

Milwaukee Property Search Dashboard

Data visualization project using SQLite, Python, MongoDB, Flask, JS, HTML, ML

Presenters











Project Objective

Interactive UI that provides information on Milwaukee property listings to help users make informed decisions for purchasing residential properties within this area.

- Milwaukee Property Insights:

- Get immediate access to detailed property information predicted 2025 house price

Customizable Searches:

- Filter properties by the number of bedrooms to find exactly what you need.

- User-Friendly:

 Designed for simplicity, our dashboard makes finding and analyzing Milwaukee properties intuitive and effective.

Dataset and Limitations

No Free Toronto dataset found for our Requirements!

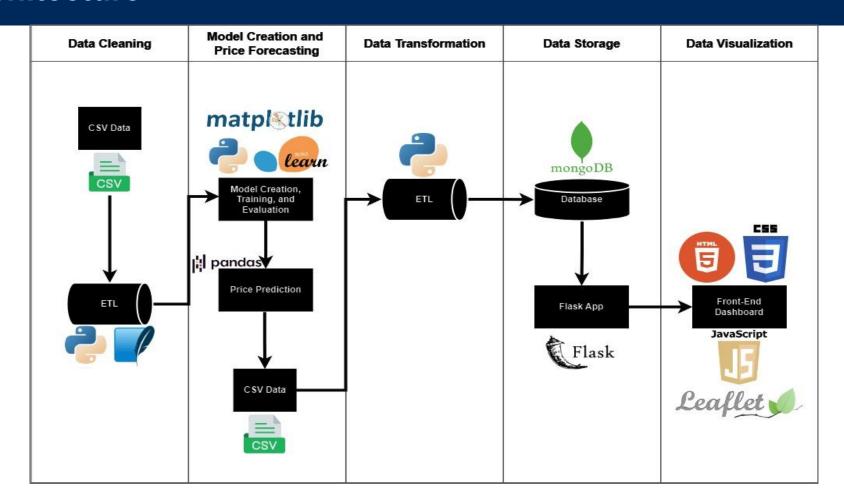
Dataset 1: Milwaukee Open Data - Individual Property Sales Data from 2013 to 2023

- ~46,000 records before cleaning → ~37,000 records after cleaning
- Columns: 'district', 'nbhd', 'style', 'extwall', 'stories', 'year_built', 'rooms', 'finishedsqft', 'units', 'bdrms', 'fbath', 'hbath', 'lotsize', 'sale_date', 'sale_price'
- Issues: null values, 0s, and incomplete/inaccurate data

Dataset 2: Kaggle - 2023 United States House Listings: Zillow Extract

- Filtered for Milwaukee
- Columns: finishedsqft', 'bdrms', 'baths', 'lotsize', 'state', 'city', 'address', 'zipcode', 'latitude', 'longitude', 'ppsq', 'convertedlot', 'lotunit', 'marketestimate', rentestimate', 'price'
- Issues: null values, 0s, incomplete data
 - Required manual population based on similar characteristics
 - Large difference in features may result in inaccurate 2025 house price forecast predictions

Architecture



Data Model Optimization

Model Information	Performance
Linear Regression	Accuracy: 0.55
- Trained with all 53 input features	Mean Squared Error: 3657767974.09

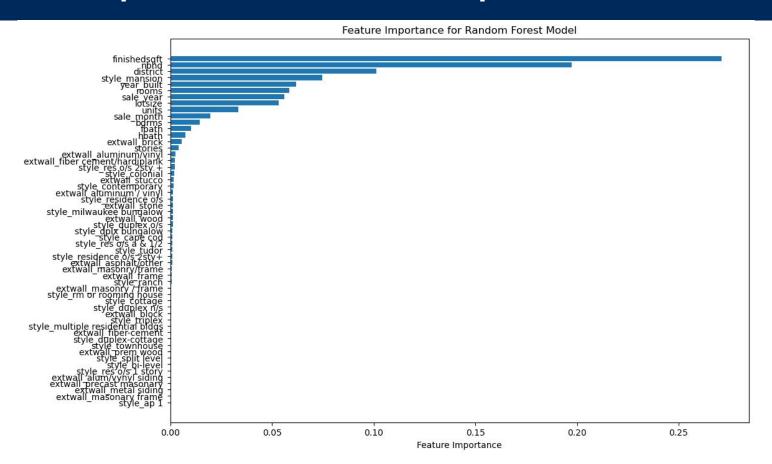
Data Model Optimization - Coefficients and P-Values

	Coefficient	p-value	
district	-2.249098e+03	1.998269e-73	
nbhd	2.292663e+01	0.000000e+00	
stories	1.322653e+04	1.220916e-12	
year_built	2.111880e+02	8.306841e-13	
rooms	1.476720e+02	3.422613e-01	
finishedsqft	7.278040e+01	0.000000e+00	
units	-4.933133e+04	3.369052e-36	
bdrms	-8.492239e+01	7.903038e-03	
fbath	3.337901e+04	0.000000e+00	
hbath	1.762184e+04	2.684586e-108	
lotsize	1.782317e+00	2.171756e-76	
sale_year	1.010315e+04	0.000000e+00	
sale_month	7.463099e+02	5.250358e-11	
style_ap 1	-5.886605e+05	1.835462e-19	
style_bi-level	-8.193872e+05	0.000000e+00	
style_cape cod	-8.121571e+05	0.000000e+00	
style_colonial	-8.114070e+05	0.000000e+00	
style_contemporary	-6.379182e+05	3.126325e-132	
style_cottage	-8.334216e+05	0.000000e+00	
style_dplx bungalow	-8.577632e+05	0.000000e+00	
style_duplex n/s	-8.492868e+05	0.000000e+00	
style_duplex o/s	-8.564366e+05	0.000000e+00	
style_duplex-cottage	-8.699977e+05	0.000000e+00	
style_mansion	-4.341570e+05	1.168804e-124	

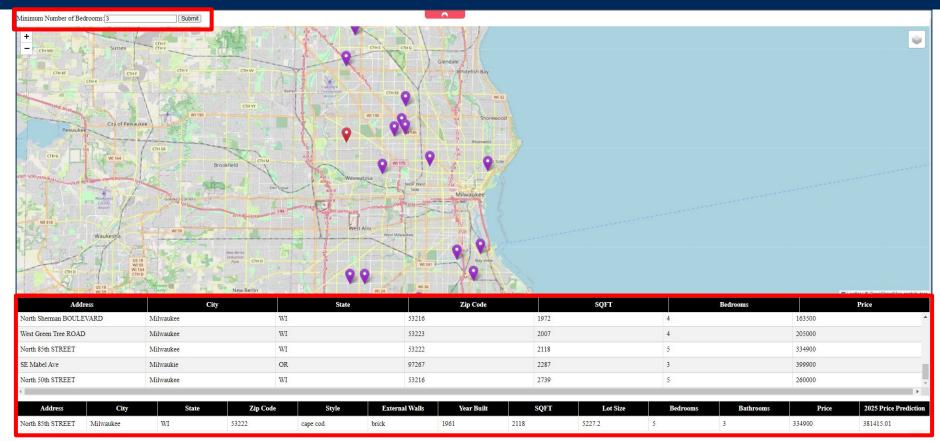
Data Model Optimization

Model Information	Performance
Random Forest Regressor	Accuracy: 0.8365
- Trained with all 53 input features	Mean Squared Error: 1497358586.32

Data Model Optimization - Feature Importance



Live Demonstration



Future Improvements

- Toronto data

- More relevance and applicability

- Paid datasource

- Data reliability and completeness

- <u>Dashboard Improvements</u>

- Additional filter logic
- Other points of interests
- Styling and layout

Thank you! Questions?