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# **Home financing loans and their relationship to real estate bubble: An analysis of the U.S. mortgage market**

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## **Abstract**

It is well reported that the much fluctuations in Real Estate (RE) markets globally are a result of unequal risk burden caused by deficiencies of financial system and speculative nature of those markets. However, the evidence is mixed in terms of whether high house prices lead to over financing of mortgages or vice versa. This work attempted to resolve this issue by using ARDL approach handy to use for time series data for the United States. The empirical results of the work have some noteworthy theoretical and policy implications. In the short run, house prices seem to be causing an increase in money supply and influencing house financing decisions. However in the long run, availability of new mortgage loans and interest impact seem to dominate. Thus, the implication from the study suggests that we have to consider both the short and the long run aspects of policy and its influence on all sectors of the economy while making such policy decisions.

**Keywords:** Mortgage loans, Real Estate bubble and Real Estate cycle.

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# **Home financing loans and their relationship to real estate bubble: An analysis of the U.S. mortgage market**

## **1. Introduction**

In his famous article 'On the efficiency of the financial system' more than 30 years ago James Tobin suggested that the financial system is becoming detached from the production of goods and services and involving more of activities that generate high private rewards which are disproportionate to their social productivity resulting in negative sum game (Tobin (1984, p. 14-15). As we can see from the aftermath of 2007 Subprime Mortgage Crisis of the U.S. this argument of Tobin is even more appropriate for the present day.

Being only second to Great Depression, subprime crisis of 2007 forced us to learn many lessons. One of those important lessons learned tells us that if methods of financing are disconnected from real economic factors it could lead to disastrous results. This crisis clearly showed it in example of home financing which got disconnected from its fundamentals through use of securitization and derivatives. Additionally, there are many other reasons critic argues that lead to this crisis which quickly swept all around the world and made millions of people much poorer in matter of few months. However, one common link that joins those reasons is detachment of finance from real economic factors, including financing and pricing of house apart from its fundamentals.

As can be observed in many countries cyclical movement of real estate (RE) prices is creating problems in economy by forming RE bubbles in some period and consequent bursting of such bubbles that could lead to prolonged recessions. Among many reasons, two important reasons for formation RE bubble could be related to low interest rates and easy to obtain loans (Taipalus, 2006; Mirakhor & Krichene, 2009). There is also theory that suggests that very high price-rent ratio for housing suggests existence of bubble (Taipalus, 2006). In fact when Subprime crisis was at verge of emerging this ratio stood at its peaks.

The above analyses suggest that the banks are acting as opportunistic profiteer to take opportunity of financing irrespective of current RE market situation. In fact it could also be argued that excessive home financing loans advanced mainly by banks for purchase of houses during economic boom periods could be one of main reasons for formation of RE bubbles. Those arguments could be considered as a good research question that needs to be scrutinized with use of proper data and research methodologies available.

## 2. Literature review

In many countries cyclical movement of Real Estate (RE) prices is creating problems in economy by forming RE bubbles in some period and consequent bursting of such bubbles that could result to recession. Taipalus (2006) point out three reasons why such bubbles maybe harmful for economy. The first one is related to the asset prices ability to signal for future inflation growth and to the overall measurement of inflation. Another reason is the linkage between asset prices and their impact to overall stability of the financial system and threat banking crises. The linkage between house prices and banking crises seems to be largely dependent on the way how banks value the collateral and how collateral appreciation affects their balance-sheets. The third reason why regulators need to pay attention to housing prices is the reason related to the overall economic development, especially due to the resource-allocating effects, wealth effects etc.

Of course concerning the wealth effect, the strength of the effect is very much dependent on whether the house price gains are perceived to be permanent or temporary. Another important point is the liquidity of the housing financing system, while it affects how well the households can take advantage of the possible capital gains in house prices (Zhu, 2005). Concerning the economic performance, Helbling & Terrones (2003) found in their research that house price busts are associated with output losses twice as large as equity bubbles. Large crashes with far reaching effects in the real estate prices are more likely to happen when the prices have been severely mispriced (i.e. in existence of bubble). In this respect sound developments in the real estate markets would be crucial.

Taipalus (2006) argue that among many reasons, one important reason for formation RE bubble could be related to low interest rates in respective countries. From her study of prices for houses both in Europe and the U.S., she found that main driver of high prices after 2000s in those markets was high demand caused by relatively low rates of interests. Another variable that could be considered among core variables in determining house prices is the availability of easy credit. Subprime lending for house purchases could lead to housing bubble as well. In fact, such easy to get housing loans are considered as main reason behind housing bubble in the U.S. which led to subsequent crisis of 2007 (Mirakhor & Krichene, 2009).

As defined by Kamil et al. (2010, p. 387) subprime lending is a general term that refers to the practice of making loans to borrowers who do not qualify for market interest rates because of problem with their credit history. As mentioned earlier, one of the reasons for the subprime crisis of 2007 was banks unscrupulous lending to unqualified borrowers. This is because housing loans are granted to the customers with low credit score, bad history of monthly repayments, little or no assets, poor income earning prospects and excessive debt from multiple banks (Johnson & Neave, 2008; Mirakhor and Krichene, 2009). Furthermore, when

prime rate was used as a benchmark for lending and were raised substantially<sup>3</sup> by the U.S. Federal Reserve Bank, it caused hardship and difficulties to the customers in making the monthly instalments (Johnson & Neave, 2008).

One way to approach the fundamental value in the real estate markets would be to examine the rent-price ratios in the markets. As known, in a sense rent-price ratios can be comparable to the dividend-yields in the stock markets as dividends and rents both represent the underlying capital component (i.e. uncertain future capital flows associated to the asset). In the financial literature the asset's fundamental value always equals the sum of its future payoffs, each being discounted back to its present value by investors using rates that reflect their preferences (see for example Krainer and Wei, 2004). In the stock markets this relationship is between discounted dividends and stock prices, but in the housing market this relationship could be thought to exist between rents and house prices. This theory suggests that high price-rent ratio of homes could indicate existence of bubble in the real estate market (Taipalus, 2006).

Krainer's and Wei's (2004) describe it as follows: "The fundamental value of a house is the present value of the future housing service flows that it provides to the marginal buyer. In a well-functioning market, the value of the housing service flow should be approximated by the rental value of the house." This meaning that the price of the house should be approximately equal to the discounted future flow of rents (if it would be rented). When using the rent-price ratios, the concept of RE bubble would become easier to define: the developments in the house prices or rents should not differ too much from each other, while otherwise this would mean that bubble would appear in the housing markets (Taipalus, 2006).

Bursting of housing bubble of 2007 subprime crisis in the U.S., transmitted by the domino effect to other developed countries, has resulted to long lasting of economic recession all around the world. From analysis of relevant literature we could see that the rates of interest and rigidity credit approval could result to housing bubbles and significant fluctuations in house prices.

### **3. Methodology**

#### **Theoretical Model and Data:**

The pricing process in the real estate markets is regarded as a relatively complex process where expectations as well as real economic variables together form the final market price. Among the core variables, which are seen to affect the pricing of the real estate are the

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<sup>3</sup> Prime rate was raised from its historical low of 4% in June of 2003 to as high as 8.25% in June of 2006 (i.e. more than double in less than two years). It was as high as 7.75% when early signs of subprime crisis emerged in October 2007. (Source: <http://www.fedprimerate.com>).

following: household incomes, interest rates, supply (especially so in the short-run), financial market institutions, demographic variables, availability of credit, taxes, public policies directed to housing etc. (see for example ECB 2003, Lamont and Stein 1999, Tsatsaronis and Zhu 2004 and IMF's WEO 2004). However, our goal is not to verify the core variables, but rather to check which variable has largest impact on house prices.

To test the main hypothesis, which has to do with identifying whether easy availability of mortgage loans and low interest rates could be considered among the main sources of housing bubbles we are proposing use model similar to Taipalus (2006).

Model will be tested using time-series regression analysis for similar set of data obtained by the author. Data can be obtained from secondary sources such as OECD main economic indicators, Bureau of Labour Statistics, Office of Federal Housing Enterprise Oversight (OFHEO) House Price Index (HPI) and the U.S. Federal Reserve Bank. The data used in this research will include price indices, housing credit, mortgage rates, national income as well as rent indices and other measures needed for analysis of the U.S. housing market.

Taipalus (2006) mentioned that there exist many weaknesses related to the usage and construction of prices and rent indices. Solely related to the price-indices we could list the following weaknesses: the underlying data comes from various sources and the statistics are compiled in various ways, the houses are heterogeneous assets and their qualities vary and in addition there are short-term fluctuations (seasonality etc.). That does not necessarily reflect any long-term changes in house price trends and the differences in statistics between different regions could be significant. Indeed, some of the variable may be not directly comparable between different states or cities in the U.S. But, assuming those differences are not major we can still use aggregate data for the entire country as representative of the general situation in all of the states.

Using same data sources used in the first objective we are going to simulate following model of house prices proposed by Égert & Mihaljek (2007). Model assumes some independent variables including demand and supply of houses to endogenous and solves for general equilibrium.

On the demand side, key factors are typically taken to be expected change in house prices ( $P^H$ ), household income ( $Y$ ), the real rate on housing loans ( $r$ ), financial wealth ( $WE$ ), demographic and labour market factors ( $D$ ), the expected rate of return on housing ( $e$ ) and a vector of other demand shifters ( $X$ ). The latter may include proxies for the location, age and state of housing, or institutional factors that facilitate or hinder households' access to the housing market, such as financial innovation on the mortgage and housing loan markets:

$$D^H = f(\overset{-}{P^H}, \overset{+}{Y}, \overset{-}{r}, \overset{+}{WE}, \overset{+/-}{D}, \overset{+}{e}, \overset{-}{X}) \quad (3.1)$$

The supply of housing is usually described as a positive function of the profitability of the construction business, which is in turn taken to depend positively on house prices and negatively on the real costs of construction (C), including the price of land ( $P^L$ ), wages of construction workers (W) and material costs (M):

$$S^H = f(P^H, C(P^L, W, M)) \quad (3.2)$$

Assuming that the housing market is in equilibrium, with demand equal to supply at all times, house prices could be expressed by the following reduced-form equation:

$$P^H = f(Y, r, WE, D, e, \bar{X}, C(P^L, W, M)) \quad (3.3)$$

The view that both the supply and demand for housing interact to determine an equilibrium level for real house prices should not be taken to imply that house prices are necessarily stable. Authors argue that in many countries it is frequently observed that house prices are significantly more volatile than would be predicted by the variation in the main determinants of supply and demand alone.

Some consider various factors that influence price ( $P_t$ ) of the house (Hott & Monnin, 2006). Among others Hott & Monnin (2006) include, mortgage rate ( $m_t$ ), a fraction ( $\phi$ ) of the house price that represents the sum of maintenance cost as, property tax and risk premium; expected capital gain which also takes in to account constant physical depreciation ( $1 - \delta$ ) which will be written as  $[E_t(P_{t+1}) - P_t]$ ; they also considered the tax influence but because of its insignificance ignored it. Thus their resulting imputed rent function ( $H_t$ ) was as follows:

$$H_t = m_t P_t + \phi P_t - (\delta E_t(P_{t+1}) - P_t) \quad (3.5)$$

However, since mortgage rate is difficult to obtain authors suggested using long-term interest rates because of its strong correlation with mortgage rate.

In Davis & Zhu (2011) we can see following equations as main determinants of housing price (Davis & Zhu (2011), p. 5-6):

$$D_t \equiv \frac{N[1 - F(P_t)]L(Y_t, i_t, P_t, \omega_t)}{P_t}, L_Y > 0, L_i < 0, L_p > 0 \quad (3.6)$$

$$K_t = (1 - \delta)K_{t-1} + I_{t-1} \quad (3.7)$$

$$I_{t-1} = \alpha \cdot B_{t-1}(Y_{t-1}, i_{t-1}, P_{t-1}, \omega_{t-1}), B_Y > 0, B_i < 0, B_p > 0 \quad (3.8)$$

$$D_t = K_t \quad (3.9)$$

Eq. (3.7) is the market demand function, which depends on the number of optimistic buyers who are willing to purchase housing at the current market price and the borrowing capacity for each of them. Eq. (3.8) is the adjustment function of the stock of market supply of buildings K, in which  $\delta$  is the depreciation rate and  $I_{t-1}$  is the completed new construction (which was started one period earlier). Most importantly, it increases in current property prices for the reasons mentioned above. Eq. (3.9) is the market-clearing condition at each

period. As we can see from above equations there are quite a few factors which play role in determination of house prices.

Ibrahim & Law (2014) looked at the long run behavior of house prices and their dynamic interactions with bank credits, real output and interest rate for the case of Malaysia. In general, they observe relation among the aggregate house prices and bank credits found to be significant and change in them has significant impacts on short-run fluctuation of the output. Most importantly they found bank credits to have a positively and the long run impact on house prices (Ibrahim & Law (2014), pp. 117-120).

Some other authors included both short and long term interest rates, as well as equity prices as determinant of house prices (Hirata et al., 2012). However, we know that house price do not only depend of financing factor but on general price level as well. Especially, house supply depends on cost of production. Thus cost of inputs used for production purposes will be critical for determination of house price. In addition we can suggest that even for demand general price level could be influential simply because as consumers tend to compare the price of house to other relative especially expansive purchases they make. Thus inclusion the general price level could be beneficial as well. But there may be strong correlation between it and real bank credit, thus we have to be careful in our analysis. However some studies have included them as well (Goodhart & Hofmann, 2008). Goodhart & Hofmann added CPI and Money supply to equation as well.

Thus we suggest to include some of the other variables to model suggested by Davis & Zhu (2011). Besides our focus variable which is index of housing price (HPI), measure of national income such as nominal GDP<sup>4</sup> (GDP), new bank loans for mortgages (NML), prime interest rates (PR), money supply (MS), started housing construction (HS) and expenditure on those constructions (CE).

Thus suggested model is as follows.

$$\text{HPI} = f(\text{GDP}, \text{NML}, \text{PR}, \text{MS}, \text{HS}, \text{CE}). \quad (3.10)$$

Each of those variables was approximated with a proxy most representative variable from available set of data for United States. Details of the used data as below:

HPI – Standard and Poor's Case-Shiller national house price index of U.S. was used as proxy

GDP – Nominal GDP of United States provide by IMF - International Financial Statistics

NML – Amount of new mortgage loans provided by Federal Housing Finance Board

PR – Quarterly average of Prime rates charges by Bank provided by U.S. Federal Reserve.

MS – Represented by Broad Money (M3) also provide by U.S. Federal Reserve.

HS – No. of new private housing units started provided by U.S. Census Bureau

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<sup>4</sup> We suggest to use nominal GDP rather than real GDP to cover for impact of price level at the same time.

CE – Total construction expenditure also provide by U.S. Census Bureau

Allow the variables were given in current U.S. dollar values and quarterly data was used. Period covered was from the first quarter of 1987 till the first quarter of 2014, which included 109 observations. The focus variable was in index form where the first quarter of 2010 was set equal to 100, one variable (PR) was in percentage returns in U.S. dollars, also one was measured in numbers (HS), and the rest were in current U.S. dollars. All of the variables except of HPI and PR were seasonally adjusted.

#### 4. Empirical outcomes

We decided to use Autoregressive Distributed Lags (ARDL) method for measurement of cointegration between those variable. The ARDL approach to cointegration involves estimating the unrestricted error correction model version of the ARDL model for U.S. house prices is given below:

LHPI, LGDP, LNML, PR, LMS, LHS, LCE

All of the variables were logged to make the stationary on variance except for PR which was not logged because of being rate variable. The first differences of those variables above were then taken to get following difference error correction model.

$$dLHPI_t = a_0 \sum_{i=1}^4 b_i dLHPI_{t-i} + \sum_{i=0}^4 c_i dLGDP_{t-i} + \sum_{i=0}^4 c_i dLNML_{t-i} + \sum_{i=0}^4 c_i dPR_{t-i} + \sum_{i=0}^4 c_i dLMS_{t-i} + \sum_{i=0}^4 c_i dLHS_{t-i} + \sum_{i=0}^4 c_i dLCE_{t-i} + \mu_i \quad (4.1)$$

Hypothesis tested are as follows:

$H_0: \delta_1 = \delta_2 = \delta_3 = \delta_4 = \delta_5 = \delta_6 = \delta_7 = 0$  Non existence of the long - run relationship

$H_0: \delta_1 \neq \delta_2 \neq \delta_3 \neq \delta_4 \neq \delta_5 \neq \delta_6 \neq \delta_7 \neq 0$  Existence of a long - run relationship

We begin our empirical testing by determining the stationarity of the variables used. In order to proceed with the testing of cointegration later, ideally, our variables should be I(1), in that in their original level form, they are non-stationary and in their first differenced form, they are stationary. The differenced form for each variable used is created by taking the difference of their log forms. For instance difference of our focus variable is calculated as:

$$DHPI_t = LHPI_t - LHPI_{t-1}$$

Only for not logged PR the difference is not the same because log symbol is not there:

$$DPR_t = PR_t - PR_{t-1}$$

Next we conducted the Augmented Dickey-Fuller, Philip-Perron and KPSS test for both level and differenced forms of each variable. The main difference between ADF or PP tests and KPSS test is while in former Null hypothesis is that variable is non-stationary in KPSS the Null is that variable is stationarity. So rejection of Null has opposite implications in those two sets of tests. The below table summarizes the results of each test.

Table 4.1.1: Augmented Dickey-Fuller (ADF) test results

<b>Variable in Level Form</b>			
Variable	Test Statistics	Critical Value	Implication
LHPI	-1.8648	-3.4535	Variable is non-stationary
LGDP	-.54652	-3.4535	Variable is non-stationary
LNML	-1.9265	-3.4535	Variable is non-stationary
PR	-3.7075	-3.4535	Variable is stationary <sup>5</sup>
LMS	-1.7834	-3.4535	Variable is non-stationary
LHS	-1.4575	-3.4535	Variable is non-stationary
LCE	-1.6638	-3.4535	Variable is non-stationary
<b>Variables in Difference Form</b>			
Variable	Test Statistics	Critical Value	Implication
DHPI	-3.2883	-2.8897	Variable is stationary
DGDP	-4.7965	-2.8897	Variable is stationary
DNML	-9.7896	-2.8897	Variable is stationary
DPR	-3.9050	-2.8897	Variable is stationary
DMS	-3.4842	-2.8897	Variable is stationary
DHS	-5.4660	-2.8897	Variable is stationary
DCE	-3.7252	-2.8897	Variable is stationary

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<sup>5</sup> Even if Prime rate (PR) is an interest rate where we do not expect to have a trend but only an intercept value, we still used the results for the Dickey-Fuller regression with an intercept and a linear trend because for given time period variable seems to have a clear downward trend. See Graph 4.1 above for visualization.

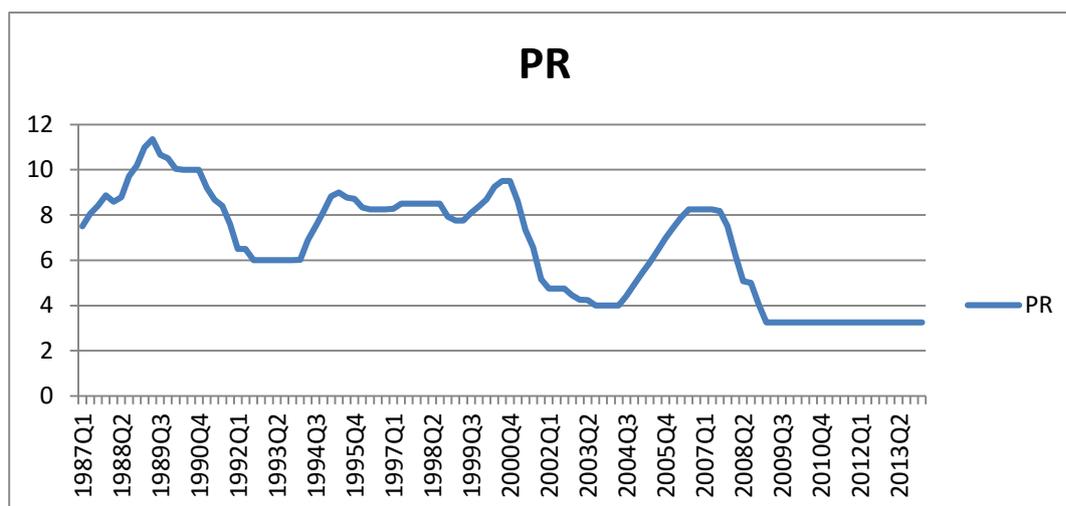


Figure 4.1: Trend of the Prime Rates (PR) of the U.S. banks from 1987Q1 to 2014Q1.

Since our results show that all of the variables are either  $I(0)$  or  $I(1)$  we can say that use of ARDL is appropriated. However, we have to have support for our findings using either Phillips-Perron Unit Root (PP) test or KPSS Stationarity test. We will start with PP test first for which results are displayed in the below table:

Table 4.1.2: Phillips-Perron Unit Root (PP) test results

<b>Variable in Level Form</b>			
Variable	Test Statistics	Critical Value	Implication
LHPI	-1.6248	-3.4515	Variable is non-stationary
LGDP	-.51704	-3.4515	Variable is non-stationary
LNML	-2.7426	-3.4515	Variable is non-stationary
PR	-2.1596	-3.4515	Variable is non-stationary
LMS	-1.2738	-3.4515	Variable is non-stationary
LHS	-1.8098	-3.4515	Variable is non-stationary
LCE	-1.3477	-3.4515	Variable is non-stationary
<b>Variables in Difference Form</b>			
Variable	Test Statistics	Critical Value	Implication
DHPI	-6.8375	-2.8884	Variable is stationary
DGDP	-7.8474	-2.8884	Variable is stationary
DNML	-12.4357	-2.8884	Variable is stationary
DPR	-3.9673	-2.8884	Variable is stationary
DMS	-6.5479	-2.8884	Variable is stationary
DHS	-8.2450	-2.8884	Variable is stationary
DCE	-6.4273	-2.8884	Variable is stationary

Last test we will perform is KPSS Stationarity test. The test results are show in table below:

Table 4.1.3: KPSS Stationarity test results

<b>Variable in Level Form</b>			
Variable	Test Statistics	Critical Value	Implication
LHPI	.093231	.14061	Variable is stationary
LGDP	.16791	.14061	Variable is non-stationary
LNML	.085987	.14061	Variable is stationary
PR	.091719	.14061	Variable is stationary
LMS	.14914	.14061	Variable is non-stationary
LHS	.12488	.14061	Variable is stationary
LCE	.12368	.14061	Variable is stationary
<b>Variables in Difference Form</b>			
Variable	Test Statistics	Critical Value	Implication
DHPI	.10997	.39896	Variable is stationary
DGDP	.50119	.39896	Variable is non-stationary
DNML	.078140	.39896	Variable is stationary
DPR	.13808	.39896	Variable is stationary
DMS	.32984	.39896	Variable is stationary
DHS	.080735	.39896	Variable is stationary
DCE	.13540	.39896	Variable is stationary

The results of KPSS test in conflicting with implications of ADF and PP tests which is normal to expect. Former two tests are consistent in their implications about variables that they are non-stationary in level form and stationary in differenced form. Only exception is Prime rate (PR), where ADF results show that it is stationary in level form, but PP results tell that it is stationary. However, KPSS results confirm ADF test's implications of stationarity in level form. Thus, in general results are mixed. Since results of our test for non-stationarity are not consistent we decided to use ARDL method.

However, before we proceed with cointegration test we want to find out the order of the vector Autoregression (VAR), which shows number of appropriate lags to be used. From the table below we can see that AIC recommends six lags while SBC recommends only one.

Table 4.1.4: Cointegration test results

	Choice of Criteria	
	AIC	SBC
Optimal Lag Order	6	1

Since results are conflicting we decided to take a number in between as four lags.

Next step would be to check for existence of cointegration among variables. If there exists at least one cointegration it would mean that results are not spurious. The results of Engle-Granger (E-G) test prove the existence of at least one cointegration:

Table 4.1.5: Engle-Granger (ADF) test statistics

Engle-Granger (ADF) test	
T statistics	Critical Value for the Test
-2.6595	-4.36*

\*Critical Value is taken from Sjo (2008).

As we can see from above table t statistics is less than critical value, which means we cannot reject the null hypothesis of non-stationarity. This means chosen variables result into non-stationary error term. Thus it indicates that there is no cointegration. However, since this outcome is not appealing we decided to proceed with Johansen cointegration test. The results of Johansen cointegration test for the Maximal Eigenvalue, Trace, AIC, SBC and HQC criteria are given in the table below:

Table 4.1.6: Johansen cointegration test results

Criteria	Number of Cointegrating Vectors
Maximal Eigenvalue	4
Trace	5
AIC	7
SBC	2
HQC	7

However, we should not forget that in VAR model Johansen's log-likelihood Maximal Eigenvalue and Trace tests are based on cointegration with an intercept but no trend. These results are both conflicting among themselves and Engle-Granger test results. This may be VAR approach's limitation dealing with mixed I(0) and I(1) variables.

Since our interest rate variable (PR) is most likely to be I(0), we will proceed with ARDL approach which takes care of VAR's limitations. First, we start with testing for the existence of long-run relationship among variables using ARDL. Results of those tests are depicted below:

Table 4.1.7: Results of Long-Run Relationship test in ARDL

Variable	F statistics	Lower CV at 5%*	Upper CV at 5%*
LHPI	8.0766	2.365	3.553
LGDP	3.5865	2.365	3.553
LNML	3.0600	2.365	3.553
PR	2.7926	2.365	3.553

LMS	1.8900	2.365	3.553
LHS	9.6011	2.365	3.553
LCE	4.3173	2.365	3.553

\* Lower and Upper Critical Values (CVs) are taken from Pesaran, Shin, & Smith (2001). The range of Lower and Upper CVs for 1% and 10% levels of significance are 3.027-4.296 and 2.035-3.153 respectively.

As we can see from the table most of F statistics are higher than upper critical value of 3.553. Only those which are lower are New Mortgage Loans (NML), Prime Rate (PR) and Money Supply (MS). Since those are usually leading (exogenous) variables therefore it is normal for us to see them being independent of other variables. Nevertheless, we find F statistics to be very significant for equation of the our focus variable, House Price Index (HPI), some variable related to house prices, such as Housing Started (HS) and Construction Expenditure (CE). This finding show that there seems to clear evidence for existence of long-run relationship among given variables. Thus, the relationship which exist is not spurious relationship, but in contrast there is founded long-run relationship among given variables.

Following step will be Error Correction Model (ECM) using ARDL approach. Distinct from VAR approaches ECM, ARDL lets ECM to choose optimal lags for each variable separately. Therefore, in this aspect ARDL is considered more advance than regular VAR approach. We chose to use AIC as criteria for choice of ECM here. Summarized table of ECM coefficients, standard error, t statistics and p-value are provided below:

Table 4.1.8: Results of Error Correction Model

Variable	Coefficient	Standard Error	T-statistics [P-value]
ecm (-1) dLHPI	-.11311	.025687	4.4035[.000]***
ecm (-1) dLGDP	-.0034788	.011105	.31326[.755]
ecm (-1) dLNML	-.60024	.086363	6.9502[.000]***
ecm (-1) dPR	-.10713	.026322	4.0698[.000]***
ecm (-1) dLMS	.022545	.017893	1.2600[.211]
ecm (-1) dLHS	-.29610	.061887	4.7845[.000]***
ecm (-1) dLCE	-.19101	.028289	6.7521[.000]***

Note: \*\*\*denotes that coefficient is significant at 1% level.

Cointegration shows existence of long-term relationship, but sometimes there could be deviations from long-run in short-term relationships. Thus, cointegration does not tell us

much about short-run relationship and how it affects long-run relationship. That is the reason we used ECM to explain effect of short-run influence on the long-run relationship. Those equations (variables) which have significant coefficient for their  $ecm(-1)$  are found to be dependent on other variables for determination their values in short run that has long term effect, thus considered endogenous in the model. Conversely, those variables which don't have significant  $ecm(-1)$  coefficient do not depend on other for determination of its value, thus considered exogenous in the model. Furthermore, the coefficient of the error-correction term tells us about the speed of short-run adjustment of the given variable.

The ECM results tell us that all variables except GDP and Money supply seem to be endogenous. Variable that converges fastest in the short-run is new mortgage loans (NML). Our focus variable House price index (HPI) is also among endogenous variables. Thus, GDP and Money supply seem to have significant impact on house price.

Even if ECM does tell us which variables are endogenous and which are exogenous it does not tell much about relative ranking of variable from the most exogenous to the most endogenous. Therefore, we need to employ the Variance Decomposition (VDC) technique to identify ranking in terms of exogeneity or endogeneity. There are two ways to apply VDC technique, namely orthogonalized and generalized. There is relative shortcoming of each. In orthogonalized ordering of the variable is vital and ranking is bias towards first in the order. Therefore, the orthogonalized VDC results are not unbiased. The issue with generalized VDCs is that sum of the lagged impacts is not normalized to 100%, even if results are unbiased. Since, results of the generalized VDCs could be normalized to 100% manually, we prefer later to the orthogonalized ones. Thus we will only concentrate and report the generalized VDC results. Following table is summary of the results generalized VDCs which are normalized to 100% for 1 year, 3 year and 10 year time horizons.

Table 4.1.9: Results of Error Correction Model for 1 year (4 quarter) time horizon

Variable	Horizon	DHPI	DGDP	DNML	DPR	DMS	DHS	DCE
DHPI	4	60.3%	7.2%	2.2%	4.5%	10.2%	7.4%	8.3%
DGDP	4	16.4%	44.0%	1.8%	3.7%	4.8%	16.7%	12.5%
DNML	4	2.6%	4.0%	73.5%	1.6%	8.1%	2.5%	7.6%
DPR	4	16.4%	11.2%	1.4%	50.0%	6.0%	4.8%	10.2%
DMS	4	8.7%	19.2%	4.2%	7.2%	51.0%	6.7%	3.0%
DHS	4	14.5%	17.0%	2.0%	3.7%	4.6%	44.1%	14.2%
DCE	4	14.8%	12.3%	3.9%	4.4%	2.9%	20.2%	41.4%

Table 4.1.10: Results of Error Correction Model for 3 year (12 quarter) time horizon

Variable	Horizon	DHPI	DGDP	DNML	DPR	DMS	DHS	DCE
DHPI	12	45.4%	5.4%	2.9%	14.5%	18.2%	7.1%	6.6%
DGDP	12	18.7%	36.5%	5.7%	6.5%	4.4%	14.8%	13.5%

DNML	12	3.9%	6.6%	63.8%	4.2%	8.7%	3.3%	9.6%
DPR	12	16.0%	10.5%	2.6%	49.1%	5.8%	6.1%	10.0%
DMS	12	7.8%	18.5%	4.3%	10.9%	48.1%	6.7%	3.7%
DHS	12	14.9%	14.4%	5.8%	8.0%	5.6%	36.4%	14.9%
DCE	12	20.3%	11.0%	5.6%	8.1%	4.5%	19.1%	31.4%

Table 4.1.11: Results of Error Correction Model for 10 year (40 quarter) time horizon

Variable	Horizon	DHPI	DGDP	DNML	DPR	DMS	DHS	DCE
DHPI	40	48.2%	5.9%	4.5%	14.5%	13.9%	7.1%	5.9%
DGDP	40	18.4%	35.2%	6.4%	7.1%	5.0%	14.6%	13.2%
DNML	40	5.8%	7.0%	60.6%	4.8%	8.6%	3.8%	9.5%
DPR	40	15.6%	10.3%	3.1%	46.6%	7.2%	7.1%	10.2%
DMS	40	10.9%	18.2%	4.6%	10.5%	45.2%	6.4%	4.2%
DHS	40	16.9%	13.3%	6.2%	9.0%	6.4%	34.2%	14.1%
DCE	40	20.2%	10.4%	6.7%	8.8%	5.2%	19.5%	29.2%

The results of the generalized VDCs are interesting, especially if we observe impact of each variable over different time horizons. For short-run (4 quarter) horizon, New Mortgage Loans seem to be the most exogenous followed by House Prices, and GDP seems to be the most endogenous. For medium-run (12 quarters) horizon, still New Mortgage Loans are the most exogenous followed by Interest rates (PR), and most endogenous one is Construction Expenses. Finally, for long-term (40 quarters) horizon, still New Mortgage Loans are the most exogenous now followed by House prices, and the most exogenous variable is still Construction expense.

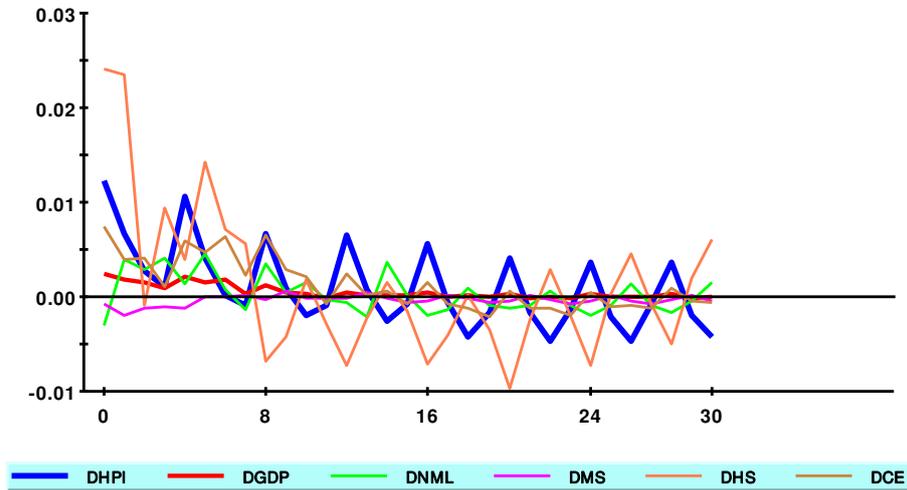
## 5. RESULTS AND POLICY RECOMMENDATIONS

Our results show that we can derive from them some good theoretical and policy implications. First of all, amount of New Mortgage Loans (bank lending for house financing) seems to be leading indicator for short, medium and long terms. Thus, bank lending seems to have impact on not only house prices, but also many other aspects of economy. In short-run GDP seem to be most follower (endogenous) variable. Which means GDP depends on other variables in short-run but later on for medium and long term it gets less endogenous and more exogenous. However, for House prices situation is reversed. In short run it is second most exogenous variable, but this exogeneity drops in medium time horizons but later on pick up again for longer time horizon.

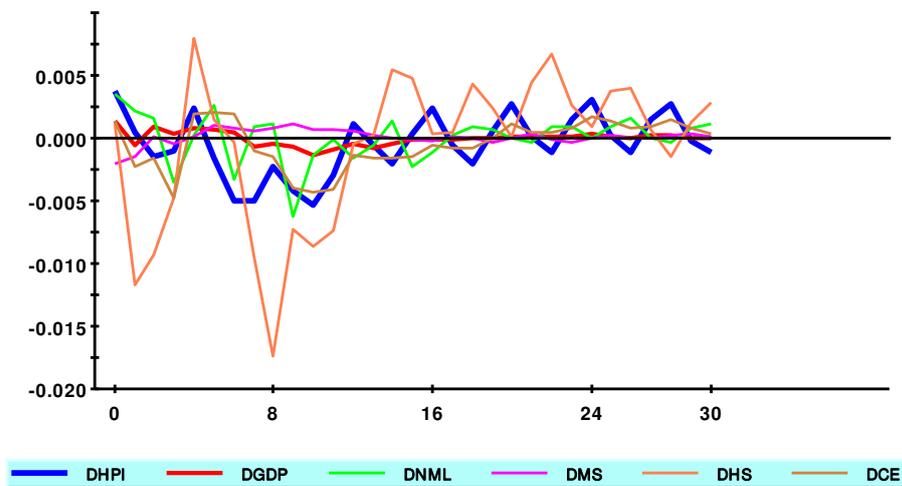
The factor that seems to influence house prices in short run are money supply, construction expense and housing started in respective order of importance. However, in medium and long run only interest rate and money supply seem to be importance. Therefore, we can say that new amount of bank lending especially for house financing seem to be the most exogenous variable that impact overall model. However, interest rate and money supply seem to have

direct impact on house prices. But, in short-run house prices seem to have influence over many other variables as well, especially GDP. Thus we can see impact of house prices on bank lending and GDP in short run, while interest rate and money supply influences it on long run. This can be seen on impulse response function depicted below:

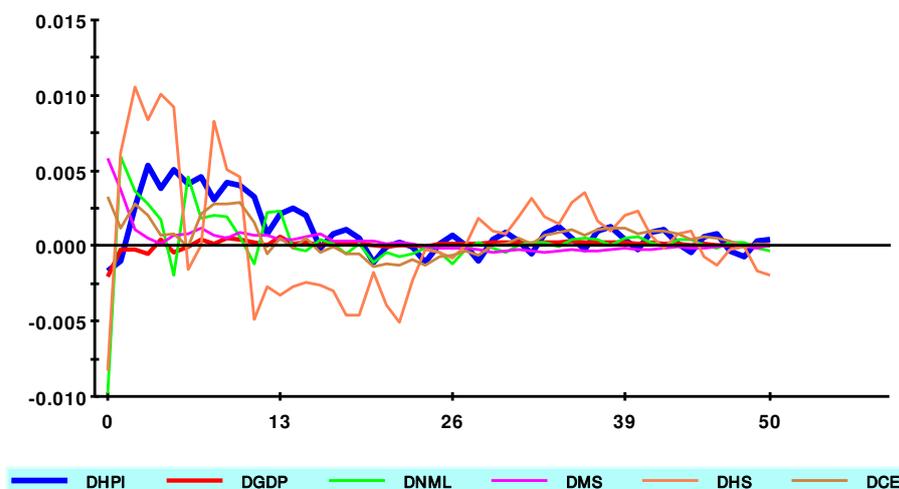
**Generalised Impulse Responses to one SE shock in the equation for DHPI**



**Generalised Impulse Responses to one SE shock in the equation for DPR**



### Generalised Impulse Responses to one SE shock in the equation for DMS



Therefore, we have to consider negative consequences of direct or indirect influence of bank lending and monetary policy on house and as consequence on national income (GDP). There is additional benefit from smoothing of house price fluctuations by reducing impact of financing on it. In one of their studies OECD (2011) summarized the impact of different changes of house financing policy among its member countries follows:

Table 5.1. The effect of policies on reducing real house price volatility<sup>1</sup>

Real house price volatility can be reduced by...	Policy experiment
25%	A further improvement in banking supervision equivalent to that observed on average in OECD over the 1990-2005 period (based on an index sourced from Abiad <i>et al.</i> 2008).
20%	Reducing the maximum loan-to-value ratio by 10 percentage points. <sup>2</sup>
19%	Increasing the estimated supply elasticity from the level observed in Ireland to the level in Canada (see Figure 4.1).
11%	Reducing the tax relief on mortgage debt financing costs from the level observed in Netherlands to the level in Sweden (see Figure 4.7).

1. The policy experiments are roughly equivalent to the impact of a one standard deviation change in the policy variables of interest on real house price volatility. Estimates are based on random effects panel regressions for between 16 and 20 OECD countries, over the period circa 1980-2005. The dependent variable is the standard deviation in annual real house price growth and the model also controls for macroeconomic volatility and time fixed effects (see Andrews (2010) for details).
2. Over the sample period, loan-to-value ratios range from a minimum of 56% to a maximum of 110% in OECD countries.

Source: Abiad, A., E. Detragiache and T. Tressel (2008), "A New Database of Financial Reforms", IMF Working Paper No. 08, Vol. 266, International Monetary Fund; Andrews, D. (2010), "Real House Prices in OECD Countries – The Role of Demand Shocks and Structural and Policy Factors", OECD Economics Department Working Papers.

From observing above table we can notice that the self imposed (or over-imposed by regulators) discipline for the banks bring along improvement in bank supervision (i.e. Row 1). Additionally, tightening access to credit would reduce the ratio of loans, also benefits economy by reducing house price volatility. As this study reports the maximum loan-to-value ratio – the maximum permitted value of the loan as a share of the market price of the property is considered as a measure of down payment constraint. It give an estimates suggesting that a

10 percentage point increase in the maximum loan-to-value ratio is associated with a 12% rise in the home ownership rate among younger low-income households (i.e. owners aged 25-34 years in the second income quartile) compared to a typical household. As we would recall one of the reasons of subprime crisis was precisely such ease of housing loan provided to household with unstable income. One of suggestions in our research is similar to effect as decreasing the maximum loan-to-value ratio.

## **6. Conclusion**

There seems to be strong relations between house prices and the ease of obtaining housing loan and its cost (as measured by interest rates). However, does increasing price make lending attractive or does easy lending lead to high prices seems to be a debatable issue. This study made an attempt to resolve this issue to some extent using data for the United States. We have employed ARDL approach for time series data to analyze available data with the use of this time series technique.

The finding of this study suggests, while house prices seem to be the exogenous (leading) variable in the short-run which influence many other variables such as, construction of new houses and its cost. However, in both short and long run, the interest rates seem to be important factor contributing to changes in house prices. New mortgage loans seem to be exogenous factors in both short and long run impacting house prices both directly and indirectly. Therefore, we have to reconsider making necessary changes into the current way of house financing which seems to be causing serious fluctuations in house prices.

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