

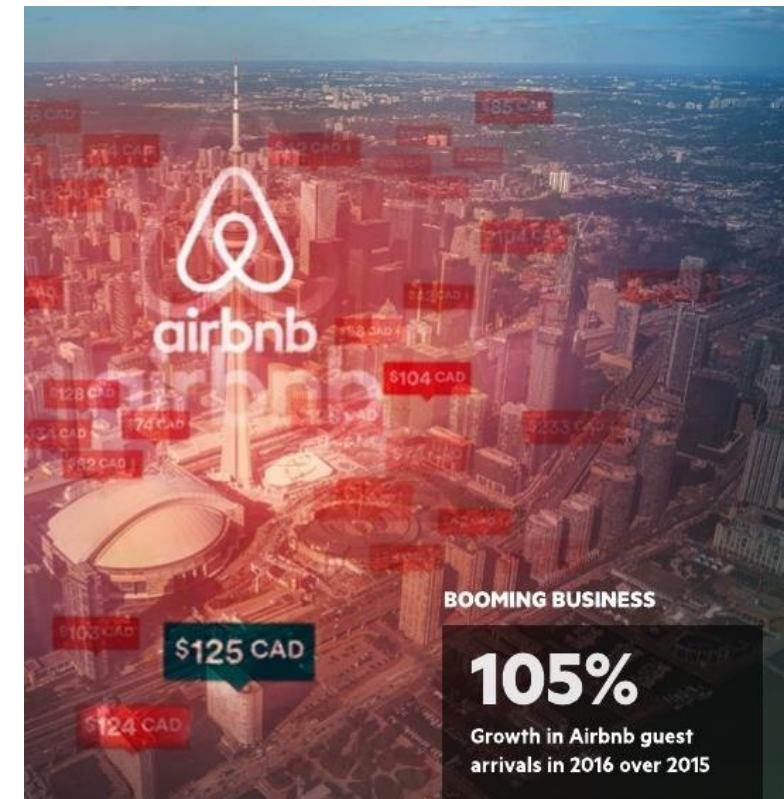
INVESTIGATING THE FACTORS THAT AFFECT THE NIGHTLY PRICE OF TORONTO AIRBNB LISTINGS

Nicholas Jalen Cheng

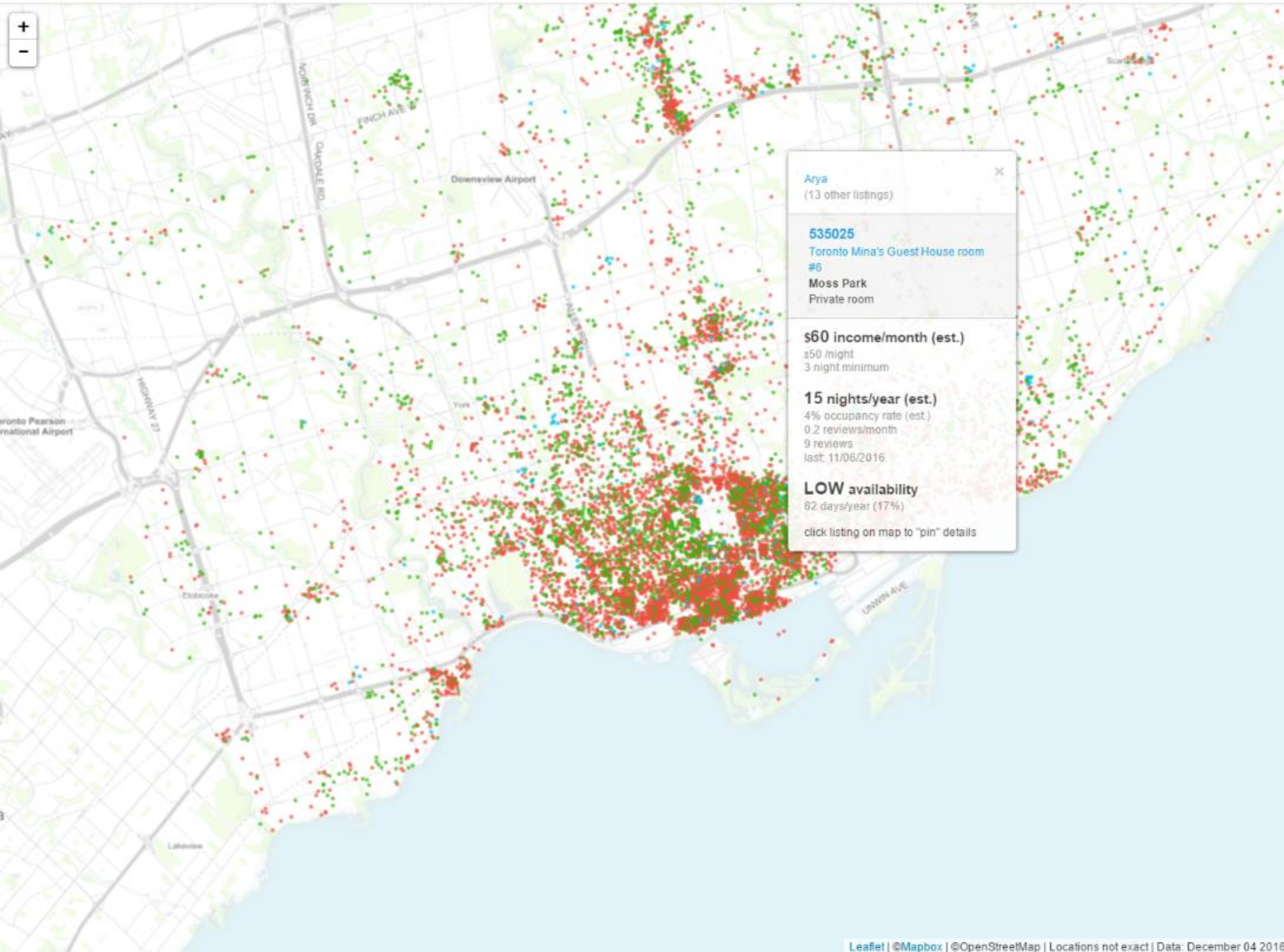
Alex Kwan

Ron Suprun

Jiming Zhou







Toronto

Filter by:
Toronto

12,029
out of 12,029 listings (100%)

[About Airbnb in Toronto](#)

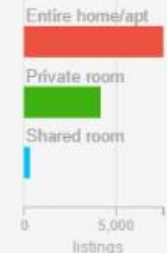
How is Airbnb really being used in and affecting your neighborhoods?

Room Type

☐ Only entire homes/apartments

Airbnb hosts can list entire homes/apartments, private or shared rooms.

Depending on the room type, [availability](#), and [activity](#), an airbnb listing could be more like a hotel, disruptive for neighbors, taking away housing, and [illegal](#).



62.7%
entire homes/apartments
\$123
price/night
7,544 (62.7%)
entire home/apartments
4,155 (34.5%)
private rooms
330 (2.7%)
shared rooms

Activity

☐ Only [recent](#) and [frequently](#) booked

Airbnb guests may leave a review after their stay, and these can be used as an indicator of airbnb activity.

The minimum stay, price and number of reviews have been used to estimate the **occupancy rate**, the number of **nights per year** and the **income per month** for each listing.

How does the income from Airbnb compare to a long-term lease?

Do the number of nights booked per year make it impossible for a listing to be used for residential housing?

And what is renting to a tourist full-time rather than a resident doing to our neighborhoods and cities?

90
estimated nights/year
1.4
reviews/listing/month
166,437
reviews
\$123
price/night
24.7%
estimated occupancy
\$844
estimated income/month

	2011	%
<i>Total</i>	29,155	100%
Children 0-14	2,195	7.5%
Youth 15-24	3,895	13.4%
Working Age 25-64	18,260	62.6%
Seniors 65+	4,805	16.5%

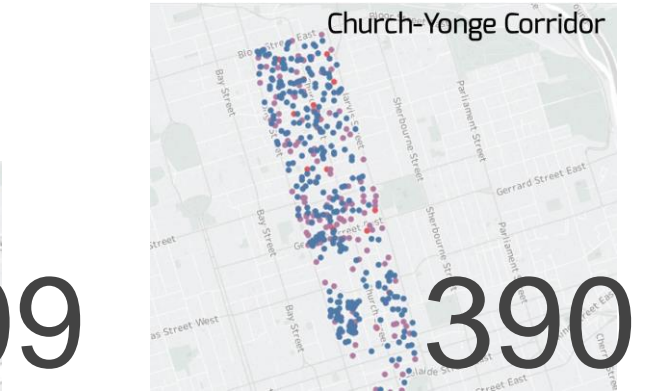
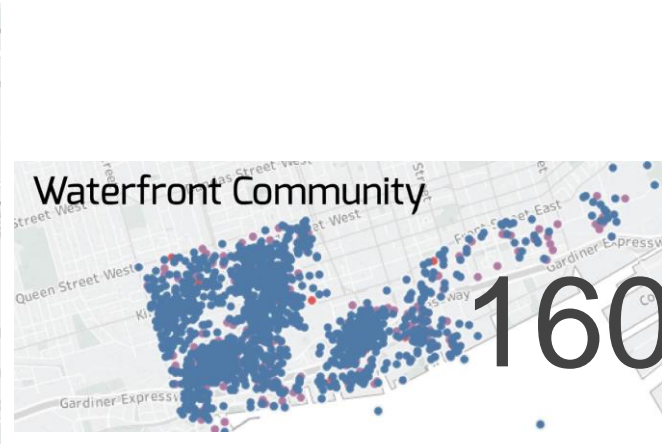
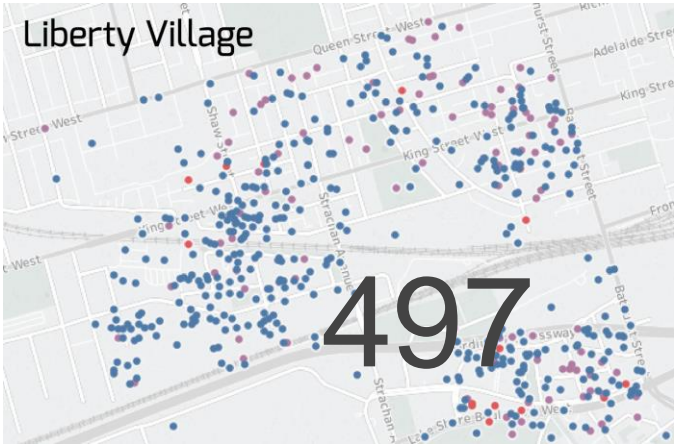
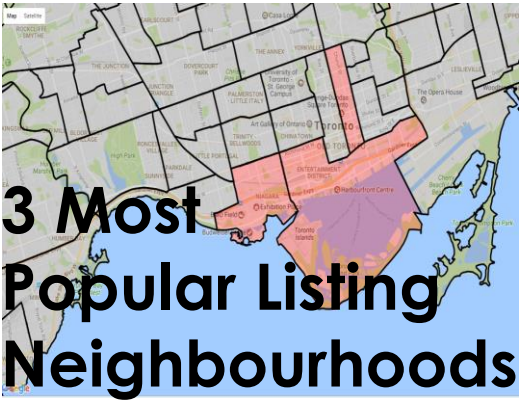
Income Distribution by Groups 2005			
Category	Census Family	Private Household	1-person Private Household
Average income - \$	181,587	112,510	65,735

Toronto Real Estate Board										
SUMMARY OF RENTAL TRANSACTIONS					APARTMENTS, FOURTH QUARTER 2016 CITY OF TORONTO MUNICIPAL BREAKDOWN					
	All Apartments		Bachelor		One-Bedroom		Two-Bedroom		Three-Bedroom	
	Total Listed ¹	Total Leased ²	Leased ²	Avg. Lease Rate ³	Leased ²	Avg. Lease Rate ³	Leased ²	Avg. Lease Rate ³	Leased ²	Avg. Lease Rate ³
TREB Total	9,545	5,717	227	\$1,512	3,379	\$1,776	1,988	\$2,415	123	\$3,283
City of Toronto Total	7,769	4,606	223	\$1,517	2,776	\$1,819	1,505	\$2,567	102	\$3,493
Toronto West	992	666	5	\$1,305	392	\$1,628	252	\$2,203	17	\$2,629
Toronto W01	94	60	0	-	40	\$1,837	20	\$2,657	0	-
Toronto W02	43	23	1	\$1,500	14	\$1,698	7	\$3,200	1	\$2,550
Toronto W03	6	4	0	-	2	\$1,275	1	\$1,900	1	\$2,250
Toronto W04	209	147	0	-	84	\$1,499	60	\$1,829	3	\$2,483
Toronto W05	44	21	2	\$1,038	13	\$1,650	5	\$2,010	1	\$2,375
Toronto W06	305	203	2	\$1,475	106	\$1,682	91	\$2,461	4	\$3,788
Toronto W07	13	6	0	-	3	\$1,667	3	\$2,333	0	-
Toronto W08	234	175	0	-	118	\$1,611	51	\$2,013	6	\$2,154
Toronto W09	8	2	0	-	0	-	1	\$1,950	1	\$2,000
Toronto W10	36	25	0	-	12	\$1,462	13	\$1,712	0	-

29.2%
Houses



63.2%
Condos



▶ EASY TO INTERPRET

▶ FLEXIBLE

▶ GOOD STARTING POINT FOR ANALYSIS

Why Multivariable Linear Regression?

	Regular Regression	➡ Add Lasso
MSE	17424	17728
R²	0.264	0.2434

G	H	I	J
Property Type	isHou	isTownHou	isApt
2	1	0	0
3	0	0	1
1	0	1	0

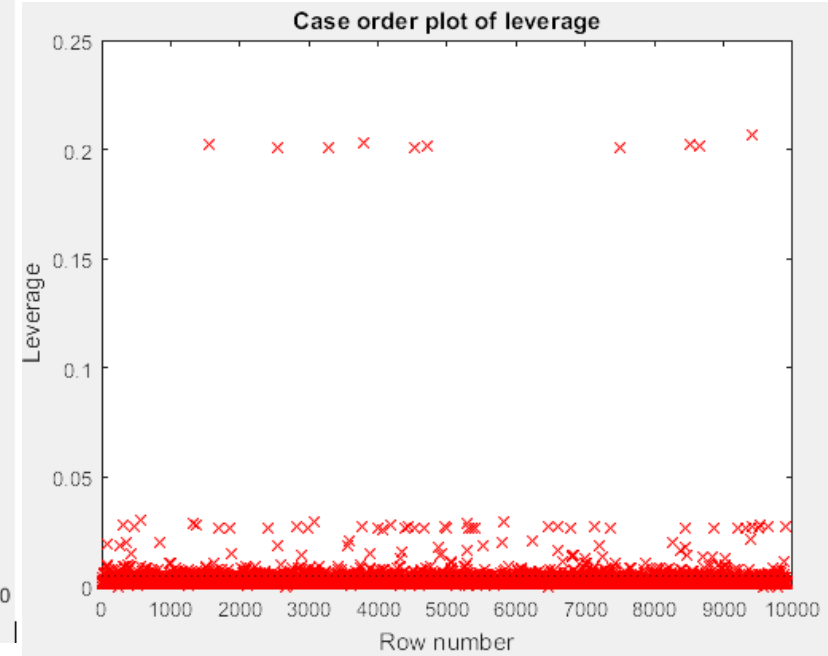
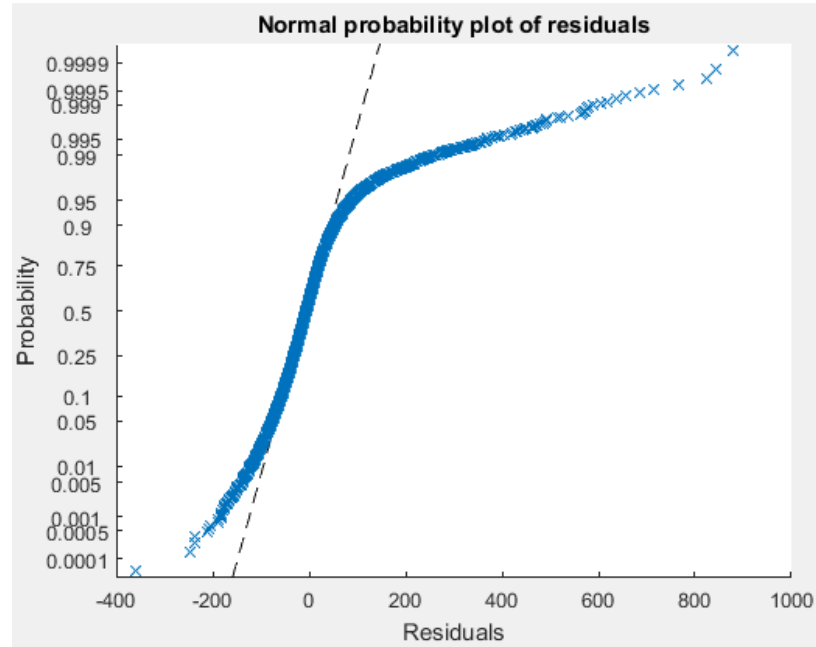
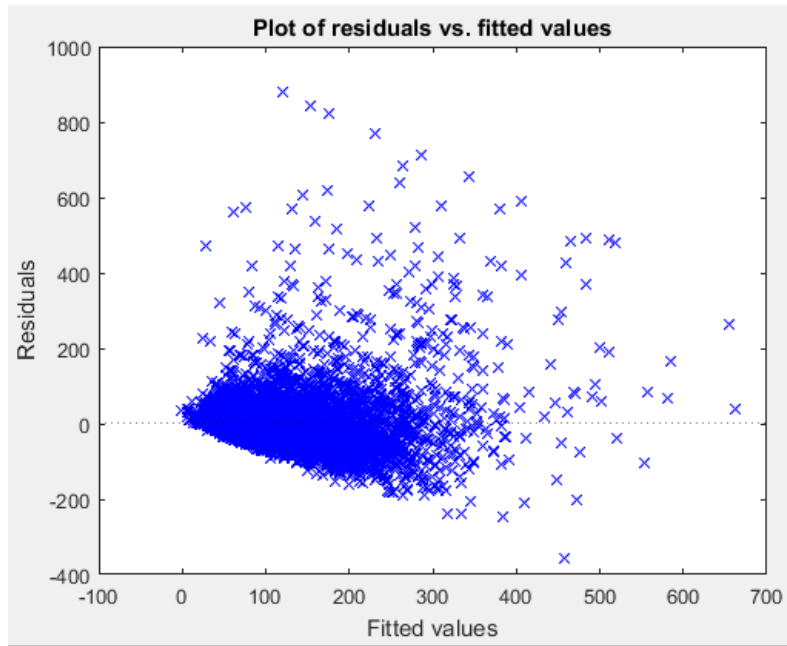
$$\begin{aligned}
 &\min_{\beta_0, \beta_1, \dots, \beta_K} \quad \frac{1}{n} \sum_{i=1}^n (\hat{y}_i - y_i)^2 + \lambda \sum_{k=1}^K |\beta_k| \\
 &\text{s.t.} \quad \hat{y}_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_K x_{iK}, \quad i = 1, \dots, n
 \end{aligned}$$

	Regular Regression	Add Lasso	Remove Lasso & Add Neighbourhood Data	Remove Insignificant Variables	Remove Outliers
MSE	17424	17728	17161	17689	4395.69
R ²	0.264	0.2434	0.285	0.258	0.504

	2001	%	2006
Total	28,835	100%	30,175
Children 0-14	4,800	16.6%	4,740
Youth 15-24	4,445	15.4%	4,325
Working Age 25-64	16,470	57.1%	16,815
Seniors 65+	3,120	10.8%	4,295



CHECKING LINEAR MODEL ASSUMPTIONS



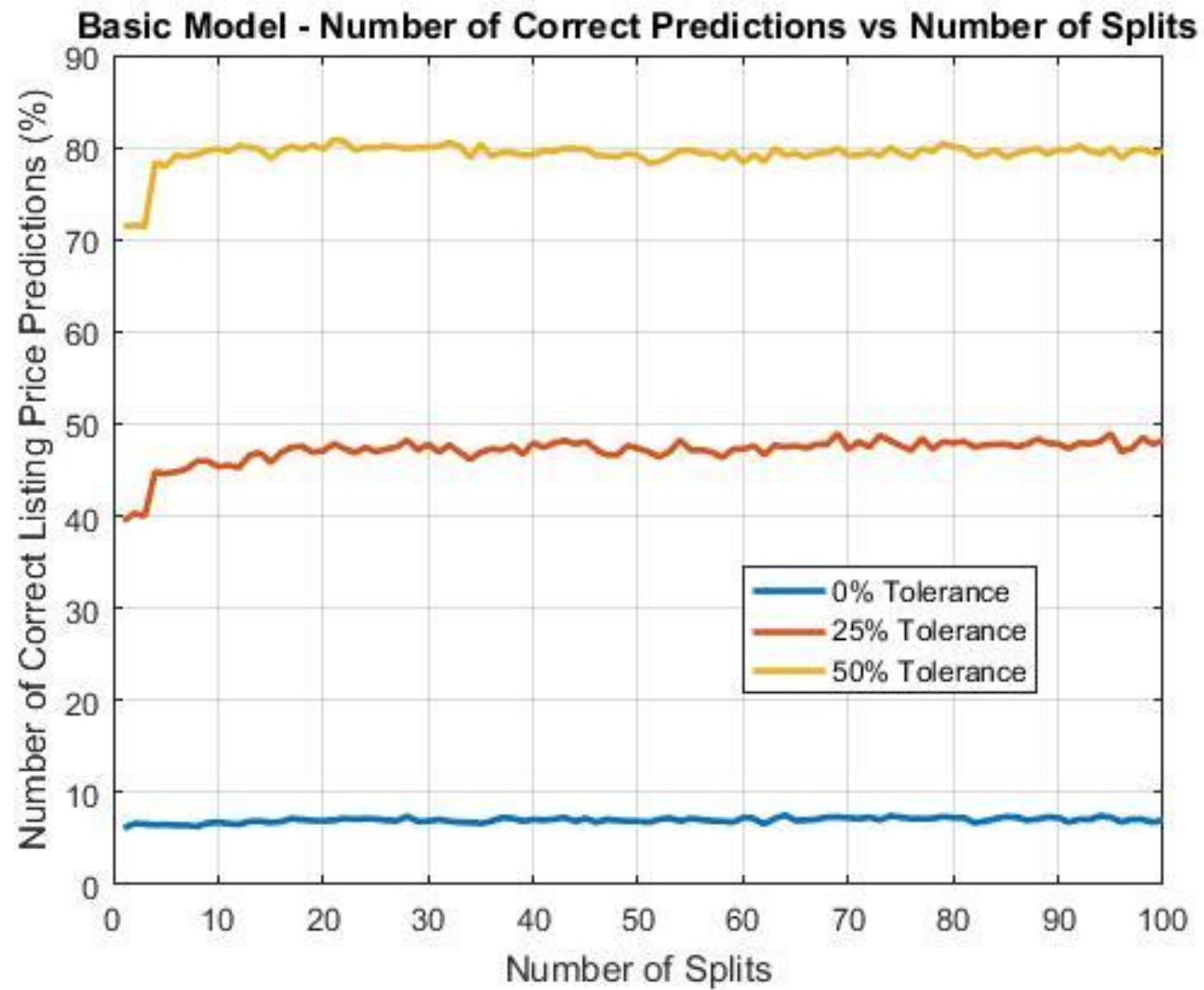
NORMALITY ASSUMPTION NOT MET - LET'S TRY NON-LINEAR MULTIPLE REGRESSION

Warning: Regression design matrix is rank deficient to within machine precision.

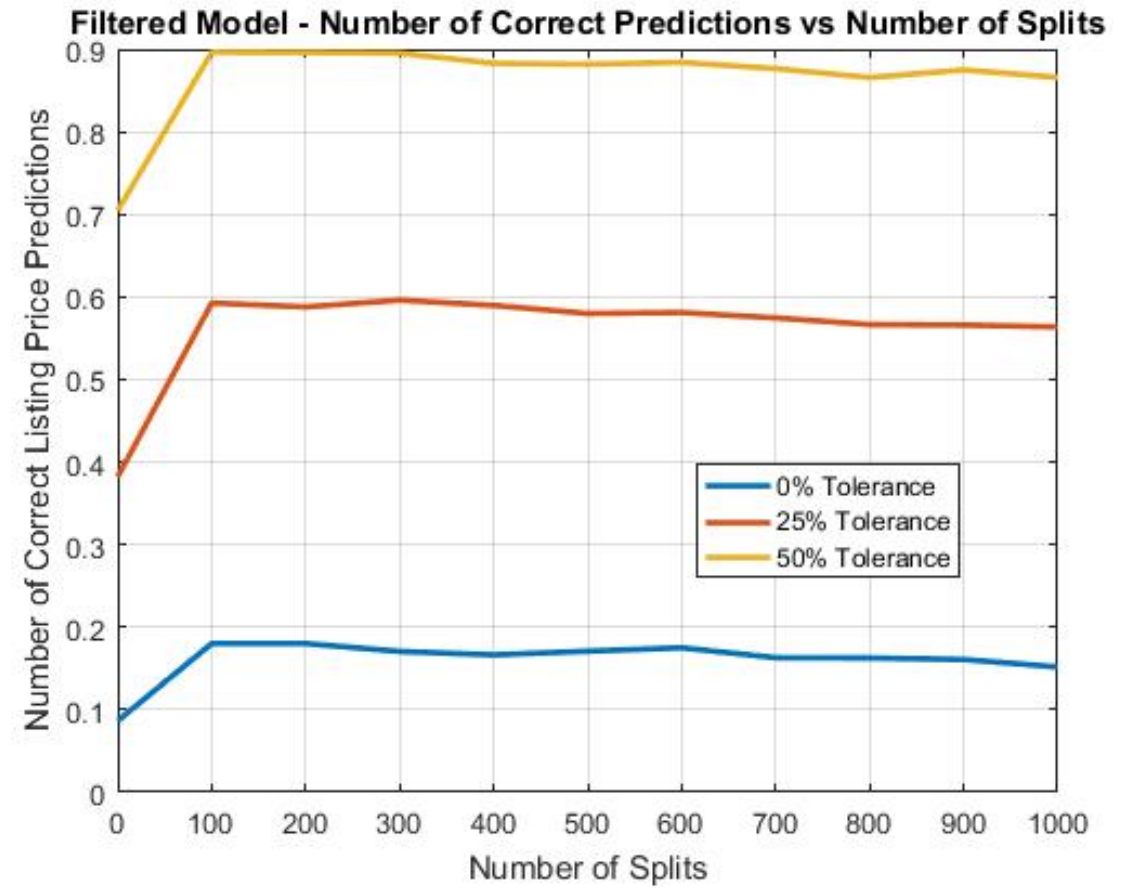
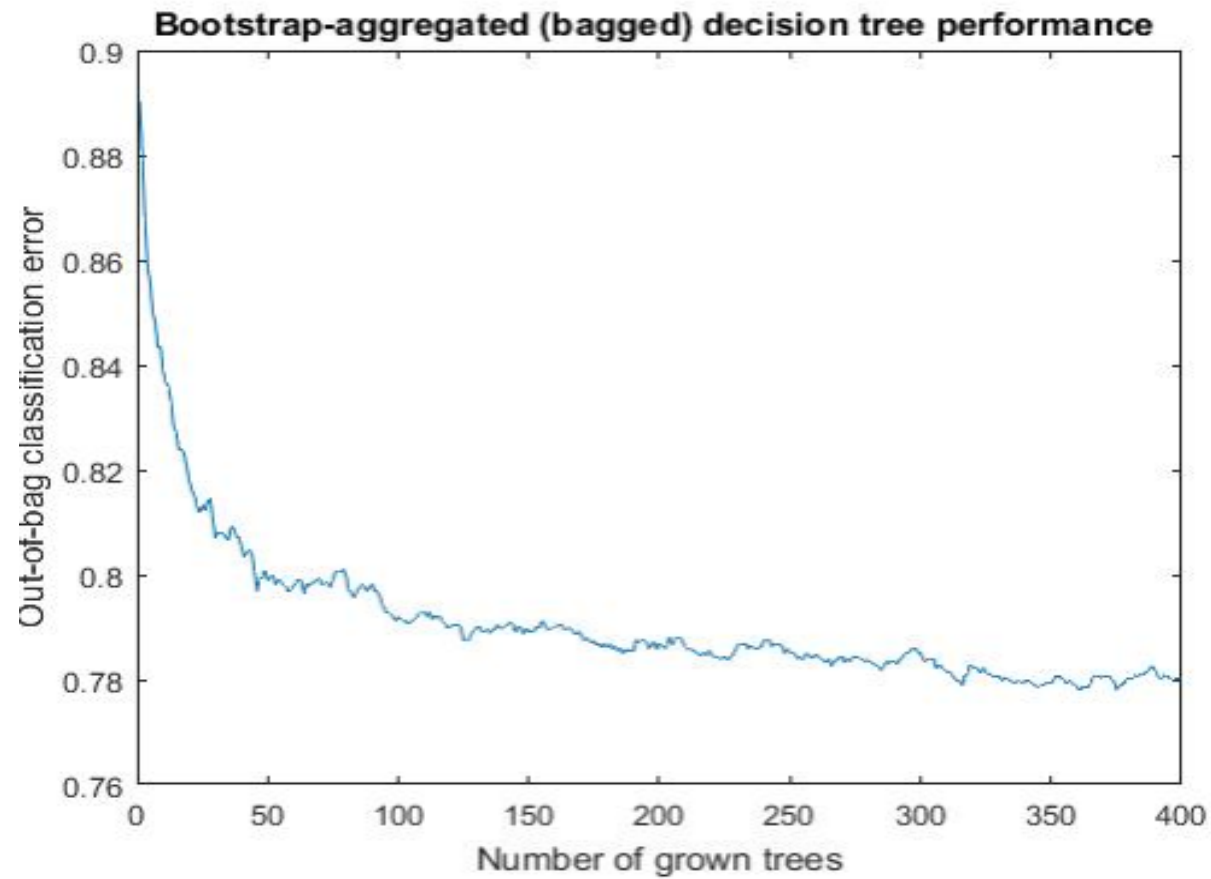


- ▶ Relatively Fast
- ▶ All Types of Variables (Numeric, Binary, Categorical)
- ▶ Resistant to irrelevant variables
- ▶ Easily interpretable model representation

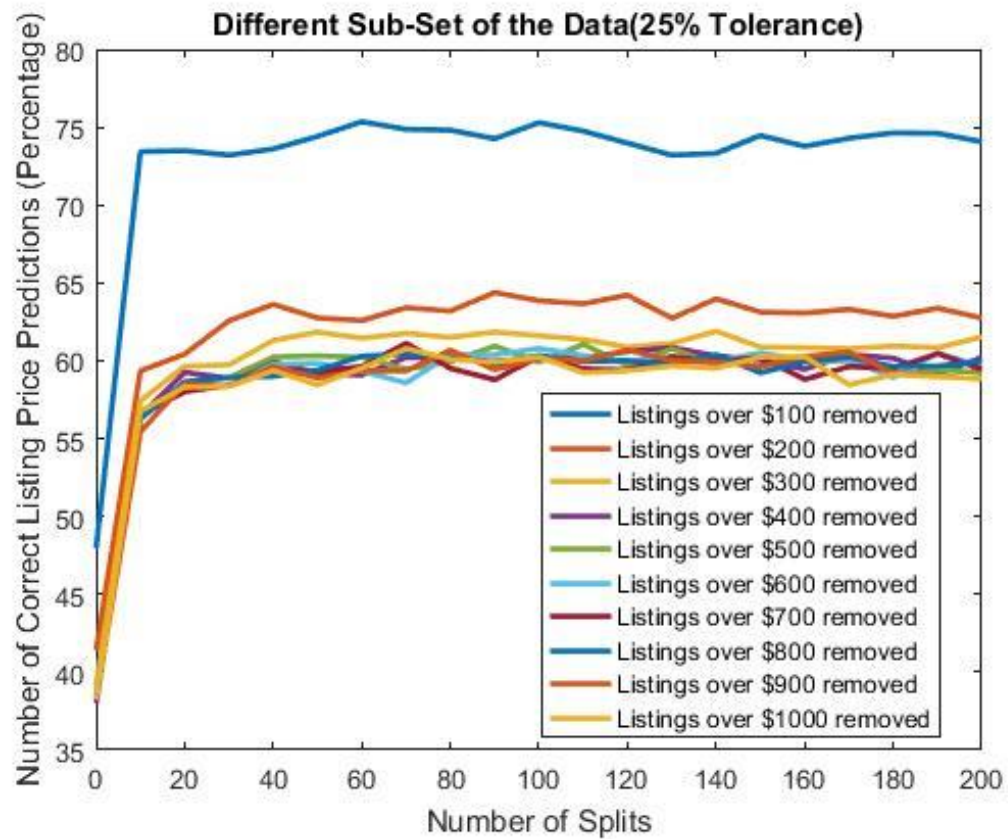
Why CART?



Basic Model that includes everything



Filtered Model that predicts with a \$10 interval, and only with listings above 40% occupancy rate



The listing traits that contribute to the price distinction beyond \$100 aren't captured within this data set.

Most Significant Variables:

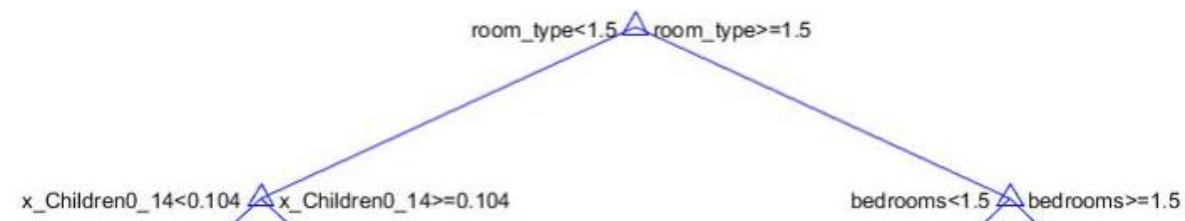
Room Type

Number of Bedrooms

Rental Price

Number of Bathrooms

Percentage of Children Between the age 0 to 14





13 Reviews ★★★★★

Search reviews

Summary

Accuracy

★★★★★

Location

★★★★★

Communication

★★★★★

Check In

★★★★★

Cleanliness

★★★★★

Value

★★★★★



Dave and Deb communication was perfect from the beginning. They are very attentive and helpful hosts. The place is absolutely nice and well equipped! It is so clean and so bright, I recommend it to anyone who visits Edmonton. Plus the location is really convenient! My friend and I will stay here again for sure ^_^





