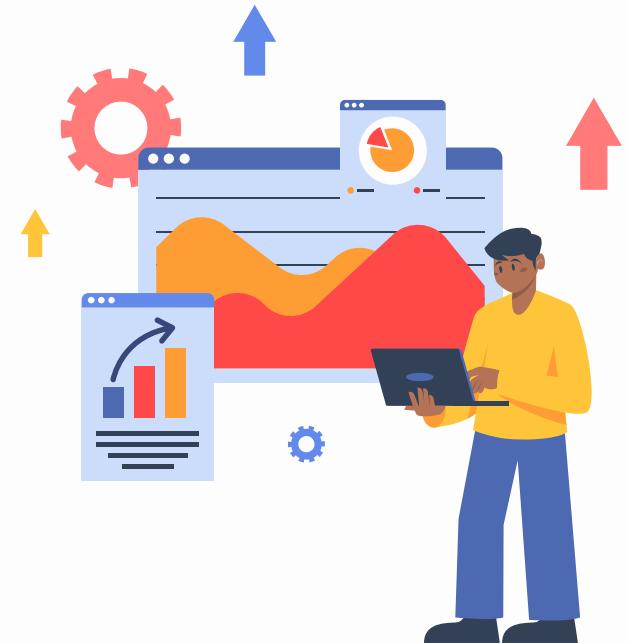


Predictive Analytics for King County Real Estate

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Introduction

- King County (Seattle area) house sales from May 2014–May 2015.
- 21,613 entries, 21 features (price, size, rooms, etc.).
- Target variable: `price` - ideal for regression and feature engineering.
- Useful for data cleaning, visualization, and model evaluation practice.
- Data limited to one year - price values are highly skewed.



Agenda

01 Data Cleaning and Transformation

02 Exploring Data Analysis (EDA)

03 ML Model Performances

04 Conclusions





01

Problem-solving strategies

Apply strategies to clean and transform data into clean dataframe



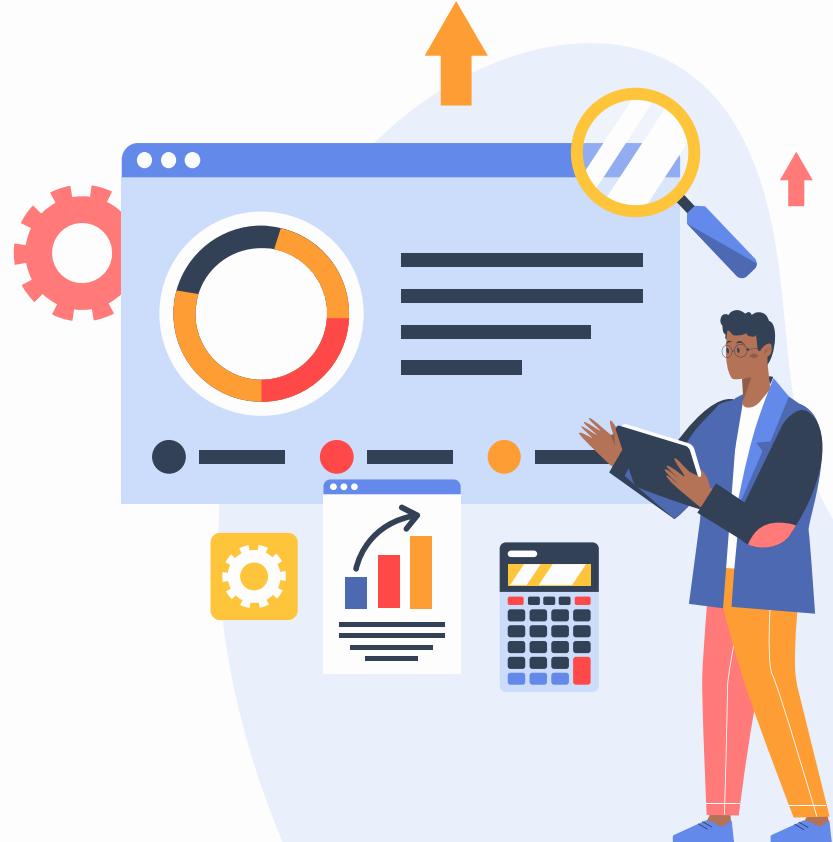
Data Cleaning Process

- Verified dataset structure: **21,613 rows × 21 columns**.
- No missing values detected → **no imputation required**.
- **Removed duplicates** to ensure data integrity.
- Adjusted data types: converted numeric fields to **integer/float**.
- **Dropped irrelevant columns** (e.g., `id`, `date`, `sqft_lot`, `condition`, `zipcode`).
- Final dataset **cleaned and ready for EDA and modeling**.



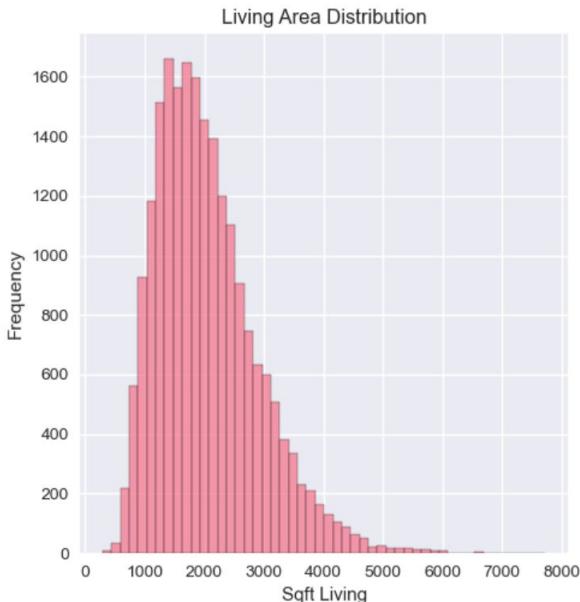
02

Exploring Data Analysis (EDA)



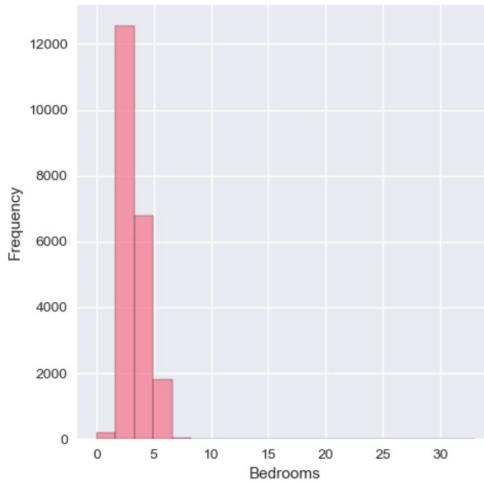


Univariate Analysis

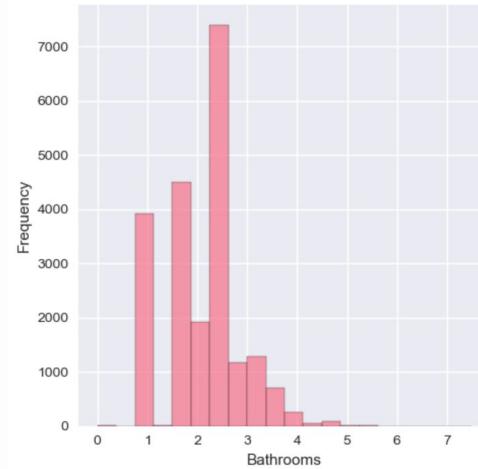


Univariate Analysis

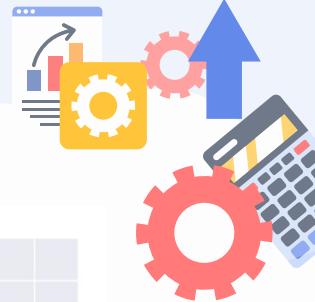
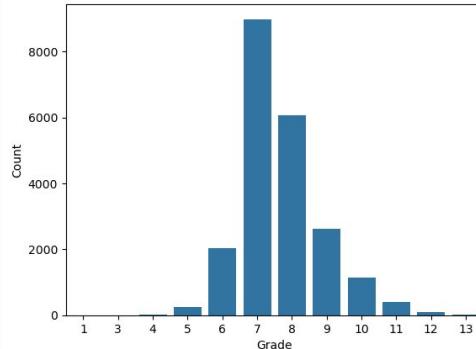
Bedrooms Distribution



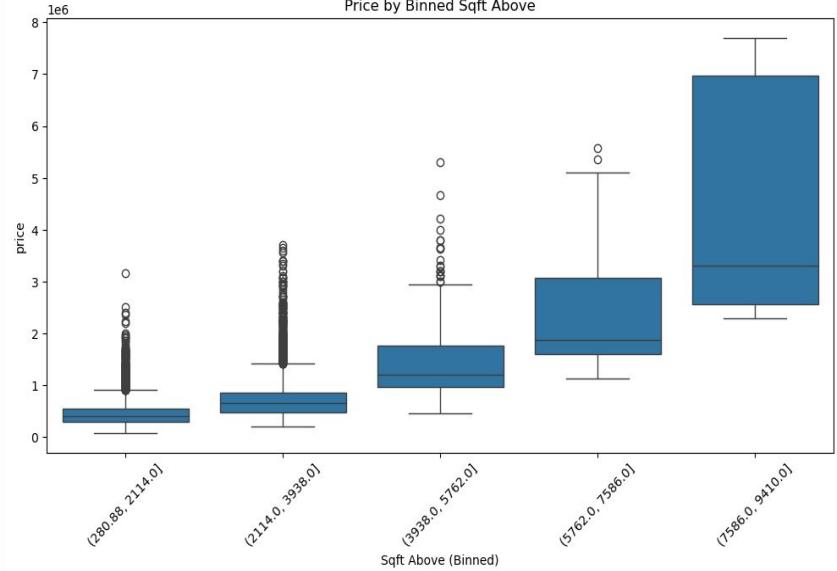
Bathrooms Distribution



House Grade Distribution



Price by Binned Sqft Above



Price vs Sqft Living



Machine learning



Chosen Dataframes (8 Versions)

#	Version	Description	Features	Rows (Train)
1	Raw_All	No cleaning	18	17,290
2	Raw_Selected	Feature selection only	14	17,290
3	Out_All	Outliers removed (IQR × 3)	18	~15,800
4	Out_Selected	Outliers + selection	14	~15,800
5	Norm_All	Scaled only	18	17,290
6	Norm_Selected	Scaled + selection	14	17,290
7	Clean_All	Outliers + scaled	18	~15,800
8	Clean_Selected	Outliers + scaled + selection	14	~15,800

Investigation

Model	Best Version	Test MAE	MAE%	Remarks
XGBoost	Clean_Selected	\$66,800	12.3%	Clean + top 14 features → perfect splits
CatBoost	Clean_All	\$65,200	12.0%	Handles grade natively
Random Forest	Clean_Selected	\$68,900	12.7%	Ensemble focused data
Gradient Boosting	Clean_Selected	\$70,100	12.9%	Similar to XGBoost
AdaBoost	Out_All	\$91,200	16.8%	Sensitive to scaling
Decision Tree	Out_Selected	\$78,900	14.5%	Overfits without selection
Linear	Clean_All	\$101,234	18.6%	Needs non-linearity

Model Performance

	Model	Version	Split	MAE	MAE%	MSE	MSE%	RMSE	RMSE%	R2
1	Random Forest	Clean_Selected	Test	55292	11.68	8827399284	3.94	93954	19.85	0.8423
			Train	24498	5.22	2022527069	0.92	44973	9.58	0.9627
2	AdaBoost	Out_All	Test	55796	11.79	9058010648	4.05	95174	20.11	0.8382
			Train	243	0.05	3591054	0.0	1895	0.4	0.9999
3	Decision Tree	Out_Selected	Test	71636	15.14	15741403750	7.03	125465	26.51	0.7188
			Train	54663	11.64	7602247851	3.45	87191	18.57	0.8599
4	Linear Regression	Clean_All	Test	88697	18.74	18394869649	8.21	135628	28.66	0.6714
			Train	87552	18.65	17084376088	7.75	130707	27.84	0.6852

Model Performance

	Model	Version	Split	MAE	MAE%	MSE	MSE%	RMSE	RMSE%	R2
5	CatBoost	Clean_All	Test	50860	10.75	6858240875	3.06	82814	17.5	0.8775
			Train	33573	7.15	2458951166	1.12	49588	10.56	0.9547
6	XGBoost	Clean_Selected	Test	52153	11.02	7431605234	3.32	86207	18.22	0.8673
			Train	36392	7.75	2962109199	1.34	54425	11.59	0.9454
7	Gradient Boosting	Clean_Selected	Test	52212	11.03	7493604635	3.35	86566	18.29	0.8661
			Train	32363	6.89	2179741937	0.99	46688	9.95	0.9598



WINNER - CATBOOST



Accuracy

Model is correct within
±10.75% on average —
that's 89.25% accurate!



MAE

Average error is 10.75%
of the true price

Test MAE

\$50,860

average mistake
on ~\$473k houses



Thanks!

