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The Hedonic Pricing Model Applied to the Housing Market

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Abstract:

Purpose: The article applies the hedonic pricing model to estimate house price in the housing market of Vietnam, which is a country with a fledgeling housing market, so the study is expected to bring interesting findings.

Design/Methodology/Approach: By applying the hedonic pricing model, most of the previous studies have reported that house price is significantly influenced by the characteristics of the house itself, its location and surrounding facilities. Based on this, the article adopts the Ordinary Least Squares (OLS) regression in combination with robustness statistics in the model estimation, so the estimated results on house price are reliable and able to be widely applied. Data are collected through the survey into housing projects in Ho Chi Minh and Ha Noi city, which are the two largest cities in Vietnam.

Findings: The findings reveal that the hedonic pricing model can be applied to estimate house price in Vietnam's housing market. This can be said to be a big success in giving first empirical evidence in Vietnam on this matter. Specifically, house price is negatively affected by its proximity to the city center. Also, factors including house size, house type, house structure, number of bedrooms, amenities around the house exert a positive influence on the price.

Practical Implications: The estimated results are typical and reliable, which can be applied universally.

Originality/Value: The study confirms the hedonic pricing model can be well-applied to accurately estimate the price of houses in Vietnam. More than that, the findings are also essential for other countries, especially those with a nascent housing market like Vietnam.

Keywords: Hedonic pricing model, house price, housing market, real estate, Vietnam.

JEL: C53, E37, R32.

Paper Type: Research article.

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1. Introduction

To make an accurate prediction on house price is really important to prospective homeowners, investors, appraisers, tax assessors and other housing market participants (Frew and Jud, 2003). Traditional house price prediction is based on cost and the relationship between demand and supply. This usually leads to the house price is much higher than its real value, even the appearance of bubbles in the housing market (Bui, 2019). Therefore, it is essential to identify determinants of house price, thereby making a precise prediction and improving the efficiency of the market. Further, operations in the housing market are performing fast, efficiently, which are beneficial for the participants.

In Vietnam, housing demand is highly increasing, especially in metropolitan areas (Nguyen *et al.*, 2019; Bui, 2020a; Nguyen and Do, 2020). This is because a great number of inter-provincial migrants tend to move to big cities (Bui, 2020b) while the availability of land is limited. On the other hand, most Vietnamese wish to possess their own house, instead of renting one. More than that, handful of investors have participated in the housing market with the expectation of a high profit from this market. This gives a considerable increase in house price, possibly occurring real estate bubbles (Bui, 2019). In fact, changes in house price in Vietnam's housing market are unpredictable. This can be clearly described through house price index in Ho Chi Minh and Ha Noi city, the two biggest metropolises of Vietnam, which is quarterly published by Savills Vietnam (Figure 1).

Indeed, after the global financial crisis, Vietnam's housing market gained significant growth in the 2009-2011 period, proving that the post-crisis policies aimed to recover the economy have achieved positive benefits in the housing market. However, right after that, in order to recover the economy as well as minimize the risks of crisis reoccurrence in the real estate market, Vietnam started to perform tight monetary policies toward real estate industry, making the housing market face lots of difficulties and considerably fell in 2013. Recently, the housing market has recovered and gradually increased. Also, transactions in the housing market have been proceeded more smoothly.



Figure 1. House price index in Ho Chi Minh and Ha Noi city

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Constant fluctuations in the housing market confuse and worry its participants (Schulz and Werwatz, 2004; Jeon, 2018; Nguyen and Bui, 2019). They also make the managers hesitate in giving suitable management policies (Nguyen *et al.*, 2020). Therefore, to determine influential factors of house price is necessary. It provides the participants a reliable base for an accurate prediction on house price. This research problem has been examined in some empirical studies, wherein the hedonic pricing model has been frequently applied in predicting the house price.

Nevertheless, these studies have their drawbacks of being unable to reach an agreement on the determinants of house price. Also, there exist lots of difficulties in collecting data on traded houses. Therefore, most of them can only obtain small-scale data which cannot fully represent for the entire market, leading to unreliable results which are unable to be used widely. This is a big gap in the current literature.

Moreover, most of them have been conducted in developed countries with a lack of empirical studies in countries with nascent housing market. By this study, the article expects to contribute to the current literature with an accurate estimation on the house price in Vietnam through applying the hedonic pricing model. The data are obtained from the survey of housing projects which were successfully traded in Ho Chi Minh and Ha Noi city, the two biggest cities of Vietnam. In Vietnam, a majority of trading in the housing market has mainly been conducted in metropolises, so the dataset utilized is expected to be highly representative. Particularly, by collecting the data in the recent period, the study is expected to accurately reflect the actual developments in the housing market as well as reveal unprecedented findings.

2. Literature Review and Research Hypothesis

Housing market is defined as the market for properties being sold, purchased, rented or differently traded, thereby forming prices (Bui, 2020b). Thus, it is traditionally measured by house price. In fact, this problem has been empirically investigated in developed countries while few empirical studies have been conducted in places with nascent housing market. The application of the hedonic pricing model has been widely done in a big number of previous studies. Specifically, house price is usually determined by internal characteristics, its location and surrounding facilities. In order to estimate the price of home by applying the hedonic pricing model, obtaining data on traded houses is a prerequisite.

One of the advantages is that hedonic pricing uses regression to estimate house price, which can overcome drawbacks of traditional valuation methods. For example, it can minimize the big subjective influence of appraisers on house value like other methods. Furthermore, the hedonic pricing model can help reduce estimation cost if being widely applied (Peterson and Flanagan, 2009; McCluskey *et al.*, 2013). As a result, this application is frequently a good choice in house price valuation (Peterson and Flanagan, 2009; Lin and Mohan, 2011; Zurada *et al.*, 2011; Antipov and Pokryshevskaya, 2012; McCluskey *et al.*, 2012; Renigier-Biłozor *et al.*,

2019). This model is even used to test the estimation results of other traditional models (Nghiep and Al, 2001; Friesen *et al.*, 2011; Kilpatrick, 2011; Kontrimas and Verikas, 2011). With lots of outstanding advantages, the hedonic pricing model has been applied in the U.S since 1920 (Abidoye and Chan, 2017). Nowadays, it has been utilized in many developed countries.

However, most empirical studies using this model have their drawbacks when unable to collect adequate dataset which can represent the entire market. It is because there still exist many difficulties in data collection of traded housing projects, particularly in a country with a young housing market. This leads to a big number of errors in the application of estimated results suggested in the previous studies. Consequently, if this matter can be well-controlled, the results of the model application are more reliable, which can be practically applied widely. Based on the current literature, it can be recognized that the price of a house is frequently affected by house size, house type, house structure, number of bedrooms, number of bathrooms, the proximity to the city center, included amenities (such as hot-tub, swimming pool) and local amenities (such as parks, squares, schools, shopping malls), which are comprised in this study.

2.1 The Impact of House Size on House Price

A house in big size is frequently more preferred, particularly by crowded families. It is because big houses are able to provide larger and more comfortable space. Some families can even utilize their house for business or offices. Hence, houses in big size are more valuable (Kain and Quigley, 1970; Selim, 2009; Amenyah and Fletcher, 2013; Shimizu, 2014; Yayar and Demir, 2014; Ndegwa, 2018). Based on this, the article suggests the research hypothesis as follows:

*H*₁: *House size* (SIZE) *has a positive impact on house price* (*PRICE*).

2.2 The Impact of the Number of Bedrooms on House Price

Houses with more bedrooms are likely more expensive because they meet the demand of big families. These houses can provide enough private space for each family member. Further, they are more convenient for their friends or relatives to stay over. Therefore, the number of bedrooms positively affects the house's value. This result is also in line with what have been found by Kain and Quigley (1970), Cebula (2009), Selim (2009), Baek *et al.* (2019). Hence, the research hypothesis is suggested as follows:

*H*₂: *The number of bedrooms (BEDROOMS) has a positive impact on house price (PRICE).*

2.3 The Impact of the Number of Bathrooms on House Price

Kain and Quigley (1970), Cebula (2009) reported that the number of bathrooms are positively correlated to house price. Indeed, the number of bathrooms meets basic needs of households. Although bathrooms are considered as a must to be present in a house, this factor has received much attention in many empirical research. Therefore, the research hypothesis is proposed as follows:

H₃: *The number of bathrooms (BATHROOMS) has a positive impact on house price (PRICE).*

2.4 The Impact of House Location on House Price

A house which is close to the center area of a city or a town tends to be a preferential choice to buyers. This facilitates the commute to the downtown area or work. This also allows families enjoy more facilities and services nearby. In addition, houses close to the center are easily sold or rented at a steep price. These houses may thus sell for a premium. Alternatively, houses with a further distance from the center may sell for less. This influence corroborates with the findings of Selim (2009), Aluko (2011), Amenyah and Fletcher (2013), Yayar and Demir (2014), Ndegwa (2018). Based on this, the article proposes the research hypothesis as follows:

H₄: The proximity to the center (CENTER) has a negative impact on house price (PRICE).

2.5 The Impact of House Type on House Price

Type of a house is a big consideration of house buyers. In fact, houses with a garden can help create green space and significantly improve the living standard. These houses, thus, are more preferred and priced (Limsombunchai *et al.*, 2004). Therefore, the research hypothesis is suggested as follows:

H₅: *Type of house (the house has a garden) (TYPE) has a positive impact on house price (PRICE).*

2.6 The Impact of House Amenities on House Price

A house is more valued with lots of amenities including pools, hot-tubs (Laurice and Bhattacharya, 2005; Sirmans *et al.*, 2005; Cebula, 2009; Yayar and Demir, 2014). The presence of swimming pools or hot-tubs obviously satisfies basic needs and improves a family's standard of living. Therefore, the research hypothesis is suggested as follows:

*H*₆: *House amenities (POOL) have a positive impact on house price (PRICE).*

2.7 The Impact of Local Amenities on House Price

Local amenities significantly influence the price of a house. Houses close to parks, squares, schools or shopping malls facilitate households' entertainment, study or shopping. Housing projects, thus, tend to pay more attention on their public amenities. Those houses are more preferred and expensive (Cebula, 2009; Selim, 2009; Amenyah and Fletcher, 2013; Ndegwa, 2018). Therefore, the research hypothesis is proposed as follows:

*H*₇: Local amenities (PARK) have a positive impact on house price (PRICE).

2.8 The Impact of House Structure on House Price

A house with a new and modern structure is easily sold at a higher value. Indeed, beside their living demands, buyers are looking forward to owning modern newbuild houses. Thus, these houses are highly priced (Cebula, 2009). Hence, the article suggests the following research hypothesis:

*H*₈: *House structure (STRUCTURE) has a positive impact on house price (PRICE).*

No	Hypothesis	
H_1	$SIZE \Longrightarrow PRICE$	(+)
H_2	$BEDROOMS \Longrightarrow PRICE$	(+)
H ₃	$BATHROOMS \implies PRICE$	(+)
H_4	$CENTER \Longrightarrow PRICE$	(-)
H ₅	$TYPE \Longrightarrow PRICE$	(+)
H ₆	$POOL \Rightarrow PRICE$	(+)
H_7	$PARK \Longrightarrow PRICE$	(+)
H ₈	$STRUCTURE \Rightarrow PRICE$	(+)

Table 1. Research Hypotheses

Source: Own study.

3. Methodology

In this study, the article adopts the hedonic pricing model to estimate the house price in Vietnam. The article collects survey data of 205 houses which were successfully traded in Ho Chi Minh and Ha Noi city in the first six months of 2019. By the collection of data survey, the study is expected to accurately reflect the actual developments in housing market as well as reveal unprecedented findings.

Regarding the analysis, the Ordinary Least Squares (OLS) regression is employed to estimate the research model. If the hypotheses are violated, the article would adopt other suitable quantitative analyses to control, such as robustness statistics, following what was suggested by White (1980).

Based on the hedonic pricing model, the previous results and the above-developed hypotheses, the research model is estimated as follows:

 $\begin{array}{l} PRICE = \beta_{0} + \beta_{1} \; SIZE + \beta_{2} \; BEDROOMS + \beta_{3} \; BATHROOMS + \beta_{4} \; CENTER + \beta_{5} \\ TYPE + \beta_{6} \; POOL + \beta_{7} \; PARK + \beta_{8} \; STRUCTURE + \epsilon \end{array}$

Where: *Dependent variable* (PRICE): House price (price per square meter).

Independent variables: SIZE: House size (square meters); BEDROOMS: Number of bedrooms; BATHROOMS: Number of bathrooms; CENTER: Proximity to the city center (kilometers); TYPE: A binary variable indicating whether the house has a garden (= 1 if yes, = 0 otherwise); POOL: A binary variable indicating whether the house has its amenities such as a hot-tub and/or a swimming pool (= 1 if yes, = 0 otherwise); PARK: A binary variable indicating whether the house has local amenities such as parks, squares, schools or shopping malls (= 1 if yes, = 0 otherwise); STRUCTURE: A binary variable indicating whether the structure is new and modern (= 1 if yes, = 0 otherwise); ε : Error term; β_i : Coefficients; β_0 : Constant.

4. Empirical Results

First, the article analyses the correlation among the variables. The result is shown in Table 2 as follows:

	PRICE	SIZE	BEDR OOMS	BATH ROOM S	CENTE R	ТҮРЕ	POOL	PARK	STRUC TURE
PRICE	1.000								
SIZE	0.128	1.000							
BEDROOMS	0.776	0.146	1.000						
BATHROOMS	0.546	0.167	0.656	1.000					
CENTER	-0.590	0.010	-0.473	-0.369	1.000				
TYPE	0.595	-0.046	0.429	0.315	-0.300	1.000			
POOL	0.011	0.017	0.025	0.005	-0.037	-0.002	1.000		
PARK	0.776	0.082	0.471	0.334	-0.413	0.615	-0.044	1.000	
STRUCTURE	0.712	-0.092	0.536	0.370	-0.422	0.435	0.017	0.661	1.000

Table 2. Variable correlations

Source: Own study.

Table 2 reveals that the distance from the house to the center (CENTER) is negatively correlated to its price (PRICE) while the other independent variables are positively associated with the house price (PRICE). This is in line with what the article has been suggested previously.

Multicollinearity	y test		
Variable	VIF	Heteroscedasticity test	
PARK	2.52		
BEDROOMS	2.36	Chi2(42) = 89.01	
STRUCTURE	2.18	$Prob>chi2 = 0.000^{***}$	
BATHROOMS	1.79		

Table 3. Results of tests on multicollinearity, heteroscedasticity

TYPE	1.73		
CENTER	1.41		
SIZE	1.14		
POOL	1.01		
Mean VIF = 1.77			

Note: *** indicates significance at the 1% level. *Source:* Own study.

The model is considered to experience no serious issues of multicollinearity (Mean VIF = 1.77). However, heteroscedasticity really exists at the 1% level of significance (Prob>chi2 = 0.000) (Table 3). This means that the estimation using the Ordinary Least Squares regression (OLS) does not provide a reliable result. Thus, robustness statistics is utilized for the estimation, following what have been suggested by White (1980) and recently performed by Bui (2020a).

Dependent variable: House price (PRICE)				
Variables	Coef.	P > z		
Constant	17.093***	0.000		
SIZE	0.001**	0.023		
BEDROOMS	0.065***	0.000		
BATHROOMS	0.003	0.640		
CENTER	-0.020***	0.000		
ТҮРЕ	0.048**	0.036		
POOL	0.002	0.709		
PARK	0.187***	0.000		
STRUCTURE	0.050^{***}	0.000		
Number of obs	205			
R-squared	85.59%			
Significance level	F(8, 196) = 276. Prob>F = 0.000	F(8, 196) = 276.08 $Prob>F = 0.000^{***}$		

 Table 4. Estimation results

Note: ** and *** indicate significance at the 5% and 1% level, respectively. *Source:* own study.

As can be seen from Table 4, the estimation results are significant at the 1% level (Prob>F = 0.000). Specifically, the independent variables can account for 85.59% variation in house price. It can be concluded that the model is appropriate and valid for the practical application. Alternatively, the hedonic pricing model is successfully applied in estimating house price in Vietnam. Furthermore, with a high explanation level of research model and highly-representative dataset, the results can be widely administered.

Particularly, CENTER exerts the negative impact ($\beta = -0.020$) on PRICE at 1% level of significance whereas PRICE is positively influenced by SIZE ($\beta = 0.001$, at 5% level of significance), BEDROOMS ($\beta = 0.065$, at 1% level of significance), TYPE ($\beta = 0.048$, at 5% level of significance), PARK ($\beta = 0.187$, a 1% level of significance), STRUCTURE ($\beta = 0.050$, at 1% level of significance). Nevertheless,

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the study has not found the significant influence of BATHROOMS and POOL on PRICE. Hence, the estimated results of the model take the following equation:

PRICE = 17.093 + 0.001 SIZE + 0.065 BEDROOMS - 0.020 CENTER + 0.048 TYPE + 0.187 PARK + 0.050 STRUCTURE + ϵ

No	Hypothesis	Results
H_1	$SIZE \Longrightarrow PRICE$	Accepted
H_2	$BEDROOMS \Longrightarrow PRICE$	Accepted
H_3	$BATHROOMS \Longrightarrow PRICE$	Rejected
H_4	$CENTER \Longrightarrow PRICE$	Accepted
H ₅	$TYPE \Longrightarrow PRICE$	Accepted
H ₆	$POOL \Rightarrow PRICE$	Rejected
H_7	$PARK \Rightarrow PRICE$	Accepted
H_8	$STRUCTURE \Rightarrow PRICE$	Accepted

 Table 5. Results of hypothesis testing

Source: Own study.

- The impact of house size (SIZE) on house price (PRICE): The results show that SIZE positively affects PRICE. This indicates that houses in bigger size are usually priced more. This finding corroborates the reality and what has been reported by Kain and Quigley (1970), Selim (2009), Amenyah and Fletcher (2013), Shimizu (2014), Yayar and Demir (2014), Ndegwa (2018). Accordingly, big-size houses can meet demand of crowded families, bringing a wide and comfortable space. These houses can also specially be used for their business or offices.

- The impact of the number of bedrooms (BEDROOMS) on house price (PRICE): The results show that BEDROOMS positively affects PRICE. This is consistent with the findings of Kain and Quigley (1970), Cebula (2009), Selim (2009), Baek *et al.* (2019). Accordingly, houses with more bedrooms suit families with many members. These houses allow their guests to stay over. As a result, they are more expensive.

- The impact of house location (CENTER) on house price (PRICE): The study supports the negative influence of CENTER on PRICE. This shows that houses far from the center may sell for less than those with a close proximity to the center. This is because people living in the former find it difficult to commute to work or town center. Also, their public amenities are more limited than the latter. This is in line with the results in the studies of Selim (2009), Aluko (2011), Amenyah and Fletcher (2013), Yayar and Demir (2014), Ndegwa (2018).

- The impact of house type (TYPE) on house price (PRICE): The findings reveal that PRICE is positively affected by TYPE. In the metropolises with a high population, the presence of garden in a house can help improve people's living standard. Consequently, this kind of house is higher priced. This finding is consistent with what has been stated by Limsombunchai *et al.* (2004).

- The impact of local amenities (PARK) on house price (PRICE): The study establishes the positive influence of PARK on PRICE. Houses which are located near parks, squares, schools or shopping malls are more preferred because of their convenience for entertainment, study or shopping. They are, thus, sold at a high value. This corroborates with the findings of Cebula (2009), Selim (2009), Amenyah and Fletcher (2013), Ndegwa (2018).

- The impact of house structure (STRUCTURE) on house price (PRICE): The study shows that PRICE is positively influenced by STRUCTURE. This result has also been found in the analysis of Cebula (2009). Houses with new and modern structure are obviously more preferred and higher priced.

5. Conclusions

By this study, the author greatly succeeded in giving first empirical evidence on applying the hedonic pricing model to estimate house value in Vietnam. In addition, the results can be practically applied widely. The results reveal that the proximity to the center negatively affects the house price. Meanwhile, houses with big size, modern new structure, more bedrooms, gardens and local amenities (like squares, parks, schools, shopping malls) are more valuable. Based on these results, housing market participants can make an accurate prediction on housing prices. Indeed, based on the results, the management in the housing market can have a comprehensive perspective on this market, especially the factors affecting house price. This is essential for them to come up with suitable solutions boosting an efficient and sustainable housing market. More than that, the results enable the management to successfully develop a highly-representative house price index in Vietnam. The results can be widely applied by appraisers at a low cost while the investors can introduce high-valued houses which meet customer's demands. To distributors, it facilitates the improvement of distribution channels, so their projects can be introduced quickly and effectively to customers. The results also allow buyers to determine the real value of a house easily. In addition, these findings are valuable reference for other countries, especially those with a nascent housing market like Vietnam.

Despite achieving its objective, the study still has its own limitation when not analyzing other determinants which may affect the house price such as macroeconomic factors, investor's prestige and housing distribution channel's performance. These may be interesting proposals for future studies.

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