

Who is this data for?



IMMIGRANTS

For those to find desirable living areas based on personal preferences



ESTATE DEVELOPERS

To scout suitable areas for housing developments and population targets

Our target analyses?



Relation between certain housing factors



Top housing factors influencing PRICE



Current segregation of homes in Boston

Data Source



1978 BOSTON HOUSING DATASET

- Collected by the U.S. government
- 506 observations and 13 attributes
- Variable examples: capita crime rate, average number of rooms per dwelling, proportion of non-retail business acres per town
- 2 outliers were removed during the preliminary linear regression

Housing Recommendation

- Problem: It's hard for new-comers in Boston to pick the right neighborhood that suit their needs from countless towns and tracts
- Solution: A housing recommendation function

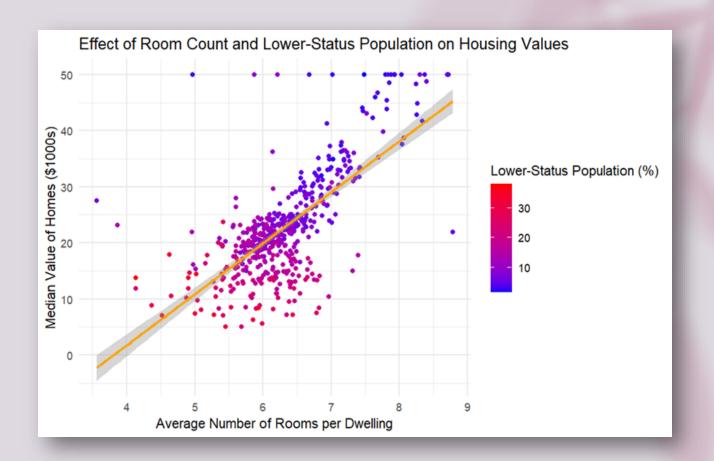
```
> # Example of customer input and recommendation
> # For a customer who has a budget of 30 and cares most about number of rooms, lower status population and crime rate:
> recommendations <- recommend_neighborhoods(</pre>
   budget = 30,
   number_of_rooms = 5.5,
   max_Lower_Status_Population = 5,
   max_Non_retail_Business_Acres = 25,
   max_Weighted_Distance_to_Employment = 13,
   max_Proportion_of_Old_Units = 90,
   max_Nitric_Oxides_Concentration = 0.7,
                                         But what about houses built in
   max_Property_Tax_Rate = 700,
   max_Pupil_Teacher_Ratio = 30,
   max_Per_Capita_Crime_Rate = 0.1,
                                         newly developed neighborhoods?
   min_Residential_Land_Proportion = 0,
   min_Index_of_Highway_Accessibility = 0
> print(recommendations)
           town tract cmedv
                              crim zn indus chas
                                                        rm age
                                                                   dis rad tax ptratio
                                                                                          b 1stat
          Sharon 4141 22.5 0.06466 70 2.24
                                              0 0.400 6.345 20.1 7.8278
                                                                        5 358
                                                                                 14.8 368.24
299
                                              0 0.398 6.290 17.8 6.6115
   North Reading 3301
                      23.5 0.03584 80 3.37
                                                                        4 337
                                                                                 16.1 396.90
          Nahant 2011 24.0 0.00632 18 2.31
                                              0 0.538 6.575 65.2 4.0900
                                                                        1 296
                                                                                15.3 396.90
293
        Westwood 4123 27.9 0.03615 80 4.95
                                              0 0.411 6.630 23.4 5.1167
                                                                        4 245
291
        Westwood 4121 28.5 0.03502 80 4.95
                                              0 0.411 6.861 27.9 5.1167
                                                                        4 245
                                                                                 19.2 396.90
300
          Sharon 4142 29.0 0.05561 70 2.24
                                              0 0.400 7.041 10.0 7.8278
                                                                        5 358
                                                                                14.8 371.58
         Bedford 3592 29.1 0.01439 60 2.93
                                              0 0.401 6.604 18.8 6.2196
195
                                                                        1 265
                                                                                 15.6 376.70 4.38
      Framingham 3839 29.6 0.08221 22 5.86
                                              0 0.431 6.957 6.8 8.9067
                                                                        7 330
                                                                                 19.1 386.09 3.53
```

Factors: Lower-Status Population, Average Number of Rooms Per Dwelling

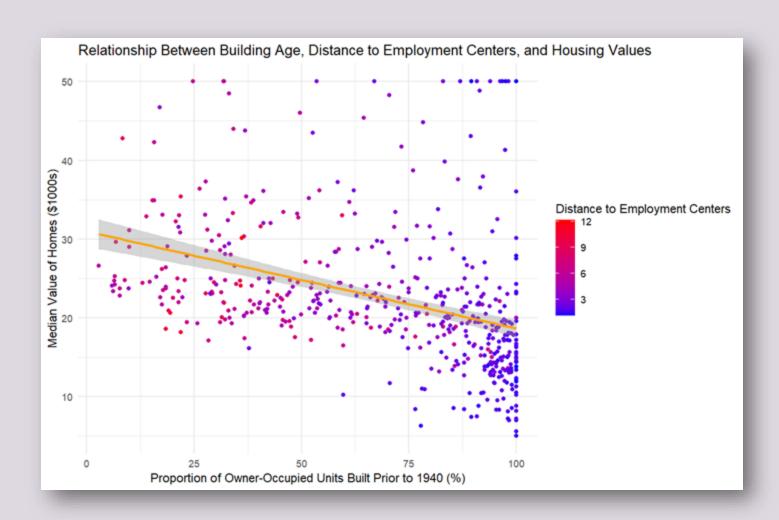
FOCUS : Socio-Economic Factors and Property Values

The results indicate that in the dataset, more valuable homes are typically found in neighbourhoods with fewer lower-status residents (Istat).

Additionally, these more valuable homes tend to have more rooms (nr). This reinforces the idea that home size and socio-economic conditions are factors influencing property values.



Factors: Building Age, Distance to employment Centers



FOCUS : Housing Building Condition and Location

Consider a neighborhood with a large number of older homes and proximity to a major employment hub. In this area, the older homes may be less attractive to buyers due to their age and potentially outdated features, leading to lower property values.

Furthermore, the close proximity to the busy employment center might contribute to a less desirable living environment, further reducing home values.

Factors: Residential Land Zoning, Non-retail Business Acres

FOCUS: Land Proportionality of Housing and Business

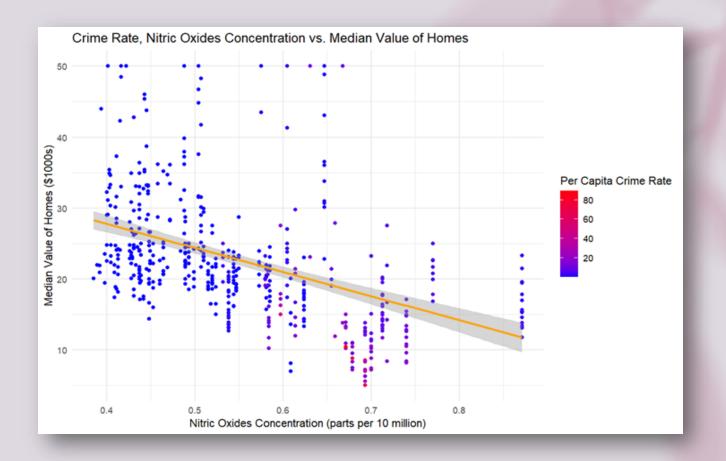
Imagine two neighborhoods: one with a high proportion of land zoned for large residential lots and few non-retail business areas, and another with smaller lot sizes and significant commercial or industrial activity. Typically, the former neighborhood is likely to have higher home values due to the spacious and quieter residential environment, making it more attractive to homebuyers looking for a desirable living area.



Factors: Crime Rate, Nitric Oxide Concentration

FOCUS: Physical and Mental Environment (crime rates and air pollution)

The analysis shows a positive relationship between environmental quality and safety with housing values. Specifically, neighborhoods with lower nitric oxides concentration and lower crime rates are associated with higher median home values. This indicates that homebuyers place a high value on both clean air and safety, and are willing to pay more for properties in such areas.



EDA - Testing Hypothesized Correlations (Linear Regression)

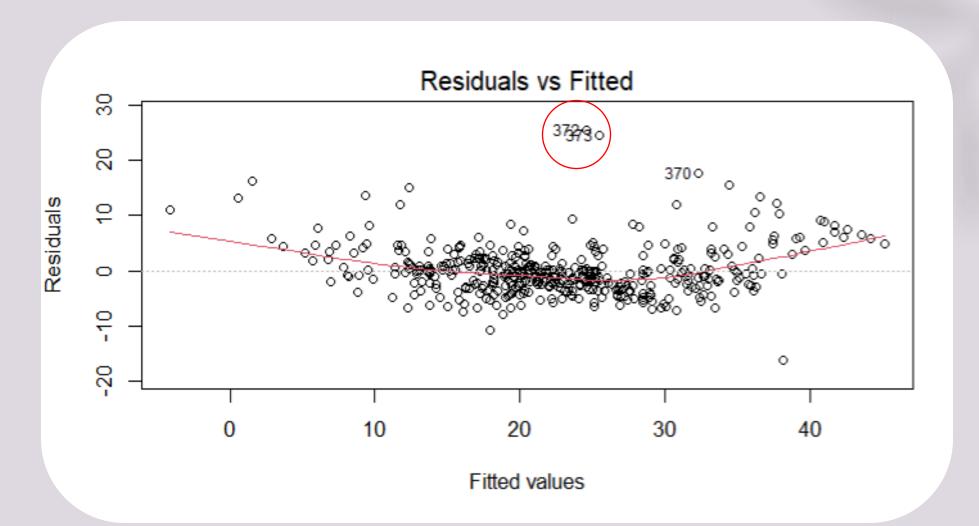
- Median Value Housing Price =
- -0.5068(lower status population)
- + 4.239(avg no. of rooms)
- -1.045(parent-teacher ratio)
- 1.607(distance to employment centers)

+ ...

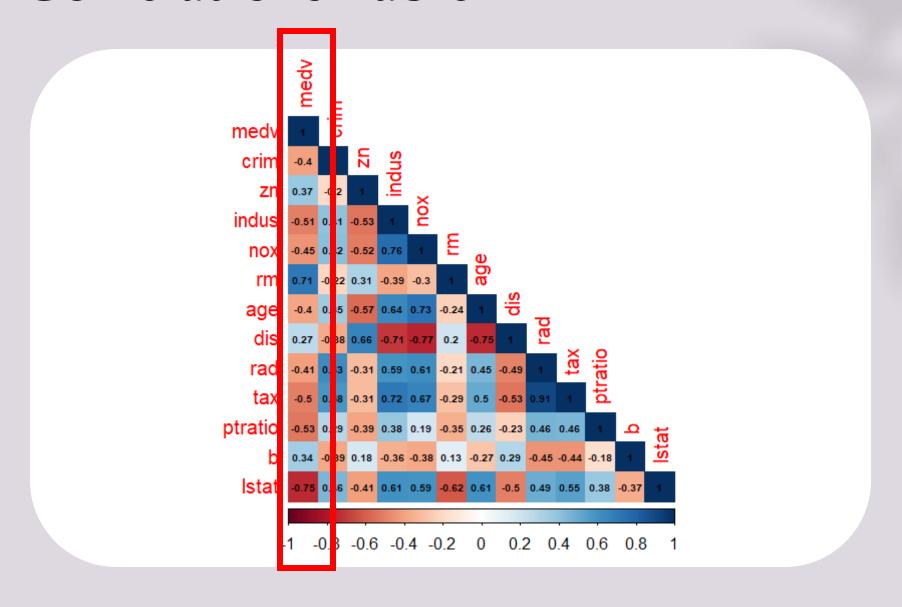
This model is able to explain 75.7% of the cases

Root-mean-squared-error: 4.9

EDA - Testing Hypothesized Correlations (Outliers)

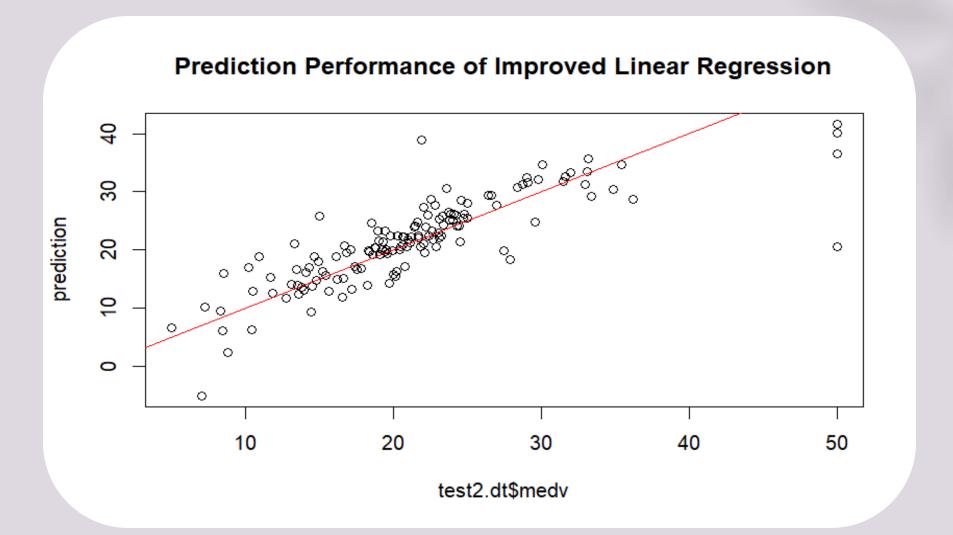


EDA - Correlations Table



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EDA - Testing Hypothesized Correlations (Improved Linear Regression)



EDA - Testing Hypothesized Correlations (Improved Linear Regression)

- Median Value Housing Price =
- -0.426(lower status population)
- + 4.92(avg no. of rooms)
- -0.892(parent-teacher ratio)

+ ...

This model is able to explain 79.2% of the cases

Root-mean-squared-error: 4.7

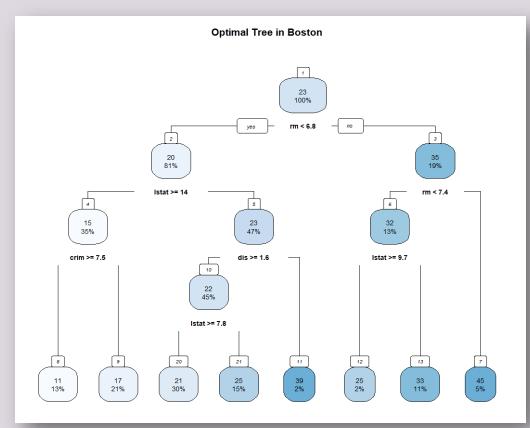
Most important variables in descending order:

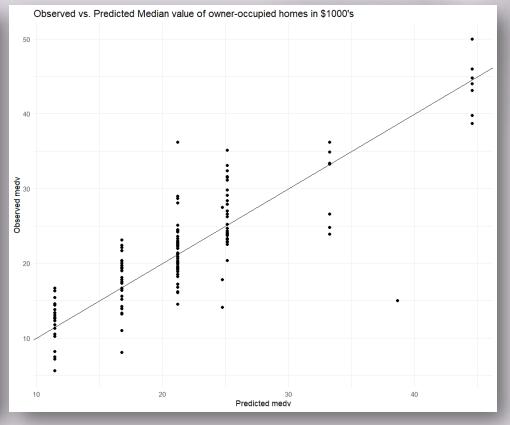
- 1. Lower Status Population
- 2. Average Number of Rooms per Dwelling
- 3. Parent Teacher Ratio

Housing Factors Analysis (CART)

City expansion and new houses

- Problem: New houses are being built in newly developed neighborhoods as Boston expands, and people don't know whether these houses can fulfill their preferences or if they are affordable.
- Solution: A CART model to predict housing prices



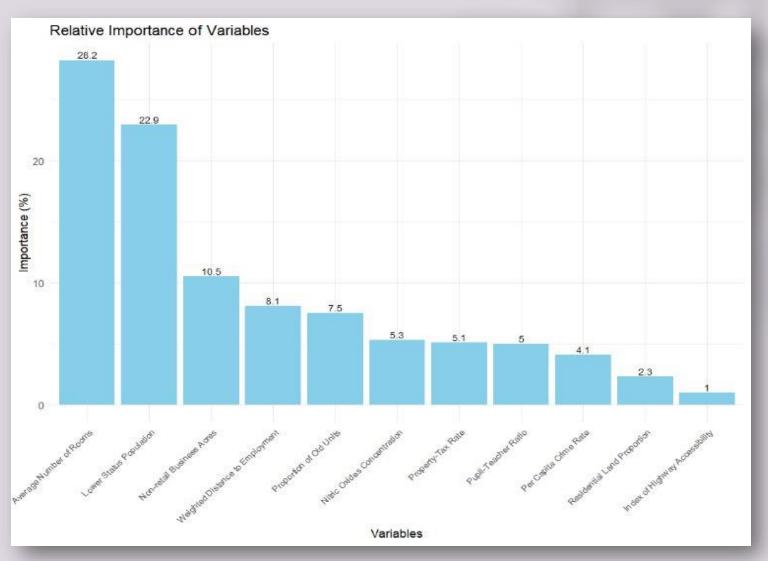


Housing Factors Analysis (CART)

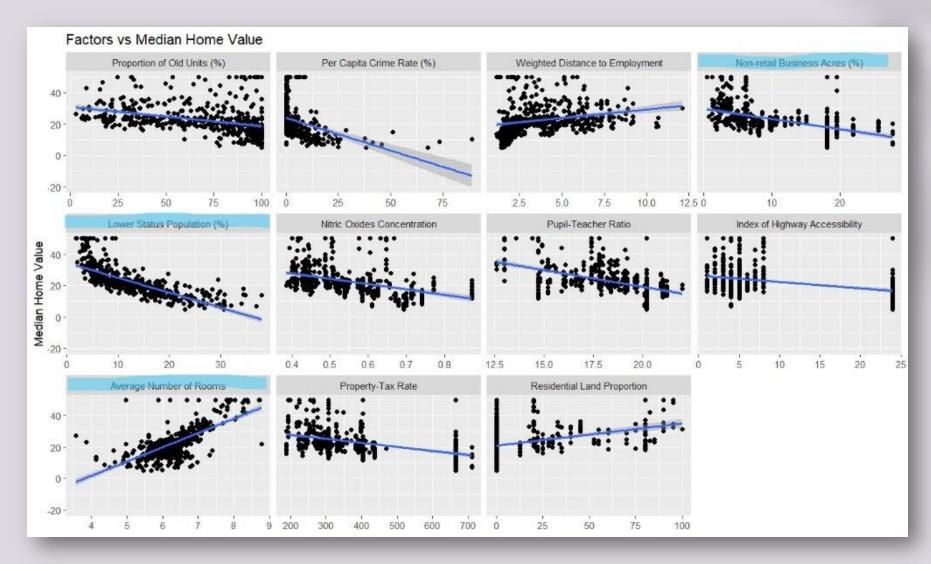
Our CART Model takes all variables and ranks their *importance* in predicting median price values of Boston houses.

Real estate developers can focus on the more significant variables when choosing locations to build new dwelling areas.

It also benefits house buyers as more houses will be built to cater their needs, instead of based on the judgement of real estate developers, which may not really capture what residents care most about.

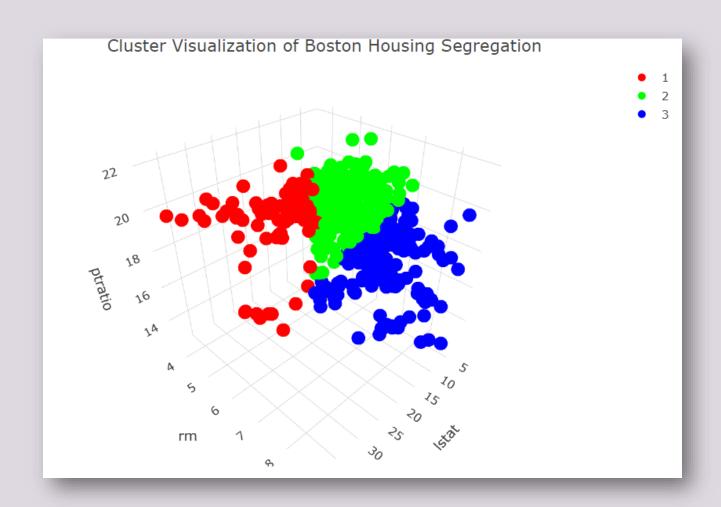


Housing Factors Analysis (CART)



Chen Zeqi

How are houses in Boston currently clustered?



Here's **how Boston housing is segregated into groups**, and the characteristics of each group.

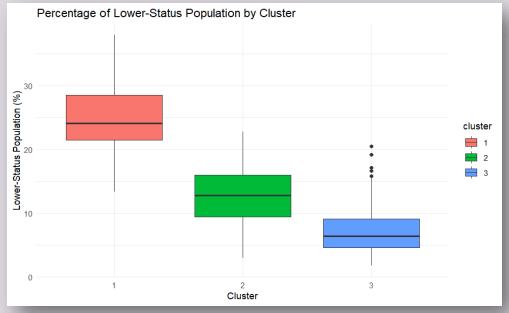
First, how many groups will exist? **THREE**.

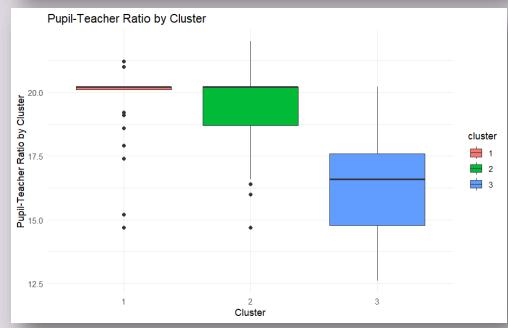
 K-Means Clustering with k=3 through Elbow Method

What will the groups be analyzed on?

TOP 3 PRIMARY HOUSING FACTORS.

- AVERAGE ROOMS PER DWELLING (rm)
- LOWER STATUS POPULATION (Istat)
- PUPIL TEACHER RATIO (ptratio)





Cluster Variable Analysis

Distinction between each housing clusters based on these factors :



Percentage of lower status population

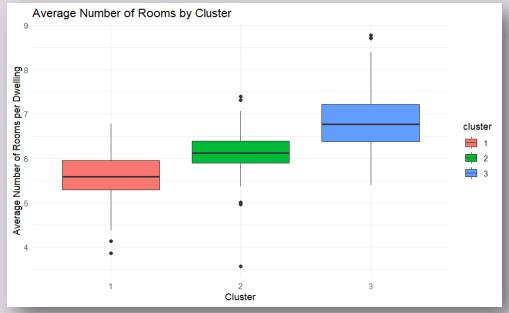
This factor influences the living environment and community.



Pupil-teacher ratio in schools

Small ratio is usually desirable as it relates to more focused learning per student and thus better education quality.

Sasha Annabel





Cluster Variable Analysis

Distinction between each housing clusters based on these factors :



Average number of rooms per dwelling

More rooms per dwelling often suggests bigger-sized (more accommodating) houses.



Median price of houses

The median price of houses per cluster is also calculated.

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Sasha Annabel

The meanings? 1978 Boston Housing clusters are as follows.

CLUSTER 1 Cheap Houses	CLUSTER 2 Moderately Priced	CLUSTER 3 Expensive Houses
Mostly are houses around	Houses located around	Houses belonging majority
lower-class communities	middle-class communities	to upper-class communities
with less # of schools (or	with decent # of schools (or mid	with more exclusive schools
major class sizes) and	class sizes) and	(or minor class sizes)
smaller houses.	mid-sized houses.	and bigger houses.

Conclusions

- Boston's housing data has been analyzed using Linear Regression, Cart and K-means Clustering
- Future buyers and investors are advised to make more informed choices based on
- Primary factors influencing house prices:
 - Average rooms per dwelling
 - Percentage of lower-status population
 - Pupil-teacher ratios in schools
 - Non-retail business in acres of land

Current grouping of Boston housing: