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https://www.linkedin.com/in/drakulic

https://github.com/drakulic/Capstone

REAL ESTATE ANALYSIS

SAN FRANCISCO

DISTRICTS 7, 8, AND 10

2013 NOVEMBER 12 – 2016 NOVEMBER 08

DISTRICTS 7, 8, AND 10

4339 DATA POINTS, 35 FEATURES

Bedroom

Bath

Parking

Sqft

home_own_ass

day on market

single_f_h

Condo

dist_no

sold_year

sold_month

Bayview Heights

Bayview

Candlestick Point

Cow Hollow

Crocker Amazon

Downtown

Excelsion

Financial District

Hunters Point

Little Hollywood

Marina

Mission Terrace

Nob Hill

North Beach

North Waterfront

Outer Mission

Pacific Heights

Portola

Presidio Heights

Russian Hill

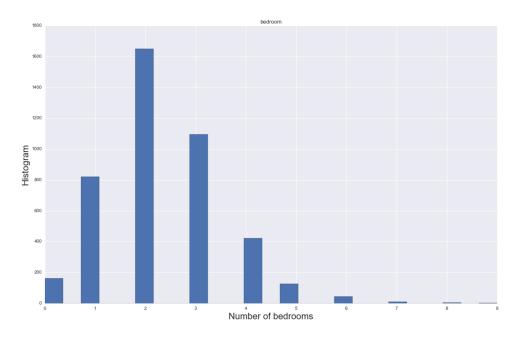
Silver Terrace

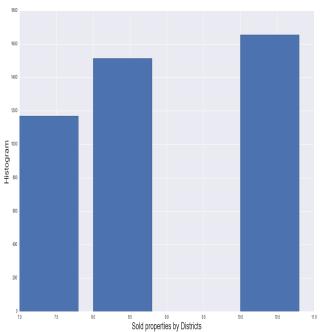
Telegraph Hill

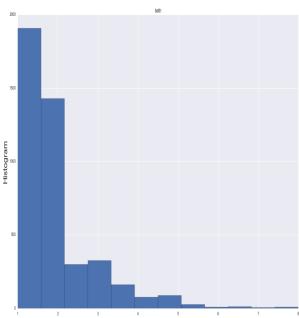
Van Ness/Civic Cente

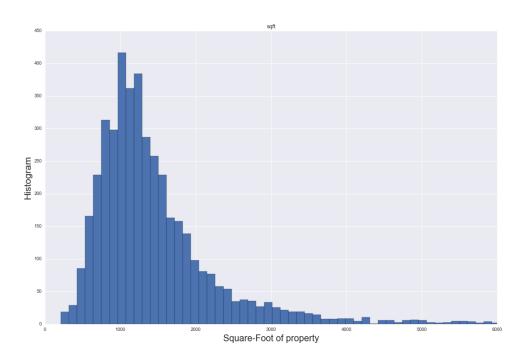
Visitacion Valley



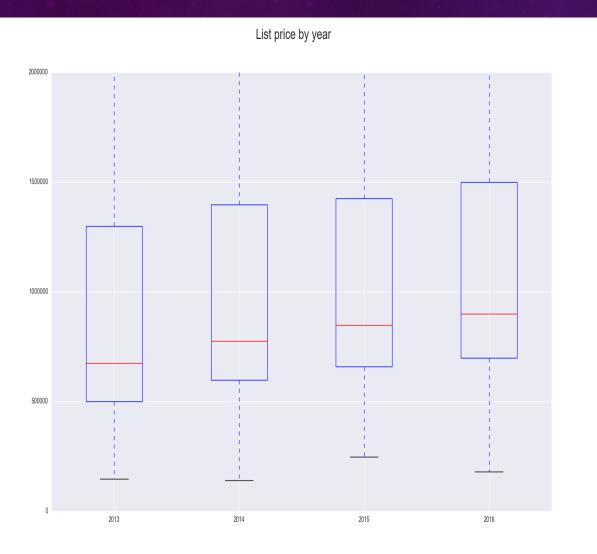


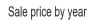


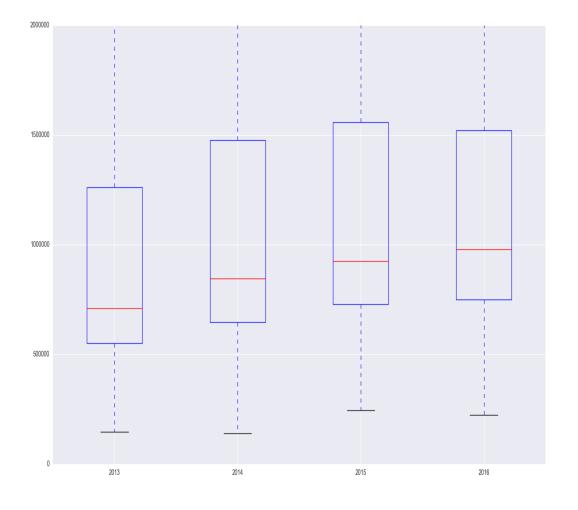




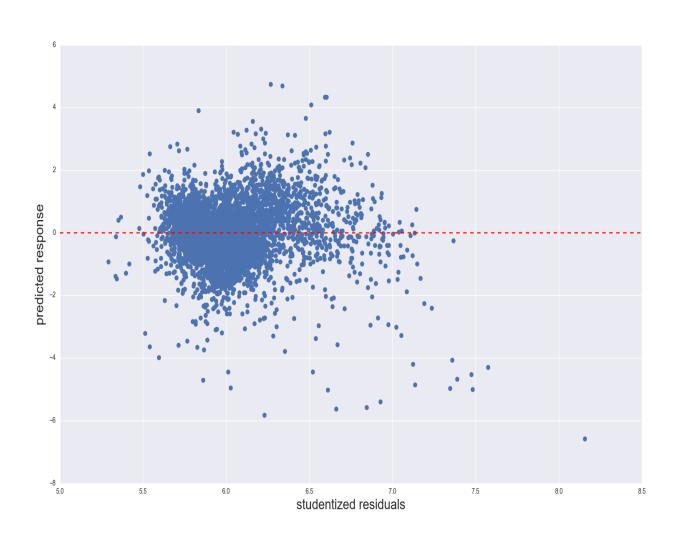
EXPLORATORY DATA ANALYSIS

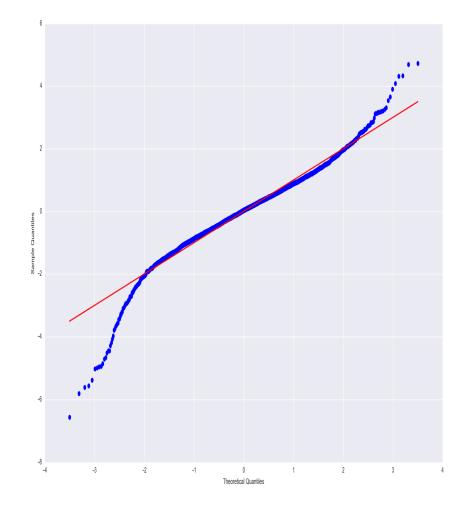






EXPLORATORY DATA ANALYSIS





LINEAR REGRESSION WITH LIST PRICE

TRAIN VARIANCE SCORE: 0.98
TEST VARIANCE SCORE: 0.96

VALIDATION VARIANCE SCORE: 0.98

model with year month columns, no log10:

OLS Regression Results

Dep. Variable: sale_price

Model: OLS

Method: Least Squares

Date: Tue, 29 Nov 2016

Time: 12:25:23

No. Observations: 4339

Df Residuals: 4305

Df Model: 33

R-squared: 0.978
Adj. R-squared: 0.978
F-statistic: 5925.
Prob (F-statistic 0.00

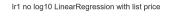
Log-Likelihood AIC: 1.200e+05

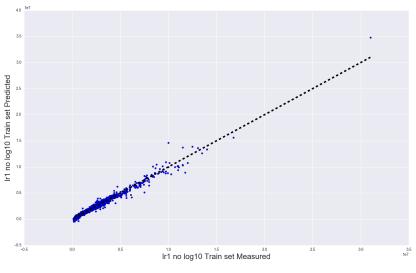
BIC: 1.202e+05



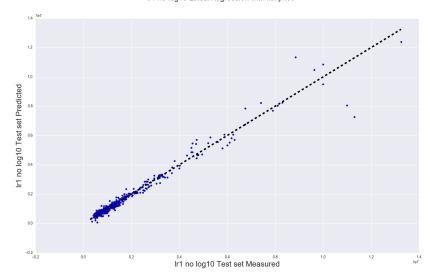
-59955.

Cross-Validated Predictions





Ir1 no log10 LinearRegression with list price



11/30/16

6

LINEAR REGRESSION WITH LIST PRICE WITH LOG10 OF PRICES

TRAIN VARIANCE SCORE: 0.98 **TEST VARIANCE SCORE: 0.96**

VALIDATION VARIANCE SCORE: 0.98

model with log10 list price:

OLS Regression Results

Dep. Variable: np.log10(sale_price) R-squared:

0.980

Model: OLS

0.980 Adj. R-squared:

Method:

F-statistic:

6302.

Tue, 29 Nov 2016

0.00

Date: Time:

12:25:23

Least Squares

Log-Likelihood

Prob (F-statistic

7569.7

No. Observations: 4339

-1.507e+04 AIC:

Df Residuals:

4305

-1.485e+04 BIC:

Df Model:

33

Cross-Validated Predictions

LINEAR REGRESSION WITHOUT LIST PRICE

model with log10 list price:

OLS Regression Results

Dep. Variable: np.log10(sale_price)

Model: OLS

Method: Least Squares

Date: Tue, 29 Nov 2016

Time: 12:25:23

No. Observations: 4339

Df Residuals: 4305

Df Model: 33

R-squared:

Adj. R-squared:

0.867

0.866

880.6

0.00

3496.9

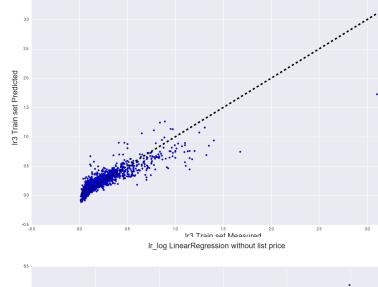
F-statistic:

Prob (F-statistic

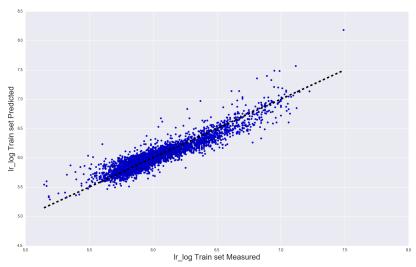
Log-Likelihood

AIC: -6928.

BIC: -6717.

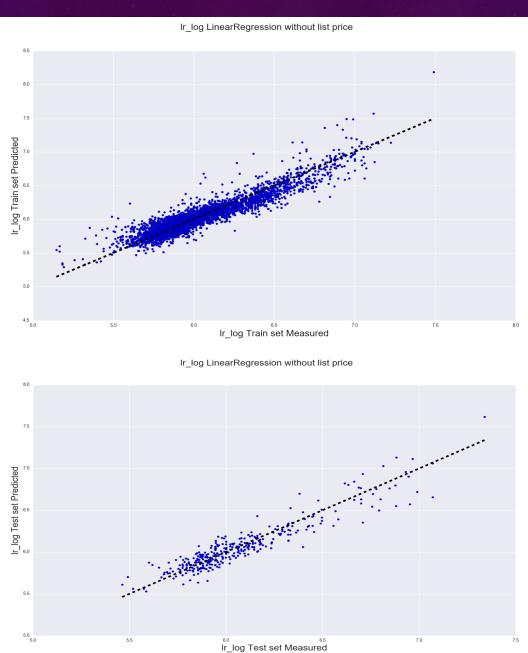


Ir3 LinearRegression without list price



Cross-Validated Predictions

LINEAR REGRESSION WITHOUT LIST PRICE, WITH LOG10 SALE PRICE



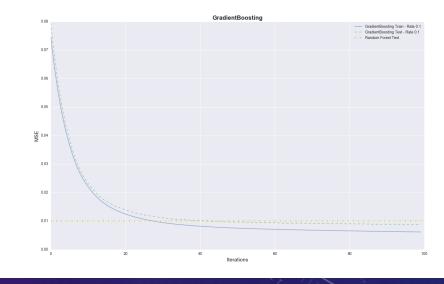
Date	Squar	e-Foot	Sale-price	Predicted price
2016-08-	-16	1050	980,000	1,113,973.86
2016-08-	-16	1312	850,000	619,429.93
2016-08-	-16	700	917,500	1,207,900.75
2016-08-	-16	760	670,000	602,939.08
2016-08-	-16	4617	9,750,000	5,256,578.14
2016-08-	-17	550	699,000	951,734.09
2016-08-	-17	527	385,000	338,296.01
2016-08-	-17	1000	775,000	639,240.18
2016-08-	-17	1210	1,050,000	824,167.77
2016-08-	-17	975	790,000	625,026.51
2016-08-	-17	1140	1,100,000	973,377.38
2016-08-	-18	1290	755,000	883,667.87
2016-08-	-18	1841	1,850,000	2,044,310.57
2016-08-	-18	1350	1,475,000	1,382,372.23
2016-08-	-18	460	760,000	820,469.61
2016-08-	-19	825	595,000	530,407.16
2016-08-	-19	1300	925,000	716,916.79

Train Variance score: 0.87 11/30/16 9

Test Variance score: 0.87

CHOOSING THE FINAL MODEL:

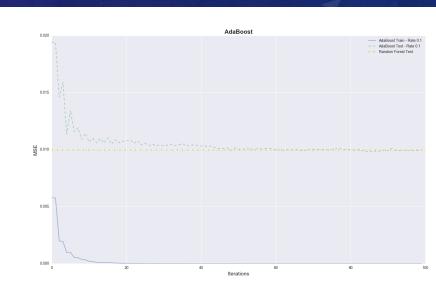
- Random Forest Regressor
- Gradient Boosting Regressor
- Ada Boosting Regressor



RandomForestRegressor Train CV | MSE: 0.008 | R2: 0.904

GradientBoostingRegressor Train CV | MSE: 0.008 | R2: 0.906

AdaBoostRegressor Train CV | MSE: 0.009 | R2: 0.901



CHOOSING THE FINAL MODEL: GRID SEARCH FOR BEST ESTIMATORS

Grid search RandomForestRegressor

Fitting 3 folds for each of 324 candidates, totalling 972

Done 52 tasks | elapsed: 2.1s

Done 352 tasks | elapsed: 14.0s

Done 657 tasks | elapsed: 28.2s

Done 972 out of 972 | elapsed: 45.5s finished

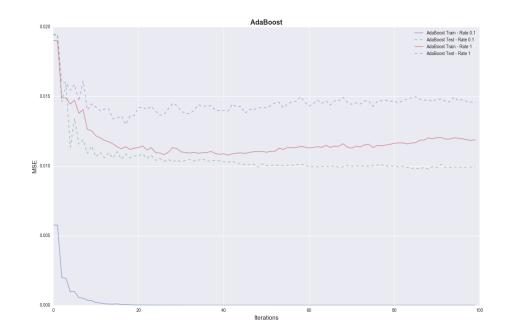
Grid search GradientBoostingRegressor

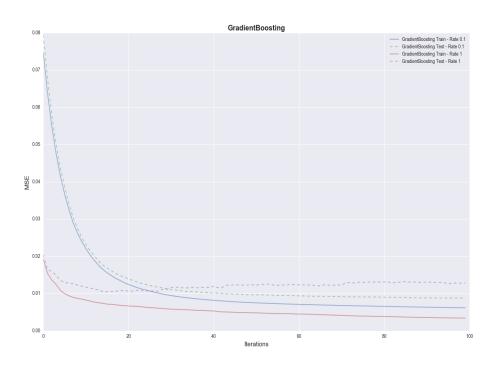
Fitting 3 folds for each of 96 candidates, totalling 288

Done 34 tasks | elapsed: 6.0s

Done 184 tasks | elapsed: 44.4s

Done 288 out of 288 | elapsed: 1.2min finished





FINAL MODEL: GRADIENT BOOSTING REGRESSOR

- GradientBoostingRegressor Train CV | MSE: 0.007 | R2: 0.919
- RandomForestRegressor Train CV | MSE: 0.007 | R2: 0.914
- Gradient Boost Test MSE: 0.00806594315088
- Gradient Boost Test R2: 0.90601555473
- Random Forest Test MSE: 0.00834159580026
- Random Forest Test R2: 0.900919571981
- Gradient Boost Test MSE: 0.00806594315088
- Gradient Boost Test R2: 0.90601555473

GradientBoostingRegressor best estimator:

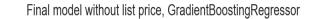
GradientBoostingRegressor(alpha=0.9, criterion='friedman_mse', init=None, learning_rate=0.02, loss='ls', max_depth=6, max_features=0.3, max_leaf_nodes=None, min_impurity_split=1e-07, min_samples_leaf=5, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=500, presort='auto', random_state=1, subsample=1/40/14 verbose=0, warm_start=False)

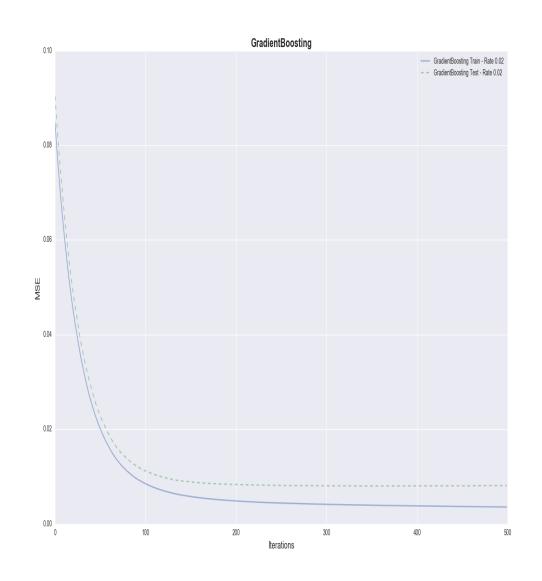
PREDICTIONS:

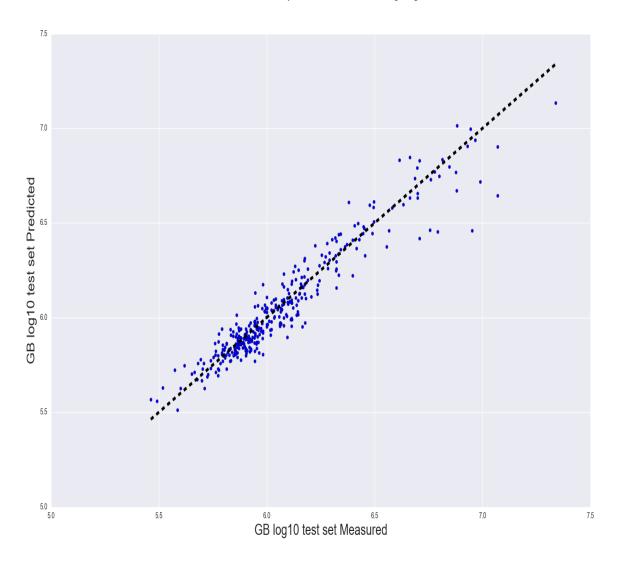
•	Date	Square-Foot	Sale-price	Predicted price
•	2016-08-16	1050	980,000	1,167,352.68
•	2016-08-16	1312	850,000	763,698.09
•	2016-08-16	700	917,500	885,745.47
•	2016-08-16	760	670,000	634,854.98
•	2016-08-16	4617	9,750,000	5,233,383.73
•	2016-08-17	550	699,000	800,931.58
•	2016-08-17	527	385,000	326,196.34
	2016-08-17	1000	775,000	691,178.71
•	2016-08-17	1210	1,050,000	950,538.13
•	2016-08-17	975	790,000	704,674.36
•	2016-08-17	1140	1,100,000	1,256,932.34
•	2016-08-18	1290	755,000	781,256.91
•	2016-08-18	1841	1,850,000	1,955,437.64
•	2016-08-18	1350	1,475,000	1,450,500.51

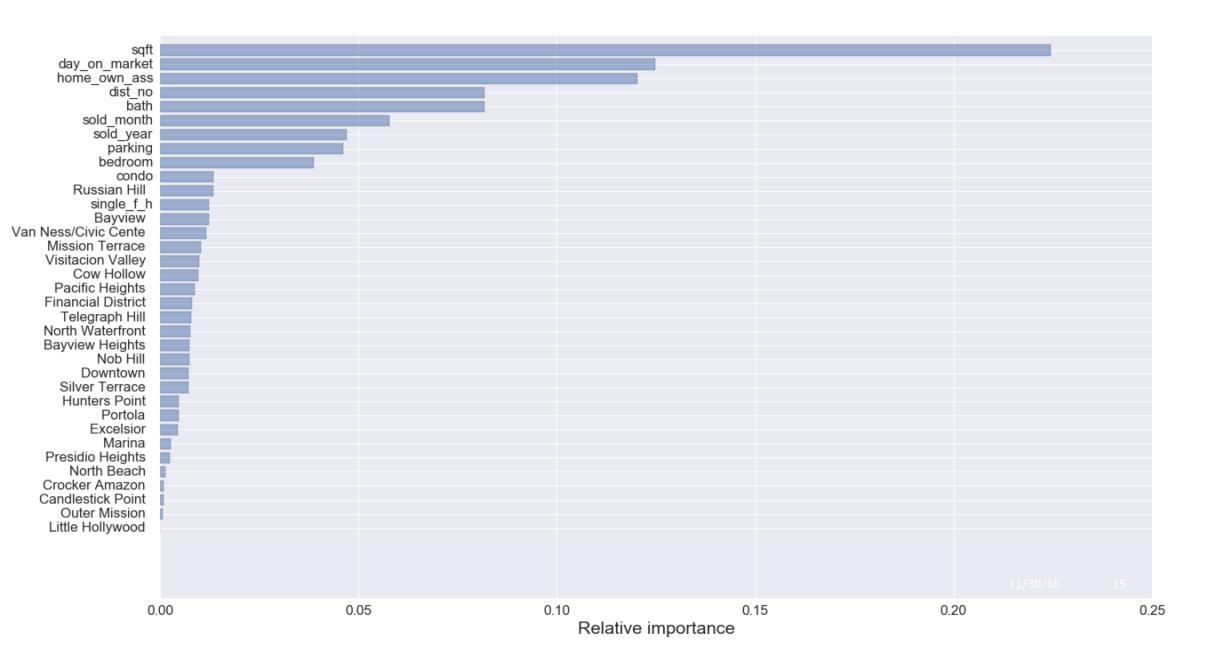


GRADIENT BOOSTING REGRESSOR









CONCLUSION AND FUTURE WORK

test set sale price avg: 1,642,855.04

test set with list price avg: 1,655,398.35

diff: -12,543.31

test set without list price avg: 1,607,460.92

diff: 35,394.12

 More data is needed: individual sales across the City, for all districts for at least 5 years. That data would allow us to analyze time series better, and the model selection through grid-search and cross-validated estimator would give us more reliable results.