

Applied Data Science Capstone

Modeling Foursquare Restaurant Ratings
Paul Anderson, 2020

Modeling Foursquare Restaurant Ratings

- ▶ Some people open a business like a restaurant based on a hunch, feelings or convenience.
- ▶ Having a model that can predict to some extent the rating a restaurant based on neighborhood and type of restaurant, can reduce some of the guesswork.

Data

1. Average commercial rental costs in Toronto per neighborhood (Web scrapping).
2. Population description of each neighborhood (Census).
3. Restaurant ratings and traits (Foursquare API).
4. Geographic data of Toronto's neighborhoods (Toronto.org).

Data Analysis, input data.

	nbh_num	Neighbourhood	Population 2016	Population density per square kilometre	Children (0-14 years)	Youth (15-24 years)	Working Age (25-54 years)	Pre-retirement (55-64 years)	Seniors (65+ years)	Older Seniors (85+ years)	After-tax income: Average amount (\$)
0	129	Agincourt North	29113	3929	3840	3705	11305	4230	6045	925	26955
1	128	Agincourt South-Malvern West	23757	3034	3075	3360	9965	3265	4105	555	27928
2	20	Alderwood	12054	2435	1760	1235	5220	1825	2015	320	39159
3	95	Annex	30526	10863	2360	3750	15040	3480	5910	1040	80138
4	42	Banbury-Don Mills	27695	2775	3605	2730	10810	3555	6975	1640	51874

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	nbh_num	CAD / ft ² ·month
0	1	4.015792
1	2	2.916667
2	3	1.333333
3	17	2.652806
4	19	0.389306

Data Analysis, correlations.

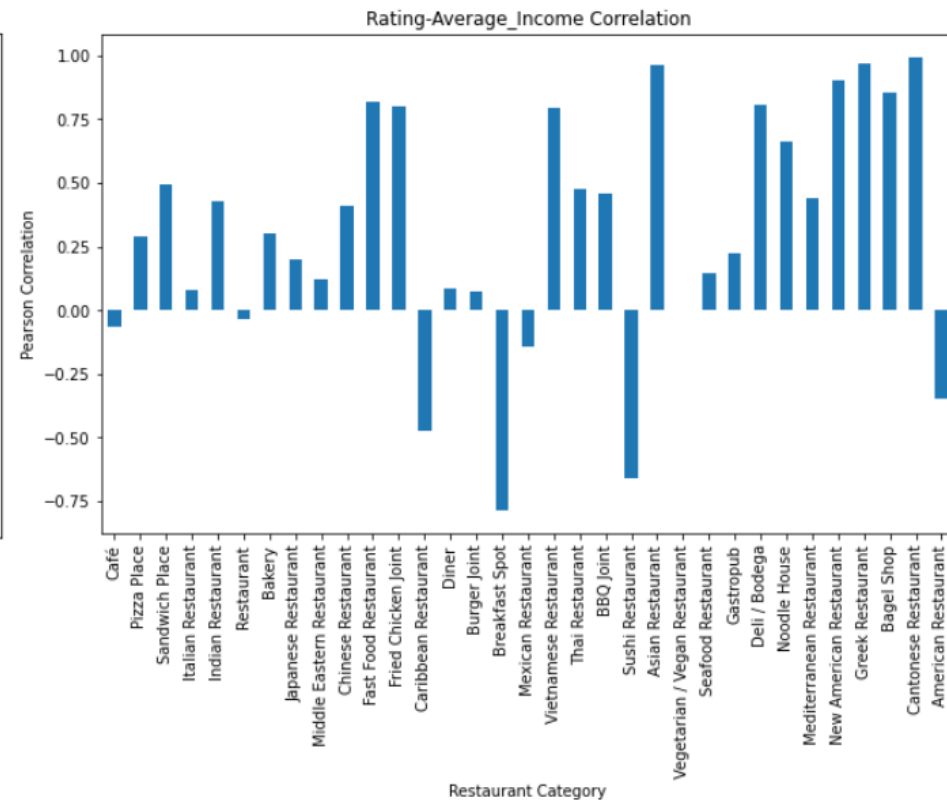
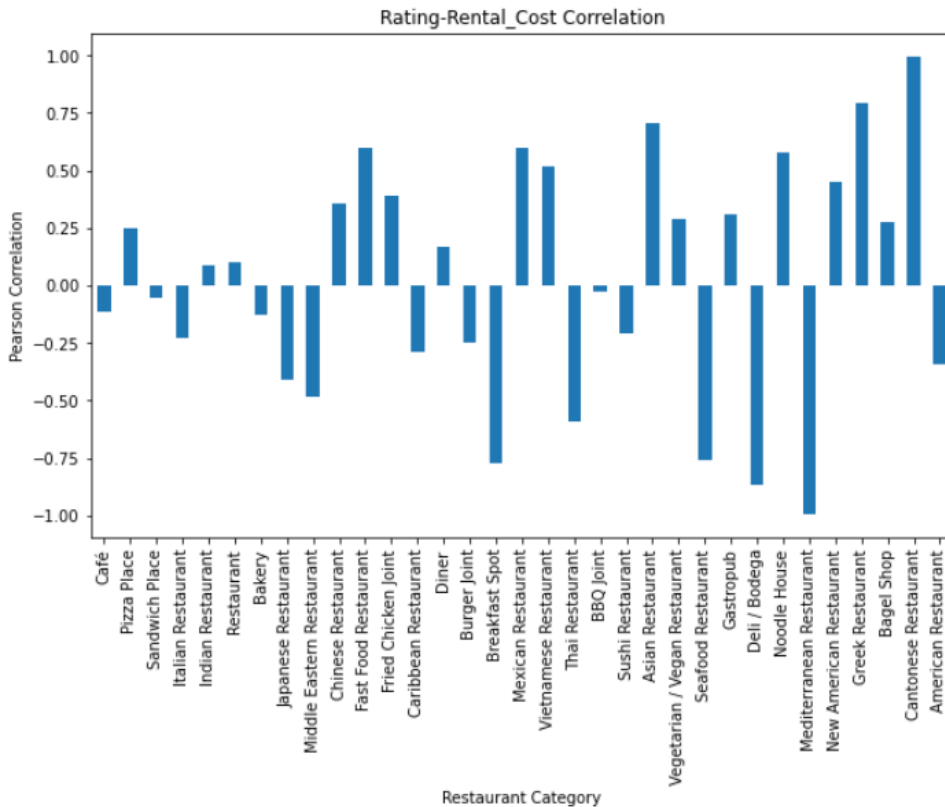
Rating correlations.

	Café	Pizza Place	Sandwich Place	Italian Restaurant	Indian Restaurant	Restaurant	Bakery	Japanese Restaurant	Middle Eastern Restaurant	Chinese Restaurant	Seafood Restaurant	Gastropub	Deli / Bodega	Noodle House	Mediterranean Restaurant	New American Restaurant	Greek Restaurant	Bagel Shop	Cantonese Restaurant	American Restaurant	
nbh_num	-0.033529	0.023723	-0.041736	0.268448	0.158477	0.043137	0.087970	0.343976	0.420890	-0.248472	...	-0.899804	0.966594	0.471060	-0.897345	0.931728	0.453921	0.972015	0.755929	-0.675983	0.344865
price_tier	0.003719	0.472979	NaN	-0.085246	0.202601	-0.040845	NaN	NaN	0.070747	0.625611	...	-0.402919	NaN	0.815368	NaN	NaN	-0.500000	NaN	0.628619	NaN	NaN
likes	0.430361	0.655373	0.652924	0.397647	0.727943	0.411355	0.533170	0.524699	0.530690	0.523081	...	0.040201	0.847961	0.850782	0.592675	0.844786	0.995423	0.995871	-0.853206	0.456584	0.781465
rating	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	...	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
CAD / ft²-month	-0.116392	0.246732	-0.054965	-0.230709	0.087005	0.101658	-0.126026	-0.406624	-0.482228	0.355192	...	-0.760830	0.307382	-0.866550	0.580449	-0.996143	0.452024	0.794359	0.276815	0.997076	-0.344865
Population 2016	-0.062478	0.149025	-0.042502	-0.356568	0.184518	-0.345600	0.211299	0.077336	0.134849	0.520493	...	-0.594236	0.140241	0.245071	-0.991775	-0.514450	-0.987265	-0.128106	-0.429163	0.672405	0.344865
Population density per square kilometre	0.015857	0.202004	0.706586	0.159904	0.448504	0.420334	0.457593	0.136287	0.129037	0.678911	...	0.356532	0.129828	0.378599	0.881905	0.460118	0.237697	0.869546	0.431372	-0.896498	0.344865
Children (0-14 years)	-0.027632	-0.188175	-0.338998	-0.449635	0.039535	-0.431631	-0.048668	-0.278733	0.033856	-0.233115	...	-0.997086	-0.288964	0.796817	-0.998925	-0.838939	-0.719551	0.500000	-0.086862	0.663823	0.344865
Youth (15-24 years)	-0.057581	-0.005456	-0.084460	-0.460592	0.039922	-0.336857	0.070518	0.010390	0.023526	0.458647	...	0.083610	0.141867	-0.334987	0.662412	-0.845010	-0.883925	-0.457804	-0.123080	-0.188982	0.344865
Working Age (25-54 years)	-0.058428	0.268444	0.297310	-0.291472	0.253465	-0.224851	0.334098	0.177149	0.248957	0.597498	...	-0.310696	0.321660	0.248349	-0.994120	0.147677	-0.148587	0.203122	-0.859003	0.509789	0.344865
Pre-retirement (55-64 years)	-0.054237	-0.020216	-0.253718	-0.317997	0.207399	-0.418492	-0.008621	-0.091397	-0.008124	-0.050106	...	-0.936640	-0.216065	0.265658	-0.990969	-0.903124	-0.497423	-0.079383	0.180070	0.550819	0.344865
Seniors (65+ years)	-0.063601	0.012366	-0.441497	-0.303730	0.235077	-0.411986	-0.056886	-0.154053	-0.061543	-0.233609	...	-0.908249	-0.423377	0.086803	-0.933886	-0.802766	0.066831	-0.388452	-0.090061	0.995611	0.344865
Older Seniors (85+ years)	-0.096606	0.037363	-0.482222	-0.220917	0.277890	-0.259412	-0.125758	-0.290659	-0.118351	-0.374863	...	-0.862582	-0.531568	0.252852	-0.772487	-0.319505	0.325561	-0.500000	-0.500000	0.995562	0.344865
After-tax income: Average amount (\$)	-0.065753	0.287957	0.494460	0.080790	0.430825	-0.036893	0.303335	0.198728	0.124050	0.408240	...	0.143171	0.226245	0.809703	0.665605	0.437444	0.904451	0.967411	0.853159	0.995076	-0.344865

Likes correlations.

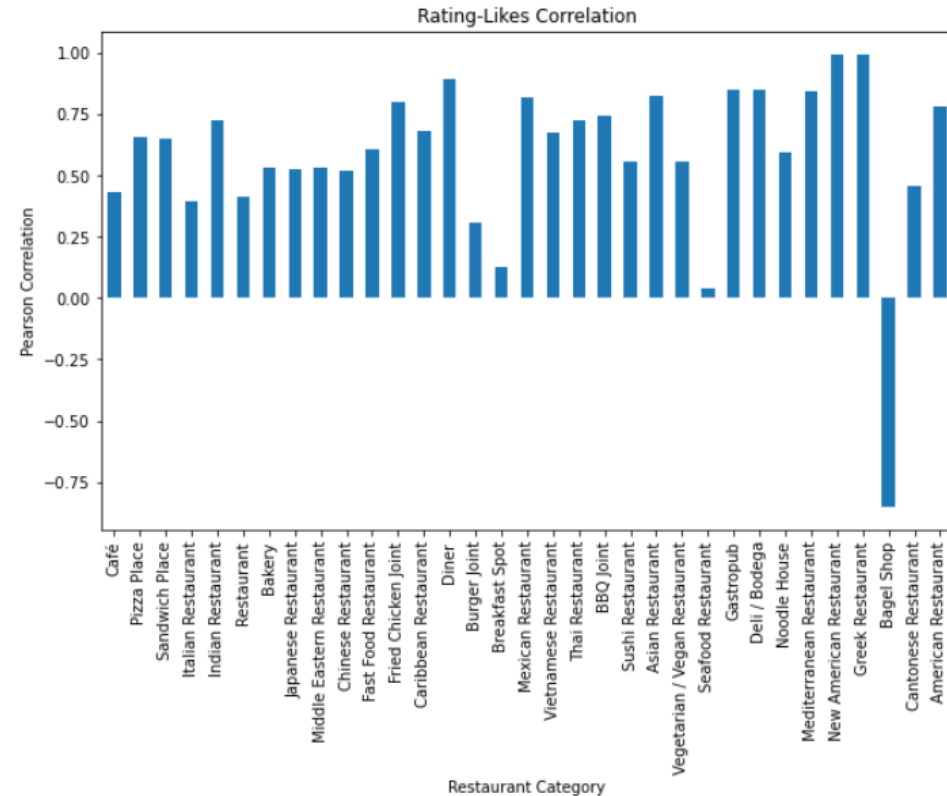
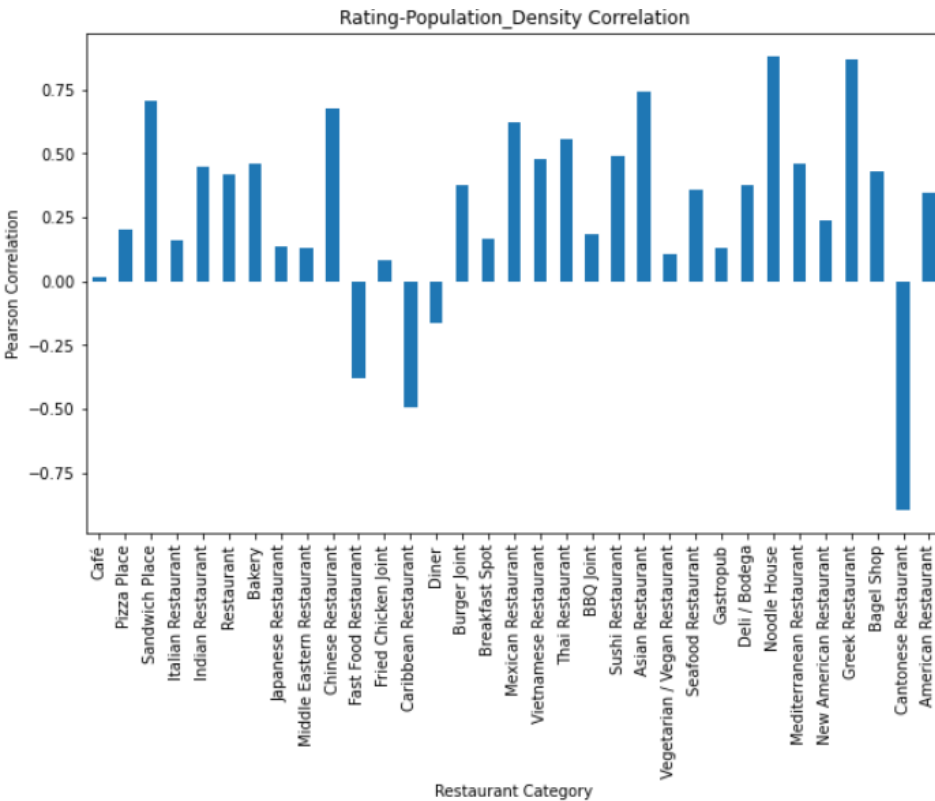
	Café	Pizza Place	Sandwich Place	Italian Restaurant	Indian Restaurant	Restaurant	Bakery	Japanese Restaurant	Middle Eastern Restaurant	Chinese Restaurant	...	Seafood Restaurant	Gastropub	Deli / Bodega	Noodle House	Mediterranean Restaurant	New American Restaurant	Greek Restaurant	Bagel Shop	Cantonese Restaurant	American Restaurant
nbh_num	-0.083987	-0.023468	-0.007340	0.225449	0.163300	0.118498	-0.016697	0.134613	0.187745	-0.375762	...	0.255416	0.832096	0.463408	-0.176369	0.592783	0.366692	0.946674	-0.986414	-0.964264	-0.316171
price_tier	0.100403	0.962765	NaN	0.343271	-0.102039	0.407008	NaN	NaN	0.086430	0.671783	...	-0.903016	NaN	0.995423	NaN	NaN	NaN	-0.576557	NaN	0.978934	NaN
likes	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	...	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
rating	0.430361	0.655373	0.652924	0.397647	0.727943	0.411355	0.533170	0.524699	0.530690	0.523081	...	0.040201	0.847961	0.850782	0.592675	0.844786	0.995423	0.995871	-0.853206	0.456584	0.781465
CAD / ft ² -month	0.090691	0.249101	0.009250	0.259647	0.162219	0.584591	0.502024	-0.115162	-0.267120	0.290358	...	0.412670	0.713246	-0.480328	-0.311850	-0.794574	0.364713	0.735932	0.265013	0.387263	0.316171
Population 2016	0.065700	-0.101951	0.266888	-0.112212	0.041865	-0.268785	0.282809	0.295162	0.037123	0.968992	...	0.779748	0.543716	0.713333	-0.690891	0.024263	-0.997949	-0.217613	0.837265	-0.351517	-0.316171
Population density per square kilometre	0.051027	0.186455	0.808661	-0.145615	0.534307	0.308086	0.367496	0.560662	-0.294994	0.715731	...	0.904356	0.594713	0.220417	0.142976	-0.086396	0.143785	0.910789	-0.838600	-0.015156	-0.316171
Children (0-14 years)	-0.104322	-0.230802	-0.325201	-0.158522	-0.083059	-0.318144	-0.125110	-0.169989	0.053477	0.055096	...	-0.097343	-0.492840	0.944406	-0.629382	-0.417506	-0.648995	0.419314	0.593715	-0.362291	-0.316171
Youth (15-24 years)	0.126912	-0.183672	0.317746	-0.076480	-0.081101	-0.246375	0.044844	0.345864	-0.246614	0.936536	...	0.990895	0.624904	0.165610	0.995983	-0.427705	-0.835191	-0.536625	0.622622	-0.959935	-0.316171
Working Age (25-54 years)	0.106855	-0.018387	0.564693	-0.193161	0.112672	-0.186409	0.524727	0.338403	0.187108	0.977930	...	0.923633	0.713625	0.719163	-0.676410	0.653993	-0.242410	0.113391	0.999937	-0.532630	-0.316171
Pre-retirement (55-64 years)	-0.082692	-0.199357	0.069358	-0.021605	0.068343	-0.337546	-0.111878	0.273673	-0.011846	0.583433	...	0.097535	-0.236167	0.730962	-0.479317	-0.533183	-0.412244	-0.169553	0.359412	-0.491054	-0.316171
Seniors (65+ years)	-0.044109	-0.137053	-0.119188	0.144930	0.074663	-0.316212	-0.286862	0.146151	-0.101300	0.076459	...	0.212374	-0.109119	0.577326	-0.841493	-0.359085	0.161875	-0.470503	0.596296	0.371313	-0.316171
Older Seniors (85+ years)	-0.018278	-0.090775	-0.255248	0.169625	0.115762	-0.289608	-0.391672	-0.135393	-0.081848	-0.441037	...	0.226561	-0.275052	0.698396	-0.969314	0.237144	0.414428	-0.576557	0.878300	0.370830	-0.316171
After-tax income: Average amount (\$)	-0.156672	0.028463	0.099985	0.450070	0.239770	0.108127	0.365303	0.039746	-0.157296	0.037121	...	-0.324331	0.434923	0.997265	-0.206617	-0.111644	0.859547	0.986404	-0.455840	0.542515	0.316171

Data results



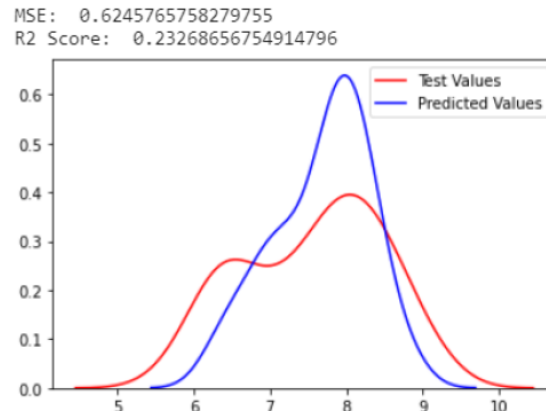
Restaurant categories are more affected by some neighborhood features than others.

Data results

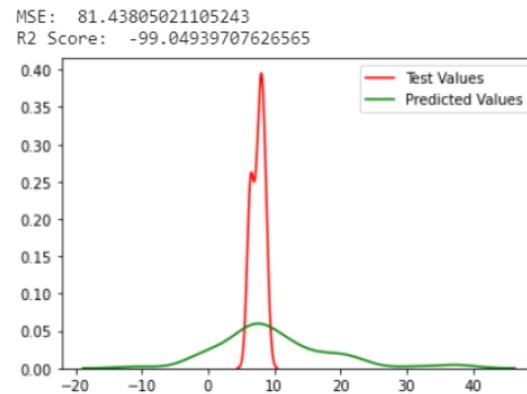


Amount of likes seems to be the best predictor for amount of likes. Likes won't be taken into account, because these come after the restaurant exists, not before.

Regression Models

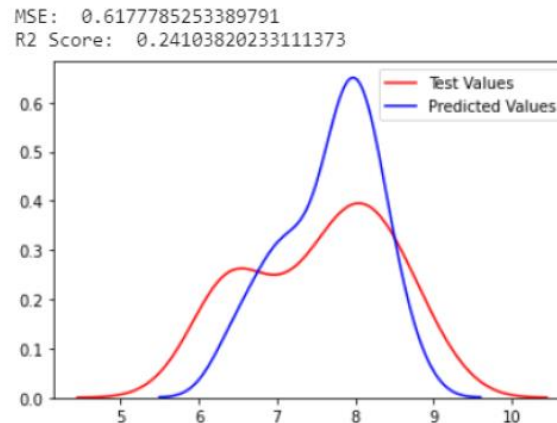


Model 1. Multiple linear regression, using label encoding for the neighborhood name.

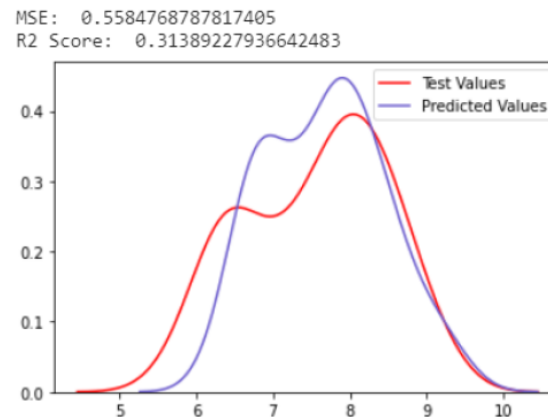


Model 2. Polynomial regression, using label encoding for the neighborhood name and a second order degree polynomial feature on the population data frame.

Regression Models

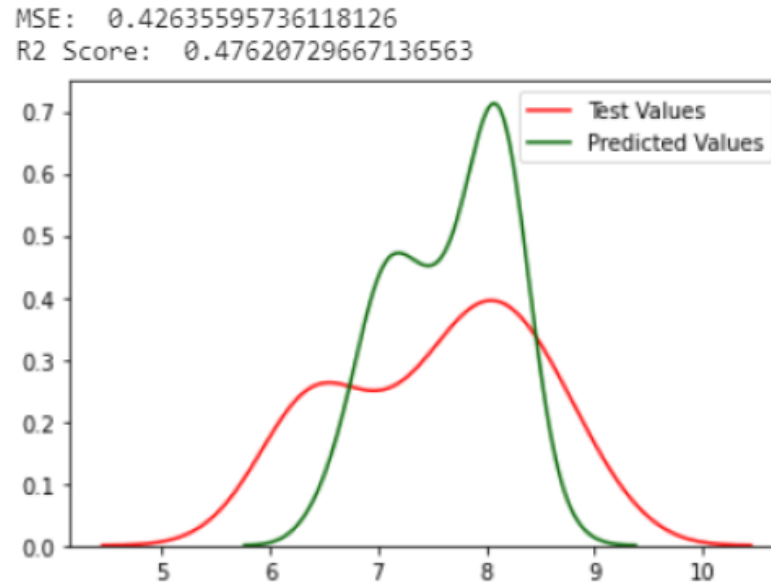


Model 3. Ridge regression, $\alpha = 0.1$. Label encoding for neighborhoods.



Model 4. Ridge regression, $\alpha = 0.1$. One-hot encoding for neighborhoods.

Regression Models



Model 5. Ridge regression after using grid search, $\alpha = 1.5$. One-hot encoding for neighborhoods.

Conclusions

- ▶ Getting enough data was the main factor limiting the quality of the model.
- ▶ Proper handling of the categorical data improves model quality.
- ▶ Based on this sample of data, the safest option is a Café.