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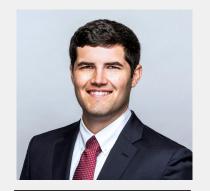


Airbnb Recommendation System

Team 5 Henry Dimlow, Pandora Shou, Zayn Sui, Cindy Zhang

Our Team





Henry Dimlow



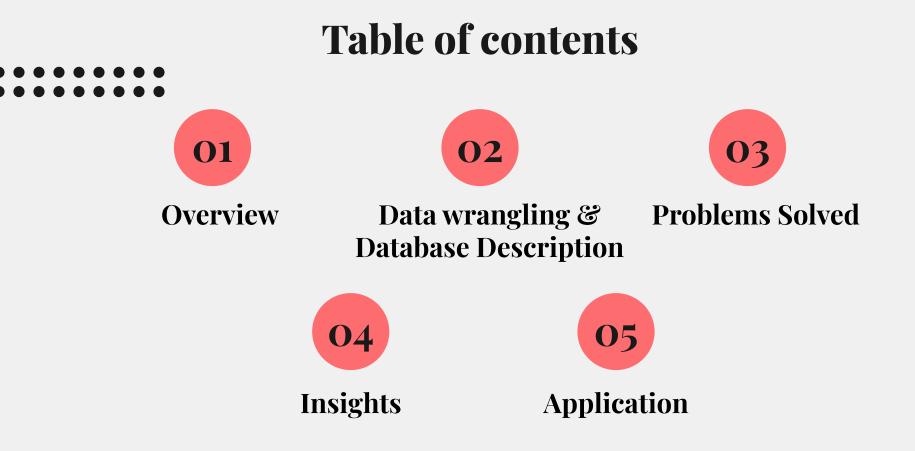
Pandora Shou



Zayn Sui



Cindy Zhang









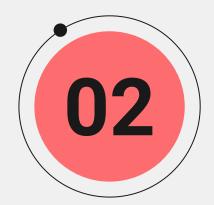


Overview



Our purpose and the corresponding tasks to be done:

- Create the database on a more corporate-friendly level to boost its usability by company employees
 - What kinds of tables should be contain and what kinds of columns in each table?
 - What's the linkage between tables?
- Make usage of the dataset to provide suggestions to 2 most important client group customers and hosts
 - Recommendation system for customers
 - Decision making system for host
- Draw insights from database to learning the current airbnb market
 - o characteristics of listings in different cities
 - host operation situation and profitability





Data wrangling & Database Description



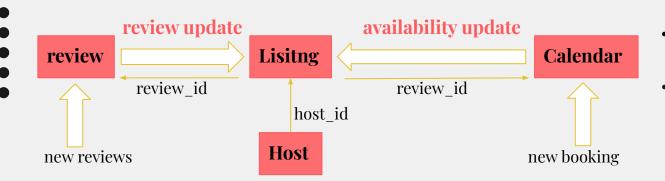
Database design & Data Wrangling

Create Tables

- Merge separate dataset for 8 cities
- Split host and listing from the original data
- Delete some text columns unhelpful (ie. listing descriptions)

Data Cleaning

- remove comma symbols in comments
- fill na value in rating-related columns with o
- fill na value with 'unknown' in text columns
- * We leave all the listing to keep dataset complete



- The Database have 4 tables: Listing, Host, Calendar, Review (2.49GB)
- Review and Calendar tables need to be updated frequently, which are also linked to some summaried columns (eg. availability in 30 days/ reviews in 30 days)

Data Source: <u>Airbnb Inside — USA</u>

Data storage and Processing

approach	RDS	Hive	
Dataset	Los Angeles(smaller)	California	
Loading Path	local file – RDS	Local file – S3 – Hive	
Processing	ER Model, Queries, Funtions	Queries,Link to Tableau	
Queries	recommendation system	recommendation system	
	(with sentiment analysis)	(without sentiment analysis)Decision-making system	

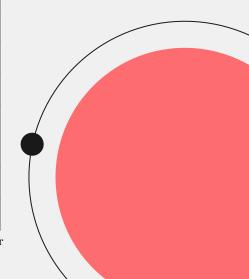
Database Description

Airbnb-California
598 neighbors 74609 listings 37975 hosts 170209 beds in total

Summary Statistics for values which is helpful to understand the overall market:

	min	Q1	median	Q3	max	stdv
	mm	V 1	median	Q)	mux	Stuv
Listings in a Neighborhood	1	13	41.5	143.5	2270	239.74
Hosts in a Neighborhood	1	18.82	3	12	771	52.96
Accommodates	О	2	3	6	16	2.85
Number of Beds	1	1	2	3	32	1.77
Price	О	\$88	\$150	\$263	\$999	\$175.81
Price per Person	О	\$33.75	\$49.5	\$71.4	\$999	\$41.45
Number of Reviews	О	1	9	46	1689	79.76
Rating	О	4.68	4.88	5.0	5.0	0.55

^{*}We don't filter out the listing with o value which is unreasonable because this is a database for queries and those listing with error values are real. So in real business scenarios, those listings shouldn't be deleted but require adjustment.







Problems Solved



Recommendation System for Customers

Customer Requirement:

City & Neighborhood Price Range Accommodates

Date Duration

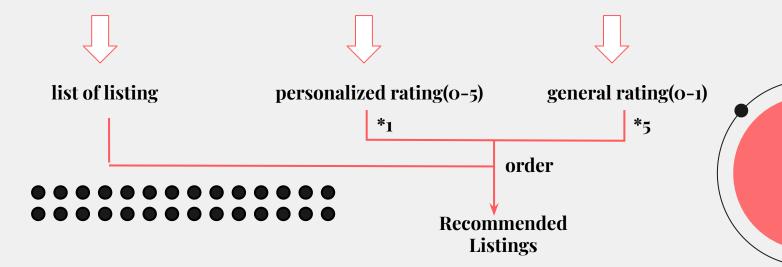
Customer Priority:

Accuray Communication Cleanliness

Location

Sentiment Analysis for comments:

positive words negative words



Recommendation System for Customers

Customer Requirement:

Location: Hollywood

Accommodate: 2

Trip date: 2022-11-30 to 2022-12-03"

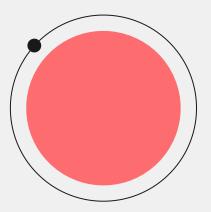
Price range: \$80-220

Customer Priority:

Accuray Cleanliness Communication Location



isting_id	overall_rating
52763527	10
50365526	10
39301021	10
48614491	10
51752110	10
33397503	10
30680366	10
25734648	9.943014569710801
47287433	9.94
27312821	9.935089302500469
39732436	9.919
53303493	9.916
14520333	9.905000000000001
54074518	9.9
51691036	9.888
52909541	9.888
2144103	9.883278084714549
49774867	9.867999999999999
2326099	9.85776433195085
5736708	9.85



Decision-Making System for Hosts

Listing Conditions:

City & Neighborhood Roomtype



Host Goal:

Price Grade(A/B/C/D/E)

Listing Conditions:

los angeles & Hollywood Private room Grade B



Results:

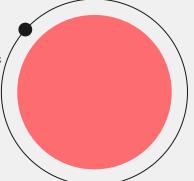
Estimated Occupany rate Estimated Price Estimated Revenue

Results:

Expected Occupancy Rate	69%		
Expected Price	\$87.93		
Expected Revnue	\$22408.94		

*base on listings in dataset in the same area and same roomtype in next 12 months







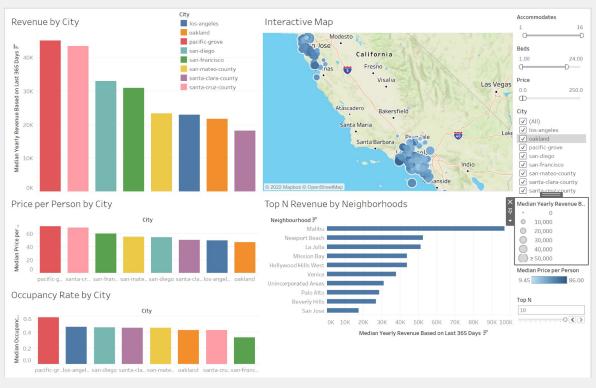


Insights from Dataset



City Insights Dashboard





Insight from City Dashboard



- Price per Person by City
 - LA has the highest median price but is the second lowest median price per person
 - Pacific Grove has the highest price per person despite having the second lowest median price
- Occupancy Rate by City
 - Pacific-grove has by far the largest median occupancy rate
 - San Francisco has by far the lowest occupancy rate despite not standing out in terms of median
 price or Price per Person

Host Insights Dashboard

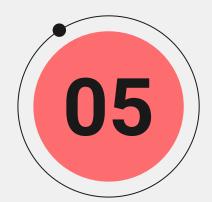




Insight from Host Dashboard



- New Hosts by Time
 - Steady increase in new hosts every year, large peaks of new hosts every summer
 - After 2016 new hosts peak and begin to decline
- Price Distribution by City and Superhost
 - LA has the largest range of IQR and the highest Median price. Oakland has the smallest range and lowest median
 - Pacific-Grove is the only city where superhosts' median price is higher than regular hosts
- Host Superhost Rating Comparison
 - o Superhost's median rating is higher than non-superhosts and range is smaller





Application & Improvement



Application & Improvement



• Application in Airbnb

- Embed Recommendation System in the **Airbnb application interface** to recommend listings
- Embed Decision-making System in the **host interface** to help price their listing and analyze the potential values

• Improvement of models

- **User priority** can be collected when they are new and used each time when they order a listing
- More complicated analysis can be exhibited for host

• Generic Application

 The logic behind our models can be generalized for many platforms for both customers and merchants, like lifestyle platforms and e-commerce websites

Thanks!

Do you have any questions?

