# Research on the Factors Contributing to Change in Housing Price in California

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Abstract: Housing price is always the popular topic in the current market. To predict the alternation in housing price, it

is necessary to initially forecast the trend of those influencing factors. This article aims to investigate the potential factors that influence alternation in housing prices. 20,640 samples from a dataset of California's housing group by 1990 will be conducted in this essay to complete further analysis. Then, this research will measure correlation coefficients and employ linear regression model based on the eight variables shown from the dataset. By searching for coefficient, unstandardized beta and variance inflation factor (VIF) value, this research can conclude the most significant causes to affect housing prices. Finally, the research indicates that income level is the main feature, other causes including housing ages, location as well as population can also affect prices in some aspects. The data and purpose outcomes can provide households and housing agencies

direction regarding the trend of housing market.

## 1 INTRODUCTION

Housing price is one of the most pivotal indicators in today's society, affecting the entire economy and people life. As housing acting as one of the fundamental properties for people to require maintaining normal life, it is the criteria to measure people's well-being and individual health, which means lots to human beings (Rolfe et al, 2020). Due to the extensive demand for shelters, property values are now a controversial topic among academics and the public. The importance of researching house prices is further emphasized by the fact that real estate is frequently seen as a worthwhile financial asset. Housing prices in the United States have increased rapidly and even dramatically over the last few decades. Nowadays, house prices have become one of the economies that employed adults care about the most. Its price trend determines when people will choose to buy or sell houses. Between 2020 and 2023, California recorded its first population decline since becoming a state in 1850 (Batdorf, 2024). This unprecedented demographic change because of multiple causes including high living expenses, increasing remote work flexibility, family considerations and potential economic challenges. shifts have highlighted the complex

relationship between changing population trends and housing affordability in the state.

While, forecasting the price trends and comprehending their wider economic impact require a grasp of the elements that affect house values. This study will examine and conclude home prices according to existing housing data from California. Besides, this research will also analyse the factors influencing change in housing prices.

The housing price is definitely determined by multiple features that surrounding housing itself, housing locations and the entire society. From the past, the theme has attracted by several researchers to explore into. Those existing research offers extensive theoretical models and empirical evidence examining the key factors influencing housing prices. In the research by Li and Yang, they employed a multiple linear programming tool to analyse the key features. By examining the linear relationship between a dependent variable and multiple independent variables, the model estimates regression coefficients, and minimizes the difference between the actual observed values and the values predicted by the model. By comparison of different variable combinations, the study aimed to identify the optimal model for accurate housing prices forecasting (Li and Yang, 2024).

Mao and Yao conducted a similar study using the King County House Sales dataset to investigate how geographic features impact housing price. In the research, they apply multiple linear regression (MLR) combined with 10-fold cross-validation to assess model performance. The findings highlight that factors including the number of bedrooms, latitude, and longitude significantly affect. Their predictive summary is both methodologically sound and interpretable, offering detailed views into the relationship between these variables and property values (Mao and Yao, 2020). Additionally, Lau analysed how population alternation homeownership rates impact California housing prices from 2020-2022 potentially using county-level data. This study applied time-series analysis and correlation theory to measure relationships between these variables, with results indicating positive, negative, or neutral associations (Lau, 2024).

Yu applied the same linear regression analysis, the common statistic method in order to predict the target variable by linear combination of characteristic variables. Evaluation of these models through different matrixes such as mean square error is the strategy she used to determine the accuracy of model, which is indeed understandable and precise (Yu, 2024). Similarly, Yan also conducted excel to build up a multi-regression model, aiming to research on influencing factors of housing value in New York in a comprehensive aspect (Yan, 2022). There are some other researchers adopt different strategies to figure out those potential causes. Zoppi et al. attempted to employ Hedonic models to analyze environmental and structural features which affects housing market in Cagliari (Zoppi et al, 2015). Huang et al. conducted models to study and collect those data of unusual variation in housing prices (Huang et al., 2010). In summary, this study will properly conduct the combination of two strategies, correlation module and linear regression, in order to investigate the potential features that influencing change in housing prices in California.

#### 2 METHODS

#### 2.1 Data Source

In this research, a dataset focuses on housing price from Kaggle website will be adopted. This dataset offers a practical starting point for exploring machine learning techniques. It covers housing information from the 1990 California census, measuring various homes across different districts. This dataset includes region-level statistics such as population, median income and housing characteristics, providing an understandable and clear background for building and evaluating machine learning models. From the original dataset, there were several missing values in the total\_bedrooms column. To deal with this issue, the missing entries were occupied using the median value. From the full set inputs, a random sample of 20640 records was selected for this research. Then, the updated dataset now includes nine features-longitude, latitude, housing median age, total rooms, total bedrooms, population, households, median income and ocean proximity, along with the target variable, median house value.

## 2.2 Variables Explanation

In order to present the further model, the study needs to initially list out all the dependent variables and the independent variable that investigating in the dataset. These variables cover longitude, latitude, housing age, total rooms, total bedrooms, population, households, income level, and the house value. Therefore, the research lists out a relevant variables table, as shown in Table 1.

Variables	Description
Longitude	House distance to the west
Latitude	House distance to the north
Housing Age	Median age of a house
Total Rooms	Total room number
Total Bedrooms	Total bedroom number
Population	Total number of residents living in a house
Households	Total household number
Income	Income level for households
House Value	Median House price

Table 1: Variables introduction

#### 2.3 Method Introduction

In this analysis, both correlation and linear regression methods are used to investigate the potential factors influencing housing prices in California. The initial step is data cleaning, referring to handling missing values, such as replacing null numbers in total bedrooms with the median calculation as well as dropping irrelevant columns like ocean proximity, since it doesn't change throughout the whole dataset.

The correlation between median house value and variables including longitude, latitude, housing median age, total rooms, total bedrooms, population, households, and median income is calculated, helping determine which variables have relatively stronger relationships with the house price.

After identifying those relevant variables through correlation approach, this paper builds a multiple linear regression model based on these indicators. This model not only quantifies the influence of each variable but also eliminate errors, for example two variables are highly correlated, which may distort the model accuracy. This research uses statistical tests including p-values and R<sup>2</sup> scores in order to evaluate the significance and explanatory power of the model.

In this study, the combination of correlation and regression model provides a clearer understanding of how those different features relate to housing prices in California.

## 3 RESULTS AND DISCUSSION

# 3.1 Correlation Analysis

To investigate the relationship between housing value and those potential influencing factors, this paper attempts to conduct Pearson correlation analysis, referring to a calculation measuring linear relationship between two continuous variables. The correlation coefficient is a result of ratio between the covariance of two variables and the product of their standard deviations (Asuero et al., 2007). Those

coefficients always are ranging from -1 to 1, a positive value represents a direct relationship while a negative value implies an inverse relationship.

After the implement of regression correlation model, the standard deviation and mean of each variable are figured out. Plugging those data into the formula mentioned before, then this paper obtains the correlation coefficient between each pair of two variables among those listed elements. Figure 1 shows the all the combinations; it indicates the indexes of linear relationship between those factors directly. From the figure 1, it is intuitive to determine which pairs of variables maintain a strong correlation.

The diagram presents a strong relationship between households and total bedrooms (0.98), total rooms and total bedrooms (0.93), as well as total rooms and population (0.92). These suggest that families with more households require greater amount of rooms and bedrooms, which is obvious.

Although using all the combination is direct to see each variable, this study should focus more on the factors influencing the housing value. Therefore, this paper wipes off the other variables being target variables in order to receive an updated correlation diagram.



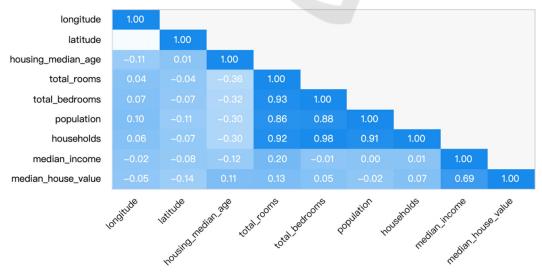
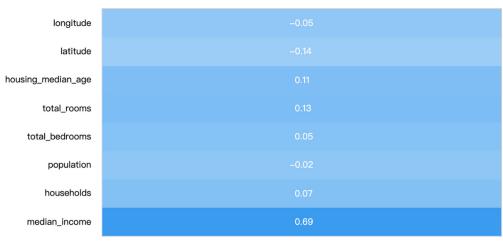


Figure 1: Pearson Coefficient Table for Variables (Picture credit: Original).

#### Pearson Coefficient



median\_house\_value

Figure 2: Pearson Coefficient Table for Median Housing Value (Picture credit: Original).

From Figure 2, it is much more direct to witness the potential causes to affect change in housing value. Looking at the chart, the strongest relationship is between income and house value. The correlation between median income coefficient median house value reaches to 0.69, which is a relatively strong positive figure. That is probably because households with higher income will prefer houses in a higher price to satisfy their requirement of higher living standard. Besides, wealthy people always pursue shelters with superb location and greater size, these factors will also impact a higher housing price. Therefore, it's reasonable that income level is the most strongly and positively linked factor to housing value.

Other factors like the number of rooms in a house (total rooms) and the age of the house (housing median age) maintain weaker positive relationships with housing prices. Sometimes housing age cannot be a determination of housing prices as new houses built in current years can be sold in a high price because of its brand-new infrastructure and novel layout. Consumers are likely to be attracted by those factors, thus larger consumer groups increase its demand, causing increase in price. However, aged homes can also be set in a high price. Those homes are often found in well-established neighborhoods and hold a larger land size because land was cheaper and more accessible at that time. Therefore, these historical houses may also be popular among consumers. Housing value, as well as can be properly impacted by the room amounts, while it is reasonable that room sizes should take into account when it comes to the number.

Some variables including total bedrooms, population as well as the households don't seem to affect house prices significantly, they maintain correlation coefficients very close to zero. This suggests that numbers of households and bedrooms cannot be the determinations that influencing housing value

Finally, location, especially latitude, seems to matter relatively. Latitude and longitude have small negative correlations with house prices. This may indicate that those houses located relatively inland or farther north in California tend to become slightly cheaper. As those homes are located far away from main cities or the coast, the transportation could be more inconvenient and time-consuming, therefore demand on those houses will reduce, causing decline in housing prices of homes.

#### 3.2 Linear Model Results

After calculation of correlation coefficients, the linear regression model should be conducted for the further analysis and conclusion. This is a model that estimates the relationship between dependent variables and independent variables. As seen from table 2, those eight dependent variables for linear regression have established a model combined with value of beta of each variable to calculate the housing value.

	В	Std. Error	Beta	t	P value	VIF	Tolerance	
Constant	-3585395.747	62900.543	-	-57.001	0.000	-	-	
Longitude	-42730.120	717.087	-0.742	-59.588	0.000	8.714	0.115	
Latitude	-42509.737	676.952	-0.787	-62.796	0.000	8.829	0.113	
Housing median age	1157.900	43.389	0.126	26.687	0.000	1.260	0.794	
Rooms number	-8.250	0.794	-0.156	-10.387	0.000	12.717	0.079	
Bedroom number	113.821	6.931	0.415	16.423	0.000	36.004	0.028	
Population	-38.386	1.084	-0.377	-35.407	0.000	6.371	0.157	
Household	47.701	7.547	0.158	6.321	0.000	35.136	0.028	
Income	40297.522	337.207	0.663	119.504	0.000	1.732	0.578	
R-squared	0.637							
Adjust R- squared	0.637							
F Test	F(8, 20424) = 4478.347, p=0.000							
D-W	0.975							

Table 2: Linear Regression Table

Then, R- squared value and adjusted R-squared value are both 0.637, implying that these several variables can entirely bring about 63.7% of the change in median house price, which is powerful result to display.

From the calculation and the table, it is obvious that median income keeps the most significant positive implications on house value with a standardized Beta of 0.663 and a significant unstandardized coefficient of 40297.52. This represents that if other variables constant, an increase in household income will probably bring a dramatic increase in housing price. This is also consistent with the idea that wealthy areas tend to develop more expensive housing markets and wealthy consumers.

In addition, the location in terms of latitude and longitude, both possess strong negative effects on house value, which standardized betas are -0.787 and -0.742 respectively. Housing median age has a positive relationship with housing value, with beta of 1157.90.

While it is quite surprising that total rooms display a negative coefficient, which B equals to -8.25 and  $\beta$  equals to -0.156, but the total bedrooms show a positive one ( $\beta$  = 0.415). This suggests multicollinearity and an overlapping influence. Indeed, their Variance Inflation Factors (VIF) are both high, showing 12.717 for total rooms and 36.004 for total bedrooms, which indicates that there may be a risk of collinearity and potential distortion in terms of estimates.

## 4 CONCLUSION

This research randomly selected 20640 samples from California Census dataset collected by the US Census Bureau in 1990. By using correlation and regression analysis as combination of strategies, this study investigates this potential feature causing alternation in housing value.

Through the analysis, it is obvious that households' income level is always the most influential feature regarding the housing value through both size of correlation coefficient and regression model. Besides, from further regression analysis, it is known that factors of location and housing age will cause limited effect on housing value, and other variables such as room numbers and households don't affect significantly.

This outcome provides a relatively clear direction regarding potential factors which influencing housing prices, which is useful for households, real estate agencies as well as the government to determine future trend on housing markets. Households ought to select appropriate housings based on their own wealth level and personal purchasing budget. However, the conclusion still cannot be the accurate result about this issue, as the sample size covers only 20000 approximately and the time for research is relatively outdated, which may affect the conclusion accuracy and reality.

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