

OVERVIEW

- Cape Town's Airbnb market is competitive and fast-growing.
- Many hosts—especially new ones—struggle to set prices and estimate revenue.
- We built a data-driven tool to help hosts:
 - -Predict expected revenue.
- Suggest improvements to increase earnings.



BUSINESS UNDERSTANDING

The Problem:

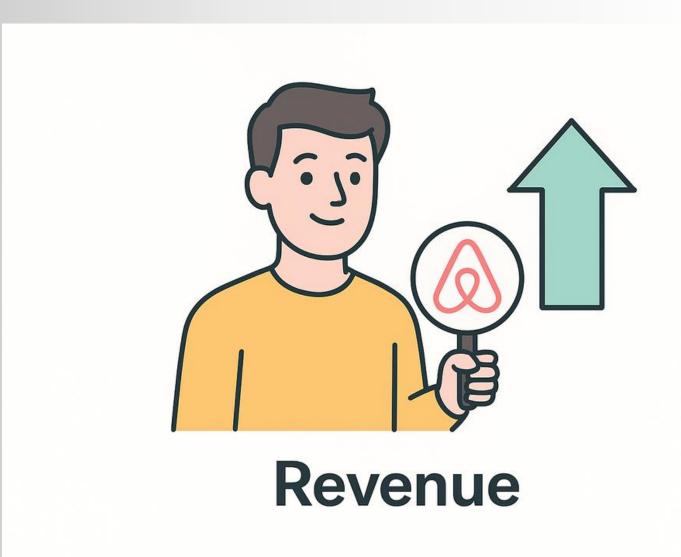
Hosts risk losing bookings or undervaluing their property without good pricing and availability strategies.

Our Solution:

A Revenue Recommender System that learns from past Airbnb data and gives clear, actionable tips.

Goals:

Identify drivers of revenue, understand guest satisfaction factors and Forecast occupancy and pricing trends





Data Understanding

Data Sources

We used Airbnb's public Cape Town datasets:



Listings.csv - Property details (location, room type, amenities)



Calendar.csv - Daily prices and availability

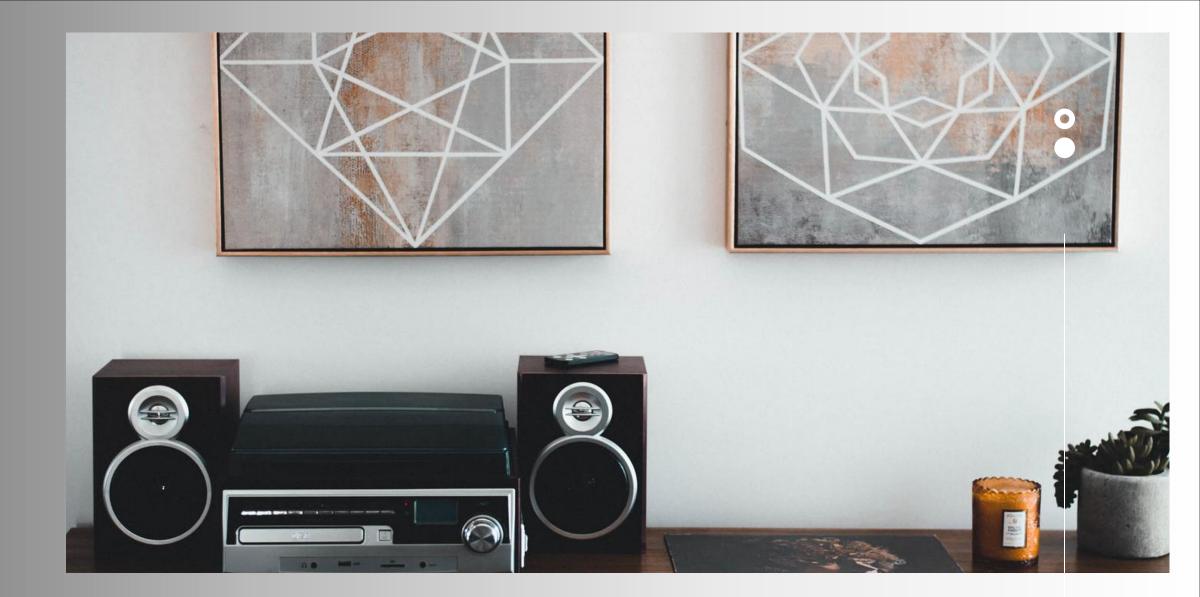


Reviews.csv - Guest ratings and comments



Neighbourhoods.csv – Mapping of areas in Cape Town

Data Preparation



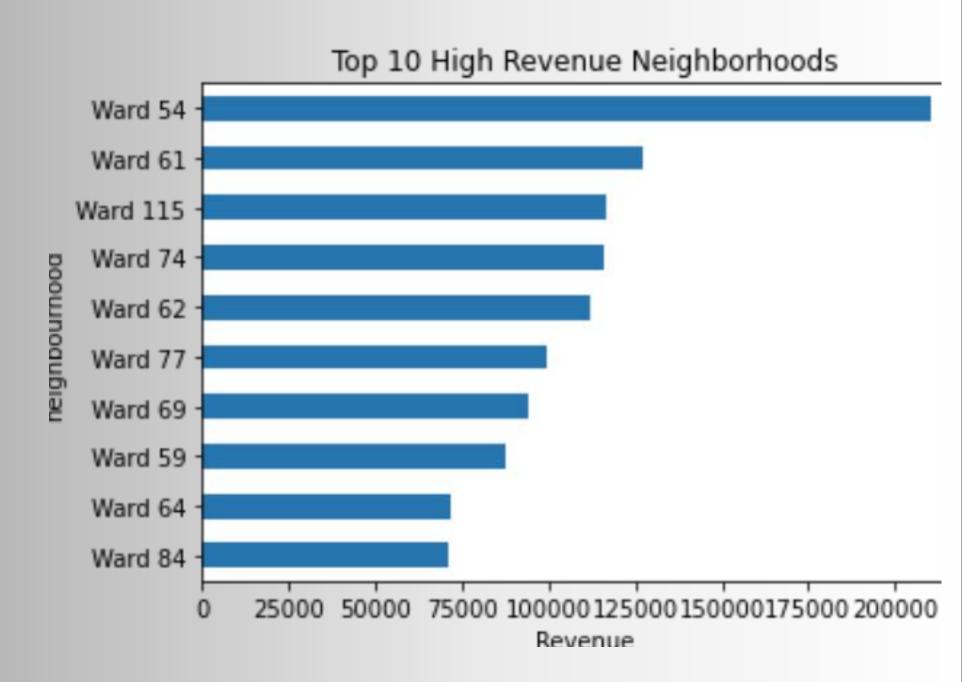
To make sure our analysis was accurate, we:

- Removed outliers (e.g., listings charging \$10,000 per night).
- Filled missing values in reviews and availability.
- Converted categories to numbers so models can understand them.
- Created new features like seasonal availability trends.

Key Data Insights

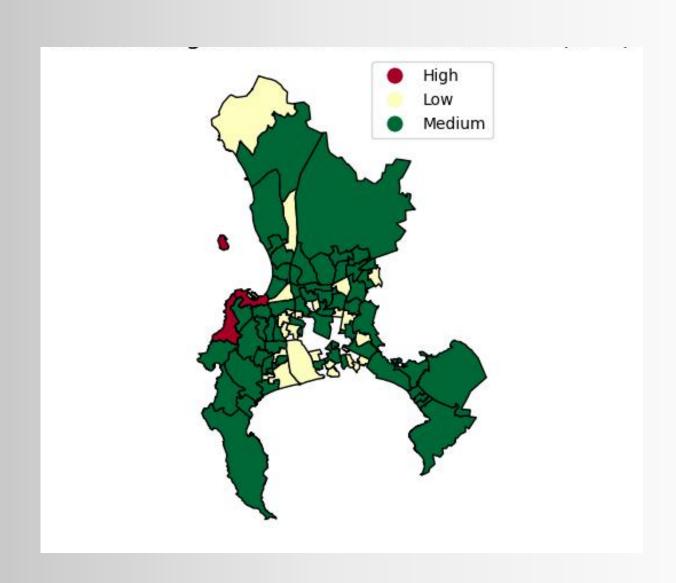
Certain neighbourhoods consistently outperform others in revenue.

- Ward 54 is the top-earning neighbourhood, with revenue far exceeding other areas.
- Other high-revenue areas include Ward 61, Ward 115, Ward 74, and Ward 62.
- Revenue differences between wards suggest location plays a major role in Airbnb success.



GEOGRAPHIC INSIGHTS

- Listings in tourist hotspots like the Waterfront and Camps Bay have higher prices.
 - Central locations attract business travelers, boosting occupancy.
- Distance from attractions influences revenue.



Modelling

We tried several "smart tools" to see which predicts revenue best:

- Simple Trend Finder (Linear Regression) Looks for straight-line relationships between factors like price, location, and revenue.
- Pattern Finder (Random Forest) Handles complex situations where many factors work together.
- High-Accuracy Predictor (Gradient Boosting) Combines many small models to make very accurate predictions.
- Group Finder (Clustering) Groups similar listings together so we can compare performance within each group.



Findings

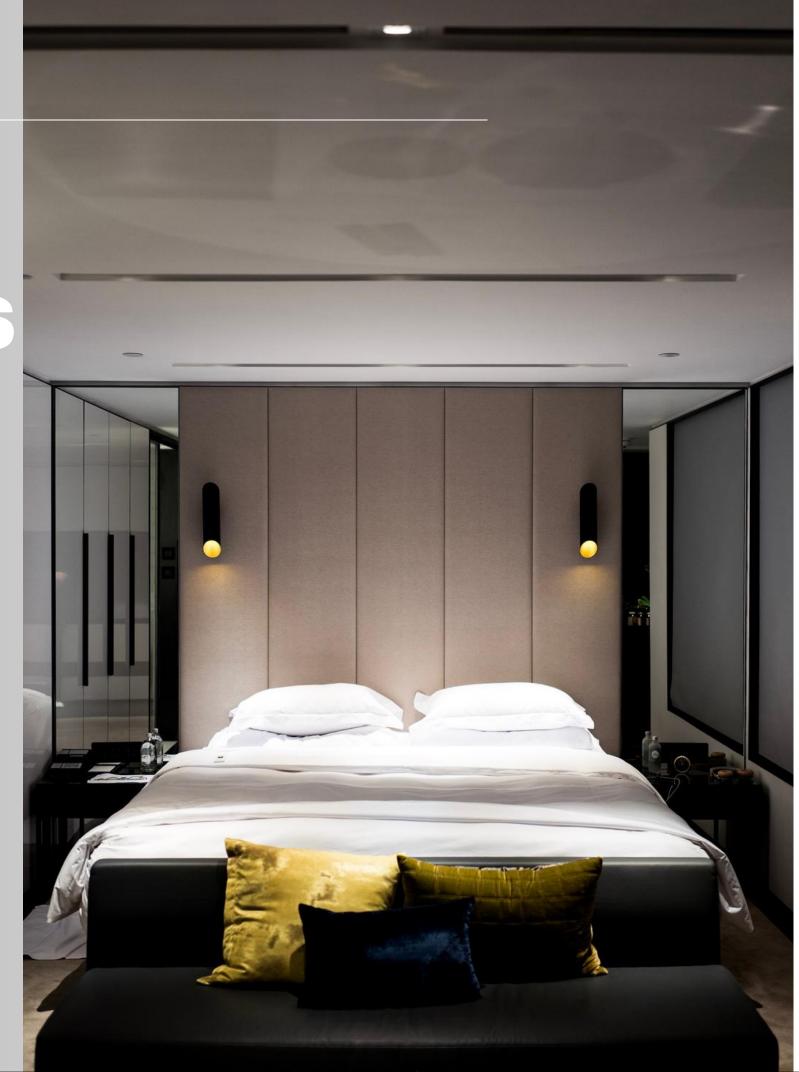
- Price, location, and room type are the biggest drivers of revenue
- Higher review scores boost revenue potential
- Availability patterns (especially during peak season) matter a lot



Recommendations

To boost revenue, hosts should:

- Offer entire homes if possible
- Increase booking availability
- Maintain high review scores
- Optimize pricing using our model's feedback



Next Steps

We plan to:

- ✓ Launch a prototype dashboard using Streamlit

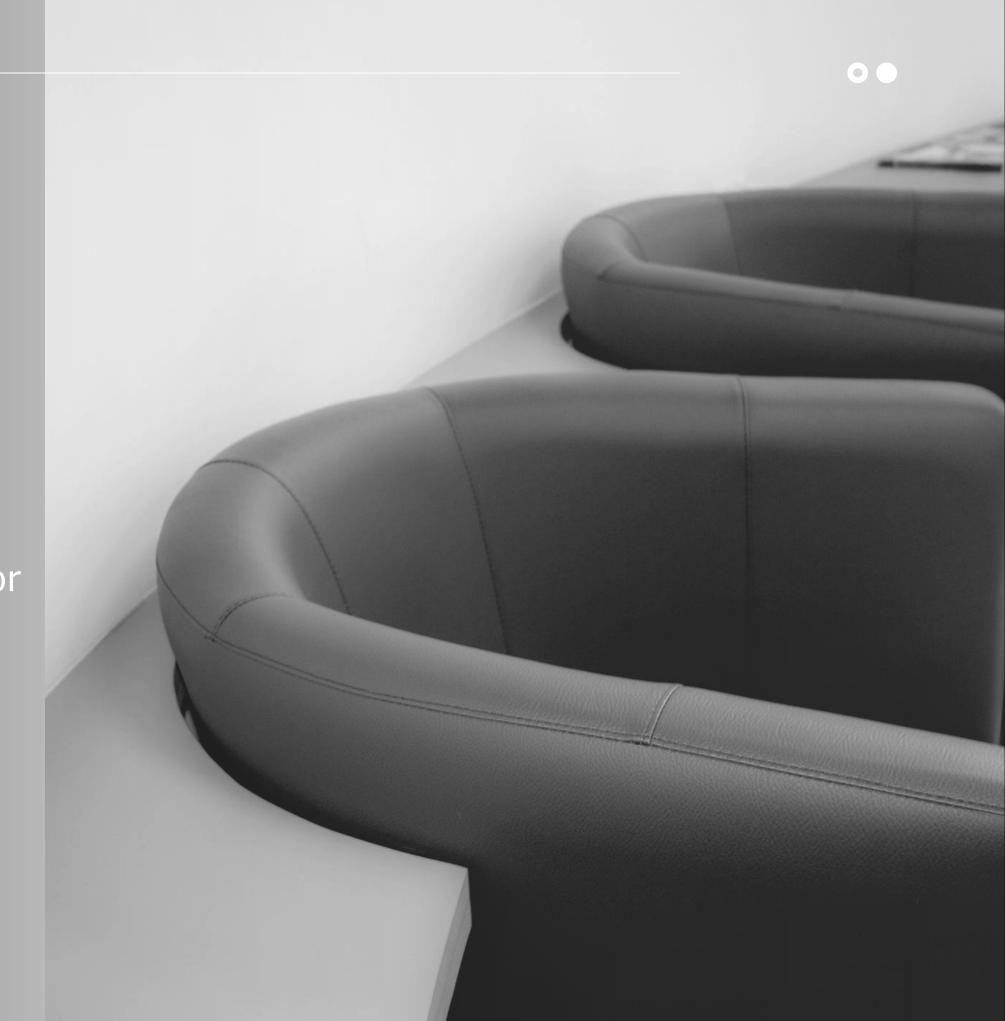
- **♦** Share with local host communities





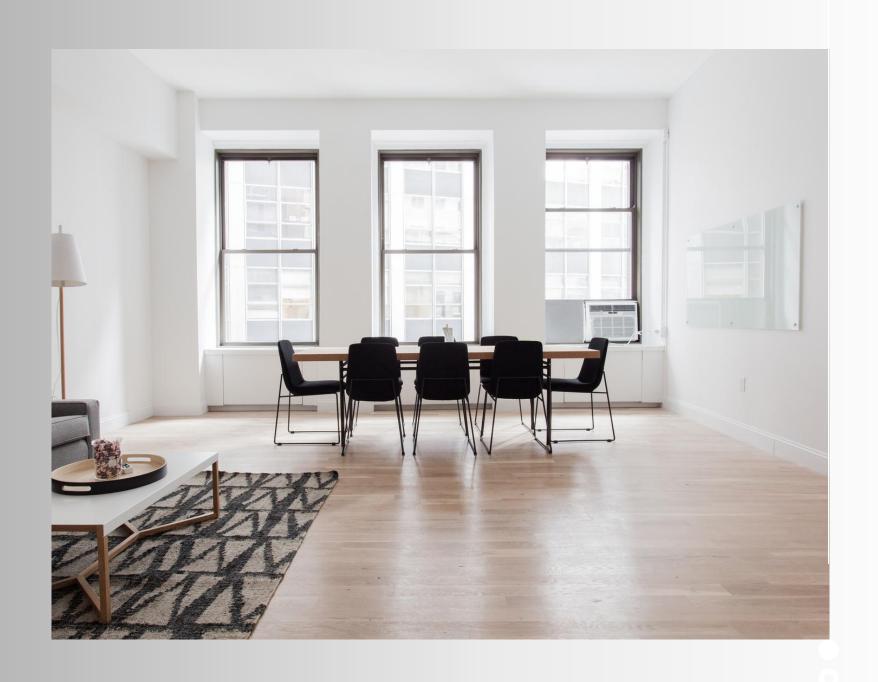
Limitations

- Data may not reflect recent Airbnb trends
- Sparse reviews or availability can reduce accuracy
- External events (e.g., travel bans) not accounted for



Real-World Impact

- Helps hosts make smarter, datainformed decisions
- Increases revenue potential, especially for new or struggling hosts
- Can scale to other cities in the future



THANK YOU

Questions and feedback

We are ready to discuss Further