Project Title Bangalore Home Price Prediction

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Objective

- Predict fair house prices in Bangalore using historical data.
- Helps buyers and sellers make data-driven decisions.
- Focus on area, BHK, bathrooms, and location.

Dataset Overview

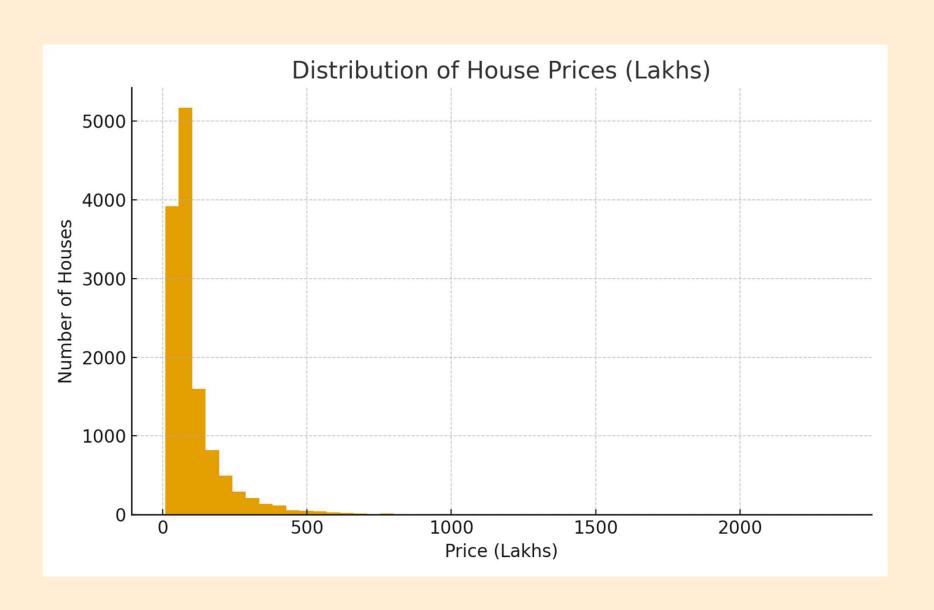
- Rows in original dataset: 13320
- Columns: area_type, availability, location, size, society, total_sqft, bath, balcony, price
- Key features used: total_sqft, BHK (size), bath, location, price

	area_type	availability	location	size	society	total_sqft	bath	balcony	price
0	Super built-up Area	19-Dec	Electronic City Phase II	2 BHK	Coomee	1056	2.0	1.0	39.07
1	Plot Area	Ready To Move	Chikka Tirupathi	4 Bedroom	Theanmp	2600	5.0	3.0	120.00
2	Built-up Area	Ready To Move	Uttarahalli	3 BHK	NaN	1440	2.0	3.0	62.00
3	Super built-up Area	Ready To Move	Lingadheeranahalli	3 BHK	Soiewre	1521	3.0	1.0	95.00
4	Super built-up Area	Ready To Move	Kothanur	2 BHK	NaN	1200	2.0	1.0	51.00

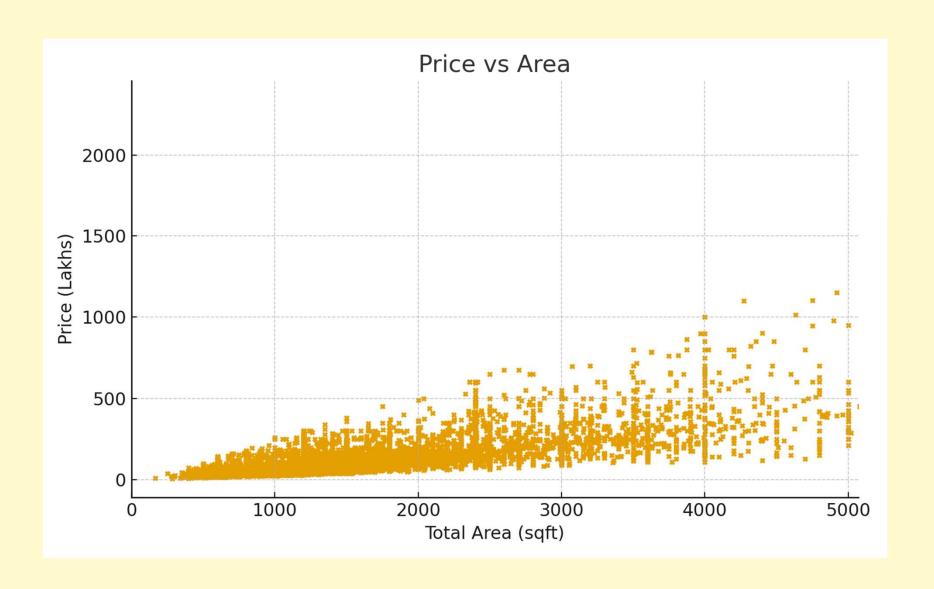
Data Preprocessing

- Converted 'total_sqft' to numeric, handled ranges and units.
- * Extracted BHK from 'size' column.
- Computed price per sqft and removed outliers (1st-99th percentile).
- Final cleaned rows used for modeling: 13040

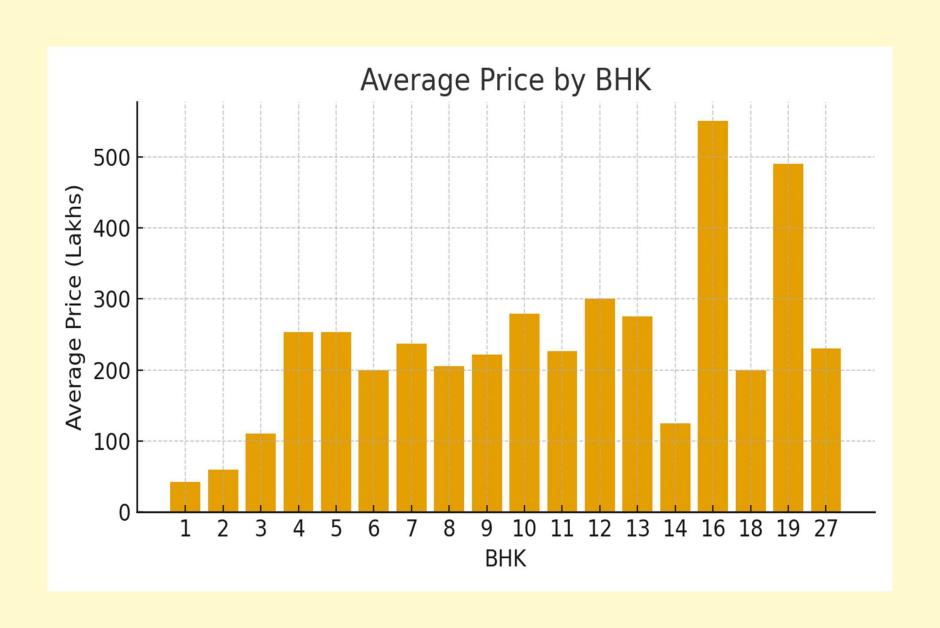
Price Distribution (Lakhs)



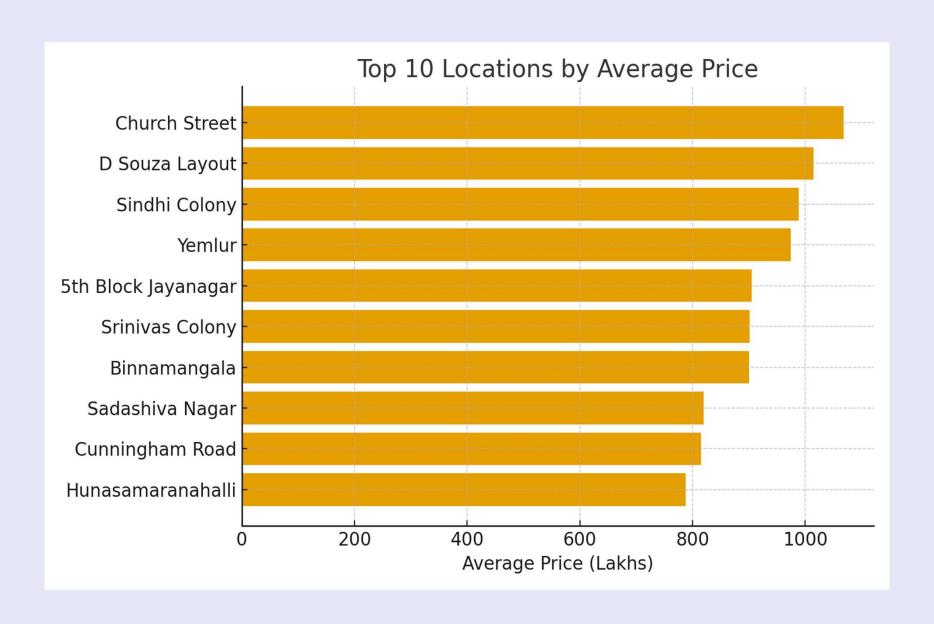
Price vs Area (sqft)



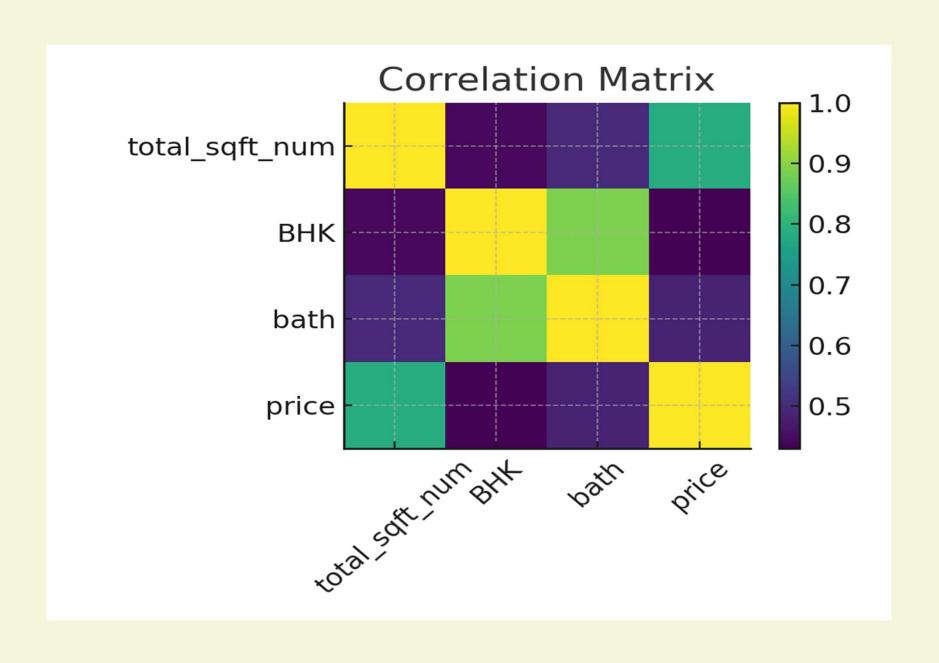
Average Price by BHK



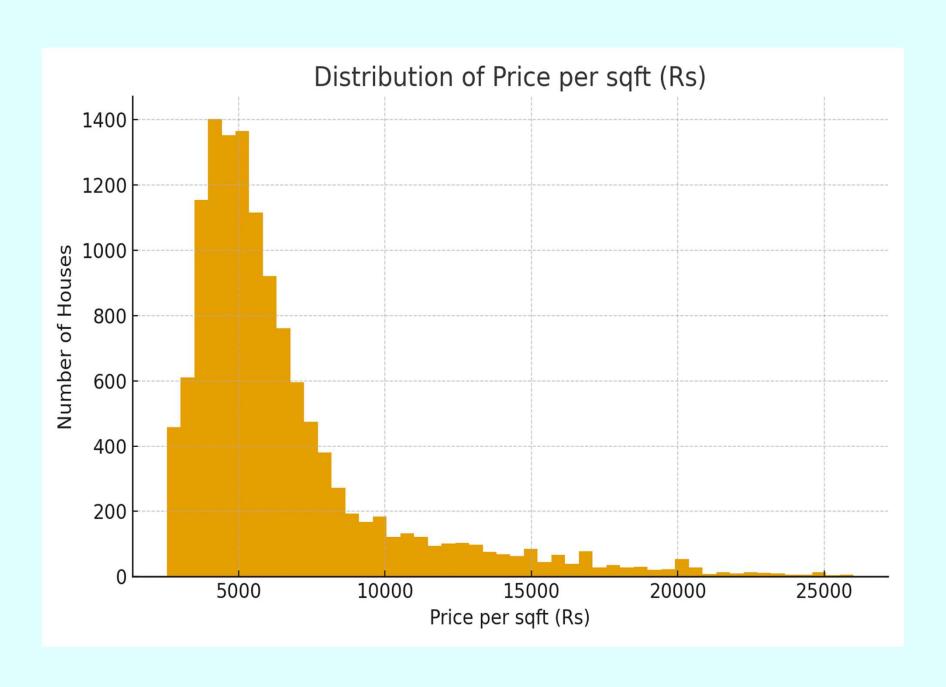
Top 10 Locations by Average Price



Correlation Matrix (Area, BHK, Bath, Price)



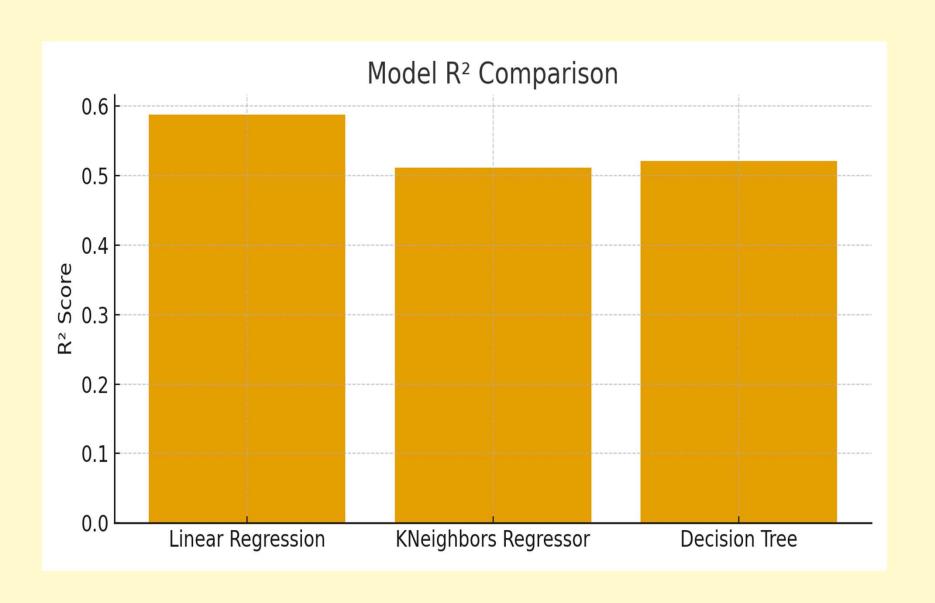
Price per sqft Distribution (Rs)



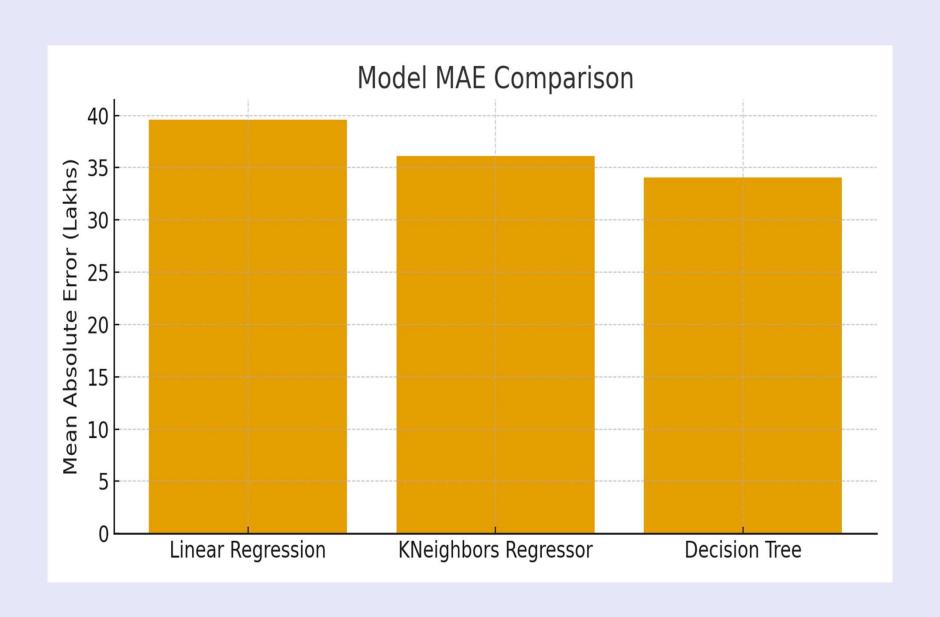
Model Building - Approach

- Selected features: total_sqft, BHK, bath, and top locations (one-hot).
- Tried multiple algorithms: Linear Regression, KNN, Decision Tree.
- Trained with 80% data and tested with 20% data.

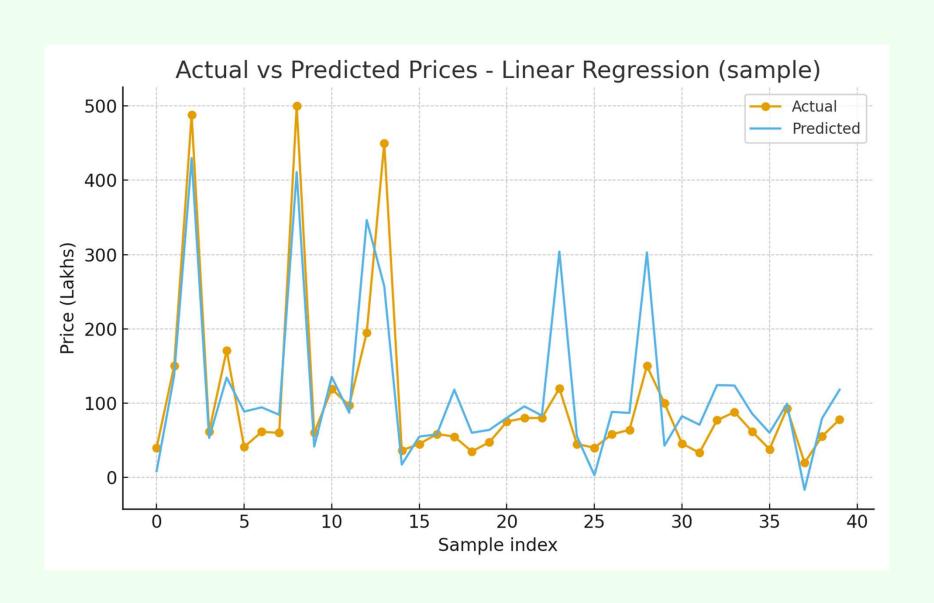
Model R² Comparison



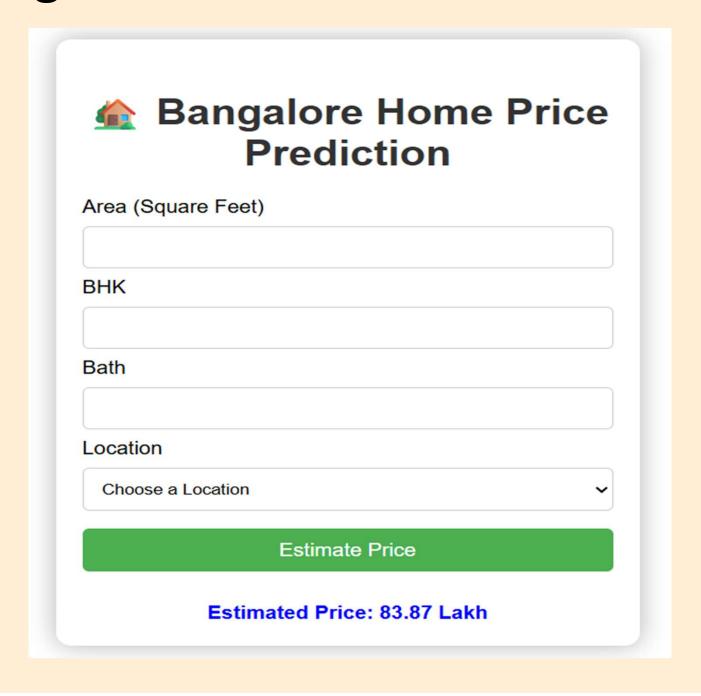
Model MAE Comparison (Lakhs)



Actual vs Predicted



Bangalore Home Price Prediction



Results & Insights

- **A** Best model (by R²): Linear Regression
- Location and area are strong drivers of price; BHK and bathrooms also matter.
- Price per sqft varies widely across locations location heavily influences price.

Conclusion & Future Work

- Model provides reasonable price estimates using basic features.
- Future improvements: add amenities, year built, proximity to metro/schools, and more features.
- Deploy model with REST API and mobile frontend for public use.

Thank you