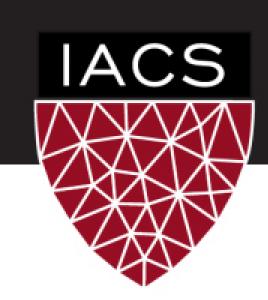
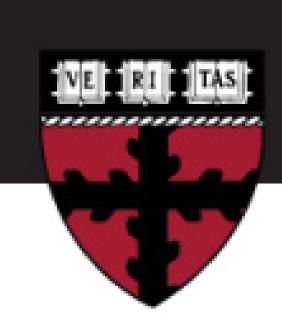
"LOCATION, LOCATION, LOCATION: CONSIDERING SPACE AND DISTANCE IN HOME PREDICTION"





WHAT PROBLEM ARE WE TRYING TO SOLVE?

PROBLEM:

ACCURATELY PREDICTING HOME SALE PRICE

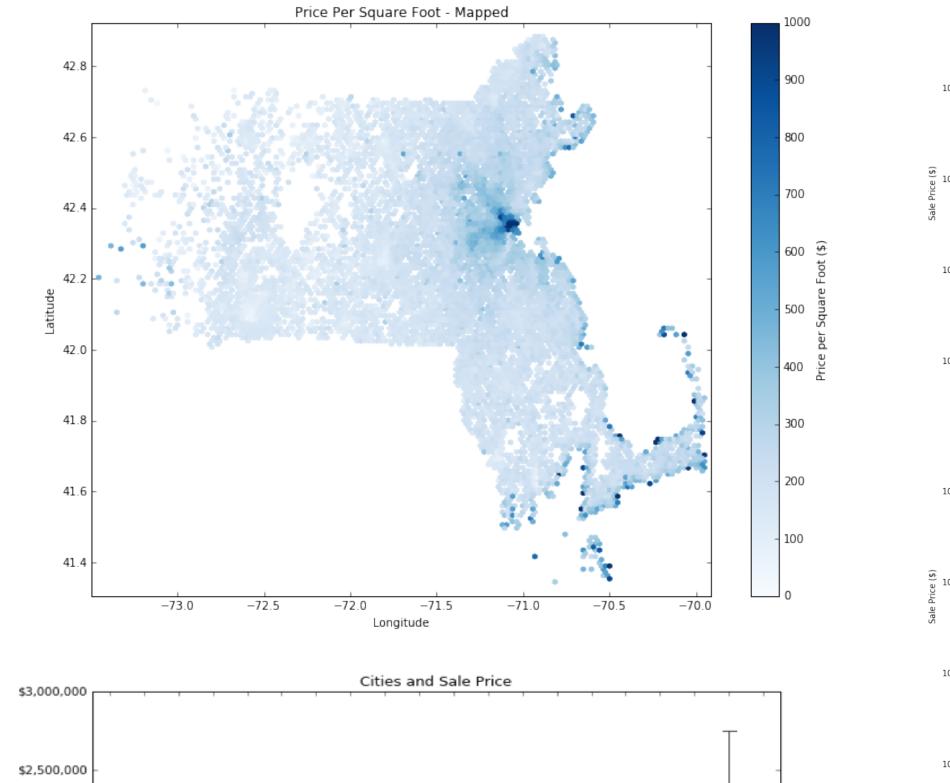
POSSIBLE SOLUTIONS:

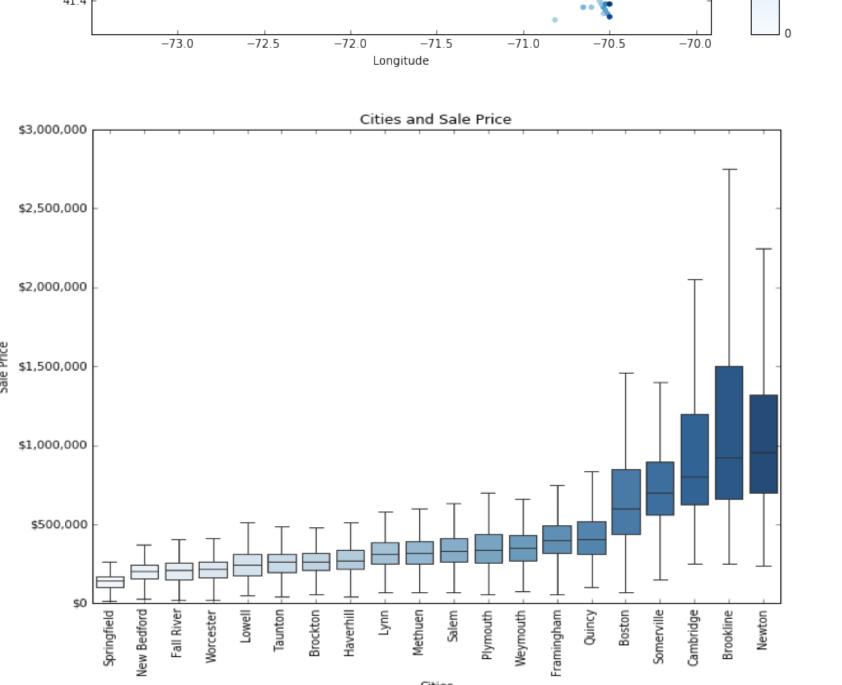
ADD NEIGHBORHOOD CHARACTERISTICS ADD UNSTRUCTURED DATA ADD ENGINEERED FEATURES **USE HEDONIC REGRESSION**

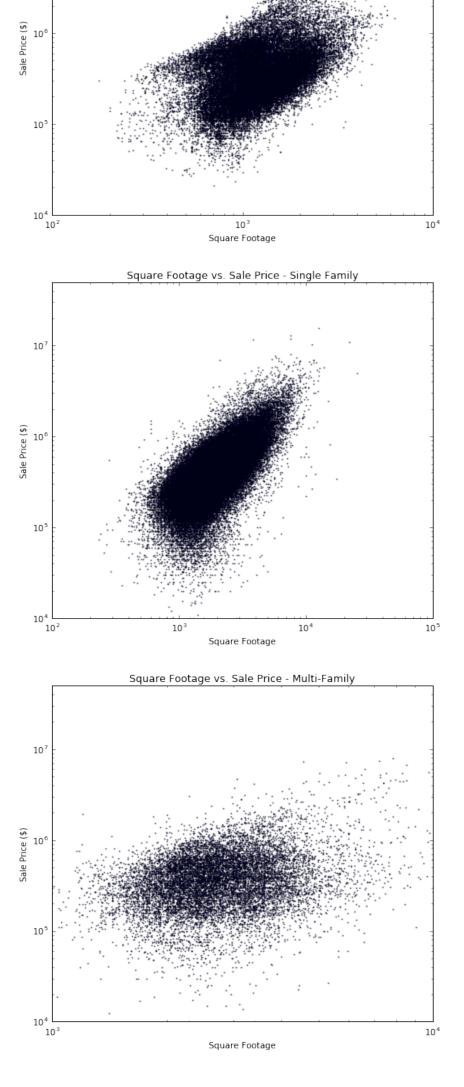
OUR APPROACH:

CONSIDER LOCATION AND DISTANCES AMONG PROPERTIES

EXPLORING OUR DATA







2 DATA PREPROCESSING

DATA COLLECTION:

GEOCODE LOCATION ArcGIS



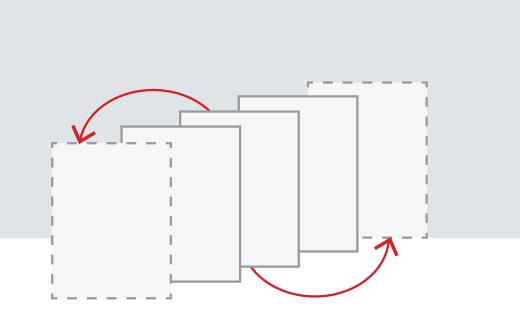
DATA CLEANING:

COMBINE DATASETS REMOVE DUPLICATES

DIVIDE INTO 2016 AND 2017, ACTIVE AND SOLD

FEATURE EXTRACTION

DUMMY VARIABLES



3 FEATURE ENGINEERING

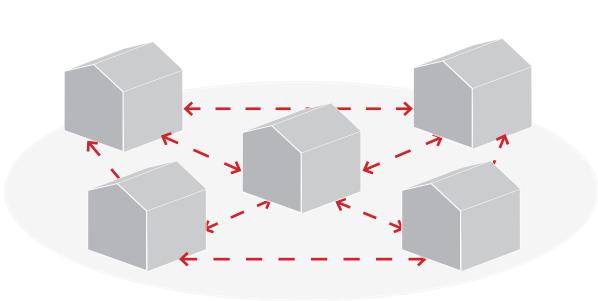
O MONTHS OF SUPPLY:

- SHARE OF ACTIVE PROPERTIES OVER PROPERTIES SOLD IN THE PREVIOUS YEAR FOR A GIVEN CATEGORY
- APPROXIMATES HOUSING MARKET SUPPLY AND DEMAND CONDITIONS
- INTUITIVE AND WIDELY USED IN PROPERTY **APPRAISAL**

#ACTIVE PROPERTIES #PROPERTIES SOLD IN PREVIOUS YEAR

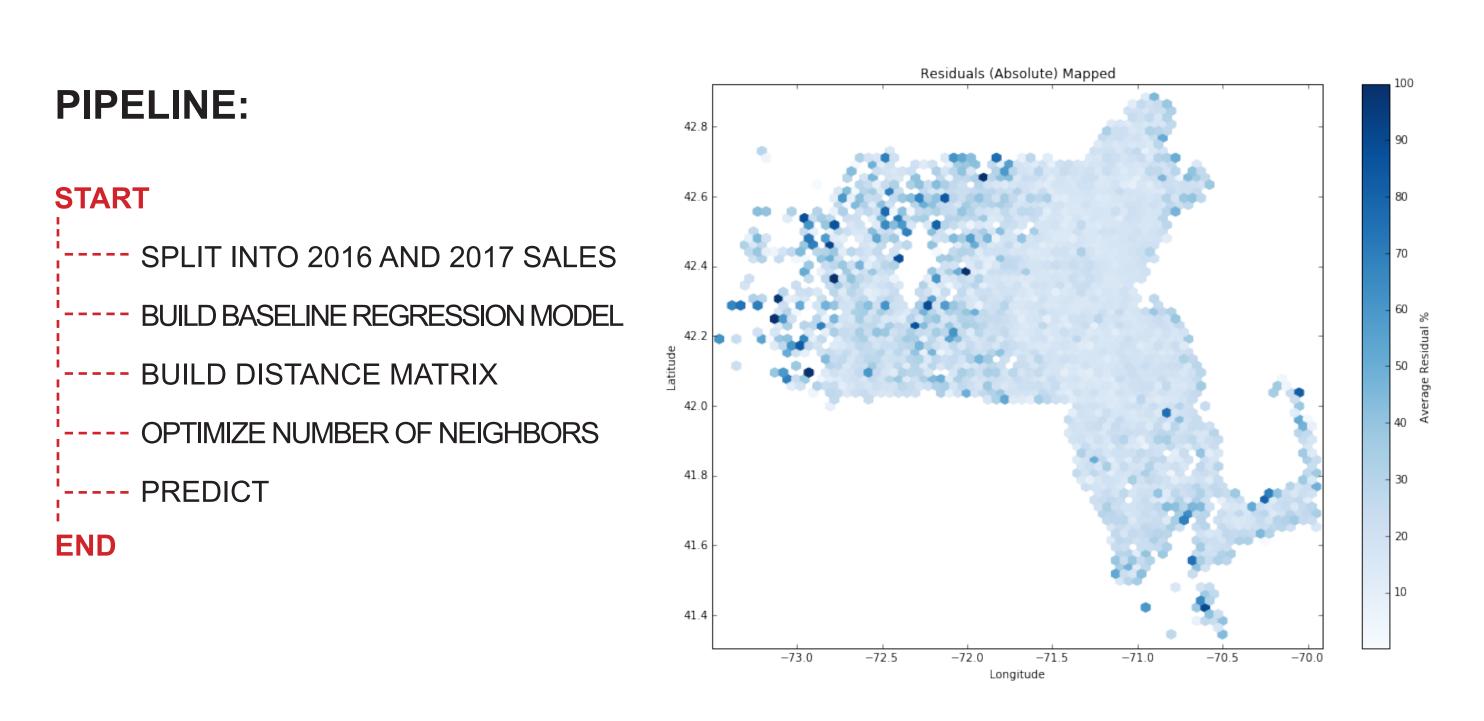
DISTANCE MATRIX:

- USE PYSAL TO COMPUTE DISTANCE AMONG ALL PROPERTIES SOLD IN 2016 AND 2017
- INTENSIVE COMPUTATION O(n²)



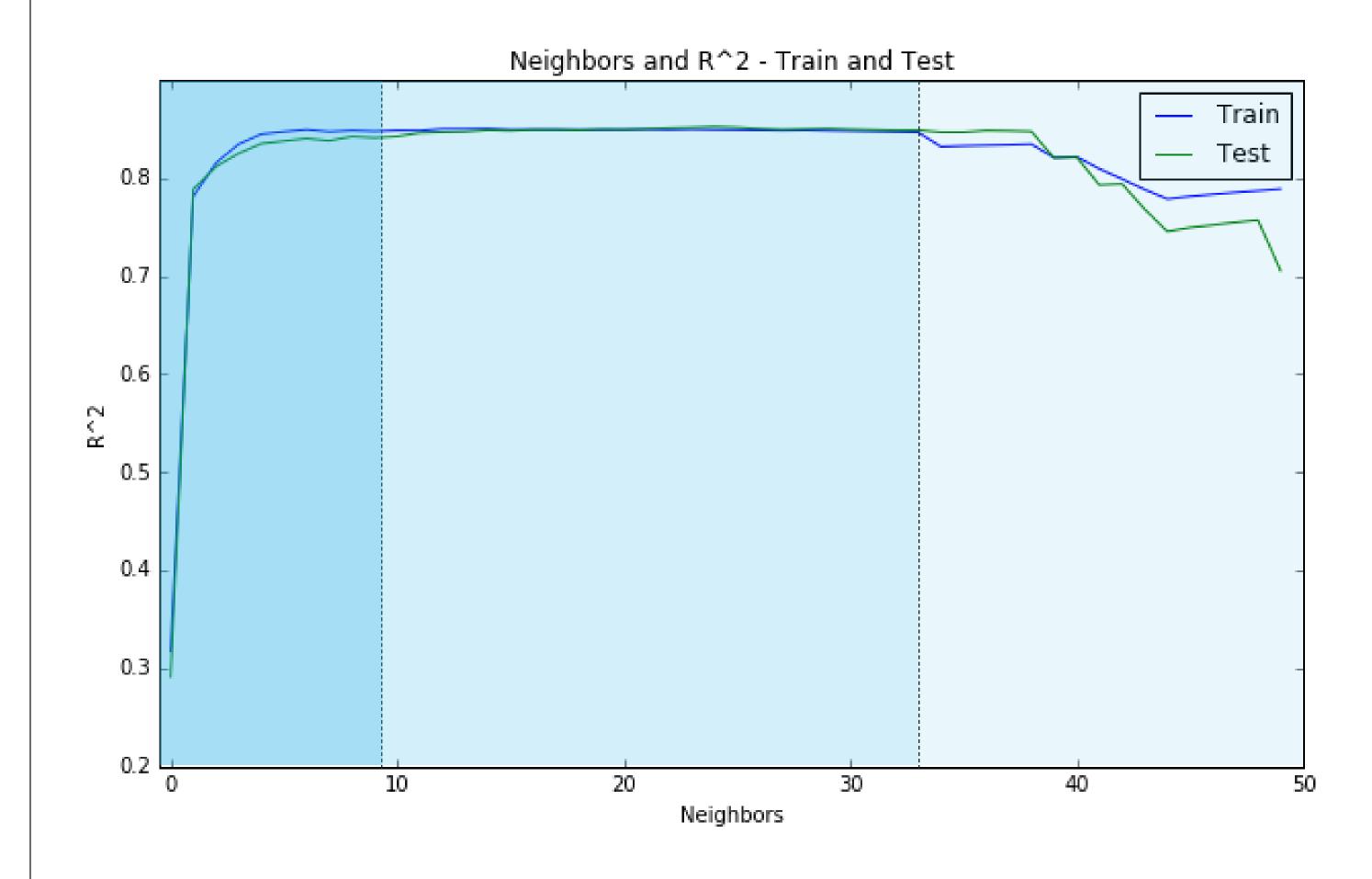
	Sample Distance Matrix - Distance between each property (Km)								
	1 Earhart St	33 Fulkerson St	31 Fulkerson St	17 Otis St	32 Sciarappa St				
1 Earhart St	0	1.043	1.035	0.251	0.542				
33 Fulkerson St	1.043	0	0.007	0.829	0.531				
31 Fulkerson St	1.035	0.007	0	0.821	0.524				
17 Otis St	0.251	0.829	0.821	0	0.389				
32 Sciarappa St	0.542	0.531	0.524	0.389	0				

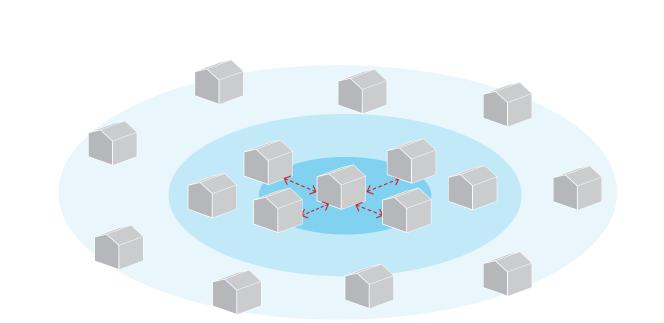
4 MODELING & ANALIZING



PRICE = $\beta_0 + \beta_1 * AGE + \beta_2 * BATHS + \beta_3 * BEDS + \beta_4 * GARAGE + \beta_5 * SQFT + \beta_6 * MOS + \beta_7 * PREDICTED_PRICES + \beta_8 * REVISED_PRICE$

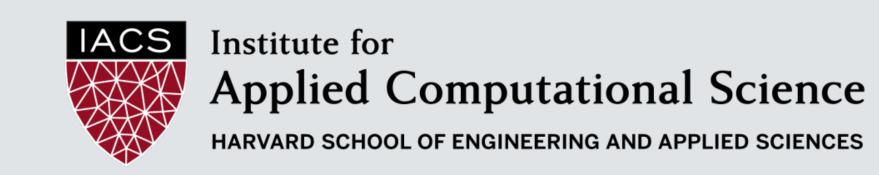
5 RESULTS





Dep. Variable:	OLS Least Squares		R-squared: Adj. R-squared: F-statistic: Prob (F-statistic):		1.381e+04 0.00	
Model:						
Method:						
Date:						
Time:		12:02:52 16655			4.461e+05	
No. Observations:						
Df Residuals:		16646	BIC:		4.461	le+05
Df Model:		8				
Covariance Type:						
	coef	std err	t	P> t	[0.025	0.9
const			3.104			
AGE	5.7261	6.382	0.897	0.370	-6.784	18.
BATHS	4.355e+04	2206.592	19.735	0.000	3.92e+04	4.79e
BEDS	2.336e+04	1283.623	18.201	0.000	2.08e+04	2.59e
GARAGE	2.353e+04	1507.415	15.611	0.000	2.06e+04	2.65e
SQFT	-114.7648	2.636	-43.541	0.000	-119.931	-109.
Category_MOS	1.512e+04	1.05e+04	1.443	0.149	-5412.258	3.57e
Predicted_Prices	1.0738	0.005	230.878	0.000	1.065	1.
Revised Price	-0.0030	0.001	-2.068	0.039	-0.006	-0.

OLS Regression Results





Basu, Sabyasachi, and Thomas G. Thibodeau. "Analysis of Spatial Autocorrelation in House Prices." The Journal of Real Estate Finance and Economics 17, no. 1 (1998): 61–85. Dubin, Robin A. "Spatial Autocorrelation and Neighborhood Quality." Regional Science and Urban Economics 22, no. 3 (1992): 433–452. Dubin, Robin, Kelley Pace, and Thomas Thibodeau. "Spatial Autoregression Techniques for Real Estate Data." Journal of Real Estate Literature 7, no. 1 (1999): 79–95.

Harring, Robert. "Spatial Data Analysis: Theory and Practice." Cambridge University Press (2003): 289–378. Legendre, Pierre. "Spatial Autocorrelation: Trouble or New Paradigm?" Ecology 74, no. 6 (1993): 1659–1673. Pace, R. Kelley, Ronald Barry, and C. F. Sirmans. "Spatial Statistics and Real Estate." The Journal of Real Estate Finance and Economics 17, no. 1 (1998): 5–13. Schernthanner, Harald, Hartmut Asche, Julia Gonschorek, and Lasse Scheele. "Spatial Modeling and Geovisualization of Rental Prices for Real Estate Portals." In International Conference on Computational Science and Its Applications, 120–133. Springer, 2016.