Airbnb Pricing Sankarhan Acharya, Paul de Fusco, Parinaz Azdavari Report 2 – Data Pipelines and Process

1. Raw Data Sources

- Summaries of each dataset description from report 1
- Table for Dataset name, source location, destination in your data pipeline, data movement and processing scripts and notebooks, and data size

Dataset Name	Source	Destination	Acquisition Notebooks, Code, Documents	Data Size	Other Notes, e.g., Confidentiality , Notes from data provider. Etc.
Affordability _Wide_2017 Q3_Public.cs v	www.kaggle.	AWS	Acquired through download	2.6 MB	N/A
MarketHealth Index_Zip.cs v	www.kaggle.	AWS	Acquired through download	2.7 MB	N/A
Zip_MedianR entalPrice_1 Bedroom.csv	www.kaggle.	AWS	Acquired through download	393 KB	N/A
Zip_MedianR entalPricePer Sqft_1Bedroo m.csv	www.kaggle.	AWS	Acquired through download	247 KB	N/A
Zip_PriceTo RentRatio_Al lHomes.csv	www.kaggle.	AWS	Acquired through download	8 MB	N/A

Zip_Zhvi_All Homes.csv	www.kaggle.	AWS	Acquired through download	25.9 MB	N/A
ZriForecast_ Public.csv	www.kaggle.	AWS	Acquired through download	4 KB	N/A
population_b y_zip_2000.c sv	www.kaggle.	AWS	Acquired through download	58.2 MB	N/A
population_b y_zip_2010.c sv	www.kaggle.	AWS	Acquired through download	59.6 MB	N/A
price.csv	www.kaggle.	AWS	Acquired through download	5.1 MB	N/A
pricepersqft.c sv	www.kaggle.	AWS	Acquired through download	5.6 MB	N/A
kaggle_inco me.csv	www.kaggle.	AWS	Acquired through download	5 MB	N/A
Air_Quality_ Measures_on _the_Nationa l_Environme ntal_Health_ Tracking_Net work.csv	www.kaggle.	AWS	Acquired through download	49.6 MB	N/A
Crime_Data_ LA_from_20 10_to_Presen t.csv	www.kaggle.	AWS	Acquired through download	385.3 MB	N/A

2016_Genera lElection_Re sults_by_prec inctcomple te_eCanvass_ datasetcsv	www.kaggle.	AWS	Acquired through download	47.9 MB	N/A
calendar.csv	www.insideai rbnb.com	AWS	Acquired through download	62.9 MB	
listings.csv	www.insideai rbnb.com	AWS	Acquired through download	29 MB	
reviews.csv	www.insideai rbnb.com	AWS	Acquired through download	38.9MB	
data_hotel .zip	www.kaggle.			434.6 MB	
Weather data TBD	WIP				
Consumer Datasets	WIP				
Addl Datasets TBD					

2. Data Exploration, Cleaning, Wrangling and Engineering

• Data Exploration Summary

Initial discussions on modeling the prediction have been focused on producing a regression model to forecast the ideal price for a home. As a consequence, the initial efforts in data

exploration have been centered around identifying variables that have an effect on the price variable.

Data Preprocessing Approach

So far, we have focused on the datasets provided by <u>insideairbnb.com</u>. These datasets consist in the calendar.csv, the listings.csv and the reviews.csv. We joined the calendar and the listings to produce a larger file named 'Master' containing listing information for each calendar day a particular listing was offered. The ultimate goal is to predict the price for each day and produce market aggregate measures.

• Approach for storing processed and/or integrated data

So far the data has been stored on ipython notebooks. In the meantime, Parinaz has been working on setting up an AWS instance which will eventually host all or most of our data.

 Processed dataset description for each processed dataset including why you want to process it that way

Each dataset is being processed to augment the Master dataset. So far the master dataset has the following columns:

id int64 listing url object scrape id int64 last scraped object name object object summary object space object description experiences offered object neighborhood overview object object notes object transit object access interaction object house rules object thumbnail url object medium url object picture url object xl picture url object

host_id	int64
host_url	object
host_name	object
host_since	object
host_location	object
host_about	object
host_response_time	object
host_response_rate	object
host_acceptance_rate	object
host_is_superhost	object
host_thumbnail_url	object
maximum_nights	int64
calendar_updated	object
has_availability	float64
availability_30	int64
availability_60	int64
availability_90	int64
availability_365	int64
calendar_last_scraped	object
number_of_reviews	int64
first_review	object
last_review	object
review_scores_rating	float64
review_scores_accuracy	float64
review_scores_cleanline	ess float64
review_scores_checkin	float64
review_scores_commun	ication float64
review_scores_location	float64
review_scores_value	float64
requires_license	object
license	float64
jurisdiction_names	object
instant_bookable	object
cancellation_policy	object
require_guest_profile_pi	
require_guest_phone_ve	erification object
calculated_host_listings	_count int64
reviews_per_month	float64
listing_id	int64
date	object
price_y	float64

In summary, this dataset produces price curves for each listing as a function of many variables.

• Table for processed data sets including processed data set name, input datasets, link to the processing scripts and notebooks, and provisional data size

Dataset Name	Input Datasets (Dependencies)	Destination	Related Notebooks, Code, Documents	Provisional Data Size	Other Notes
Master	Listings, Reviews, Calendar. Eventually more with different structure.	AWS	listings_explo .ipynb	125 MB	

3. Approach for Feature Engineering and Data Modeling

- Summary of feature sets
- Table for feature set including links to input datasets, feature engineering scripts and notebooks, and provisional data size

Feature	Input Datasets (Dependencies)	Destination	Related Notebooks, Code, Documents	Other Notes
The features are listed section 2	listings, reviews, calendar	listings_explo .ipynb	listings_explo .ipynb	

4. Approach for Data Access

- Initial design for data querying interfaces
- Justification for manual vs. programmatic access

The data querying interfaces will ultimately provide access to the central logical schema. At this point there hasn't been a need for designing them as the datasets used are all imported manually into the ipython notebook.

Ideally we would like to opt for programmatic access

5. Data Pipeline

- Description of the needs, approach, and data access and refresh frequency
- Logical diagram showing major data pipeline components for data sources and sinks

Predictive Models

Central Storage (AWS)

InsideAirbn b Datasets

Hotel Data

Zillow Rental & Home Data Other datasets (weather, consumer, etc)

Refresh Frequency:

Inside Airbnb: Depends on scraping abilities. Ideally once a day, maybe once a week

Hotel Data: No refresh planned at this time Zillow Data: No refresh planned at this time

Other datasets: depends on the nature of the dataset. Weather could be daily, consumer more rare

6. Set up for your data environment

• Cloud vs. local, database vs. flat files, etc. (see lecture slides for further suggestions)

7. Bullets for each team member's individual contributions in Step 2

Paul: produced code to continue data exploration and build predictive model

Sankarshan: produced code to continue data exploration and build predictive model

Parinaz: investigated AWS account opening

8. Any major updates to Step 1 as a result of data pipeline step