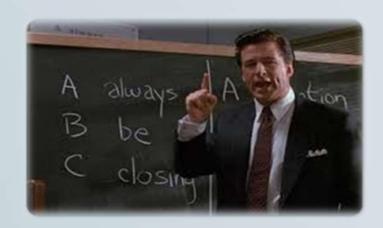


Predicting Ames, IA Housing Prices



Group: Glengarry Glen Ross

- Liam McDermott
- Jan Ruffner
- David Wasserman
- Ethan Zien



Agenda

01

Setting the Stage

02

Feature Analysis and Selection

03

Model Methodology

04

Results & Conclusion

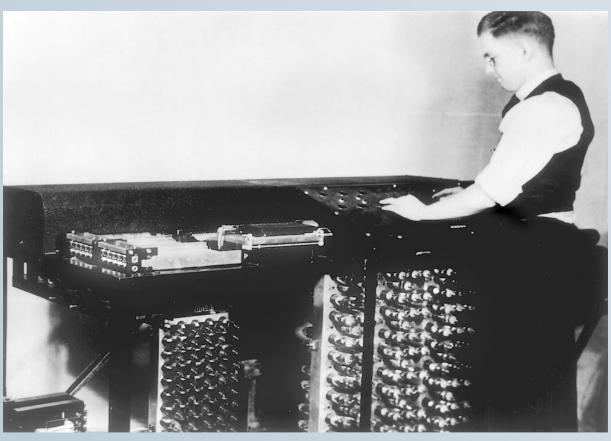
Objective: Deliver a model to describe housing market in Ames, IA







Only US City with a population greater than 65,000 that has 45% of residents enrolled in college or graduate school



The Atanasoff–Berry Computer (ABC)
First Automatic Electronic Computer
Successfully tested 1942 - Iowa State College



Green Hill: \$198.65 Per Square Foot



South West of ISU: \$89.89 per square foot

Model Preparation (Data Cleaning)

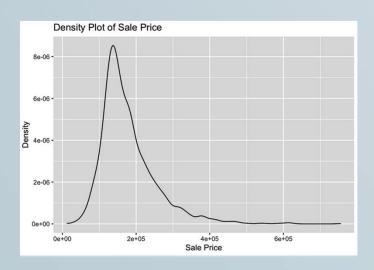
- A. Removed duplicate records from the training set
- B. Imputed **zero (0)** for **numerical** variables where NA (except Year)
- C. Imputed None for categorical variables where NA

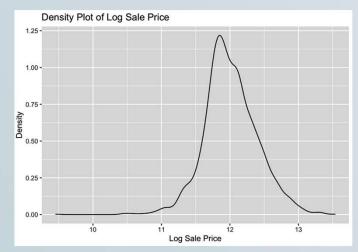


Feature Analysis & Selection

Dependent Variable

• Chose to analyze log sale price to due to having a more normal distribution



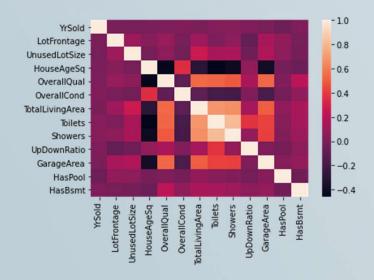


Feature Creation

Feature	Code Name	Calculation
Total Living Area	TotalLivingArea	GrLivArea + TotalBsmttSF - BsmtUnSF
Unused Lot Size	UnusedLotSize	LotArea - 1stFlrSF
Has Pool	HasPool	PoolArea > 0
Has Basement	HasBsmt	BsmtQual != None
Toilets	Toilets	HalfBath + FullBath + BsmtHalfBath + BsmtFullBath
Showers	Showers	FullBath + BsmtFullBath
House Age	HouseAge	max(YearBuilt) - YearBuilt + 1
House Age Squared	HouseAgeSq	HouseAge ^ 2
Up/Down Ratio	UpDownRatio	2ndFlrSF / 1stFlrSF

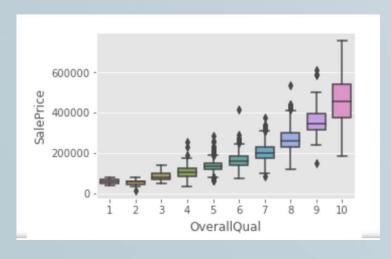
Feature Selection

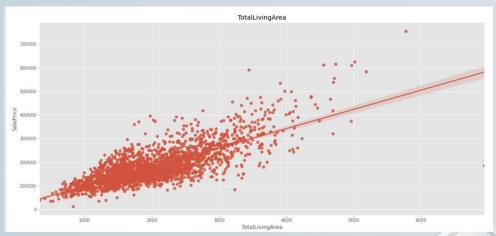
Looked at correlation across metrics



Feature Selection

Looked at different variables and how they correlated with sale price





Model Variable Selection

- Basement Quality
- Building Type
- Garage Area
- Has Basement
- Has Pool
- House Age (Squared)
- Lot Frontage
- Neighborhood
- Overall Condition

- Overall Quality
- Sale Condition
- Showers
- Toilets
- Total Living Area
- Unused Lot Size
- Up/Down Ratio
- Year Sold

These variables showed a strong correlation to the sale price on a house and limited multi-collinearity.

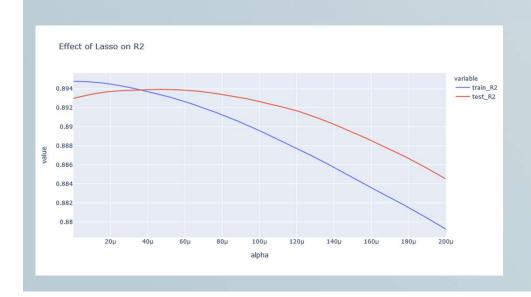


Model Methodology

Regression Models Tested

- 1) Multiple Linear Regression with Lasso Regression
- 2) Random Forest
- 3) XGBoost
- 4) LightGBM
- 5) CatBoost
- 6) Support Vector Machine

Multiple Linear Regression



	coef
const	10.4162
BQ Ex	0.1175
BQ Gd	0.0893
BT_1Fam	0.0970
GarageArea	0.0003
LotFrontage	0.0002
Nbhd_Blmngtn	0.1708
Nbhd_ClearCr	0.0941
Nbhd_Crawfor	0.1362
Nbhd_Gilbert	0.0811
Nbhd_GrnHill	0.6054
Nbhd_NoRidge	0.1008
Nbhd_NridgHt	0.1361
Nbhd_Somerst	0.1168
Nbhd_StoneBr	0.1774
OverallCond	0.0378
OverallQual	0.1064
SC_Partial	0.0375
Toilets	0.0308
TotalLivingArea	0.0002
UnusedLotSize	2.736e-06



The Power of Boosting Models

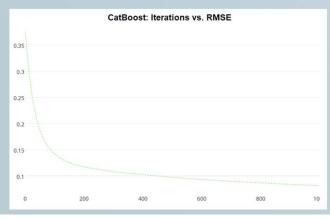


CatBoost



XGBoost







Cross-Model Comparison

Model	Training R ²	Testing R ²	Best RMSE
CatBoost	94.65%	91.62%	0.113
LightGBM	97.40%	91.00%	0.117
XGBoost	99.35%	90.89%	0.118
Random Forest	97.86%	89.66%	0.126
Support Vector Machine	90.38%	88.30%	0.134
Linear Regression	87.00%	88.24%	0.138

Modeling Lessons Learned

- 1) Boosting models perform well but take long to train (especially with CV)
- 2) Linear regression performs poorly when you have many features compared to other models
- 3) While the models are not in complete agreement, some variables always show high importance



Results & Conclusion

Conclusion

Most successful model:

CatBoost Model

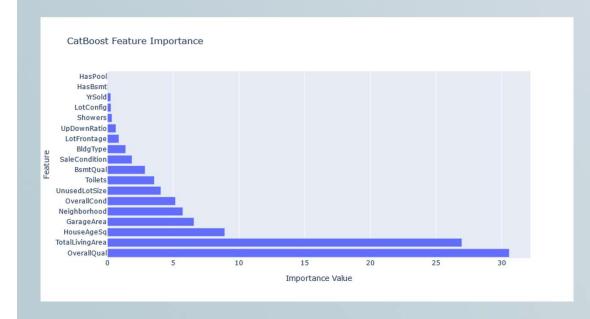
Final Performance:

- R²: 91.6%,

- Root mean squared error: 0.1135



Feature Importances from CatBoost



Feature	Importance
Overall Quality (OverallQual)	31%
Total Living Area (TotalLivingArea)	27%
House Age Squared (HouseAgeSq)	9%
Garage Area (GarageArea)	7%
Neighborhood (Neighborhood)	6%
Overall Condition (OverallCond)	5%
Unused Lot Size (UnusedLotSize)	4%
# of Toilets (Toilets)	4%
Basement Quality (BsmtQual)	3%
Sale Condition (SaleCondition)	2%
Building Type (BldgType)	1%
Lot Frontage (LotFrontage)	1%
Upstairs/Downstairs Ratio (UpDownRatio)	1%

Next steps

Compare offering prices to what the model predicts to find bargains

Sale Price: \$280,000

4 bedrooms

2 full baths + 2 half baths

2,241 sqft

Our estimate: \$324,000

