

# HOUSE PRICES EDA REPORT

Exploratory Data Analysis

Kaggle House Prices Competition

Generated: 2025-11-09

# EXECUTIVE SUMMARY

## DATA OVERVIEW:

Samples: 1,460

Features: 81

Price Range: 34,900 – 755,000

## DATA QUALITY:

Columns with Missing Values: 19

Highest Missing: PoolQC (99.52%)

## KEY FINDINGS:

OverallQual: Correlation 0.791

GrLivArea: Correlation 0.709

GarageCars: Correlation 0.640

## CONCLUSIONS:

- Target variable is right-skewed, log transform recommended
- Strong correlated features identified for modeling
- Missing values need to be handled

# **DATASET OVERVIEW**

## **BASIC INFORMATION:**

- Data Shape: (1460, 81)
- Numeric Features: 38
- Categorical Features: 43
- Memory Usage: 3.4 MB

## **TARGET STATISTICS:**

- Mean: \$180,921
- Median: \$163,000
- Standard Deviation: \$79,443
- Skewness: 1.883

# MISSING VALUES ANALYSIS

## SUMMARY:

- Columns with Missing Values: 19
- Total Missing Values: 7829
- Average Missing %: 28.2%

## TOP MISSING COLUMNS:

1. PoolQC: 99.52%
2. MiscFeature: 96.3%
3. Alley: 93.77%
4. Fence: 80.75%
5. MasVnrType: 59.73%
6. FireplaceQu: 47.26%
7. LotFrontage: 17.74%
8. GarageYrBlt: 5.55%

# TARGET VARIABLE ANALYSIS

## STATISTICS:

- Skewness: 1.883
- Kurtosis: 6.536
- Coefficient of Variation: 0.439
- Recommendation: Log transformation

## DISTRIBUTION NOTES:

- Distribution is right-skewed
- Few high-priced houses pull mean upward
- Log transformation will help

# CORRELATION ANALYSIS

## CORRELATION STRENGTH:

- Strong ( $>0.5$ ): 11 features
- Moderate (0.3-0.5): 4 features
- Weak ( $<0.3$ ): 0 features

## TOP FEATURES:

1. OverallQual: 0.791

*(Very Strong)*

2. GrLivArea: 0.709

*(Very Strong)*

3. GarageCars: 0.640

*(Strong)*

4. GarageArea: 0.623

*(Strong)*

5. TotalBsmtSF: 0.614

*(Strong)*

# FEATURE ENGINEERING

## NEW FEATURES:

- HouseAge
- TotalArea
- HasPool
- TotalBath
- log\_SalePrice
- log\_GrLivArea
- log\_LotArea

# CONCLUSIONS AND RECOMMENDATIONS

## MAIN CONCLUSIONS:

- Good data quality with some missing values
- Target variable needs log transformation
- Strong features identified for modeling
- Feature engineering can improve performance

## NEXT STEPS:

1. Advanced feature engineering
2. Handle categorical variables
3. Build predictive models
4. Model evaluation and tuning
5. Results interpretation

*Report completed*