# Data Visualization and Dashboards with Tableau

GOAL: THE PURPOSE OF THIS PROJECT WAS TO PREPARE IMPACTFUL VISUALIZATION USING TABLEAU FOR A GIVEN DATASET AND MAKE A MEANINGFUL DASHBOARD TO COMMUNICATE BUSINESS INSIGHTS ABOUT AIRBNB DATASET.

# Processes:

**Data Connection**: Collect/download the dataset to be analyzed and connect in Tableau using Datasource

**Detect data types** Review data and check if all datatypes are appropriate according to column names.

- Host Id :Number

-Host Since :Date

-Name :String

-Neighbourhood :String

-Property Type:String

-Review Scores Rating (bin): Number

-Room Type :String

-Zipcode :Number (Geographic Role)

-Beds :Number

-Number of Records: Number

-Number Of Reviews : Number

-Price :Number

-Review Scores Rating: Number

Check for null values, duplicate data, invalid or incorrect data. Such data can be easily filtered out in Tableau.

# **Categorical Features:**

- •Room Type
- •Beds
- Price
- Neighborhood
- •Host Since

## **Build Visualization for EDA:**

Following slides consist all the visualization.

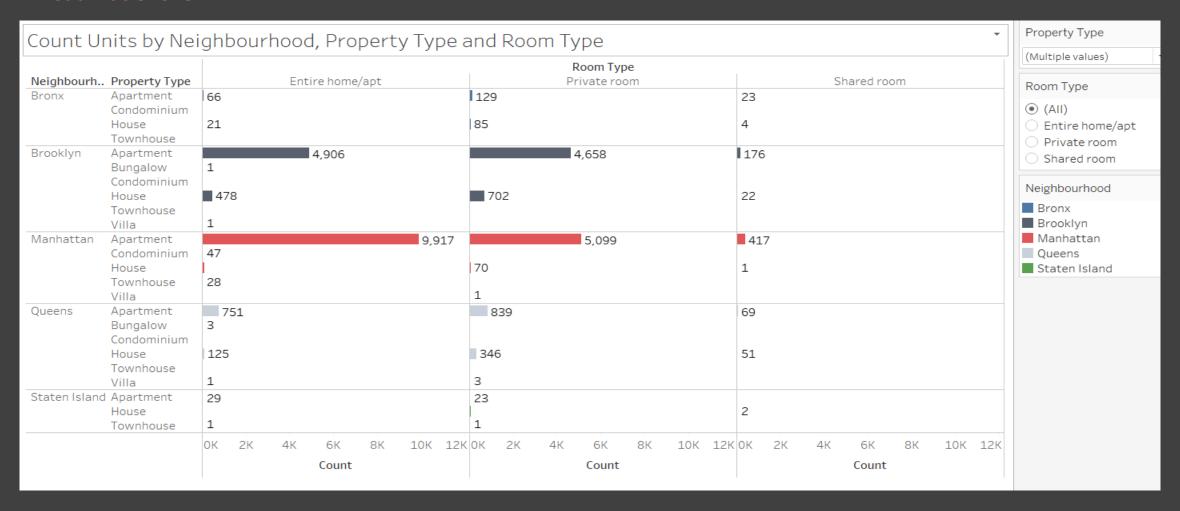
Find an interesting pattern, trend, outlier, etc. from the data used in the above steps.

## **Create dashboard**

# **Data Interpretation:**

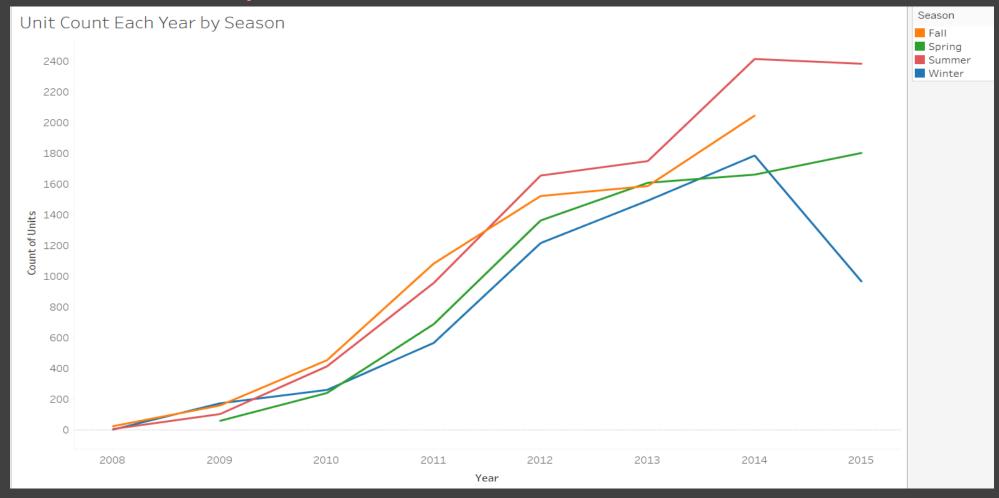
Detect Meaningful keypoints and answer relevant questions observed from visualization.

## **Visualizations for EDA**



Above graph is to count units according to Neighborhood, Property type and Room type. Here I have used two filters to search for specific Property type and room type. E.g. in Brooklyn there are 478 Houses as 'Entire home', 702 Houses has 'Private Room' and in 22 Houses 'Shared Room' has been put as Airbnb unit.

# **Unit Count Each Year by Season**



This graph shows the trend as how many units were added as Airbnb units during each season over the years. Mostly people add their property as Airbnb during Summer when holiday season is on the peak. Here I have used 'Season' Calculated field on Year column.

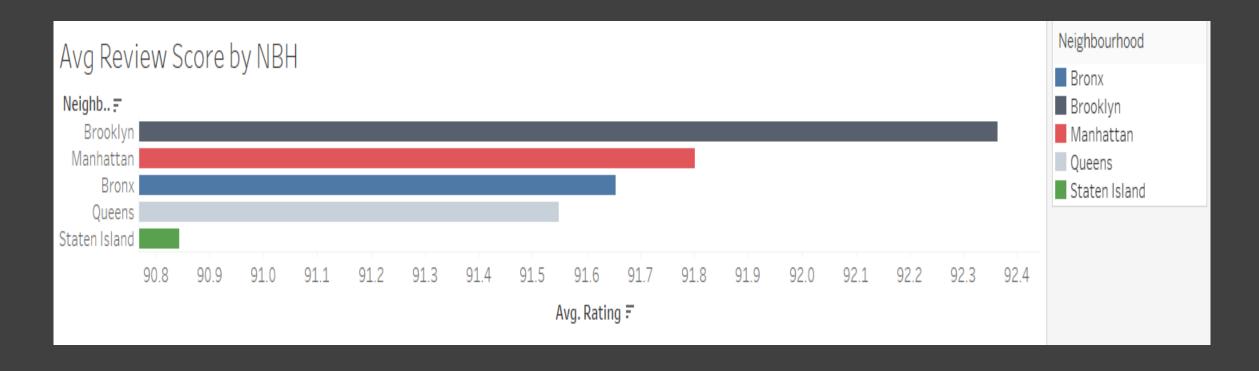
# **Count of Units by Price Category**

Count of Units by Price Category							
Neighbourhood							
							Ct-t-
Property Type Apartment	Category Affordable	Room Type Entire home/apt	Bronx 31	847	Manh 500	272	State
Apartment	Allordable	Private room	118	4.082	3.063	767	22
		Shared room	22	168	3,003	65	22
	Costly		35			464	16
	Costry	Entire home/apt Private room		3,850	7,965		10
			10	569	1,992	72	Т
	_	Shared room	1	7	79	4	
	Expensive	Entire home/apt		160	1,041	12	
		Private room		5	29		
		Shared room			4		
	Luxury	Entire home/apt		49	411	3	
		Private room	1	2	15		
		Shared room		1	1		
Bed & Breakfast	t Affordable	Entire home/apt		2		1	
		Private room	4	54	23	28	
		Shared room		5	4		1
	Costly	Entire home/apt		4	3	1	1
		Private room	2	9	29	6	
		Shared room		1			
	Luxury	Entire home/apt			1		
	,	Shared room		1			
Boat	Affordable	Entire home/apt		1			
2000	Costly	Entire home/apt			1	5	
	cosciy	Private room			1		
Bungalow	Affordable	Entire home/apt				1	
bungalow	Costly	Entire home/apt		1		2	
Cabin	Affordable	Private room				1	
Cabin							
C/D\/	Costly	Entire home/apt		4		2	
Camper/RV	Affordable	Entire home/apt		1		3	
		Private room				1	
	Costly	Entire home/apt				1	
	Expensive	Entire home/apt			1		
Cactle	Costly	Drivate room			1		

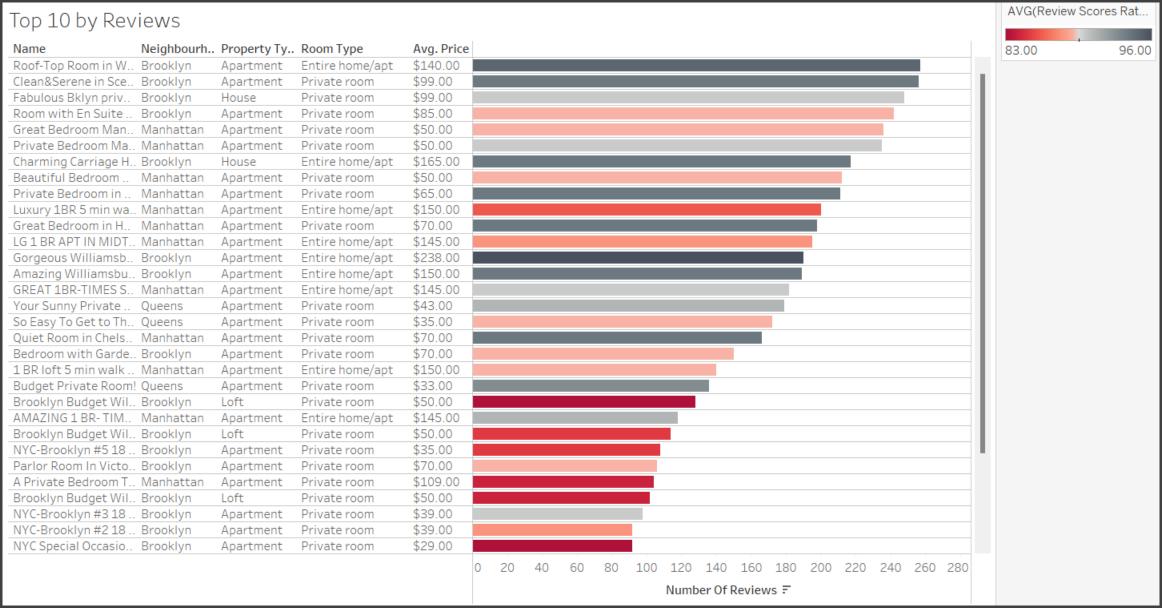
This graph shows count of units according to price. E.g. in Brooklyn the count of affordable apartment (entire house) is 31. Here I have created 'Category' as calculated field based on price column.

# **Average Review Score rating by Neighborhood**

As per average Review Score Rating, Brooklyn has the highest rating than other neighborhoods



# **Top 10 Airbnb by Reviews**



# **Maximum and Minimum Price based on Room Type in each Neighborhood**



# **Clustering:**

**Inputs for Clustering** 

Variables: Count of Beds, Avg. Review Scores Rating, Count of Number of Records, Count of Review

Scores Rating (bin)

**Level of Detail:** Neighbourhood, Property Type, Room Type

**Scaling:** Normalized

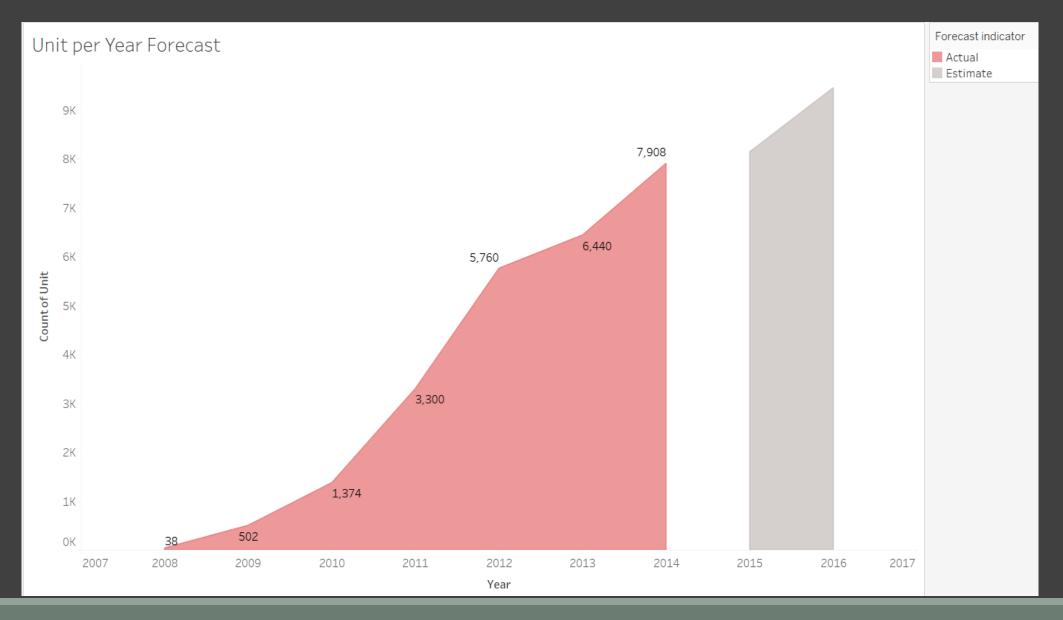
		Centers							
Clusters	Number of Items	Count of Beds Av	vg. Review Scores Rating	Count of Number of Records	Count of Review Scores Rating (bin)				
Cluster 1	52	100.81	90.802	101.08	71.558				
Cluster 2	23	23.087	96.59	23.087	17.478				
Cluster 3	14	5.7857	82.833	5.8571	4.1429				
Cluster 4	3	4870.3	92.006	4887.7	3516.3				
Cluster 5	1	9899.0	92.167	9917.0	7422.0				
Cluster 6	3	1.3333	70.667	1.3333	1.0				

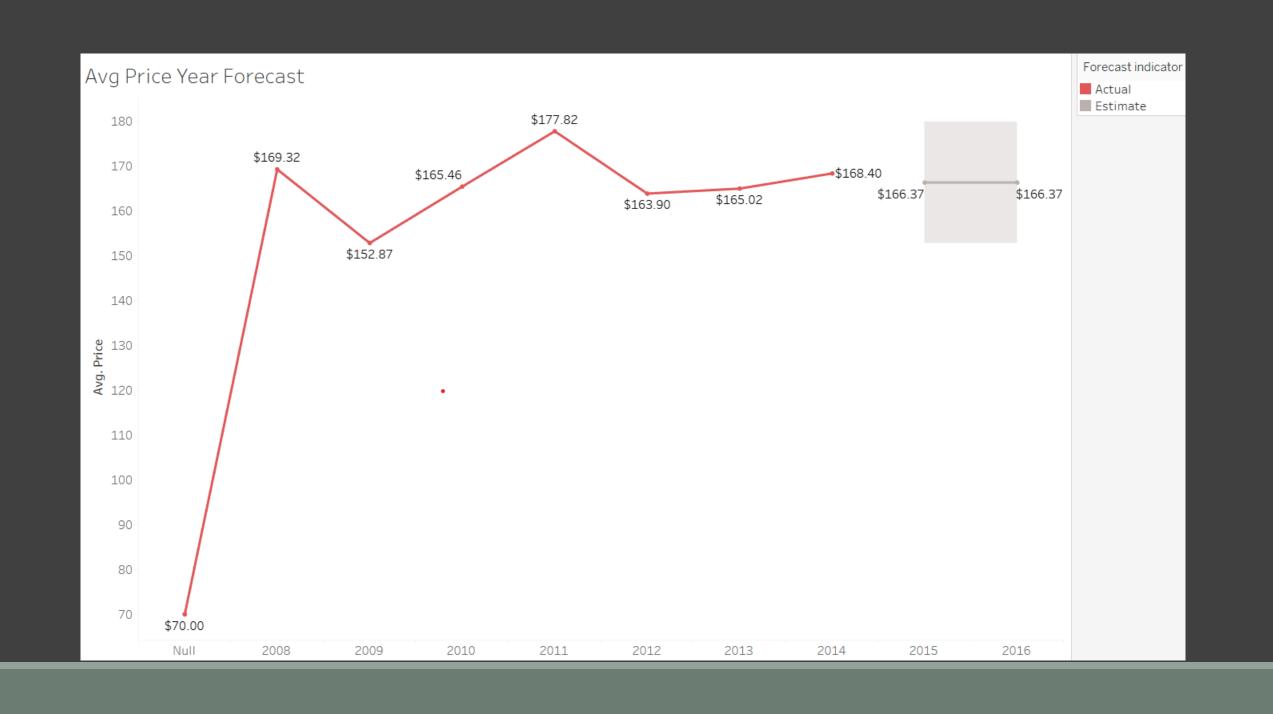
Madal

# **Analysis of Variance:**

			Model		Sum of Squares DF		
Variable	F-statistic	p-value	Sum of Square	s DF			
Count of Beds	17.77	3.251e-12	1.634	5	1.655	90	
Count of Number of Records	17.77	3.251e-12	1.636	5	1.658	90	
Count of Review Scores Rating (bin)	17.73	3.393e-12	1.586	5	1.61	90	
Avg. Review Scores Rating	15.91	3.305e-11	2.636	5	2.982	90	

# Forecast:





8K

6K

4K-

2K

Count of Unit

# Avg Price by Zipcode in each Neighbourhood



# Avg Price Year Forecast

Unit per Year Forecast

38 502



5,760

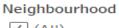
3,300

2008 2009 2010 2011 2012 2013 2014 2015 2016

Year

7,908

6,440





- ✓ Bronx
- ✓ Brooklyn✓ Manhattan
- ✓ Queens
- ✓ Staten Island

#### Season

- Fall
  Spring
- Summer
- Winter

#### Date Parameter

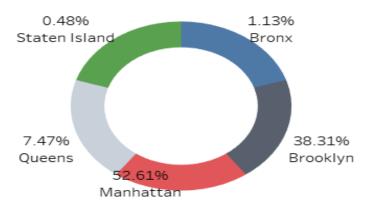


- Bronx
- Brooklyn
- Manhattan
- Queens
- Staten Island

#### Forecast indicator

- Actual
- Estimate

# Percentage of Units by Neighbourhood

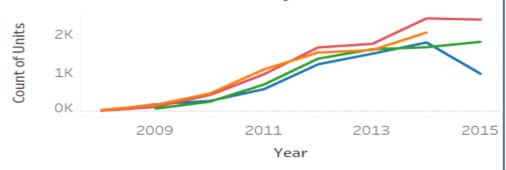


# Unit Count by Year/Quarter/Month

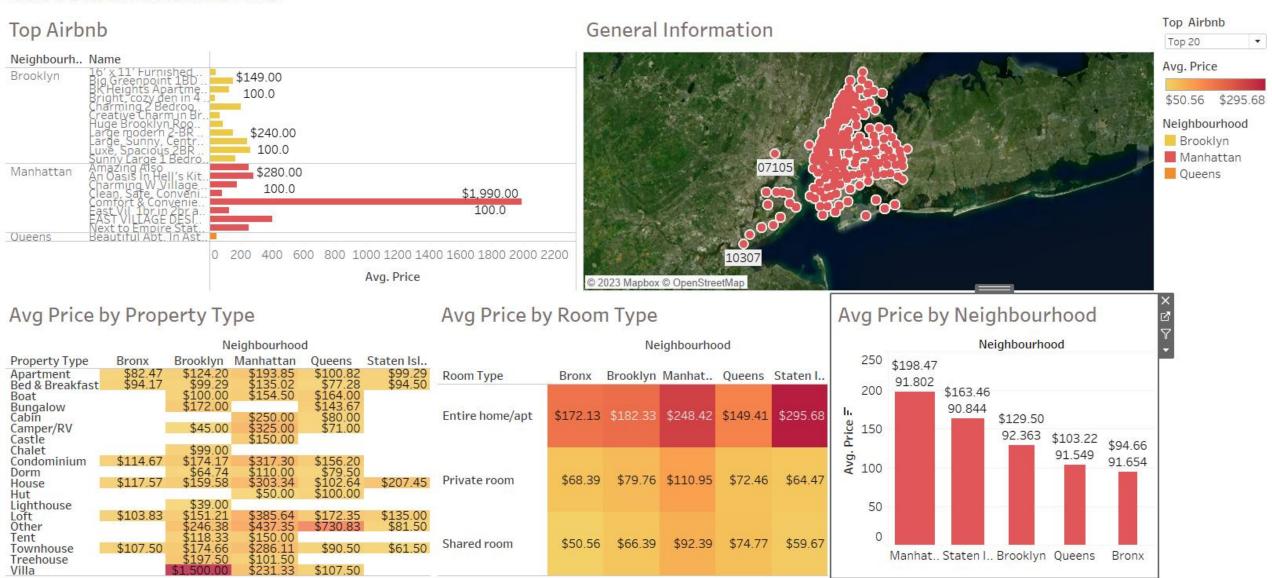
Neighbo			2008	2009	2010	2011	2012	2013	2014	2015
Bronx	S	5K-		2	32	27	29	59	106	90
Brooklyn	Ö	5K-	11	229	619	1,518	2,401	2,340	2,798	1,757
Manhatt	S	5K-	22	245	674	1,581	2,909	3,595	4,257	2,749
Queens	Ö	5K	5	24	40	162	388	418	713	528
Staten Is	Ö	5K-		2	9	12	33	28	34	29
			2008	2009	2010	2011	2012	2013	2014	2015

# Unit Count Each Year by Season

1,374



# **CUSTOMER DASHBOARD**



# **Results:**

I have chosen the Airbnb dataset as I was familiar with the information provided in this and could conclude below observations:

- Manhattan is the most expensive neighbourhood and Bronx is cheapest
- Manhattan has the major share of airbnbs of total Airbnb units available in New York.
- -Airbnb gained significant popularity over the years starting from 2008 to 2014.
- Most units were added during Summer season and least during winter season
- Property type 'Apartment' has the maximum count in all units added as Airbnb.

# **Challenges:**

- Tableau automatically does certain aggregations which are not required
- It is a bit difficult to connect each viz and create meaningful interpretation
- Data needs to be formatted for each column
- Tooltip needs to be updated carefully so that it shows correct data with understandable names

### **Future Goals:**

If given more time I will explore more ways to improve visualization and add more insights to the current analysis. I would like to learn more about animation which can be added in this scenario.

Thank You!!