

Airbnb Project

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Agenda

Hypothesis

Data

Inspiration

Analysis

Projection

Limitations & Future Analysis

Q& A



Find places to stay on Airbnb

Discover entire homes and private rooms perfect for any trip.

LOCATION

Nearby

CHECK IN

Add Date

CHECK OUT

Add Date

ADULTS

2



CHILDREN

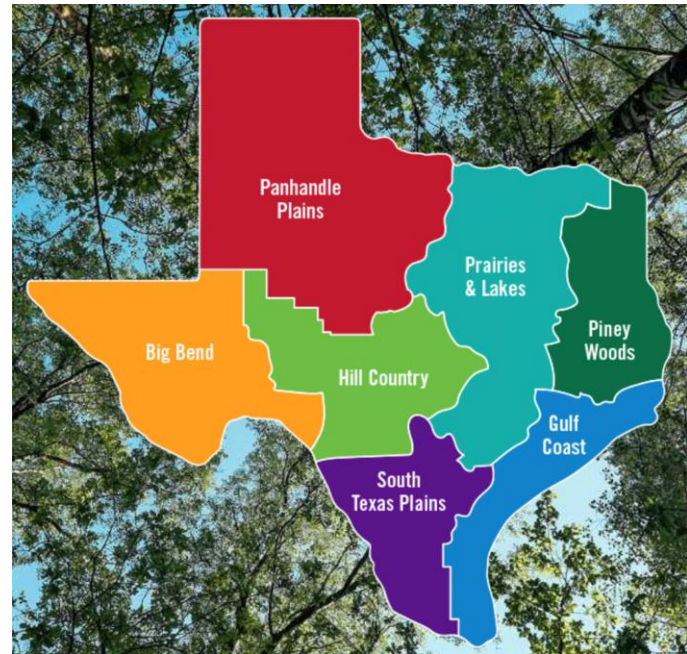
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Q Search

Hypotheses

- There is less competition in rural areas due to higher rates.
- There are more rentals in areas with tourism attractions.
- Population and income affect Airbnb rates.



Datasets

kaggle

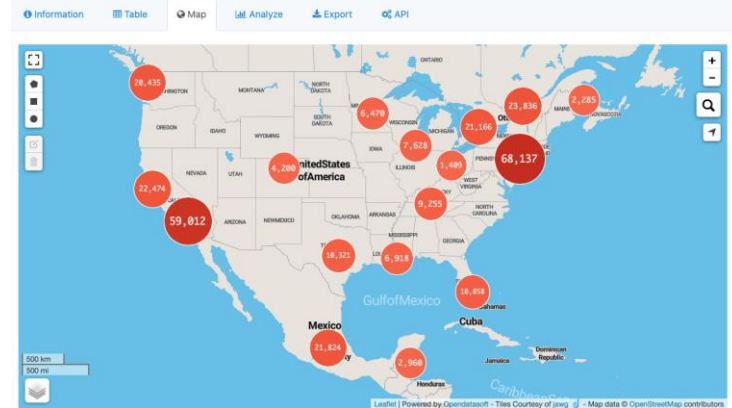


Inspiration

opendata**soft**

KDnuggets™

Airbnb listings



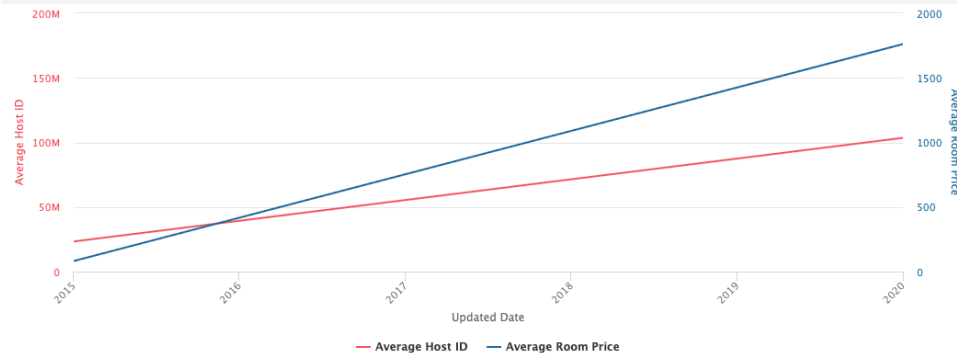
Airbnb listings

Information Table Map Analyze Export API

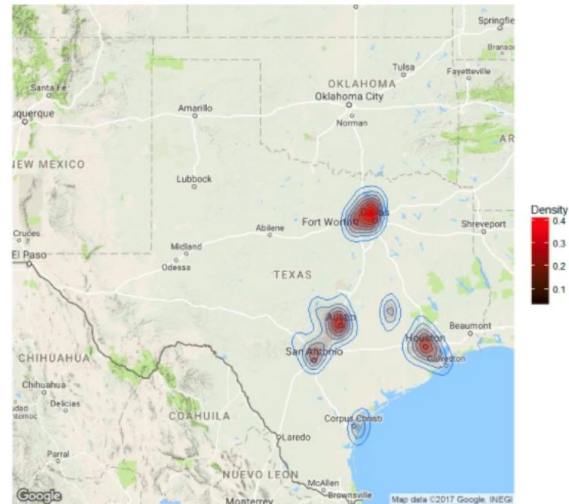
X axis: Updated Date Year Max. number of points: Default ordering

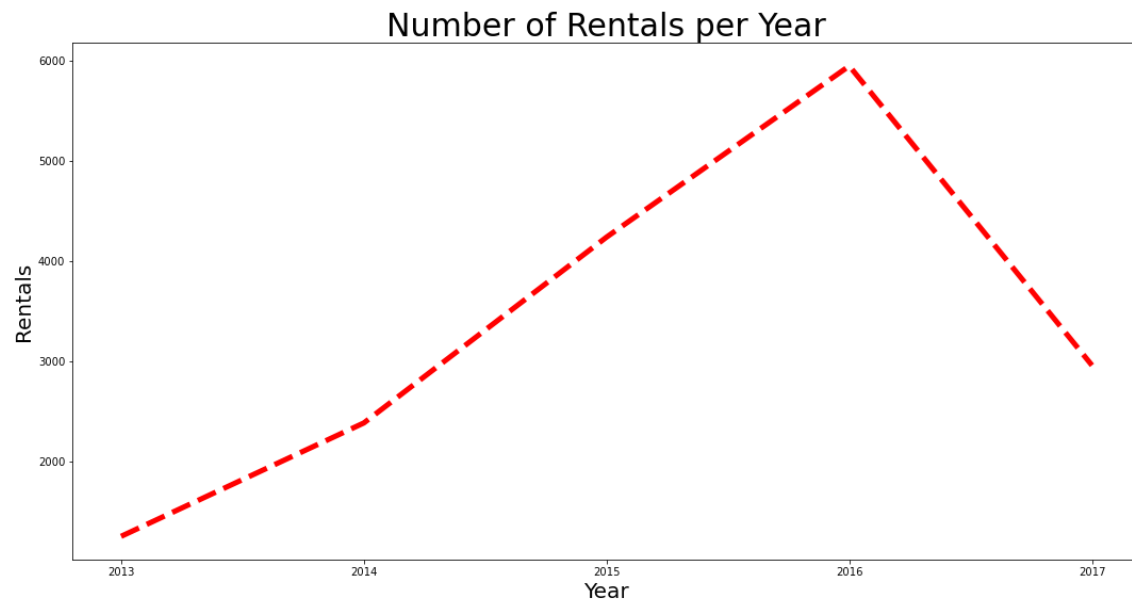
Break down series: Stack: No stack Merge Y axes

Line Y axis: Average Host ID Host ID Use for ordering
Line Y axis: Average Room Price Room Price Use for ordering

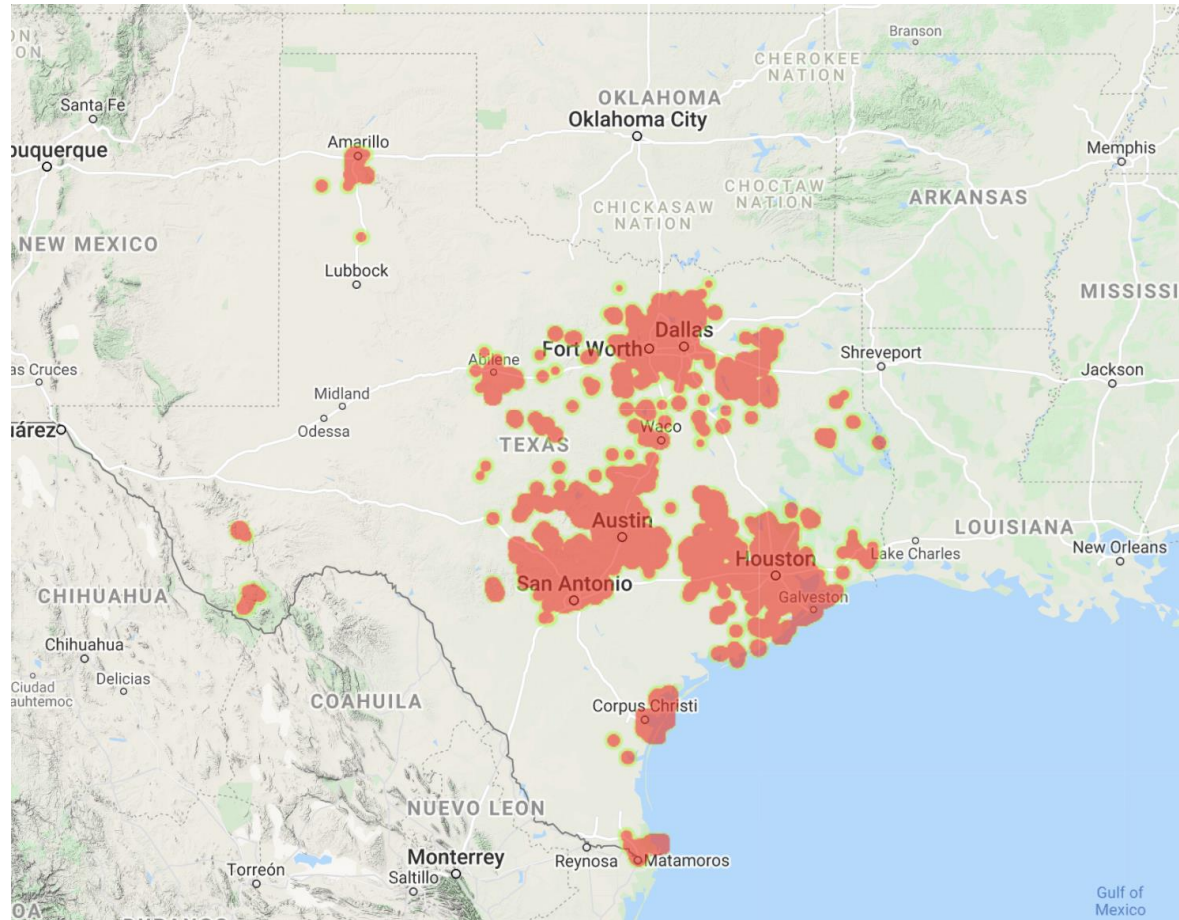


Density Distribution of Airbnb Properties in Texas





Texas Airbnb Rentals



Top 5 Urban & Rural Rentals

Urban Counties

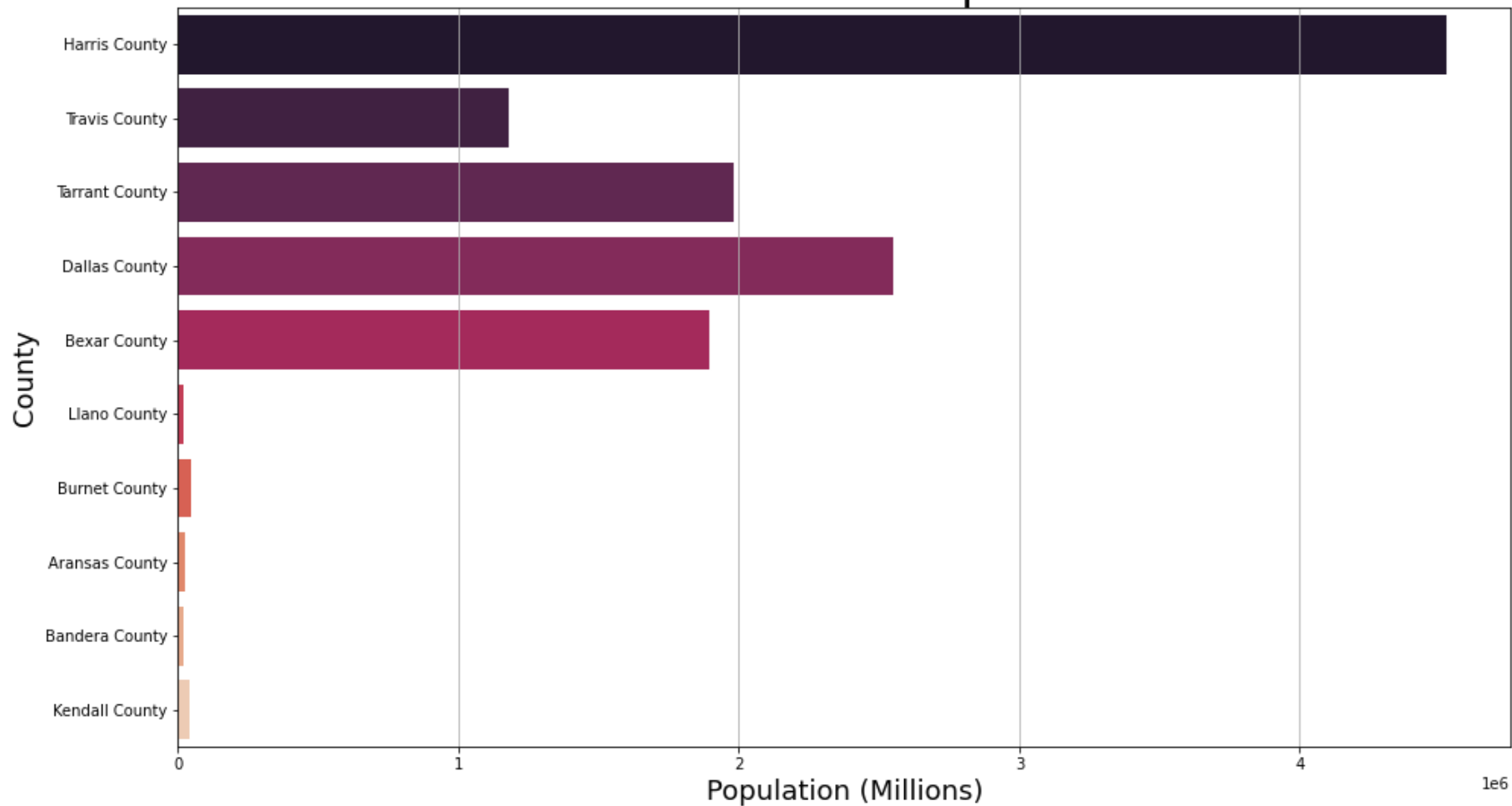
- Harris County
- Travis County
- Tarrant County
- Dallas County
- Bexar County

Rural Counties

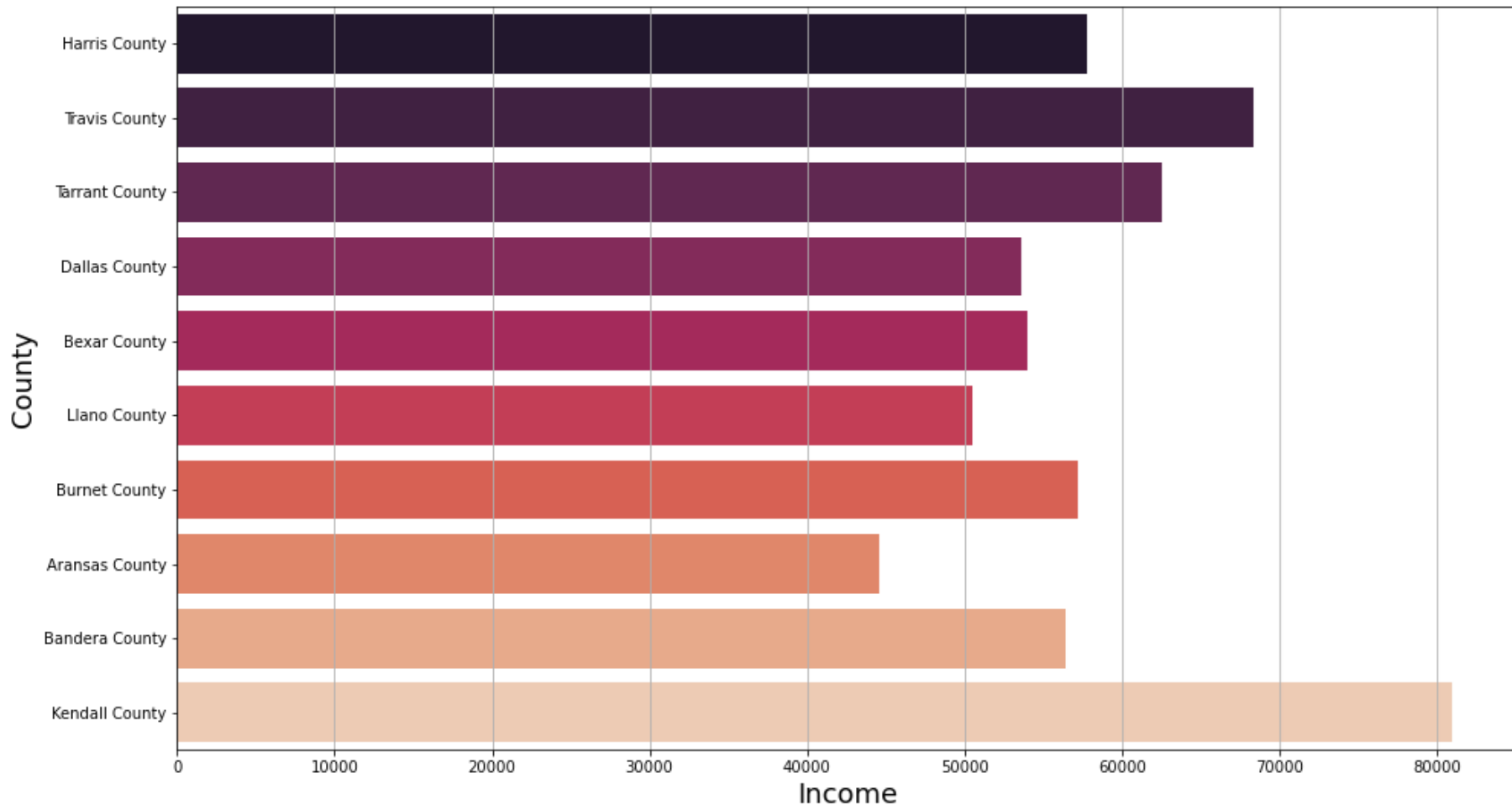
- Llano County
- Burnet County
- Aransas County
- Bandera County
- Kendall County



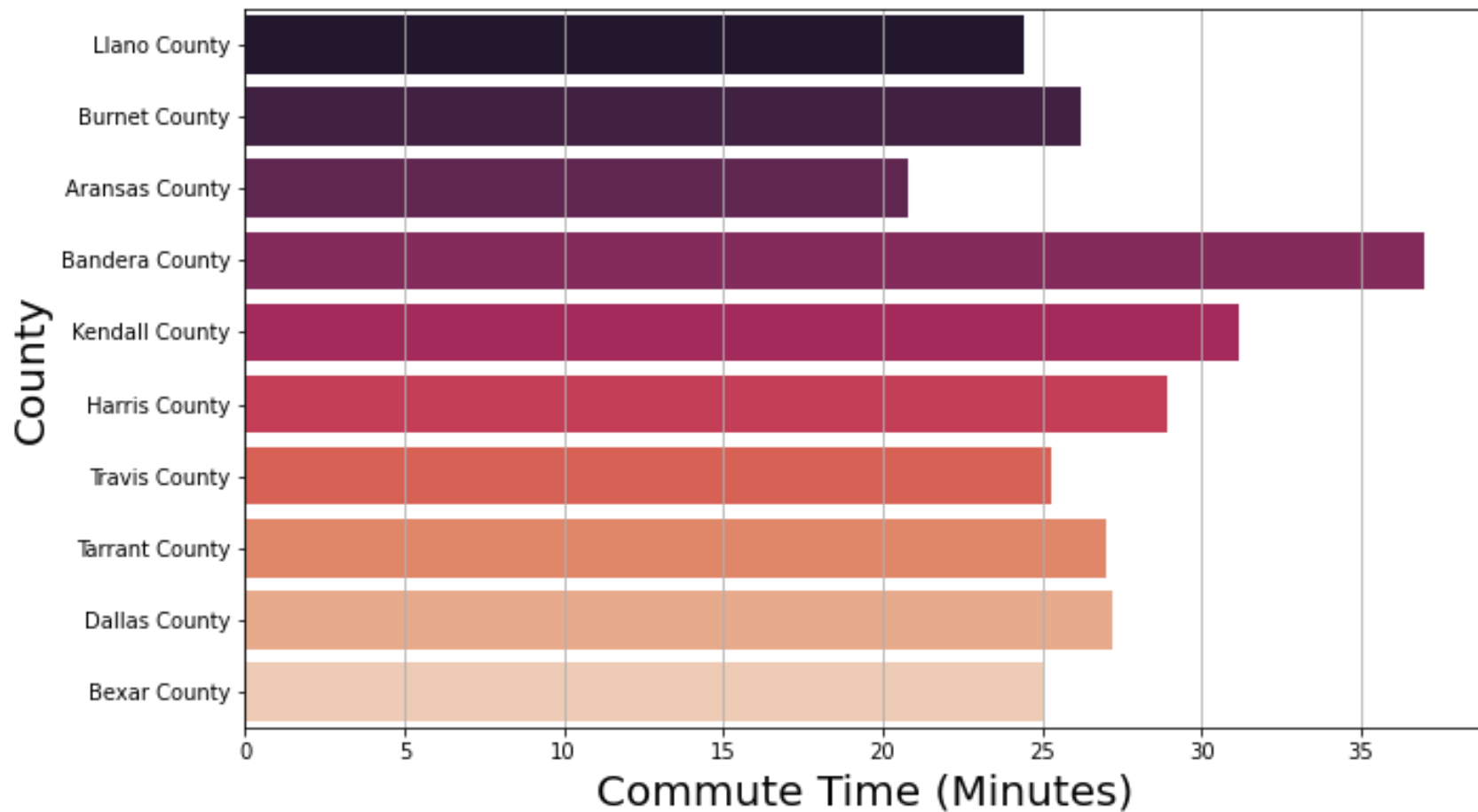
Urban & Rural - Population



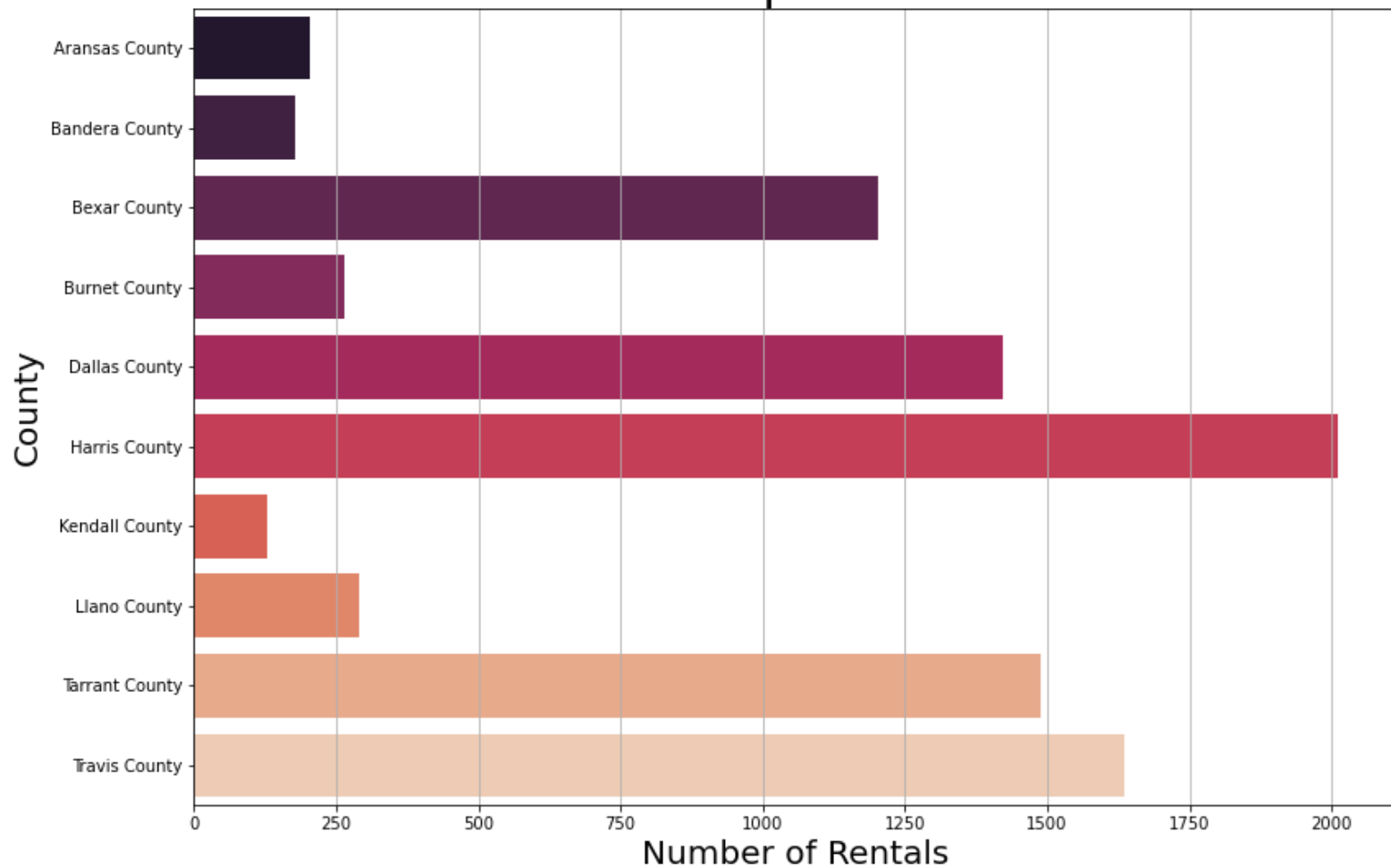
Urban & Rural - Household Income



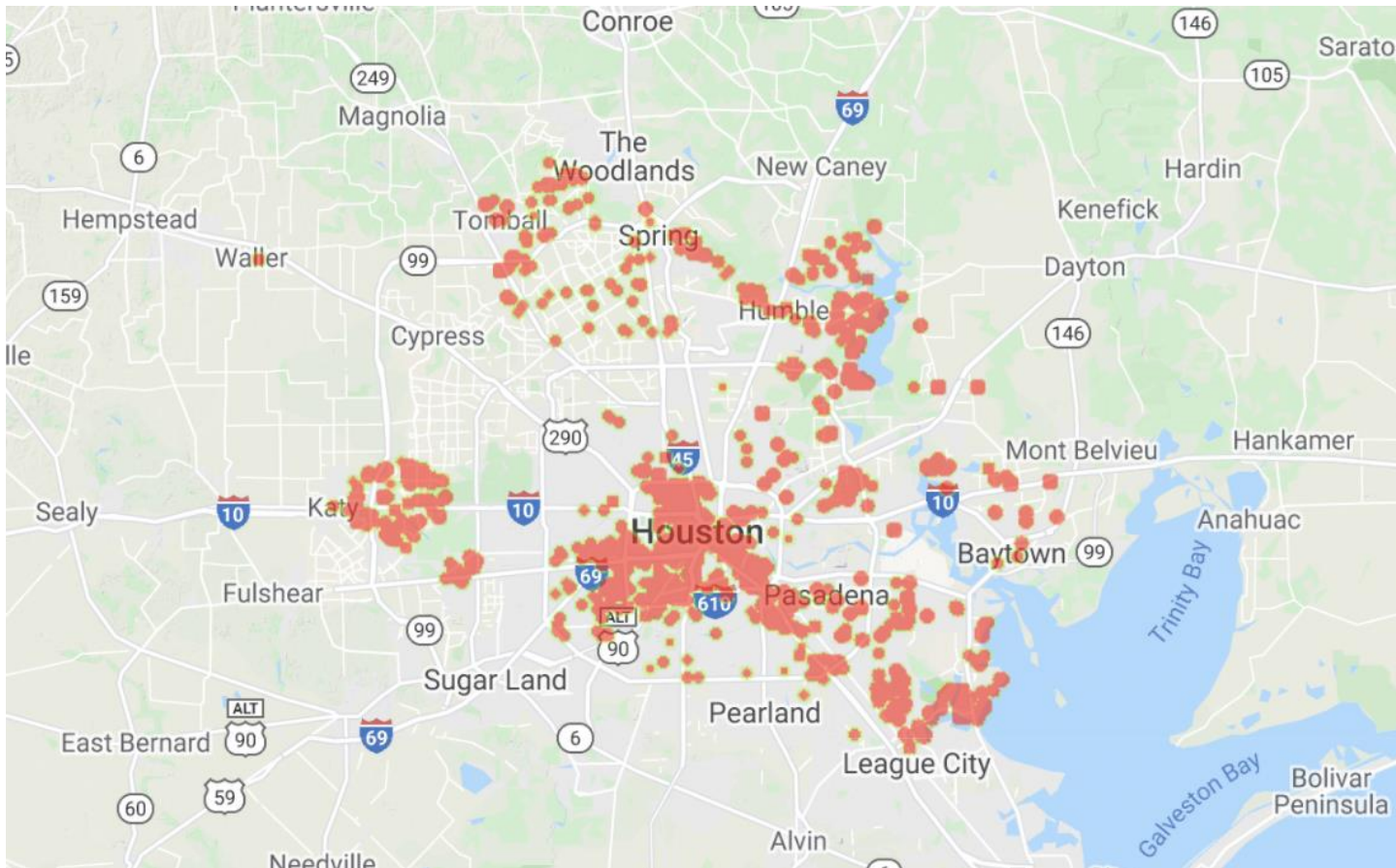
Commute Time



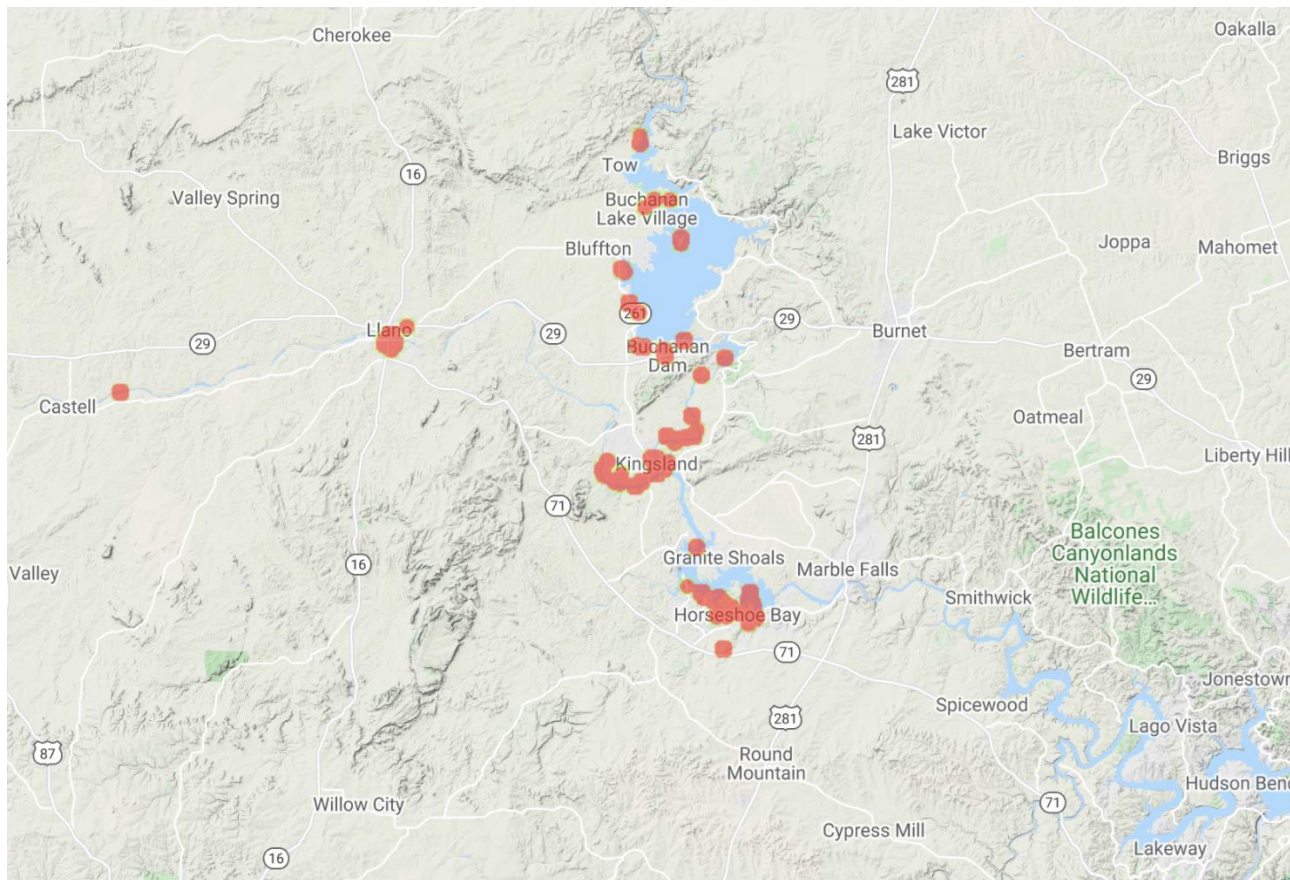
Number of Rentals - Top Urban & Rural Counties



Harris County - Urban



Llano County - Rural



Urban & Rural - Average Rates

County

Harris County

Travis County

Tarrant County

Dallas County

Bexar County

Llano County

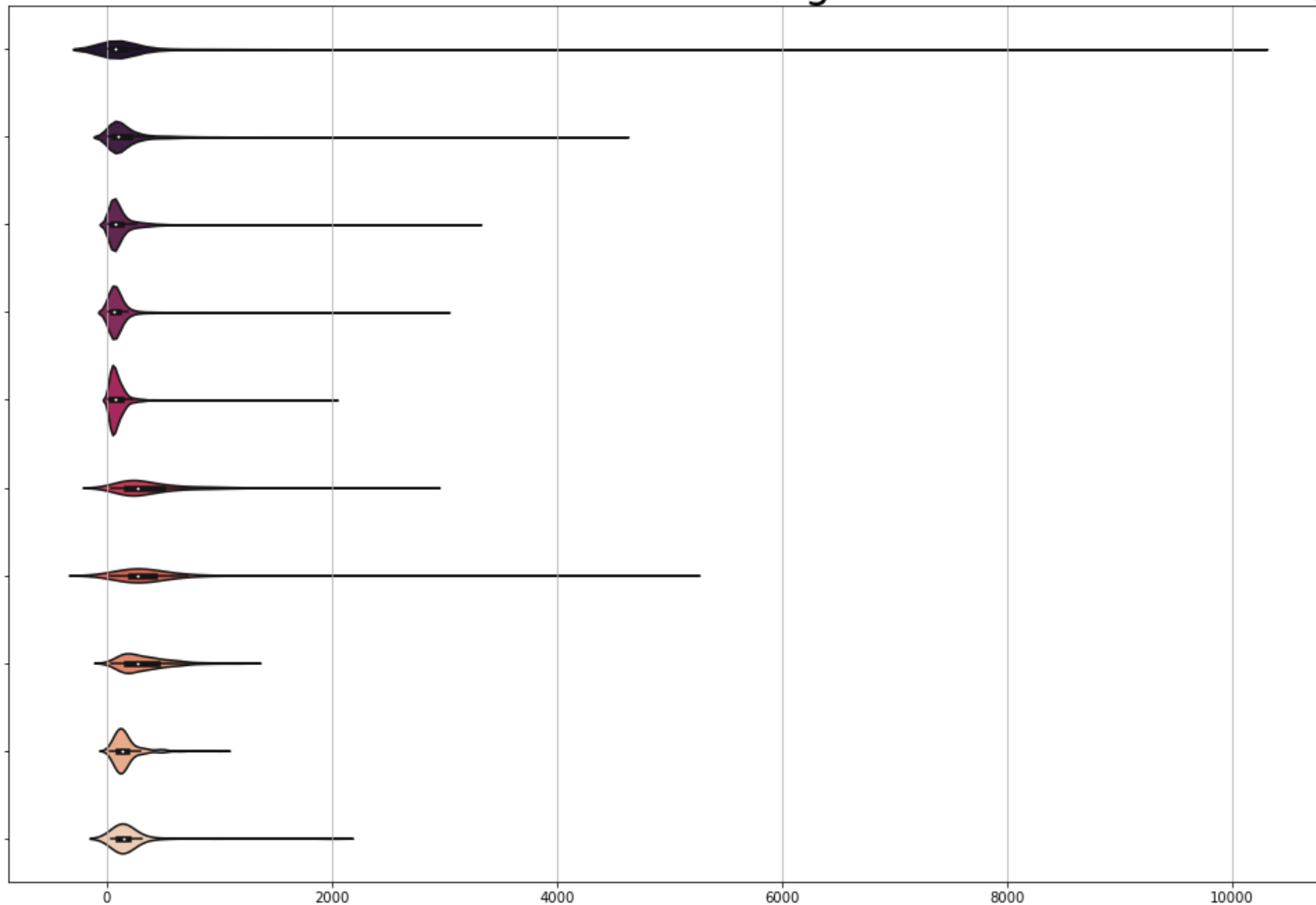
Burnet County

Aransas County

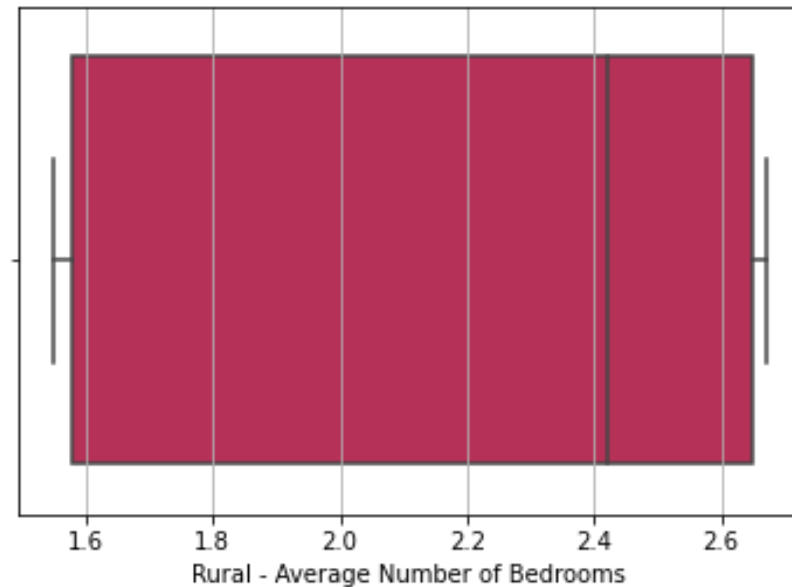
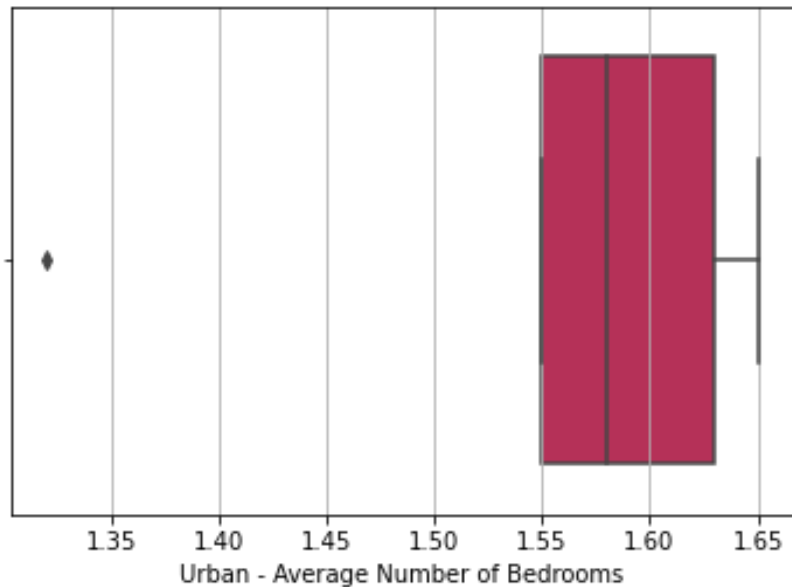
Bandera County

Kendall County

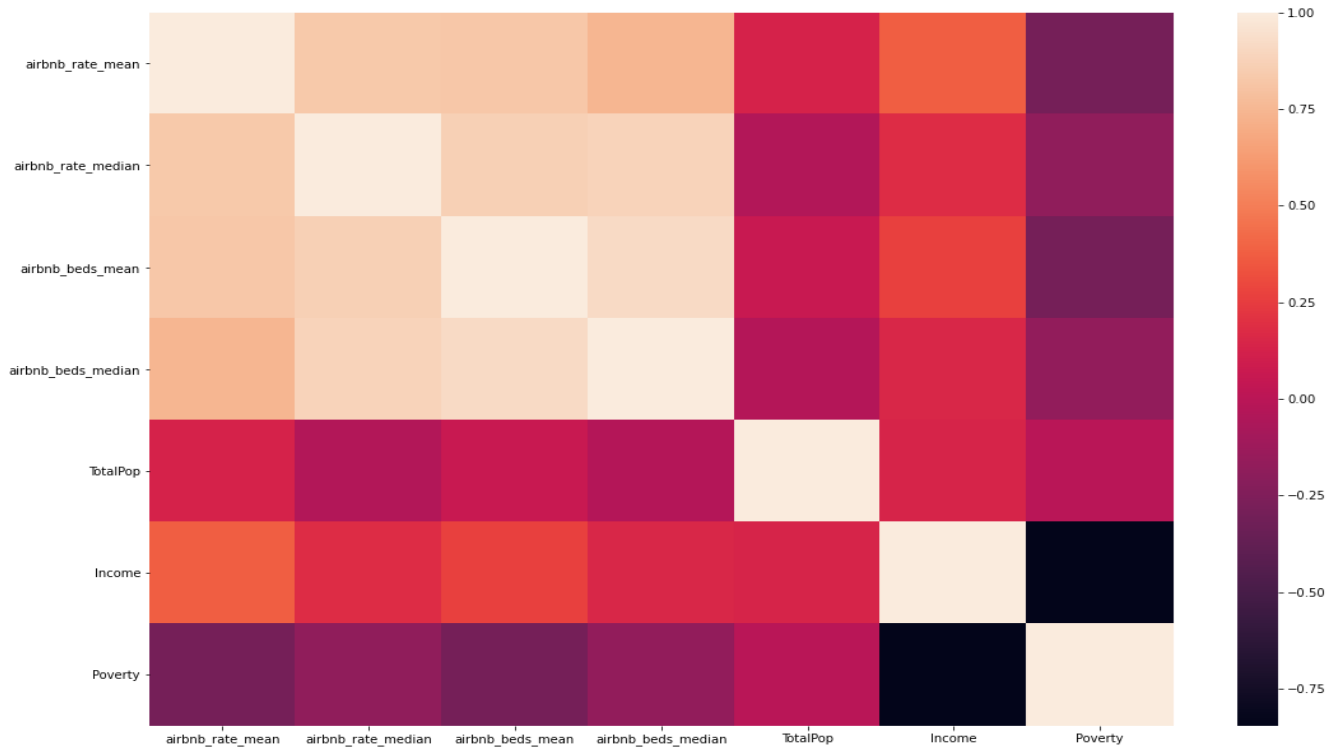
Rates



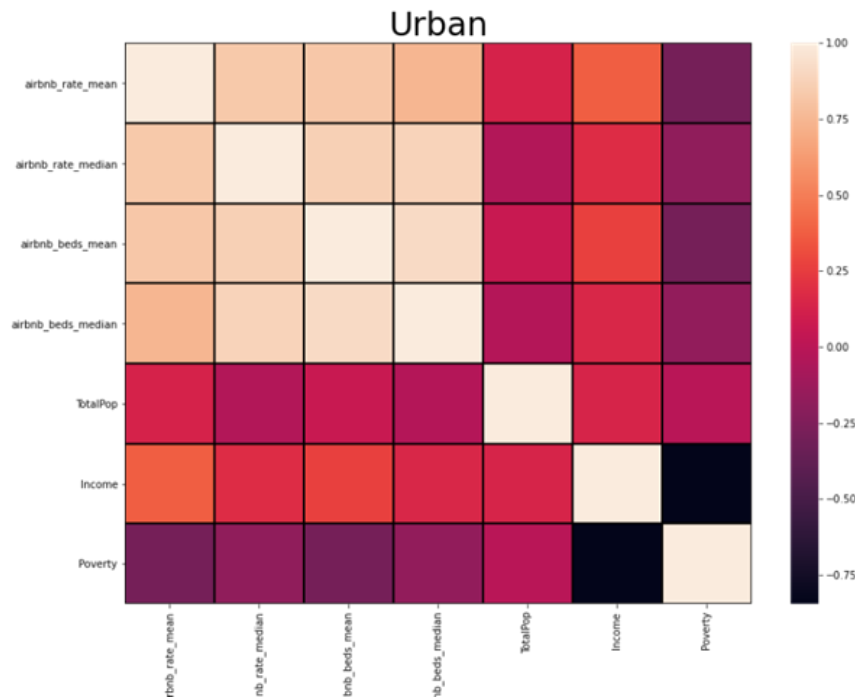
Average Number of Bedrooms



Correlation Heatmap - Urban



Correlation Heatmaps Urban vs Rural



OLS Regression Results-Urban

OLS Regression Results

```

=====
Dep. Variable:      airbnb_rate_mean      R-squared:      0.140
Model:              OLS                  Adj. R-squared:  0.126
Method:             Least Squares        F-statistic:     9.907
Date:               Wed, 04 Aug 2021      Prob (F-statistic): 0.00255
Time:               14:01:18             Log-Likelihood:  -387.89
No. Observations:   63                  AIC:             779.8
Df Residuals:       61                  BIC:             784.1
Df Model:           1
Covariance Type:    nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-47.8696	60.496	-0.791	0.432	-168.839	73.100
Income	0.0033	0.001	3.147	0.003	0.001	0.005

```

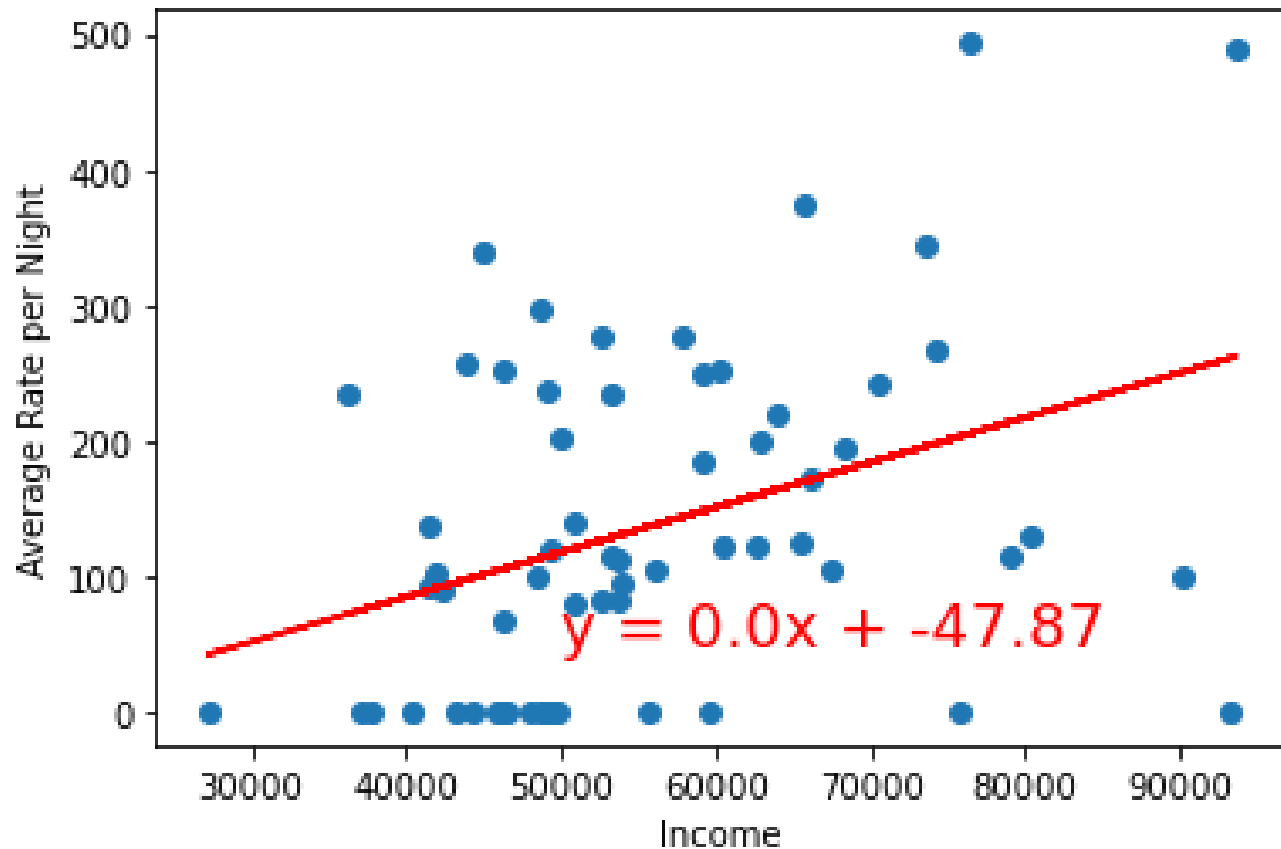
=====
Omnibus:                2.085      Durbin-Watson:      1.911
Prob(Omnibus):           0.352      Jarque-Bera (JB):    2.035
Skew:                   0.413      Prob(JB):            0.362
Kurtosis:               2.695      Cond. No.            2.38e+05
=====

```

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 2.38e+05. This might indicate that there are strong multicollinearity or other numerical problems.

Regression Chart-Urban



OLS Regression Results-Rural

OLS Regression Results

Dep. Variable:	airbnb_rate_mean	R-squared:	0.104
Model:	OLS	Adj. R-squared:	0.099
Method:	Least Squares	F-statistic:	21.85
Date:	Wed, 04 Aug 2021	Prob (F-statistic):	5.59e-06
Time:	16:38:15	Log-Likelihood:	-1208.6
No. Observations:	191	AIC:	2421.
Df Residuals:	189	BIC:	2428.
Df Model:	1		
Covariance Type:	nonrobust		

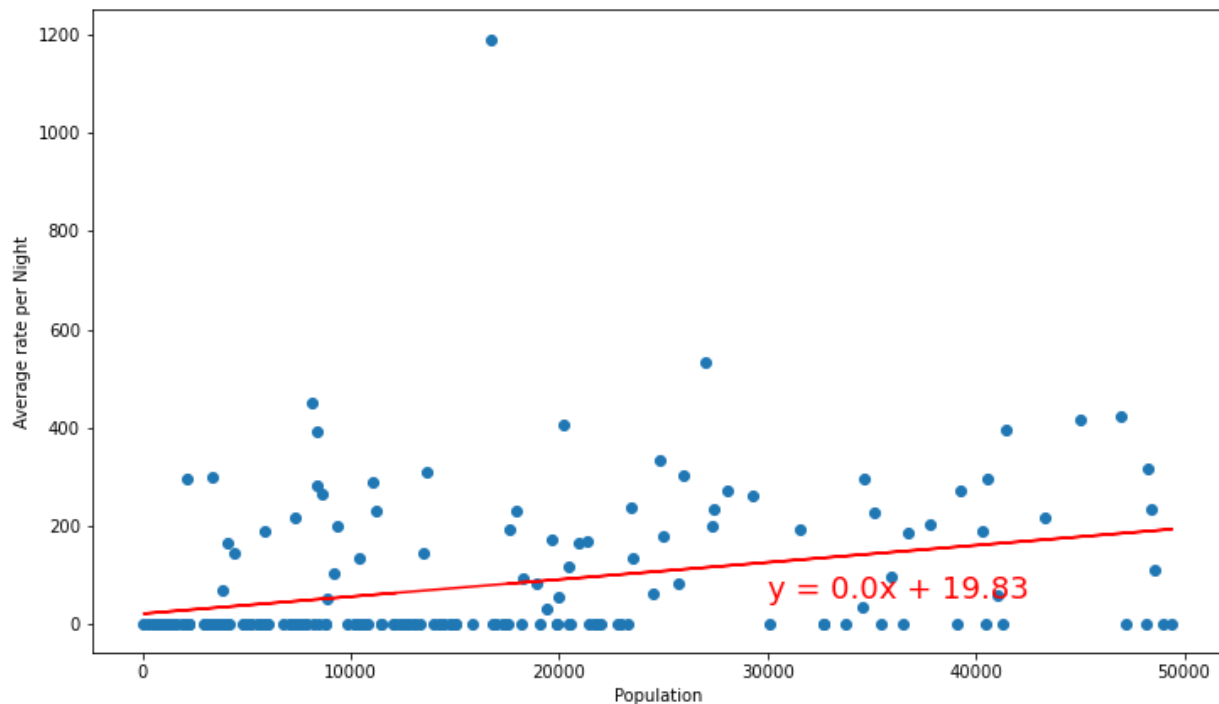
	coef	std err	t	P> t	[0.025	0.975]
const	19.8261	15.174	1.307	0.193	-10.107	49.759
TotalPop	0.0035	0.001	4.674	0.000	0.002	0.005

Omnibus:	183.991	Durbin-Watson:	2.071
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4608.124
Skew:	3.584	Prob(JB):	0.00
Kurtosis:	25.971	Cond. No.	3.13e+04

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 3.13e+04. This might indicate that there are strong multicollinearity or other numerical problems.

Regression Chart-Rural



Conclusion

We recommend operating an airbnb rental in a rural area outside a major metropolitan center that has outdoor destinations to maximize profit.



- **There will be higher rates in rural areas due to less competition.**

- Rates are slightly higher in rural areas. We also found that the rental size is larger which could contribute to the higher rates.



There are more rentals in areas with tourist attractions.

- (i.e., Houston Theme Parks, Museums and business travel)



Population and income affect Airbnb rates.

- From the correlation data, we can see that rural areas that have a larger total population have higher rates than urban areas. Moreover, urban area rates are affected by income rather than population. This would mean that it is best to target rural counties with a total population of 40,000 to 50,000 people.

Limitations & Future Work

- Data: Not current
 - Airbnb data is from 2013-2017 (only 6 months of 2017 data available)
 - Census data is from 2017
 - Airbnb data when compiled only shows data from 106 of 254 counties in Texas.
- Future work:
 - Look at the effects of rentals during COVID
 - How long will it take for the market to recover?
 - Seasonality of Airbnb rentals

References

Datasets:

<https://www.kaggle.com/PromptCloudHQ/airbnb-property-data-from-texas>

<https://www.kaggle.com/muonneutrino/us-census-demographic-data>

Inspiration:

<https://www.kdnuggets.com/2017/08/insights-data-mining-airbnb.html>

<https://public.opendatasoft.com/explore/dataset/air-bnb-listings/map/>

Graphics:

https://www.traveltexas.com/things-to-do/outdoor-adventure/?gclid=CjwKCAjw0qOIbBhEiwAyvVcfzmOWPSjpt_oIR-it5UaiZhYC_qyne15B3Tc1uD5HuzdE6kB0FIRtBoCUPAQAvD_BwE

<https://www.airbnb.com>

Definition of Urban vs Rural

<https://www.census.gov/programs-surveys/geography/guidance/geo-areas/urban-rural/2010-urban-rural.html>

Q & A Time!