



# Socio-Economic Factors and U.S. County-Level Housing Prices

*Exploring how income, poverty, education, and unemployment relate to housing values across U.S. counties using machine learning and statistical clustering.*

# Research Objectives and Methodology



## Primary Goal

*Link county-level median home values with key socio-economic characteristics including income, poverty rates, educational attainment and unemployment figures*



## Analytical Approach

*Employ correlation analysis, K-means clustering, outlier detection and tree-based feature importance models to reveal underlying patterns*



## Geographic Focus

*Examine regional inequalities and distinctive market characteristics across 3,052 US counties with complete data coverage*



# Data Sources: Detailed Overview

<i>Data Source</i>	<i>Description</i>	<i>Processing Approach</i>	<i>Purpose</i>
<i>Zillow Home Value Index</i>	<i>County-level median home values including RegionName, State, and FIPS codes</i>	<i>Downloaded CSV from Zillow research portal; cleaned FIPS codes using .str.zfill() to create uniform 5-digit identifiers; filtered for latest available date</i>	<i>Analyze housing price distribution and serve as primary dependent variable</i>
<i>US Census Bureau (ACS API)</i>	<i>Socio-economic data from American Community Survey including median income, population, poverty statistics, and educational attainment</i>	<i>Accessed via Census API with authentication; decoded variable names (B19013_001E, B15003_022E, etc.); calculated derived metrics (poverty rate, college-educated percentage); standardized FIPS codes</i>	<i>Examine correlations between housing prices and socio-economic factors</i>
<i>Bureau of Labor Statistics (LAUS)</i>	<i>County-level unemployment data , including unemployment rate and labor force statistics</i>	<i>Downloaded Excel file from BLS website; cleaned column names and filtered for 2022 data; standardized state and county FIPS codes; merged with other datasets using FIPS keys</i>	<i>Investigate relationship between labor market conditions and housing markets</i>

*After merging datasets by county FIPS codes and cleaning missing values, the final analytical dataset comprises 3,052 counties with complete information across all variables.*



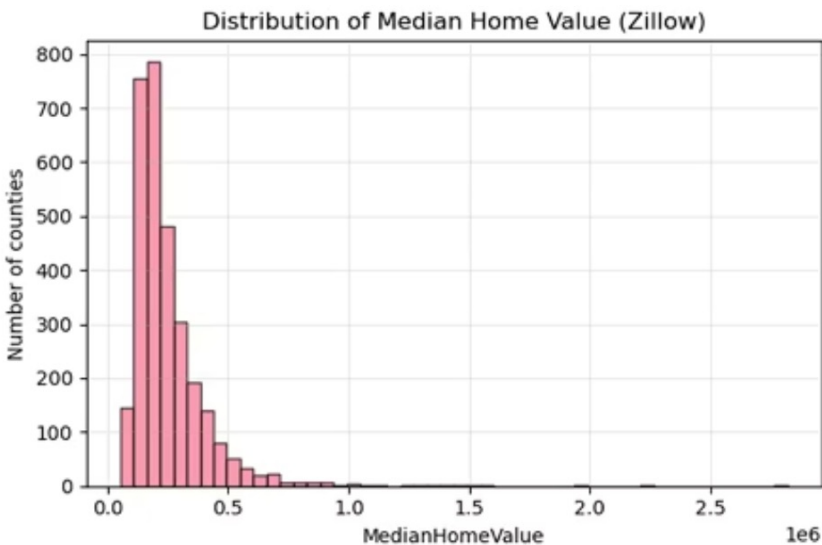
# Descriptive Overview of US Counties in 2022

Descriptive statistics reveal extraordinary diversity in housing markets and socio-economic conditions across America's 3,052 counties.

\$248K

Mean Home Value

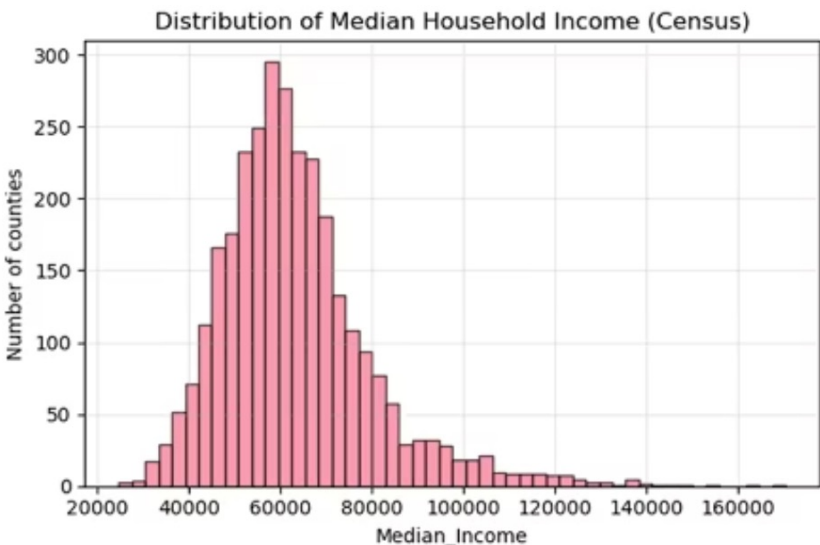
Range: \$51K – \$2.82M (highly right-skewed distribution)



\$63.4K

Mean Income

Range: \$24.6K – \$170.5K across counties



13.7%

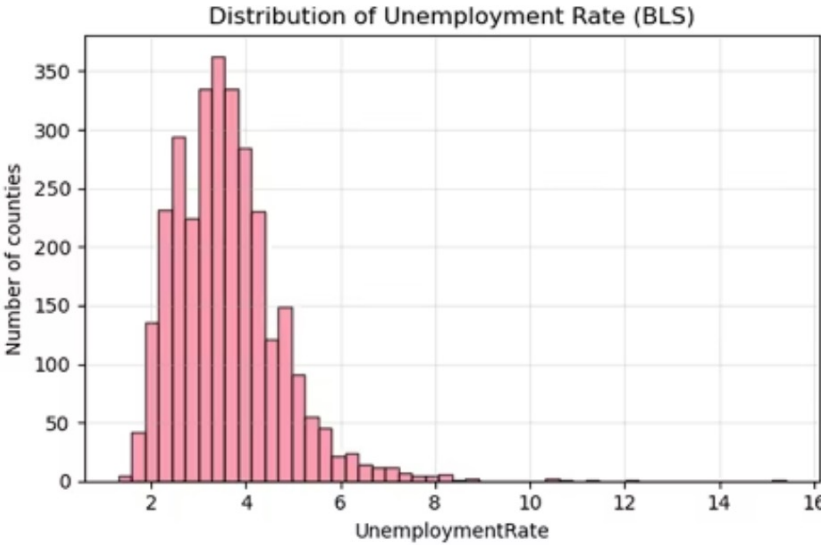
Mean Poverty Rate

Range: 1.7% – 43.2% showing stark disparities

3.6%

Mean Unemployment

Range: 1.3% – 15.4% across labour markets

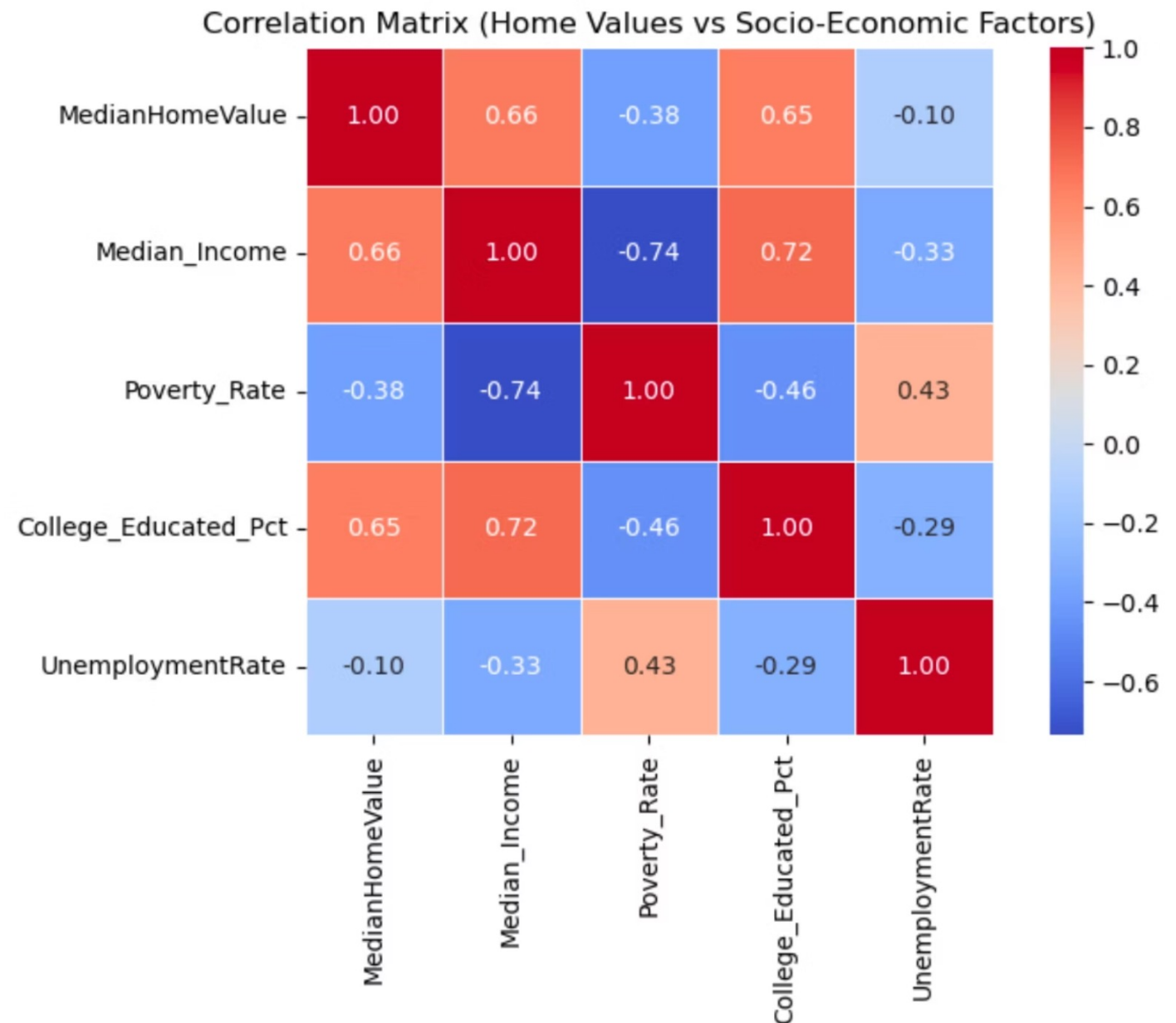


# Clear Relationships Between Housing and Socio-Economic Factors

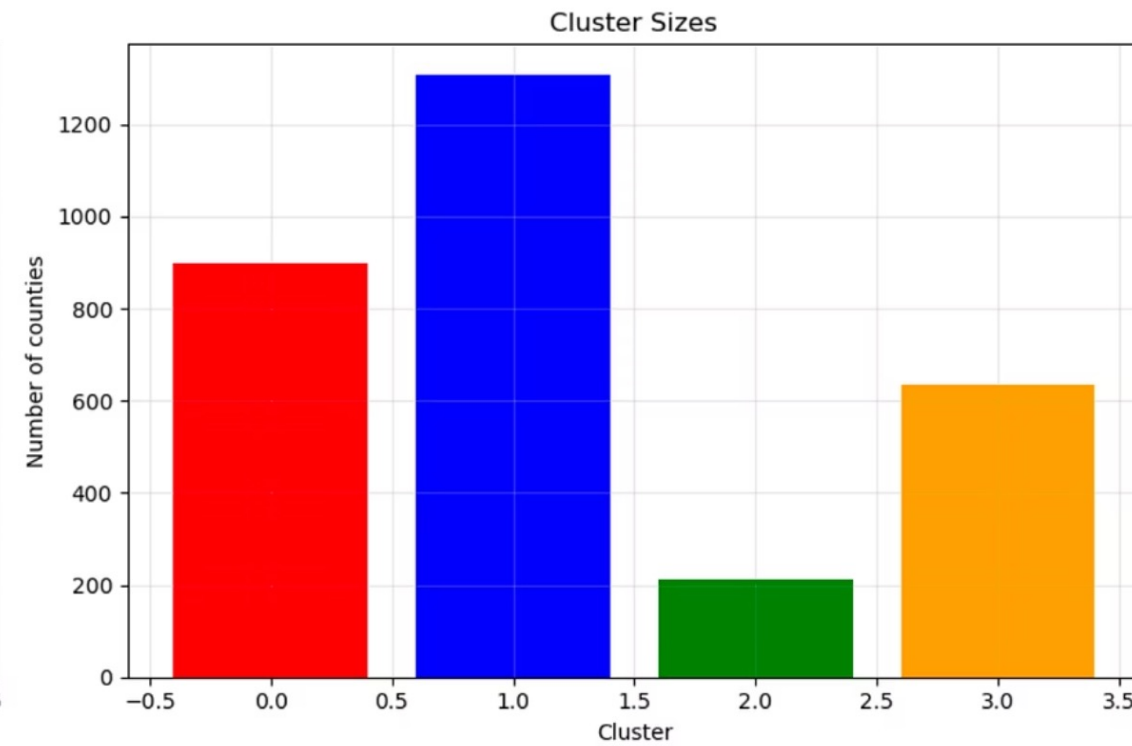
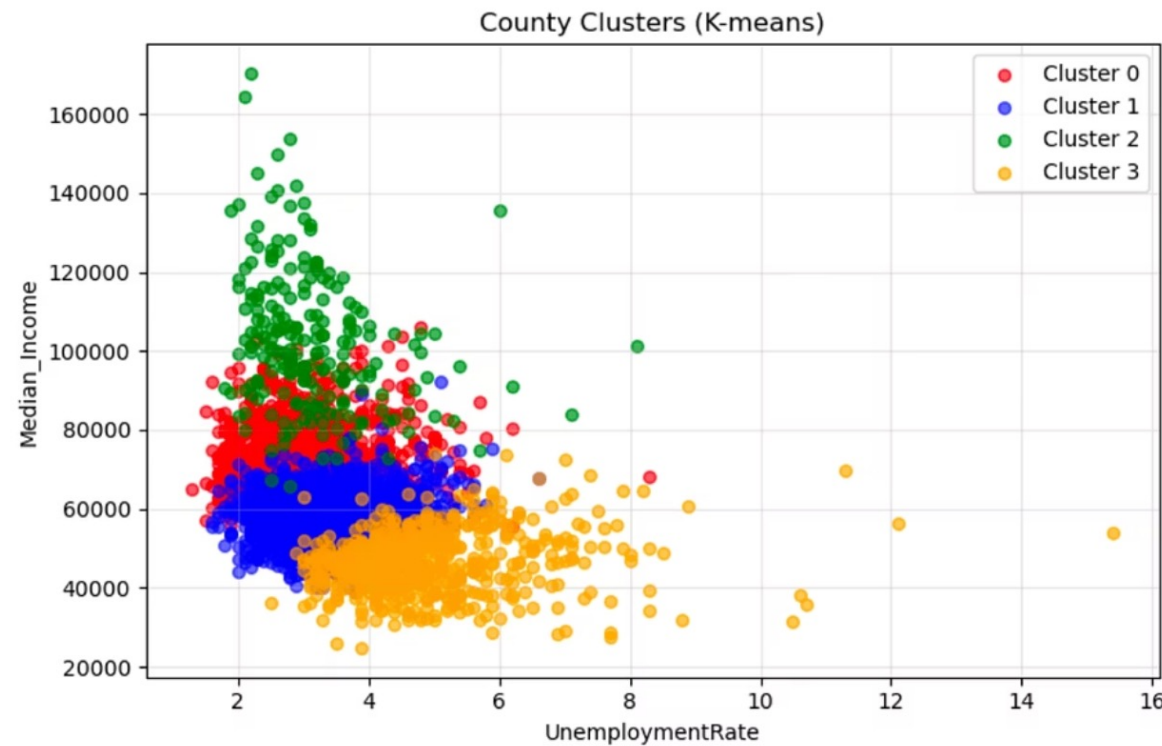
## Correlations with Median Home Value

- Median home value has a strong positive correlation with median household income ( $r \approx 0.66$ ) and the share of college-educated adults ( $r \approx 0.65$ ).
- Poverty rate shows a moderate negative correlation with home values ( $r \approx -0.38$ ).
- Unemployment rate has only a weak negative correlation ( $r \approx -0.10$ ).

Overall, richer and more educated counties tend to have much higher home values, while unemployment alone explains very little.

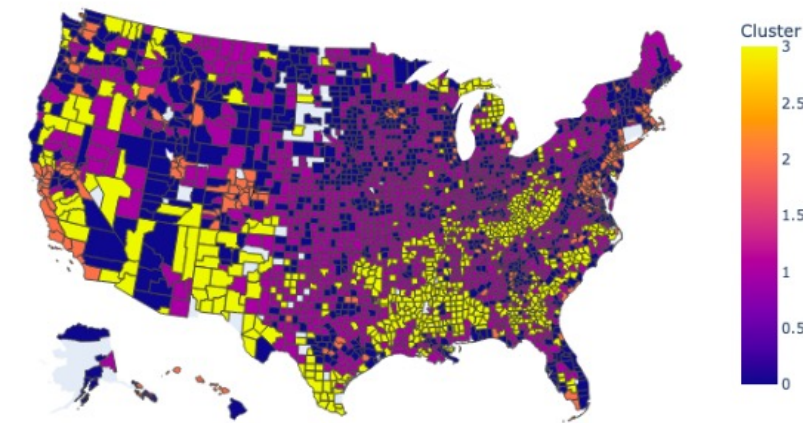


# Four Distinct Socio-Economic Profiles (K-means)

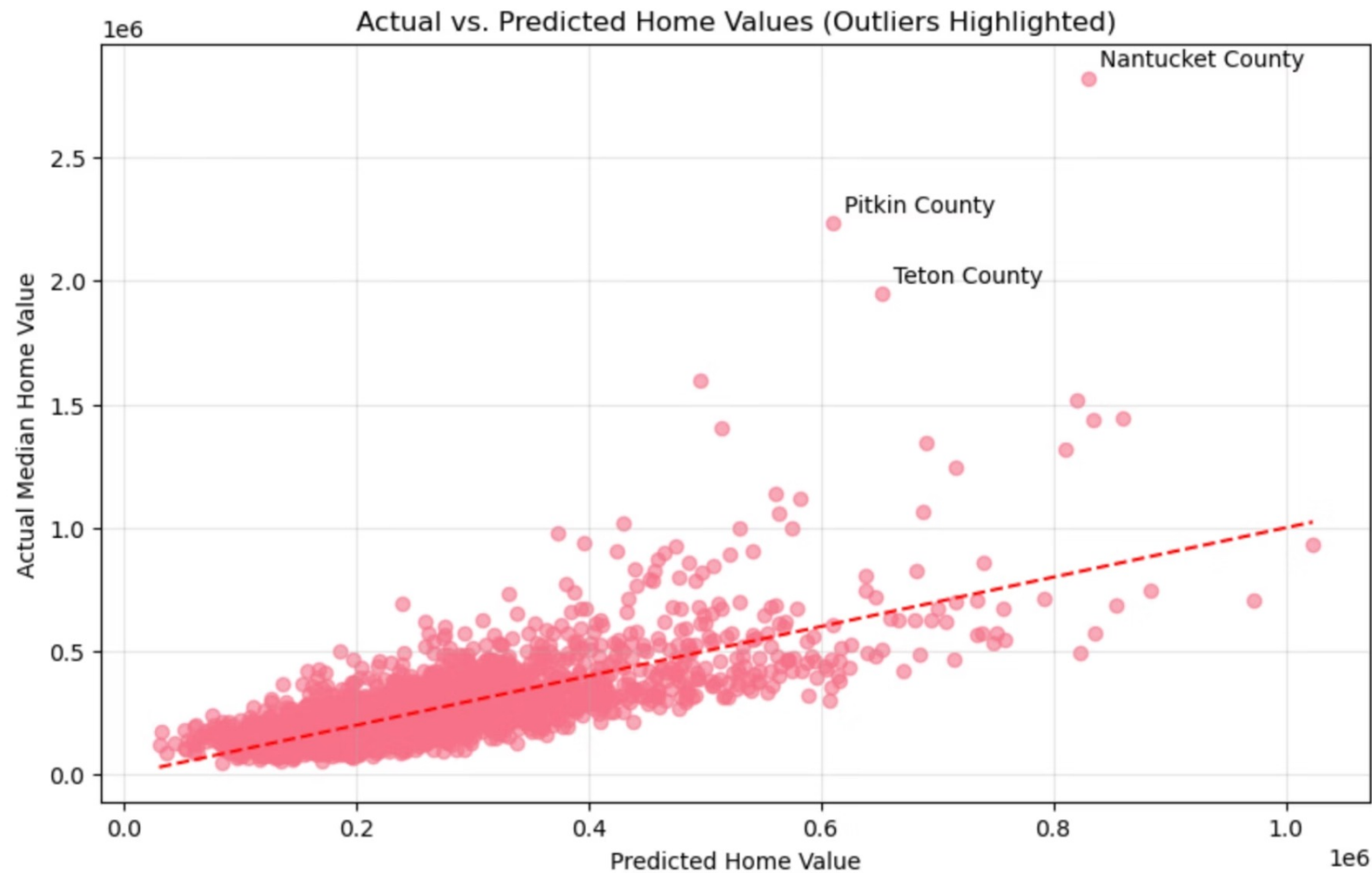


County Clusters: Housing & Socio-Economic Profiles

- **Cluster 0 - Prosperous Counties (898 counties):**
  - Above average income (\$72K), good home values (\$292K)
  - Moderate education (29% college+), low poverty (9.6%)
- **Cluster 1 - Average Counties (1,308 counties):**
  - Middle income (\$58K), moderate home values (\$179K)
  - Average education (19% college+), typical poverty (13.5%)
- **Cluster 2 - Affluent Counties (212 counties):**
  - High income (\$100K+), expensive homes (\$545K+)
  - Highly educated (45% college+), low poverty (7%)
- **Cluster 3 - Struggling Counties (634 counties):**
  - Low income (\$47K), affordable homes (\$150K)
  - Low education (15% college+), high poverty (20.6%)



# Market Exceptions and Key Insights



## Top Overpriced Counties:

- Nantucket, MA: +\$1.99M above predicted
- Pitkin, CO (Aspen): +\$1.62M above predicted
- Teton, WY: +\$1.29M above predicted

## Key Conclusions:

- Income and education are the strongest predictors of county-level home values.
- There is clear economic stratification in the US housing market, with four distinct county profiles.
- Some local markets, especially tourist and resort areas, follow “different rules” and are priced far above socio-economic fundamentals.
- Regional disparities in both housing affordability and socio-economic conditions are clearly visible.

# Challenges and Lessons Learnt

## Data Access and Cleaning

- *BLS county unemployment file required manual download due to 403 errors*
- *Careful FIPS code formatting and alignment across Zillow, Census, and BLS sources*
- *Substantial data engineering effort to merge disparate public datasets*

## Statistical Challenges

- *Managing highly skewed distributions, especially for home values*
- *Handling outliers whilst preserving meaningful variation*
- *Balancing statistical rigour with interpretability*

## Methodological Decisions

- *Selecting optimal number of clusters through elbow method and domain judgement*
- *Iterating on cluster interpretation to ensure meaningful, actionable insights*
- *Combining statistical analysis with careful qualitative assessment*

*This project demonstrated the importance of robust data engineering when working with real public data and the value of combining multiple analytical approaches to understand complex socio-economic relationships.*



Thank You!

Questions & Discussion