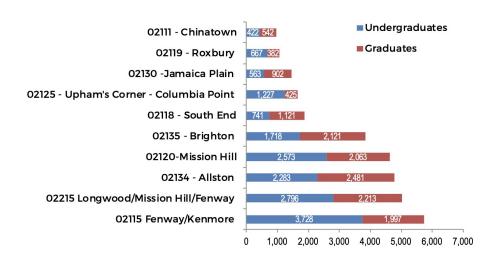
Predicting Housing Prices

Sam Delfino, Katherine Chapkis, Megan Chin, Lichen Yang

Background

- Housing is expensive in Boston, especially for students
- Boston ranked 25th most expensive city by US News
- 250,000 students attend college in Boston
- Almost 37,000 students are living in private housing in Boston
- Factors that affect price: location, size of property, accessibility, and transportation



Our Target Student

- Northeastern student
- Three bed apartment (with 2 roomates)
- Budget of \$1,727 per month/student = \$5,181 per month



Problems to address

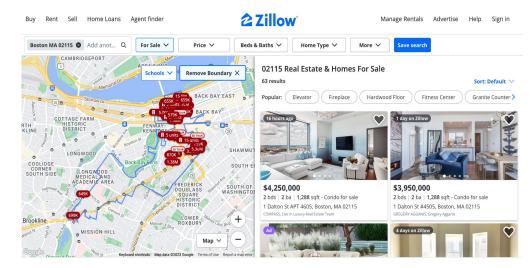
- What is the most accurate average price of a home in various Boston zip codes based on property features?
- What factors affect the price of a home?
- Are there areas where the price is overly inflated, and why?
- How many current Zillow listings are suitable for our target student, and how many of them would be in their price range?

Related works

- Bias-variance tradeoff in machine learning: Theoretical formulation and implications to structural engineering applications(X. Q. Guan and H. Burton)
- Analysis and Prediction of Real Estate Prices: A Case of the Boston Housing Market (Muralidharan et al., 2018).
 - o Decision tree and Neural network for property price prediction
- Housing Price Prediction via Improved Machine Learning Techniques (Truong et al., 2020)
 - Random Forest, Extreme Gradient Boosting, Light Gradient Boosting Machine, Hybrid Regression, and Stacked Generalization for price prediction.

Data

- Massachusetts Government property data (2020-2022)
 - Zip code, value of property, area of property
- FRED (Federal Reserve Economic Data)
 - o Unemployment rate in Boston
 - Housing Price Index (HPI) in Boston
- Zillow
 - House listings
 - Prices



Methods + Results

Methods

REGRESSION

- Data from Massachusetts
 Government and FRED
- 8 zip codes closest to Northeastern
- Goal is to find which zip code is the cheapest and which features are the best for our model

CLASSIFICATION

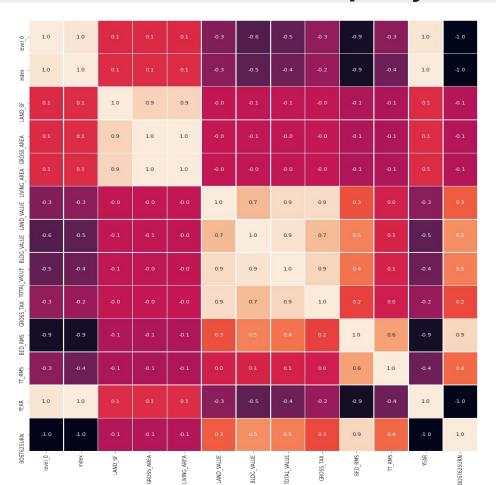
- Data from Zillow
- 4 zip codes closest to Northeastern
- O for not affordable and 1 for affordable
- Purpose is to find how many properties satisfy our budget

Initial Zip Code Analysis and Fitting

• Purpose: to find which of the 8 nearest zip code in Boston has the cheapest housing prices

	02108	02111	<mark>02115</mark>	02116	02120	02125	02127	02215
Best Method	Ridge	Lasso	Ridge	Ridge	OLS	OLS	Ridge	OLS
Average Total Housing Value (\$)	1,113,002	1,278,415	<mark>526,719</mark>	1,197,922	1,173,301	1,091,188	2,034,498	1,498,356

Feature Selection: Property Data



- Factors that affect total value the most are:
 - GROSS_TAX

- 0.75

- 0.25

0.00

-0.25

-0.50

- -0.75

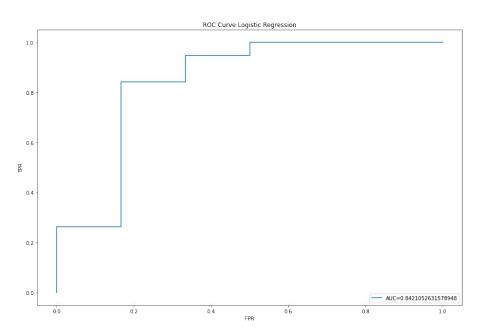
- LAND_VALUE
- BLDG_VALUE
- Regressions all had >99% accuracy
- Model fit with Linear Regression:
 - 0.9998997524032356
- Model fit with Ridge Regression:
 - o 0.9998997524032361
- Model fit with Lasso Regression (with CV):
 - 0.9999250309716614

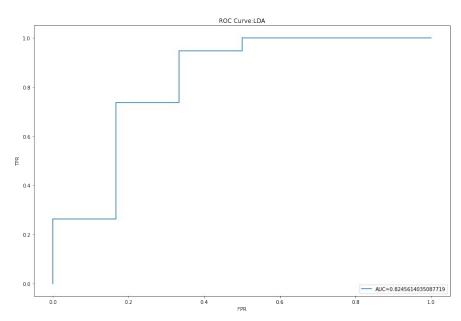
EDA: Unemployment Rates with HPI

- Tried to see if there was a correlation between unemployment rate and housing price index
- Didn't prove to be helpful
 - R^2 values were all low for linear, lasso, and ridge regressions
- Could've skewed our regressions if we used it

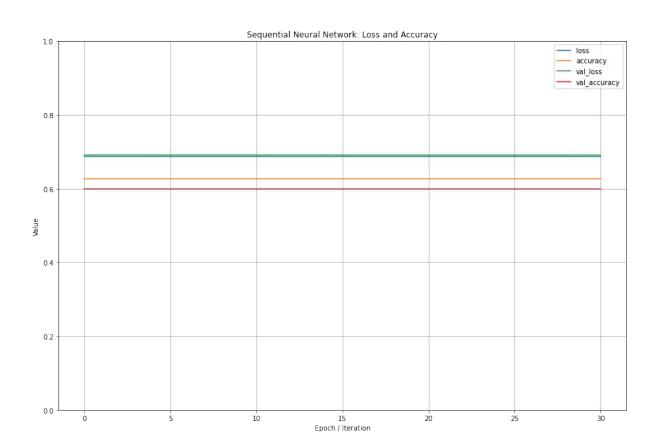
Logistic Regression and LDA

Logistic Regression and LDA: Accuracy Score: 0.95 AUC: 0.842





Deep Neural Network



- Accuracy score: 0.6
- Validation Accuracy: 0.6267

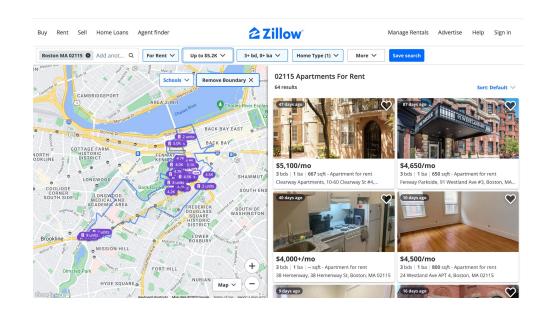
Conclusion

Best Model: Linear Discriminant Analysis & Logistic Regression

Accuracy Score: 0.95

• Area Under the Curve: 0.842

Most Affordable Neighborhood: 02115



Next Steps...

- Inclusion of more analysis fields
 - Crime rate
 - Proximity to campus
 - Other tangible factors that might affect a renter's decision
- Applying model to different/more zip codes to continually test the models

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- Jafari, Amirhosein, and Reza Akhavian. "Driving Forces for the US Residential Housing Price: A Predictive Analysis." Built Environment Project and Asset Management 9.4 (2019): 515-29. ProQuest. Web. 5 Feb. 2023.
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- Muralidharan, Sharmila, et al. "Analysis and Prediction of Real Estate Prices: A Case of the Boston Housing Market." Issues In Information Systems 19.2 (2018): 109–118. Web 5 Feb. 2023.
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- Truong, Quang, et al. "Housing Price Prediction via Improved Machine Learning Techniques." Procedia Computer Science 174 (2020): 433–442. Science Direct. Web. 5 Feb. 2023.
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Q&A

Appendix

Figure A.1: Excerpt of Property Data

	index	ZIPCODE	LAND_SF	GROSS_AREA	LIVING_AREA	LAND_VALUE	BLDG_VALUE	TOTAL_VALUE	GROSS_TAX	BED_RMS	TT_RMS	YEAR	BOST625UR
0	0	2115.0	1563.0	3320.0	3058.0	908800.0	1512300.0	2421100.0	25566.82	11.0	7.0	2020	9.87
1	1	2115.0	1451.0	3332.0	3100.0	1085100.0	1591300.0	2676400.0	28262.78	8.0	4.0	2020	9.87
2	2	2115.0	1451.0	3452.0	3100.0	947400.0	2518000.0	3465400.0	36594.62	8.0	4.0	2020	9.87
3	3	2115.0	1451.0	3308.0	3076.0	1076600.0	1561400.0	2638000.0	27857.28	8.0	3.0	2020	9.87
4	4	2115.0	1648.0	3328.0	2694.0	1056100.0	1455700.0	2511800.0	26524.60	10.0	4.0	2020	9.87

Figure A.2: Excerpt of Unemployment Data

BOST625URN

75
75
25

Figure A.3: Excerpt of Building Violations Data

	status_dttm	description	violation_city	violation_zip	ward
0	2023-03-13 15:19:03	Failed to comply w permit term	East Boston	02128	01
1	2023-03-13 13:38:48	Failure to Obtain Permit	East Boston	02128	01
2	2023-03-13 11:55:12	Unsafe Structures	Dorchester	02124	14
3	2023-03-13 11:54:38	Testing & Certification	Boston	02115	05
4	2023-03-13 10:35:08	Failure to Obtain Permit	Hyde Park	02136	18

Figure A.4: Excerpt of Zillow Prices Data

```
2 6727.0
4 10500.0
5 6698.0
7 4650.0
8 4000.0
```

Name: hdpData.homeInfo.price, dtype: float64

Figure A.4: Excerpt of Zillow Listings Data

	beds	baths	hdpData.homeInfo.zpid	hdpData.homeInfo.latitude	hdpData.homeInfo.longitude	hdpData.homeInfo.bathrooms	hdpData.homeInfo.bedrooms	hdpDa
2	3.0	2.0	2.061651e+09	42.352215	-71.059060	2.0	3.0	
4	3.0	3.0	2.077718e+09	42.362114	-71.059364	3.0	3.0	
5	3.0	3.0	2.080586e+09	42.351814	-71.062454	3.0	3.0	
7	3.0	1.0	2.063494e+09	42.344230	-71.089560	1.0	3.0	
8	3.0	1.0	2.058860e+09	42.345505	-71.089195	1.0	3.0	

Appendix B: Regression Analysis Results for Correlation Between Housing Price Index and Unemployment Rate

Model fit with Linear Regression:

-0.01496812169296513

Mean Squared Error: 3028.0154716663883

Model fit with Ridge Regression:

-0.014967838653744625

Mean Squared Error: 3028.01462725845

Model fit with Lasso Regression (with CV):

-0.014784955858999638

Mean Squared Error: 3027.4690220121834

Appendix B.2.i: Metric Analysis for Top 8 Zip Codes

```
Metric Analysis for Zipcode 02108
Model fit with Linear Regression:
0.9997809920745001
Mean Squared Error: 577150827.0786208
Model fit with Ridge Regression:
0.9997812762850615
Mean Squared Error: 576401847.9712161
Model fit with Lasso Regression (with CV):
0.9994691555242071
Mean Squared Error: 1398932607.3689961
The regression method most accurate is: Ridge
Average total housing value with Ridge for zipcode 02108 (2020-2022): 1113000
 Metric Analysis for Zipcode 02111
Model fit with Linear Regression:
0.999987380752739
Mean Squared Error: 48046526,60588291
Model fit with Ridge Regression:
0.999987382116424
Mean Squared Error: 48041334.51078087
Model fit with Lasso Regression (with CV):
0.9999933509089132
Mean Squared Error: 25315751.819274098
The regression method most accurate is: Lasso
Average total housing value with Lasso for zipcode 02111 (2020-2022): 1278415
```

```
Metric Analysis for Zipcode 02115
          Model fit with Linear Regression:
          /Users/sdelfino/opt/anaconda3/lib/python3.8/site-packages/sklearn/utils/validati
          lumn-vector y was passed when a 1d array was expected. Please change the shape o
           ng ravel().
            return f(**kwargs)
           0.9990713562107881
          Mean Squared Error: 76720144.6341697
          Model fit with Ridge Regression:
          0.9990708511946422
           Mean Squared Error: 76761866.67249368
           Model fit with Lasso Regression (with CV):
           0.9988891083141916
          Mean Squared Error: 91776601.3171251
          The regression method most accurate is: OLS
          Average total housing value with OLS for zipcode 02115 (2020-2022): 526701
Metric Analysis for Zipcode 02116
Model fit with Linear Regression:
0.9999999178745604
Mean Squared Error: 24436.91387194663
Model fit with Ridge Regression:
0.9999999191186318
Mean Squared Error: 24066.7330144201
Model fit with Lasso Regression (with CV):
0.9999922891210966
Mean Squared Error: 2294417.962862132
The regression method most accurate is: Ridge
Average total housing value with Ridge for zipcode 02116 (2020-2022): 1197922
```

Appendix B.2.ii: Metric Analysis for Top 8 Zip Codes

Metric Analysis for Zipcode 02127

```
Model fit with Linear Regression:
0.9999172507593761
Mean Squared Error: 1310390449.050677
Model fit with Ridge Regression:
0.9999172507228338
Mean Squared Error: 1310391027.7245505
Model fit with Lasso Regression (with CV):
0.9999040022393149
Mean Squared Error: 1520189765.8959105
The regression method most accurate is: OLS
Average total housing value with OLS for zipcode 02120 (2020-2022): 1173301
Metric Analysis for Zipcode 02125
Model fit with Linear Regression:
0.9999776343111962
Mean Squared Error: 297499174.94799906
/Users/sdelfino/opt/anaconda3/lib/python3.8/site-packages/sklearn/linear model/ ri
itioned matrix (rcond=2.17491e-18): result may not be accurate.
  return linalg.solve(A, Xy, sym pos=True,
Model fit with Ridge Regression:
0.9999776342083495
Mean Squared Error: 297500542.97277
Model fit with Lasso Regression (with CV):
0.9999713247182036
Mean Squared Error: 381426780.5789528
The regression method most accurate is: OLS
Average total housing value with OLS for zipcode 02125 (2020-2022): 1091188
```

Metric Analysis for Zipcode 02120

```
Model fit with Linear Regression:
 0.9999958243908091
 Mean Squared Error: 24309811.236456733
 Model fit with Ridge Regression:
 0.9999958623660862
 Mean Squared Error: 24088724.49832089
 Model fit with Lasso Regression (with CV):
 0.9999909703684429
 Mean Squared Error: 52569248.85757052
 The regression method most accurate is: Ridge
 Average total housing value with Ridge for zipcode 02127 (2020-2022): 2034498
 Metric Analysis for Zipcode 02215
Model fit with Linear Regression:
0.9999256856212774
Mean Squared Error: 2500238094.2289367
Model fit with Ridge Regression:
0.9999251499501555
Mean Squared Error: 2518260250.480517
Model fit with Lasso Regression (with CV):
0.9999048530687719
Mean Squared Error: 3201129930.6344295
The regression method most accurate is: OLS
Average total housing value with OLS for zipcode 02215 (2020-2022): 1498356
```

Appendix B.3: Regression Analysis of the Three Cheapest Neighborhoods and the Three Most Significant Columns

Model fit with Linear Regression:
0.999971810315485

Mean Squared Error: 26866188.399756532

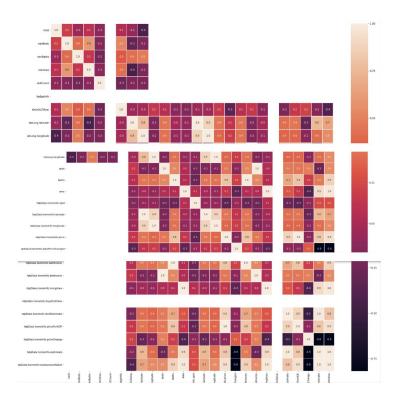
Model fit with Ridge Regression:
0.9999718103154851

Mean Squared Error: 26866188.399605617

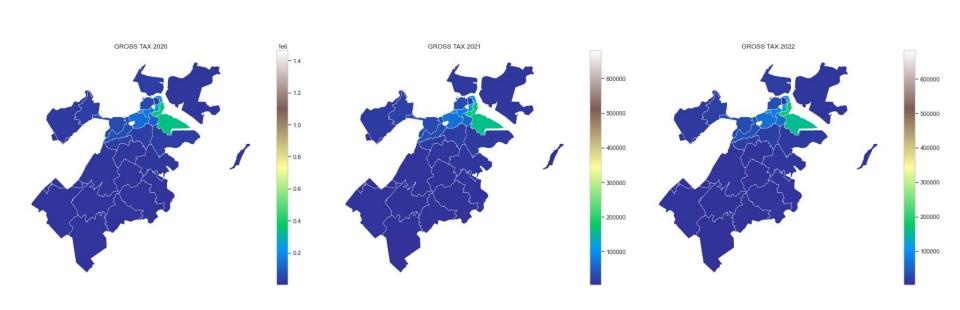
Model fit with Lasso Regression (with CV):
0.9999231844543632

Mean Squared Error: 73209081.85429057

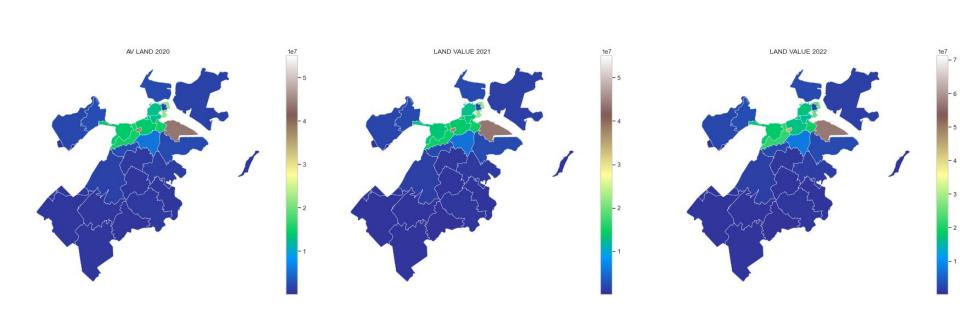
Appendix B.4:
Correlation Heatmap
Between Zillow
Property Features
and Price



Appendix B.5: Average Gross Tax Visualization



Appendix B.6: Average Land Value Visualization



Appendix B.7: Training and Testing Data Size and Logistic Regression Accuracy Scores

Shape of X_train, y_train: (57, 8) (57,) Shape of X_test, y_test: (20, 8) (20,)

Classifying listings from Zillow with Logistic Regression:

[[6 0] [1 13]]	precision	recall	f1-score	support
0 1	0.86 1.00	1.00 0.93	0.92 0.96	6 14
accuracy macro avg weighted avg	0.93 0.96	0.96 0.95	0.95 0.94 0.95	20 20 20

Appendix B.8: Linear Discriminant Analysis Accuracy Scores

Classifying listings from Zillow with Linear Discriminant Analysis:

[0 14]]]	precision	recall	f1-score	support
	0 1	1.00 0.93	0.83 1.00	0.91 0.97	6 14
accurac macro av weighted av	g	0.97 0.95	0.92 0.95	0.95 0.94 0.95	20 20 20