

Real Estate Sales Price Prediction Project

Predictive Modeling for Enhanced Real Estate Valuation

Presented By: Donn Bryan Julian

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Project Overview

Goal:

- * Predict real estate sales prices using historical data.

Scope:

- * Focused on properties from 2001 to 2022, using a variety of statistical and machine learning methods to develop an accurate prediction model.

Key Takeaway:

- * The best model has an R-squared of 0.949, demonstrating strong predictive capabilities.

Source Data

- * **Data Source:**

Acquired from <https://catalog.data.gov/dataset/real-estate-sales-2001-2018> covering real estate sales between 2001-2022.

- * **Details:**

Over 1 million records with features like assessed value, property type, town, and sale amount.

Exploratory Data Analysis (EDA) Overview

- * **Objective:**

Understand data distribution, identify outliers, and recognize relationships.

- * **Initial Insights:**

Data contains multiple types, non-uniform missing values, and varying distribution across categorical features.

Data Cleansing

Actions Taken:

- * Removed towns with fewer than 500 entries to eliminate noise.
- * Dropped non-numeric fields like 'Address' and 'Date Recorded' that didn't contribute to prediction.

Outcome:

- * Focused on significant features to enhance model performance.

Statistical Tests Conducted

Multicollinearity (VIF Analysis):

Identified and removed highly correlated features to reduce redundancy.

Autocorrelation: Performed Ljung-Box test to verify residual independence.

Heteroskedasticity: Breusch-Pagan test confirmed minimal variance issues.

Normality: Applied Lilliefors and Shapiro-Wilk tests; deviations from normality informed the selection of ensemble models.

OLS Regression Results

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Dep. Variable:          Sale Amount      R-squared:                0.021
Model:                  OLS              Adj. R-squared:           0.021
Method:                 Least Squares    F-statistic:             823.6
Date:                  Thu, 05 Dec 2024  Prob (F-statistic):      0.00
Time:                  10:15:43          Log-Likelihood:          -1.8494e+07
No. Observations:      1096796          AIC:                    3.699e+07
Df Residuals:          1096767          BIC:                    3.699e+07
Df Model:              28
Covariance Type:       nonrobust
    
```

List Year	1.028e+04	1303.971	7.884	0.000	7725.305	1.28e+04
Assessed Value	0.3676	0.003	122.475	0.000	0.362	0.373
Town_Darien	1.013e+06	5.99e+04	16.894	0.000	8.95e+05	1.13e+06
Town_Fairfield	3.51e+05	3.84e+04	9.134	0.000	2.76e+05	4.26e+05
Town_Greenwich	9.673e+05	3.86e+04	25.054	0.000	8.92e+05	1.04e+06
Town_New Canaan	9.124e+05	6.08e+04	15.001	0.000	7.93e+05	1.03e+06
Town_Ridgefield	3.508e+05	5.52e+04	6.355	0.000	2.43e+05	4.59e+05
Town_Rocky Hill	4.122e+05	6.53e+04	6.316	0.000	2.84e+05	5.4e+05
Town_Stamford	4.071e+05	2.72e+04	18.260	0.000	4.44e+05	5.5e+05
Town_Washington	4.518e+05	1.2e+05	3.773	0.000	2.17e+05	6.86e+05
Town_Weston	5.06e+05	8.37e+04	6.047	0.000	3.42e+05	6.7e+05
Town_Westport	9.349e+05	5.04e+04	18.552	0.000	8.36e+05	1.03e+06
Town_Willington	3.834e+06	1.35e+05	28.459	0.000	3.57e+06	4.1e+06
Town_Wilton	4.802e+05	6.59e+04	7.284	0.000	3.51e+05	6.09e+05
Property Type_Commercial	-5.049e+06	1.55e+05	-32.678	0.000	-5.35e+06	-4.75e+06
Property Type_Condo	-3.303e+06	3.27e+05	-10.105	0.000	-3.94e+06	-2.66e+06
Property Type_Four Family	-3.2e+06	3.41e+05	-9.394	0.000	-3.87e+06	-2.53e+06
Property Type_Industrial	-4.345e+06	2.28e+05	-19.027	0.000	-4.79e+06	-3.9e+06
Property Type_Public Utility	-5.972e+06	1.62e+06	-3.695	0.000	-9.14e+06	-2.8e+06
Property Type_Residential	-3.326e+06	3.37e+05	-9.865	0.000	-3.99e+06	-2.67e+06
Property Type_Single Family	-3.369e+06	3.27e+05	-10.290	0.000	-4.01e+06	-2.73e+06
Property Type_Three Family	-3.413e+06	3.36e+05	-10.148	0.000	-4.07e+06	-2.75e+06
Property Type_Two Family	-3.317e+06	3.28e+05	-10.120	0.000	-3.96e+06	-2.67e+06
Property Type_Unknown	-5.759e+06	1.42e+05	-40.648	0.000	-6.04e+06	-5.48e+06
Property Type_Vacant Land	-5.854e+06	1.51e+05	-38.695	0.000	-6.15e+06	-5.56e+06
Residential Type_Single Family	1.499e+05	2.98e+04	5.038	0.000	9.16e+04	2.08e+05
Residential Type_Three Family	1.073e+05	8.46e+04	1.268	0.205	-5.85e+04	2.73e+05
Residential Type_Unknown	2.654e+06	3.58e+05	7.417	0.000	1.95e+06	3.36e+06

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Omnibus:                7855902.503      Durbin-Watson:            2.000
Prob(Omnibus):          0.000           Jarque-Bera (JB):        32283762171578400.000
Skew:                   861.949           Prob(JB):                0.00
Kurtosis:               840495.511        Cond. No.                4.02e+15
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	Feature	VIF
1	Serial Number	1.011111
2	List Year	3.299387
3	Assessed Value	1.047965
4	Town_Darien	1.004110
5	Town_Fairfield	1.004700
6	Town_Greenwich	1.036704
7	Town_New Canaan	1.003985
8	Town_Ridgefield	1.002206
9	Town_Rocky Hill	1.003642
10	Town_Stamford	1.011763
11	Town_Washington	1.000562
12	Town_Weston	1.001897
13	Town_Westport	1.004577
14	Town_Willington	1.000511
15	Town_Wilton	1.001744
16	Property Type_Commercial	5.479774
19	Property Type_Industrial	1.598003
20	Property Type_Public Utility	1.007533
26	Property Type_Vacant Land	6.823659
27	Residential Type_Single Family	9.309094
28	Residential Type_Three Family	4.502980

Durbin-Watson statistic: 2.000090289364537

Breusch-Pagan test for heteroskedasticity:
Statistic: 1640.911699957239, p-value: 0.0

Shapiro-Wilk Test for Normality of Residuals:
Statistic: 0.009829581171743484, p-value: 1.2134210007860006e-239

Jarque-Bera Test for Normality of Residuals:
Statistic: 3.22837621715784e+16, p-value: 0.0

Lilliefors Test for Normality of Residuals:
Statistic: 0.43000719787387415, p-value: 0.00099999999999998899

Ljung-Box Test for Autocorrelation of Residuals:

	lb_stat	lb_pvalue
10	0.289082	1.0

Feature Selection Techniques

- * **Variance Thresholding:**

Removed features with low variance that provided little predictive information.

- * **ANOVA F-test:**

Dropped statistically insignificant features.

- * **Outcome:**

Reduced dimensionality to retain the most impactful predictors.

Data Preparation

One-Hot Encoding:

- * Converted categorical features like 'Town' to dummy variables.

Sampling:

- * Used 10% of the dataset to speed up the model training.

Final Dataset Size:

- * Over 100,000 records used for modeling.

Model Testing and Selection

Models Tested:

- * Linear Regression, Ridge, Lasso, Random Forest, Gradient Boosting, XGBoost.

Selection Criteria:

- * R-squared, RMSE, and MAE for performance evaluation.

	Model	R-squared	RMSE	MAE
0	Linear Regression	-0.1832	1215916.53	206676.36
1	Ridge Regression	-0.1832	1215912.07	206616.14
2	Lasso Regression	-0.1832	1215915.04	206667.30
3	Random Forest Regressor	0.9188	318506.28	12685.29
4	Gradient Boosting Regressor	0.4392	837115.20	214755.63
5	XGBoost Regressor	0.5735	730000.71	61023.49
Mean Cross-Validation R-squared				
0		0.1971		
1		0.1971		
2		0.1971		
3		0.7069		
4		0.4528		
5		0.4014		

Model Performance Overview

Random Forest Results: R-squared: 0.9042

Root Mean Squared Error (RMSE): 345,908

Mean Absolute Error (MAE): 11,250

Insight: Random Forest was the best-performing model, indicating strong predictive power.

Random Forest Model - Detailed Evaluation

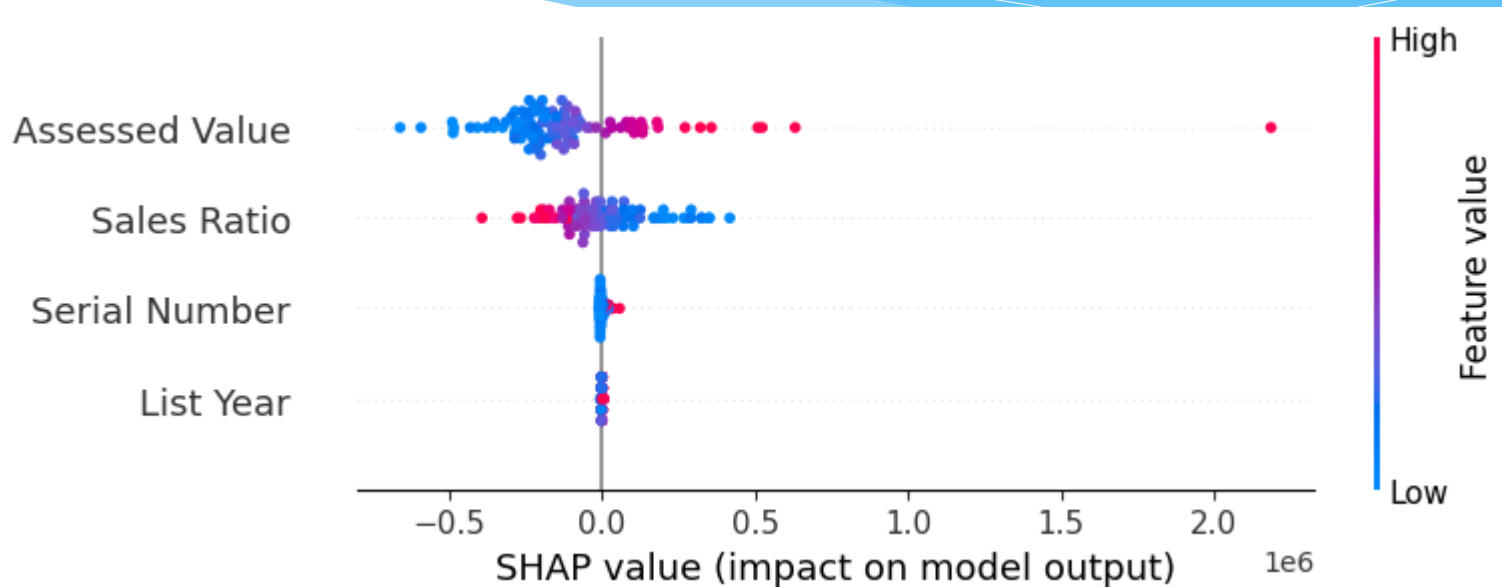
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Fitting 3 folds for each of 24 candidates, totalling 72 fits
Best Parameters for Random Forest: {'max_depth': 20, 'min_samples_leaf': 1, 'min_samples_split': 2, 'n_estimators': 100}

Random Forest Regressor with Best Parameters:
R-squared: 0.9042
Root Mean Squared Error (RMSE): 345908.40
Mean Absolute Error (MAE): 11250.41

Feature Importances:
Sales Ratio                4.787942e-01
Assessed Value             4.272614e-01
Property Type_Vacant Land  2.722328e-02
Serial Number              2.220814e-02
Town_Stamford              1.207017e-02
...
Town_Hampton               1.833870e-10
Town_Thomaston             1.801281e-10
Town_Chaplin               1.794898e-10
Town_Voluntown             7.566988e-11
Property Type_Public Utility 0.000000e+00
Length: 186, dtype: float64
```

- * **Cross-Validation:** Average R-squared of 0.788 across 5-fold validation.
- * **Residual Analysis:** Confirmed no visible pattern in residuals—indicative of a well-fitting model.

Understanding Predictions with SHAP



- * **SHAP Analysis:**

Key features driving model predictions were Assessed Value and Sales Ratio.

- * **Interpretation:**

SHAP values help visualize each feature's contribution—red bars indicate higher values pushing the prediction up, while blue reduces it.

This provides transparency into model decision-making, aligning predictions with real-world expectations.

Deployment Preparation

Model Packaging:

- * Saved the model as a .pkl file.

Deployment Plan:

- * Created a REST API using Flask for real-time predictions.
- * Future deployment on Hugging Face for broader accessibility.

Business Insights & Next Steps

Insights Gained:

- * Key predictors like Assessed Value can be leveraged for targeted marketing and strategic pricing.

Next Steps:

- * Deploy the model on Hugging Face.
- * Monitor performance and plan for retraining.