

Airbnb Predicted Pricing

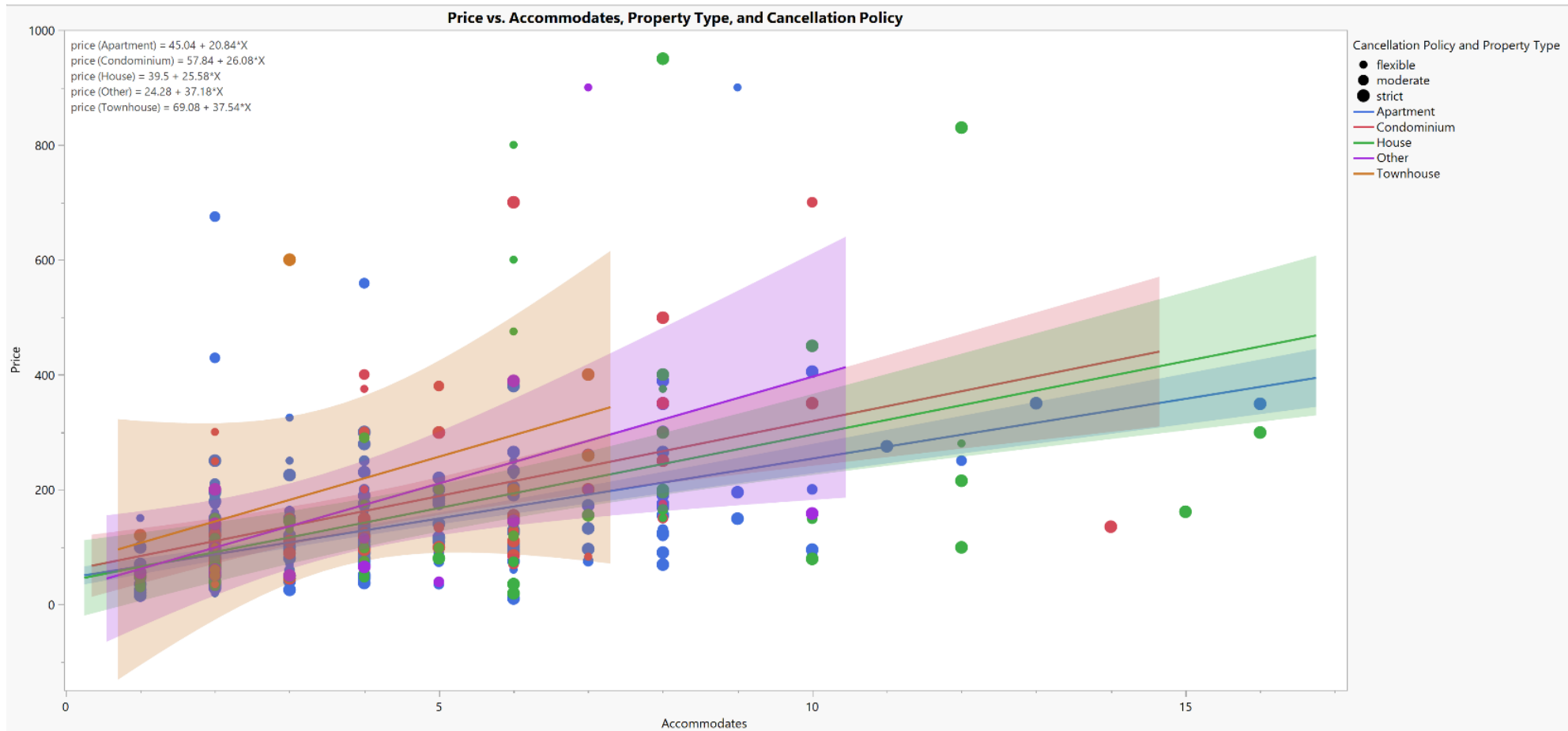
Airbnb Dataset

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Project Background and Data Description

- Dataset: Airbnb listings subset
- Random sample from Airbnb hosts
- Listings in Chicago, IL from Aug 2008-May 2017
- Response variable: price
- Quantitative explanatory variable:
accommodates
- Qualitative explanatory variables:
cancellation_policy (strict as reference group),
property_type (house as reference group)

Scatterplot



Results from Multiple Regression Analysis

	Full Model	Interaction Model	Transformation Model	Quadratic Model	Transformation +Interaction Model
Number of Explanatory Variables	7	11	7	8	11
Adjusted R^2	0.257539	0.257571	0.379474	0.269536	0.376293
Mallow's Cp	8	12	8	9	12
RMSE	108.49583	108.49354	0.5716368	107.61571	0.5731005
F-test p-value	<0.0001 F-test: 25.7270	<0.0001 F-test: 16.7380	<0.0001 F-test: 44.5939	<0.0001 F-test: 24.0159	<0.0001 F-test: 28.3686
Largest p-value for all individual coefficient t-tests: $\beta_i = 0$ vs $\beta_i \neq 0$	0.8791 (moderate)	0.8748 (condo*accommodates)	0.8521 (moderate)	0.9493 (moderate)	0.9185 (moderate)

Models

Full: property_type, cancellation_policy, accommodates, (response price)

Interaction: property_type*accommodates + Full

Transformation: property_type, cancellation_policy, $\ln(\text{accommodates})$, (response $\ln(\text{price})$)

Quadratic: accommodates*accommodates + Full

Transformation + Interaction:

property_type* $\ln(\text{accommodates})$, property_type, cancellation_policy, $\ln(\text{accommodates})$, (response $\ln(\text{price})$)

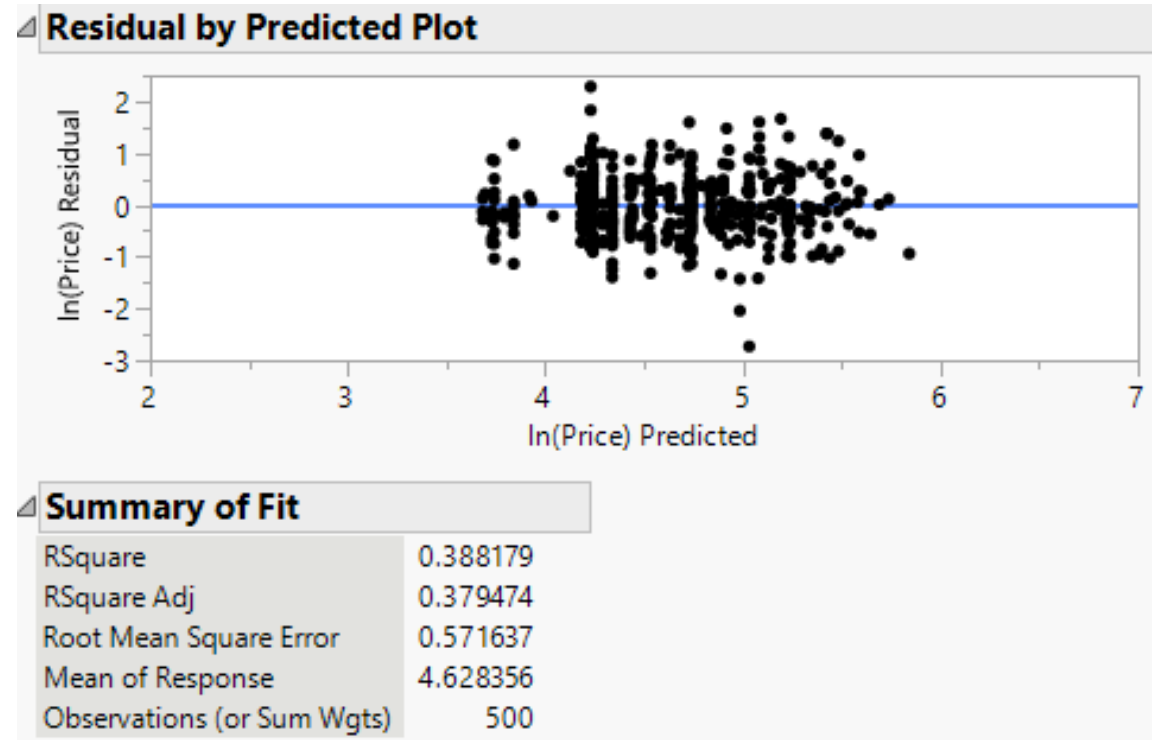
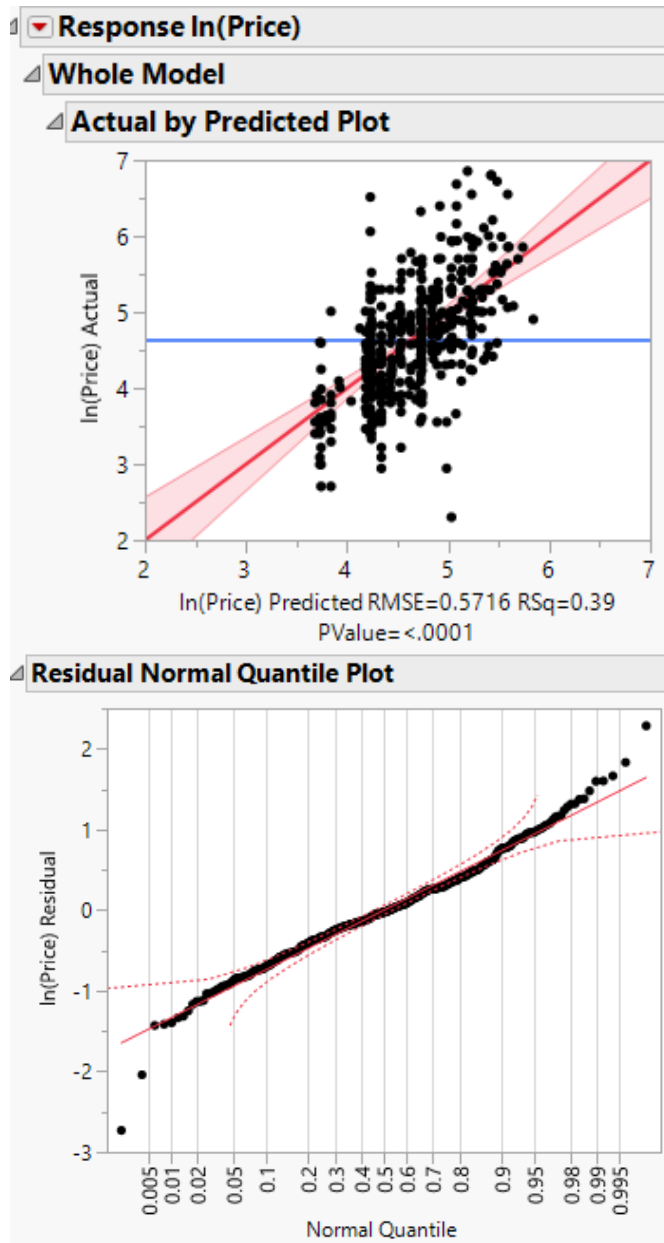
Chosen Model

- Best model: transformation model
- Highest adjusted R^2 of 0.38
- Lowest RMSE of 107.62
- Lowest Mallow's C_p of 8
- Lowest F-test value of 44.59, with very small p-value of <0.0001
- Fewest variables

Chosen Model Results

- predicted $\ln(\text{price}) = 3.695 + 0.047 * \text{Apartment} + 0.720 * \ln(\text{Accommodates}) - 0.011 * \text{moderate} + 0.098 * \text{flexible} + 0.432 * \text{Townhouse} + 0.235 * \text{Other} + 0.247 * \text{Condominium}$
- Sample size: 500
- 37.95% of variability in $\ln(\text{price})$ can be explained by a linear model with `property_type`, $\ln(\text{accommodates})$, and `cancellation_policy` after adjusting for the complexity of the model

Graphs for Checking Conditions

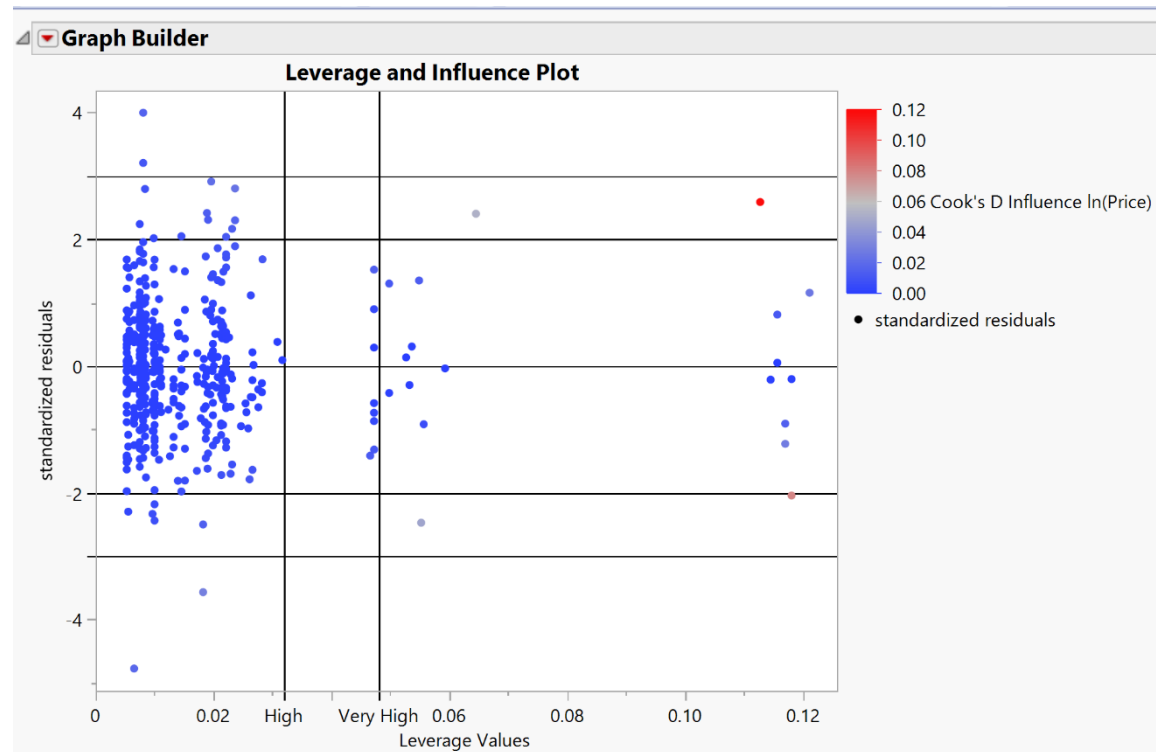


Checking Conditions

- Independence met: random sample
- Linearity met: no curve in scatterplot
- Normality met: data within recommended boundaries of normal quantile plot
- Equal Variance met: no fan in residual by predicted plot, means equally distributed above and below mean of zero

Influence and Leverage Plot

- ≈ 20 outliers indicating high leverage
- ≈ 10 outliers with standardized residual > 2
- ≈ 5 outliers with standardized residual < -2



Hypothesis Tests

H₀: $\beta_2 = \beta_3 = \beta_4 = \beta_5 = 0$ where
 β_2 = apartment slope, β_3 =
condominium slope, β_4 =
other slope, β_5 = townhouse
slope.

H_a: At least one $\beta_i \neq 0$

Test statistic = 3.2264

p-value = 0.01248

Conclusion: There is strong evidence to suggest
that the Property Type is a predictor of $\ln(\text{Price})$.

Custom Test				
Parameter				
Intercept	0	0	0	0
Apartment	1	0	0	0
Condominium	0	1	0	0
Other	0	0	1	0
Townhouse	0	0	0	1
flexible	0	0	0	0
moderate	0	0	0	0
ln(Accommodates)	0	0	0	0
=	0	0	0	0
Value	0.0472836546	0.2472841186	0.2353550284	0.4319124037
Std Error	0.0765972582	0.099694898	0.1423622248	0.2033830822
t Ratio	0.6173021821	2.4804089619	1.653212632	2.123639779
Prob> t	0.5373210066	0.0134565662	0.0989254931	0.0341986394
SS	0.1245190982	2.0104206021	0.8930952355	1.4736761195
Sum of Squares	4.2171745315			
Numerator DF	4			
F Ratio	3.2264225269			
Prob > F	0.0124822497			

H0: $\beta_6 = \beta_7 = 0$ where $\beta_6 =$ flexible cancellation policy slope, $\beta_7 =$ moderate cancellation policy slope

H_a: At least one $\beta_i \neq 0$

Test statistic = 1.405

p-value = 0.246

Conclusion: There is little to no evidence to suggest that the Cancellation Policy is a predictor of $\ln(\text{Price})$.

Custom Test		
Parameter		
Intercept	0	0
Apartment	0	0
Condominium	0	0
Other	0	0
Townhouse	0	0
flexible	1	0
moderate	0	1
$\ln(\text{Accommodates})$	0	0
=	0	0
Value	0.0978346746	-0.011431748
Std Error	0.0676016746	0.0612710773
t Ratio	1.4472226484	-0.186576583
Prob> t	0.1484714173	0.8520695348
SS	0.6844016428	0.0113750839
Sum of Squares	0.9185098159	
Numerator DF	2	
F Ratio	1.4054437344	
Prob > F	0.2462410906	

H0: $\beta_8=0$ where β_8 = slope for $\ln(\text{Accommodates})$

H_a: $\beta_8 \neq 0$

Test statistic = 278.94

p-value = 6.18e-50

Conclusion: There is overwhelming evidence to suggest that the $\ln(\text{Accommodates})$ is a predictor of $\ln(\text{Price})$.

Custom Test	
Parameter	
Intercept	0
Apartment	0
Condominium	0
Other	0
Townhouse	0
flexible	0
moderate	0
$\ln(\text{Accommodates})$	1
=	0
Value	0.7200794257
Std Error	0.0431143265
t Ratio	16.701627616
Prob> t	6.175901e-50
SS	91.150264891
Sum of Squares	91.150264891
Numerator DF	1
F Ratio	278.94436503
Prob > F	6.175901e-50

Overall Conclusions

The number of accommodates that an Airbnb can support, as well as the type of property of the Airbnb influence the price of the Airbnb.

Cancellation price does not influence the price of the Airbnb.