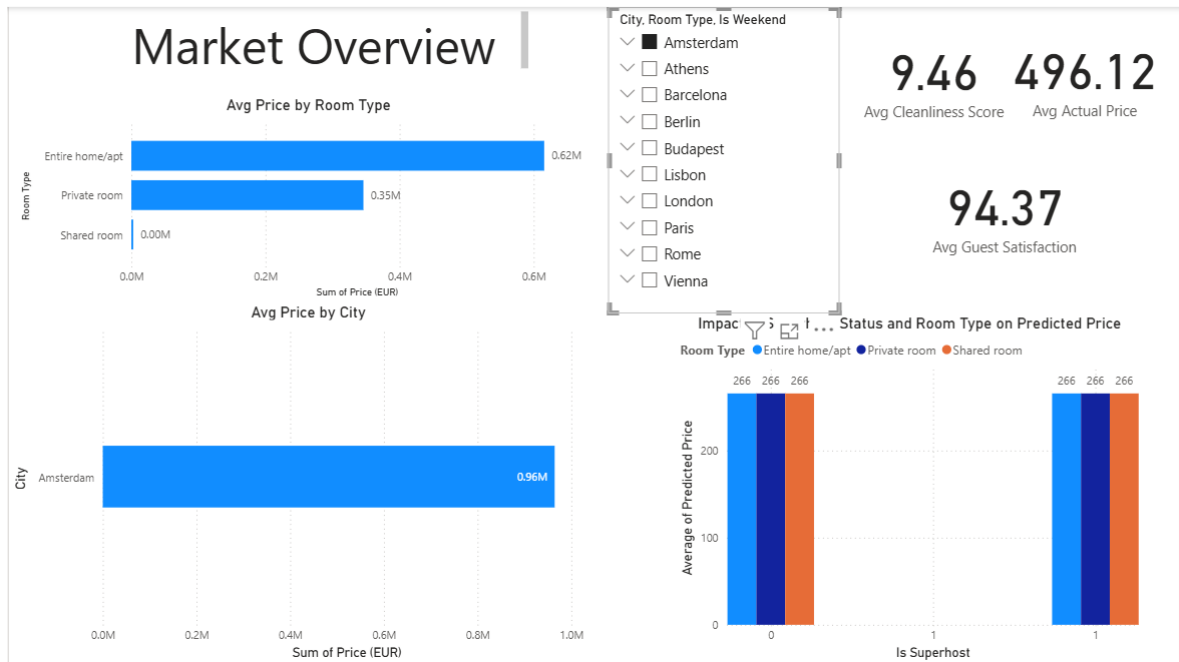
- How?
  - The model factored in location, room type, guest capacity, cleanliness score, and reviews.
  - Compared with market averages and adjusted for demand patterns.
- Why?
  - Ensures competitive positioning against similar listings.
  - Maximizes host revenue without reducing guest booking probability.
  - Aligns pricing with quality signals (reviews, Superhost, room type).

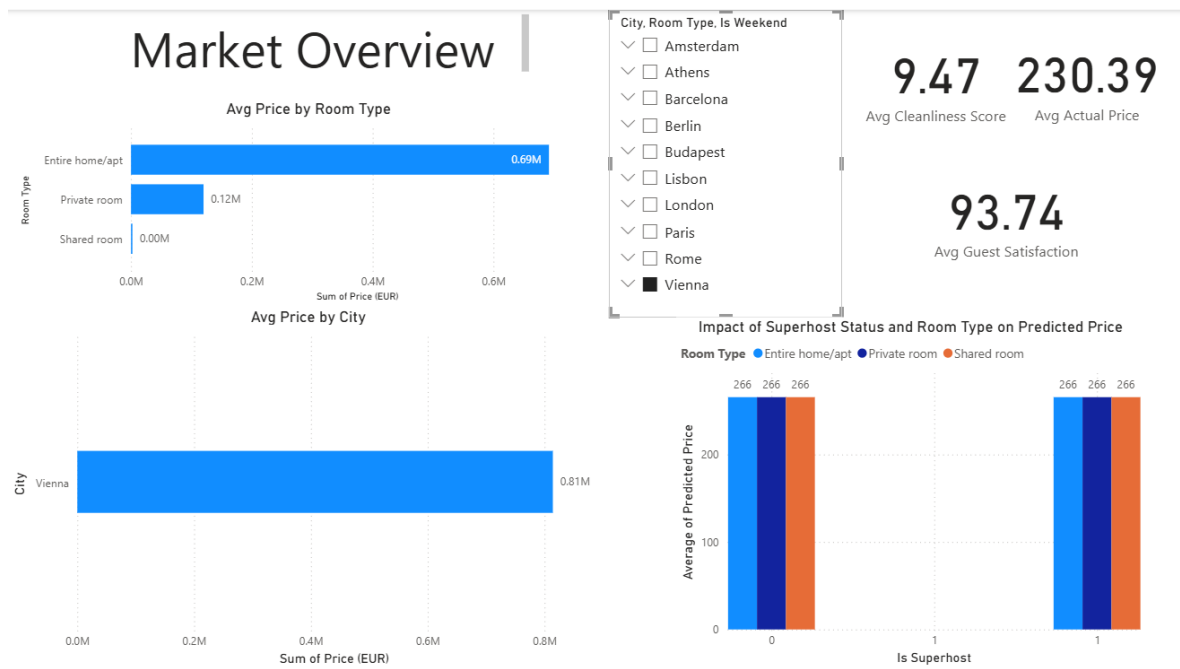
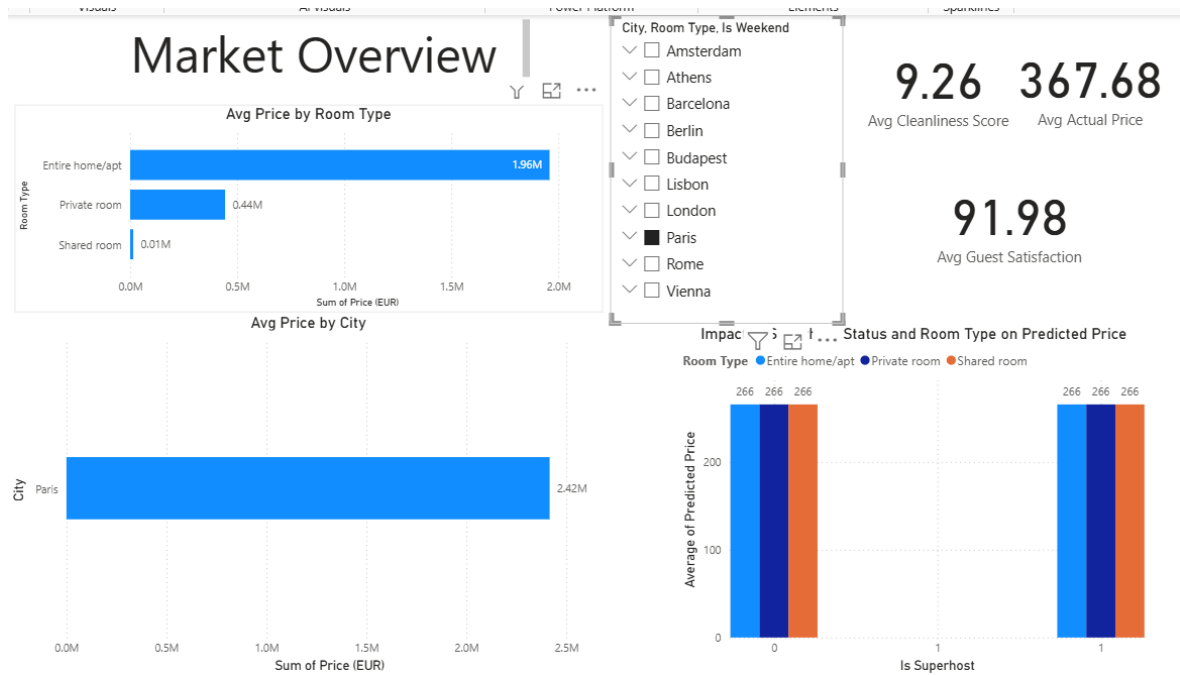
# Dashboard Snapshots

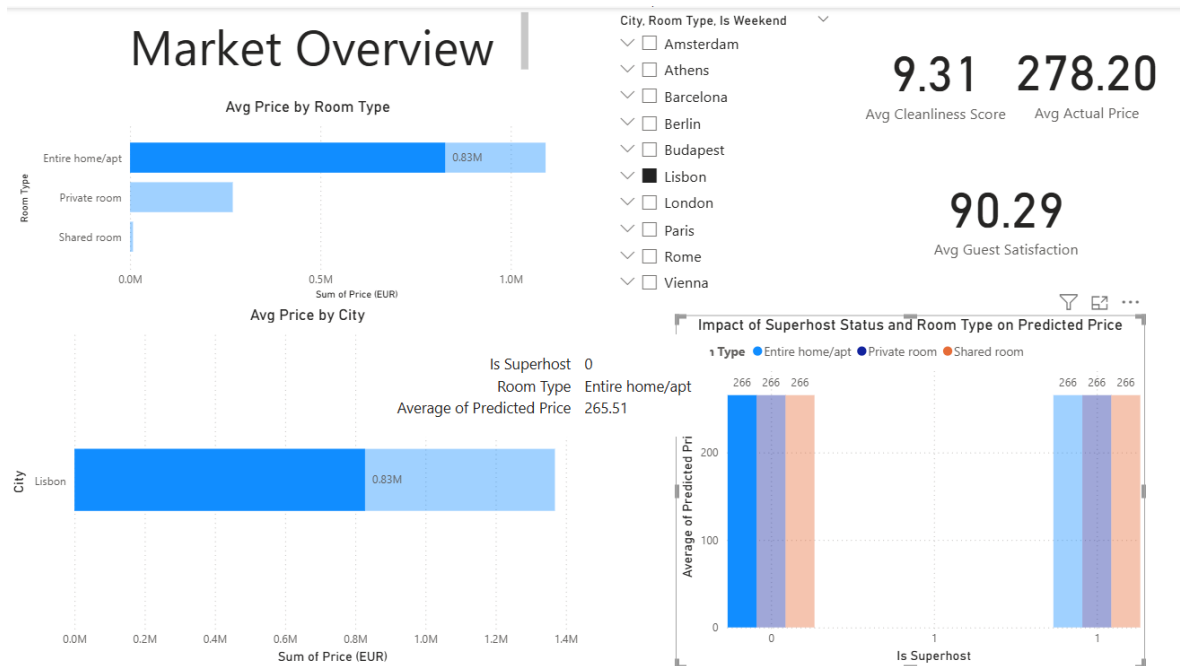
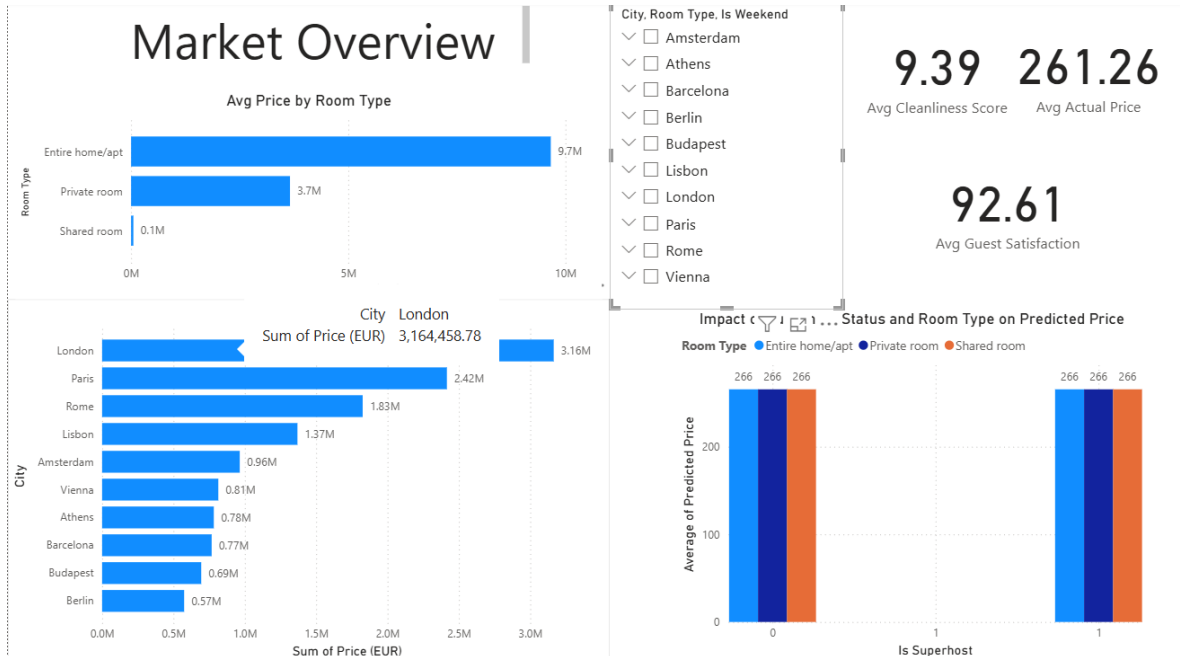
Dashboards provide actionable insights for hosts and the platform:

- Market Overview Dashboard → City-wise pricing distribution, top-performing cities.
- Pricing Analysis Dashboard → Actual vs Predicted Prices, Price Gaps, and Room Type Trends.
- Superhost & Guest Insights → Correlation of ratings, cleanliness, and Superhost status with pricing.

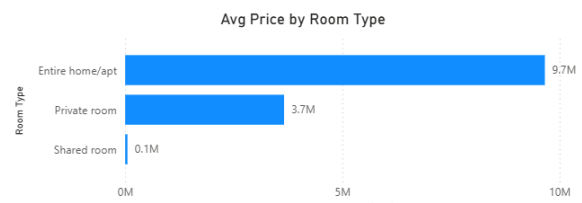








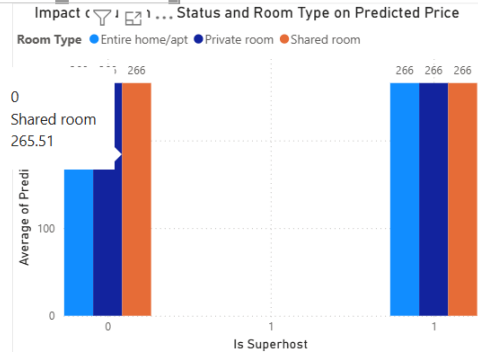
# Market Overview



- City, Room Type, Is Weekend
- ☐ Amsterdam
  - ☐ Athens
  - ☐ Barcelona
  - ☐ Berlin
  - ☐ Budapest
  - ☐ Lisbon
  - ☐ London
  - ☐ Paris
  - ☐ Rome
  - ☐ Vienna

9.39 261.26  
Avg Cleanliness Score Avg Actual Price

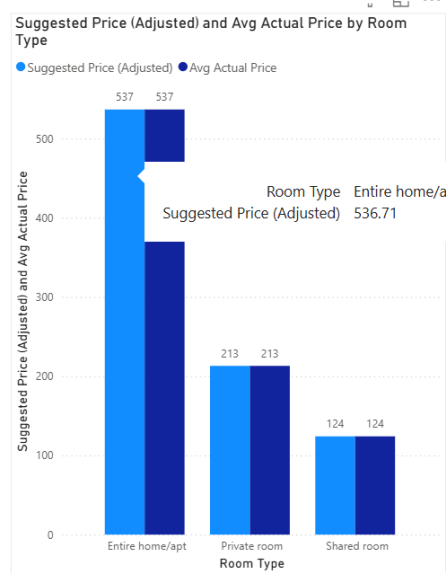
92.61  
Avg Guest Satisfaction



# Pricing Analysis

272.97  
Suggested Price (Adjusted)

- City, Room Type, Is Weekend, Clea...
- ☐ Amsterdam
  - ☐ Athens
  - ☒ Barcelona
  - ☐ Berlin
  - ☐ Budapest
  - ☐ Lisbon
  - ☐ London
  - ☐ Paris
  - ☐ Rome
  - ☐ Vienna



Listing ID	Room Type	City	Sum of Price (EUR)	Predicted Price	Price Gap
929	Private room	Barcelona		-73.05	-73.05
1018	Private room	Barcelona		-73.05	-73.05
1025	Private room	Barcelona		-73.05	-73.05
1042	Private room	Barcelona		-73.05	-73.05
1090	Private room	Barcelona		-73.05	-73.05
1321	Private room	Barcelona		-73.05	-73.05
1322	Private room	Barcelona		-73.05	-73.05
1331	Private room	Barcelona		-73.05	-73.05
1336	Private room	Barcelona		-73.05	-73.05
1385	Private room	Barcelona		-73.05	-73.05
1404	Private room	Barcelona		-73.05	-73.05
1424	Private room	Barcelona		-73.05	-73.05
1441	Shared room	Barcelona		-73.05	-73.05
1456	Private room	Barcelona		-73.05	-73.05
1457	Private room	Barcelona		-73.05	-73.05
1471	Private room	Barcelona		-73.05	-73.05
1472	Private room	Barcelona		-73.05	-73.05
1477	Private room	Barcelona		-73.05	-73.05
1484	Private room	Barcelona		-73.05	-73.05
1487	Private room	Barcelona		-73.05	-73.05
1488	Private room	Barcelona		-73.05	-73.05
1491	Private room	Barcelona		-73.05	-73.05
929	Private room	Barcelona		-21.48	-21.48
1018	Private room	Barcelona		-21.48	-21.48
1025	Private room	Barcelona		-21.48	-21.48
1042	Private room	Barcelona		-21.48	-21.48
1090	Private room	Barcelona		-21.48	-21.48
1321	Private room	Barcelona		-21.48	-21.48
1322	Private room	Barcelona		-21.48	-21.48
1331	Private room	Barcelona		-21.48	-21.48
1336	Private room	Barcelona		-21.48	-21.48
1385	Private room	Barcelona		-21.48	-21.48
Total			2,348.98		171.55