



**RELATIONSHIP BETWEEN STOCK MARKET AND
REAL ESTATE MARKET: EVIDENCE IN THAILAND**

BY


MISS CHIDCHANOK SUVICHACHERDCHOO

**AN INDEPENDENT STUDY SUBMITTED IN PARTIAL
FULFILLMENT OF THE REQUIREMENTS FOR
THE DEGREE OF MASTER OF SCIENCE
PROGRAM IN FINANCE (INTERNATIONAL PROGRAM)
FACULTY OF COMMERCE AND ACCOUNTANCY
THAMMASAT UNIVERSITY
ACADEMIC YEAR 2016
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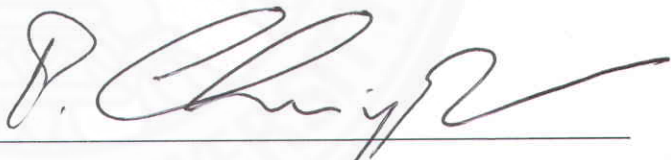
ENTITLED

RELATIONSHIP BETWEEN STOCK MARKET AND REAL ESTATE MARKET:
EVIDENCE IN THAILAND

was approved as partial fulfillment of the requirements for
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ABSTRACT

This paper study about relationship between stock market, which represented by SET Index, and real estate market, which represented by price index of land, condo, townhouse, and single detach house. Based on two theories which are wealth effect and credit price effect. This paper using bivariate VAR model, also granger causality, and forecast error variance decomposition. The result was found support both theories. To explain, return of land price index has affect return of SET index which the result support credit price effect. While return of SET index was found support return of price index of townhouse and single detach house, which the result support credit price effect.

Keywords: Real Estate Market, Stock Market, Wealth Effect, Credit Price Effect

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To all my classmates who taught life lesson that will never be forgotten.

Miss Chidchanok Suvichacherdchoo

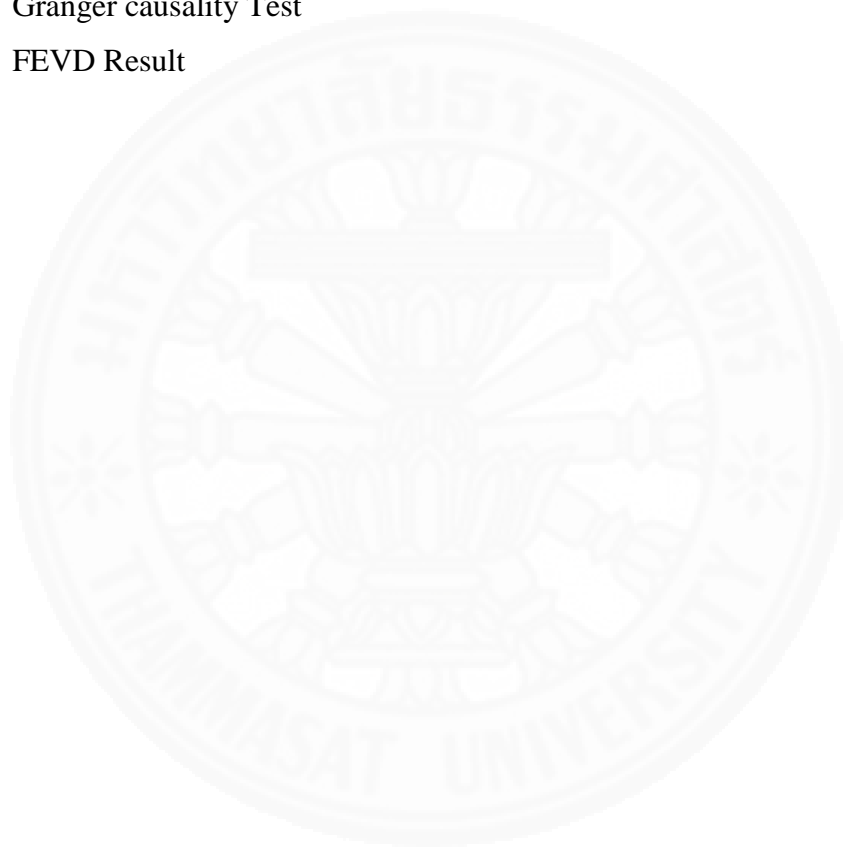


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CHAPTER 1

INTRODUCTION

Real estate investment is another alternative investment that can be chosen to diversify investor's portfolio. Knowing relationship between stock price and real estate price can help investors as another possibility in order to consider future movement of the markets over historical data of one another.

Impact of wealth on consumption had been studied (green,2002; Piazzesi *et al.*, 2007; Chen,2001; Sutton,2002; Kakes and Van Den End,2004; Ibrahim,2010) and shows the mechanism that augment the relationship between stock and real estate price which is the wealth effect. Wealth effect states that when there is an increasing in the stock market. Investors with unanticipated gain will feel wealthier and more comfortable to purchase properties; this will push up demand on real estate. Hence the stock market will lead real estate market.

Another effect that had been studied call credit-price effect (Sim and chang,2006) shows that when real estate price is up, firms will effected first on the statement of balance sheet imply that when property value rise then there is an unreleased gain and will end up with rising in equity. Firm will have lower cost in case of reinvestment, which leads stock market to be increased also.

Earlier researches study about dynamic linkage between stock price and real estate price, most has examined the correlation of two asset returns and data is the evidence of U.S. and U.K. (Ibbotson and Siegel,1984; Hartzell,1986; Eichholtz and Hartzell,1996). Therefore this paper will be conduct in order to answer the question about how the two markets; stock market and real estate market, interrelate to one another.

Many studies have examined the linkage of these two markets, stock market and real estate market, but most the studies were evidence on developed countries, for example, the United State of America, United Kingdom, Turkey. While studies base on developing countries, which published, were limited in the small amount as well as the applicable of the research result.

Therefore, objective of this study is to examine the relationship between stock market and real estate price using land price data available in BOT as real estate price, and using Stock Exchange of Thailand Index as stock market price. Data of this paper will be between 2008-2016, This research will apply Vector Autoregressive Model (VAR) as early research had been applied (Green,2002; Kapopoulos and Siokis,2005; Chen,2001; Ibrahim,2010). Moreover, Granger Causality, Impulse Response Function (IRF), and Forecast Error-Variance Decomposition (FEVD) will also be applied.



CHAPTER 2

REVIEW OF LITERATURE

There are some researches found evidence supported wealth effect and credit-price effect as these two effects can describe the relationship of stock market and real estate of each particular data. For example of wealth effect, the study of (Green, 2002) applied concept of Granger causality using single-equation framework. The paper was employed data of four different sources to represent real estate price in California and using Russell 2000 to represent Stock Market. The result according to the paper (Green, 2002), finds evidence shows that stock market value affects housing consumption.

Earlier researches study about dynamic linkage between stock price and real estate price, most have examined the correlation of two asset returns and evidence data shows both negative relation and positive relation. Researches based on United Kingdom evidence are (Worzala and Vandell, 1993), (Eichholtz and Hartzell, 1996), and (Ibbotson and Siegal, 1984) which mainly focus on correlation of the return between the stock market and real estate market. (Worzala and Vandell, 1993) found positive correlation of 0.039 while (Eichholtz and Hartzell, 1996) found negative correlation of -0.08

Another interested research of (Quan and Titman, 1999) based on set of data that comprised of 17 developed and emerging markets in order to examine relationship of stock price and real estate price whether they move together under the condition of cross-sectional and panel regression. The result showed positive correlation under the cross-sectional regression. However, once controlled variable was added, the result of positive correlation had changed. (Chen, 2001) also have the result support wealth effect which found evidence support wealth effect by using bivariate VAR model. The paper based on quarterly data of Taiwan, data period from 1973 – 1992.

Relationship of the two markets wasn't limited within the same country, there was a research studies the relationship of these two markets across country. (Sutton, 2002) studied the relationship between six economies of developed countries; UK, USA, Canada, Netherland, Ireland, and Australia under VAR framework which the

result was support wealth effect that house price have response to change in stock price positively.

In the other hand, (Sim and Chang, 2006) using VAR model but they found the result in the opposite. Their research using VAR model and found the result supports credit-price effect. While (Miller,2014) shows that real estate and stock returns are correlated generally across lower frequencies but not for all the period of the sample. They use Vector Autoregressive model, Vector Error Correction model, and also threshold error correlation model to examine the linkage between these two markets. Another research that have been resulted both positive and negative effect (Chang, 2013) the paper shows positive relationship for the period of 1890-2012 and negative relation between 1998-2002. This paper using GDP growth as controlled variable and also found that stock price and real estate price are both having positive response to the GDP growth more than one another. Therefore, adding GDP growth will help the two markets avoid suffering some inaccuracy occurred by GDP growth.

One research, study relationship between these two markets, in southeast region (Lean and Smyth, 2012). The study using the standard augmented dickey fuller unit root test found evidence support wealth effect for some REITs. For the evidence of Thailand there is one research (Ibrahim, 2010) studies the relationship between these two market using the data from 1995-2006 and focus on using VAR model with two controlled variables, which are real GDP and aggregate price level. The paper found the evidence support wealth effect. Another research evidenced in Thailand using monthly data from 2008-2015 with controlled variable of interest rate, inflation, and real effective exchange rate under methodologies of bivariate VAR and DCC-GRACH. The paper proposed, “low correlation coefficients between the real estate and stock markets prevail and negative correlation between townhouse and stock indices as well as condominium and stock indices” as the result (Padungsaksawasdi and Jaroenjitrkam, 2016). Research of Nittayagasetwat, Aekkachai, and Jiroj Buranasiri was about the relationship between property fund and stock market which indicated that “property fund’s return is similar to the overall stock’s return by 26%, including value stock 0%, growth stock 4%, medium capitalization stock 9%, and small capitalization stock 13%”.

Recent researches find the relationship between stock and real estate price evidence in Turkey (Yuksel, 2016) using VAR model found evidence support wealth

effect and credit price effect for the period of pre-crisis. During the crisis period, credit price effect was found but no wealth effect evidence for that period.



CHAPTER 3

DATA

In this study, we use the Stock Exchange of Thailand (SET) set index, as a stock market price. For real estate sector, there are four different price indexes which are land price index, single detached house price index, townhouse price index, and condominium price index. This study investigates these four price indexes as a real estate price index. These data is from the Bank of Thailand. Because of the availability of the real estate price index, the data are ranging from 2008 and 2016. The stock price index and real estate price indexes are monthly data. We also use the macroeconomic variables such as manufacturing index and interest rate, which is 1 year T-Bill interest rate, as controlled variables. Previous study on Thai data, such as (Ibrahim, 2010), employs GDP growth and consumer price index as controlled variables with quarterly frequency. Since this paper using monthly data, we use the percentage change of manufacturing production index to match monthly frequency with real estate price index and set index. By using percentage change of manufacturing production index can be reliable according to published research of (Mitchell, Smith, Weale, Wright, and Salazar, 2005). The research has refer that “A range of monthly series is currently available giving indications of short-term movements in output. As the only available information, they are already exploited in various ways: financial commentators routinely examine monthly data on retail sales, the trade figures, and the output of the production industries in order to assess the state of the economy and likely developments in monetary policy; academic researchers exploiting high frequency econometric techniques make use of one another of these series as the best available proxy for a broader measure of demand or output.” Therefore, percentage change of manufacturing production index is appropriate to be used alone as GDP growth for monthly data.

Figure 3.1: Price Index of Real Estate Indexes of Thailand



*Noted that land condo twh and sdh are price index.

Figure 3.2: Price Index of SET

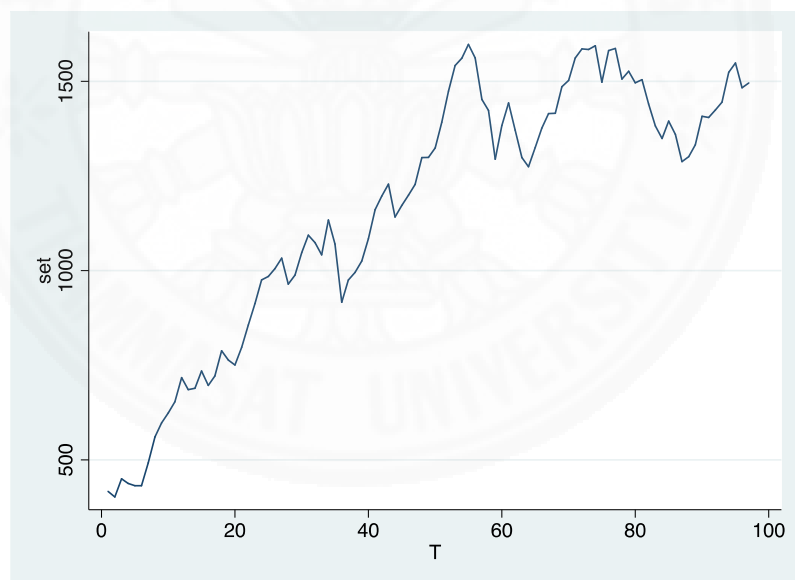
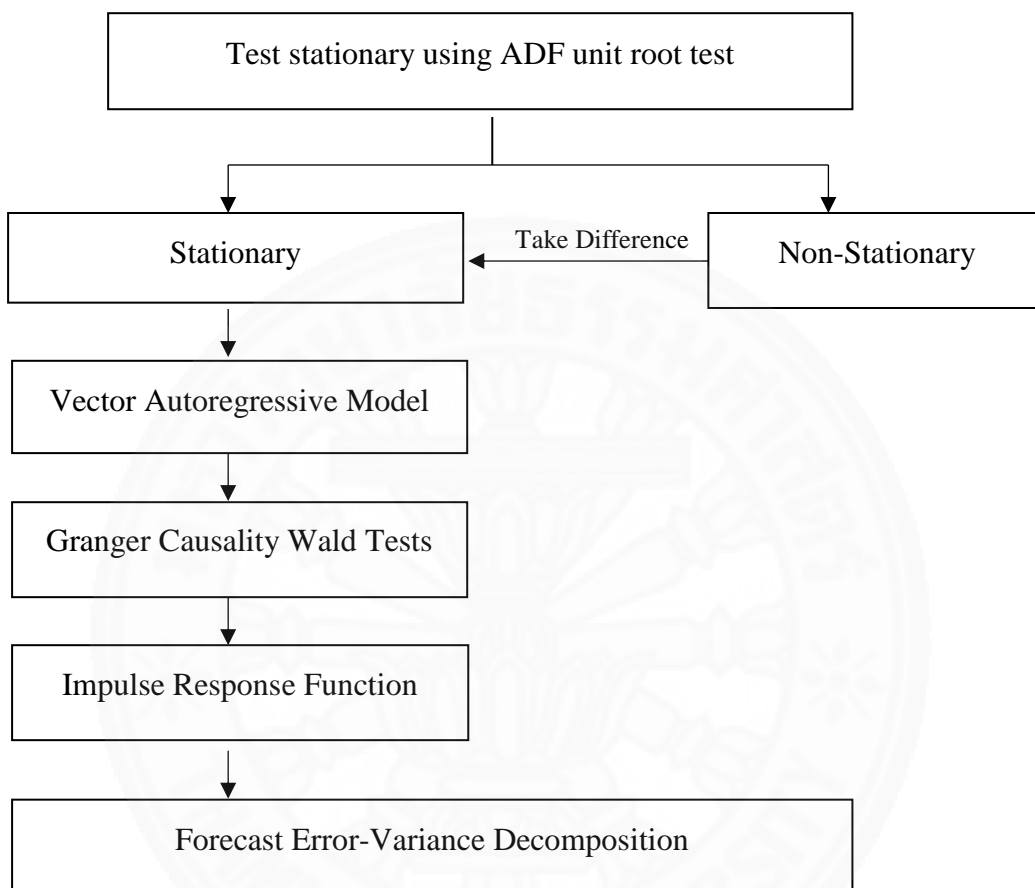


Figure 3.1 showed the illustration of four different kinds of real estate price index, which are land price index, condominium price index, town house price index, and single detach house price index. While the graph showed 97 periods of time begin from October 2008 to September 2016.

Figure 3.2 showed the illustration of Stock Price Index which both figure can tell that the data sets has trend which is not stationary.

CHAPTER 4

RESEARCH METHODOLOGY



In this study, we investigate the dynamic relationship between the stock price and real estate price using the Vector Autoregressive (VAR) model. After we estimate the VAR model, we will use the Granger Causality test, Impulse Response Function (IRF) and Variance decomposition to explain their relationship. Before we estimate the VAR we need perform the Augmented Dickey-Fuller unit root test on variables using in the model (Green,2002) (Chen,2001).

In the case that data is not stationary then we have to make the data stationary by taking the different follow the lag length and order of the model that has to be estimated before running any model. After making data stationary there are many tests that can be apply. In our case, we suspect that the price indexes are nonstationary and we need to transform them into return series which are stationary.

In case the data is already stationary then granger causality can be apply next to see whether there is any relationship between stock market and real estate market and if the relationship exists then what direction the relationship is. Long-run Granger causality of stock market and real estate market can also be tested though the null hypothesis $\beta_{k,i} = \dots = \beta_{k,j} = 0$ to see the relationship between the two market.

Before apply VAR test, the appropriate lag and order of the variables need to be determined. This paper considers appropriate lag and order by minimize of AIC, BIC, likelihood-ratio, and FPE. Reduce-form of VAR is widely use which is shown below. This paper will do the estimation using the bivariate form of VAR

$$R_SET_t = \Gamma_{1t} + \sum_{p=1}^P \theta_{1p} SET_{t-p} + \sum_{g=0}^G \theta_{1g} X_{t-g} + \lambda_1 Z_t + \varepsilon_{st}$$

$$R_X_t = \Gamma_{2t} + \sum_{j=0}^J \theta_{2j} SET_{t-j} + \sum_{k=1}^K \theta_{2k} X_{t-k} + \lambda_2 Z_t + \varepsilon_{jt}$$

where R_SET_t is changes in Return of stock price index at time t
 R_X_t is changes in Return of real estate price index at time t
 ;(LAND, CONDO, SDH, TWH)

Γ_{1t}, Γ_{2t} are constants

θ_{1p} is parameters of auto regressive optimal at lag p

θ_{1g} is parameters of auto regressive optimal at lag g

θ_{1j} is parameters of auto regressive optimal at lag j

θ_{1k} is parameters of auto regressive optimal at lag k

λ_1, λ_2 is parameters of controlled variables

Z_t is controlled variable ;(MPI, R)

$\varepsilon_{st}, \varepsilon_{jt}$ is error term

This paper will test the relationship between the two markets with different set of controlled variables. In models 1-4 we include manufacturing production index

(MPI). In models 5 to 8, we include MPI and Interest rate. In models 9 to 12, we include Interest rate.

- Noted that
- Model 1: Represents relationship between Return of SET Index and return of Land Price Index with MPI as controlled variable.
 - Model 2: Represents relationship between Return of SET Index and return of Condo Price Index with MPI as controlled variable.
 - Model 3: Represents relationship between Return of SET Index and return of TWH Price Index with MPI as controlled variable.
 - Model 4: Represents relationship between Return of SET Index and return of SDH Price Index with MPI as controlled variable.
 - Model 5: Represents relationship between Return of SET Index and return of Land Price Index with MPI and R as controlled variables.
 - Model 6: Represents relationship between Return of SET Index and return of Condo Price Index with MPI and R as controlled variables.
 - Model 7: Represents relationship between Return of SET Index and return of TWH Price Index with MPI and R as controlled variables.
 - Model 8: Represents relationship between Return of SET Index and return of SDH Price Index with MPI and R as controlled variables.
 - Model 9: Represents relationship between Return of SET Index and return of Land Price Index with R as controlled variable.
 - Model 10: Represents relationship between Return of SET Index and return of Condo Price Index with R as controlled variable.
 - Model 11: Represents relationship between Return of SET Index and return of TWH Price Index with R as controlled variable.
 - Model 12: Represents relationship between Return of SET Index and return of SDH Price Index with R as controlled variable.

Then apply Granger Causality to see the relationship between the two markets then test for Impulse Response Function in order to analyze the reaction of a system to a shock. To see response of endogenous variables in the VAR model to one-time exogenous impulse with all other variable dated t or earlier held constant and it is better analyzed by using graph. Therefore this paper will also estimate a unit of shock to the system. Finally, FEVD will be applied in order to aid in the interpretation of a vector autoregression (VAR) model once it has been fitted.



CHAPTER 5

RESULT

Table 5.1 displays the summary statistics for the SET index return (R_SET), the land price return (R_LAND), the condominium price return (R_CONDO), the townhouse price return (R_TWH) and the single-detached house price return (R_SDH). Table 5.1 also indicate that Set gives the highest return and highest risk while Single Detach house gives the lowest return but Town House gives the lowest risk. Moreover, the table shows that the data set (Return of price index) is stationary.

Table 5.1: Descriptive Statistic

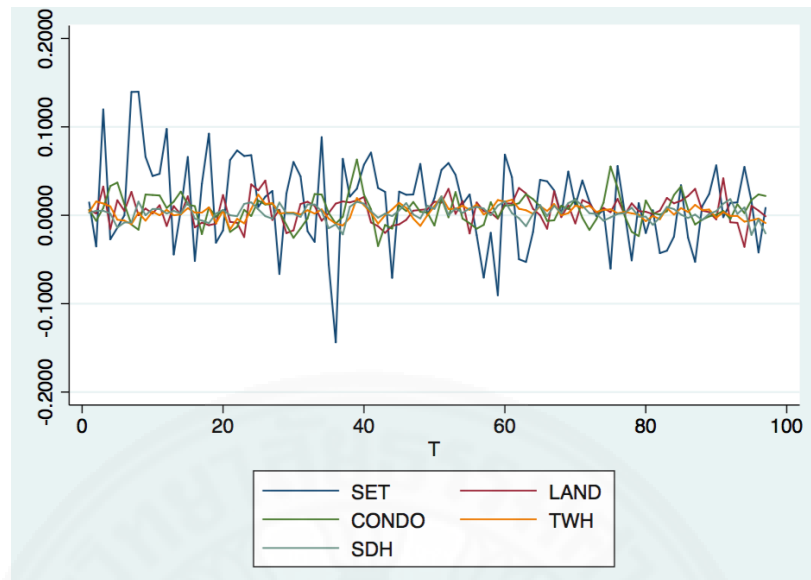
VARIABLES	MEAN	STD. DEV.	MIN	MAX	Skewness	Kurtosis
R_SET	0.0146	0.0493	-0.4376	0.1397	0.4687	0.2338
R_LAND	0.0058	0.0148	-0.0357	0.0418	0.9241	0.8871
R_CONDO	0.0062	0.0166	-0.0348	0.0630	0.0793	0.0862
R_TWH	0.0036	0.0077	-0.0159	0.0233	0.8952	0.9776
R_SDH	0.0028	0.0086	-0.0222	0.0214	0.0328	0.2232

Begin with the augmented dickey fuller unit root test in order to test weather the data set is stationary or not. Data set was not stationary, thus, the first difference is needed to be applied. Resulted in Table 5.2 that all variables are significance at 1% level critical value, 5% level critical value, and 10% level critical value, respectively.

Table 5.2: Unit Root Test

Variables	Test Statistic	1% Critical Value	5% Critical Value	10% Critical Value
R_SET	-8.528	-3.516	-2.893	-2.582
R_LAND	-8.967	-3.516	-2.893	-2.582
R_CONDO	-6.247	-3.516	-2.893	-2.582
R_TOWNHOUSE	-6.044	-3.516	-2.893	-2.582
R_SDH	-6.890	-3.516	-2.893	-2.582

Figure 5.1: Return of Price Index of each variable



*Noted that SET LAND CONDO TWH SDH are return of price index

Figure 5.1 shows that the returns are stationary. Since data set using in VAR model is time-series data therefore appropriate lag length is needed. In this paper, the appropriate lags are chosen by the minimum of AIC. The suitable lags for each models are presented in Table 5.3.

Table 5.3: Appropriate Lag

Endogeneous Variable		Appropriate Lag
SET	LAND	4
SET	CONDO	3
SET	SDH	3
SET	TWH	3

Table 5.4: VAR Result

	R_SET		R_LAND	
	Coefficient	SD	Coefficient	SD
R_SET _{t-1}	0.2159*	0.1024	0.0052	0.0263
R_SET _{t-2}	-0.1565	0.1004	0.0382	0.0258
R_SET _{t-3}	0.1822	0.1013	-0.0244	0.0260
R_SET _{t-4}	-0.0053	0.1010	-0.0858	0.0261
R_LAND _{t-1}	-0.2129	0.3724	0.2572*	0.0956
R_LAND _{t-2}	0.3571	0.3401	0.1621	0.0873
R_LAND _{t-3}	0.0539	0.3323	-0.4434*	0.0853
R_LAND _{t-4}	-0.2533	0.3810	0.2272*	0.0980
CONS	0.0119*	0.0068	0.0053*	0.0017

	R_SET		R_CONDO	
	Coefficient	SD	Coefficient	SD
R_SET _{t-1}	0.1759	0.0987	0.0147	0.0277
R_SET _{t-2}	-0.1307	0.0992	0.0097	0.0278
R_SET _{t-3}	0.1566	0.0985	-0.0220	0.0276
R_CONDO _{t-1}	0.0491	0.3357	0.3883*	0.0942
R_CONDO _{t-2}	0.1701	0.3574	-0.0524	0.1003
R_CONDO _{t-3}	0.1351	0.3310	-0.4546*	0.0929
CONS	0.0090	0.0059	0.0067*	0.0016

	R_SET		R_TWH	
	Coefficient	SD	Coefficient	SD
R_SET _{t-1}	0.1623	0.0957	0.0001	0.0137
R_SET _{t-2}	-0.1581	0.0962	-0.0137	0.0138
R_SET _{t-3}	0.1482	0.0951	0.0029	0.0136
R_TWH _{t-1}	-0.7392	0.6680	0.4876*	0.0959
R_TWH _{t-2}	0.3478	0.7183	0.1848	0.1032
R_TWH _{t-3}	-1.5992*	0.6722	-0.3897*	0.8965
CONS	0.0194*	0.0062	0.0028*	0.0008

	R_SET		R_SDH	
	Coefficient	SD	Coefficient	SD
R_SET _{t-1}	0.1620	0.0970	0.0259	0.0164
R_SET _{t-2}	-0.1265	0.0977	0.0040	0.0166
R_SET _{t-3}	0.1882	0.0965	0.0018	0.0164
R_SDH _{t-1}	1.2023*	0.5782	0.2335*	0.0982
R_SDH _{t-2}	-0.4822	0.6075	0.0843	0.1032
R_SDH _{t-3}	-1.1629	0.6144	-0.4170*	0.1044
CONS	0.0125*	0.0055	0.0027*	0.0009

According to the methodology discussed above, this paper estimates bivariate VAR with controlled variable of MPI or Manufacturing Production Index (Mitchell, Smith, Weale, Wright, and Salazar, 2005). We estimate the bivariate VAR with different proxies for real estate price. The results are presented in Table 5.4. Then, we use the Granger causality to explore dynamic relationship between stock return and real estate return.

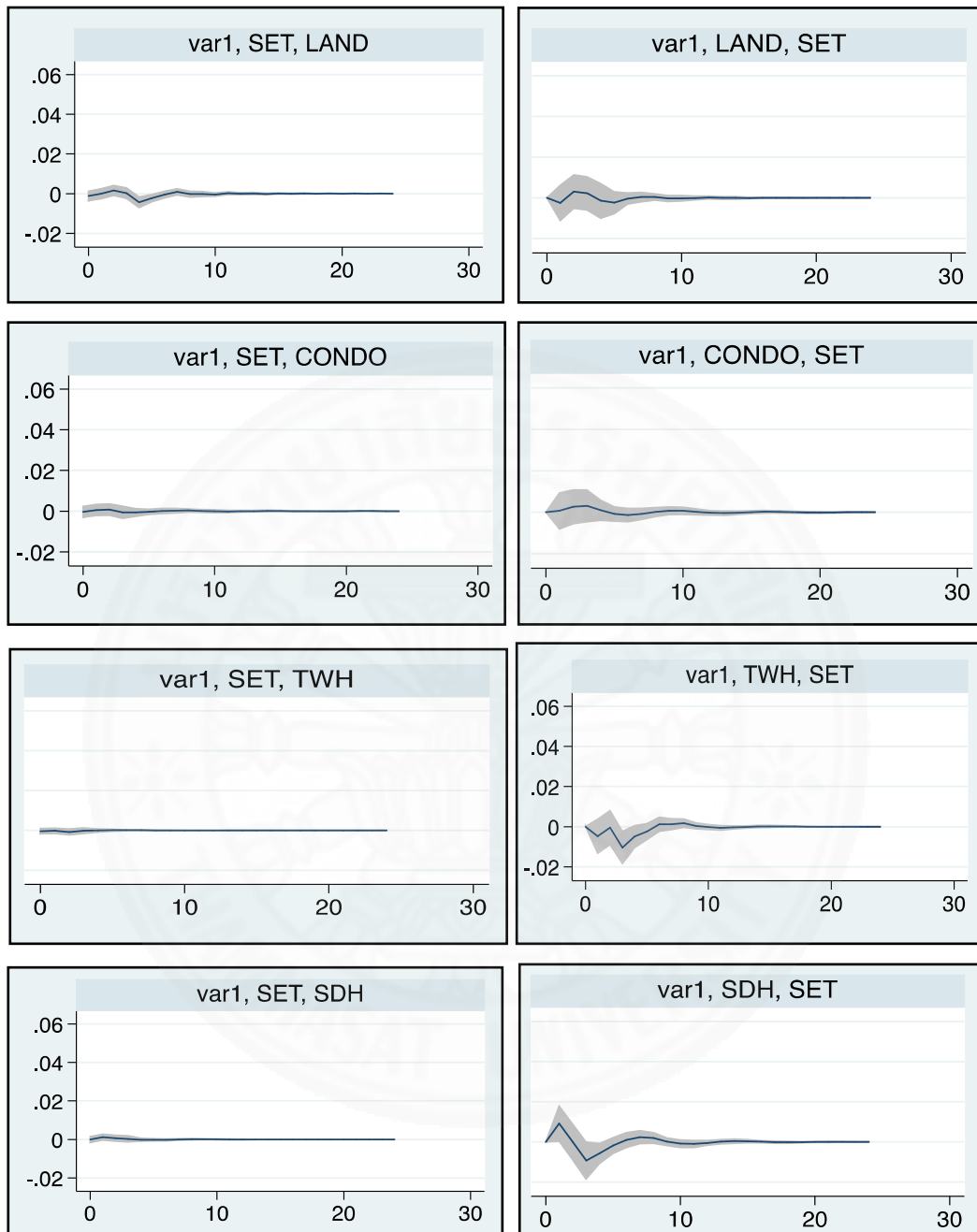
Table 5.5: Granger Causality Test

Model No.	Control Variable	Granger Causality	df	Prob > chi2
Model 1	MPI	$R_{SET} \not\rightarrow R_{LAND}$	4	0.054
	MPI	$R_{LAND} \rightarrow R_{SET}$	4	0.003
Model 2	MPI	$R_{SET} \not\rightarrow R_{CONDO}$	3	0.833
	MPI	$R_{CONDO} \not\rightarrow R_{SET}$	3	0.774
Model 3	MPI	$R_{SET} \rightarrow R_{SDH}$	3	0.025
	MPI	$R_{SDH} \not\rightarrow R_{SET}$	3	0.446
Model 4	MPI	$R_{SET} \rightarrow R_{TWH}$	3	0.047
	MPI	$R_{TWH} \not\rightarrow R_{SET}$	3	0.799

*Noted that \rightarrow means granger causality, $\not\rightarrow$ means not granger causality.

From Table 5.5, the return of LAND Granger causes the return of SET for all types of controlled variables, while the return of SET Granger causes the return of Townhouse and the return of Single-Detached House for all types of controlled variables condition. The result supports credit price effect in model that LAND is the representative of Real Estate Market and Wealth effect was supported by model of SDH and TWH which are the representative of Real Estate Market. However, there's no relationship between SET and CONDO which might be because that most of purchasers of LAND, TWH, and SDH were end-user which caused the price of these variables to reflect the rational price of the market, while CONDO price appeared under many types of purchasers (inspectors, investors, and end-user) which are not able to reflect market price rationally.

Figure 5.2: IRF Result



*Noted that SET is Return of Stock price index, CONDO is Return of condo price index, TWH is return of town-house price index, SDH is return of Single Detach house price index.

The IRF graph (Refer to Figure 5.2) which are the relationship of two markets with controlled variable of MPI. The graph shows the impulse by row and response by column with 24 monthly periods. For model1, the graph shows that effect of one S.D. impulse to LAND, SET was response by the small fluctuation in both directions less

than 0.01% changes and move back to normal with 8 periods, while response of LAND when impulse on SET show the similar result. For model2, the effect of one S.D. impulse to CONDO, SET was response with very small change, while response on CONDO was close to unchanged when impulse on SET. For model3, the effect of one S.D. impulse to TWH, SET show negatively fluctuate response less than 0.02% and move back to normal within 12 periods of time, while response on TWH when impulse on SET doesn't show any significant changing. For model4, response of SET when impulse on SDH show rising in SET less than 0.02% for the first 2 periods then decline to -0.01% at the fourth period and move back to normal for within 12 periods approximately. While SDH haven't showed any significant response when impulse on SET.

For Model 5-8, which has controlled variables of MPI and R gives the similar response to the set of model 1-4, respectively. As well as, the set of model 9-12 which has controlled variables of R gives the similar result of direction, magnitude, and effect period

Table 5.6: FEVD Result

FEVD	Forecast Errors Variance of SET		Forecast Errors Variance of LAND	
	Period ahead	Due to SET	Due to LAND	Period ahead
1	1	0	1	1
2	0.9962	0.0037	2	0.9962
3	0.9953	0.0046	3	0.9953
4	0.9953	0.0046	4	0.9953
5	0.9940	0.0059	5	0.9940
10	0.9922	0.0077	10	0.9922
20	0.9922	0.0077	20	0.9922

FEVD	Forecast Errors Variance of SET		Forecast Errors Variance of CONDO	
	Period ahead	Due to SET	Due to CONDO	Due to SET
1	1	0	0.0002	0.9997
2	0.9999	2.4e-06	0.0017	0.9982
3	0.9988	0.0011	0.0030	0.9969
4	0.9955	0.0044	0.0060	0.9939
5	0.9949	0.0050	0.0078	0.9921
10	0.9940	0.0059	0.0085	0.9914
20	0.9937	0.0062	0.0087	0.9912

FEVD	Forecast Errors Variance of SET		Forecast Errors Variance of SDH	
Period ahead	Due to SET	Due to SDH	Period ahead	Due to SET
1	1	0	1	1
2	0.9648	0.0351	2	0.9648
3	0.9649	0.0350	3	0.9649
4	0.9319	0.0680	4	0.9319
5	0.9206	0.0793	5	0.9206
10	0.9162	0.0837	10	0.9162
20	0.9155	0.0844	20	0.9155

FEVD	Forecast Errors Variance of SET		Forecast Errors Variance of TWH	
Period ahead	Due to SET	Due to TWH	Due to SET	Due to TWH
1	1	0	0.0121	0.9878
2	0.9855	0.0144	0.0123	0.9876
3	0.9839	0.0160	0.0199	0.9800
4	0.9284	0.0715	0.0196	0.9803
5	0.9185	0.0014	0.0196	0.9803
10	0.9135	0.0864	0.0202	0.9797
20	0.9131	0.0868	0.0202	0.9797

Forecast Error Variance Decomposition (Refer to Table 5.6), which it reveals the proportion of the movements in a sequence due to its own shocks versus shocks to the other variable given the result that similarly to IRF. Change on SET for one standard deviation was from the combination of SDH price had changed and combination of TWH price had changed. Therefore, the fluctuation of return on SET Index was determined by the fluctuation of return on SDH and the fluctuation of return on TWH.

CHAPTER 6

CONCLUSION

This paper tries to investigate the relationship between stock market and real estate market supporting by theories of wealth effect and credit-price effect. The paper had applied the VAR model to monthly data on SET index return and real estate return.

The result was support both wealth effect and credit-price effect that the return of SET Index has significant effect the return of Land price. In the other hand, the return of LAND granger causes the return of SET for all types of controlled variables condition, while the return of SET granger causes the return of Townhouse and the return of Single-Detached House for all types of controlled variables condition. The result shows positive relationship between return of stock index and return of single detach house price index while relationship between return of set index and return of town house price index was negative.

The result of relationship between return of SET index and return of land price index was support wealth effect that an anticipate gain in rising of stock price will causes higher purchasing power to invest in land as alternative investment.

Moreover, granger causality and IRF has support the result of VAR model with different controlled variables. This prove the result of the relationship between stock market and real estate market as mentioned.

However, the same data set of real estate price was contain from year 2008 to 2016, which affect sample size to be quite small. Further paper can apply by using longer period of data set.

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APPENDICES

APPENDIX A

EXOGENOUS VARIABLE OF MPI

VAR Result with exogeneous variable of MPI

```
. var SET LAND, lags(1/4) exog(MFP)
```

Vector autoregression

```
Sample: 5 - 97
Log likelihood = 435.162
FPE = 4.55e-07
Det(Sigma_ml) = 2.96e-07
No. of obs = 93
AIC = -8.928214
HQIC = -8.708302
SBIC = -8.383569
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	10	.048822	0.0944	9.694391	0.3758
LAND	10	.012542	0.3447	48.92266	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.2159205	.1024102	2.11	0.035	.0152002	.4166407
L2.	-.1565149	.1004821	-1.56	0.119	-.3534562	.0404265
L3.	.1822858	.1013045	1.80	0.072	-.0162673	.3808389
L4.	-.0053088	.1018051	-0.05	0.958	-.2048431	.1942255
LAND						
L1.	-.2129968	.3724501	-0.57	0.567	-.9429856	.5169919
L2.	.357136	.3401556	1.05	0.294	-.3095567	1.023829
L3.	.0539135	.3323032	0.16	0.871	-.5973888	.7052157
L4.	-.2533679	.3818215	-0.66	0.507	-1.001724	.4949885
MFP	-.0284127	.0286076	-0.99	0.321	-.0844826	.0276572
_cons	.0119936	.0068659	1.75	0.081	-.0014632	.0254504
LAND						
SET						
L1.	.0052788	.0263087	0.20	0.841	-.0462854	.056843
L2.	.0382966	.0258134	1.48	0.138	-.0122968	.08889
L3.	-.0244771	.0260247	-0.94	0.347	-.0754845	.0265304
L4.	-.0858456	.0261533	-3.28	0.001	-.1371052	-.0345861
LAND						
L1.	.2572905	.0956809	2.69	0.007	.0697594	.4448215
L2.	.1621639	.0873846	1.86	0.063	-.0091067	.3334345
L3.	-.4434693	.0853673	-5.19	0.000	-.6107862	-.2761525
L4.	.2272256	.0980884	2.32	0.021	.0349759	.4194753
MFP	.0105712	.0073492	1.44	0.150	-.0038329	.0249753
_cons	.0053783	.0017638	3.05	0.002	.0019212	.0088353

```
. var SET CONDO,lag(1/3) exog(MFP)
```

Vector autoregression

```
Sample: 4 - 97
Log likelihood = 428.97
FPE = 5.24e-07
Det(Sigma_ml) = 3.73e-07
No. of obs = 94
AIC = -8.786596
HQIC = -8.611736
SBIC = -8.353695
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.048757	0.0715	7.239269	0.4044
CONDO	8	.01369	0.3926	60.76029	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1759054	.0987919	1.78	0.075	-.0177232	.369534
L2.	-.1307833	.0992956	-1.32	0.188	-.325399	.0638325
L3.	.1566259	.0985593	1.59	0.112	-.0365467	.3497985
CONDO						
L1.	.0491425	.3357457	0.15	0.884	-.6089069	.7071919
L2.	.1701816	.3574659	0.48	0.634	-.5304387	.8708019
L3.	.1351115	.3310648	0.41	0.683	-.5137635	.7839865
MFP	-.0176011	.0283661	-0.62	0.535	-.0731975	.0379954
_cons	.0090927	.0059619	1.53	0.127	-.0025924	.0207778
CONDO						
SET						
L1.	.0147167	.0277383	0.53	0.596	-.0396494	.0690828
L2.	.0097337	.0278798	0.35	0.727	-.0449096	.064377
L3.	-.0220536	.027673	-0.80	0.425	-.0762917	.0321845
CONDO						
L1.	.3883572	.0942691	4.12	0.000	.2035932	.5731213
L2.	-.0524187	.1003676	-0.52	0.601	-.2491356	.1442982
L3.	-.4546071	.0929548	-4.89	0.000	-.6367952	-.272419
MFP	.0057155	.0079645	0.72	0.473	-.0098946	.0213256
_cons	.0067449	.001674	4.03	0.000	.003464	.0100257

. var SET SDH,lag(1/3) exog(MFP)

Vector autoregression

Sample:	4 - 97	No. of obs	=	94
Log likelihood =	484.1489	AIC	=	-9.960615
FPE	= 1.62e-07	HQIC	=	-9.785754
Det(Sigma_ml)	= 1.15e-07	SBIC	=	-9.527713

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.046716	0.1476	16.27549	0.0227
SDH	8	.00794	0.2517	31.61457	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1620828	.0970078	1.67	0.095	-.028049	.3522147
L2.	-.1265413	.0977602	-1.29	0.196	-.3181477	.0650651
L3.	.1882429	.096539	1.95	0.051	-.0009701	.3774559
SDH						
L1.	1.202353	.5782273	2.08	0.038	.0690485	2.335658
L2.	-.4822306	.6075773	-0.79	0.427	-1.67306	.708599
L3.	-1.162967	.6144559	-1.89	0.058	-2.367278	.0413447
MFP	-.0060441	.0266607	-0.23	0.821	-.0582981	.0462099
_cons	.0125072	.0055937	2.24	0.025	.0015438	.0234706
SDH						
SET						
L1.	.025998	.0164886	1.58	0.115	-.006319	.058315
L2.	.004004	.0166165	0.24	0.810	-.0285637	.0365717
L3.	.0018399	.0164089	0.11	0.911	-.030321	.0340008
SDH						
L1.	.2335069	.0982823	2.38	0.018	.0408771	.4261366
L2.	.0843085	.103271	0.82	0.414	-.1180989	.2867159
L3.	-.4170599	.1044401	-3.99	0.000	-.6217588	-.212361
MFP	-.0038434	.0045316	-0.85	0.396	-.0127251	.0050383
_cons	.0027449	.0009508	2.89	0.004	.0008815	.0046084

```
. var SET TWH,lag(1/3) exog(MFP)
```

Vector autoregression

```
Sample: 4 - 97                      No. of obs   =      94
Log likelihood = 498.8152            AIC          = -10.27266
FPE           = 1.19e-07            HQIC        = -10.0978
Det(Sigma_ml) = 8.43e-08            SBIC        = -9.839763
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.047037	0.1358	14.77624	0.0390
TWH	8	.006758	0.2972	39.75427	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1623455	.0957998	1.69	0.090	-.0254187	.3501097
L2.	-.1581798	.0962151	-1.64	0.100	-.3467579	.0303982
L3.	.1482545	.0951471	1.56	0.119	-.0382305	.3347394
TWH						
L1.	-.7392676	.6680702	-1.11	0.268	-2.048661	.5701258
L2.	.3478719	.7183442	0.48	0.628	-1.060057	1.755801
L3.	-1.599247	.672257	-2.38	0.017	-2.916847	-.2816479
MFP	-.0085424	.0268265	-0.32	0.750	-.0611215	.0440366
_cons	.0194319	.006222	3.12	0.002	.007237	.0316269
TWH						
SET						
L1.	.0001381	.013764	0.01	0.992	-.0268388	.027115
L2.	-.013718	.0138236	-0.99	0.321	-.0408118	.0133759
L3.	.0029859	.0136702	0.22	0.827	-.0238073	.029779
TWH						
L1.	.4076149	.0959845	4.25	0.000	.2194886	.5957411
L2.	.1848406	.1032076	1.79	0.073	-.0174426	.3871238
L3.	-.3897767	.0965861	-4.04	0.000	-.5790819	-.2004714
MFP	.0015914	.0038543	0.41	0.680	-.0059628	.0091457
_cons	.0028726	.0008939	3.21	0.001	.0011205	.0046247

GRANGER CASUALITY Result with exogenous variable of MPI

. vargranger

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	LAND	1.3406	4	0.854
SET	ALL	1.3406	4	0.854
LAND	SET	16.036	4	0.003
LAND	ALL	16.036	4	0.003

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	CONDO	.86814	3	0.833
SET	ALL	.86814	3	0.833
CONDO	SET	1.111	3	0.774
CONDO	ALL	1.111	3	0.774

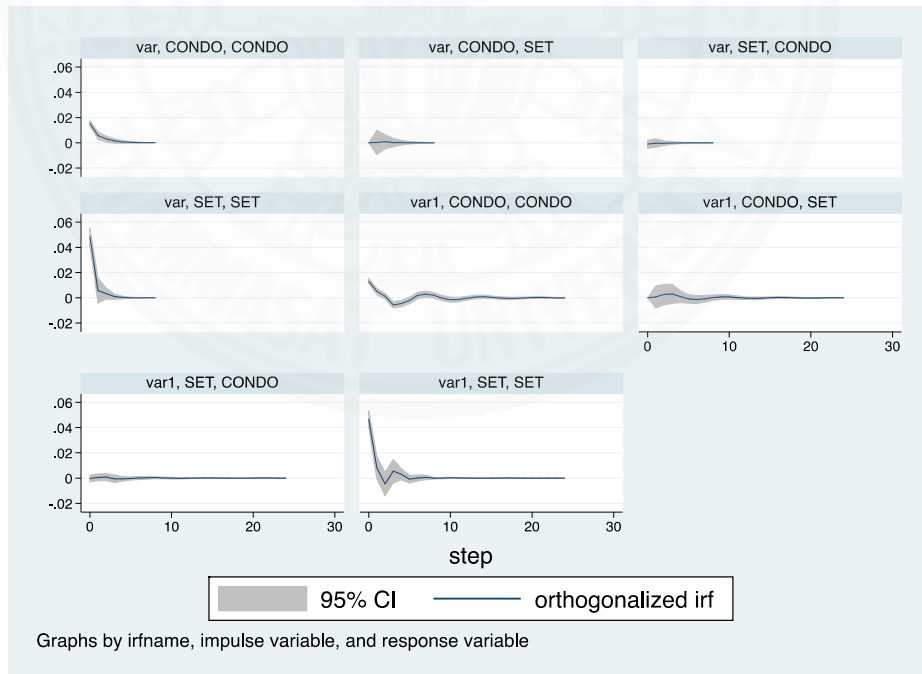
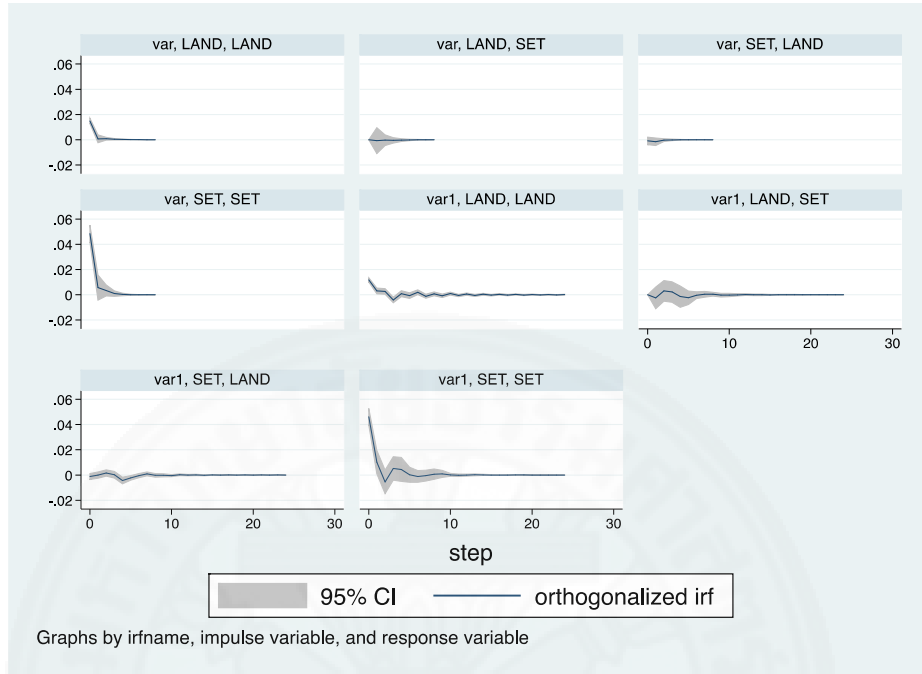
Granger causality Wald tests

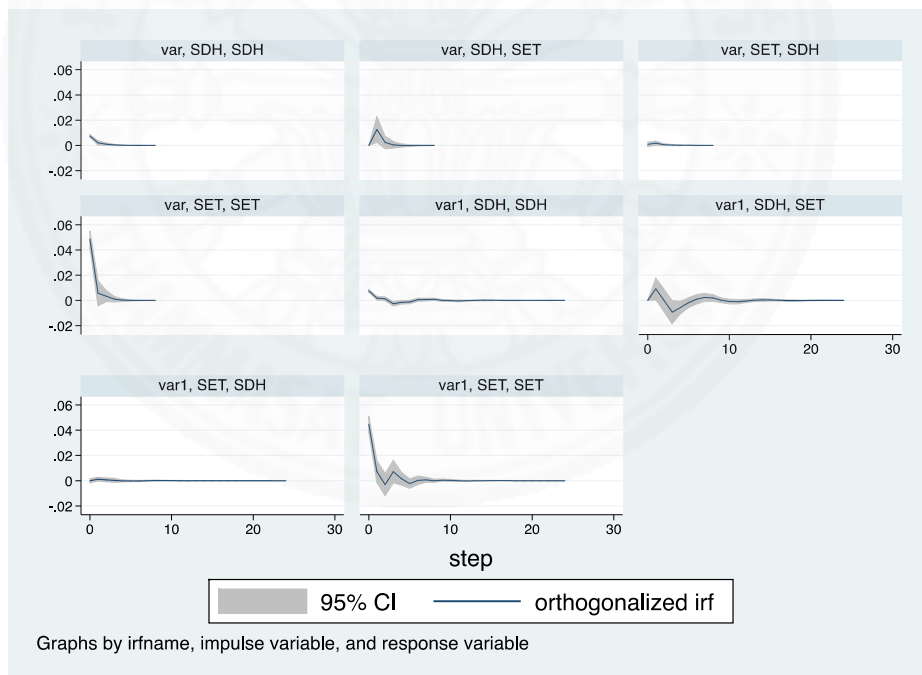
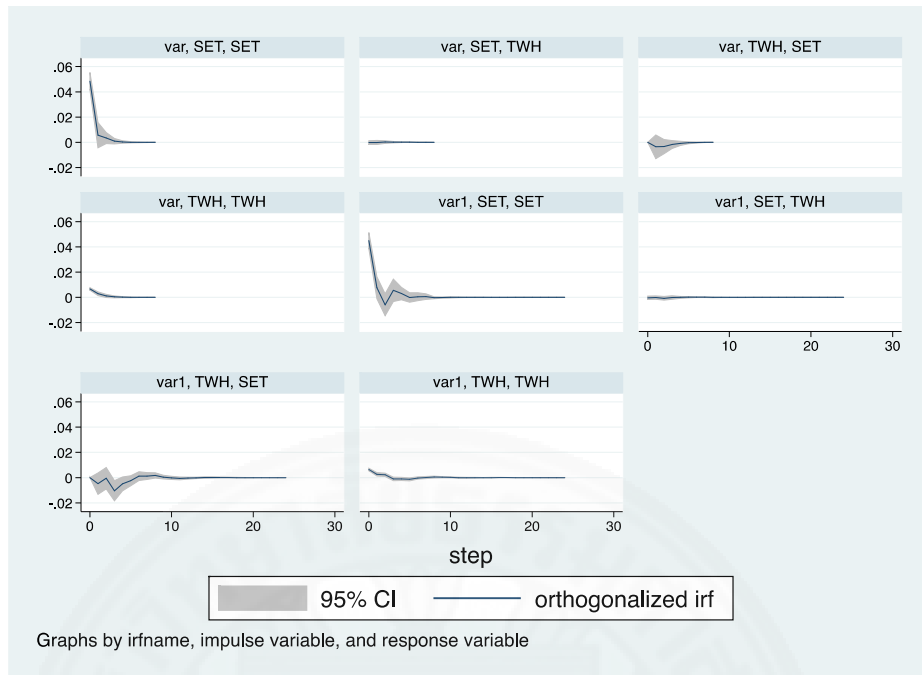
Equation	Excluded	chi2	df	Prob > chi2
SET	SDH	9.3357	3	0.025
SET	ALL	9.3357	3	0.025
SDH	SET	2.6645	3	0.446
SDH	ALL	2.6645	3	0.446

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	TWH	7.9308	3	0.047
SET	ALL	7.9308	3	0.047
TWH	SET	1.0073	3	0.799
TWH	ALL	1.0073	3	0.799

IRF Result with exogenous variable of MPI





APPENDIX B

EXOGENOUS VARIABLE OF MPI AND R

VAR Result with exogenous variable of MPI and R

```
. var SET LAND,lag(1/4) exog(MFP R)
```

Vector autoregression

```
Sample: 5 - 97                      No. of obs   =      93
Log likelihood = 439.0477             AIC          = -8.968768
FPE            = 4.37e-07             HQIC         = -8.726865
Det(Sigma_ml) = 2.72e-07             SBIC         = -8.369659
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	11	.048682	0.1104	11.54393	0.3167
LAND	11	.012264	0.3810	57.24564	0.0000

		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET							
SET							
	L1.	.2161972	.1015005	2.13	0.033	.01726	.4151345
	L2.	-.1402215	.1003819	-1.40	0.162	-.3369665	.0565235
	L3.	.1903392	.100597	1.89	0.058	-.0068273	.3875056
	L4.	.0023403	.1010735	0.02	0.982	-.1957601	.2004407
	LAND						
	L1.	-.2561711	.3706451	-0.69	0.489	-.9826222	.47028
	L2.	.2857135	.3416203	0.84	0.403	-.38385	.9552769
	L3.	-.0146157	.33358	-0.04	0.965	-.6684206	.6391892
	L4.	-.3122862	.3811575	-0.82	0.413	-1.059341	.4347687
	MFP	-.0270375	.0283733	-0.95	0.341	-.0826482	.0285732
	R	-.0827225	.0639181	-1.29	0.196	-.2079997	.0425547
	_cons	.0139406	.0069692	2.00	0.045	.0002813	.0276
LAND							
LAND							
	SET						
	L1.	.0054046	.0255697	0.21	0.833	-.0447112	.0555203
	L2.	.0457033	.0252879	1.81	0.071	-.0038602	.0952668
	L3.	-.0208162	.0253421	-0.82	0.411	-.0704858	.0288535
	L4.	-.0823685	.0254622	-3.23	0.001	-.1322734	-.0324635
	LAND						
	L1.	.2376641	.0933719	2.55	0.011	.0546585	.4206698
	L2.	.1296964	.0860601	1.51	0.132	-.0389782	.2983711
	L3.	-.4746215	.0840346	-5.65	0.000	-.6393263	-.3099168
	L4.	.2004423	.0960202	2.09	0.037	.0122462	.3886384
	MFP	.0111963	.0071477	1.57	0.117	-.002813	.0252056
	R	-.0376043	.0161021	-2.34	0.020	-.0691638	-.0060448
	_cons	.0062633	.0017557	3.57	0.000	.0028223	.0097044

. var SET CONDO, lag(1/3) exog(MFP R)

Vector autoregression

Sample: 4 - 97
 No. of obs = 94
 Log likelihood = 430.4176
 AIC = -8.774842
 FPE = 5.31e-07
 HQIC = -8.578124
 Det(Sigma_ml) = 3.61e-07
 SBIC = -8.287828

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	9	.048666	0.0857	8.812041	0.3584
CONDO	9	.01366	0.4023	63.26131	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1770932	.0980383	1.81	0.071	-.0150583	.3692447
L2.	-.1148474	.0994117	-1.16	0.248	-.3096908	.0799959
L3.	.1675078	.0982162	1.71	0.088	-.0249923	.360008
CONDO						
L1.	.0458181	.3331791	0.14	0.891	-.6072009	.6988371
L2.	.1134562	.3578137	0.32	0.751	-.5878457	.814758
L3.	.1545065	.3289146	0.47	0.639	-.4901543	.7991672
MFP	-.0156963	.0281923	-0.56	0.578	-.0709523	.0395597
R	-.0751702	.0622047	-1.21	0.227	-.1970891	.0467488
_cons	.0097961	.0059447	1.65	0.099	-.0018553	.0214475
CONDO						
SET						
L1.	.0143766	.0275183	0.52	0.601	-.0395582	.0683115
L2.	.0051715	.0279038	0.19	0.853	-.0495189	.0598619
L3.	-.0251689	.0275682	-0.91	0.361	-.0792016	.0288637
CONDO						
L1.	.389309	.0935197	4.16	0.000	.2060137	.5726042
L2.	-.0361792	.1004343	-0.36	0.719	-.2330269	.1606685
L3.	-.4601595	.0923227	-4.98	0.000	-.6411087	-.2792104
MFP	.0051702	.0079133	0.65	0.514	-.0103396	.0206799
R	.0215199	.0174602	1.23	0.218	-.0127014	.0557413
_cons	.0065435	.0016686	3.92	0.000	.0032731	.0098139

```
. var SET SDH,lag(1/4) exog(MFP R)
```

Vector autoregression

```
Sample: 5 - 97                      No. of obs   =      93
Log likelihood = 483.153              AIC          = -9.917269
FPE            = 1.69e-07             HQIC        = -9.675366
Det(Sigma_ml) = 1.05e-07             SBIC        = -9.31816
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	11	.046307	0.1951	22.54315	0.0126
SDH	11	.007952	0.2844	36.95774	0.0001

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1620331	.1003752	1.61	0.106	-.0346986	.3587648
L2.	-.1282607	.0984714	-1.30	0.193	-.3212611	.0647398
L3.	.2080906	.0958168	2.17	0.030	.020293	.3958881
L4.	.0829104	.0968944	0.86	0.392	-.1069991	.27282
SDH						
L1.	.837311	.6063118	1.38	0.167	-.3510383	2.02566
L2.	-.4074391	.5942867	-0.69	0.493	-1.57222	.7573415
L3.	-.9707124	.6342785	-1.53	0.126	-2.213875	.2724506
L4.	-.9568599	.6591811	-1.45	0.147	-2.248831	.3351113
MFP	-.0147613	.0264124	-0.56	0.576	-.0665287	.0370061
R	-.0484502	.058824	-0.82	0.410	-.1637431	.0668426
_cons	.0158519	.0059136	2.68	0.007	.0042615	.0274424
SDH						
SET						
L1.	.0273755	.0172363	1.59	0.112	-.0064071	.0611581
L2.	.0082168	.0169094	0.49	0.627	-.0249251	.0413586
L3.	-.0007174	.0164536	-0.04	0.965	-.0329658	.031531
L4.	.0210017	.0166386	1.26	0.207	-.0116094	.0536128
SDH						
L1.	.2448795	.1041153	2.35	0.019	.0408172	.4489418
L2.	.0968474	.1020504	0.95	0.343	-.1031677	.2968625
L3.	-.4573558	.1089178	-4.20	0.000	-.6708307	-.2438809
L4.	.0801302	.113194	0.71	0.479	-.141726	.3019863
MFP	-.0045049	.0045355	-0.99	0.321	-.0133944	.0043845
R	-.0133279	.0101012	-1.32	0.187	-.0331259	.00647
_cons	.0024144	.0010155	2.38	0.017	.000424	.0044047

. var SET TWH,lag(1/4) exog(MFP R)

Vector autoregression

Sample:	5 - 97	No. of obs	=	93
Log likelihood =	500.4197	AIC	=	-10.2886
FPE	= 1.17e-07	HQIC	=	-10.04669
Det(Sigma_ml)	= 7.27e-08	SBIC	=	-9.689486

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	11	.046635	0.1837	20.92313	0.0216
TWH	11	.006591	0.3575	51.73977	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1862463	.1011448	1.84	0.066	-.0119939	.3844864
L2.	-.1565188	.0965793	-1.62	0.105	-.3458107	.0327732
L3.	.1606453	.0956372	1.68	0.093	-.0268001	.3480907
L4.	.0301467	.0948527	0.32	0.751	-.1557611	.2160545
TWH						
L1.	-.8819281	.722625	-1.22	0.222	-2.298247	.5343909
L2.	.3534488	.7194618	0.49	0.623	-1.05667	1.763568
L3.	-1.647912	.7205298	-2.29	0.022	-3.060124	-.2356993
L4.	-.2062271	.7678735	-0.27	0.788	-1.711232	1.298777
MFP	-.0154918	.0273271	-0.57	0.571	-.069052	.0380684
R	-.1144654	.0607811	-1.88	0.060	-.2335941	.0046633
_cons	.0221642	.0071516	3.10	0.002	.0081473	.0361811
TWH						
SET						
L1.	.0074804	.0142938	0.52	0.601	-.0205349	.0354958
L2.	-.009249	.0136486	-0.68	0.498	-.0359998	.0175018
L3.	.0110694	.0135155	0.82	0.413	-.0154204	.0375593
L4.	-.0109013	.0134046	-0.81	0.416	-.0371739	.0153712
TWH						
L1.	.4613589	.1021216	4.52	0.000	.2612042	.6615135
L2.	.1272869	.1016746	1.25	0.211	-.0719916	.3265654
L3.	-.4751199	.1018255	-4.67	0.000	-.6746942	-.2755456
L4.	.1907651	.1085161	1.76	0.079	-.0219226	.4034528
MFP	.0038273	.0038619	0.99	0.322	-.0037419	.0113964
R	-.0188111	.0085896	-2.19	0.029	-.0356464	-.0019758
_cons	.0025207	.0010107	2.49	0.013	.0005399	.0045016

GRANGER CASUALITY Result with exogenous variable of MPI and R

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	LAND	1.2582	4	0.868
SET	ALL	1.2582	4	0.868
LAND	SET	16.723	4	0.002
LAND	ALL	16.723	4	0.002

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	CONDO	.66236	3	0.882
SET	ALL	.66236	3	0.882
CONDO	SET	1.2352	3	0.745
CONDO	ALL	1.2352	3	0.745

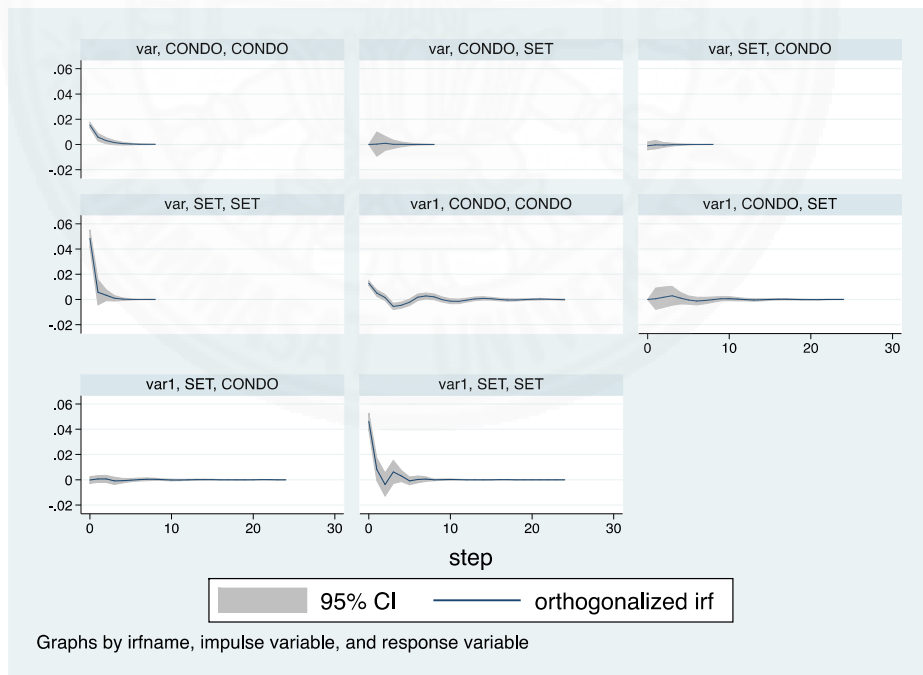
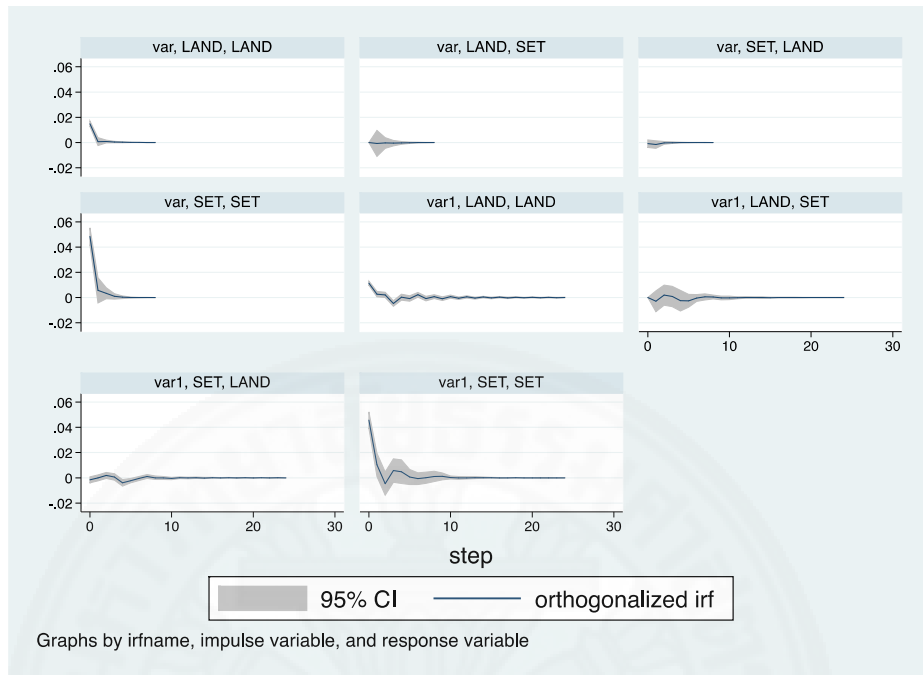
Granger causality Wald tests

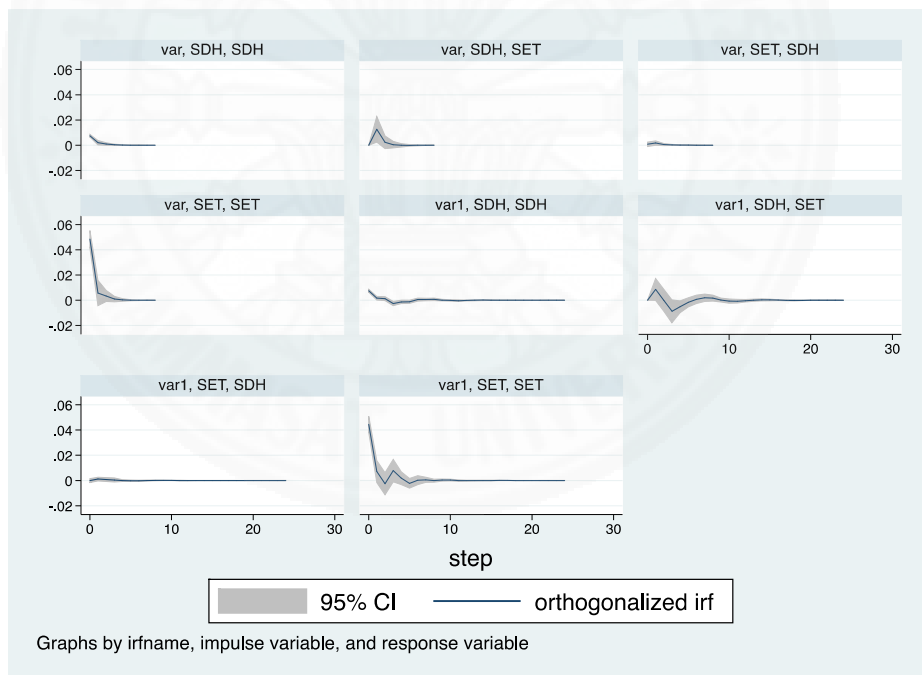
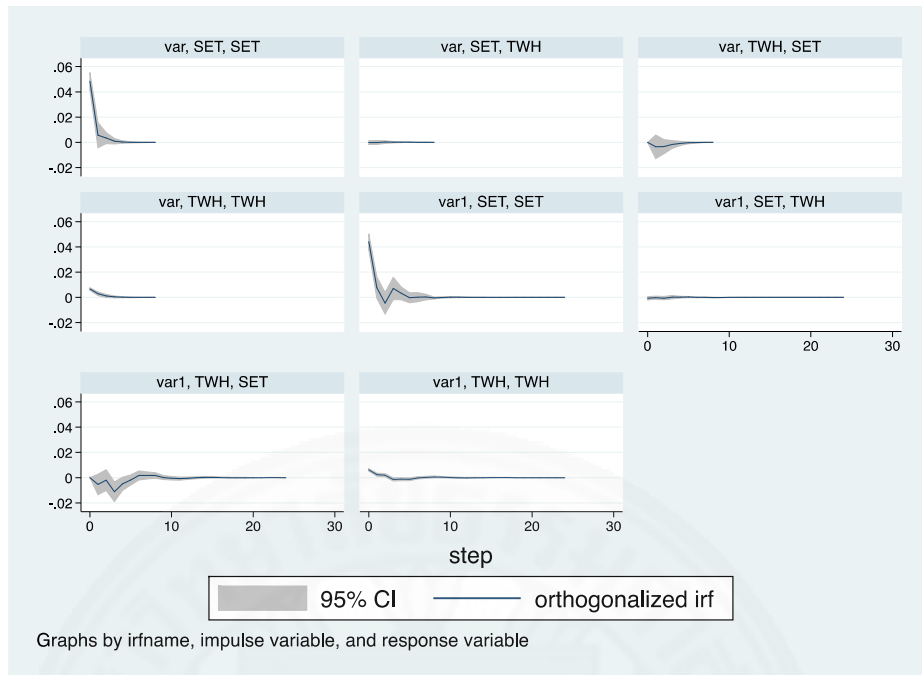
Equation	Excluded	chi2	df	Prob > chi2
SET	SDH	8.4158	3	0.038
SET	ALL	8.4158	3	0.038
SDH	SET	2.9202	3	0.404
SDH	ALL	2.9202	3	0.404

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	TWH	9.9973	3	0.019
SET	ALL	9.9973	3	0.019
TWH	SET	.61094	3	0.894
TWH	ALL	.61094	3	0.894

IRF Result with exogenous variable of MPI and R





APPENDIX C

EXOGENOUS VARIABLE OF R

VAR Result with exogenous variable of R

```
. var SET LAND,lag(1/4) exog(R)
```

Vector autoregression

```
Sample: 5 - 97                               No. of obs   =      93
Log likelihood = 437.5665                     AIC          = -8.979924
FPE            = 4.32e-07                     HQIC        = -8.760012
Det(Sigma_ml) = 2.81e-07                     SBIC        = -8.435279
```

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	10	.048624	0.1017	10.53303	0.3091
LAND	10	.01235	0.3647	53.38353	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.2158618	.1019942	2.12	0.034	.0159569	.4157667
L2.	-.1328243	.1005687	-1.32	0.187	-.3299354	.0642868
L3.	.1875342	.1010436	1.86	0.063	-.0105077	.385576
L4.	-.001988	.1014631	-0.02	0.984	-.2008521	.196876
LAND						
L1.	-.2510017	.3724103	-0.67	0.500	-.9809126	.4789091
L2.	.2355854	.3391897	0.69	0.487	-.4292142	.9003849
L3.	-.0139839	.335204	-0.04	0.967	-.6709716	.6430038
L4.	-.2426977	.3759183	-0.65	0.519	-.9794841	.4940886
R	-.0850036	.0641844	-1.32	0.185	-.2108026	.0407954
_cons	.0133813	.0069782	1.92	0.055	-.0002958	.0270584
LAND						
SET						
L1.	.0055435	.0259047	0.21	0.831	-.0452287	.0563158
L2.	.0426401	.0255427	1.67	0.095	-.0074226	.0927028
L3.	-.0196546	.0256633	-0.77	0.444	-.0699537	.0306445
L4.	-.0805761	.0257698	-3.13	0.002	-.131084	-.0300682
LAND						
L1.	.2355235	.0945855	2.49	0.013	.0501392	.4209077
L2.	.1504547	.0861481	1.75	0.081	-.0183924	.3193018
L3.	-.4748832	.0851358	-5.58	0.000	-.6417462	-.3080201
L4.	.1716254	.0954765	1.80	0.072	-.015505	.3587559
R	-.0366596	.0163017	-2.25	0.025	-.0686103	-.004709
_cons	.006495	.0017723	3.66	0.000	.0030212	.0099687

. var SET CONDO, lag(1/3) exog(R)

Vector autoregression

Sample: 4 - 97 No. of obs = 94
 Log likelihood = 430.0553 AIC = -8.809688
 FPE = 5.12e-07 HQIC = -8.634828
 Det(Sigma_ml) = 3.64e-07 SBIC = -8.376787

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.048462	0.0827	8.474118	0.2926
CONDO	8	.013611	0.3996	62.55039	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1788918	.0981465	1.82	0.068	-.0134718	.3712554
L2.	-.1104976	.0992675	-1.11	0.266	-.3050583	.0840631
L3.	.1665501	.0983629	1.69	0.090	-.0262377	.3593378
CONDO						
L1.	.0055578	.3257731	0.02	0.986	-.6329456	.6440613
L2.	.1220249	.3580715	0.34	0.733	-.5797823	.823832
L3.	.1373596	.328009	0.42	0.675	-.5055263	.7802455
R	-.0771065	.0622097	-1.24	0.215	-.1990353	.0448223
_cons	.0098483	.0059537	1.65	0.098	-.0018208	.0215174
CONDO						
SET						
L1.	.0137842	.0275657	0.50	0.617	-.0402436	.067812
L2.	.0037387	.0278805	0.13	0.893	-.0509061	.0583836
L3.	-.0248535	.0276265	-0.90	0.368	-.0790004	.0292935
CONDO						
L1.	.4025702	.0914975	4.40	0.000	.2232383	.5819021
L2.	-.0390016	.100569	-0.39	0.698	-.2361132	.1581099
L3.	-.4545116	.0921255	-4.93	0.000	-.6350743	-.2739488
R	.0221577	.0174724	1.27	0.205	-.0120875	.056403
_cons	.0065263	.0016722	3.90	0.000	.0032489	.0098037

. var SET SDH,lag(1/3) exog(R)

Vector autoregression

Sample: 4 - 97
 Log likelihood = 484.8961
 FPE = 1.59e-07
 Det(Sigma_ml) = 1.13e-07
 No. of obs = 94
 AIC = -9.976513
 HQIC = -9.801652
 SBIC = -9.543612

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.046525	0.1546	17.18684	0.0162
SDH	8	.007911	0.2573	32.56255	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1641067	.0965798	1.70	0.089	-.0251863	.3533997
L2.	-.1126213	.097921	-1.15	0.250	-.3045429	.0793004
L3.	.1941802	.0963717	2.01	0.044	.005295	.3830653
SDH						
L1.	1.137584	.5804069	1.96	0.050	7.25e-06	2.27516
L2.	-.4424961	.6066783	-0.73	0.466	-1.631564	.7465715
L3.	-1.145455	.6082711	-1.88	0.060	-2.337645	.0467343
R	-.0544389	.0598028	-0.91	0.363	-.1716502	.0627725
_cons	.0127729	.0055768	2.29	0.022	.0018426	.0237032
SDH						
SET						
L1.	.0267164	.0164217	1.63	0.104	-.0054695	.0589023
L2.	.0078802	.0166497	0.47	0.636	-.0247527	.0405131
L3.	.0031134	.0163863	0.19	0.849	-.0290031	.03523
SDH						
L1.	.219676	.0986879	2.23	0.026	.0262513	.4131007
L2.	.093065	.1031549	0.90	0.367	-.1091148	.2952449
L3.	-.4203822	.1034257	-4.06	0.000	-.6230928	-.2176715
R	-.0121774	.0101684	-1.20	0.231	-.0321071	.0077524
_cons	.0027823	.0009482	2.93	0.003	.0009238	.0046408

. var SET TWH,lag(1/3) exog(R)

Vector autoregression

Sample:	4 - 97	No. of obs	=	94
Log likelihood	= 503.6808	AIC	=	-10.37619
FPE	= 1.07e-07	HQIC	=	-10.20133
Det(Sigma_ml)	= 7.60e-08	SBIC	=	-9.943286

Equation	Parms	RMSE	R-sq	chi2	P>chi2
SET	8	.046175	0.1672	18.87665	0.0086
TWH	8	.006566	0.3365	47.68347	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
SET						
SET						
L1.	.1616104	.0939809	1.72	0.086	-.0225888	.3458096
L2.	-.1329093	.0951231	-1.40	0.162	-.3193471	.0535285
L3.	.1622023	.0936933	1.73	0.083	-.0214331	.3458377
TWH						
L1.	-.8696395	.6588531	-1.32	0.187	-2.160968	.4216889
L2.	.1745914	.7089447	0.25	0.805	-1.214915	1.564097
L3.	-1.68564	.6570304	-2.57	0.010	-2.973396	-.3978836
R	-.1158745	.0606629	-1.91	0.056	-.2347715	.0030226
_cons	.0216196	.0062156	3.48	0.001	.0094373	.0338019
TWH						
SET						
L1.	-.0004165	.0133644	-0.03	0.975	-.0266102	.0257771
L2.	-.009805	.0135268	-0.72	0.469	-.036317	.016707
L3.	.0057198	.0133235	0.43	0.668	-.0203937	.0318333
TWH						
L1.	.3863489	.0936908	4.12	0.000	.2027182	.5699795
L2.	.1452869	.100814	1.44	0.150	-.0523049	.3428787
L3.	-.3968911	.0934316	-4.25	0.000	-.5800137	-.2137684
R	-.020691	.0086264	-2.40	0.016	-.0375985	-.0037835
_cons	.0033058	.0008839	3.74	0.000	.0015734	.0050381

GRANGER CASUALITY Result with exogenous variable of R

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	LAND	.93176	4	0.920
SET	ALL	.93176	4	0.920
LAND	SET	15.328	4	0.004
LAND	ALL	15.328	4	0.004

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	CONDO	.5704	3	0.903
SET	ALL	.5704	3	0.903
CONDO	SET	1.1708	3	0.760
CONDO	ALL	1.1708	3	0.760

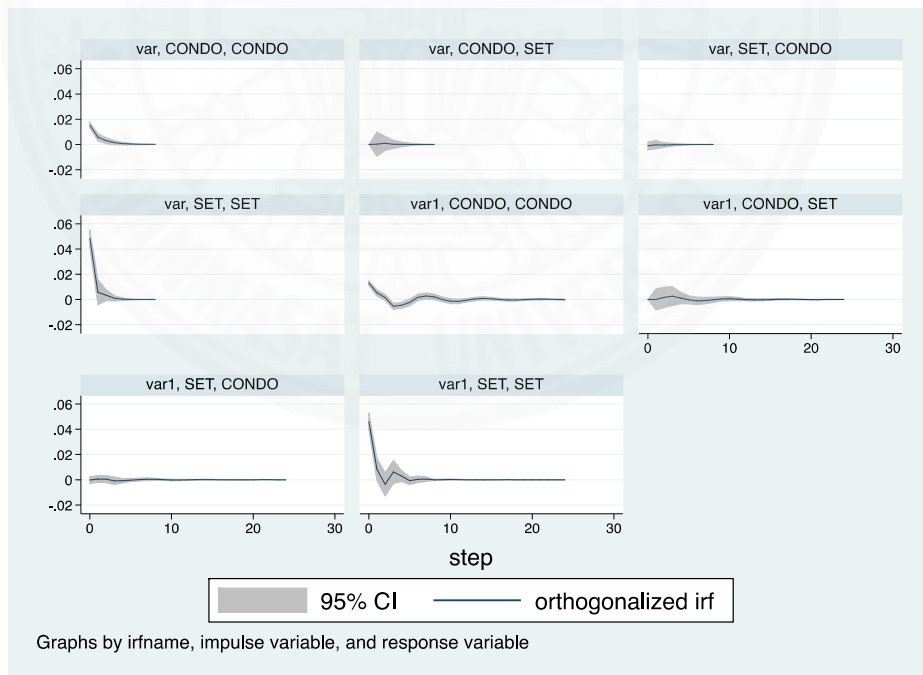
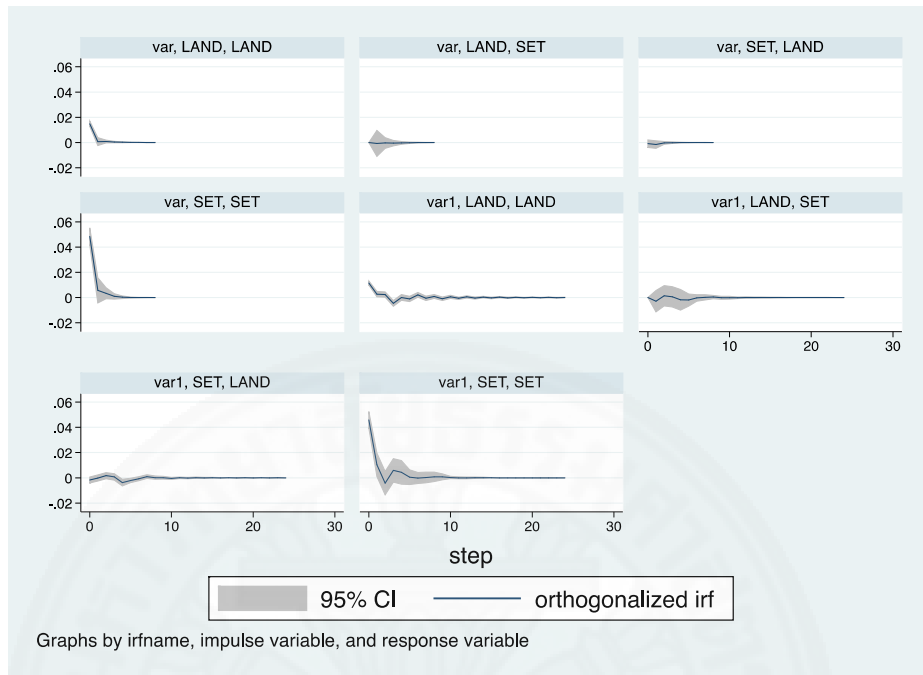
Granger causality Wald tests

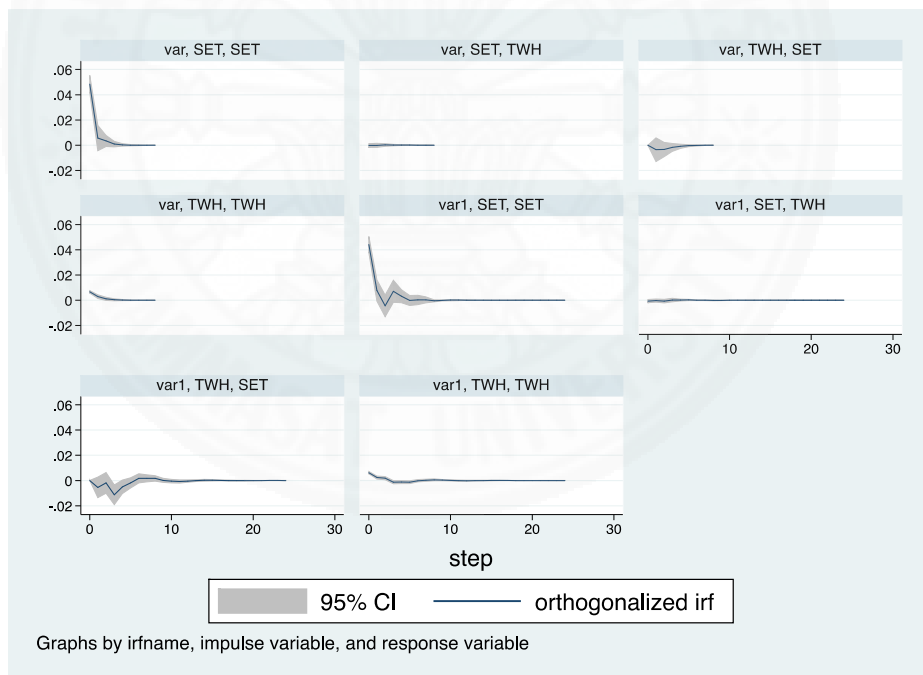
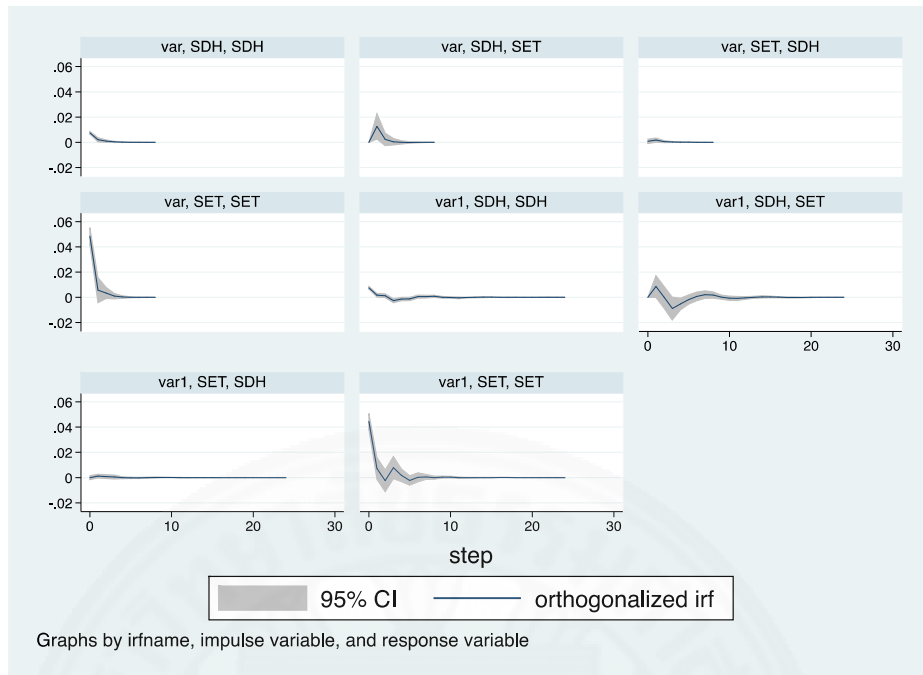
Equation	Excluded	chi2	df	Prob > chi2
SET	SDH	8.6111	3	0.035
SET	ALL	8.6111	3	0.035
SDH	SET	3.0803	3	0.379
SDH	ALL	3.0803	3	0.379

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
SET	TWH	10.171	3	0.017
SET	ALL	10.171	3	0.017
TWH	SET	.66613	3	0.881
TWH	ALL	.66613	3	0.881

IRF Result with exogenous variable of R





. irf table fevd, impulse (SET LAND) response (SET LAND)

Results from var var1

step	(1) fevd	(1) Lower	(1) Upper	(2) fevd	(2) Lower	(2) Upper	(3) fevd	(3) Lower	(3) Upper
0	0	0	0	0	0	0	0	0	0
1	1	1	1	.003849	-.020876	.028574	0	0	0
2	.918649	.815714	1.02158	.014591	-.033206	.062389	.000194	-.005408	.005795
3	.910793	.800043	1.02154	.014766	-.033784	.063315	.000244	-.005799	.006286
4	.909594	.797269	1.02192	.014804	-.034048	.063655	.000341	-.006232	.006915
5	.909276	.796491	1.02206	.014803	-.034078	.063684	.000384	-.006347	.007115
6	.909208	.796318	1.0221	.014803	-.034082	.063687	.000394	-.006372	.007161
7	.909194	.796282	1.02211	.014802	-.034082	.063687	.000396	-.006377	.00717
8	.909192	.796275	1.02211	.014802	-.034082	.063687	.000397	-.006378	.007172
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step	(4) fevd	(4) Lower	(4) Upper	(5) fevd	(5) Lower	(5) Upper	(6) fevd	(6) Lower	(6) Upper
0	0	0	0	0	0	0	0	0	0
1	.996151	.971426	1.02088	1	1	1	.022469	-.037092	.08203
2	.945617	.860275	1.03096	.996234	.974372	1.0181	.021482	-.036477	.079441
3	.939494	.845663	1.03332	.995381	.974678	1.01608	.039647	-.027974	.107268
4	.937925	.841061	1.03479	.995306	.973503	1.01711	.037675	-.033469	.108818
5	.937588	.839953	1.03522	.994057	.968985	1.01913	.113192	.001959	.224424
6	.937511	.839672	1.03535	.992367	.960246	1.02449	.132873	.007822	.257925
7	.937494	.839602	1.03539	.992328	.95991	1.02475	.132034	.008215	.255852
8	.93749	.839585	1.03539	.992312	.959731	1.02489	.135518	.006902	.264135
9992256	.959513	1.025	.135021	.006796	.263246
10992252	.959496	1.02501	.134523	.006652	.262395
11992247	.959435	1.02506	.135259	.006583	.263936
12992234	.959404	1.02506	.135061	.006511	.263612
13992233	.959404	1.02506	.135006	.006465	.263548
14992233	.959402	1.02506	.134984	.006468	.2635
15992232	.959407	1.02506	.134948	.006406	.263489
16992231	.959405	1.02506	.134916	.006378	.263455
17992231	.959405	1.02506	.134905	.006375	.263435
18992231	.959406	1.02506	.134888	.006357	.26342
19992231	.959406	1.02506	.134881	.006345	.263417
20992231	.959405	1.02506	.134873	.006341	.263405
2199223	.959406	1.02506	.134868	.006335	.263401
2299223	.959406	1.02506	.134864	.00633	.263398
2399223	.959406	1.02506	.134861	.006328	.263395
2499223	.959406	1.02505	.134859	.006326	.263393

. irf table fevd, impulse (SET LAND) response (SET LAND)

Results from var var1

step	(1)			(2)			(3)		
	fevd	Lower	Upper	fevd	Lower	Upper	fevd	Lower	Upper
0	0	0	0	0	0	0	0	0	0
1	1	1	1	.003849	-.020876	.028574	0	0	0
2	.918649	.815714	1.02158	.014591	-.033206	.062389	.000194	-.005408	.005795
3	.910793	.800043	1.02154	.014766	-.033784	.063315	.000244	-.005799	.006286
4	.909594	.797269	1.02192	.014804	-.034048	.063655	.000341	-.006232	.006915
5	.909276	.796491	1.02206	.014803	-.034078	.063684	.000384	-.006347	.007115
6	.909208	.796318	1.0221	.014803	-.034082	.063687	.000394	-.006372	.007161
7	.909194	.796282	1.02211	.014802	-.034082	.063687	.000396	-.006377	.00717
8	.909192	.796275	1.02211	.014802	-.034082	.063687	.000397	-.006378	.007172
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step	(4)			(5)			(6)		
	fevd	Lower	Upper	fevd	Lower	Upper	fevd	Lower	Upper
0	0	0	0	0	0	0	0	0	0
1	.996151	.971426	1.02088	1	1	1	.022469	-.037092	.08203
2	.945617	.860275	1.03096	.996234	.974372	1.0181	.021482	-.036477	.079441
3	.939494	.845663	1.03332	.995381	.974678	1.01608	.039647	-.027974	.107268
4	.937925	.841061	1.03479	.995306	.973503	1.01711	.037675	-.033469	.108818
5	.937588	.839953	1.03522	.994057	.968985	1.01913	.113192	.001959	.224424
6	.937511	.839672	1.03535	.992367	.960246	1.02449	.132873	.007822	.257925
7	.937494	.839602	1.03539	.992328	.95991	1.02475	.132034	.008215	.255852
8	.93749	.839585	1.03539	.992312	.959731	1.02489	.135518	.006902	.264135
9992256	.959513	1.025	.135021	.006796	.263246
10992252	.959496	1.02501	.134523	.006652	.262395
11992247	.959435	1.02506	.135259	.006583	.263936
12992234	.959404	1.02506	.135061	.006511	.263612
13992233	.959404	1.02506	.135006	.006465	.263548
14992233	.959402	1.02506	.134984	.006468	.2635
15992232	.959407	1.02506	.134948	.006406	.263489
16992231	.959405	1.02506	.134916	.006378	.263455
17992231	.959405	1.02506	.134905	.006375	.263435
18992231	.959406	1.02506	.134888	.006357	.26342
19992231	.959406	1.02506	.134881	.006345	.263417
20992231	.959405	1.02506	.134873	.006341	.263405
2199223	.959406	1.02506	.134868	.006335	.263401
2299223	.959406	1.02506	.134864	.00633	.263398
2399223	.959406	1.02506	.134861	.006328	.263395
2499223	.959406	1.02505	.134859	.006326	.263393

FEVD Result with Exogenous Variable of R

step	(7) fevd	(7) Lower	(7) Upper	(8) fevd	(8) Lower	(8) Upper
0	0	0	0	0	0	0
1	0	0	0	.977531	.91797	1.03709
2	.003766	-.018096	.025628	.978518	.920559	1.03648
3	.004619	-.016085	.025322	.960353	.892732	1.02797
4	.004694	-.01711	.026497	.962325	.891182	1.03347
5	.005943	-.019129	.031015	.886808	.775576	.998041
6	.007633	-.024489	.039754	.867127	.742075	.992178
7	.007672	-.024746	.04009	.867966	.744148	.991785
8	.007688	-.024892	.040269	.864482	.735865	.993098
9	.007744	-.024999	.040487	.864979	.736754	.993204
10	.007748	-.025007	.040504	.865477	.737605	.993348
11	.007753	-.025059	.040565	.864741	.736064	.993417
12	.007766	-.025063	.040596	.864939	.736388	.993489
13	.007767	-.025062	.040596	.864994	.736452	.993535
14	.007767	-.025063	.040598	.865016	.7365	.993532
15	.007768	-.025057	.040593	.865052	.736511	.993594
16	.007769	-.025058	.040595	.865084	.736545	.993622
17	.007769	-.025057	.040595	.865095	.736565	.993625
18	.007769	-.025056	.040594	.865112	.73658	.993643
19	.007769	-.025056	.040594	.865119	.736583	.993655
20	.007769	-.025056	.040595	.865127	.736595	.993659
21	.00777	-.025055	.040594	.865132	.736599	.993665
22	.00777	-.025055	.040594	.865136	.736602	.99367
23	.00777	-.025055	.040594	.865139	.736605	.993672
24	.00777	-.025055	.040594	.865141	.736607	.993674

95% lower and upper bounds reported

- (1) irfname = var, impulse = SET, and response = SET
- (2) irfname = var, impulse = SET, and response = LAND
- (3) irfname = var, impulse = LAND, and response = SET
- (4) irfname = var, impulse = LAND, and response = LAND
- (5) irfname = var1, impulse = SET, and response = SET
- (6) irfname = var1, impulse = SET, and response = LAND
- (7) irfname = var1, impulse = LAND, and response = SET
- (8) irfname = var1, impulse = LAND, and response = LAND

. irf table fevd, impulse (SET CONDO) response (SET CONDO)

Results from var var1

step	(1) fevd	(1) Lower	(1) Upper	(2) fevd	(2) Lower	(2) Upper	(3) fevd	(3) Lower	(3) Upper
0	0	0	0	0	0	0	0	0	0
1	1	1	1	.004917	-.022998	.032831	0	0	0
2	.918649	.815714	1.02158	.004312	-.021604	.030228	.000034	-.002148	.002215
3	.910793	.800043	1.02154	.004659	-.023011	.03233	.000332	-.005206	.005869
4	.909594	.797269	1.02192	.00475	-.023544	.033045	.000335	-.005422	.006092
5	.909276	.796491	1.02206	.004787	-.023726	.0333	.000337	-.005515	.006188
6	.909208	.796318	1.0221	.004798	-.023776	.033373	.000337	-.005526	.006199
7	.909194	.796282	1.02211	.004802	-.023789	.033393	.000337	-.005527	.006201
8	.909192	.796275	1.02211	.004803	-.023793	.033398	.000337	-.005527	.006201
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step	(4) fevd	(4) Lower	(4) Upper	(5) fevd	(5) Lower	(5) Upper	(6) fevd	(6) Lower	(6) Upper
0	0	0	0	0	0	0	0	0	0
1	.984318	.935003	1.03363	1	1	1	.000294	-.006637	.007225
2	.935363	.832733	1.03799	.999998	.999455	1.00054	.001781	-.012145	.015707
3	.92933	.817973	1.04069	.998807	.986955	1.01066	.003093	-.020318	.026505
4	.927917	.813932	1.0419	.995585	.972266	1.0189	.006067	-.019686	.03182
5	.927565	.812844	1.04229	.994973	.968923	1.02102	.007832	-.026356	.042021
6	.927479	.812558	1.0424	.994939	.968687	1.02119	.008133	-.027469	.043735
7	.927458	.812482	1.04243	.994457	.965677	1.02324	.008121	-.027556	.043798
8	.927452	.812462	1.04244	.99413	.963718	1.02454	.008313	-.028626	.045251
9994112	.963621	1.0246	.008543	-.029515	.046602
1099402	.963051	1.02499	.008551	-.029527	.046628
11993862	.962097	1.02563	.008591	-.029757	.046939
12993819	.961843	1.0258	.008675	-.030121	.04747
13993815	.961817	1.02581	.008698	-.030199	.047595
14993772	.961555	1.02599	.008698	-.03021	.047605
15993741	.961373	1.02611	.008719	-.030313	.047751
1699374	.961367	1.02611	.008735	-.030377	.047847
17993733	.961322	1.02614	.008736	-.030378	.04785
1899372	.961245	1.02619	.008739	-.030396	.047874
19993717	.961225	1.02621	.008746	-.030425	.047916
20993716	.961223	1.02621	.008748	-.030431	.047926
21993713	.961201	1.02622	.008748	-.030432	.047927
2299371	.961186	1.02623	.008749	-.03044	.047939
2399371	.961186	1.02623	.008751	-.030446	.047947
24993709	.961182	1.02624	.008751	-.030446	.047947

step	(7) fevd	(7) Lower	(7) Upper	(8) fevd	(8) Lower	(8) Upper
0	0	0	0	0	0	0
1	0	0	0	.999706	.992775	1.00664
2	2.4e-06	-.00054	.000545	.998219	.984293	1.01215
3	.001193	-.010659	.013045	.996907	.973495	1.02032
4	.004415	-.018904	.027734	.993933	.96818	1.01969
5	.005027	-.021023	.031077	.992168	.957979	1.02636
6	.005061	-.021192	.031313	.991867	.956265	1.02747
7	.005543	-.023237	.034323	.991879	.956202	1.02756
8	.00587	-.024541	.036282	.991687	.954749	1.02863
9	.005888	-.024604	.036379	.991457	.953398	1.02952
10	.00598	-.024989	.036949	.991449	.953372	1.02953
11	.006138	-.025627	.037903	.991409	.953061	1.02976
12	.006181	-.025796	.038157	.991325	.95253	1.03012
13	.006185	-.025813	.038183	.991302	.952405	1.0302
14	.006228	-.025988	.038445	.991302	.952395	1.03021
15	.006259	-.026109	.038627	.991281	.952249	1.03031
16	.00626	-.026114	.038633	.991265	.952153	1.03038
17	.006267	-.026143	.038678	.991264	.95215	1.03038
18	.00628	-.026195	.038755	.991261	.952126	1.0304
19	.006283	-.026209	.038775	.991254	.952084	1.03042
20	.006284	-.02621	.038777	.991252	.952074	1.03043
21	.006287	-.026224	.038799	.991252	.952073	1.03043
22	.00629	-.026234	.038814	.991251	.952061	1.03044
23	.00629	-.026234	.038814	.991249	.952053	1.03045
24	.006291	-.026237	.038818	.991249	.952053	1.03045

95% lower and upper bounds reported

- (1) irfname = var, impulse = SET, and response = SET
- (2) irfname = var, impulse = SET, and response = CONDO
- (3) irfname = var, impulse = CONDO, and response = SET
- (4) irfname = var, impulse = CONDO, and response = CONDO
- (5) irfname = var1, impulse = SET, and response = SET
- (6) irfname = var1, impulse = SET, and response = CONDO
- (7) irfname = var1, impulse = CONDO, and response = SET
- (8) irfname = var1, impulse = CONDO, and response = CONDO

. irf table fevd, impulse (SET SDH) response (SET SDH)

Results from var var1

step	(1) fevd	(1) Lower	(1) Upper	(2) fevd	(2) Lower	(2) Upper	(3) fevd	(3) Lower	(3) Upper
0	0	0	0	0	0	0	0	0	0
1	1	1	1	.007968	-.027459	.043395	0	0	0
2	.918649	.815714	1.02158	.042502	-.042285	.12729	.060932	-.03284	.154704
3	.910793	.800043	1.02154	.045596	-.045769	.136961	.062197	-.033998	.158391
4	.909594	.797269	1.02192	.046267	-.046718	.139252	.062157	-.034275	.158588
5	.909276	.796491	1.02206	.046364	-.046855	.139584	.062138	-.034205	.15848
6	.909208	.796318	1.0221	.046377	-.046869	.139622	.062141	-.034172	.158455
7	.909194	.796282	1.02211	.046378	-.046869	.139625	.062144	-.034165	.158453
8	.909192	.796275	1.02211	.046378	-.046869	.139625	.062145	-.034164	.158453
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step	(4) fevd	(4) Lower	(4) Upper	(5) fevd	(5) Lower	(5) Upper	(6) fevd	(6) Lower	(6) Upper
0	0	0	0	0	0	0	0	0	0
1	.786279	.640853	.931705	1	1	1	.000159	-.004938	.005256
2	.749729	.593628	.90583	.964827	.895367	1.03429	.022372	-.032462	.077206
3	.748168	.59054	.905797	.964927	.895726	1.03413	.031373	-.040594	.10334
4	.747743	.589983	.905503	.931991	.843111	1.02087	.031001	-.040388	.102389
5	.747631	.589881	.90538	.920612	.81863	1.02259	.030338	-.039139	.099815
6	.747599	.589852	.905346	.919569	.816453	1.02268	.030215	-.039152	.099583
7	.747591	.589844	.905339	.919336	.815856	1.02282	.031004	-.040488	.102495
8	.74759	.589841	.905338	.917547	.811358	1.02374	.030888	-.040441	.102218
9916232	.80788	1.02458	.030687	-.040116	.101491
10916225	.807851	1.0246	.03089	-.040375	.102155
11916021	.807347	1.0247	.030932	-.040495	.102359
12915726	.806516	1.02494	.030878	-.04044	.102195
1391565	.80628	1.02502	.030893	-.040438	.102224
14915649	.806281	1.02502	.030911	-.040469	.10229
15915604	.80615	1.02506	.030909	-.040478	.102295
1691557	.806044	1.0251	.030906	-.040471	.102283
17915567	.806029	1.0251	.030908	-.040472	.102287
18915564	.806023	1.02511	.03091	-.040477	.102297
19915557	.806002	1.02511	.030909	-.040477	.102295
20915554	.80599	1.02512	.030909	-.040476	.102294
21915554	.805989	1.02512	.03091	-.040476	.102295
22915553	.805987	1.02512	.03091	-.040477	.102296
23915552	.805984	1.02512	.030909	-.040477	.102296
24915552	.805983	1.02512	.03091	-.040477	.102296

step	(7) fevd	(7) Lower	(7) Upper	(8) fevd	(8) Lower	(8) Upper
0	0	0	0	0	0	0
1	0	0	0	.999841	.994744	1.00494
2	.035173	-.034287	.104633	.977628	.922794	1.03246
3	.035073	-.034128	.104274	.968627	.89666	1.04059
4	.068009	-.02087	.156889	.968999	.897611	1.04039
5	.079388	-.022594	.18137	.969662	.900185	1.03914
6	.080431	-.022684	.183547	.969785	.900417	1.03915
7	.080664	-.022817	.184144	.968996	.897505	1.04049
8	.082453	-.023736	.188642	.969112	.897782	1.04044
9	.083768	-.024583	.19212	.969313	.898509	1.04012
10	.083775	-.024598	.192149	.96911	.897845	1.04038
11	.083979	-.024696	.192653	.969068	.897641	1.0405
12	.084274	-.024936	.193484	.969122	.897805	1.04044
13	.08435	-.02502	.19372	.969107	.897776	1.04044
14	.084351	-.025018	.193719	.969089	.89771	1.04047
15	.084396	-.025057	.19385	.969091	.897705	1.04048
16	.08443	-.025097	.193956	.969094	.897717	1.04047
17	.084433	-.025104	.193971	.969092	.897713	1.04047
18	.084436	-.025105	.193977	.96909	.897703	1.04048
19	.084443	-.025113	.193998	.969091	.897705	1.04048
20	.084446	-.025119	.19401	.969091	.897706	1.04048
21	.084446	-.025119	.194011	.96909	.897705	1.04048
22	.084447	-.02512	.194013	.96909	.897704	1.04048
23	.084448	-.025121	.194016	.969091	.897704	1.04048
24	.084448	-.025122	.194017	.96909	.897704	1.04048

95% lower and upper bounds reported

- (1) irfname = var, impulse = SET, and response = SET
- (2) irfname = var, impulse = SET, and response = SDH
- (3) irfname = var, impulse = SDH, and response = SET
- (4) irfname = var, impulse = SDH, and response = SDH
- (5) irfname = var1, impulse = SET, and response = SET
- (6) irfname = var1, impulse = SET, and response = SDH
- (7) irfname = var1, impulse = SDH, and response = SET
- (8) irfname = var1, impulse = SDH, and response = SDH

. irf table fevd, impulse (SET TWH) response (SET TWH)

Results from var var1

step	(1) fevd	(1) Lower	(1) Upper	(2) fevd	(2) Lower	(2) Upper	(3) fevd	(3) Lower	(3) Upper
0	0	0	0	0	0	0	0	0	0
1	1	1	1	.0012	-.012644	.015045	0	0	0
2	.918649	.815714	1.02158	.001181	-.013216	.015578	.004875	-.020756	.030506
3	.910793	.800043	1.02154	.001586	-.010285	.013457	.009299	-.028597	.047194
4	.909594	.797269	1.02192	.001884	-.009844	.013611	.010452	-.03095	.051854
5	.909276	.796491	1.02206	.001979	-.009855	.013813	.010725	-.031462	.052911
6	.909208	.796318	1.0221	.002002	-.009869	.013874	.010776	-.031548	.053101
7	.909194	.796282	1.02211	.002007	-.009873	.013887	.010785	-.031559	.053128
8	.909192	.796275	1.02211	.002007	-.009874	.013889	.010786	-.031559	.053131
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step	(4) fevd	(4) Lower	(4) Upper	(5) fevd	(5) Lower	(5) Upper	(6) fevd	(6) Lower	(6) Upper
0	0	0	0	0	0	0	0	0	0
1	.919646	.81535	1.02394	1	1	1	.012125	-.031855	.056106
2	.838825	.697258	.980392	.985554	.943021	1.02809	.012347	-.03429	.058984
3	.826623	.673665	.979581	.983999	.937084	1.03091	.019909	-.044286	.084105
4	.823554	.667516	.979592	.928444	.83944	1.01745	.019687	-.041764	.081138
5	.822937	.666269	.979606	.918522	.817986	1.01906	.019647	-.041463	.080758
6	.822829	.66606	.979598	.91708	.814383	1.01978	.02015	-.042362	.082662
7	.822812	.666031	.979593	.915863	.812599	1.01913	.020217	-.042581	.083016
8	.82281	.666028	.979592	.914723	.810079	1.01937	.020202	-.042545	.082949
9913574	.807587	1.01956	.020224	-.042568	.083017
10913565	.807535	1.0196	.020252	-.042641	.083145
11913476	.807409	1.01954	.020271	-.042675	.083217
12913231	.806937	1.01952	.020267	-.042666	.0832
13913162	.806761	1.01956	.020266	-.042668	.0832
14913158	.806745	1.01957	.020271	-.042677	.083219
15913142	.806727	1.01956	.020272	-.042679	.083223
16913123	.806689	1.01956	.020272	-.042678	.083221
17913114	.806668	1.01956	.020272	-.042678	.083222
18913114	.806667	1.01956	.020272	-.042679	.083223
19913113	.806666	1.01956	.020272	-.042679	.083224
20913111	.806662	1.01956	.020272	-.042679	.083224
2191311	.80666	1.01956	.020272	-.042679	.083224
2291311	.80666	1.01956	.020272	-.042679	.083224
2391311	.80666	1.01956	.020272	-.042679	.083224
2491311	.806659	1.01956	.020272	-.042679	.083224

step	(7) fevd	(7) Lower	(7) Upper	(8) fevd	(8) Lower	(8) Upper
0	0	0	0	0	0	0
1	0	0	0	.987875	.943894	1.03186
2	.014446	-.028087	.056979	.987653	.941016	1.03429
3	.016001	-.030914	.062916	.980091	.915895	1.04429
4	.071556	-.017448	.16056	.980313	.918862	1.04176
5	.081478	-.019058	.182014	.980353	.919242	1.04146
6	.08292	-.019777	.185617	.97985	.917338	1.04236
7	.084137	-.019128	.187401	.979783	.916984	1.04258
8	.085277	-.019367	.189921	.979798	.917051	1.04254
9	.086426	-.019562	.192413	.979776	.916983	1.04257
10	.086435	-.019596	.192465	.979748	.916855	1.04264
11	.086524	-.019544	.192591	.979729	.916783	1.04268
12	.086769	-.019525	.193063	.979733	.9168	1.04267
13	.086838	-.019562	.193239	.979734	.9168	1.04267
14	.086842	-.019571	.193255	.979729	.916781	1.04268
15	.086858	-.019558	.193273	.979728	.916777	1.04268
16	.086877	-.019557	.193311	.979728	.916779	1.04268
17	.086886	-.019561	.193332	.979728	.916778	1.04268
18	.086886	-.019561	.193333	.979728	.916777	1.04268
19	.086887	-.01956	.193334	.979728	.916776	1.04268
20	.086889	-.01956	.193338	.979728	.916776	1.04268
21	.08689	-.01956	.19334	.979728	.916776	1.04268
22	.08689	-.01956	.19334	.979728	.916776	1.04268
23	.08689	-.01956	.19334	.979728	.916776	1.04268
24	.08689	-.01956	.193341	.979728	.916776	1.04268

95% lower and upper bounds reported

- (1) irfname = var, impulse = SET, and response = SET
- (2) irfname = var, impulse = SET, and response = TWH
- (3) irfname = var, impulse = TWH, and response = SET
- (4) irfname = var, impulse = TWH, and response = TWH
- (5) irfname = var1, impulse = SET, and response = SET
- (6) irfname = var1, impulse = SET, and response = TWH
- (7) irfname = var1, impulse = TWH, and response = SET
- (8) irfname = var1, impulse = TWH, and response = TWH

BIOGRAPHY

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