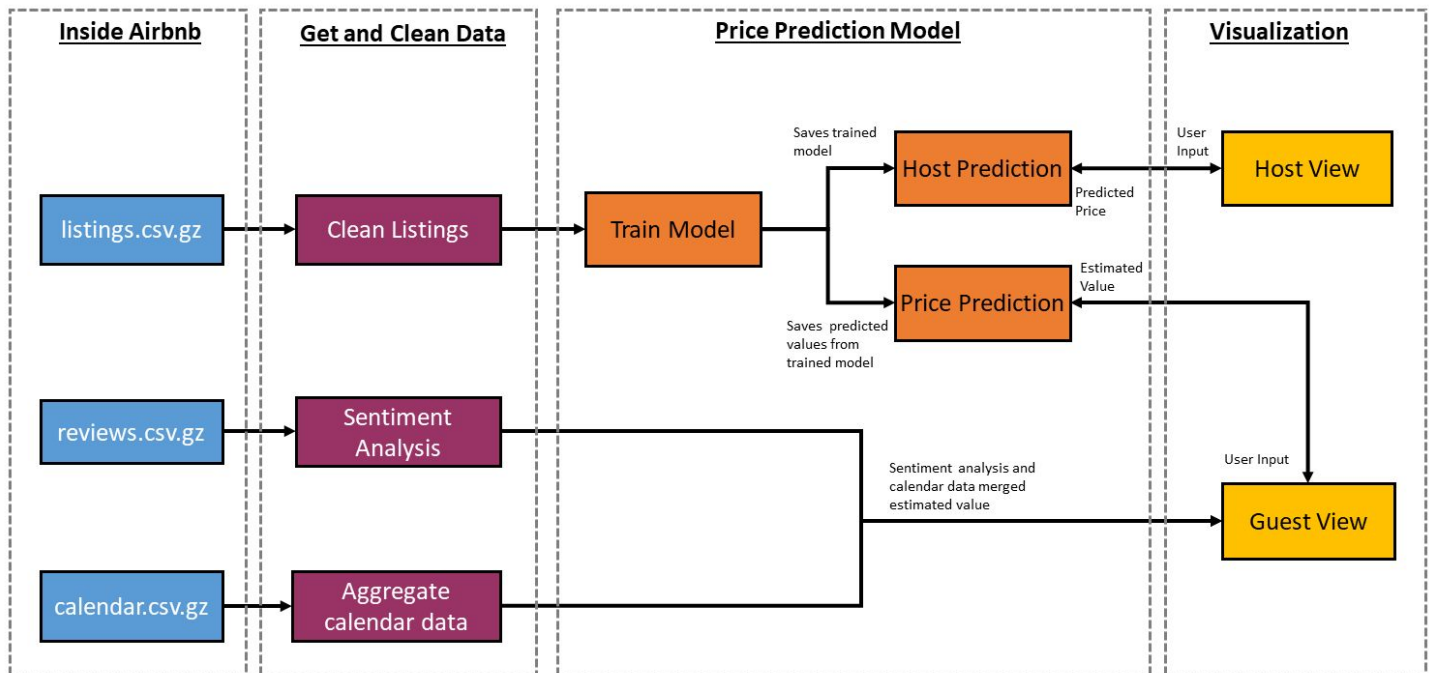


Component Specification & Project Plan

Component Interactions



Components

Visualization:

Description: The visualization is the interface the user uses when deciding if they are using the Guest Visualization or the Host Visualization. Uses a bash command to run and open

zillowbnb.py

Description: User interface for the Guest to view the price listings and Hosts to estimate the listing price for their listing.

Inputs: cleaned listings file

Outputs: data visualization and user interface

Data Gathering and Cleaning:

Description: A group of modules used to read in data from [Inside Airbnb](#) and clean and aggregate the data into forms that the machine learning model can use.

convert_to_matrix.py

Description: Converts dataframe into numerical array to be used in our model

Inputs: Cleaned dataframe of airbnb listings

Outputs: Numerical array of features and array of prices

example.py

Description: Script to run to read in data, identify specific columns to use in the predicting price model, run cleaning procedures, and sentiment analysis.
Inputs: city, state abbreviation, country, date of data compilation (yyyy_mm_dd), and filename
Outputs: a merged dataset of the sentiment analysis, cleaned listings data, and price data

get_calendar_summary.py

Description: creates aggregate prices for each listing by weekday price, weekend price, and seasonal price (fall, spring, summer, winter)
Inputs: a dataframe of calendar information
Outputs: a dataframe and .csv with average price by weekday, weekend, summer, winter, spring and fall

get_cleaned_listings.py

Description: cleans listing data pulled from [Inside Airbnb](#), and splits the combined amenities column into individual amenities columns
Inputs: listings data
Outputs: a cleaned dataframe and a .csv file

get_data.py

Description: used to read in datasets off of [Inside Airbnb](#)
Inputs: city, state abbreviation, country, date of data compilation (yyyy_mm_dd), and filename.
Outputs: a dataframe

sentiment.py

Description: performs a sentiment analysis on the reviews of the Airbnb listings
Inputs: a dataframe of the listings reviews
Outputs: a dataframe and .csv of the mean, variance and count of the reviews by listing

Price Prediction Model:

Description: Machine Learning model used to predict listing prices of Airbnb listings

dataset_prediction.py

Description: reads the dataset and predicts prices with the different models for the guest view
Inputs: Data of the listings in array form and name of the city
Outputs: Array of predicted prices

detect_outliers.py

Description: Detects outliers to choose which model to run
Inputs: Data of prices as an array
Outputs: List of outliers

host_predict.py

Description: predicts price using the saved model from train_model.py
Inputs: array of trained model, city name
Outputs: predicted price

train_model.py

Description: trains a boosted tree regressor
Inputs: array of feature, array of prices and the name of the city
Outputs: the model saved as a .dat file in the data folder

price_prediction.py

Description: Creates price predictions off of the listing data provided for the host view
Inputs: Data of the listings in array form and name of the city
Outputs: Array of predicted prices

Project Plan

Week 1:

- Visualization tool Tech Review -
 - Create preliminary visualizations for each technology
 - Dash, Tableau, Bokeh
- Clean Airbnb datasources
 - listings
 - calendar
 - reviews
- Import data sources into format for model

Week 2:

- PCA (Feature Selection) for Listings Data
- Simplify Calendar Data
- Perform Sentiment Analysis
- More Bokeh - commit initial work
- Begin creating unit tests
 - Input file matches the correct criteria
 - Valid address input from host
 - Typing filter on visualization returns error when invalid value
- Create model and define features users can filter by
 - Validation
 - Start visualizations and user filters

Week 3:

- Update Bokeh visualizations using features selected by PCA
- Convert user input into array for model
- Finalize model and store model coefficients
- Run evaluation metric on listing dataset: Good, bad, meh metric (Too Low (-50%), Low (-25%), Average(+-10%), High (+25%) , Too High (+ 50%)) Probably not these values, but something like this. Don't really know until we get model running
- Begin drafting Final Presentation
 - Future work: increase scope (other cities)

Week 4:

- Final Presentation dry run
- Complete user test
- Clean up git repo and finalize Functional Specifications and Component Specifications

Week 5:

- Final presentation