

The Effects of Gas Prices on the U.S. Personal Saving Rate

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I. Introduction

Automobiles are the main method of transportation for the majority of Americans. A major variable factor in the cost of driving is the price of gasoline. Gasoline prices can fluctuate vastly from year to year, state to state, or even station to station. There has always been speculation that when gas prices decrease or remain lower than usual, consumers take the extra savings at the pump and use it for personal savings. It is unclear just how much of that additional savings is actually making it to their personal savings and how much ends up being spent as consumption. The JP Morgan Chase Institute claims that "consumers spend more of their gas savings than initially estimated" (Farrell, 2017), but just how much is considered "more…than initially estimated"? Since the personal saving rate has a large impact on long run economic growth, this paper will look at what effects the average annual gas price has had on the U.S. personal saving rate. We begin by looking at the statistical significance of the historical average gas price per year, along with a selection of other theoretically important variables, to the U.S. personal saving rate using regression analysis. We then examine the effects of adding lags to the model, and decide whether to leave all of the variables, or to remove some in an attempt to improve the model. From this, we determine what effect gas prices have had on the U.S. personal saving rate.

II. Literature Review

Personal saving is not a new topic of study. Economists have been analyzing it in order to try to predict various factors and fluctuations in the long run economic growth of the economy. Gough (2011) specifically looked into the various determinants of the U.S. household saving rate due to its significant contribution to the U.S.'s aggregate saving. He references Carroll's (2001) theory of consumption behavior, where it is argued that the only reason households will change their consumption behavior is if "they expect a *permanent* change in their future income" (Gough, 2011, pg.28). Gough uses this as a basis for his choice of relevant variables in his regression model. He analyzes the effects of real personal disposable income, household net worth, real interest rate, and a proxy for household uncertainty on the household saving rate for their potential impact on future income. He also looks at the retired, youth, dependent youth, and college populations, since the time of life a person is in will also drastically affect what their future income will be – if an individual is in college, their future income expectations will be much different than an individual who is retired and living on a presumably fixed income. Due to Gough's study of the U.S. household saving rate, his results are likely to be affected by similar, if not the same, factors as our dependent variable: the U.S. personal saving rate.

Franz (2011) took a different approach to the one of the variables in this study, where she researched the effects of gas prices on single mothers' time use. She analyzed whether or not the price of gas had an effect on how much time a single mother would have to care for her child. Since a child's development is largely impacted by how much time he spends with his parents, a single mother's time would be that much more important since she is the only parent giving attention and care to the child. Franz suggested that a single mother would have to allocate her time between caring for the child and working to earn money. The amount that she would have to work was hypothesized to be contingent upon the gas price – the higher the price, the more time the mother would have to spend at work, thus less time is available to be spent caring for the child. Through her regression analysis, however, she found that gas prices were insignificant even at a .10 level. So, while the idea was good in theory, she could not find significant correlation between gas prices and how much time a mother has to spend on her child. This could have been, in part, due to a relatively small sample size. Even though her work did not result in statistical

significance of her gas price variable, her approach has given insight on how we will include our gas price variable in our model.

Another component to the U.S. personal saving rate that was examined was consumer sentiment. Ewing and Payne (1998) looked into this relationship and determined that "households reduce their savings rate when consumer sentiment is high, but the two variables do not drift arbitrarily far apart" (pg. 94). They then focused on what implications this had for financial planners and counselors. They saw that financial planners and counselors would have to take into account that their customers would be making investment and saving decisions based on how they perceived their future would occur. This results in them keeping most of their financial assets as the more-liquid money instead of investing it. This will effectively lower the personal savings rate. While our model does not explicitly include consumer sentiment as a variable of interest, it is important to still consider its implications on the variables we do include.

III. Model Specification and Variable Definitions

We will use an Ordinary Least Squares Regression to examine the statistical significance of relationships between the U.S. personal saving rate and the hypothetically impactful variables. Our initial hypothesis is that the most impactful variable will be disposable personal income. The data is a time-series set from 1960-2015. The following is our original model:

(1) $SAVRT_{t} = \beta_{0} + \beta_{1}GASPR_{t} + \beta_{2}INTRT_{t} + \beta_{3}TAXLOW_{t} + \beta_{4}TAXHIGH_{t} + \beta_{5}UNEMP_{t} + \beta_{6}RETPOP_{t} + \beta_{7}DISPINC_{t}$

SAVRT = U.S. Personal Saving Rate GASPR = Average U.S. Gas Price INTRT = U.S. Interest Rate TAXLOW = Personal Income Tax (lowest tax bracket) TAXHIGH = Personal Income Tax (highest tax bracket) UNEMP = U.S. Civilian Unemployment Rate RETPOP = Retired U.S. Population (% of U.S. Population 65 or older) DISPINC = Disposable Personal Income

IV. Data Description

For this study, we will compare the average annual gas price in the U.S., along with other relevant variables, to the U.S. personal saving rate. We will be referencing the years $1960-2015^1$ for all of our datasets and regressions, and it is all annualized data from the United States – no other countries are examined in this study. The personal saving sate, interest rate, income tax rate, unemployment rate, and disposable personal income data was retrieved from the St. Louis Fed's database. The average personal saving rate for the period we are looking at was 8.48%. The data for the average annual gas price was found on the U.S. Department of Energy's website, specifically the Vehicle Technology Office. It is measured in "constant 2015 dollars per gallon", meaning it is measured by the dollar amount it cost in 2015 after inflation is considered. With this said, the average annual gas price was \$2.16 per gallon. The retired population data is the percentage of U.S. citizens that are aged 65 or older and comes from the World Development Indicators database. **Table 2** shows descriptive statistics for all of the included variables.

¹ 1960-2015 is the range where data is available for all variables, producing the most comprehensive results

Table 1

Variable	Units	Hypothesized Effect on Personal Saving Rate N/A	
Personal Saving Rate	% of total income added to personal saving		
Average Gas Price	Constant 2015 dollars per gallon	Negative	
Interest Rate	% per annum	Positive	
Tax (lowest tax bracket)	Percent	Negative	
Tax (highest tax bracket)	Percent	Negative	
Unemployment Rate	Percent	Negative	
Retired Population	% of total	Negative	
Disposable Income	Billions of dollars	Positive	

Table 2

	PERS_SAVING		INTEREST_				RETIRED_	
	_RATE	GAS_PRICE	RATE	TAX_LOWEST	TAX_HIGHEST	UNEMP_RATE	POPULATION	DISPOSABLE_INCOME
Mean	8.480357	2.161964	4.906964	13.32143	52.53438	6.094643	11.69841	4924.138
Median	8.75	1.9	4.875	14	39.6	5.85	12.26259	3758.85
Maximum	14.4	3.8	12.87	20	91	10.8	14.64034	13801.5
Minimum	2.8	1.47	0.5	10	28	3.4	9.125636	378.4
Std. Dev.	3.188184	0.625429	2.929254	2.777226	19.30308	1.6047	1.377468	4187.042
Skewness	-0.107197	1.294658	0.574931	0.56791	0.509193	0.662564	-0.345003	0.653507
Kurtosis	1.833017	3.538461	3.394833	3.139819	1.872216	3.333814	2.328617	2.120607
Jarque-Bera	3.2849	16.32051	3.448841	3.055821	5.387685	4.357262	2.16268	5.790437
Probability	0.193505	0.000286	0.178276	0.216989	0.067621	0.113196	0.339141	0.055287
Sum	474.9	121.07	274.79	746	2941.925	341.3	655.1111	275751.7
Sum Sq. Dev	559.0484	21.51388	471.929	424.2143	20493.48	141.6284	104.358	9.64E+08
Observation	56	56	56	56	56	56	56	56

V. Model Estimation and Specification Testing

As mentioned before, we used an Ordinary Least Squares regression with a Newey-West adjustment to account for heteroscedasticity. From the initial model, the following regression results were achieved (**Table 3**). Only the disposable income variable was significant in this first model, with a p-value of 0.0000. Item **A** in the Appendix shows the variance inflation factors for the initial model, with no variables showing initial signs of multicollinearity. As **Table 4** shows, a ramsey reset test also concluded that there was not an issue with the original model with a p-value of .1826. A unit root test was run on each of the variables, and after necessary adjustments for seasonality and trends, are all showing stationarity in the regression's variables (see **B** through **I** in Appendix).

Table 3

Dependent Variable: PERS_SAVING_RATE_DIF Method: Least Squares Date: 12/14/17 Time: 17:59 Sample (adjusted): 1961 2015 Included observations: 55 after adjustments HAC standard errors & covariance (Bartlett kernel, Newey-West fixed bandwidth = 4.0000)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
GAS_PRICE_DIF INTEREST_RATE_DIF TAX_LOWEST TAX_HIGHEST UNEMP_RATE RETPOP_DT DISP_INCOME_DIF_DT C	0.599433 -0.272957 -0.011099 0.006537 0.002071 -0.210614 0.006055 -0.302381	0.503002 0.173648 0.061075 0.011262 0.119085 0.442923 0.001071 1.089047	1.191710 -1.571903 -0.181725 0.580430 0.017395 -0.475510 5.652283 -0.277656	0.2394 0.1227 0.8566 0.5644 0.9862 0.6366 0.0000 0.7825
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic) Prob(Wald F-statistic)	0.478414 0.400731 1.262275 74.88684 -86.52936 6.158543 0.000039 0.000000	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin Durbin-Watso Wald F-statis	lent var ent var iterion rion in criter. on stat tic	-0.083636 1.630583 3.437431 3.729407 3.550340 1.789740 12.34200

Table 4

 Ramsey RESET Test

 Equation: EQ01

 Specification: PERS_SAVING_RATE_DIF GAS_PRICE_DIF

 INTEREST_RATE_DIF TAX_LOWEST TAX_HIGHEST UNEMP_RATE

 RETPOP_DT DISP_INCOME_DIF_DT C

 Omitted Variables: Powers of fitted values from 2 to 7

 F-statistic
 Value
 df
 Probability

 F-statistic
 1.562715
 (6, 41)
 0.1826

Likelihood ratio	11.32717	(8, 41) 6	0.0788	
F-test summary:				
	Sum of Sq.	df	Mean Squares	
Test SSR	13.93832	6	2.323053	
Restricted SSR	74.88684	47	1.593337	
Unrestricted SSR	60.94852	41	1.486549	
LR test summary:				
	Value	df		
Restricted LogL	-86.52936	47		
Unrestricted LogL	-80.86577	41		

Due to only one of the independent variables being statistically significant, we ran a correlation/covariance analysis on the variable group (**Table 5**). In doing this, we found the DISPINC variable to have a .639548 correlation with the personal saving rate. This was the only relatively high correlation that was found and it will not affect the model.

Table 5

Included observations: 55 Balanced sample (listwise missing value deletion) Covariance Correlation PERS_SAVI... GAS_PRICE... INTEREST_... TAX_LOWEST_TAX_HIGHE... UNEMP_RA... RETPOP_DT_DISP_INCO... PERS_SAVING_R.. 2.610460 1.000000 GAS_PRICE_DIF 0.100997 0.093216 0.204741 1.000000 INTEREST_RATE_ -0.2295870.093016 2.129467 -0.097376 0.208775 1.000000 TAX_LOWEST -0.041455 0.036000 0.678727 6.887273 -0.009777 0.044930 0.177230 1.000000 TAX_HIGHEST 0.267836 -0.057364 5.110795 27.89164 345.2177 1.000000 0.008922 -0.010112 0.188498 0.572010 -4.479764 UNEMP_RATE 0.005147 -0.053031 -0.895638 -1.164364 2.570334 0.001987 1.000000 -0.108341-0.382827 -0.276739 -0.150388 RETPOP_DT -0.100806 -0.041701 -0.027106 0.144961 -1.611136 0.211494 0.245315 1.000000 -0.125970 -0.275764 -0.037504 0.111523 -0.175075 0.266343 DISP_INCOME_DI... 182.0213 10.33967 43.95746 8.140504 -95.74934 -25.09415 -3.276253 31029.93 0.639548 0.192252 0.171004 0.017609 -0.029255 -0.088856 -0.037551 1.000000

Covariance Analysis: Ordinary Date: 12/14/17 Time: 18:56 Sample: 1961 2015 Included observations: 55 Balanced sample (listwise missing value deletion) We then introduced lags to the variables INTRT, RETPOP, and UNEMP RATE in an attempt to improve the significance in the model. The resulting model brought us closer, but had insignificant lags that had to be removed (**Table 6**). After removing various combinations of lags to find the best fit, we also removed the variable TAX HIGHEST, as it was insignificant and it improved our model by being left out. Seeing the improvement, we also removed TAX_LOWEST since it was insignificant as well. This landed us with our final version of the model, Table 7, and improved the statistical significance of all variables. The new model is a better fit for the data, as shown by its AIC and SIC values of 3.1633 and 3.4948 respectively. This is in comparison to Table 6's 3.2855 AIC and 3.8859 SIC, and the original model's 3.4374 AIC and 3.7294 SIC. Since both the final model's AIC and SIC values are lower, it is the better model for our data. Its .6231 R-squared and .5561 adjusted R-squared are also higher than the original models' values .4784 R-squared and .4007 adjusted R-squared. Even though these are slightly lower than **Table 6**'s values, the final model is still the best since all of its lags are significant. Also, by having more variables such as the two tax rates and the insignificant lags, Table 6's regression would show higher R-squared and adjusted R-squared values simply because it has more variables. A residual analysis for each of the explanatory variables showed that there were no abnormalities with the final model. Each residual plot can be found in J through N in the Appendix.

Table 6

Dependent Variable: PERS_SAVING_RATE_DIF Method: Least Squares Date: 12/14/17 Time: 19:02 Sample (adjusted): 1964 2015 Included observations: 52 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
GAS_PRICE_DIF INTEREST_RATE_DIF INTEREST_RATE_DIF(-1) INTEREST_RATE_DIF(-2) INTEREST_RATE_DIF(-3) TAX_LOWEST TAX_HIGHEST UNEMP_RATE UNEMP_RATE(-1) UNEMP_RATE(-1) UNEMP_RATE(-2) UNEMP_RATE(-3) RETPOP_DT RETPOP_DT RETPOP_DT(-1) RETPOP_DT(-2) DISP_INCOME_DIF_DT C	2.117227 -0.272015 -0.466044 -0.234651 0.135008 0.117992 -0.002699 0.377109 -0.897425 0.248313 0.091585 -7.223759 20.65049 -14.04204 0.006485 -0.489911	0.712760 0.150122 0.171110 0.149760 0.131541 0.134267 0.013139 0.202880 0.312760 0.310231 0.240353 13.40415 26.93392 14.13490 0.000890 2.365212	2.970461 -1.811955 -2.723643 -1.566849 1.026350 0.878780 -0.205457 1.858778 -2.869375 0.800413 0.381043 -0.538919 0.766709 -0.993431 7.288592 -0.207132	0.0053 0.0783 0.0099 0.1259 0.3116 0.3853 0.8384 0.0712 0.0068 0.4287 0.7054 0.5933 0.4483 0.3271 0.0000 0.8371
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic)	0.686965 0.556533 1.105139 43.96799 -69.42247 5.266865 0.000022	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin Durbin-Watso	lent var ent var rion in criter. on stat	-0.092308 1.659535 3.285480 3.885862 3.515652 2.060566

Table 7

Dependent Variable: PERS_SAVING_RATE_DIF Method: Least Squares Date: 12/14/17 Time: 17:39 Sample (adjusted): 1962 2015 Included observations: 54 after adjustments HAC standard errors & covariance (Bartlett kernel, Newey-West fixed bandwidth = 4.0000)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
GAS_PRICE_DIF INTEREST_RATE_DIF INTEREST_RATE_DIF(-1) UNEMP_RATE UNEMP_RATE(-1) RETPOP_DT RETPOP_DT(-1) DISP_INCOME_DIF_DT C	1.902655 -0.145869 -0.459705 0.283456 -0.633871 5.552806 -5.262687 0.006412 1.877879	0.657671 0.109241 0.189784 0.149377 0.164198 1.544681 1.462769 0.000628 0.692225	2.893018 -1.335298 -2.422252 1.897583 -3.860417 3.594790 -3.597756 10.20816 2.712816	0.0059 0.1885 0.0195 0.0642 0.0004 0.0008 0.0008 0.0008 0.0009
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic) Prob(Wald F-statistic)	0.623093 0.556088 1.091143 53.57668 -76.41019 9.299112 0.000000 0.000000	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin Durbin-Watso Wald F-statis	lent var ent var iterion rion in criter. on stat tic	-0.105556 1.637695 3.163340 3.494838 3.291186 2.071988 32.92678

VI. Results

(2) $SAVRT_{t} = \beta_{0} + \beta_{1}GASPR_{t} + \beta_{2}INTRT_{t} + \beta_{3}INTRT_{t-1} + \beta_{4}UNEMP_{t} + \beta_{5}UNEMP_{t-1} + \beta_{6}RETPOP_{t} + \beta_{7}RETPOP_{t-1} + \beta_{8}DISPINC_{t}$

The various alterations to the original model resulted in the model shown above – the final model. In the final model, the only variable that is not statistically significant is the interest rate, with a p-value of .1885. The unemployment rate is significant at a 10% level with a p-value of .0642, the interest rate lag (p-value of .0195) is significant at a 5% level, and all of the remaining variables are significant at a 1% confidence level. This makes this final version a much more substantial model. From the final model's regression results (Table 7) we can see that for every one dollar change in the average price of gasoline in the U.S., the personal saving rate increases by 1.90%. This goes against the hypothesized effect of a rise in gas prices reducing the rate of personal saving. This is most likely due to the fact that the personal saving rate has been increasing in the U.S. over time at the same time that gas prices have also been increasing. Even though the interest rate variable was not statistically significant, the results from the lag on the interest rate show that a one percent increase in the U.S. interest rate today will decrease the personal saving rate next period by approximately .46% – which is also against the hypothesized positive effect. The unemployment rate's short run multiplier shows that with every one percent increase, the unemployment rate increases the saving rate by .28%. However, the long run multiplier shows the hypothesized negative effect, where every one percent increase in the interest rate today results in a .35% decrease in the saving rate next period. The retired population's short run multiplier was 5.55, meaning that for every 1% increase in the retired population, it increases the personal saving rate by 5.55%. This goes against the hypothesized effect as well, and while the long run multiplier of .29 still shows that the

retired population has a positive effect on the saving rate, it is much closer to what we estimated it would be. Finally, every one billion dollar increase in disposable income for the U.S. is shown to increase the saving rate by only .006%. As mentioned before, the final model has an R-squared value of .6231 and an adjusted R-squared value of .5561.

VII. Conclusion

In the end, we found that the average price of gasoline in the U.S. has a positive correlation with the personal saving rate. However, since the average gas price and the personal saving rate have been rising slightly throughout the time range that has been observed due to a growing economy, that could have caused this positive correlation. It could also be the case that rising gas prices are a sign of improving economic times, which would, in fact, yield a higher average personal saving rate. A time frame that is more recent and includes newer data might correct for this and cause gas prices to have the hypothesized negative effect on the personal saving rate. We also believe that if we could get the tax rates to be significant, it would further benefit the model. If we made the adjustment to only include more recent years in the study, this could potentially bring the tax rates into significance since it would not have to account for the large fluctuations in the personal income tax rate experienced from the 1930's to the early 1980's. A possible extension of this research could be to add more relevant variables such as consumer sentiment and consumption to see how they affect the model and if they begin to change the effects the other variables seem to have on the personal saving rate.

Nonetheless, the results show that the gas price, interest rate, unemployment rate, retired population, and disposable income all have significant effects on the personal saving rate. Even though the variables do not all have the effect that we originally hypothesized they would, their significance means that they are still possible triggers that can alter the average personal saving rate in the United States. For instance, if we observe that the interest rate is beginning to increase, we can expect to experience an overall decrease in the personal saving rate in the coming periods and better prepare for it. According to the regression results, we could also look at the average gas price – which is already monitored – and know that during times when it is rising we will also see a rise in the personal saving rate. Knowing what triggers to look for can help in identifying when we will be entering periods of higher or lower saving.

VIII. Appendix

Α.

Variance Inflation Factors Date: 12/14/17 Time: 18:01 Sample: 1960 2015 Included observations: 55

Variable	Coefficient	Uncentered	Centered
	Variance	VIF	VIF
GAS_PRICE_DIF	0.253011	1.525362	1.517231
INTEREST_RATE_DIF	0.030154	2.004298	1.936757
TAX_LOWEST	0.003730	38.66147	1.781641
TAX_HIGHEST	0.000127	22.79127	2.419955
UNEMP_RATE	0.014181	26.46839	2.007355
RETPOP_DT	0.196180	2.257748	2.208745
DISP_INCOME_DIF_DT	1.15E-06	1.808700	1.806681
C	1.186024	61.11397	NA

В.

Exogenous, None Lag Length: 0 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fu	ller test statistic	-5.857184	0.0000
Test critical values:	1% level	-2.608490	
	5% level	-1.946996	
	10% level	-1.612934	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(GAS_PRICE_DIF) Method: Least Squares Date: 12/14/17 Time: 18:09 Sample (adjusted): 1962 2015 Included observations: 54 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
GAS_PRICE_DIF(-1)	-0.880915	0.150399	-5.857184	0.0000
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.391827 0.391827 0.309285 5.069829 -12.74940 1.793157	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin	lent var ent var iterion rion n criter.	-0.016852 0.396593 0.509237 0.546070 0.523442

Null Hypothesis: DISP_INCOME_DIF_DT has a unit root Exogenous: None Lag Length: 10 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic		-4.196605	0.0001
Test critical values:	1% level	-2.618579	
	5% level	-1.948495	
	10% level	-1.612135	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(DISP_INCOME_DIF_DT) Method: Least Squares Date: 11/29/17 Time: 19:46 Sample (adjusted): 1972 2015 Included observations: 44 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
DISP_INCOME_DIF_DT(-1) D(DISP_INCOME_DIF_DT(-1)) D(DISP_INCOME_DIF_DT(-2)) D(DISP_INCOME_DIF_DT(-3)) D(DISP_INCOME_DIF_DT(-4)) D(DISP_INCOME_DIF_DT(-5)) D(DISP_INCOME_DIF_DT(-6)) D(DISP_INCOME_DIF_DT(-6)) D(DISP_INCOME_DIF_DT(-8)) D(DISP_INCOME_DIF_DT(-9)) D(DISP_INCOME_DIF_DT(-10))	-3.514513 2.137274 1.960012 2.045731 1.734506 1.706781 1.509634 1.257275 2.448302 1.549308 0.873131	0.837466 0.770340 0.708800 0.682975 0.652205 0.573075 0.510703 0.478836 0.432845 0.483878 0.383002	-4.196605 2.774457 2.765254 2.969240 2.659451 2.978282 2.955995 2.625688 5.656303 3.201857 2.279706	0.0002 0.0090 0.0055 0.0120 0.0054 0.0137 0.0130 0.0000 0.0000 0.0030 0.0292
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.926245 0.903895 105.6723 368499.0 -261.1594 1.927164	Mean depend S.D. depende Akaike info cri Schwarz criter Hannan-Quint	lent var Int var Iterion ion n criter.	-0.362121 340.8701 12.37088 12.81693 12.53630

D.

Exogenous: None

Lag Length: 1 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fulle	r test statistic	-6.327594	0.0000
Test critical values:	1% level	-2.609324	
	5% level	-1.947119	
	10% level	-1.612867	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(INTEREST_RATE_DIF) Method: Least Squares Date: 11/29/17 Time: 19:51 Sample (adjusted): 1963 2015 Included observations: 53 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
INTEREST_RATE_DIF(-1) D(INTEREST_RATE_DIF(-1))	-0.990900 0.369330	0.156600 0.130171	-6.327594 2.837257	0.0000 0.0065
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.448701 0.437891 1.353938 93.49059 -90.24432 2.138178	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin	lent var ent var iterion rion n criter.	0.004717 1.805879 3.480918 3.555268 3.509509

С.

E. Null Hypothesis: RETPOP_DT has a unit root Exogenous: None Lag Length: 2 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
<u>Augmented Dickey-Fu</u> Test critical values:	ller test statistic 1% level 5% level 10% level	-3.890161 -2.609324 -1.947119 -1.612867	0.0002
	10,010101	1.012001	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(RETPOP_DT) Method: Least Squares Date: 11/30/17 Time: 09:07 Sample (adjusted): 1963 2015 Included observations: 53 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
RETPOP_DT(-1) D(RETPOP_DT(-1)) D(RETPOP_DT(-2))	-0.023115 1.346075 -0.308117	0.005942 0.128478 0.144039	-3.890161 10.47709 -2.139116	0.0003 0.0000 0.0373
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.980345 0.979559 0.015096 0.011394 148.5872 2.127592	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quin	lent var ent var iterion rion ın criter.	0.022714 0.105585 -5.493857 -5.382331 -5.450970

F.

Null Hypothesis: TAX_LOWEST has a unit root Exogenous: None Lag Length: 0 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ller test statistic 1% level 5% level 10% level	-1.675990 -2.607686 -1.946878 1.612000	0.0883

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(TAX_LOWEST) Method: Least Squares Date: 11/29/17 Time: 20:03 Sample (adjusted): 1961 2015 Included observations: 55 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
TAX_LOWEST(-1)	-0.017833	0.010640	-1.675990	0.0995
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.022518 0.022518 1.077863 62.73660 -81.66094 1.753740	Mean depenc S.D. depende Akaike info cr Schwarz crite Hannan-Quin	lent var ent var iterion rion n criter.	-0.181818 1.090207 3.005853 3.042350 3.019966

G. Null Hypothesis: TAX_HIGHEST has a unit root Exogenous: None Lag Length: 0 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	iller test statistic 1% level 5% level 10% level	-2.221102 -2.607686 -1.946878 -1.612999	0.0266

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(TAX_HIGHEST) Method: Least Squares Date: 11/29/17 Time: 20:02 Sample (adjusted): 1961 2015 Included observations: 55 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
TAX_HIGHEST(-1)	-0.022325	0.010051	-2.221102	0.0306
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.039017 0.039017 4.186399 946.4006 -156.2883 1.624780	Mean depenc S.D. depende Akaike info cr Schwarz crite Hannan-Quin	lent var ent var iterion rion n criter.	-0.934545 4.270540 5.719574 5.756070 5.733687

H.

Null Hypothesis: UNEMP_RATE has a unit root Exogenous: Constant, Linear Trend Lag Length: 1 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fu Test critical values:	ler test statistic 1% level 5% level 10% level	- <u>3.775610</u> -4.137279 -3.495295 -3.176618	0.0256

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(UNEMP_RATE) Method: Least Squares Date: 11/29/17 Time: 20:05 Sample (adjusted): 1962 2015 Included observations: 54 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
UNEMP_RATE(-1) D(UNEMP_RATE(-1)) C @TREND("1960")	-0.319316 0.463762 1.797400 0.004992	0.084573 0.126083 0.529660 0.008262	-3.775610 3.678220 3.393498 0.604218	0.0004 0.0006 0.0014 0.5484
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic)	0.294758 0.252444 0.919031 42.23091 -69.98523 6.965898 0.000525	Mean depend S.D. depende Akaike info cri Schwarz crite Hannan-Quin Durbin-Watso	ent var nt var terion rion n criter. n stat	-0.018519 1.062939 2.740194 2.887526 2.797014 1.898483

I.

Null Hypothesis: PERS_SAVING_RATE_DIF has a unit root Exogenous: None Lag Length: 0 (Automatic - based on SIC, maxlag=10)

		t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic		-9.784021	0.0000
Test critical values:	1% level	-2.608490	
	5% level	-1.946996	
	10% level	-1.612934	

*MacKinnon (1996) one-sided p-values.

Augmented Dickey-Fuller Test Equation Dependent Variable: D(PERS_SAVING_RATE_DIF) Method: Least Squares Date: 11/29/17 Time: 20:15 Sample (adjusted): 1962 2015 Included observations: 54 after adjustments

Variable	Coefficient	Std. Error	t-Statistic	Prob.
PERS_SAVING_RATE_DIF(-1)	-1.283381	0.131171	-9.784021	0.0000
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood Durbin-Watson stat	0.643607 0.643607 1.573343 131.1966 -100.5909 2.072482	Mean depend S.D. depende Akaike info cr Schwarz crite Hannan-Quir	lent var ent var iterion rion in criter.	-0.025926 2.635475 3.762627 3.799460 3.776832





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