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The drivers of household indebtedness re-considered: an empirical evaluation of competing arguments on the macroeconomic determinants of household indebtedness in OECD countries

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Abstract: Household debt is at a record high in most OECD countries and it played a crucial role in the recent financial crisis. Several arguments on the macroeconomic drivers of household debt have been put forward, and most have been empirically tested, albeit in isolation of each other. This paper empirically tests seven competing hypotheses on the macroeconomic determinants of household indebtedness together in one econometric study. Existing arguments suggest that residential house prices, upward movements in the prices of assets demanded by households, the income share of the top 1%, falling wages, the rolling back of the welfare state, the age structure of the population and the short-term interest rate drive household indebtedness. We formulate these arguments as hypotheses and test them for a panel of 13 OECD countries over the period 1993 - 2011 using error correction models. We also investigate whether effects differ in boom and bust phases of the debt and house price cycles. The results show that the most robust macroeconomic determinant of household debt is real residential house prices, and that the phase of the debt and house price cycles plays a role in household debt accumulation.

Keywords: household debt, house prices, cycles

JEL classifications: E19, E21, R20

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1. Introduction

Household debt has risen at unprecedentedly high rates and reached record levels since the 1980's in most OECD countries, and is said to have played a crucial role in the 2007-2009 financial crisis (Mian and Sufi, 2014). Although household debt has not played a prominent role in macroeconomic theory until recently, the recognition of its importance to financial and macroeconomic stability has stimulated empirical research on the macroeconomic consequences of household debt. Such research looks at household debt as a cause of financial crises (e.g. Schularick and Taylor, 2012; Bordo and Meissner, 2012), debt-led consumption (e.g. Stockhammer, 2012, 2015; Hein, 2012; Kim et al 2012; Stockhammer and Wildauer, 2016) and debt-deflation (e.g. Koo, 2011; Mian and Sufi, 2014). In comparison, the macroeconomic drivers of household debt have been empirically studied relatively little. Several arguments co-exist in, and can be developed from, the literature. Ryoo's (2016) and Godley and Lavoie's (2007) post-Keynesian models can be used to show how house prices can fuel household borrowing. In the wealth effects and consumption literature, a discussion by Cooper and Dynan (2014) can be developed to illustrate that household debt is related to leverage used to purchase financial assets. Frank et al. (2014) argue that in the face of upward looking consumption norms, income inequality drives households who experience real income losses to accumulate debt to maintain relative consumption with their richer peers. Barba and Pivetti (2009) and Stockhammer (2012, 2015) indicate that falling wages drive households to take on debt to meet necessitous consumption. Lapavitsas (2013) argues that rising household indebtedness is due to the rolling back of the welfare state. The life cycle model (Modigliani and Brumberg, 1954) suggests that the age structure of the population propels household indebtedness. Taylor (2009) has argued that low federal funds interest rates are the prime reason for the increase in household borrowing associated with the housing boom. Finally, Justiniano et al. (2015) discuss that progressive relaxation of lending constraints led to a significant expansion in the supply of mortgages, the largest component of household debt.

Most of these arguments have been tested empirically, albeit in isolation of each other. For example, by controlling for the effect of house prices in estimating a panel household debt equation, Rubaszek and Serwa (2014) test the argument that house prices drive household debt accumulation. However, Rubaszek and Serwa's panel study doesn't test the other above-mentioned arguments on the drivers of household indebtedness. Malinen (2014) includes the top 1% income share and the short-term interest rate, which capture the arguments by Frank et al. (2014) and Taylor (2009) respectively, but not other potential drivers of household indebtedness. Similarly, Klein (2015) seeks

to test the effects of various measures of income inequality on household indebtedness. He tests the effects of the top 1% income share, which corresponds to a testing of Frank et al.'s argument, as well as the effects of the inverted Pareto-Lorenz coefficient, the wage share and the Gini index on household debt accumulation, but he does not consider any of the other existing arguments. More recently, Stockhammer and Wildauer (2017) test the arguments that the top 1% income share, house prices, financial deregulation, and the age structure a country's population drive household debt.

The lack of a comprehensive empirical study on the drivers of household debt is a concern because the exclusion of relevant variables results in a partial understanding of the macroeconomic drivers of household indebtedness, and econometrically can lead to omitted variable bias. Empirically assessing the causes of household indebtedness is crucial for designing policies which maintain financial and macroeconomic stability via the household sector by keeping households' balance sheets in check. For instance, if house prices are found as the core driver of household debt, restricting mortgage equity withdrawals, introducing caps on house price levels and increasing the supply of public housing would be relevant policies. On the other hand, if, for example, falling wages are found as the main driver of household debt accumulation, policies which address the lack of wage growth would have to be prioritised to avoid unsustainable debt accumulation.

In this context, the contribution of this paper lies in considering seven arguments on the drivers of household indebtedness together in one empirical study, thus developing a more complete understanding of the macroeconomic drivers of household debt accumulation. We formulate testable hypotheses from the arguments that house prices, upward movements in the prices of assets demanded by households, the income share of the top 1%, falling wages, the reduction of welfare state spending, the age structure of the population and the short-term interest rate drive household indebtedness. The hypotheses stemming from these arguments are referred to as the house price hypothesis, the financial asset hypothesis, the expenditure cascades hypothesis, the falling wages hypothesis, the welfare retrenchment hypothesis, the age structure hypothesis and the low interest rate hypothesis respectively. We test these hypotheses using a panel error correction model for 13 OECD countries over 1993 – 2011. We also investigate whether the determinants of household debt work symmetrically in the build-up of debt and during periods of deleveraging, by isolating phases of booms and busts of the debt and house price cycles.

The paper is structured as follows: in section 2 we explain the testable hypotheses on the drivers of household debt. Section 3 reviews the closely related empirical literature. Section 4 presents the econometric results. In section 5 we conclude.

2. Hypotheses on the determinants of household debt

The hypotheses on the drivers of household debt can be grouped into asset-transaction explanations, consumption-oriented explanations and explanations related to monetary policy and the credit supply.

Two of our asset-transaction explanations are derived from post-Keynesian models: Ryoo (2016) uses an application of Minsky's Financial Instability Hypothesis (FIH) to households, and Godley and Lavoie (2007) use a stock flow consistent (SFC) model. The third asset-transaction explanation is developed from Cooper and Dynan's (2014) discussion related to household leveraging in the wealth effects literature. Ryoo (2016, pp.976 - 978) builds a model of workers' debt accumulation behaviour where workers' debt is determined by their income and their net worth.¹ As houses are households' only asset in the model, house prices determine households' net worth. Starting from the assumption that households are credit-constrained, Ryoo's model shows how house prices can fuel household debt via a collateral effect, as houses serve as collateral for borrowing. A rise in house prices relaxes households' credit constraints and allows borrowing. In a SFC model, which tracks all flows and stocks in a macroeconomic model and enforces consistency between the two, Godley and Lavoie (2007) assume that households consume out of income and wealth and show that this corresponds to households attempting to reach a target wealth-to-income ratio. If we add an auxiliary assumption that households hold only housing wealth, we can expound that if households' wealth surpasses the target ratio, and if households do not sell their assets, households will consume this excess wealth by borrowing against their residential property in the form of mortgage equity withdrawals. It is through using mortgage equity withdrawals that households take on debt in the asset-transaction explanation constructed from Godley and Lavoie's SFC model.

In the asset-transaction explanations derived from the post-Keynesian models described above households react to trends in house prices. The difference between the explanations of household debt accumulation stemming from Ryoo's and Godley and Lavoie's works is that in the former house prices fuel household debt accumulation because they relax credit constraints via a collateral effect, while in the latter the mechanism is that house prices drive household indebtedness via a wealth effect. In the explanation developed from Godley and Lavoie's SFC model households are not credit constrained but increase their consumption due to increased wealth, which can be realised

¹ In Ryoo's (2016) baseline model, the household sector is divided into worker and capitalist households. Workers have housing wealth, while capitalist households hold stocks. As Capitalist households do not hold housing wealth, and as we are interested in highlighting channels through which household debt arises from house prices, we focus on Ryoo's model of worker households.

by borrowing using mortgage equity withdrawals. Despite the different mechanisms, at the macroeconomic level, the two explanations give us a testable hypothesis on the cause of household indebtedness: house prices drive household debt accumulation.

Another asset-transaction explanation of household debt accumulation can be constructed from the discussion on the relationship between households' assets and liabilities found in Cooper and Dynan's (2014) review of the so-called consumption wealth effects literature. In highlighting the observation that "a household's net worth is also a function of the debt that it holds" (Cooper and Dynan, 2014, p. 47), the authors discuss that household leverage and movements in the prices of assets held by households are related. From this, we develop what we call the financial asset hypothesis, which is based on the rationale that upward movements in the prices of assets that households demand drive households to take on debt as leverage to purchase such assets. We centre our financial asset hypothesis on non-housing wealth. In the wealth effects empirical literature, stock price indices are used a proxy for household non-housing financial wealth. Thus, our financial asset hypothesis states that upward movements in stock prices encourage household indebtedness.

Consumption-oriented explanations of household debt consist of four arguments. The first argument is Frank et al.'s (2014) expenditure cascades hypothesis. Building on the work of Veblen (1899) and Duesenberry (1949), Frank et al. argue that in the face of upward-looking consumption norms, income inequality drives households who have become relatively poorer, due to real income losses, to accumulate debt to maintain relative consumption with their richer peers. Rapidly growing top incomes embedded in a consumer environment in which consumers make spending decisions aspiring to the lifestyle and consumption behaviour of richer peers, leads to rising household debt as relatively poorer households who lose out due to the rise in top incomes use debt as an income substitute to "keep up with the Joneses". This argument is founded on a behavioural economics approach that stresses the role of socio-psychological motives in economic decision making.

The second consumption-oriented argument is what we call the falling wages hypothesis. Put forward by Barba and Pivetti (2009) and Stockhammer (2012, 2015), the falling wages hypothesis states that households who experience reduced wage incomes take on debt to maintain path-dependent, backward-looking consumption norms. The falling wages hypothesis is consistent with the standard post-Keynesian consumption models, which assume autonomous consumption and given marginal propensities of consumption which differ according to one's position within the income distribution. These assumptions imply that if there is an income shock, those that are affected negatively by the income shock will have to accumulate debt to maintain autonomous consumption.

Importantly, at the macroeconomic level both the expenditure cascades hypothesis and the falling wages hypothesis state that income inequality drives household indebtedness, and that the increase in household debt is based on the growth of consumption expenditures exceeding that of income. These two arguments are distinguished from each other with regards to the type of consumption norms at question, the category of households affected by income inequality and their predictions regarding consumption expenditures. Regarding consumption norms, the falling wages hypothesis holds that income inequality triggers households to accumulate debt in a struggle to maintain backward-looking, self-regarding consumption norms, while the expenditure cascades hypothesis holds that households borrow to fulfil upward looking, emulative consumption aspirations. Regarding the category of households affected by income inequality, the falling wages hypothesis suggests that income inequality negatively affects wage-dependent households who use their incomes for necessitous consumption; the analytical focus is on average wages. On the other hand, the expenditure cascades hypothesis focuses on households at the top-end of the distribution, whose incomes rose relative to those below them in the distribution; the analytical focus is the concentration of income of households at the top end of the distribution.² Finally, regarding consumption expenditures, the expenditure cascades hypothesis suggests rising consumption to GDP ratios, while and the falling wages hypothesis implies stable consumption to GDP ratios.

The third consumption-oriented explanation of household indebtedness is what we refer to as the welfare retrenchment hypothesis. Lapavistas (2013, p.240) argues that “rising household indebtedness has been associated with changes in the social provision of basic services including housing, health, education, transport and so on. To the degree to which social provision has retreated, or failed to expand, private provision has taken its place, mediated by finance.” This hypothesis is straightforward: stagnant or reduced welfare spending in key areas such as housing, health, education and transport cause households to borrow as these areas of welfare are a part of households’ basic spending. When not provided by the state, household spending on these areas fall into households’ basic consumption basket and thus are a part of households’ struggle to maintain consumption norms.

The final consumption-oriented explanation of household debt accumulation derives from the life cycle model (LCM) (Modigliani and Brumberg, 1954), which seeks to explain the consumption behaviour of households over their life time. It relies on perfectly informed, forward-looking households that are capable of detailed calculations of income and consumption, based on past,

² Top incomes have experienced a spectacular growth (Piketty and Saez, 2003; Atkinson and Piketty, 2010; OECD, 2008). For details, see the World Wealth and Income Database <http://wid.world/>

current and perfectly predicted future information on their consumption and real income levels. These households are experts in balancing their finances over their entire life, as they know all the future income streams that they will earn for their entire lives, and they enjoy access to perfectly functioning credit and capital markets. As a result, they maximise life-time utility through rational, life-time consumption smoothing. In a standard LCM borrowing can arise for two main reasons. The first is due to the age structure of the population. Given that young people do not earn the income required for their optimal life-time consumption, and assuming access to credit, they will borrow for consumption smoothing. Consequently, if a major part of the population is young, and if they are not credit-constrained, households would accumulate debt. The second reason for borrowing is due to transitory (non-permanent) income shocks. If households' transitory income is unexpectedly reduced, and if households do not face credit-constraints, they will borrow to maintain optimal consumption. Both the first and second drivers of household debt in the LCM critically depend on whether households have existing savings; households will only borrow if they have no savings or if they have depleted existing savings. Accordingly, household debt accumulation in the LCM depends on the age structure of the population, transitory income shocks and the stock of savings. Transitory income is not measurable, and it is an empirical fact that households borrow even though they have savings. Indeed, high income households with relatively high savings borrow more than their low-income counterparts. Therefore, the LCM gives rise to one testable hypothesis on the determinants of household indebtedness: the age structure of the population drives household debt accumulation.

Finally, the explanations of household debt accumulation related to monetary policy and the credit supply consist of what we call the low interest rate hypothesis and the credit supply hypothesis respectively. The low interest rate hypothesis stems from Taylor (2009), who has argued that low federal funds interest rates are the prime reason for the increase in household borrowing associated with the housing boom. According to Taylor, monetary policy, reflected by the federal funds interest rate, was too "loose" during the housing boom. Taylor argues that in the environment of rising house prices, the low federal funds interest rate (which he refers to as "monetary excess") encouraged risk-taking in the demand and supply of mortgages (which he refers to as "risk-taking excesses") (Taylor 2009, pp.11 – 12). Low federal funds interest rates increase household debt because the low short-term interbank interest rate means cheap borrowing for banks, which is passed onto households in the form of cheap borrowing. This argument essentially holds the central bank responsible for the rise in household debt. The low interest rate hypothesis thus states that a low short-term interest rate causes household indebtedness.

The credit supply hypothesis states that households take on debt because banks increase their willingness to lend, and thus supply more loans to households. The mechanism at work here is that the credit constraints that households hitherto faced are removed, allowing them to borrow more than previously permitted. Justiniano et al. (2015, p. 5) discuss that the increase in credit supply is fed by securitisation, market-based financial intermediation, and changes in financial regulation.

From the discussion above, we identify eight testable hypotheses on the macroeconomic drivers of household debt. We test seven hypotheses econometrically, and in doing so, we account for whether the determinants of household debt work symmetrically in the build-up of debt, i.e., during the boom phases of the debt and housing cycles, and during phases of deleveraging, i.e., during bust phases of debt and house price cycles. The eight hypotheses are summarised and empirically instrumentalised in Table 1. We are unable to test the credit supply hypothesis due to lack of data. This is a shortcoming of our econometric study, which future work should address. Data sources and definitions are discussed further in section 4.

Table 1: Testable hypotheses on the macroeconomic determinants of household debt

	Hypothesis	Theoretical argument	Corresponding variable and predicted sign
1	House price hypothesis	Household debt is driven by house prices, as an increase in house prices increases collateral, which relaxes credit constraints, and an increase in house prices increases household wealth, which prompts consumption which is realised by borrowing against the value of the residential property.	House prices (<i>HP</i>) $\frac{\partial HHD}{\partial HP} > 0$
2	Financial asset hypothesis	Upward movements in stock prices drive households to take on debt as leverage to purchase stocks.	Stock prices (<i>SP</i>) $\frac{\partial HHD}{\partial SP} > 0$
3	Expenditure cascades hypothesis	An increase in the income of households at the top end of the distribution drives household debt, because households at the lower end of the distribution take on debt to emulate the consumption of richer households.	Top 1% share of income (<i>TOP1</i>) $\frac{\partial HHD}{\partial TOP1} > 0$
4	Falling wages hypothesis	Households use debt as a substitute for reduced wage income to maintain path-dependent, backward looking consumption norms.	Average wages (<i>WAGES</i>) $\frac{\partial HHD}{\partial WAGES} < 0$

5	Welfare retrenchment hypothesis	Reduced welfare spending causes households to take on debt for spending on their basic welfare needs.	State welfare spending (<i>WELFARE</i>) $\frac{\partial HHD}{\partial WELFARE} < 0$
6	Age structure hypothesis	The age structure of the population determines household debt because the young accumulate debt while the elderly dissave.	Fraction of population aged 65 and older (<i>AGE</i>) $\frac{\partial HHD}{\partial AGE} < 0$
7	Low interest rate hypothesis	A low short-term interest rate drives household indebtedness because borrowing becomes cheaper.	Short-term interest rate (<i>i</i>) $\frac{\partial HHD}{\partial i} < 0$
8	Credit supply hypothesis	Banks supply more loans to households, allowing households to take on more debt than previously permitted.	Credit supply (<i>CRED</i>) $\frac{\partial HHD}{\partial CRED} > 0$

Table 1 summarises the hypotheses in isolation and this is how they will be tested. However, several of these arguments may be interlinked. For example, one can posit a connection between the falling wages hypothesis and the credit supply hypothesis under the following rationale: in an environment of falling wages, households can use debt as a substitute for reduced wage income to maintain path-dependent, backward looking consumption norms only if they have access to credit. This means that the falling wage hypothesis works in interaction with the credit supply hypothesis. An exhaustive discussion of these possible interactions is beyond the scope of this paper. Our principal interest is the comparative explanatory relevance of each hypothesis that we have extracted from the literature.

3. Related empirical literature

Empirical studies on the drivers of household indebtedness have econometrically tested most of the abovementioned hypotheses. Studies on the macroeconomic determinants of private credit have also tested some of the discussed hypotheses (e.g Hofmann, 2004; Égert, Backé, and Zumer, 2007; Goodhart and Hofmann, 2008; Bordo and Meissner, 2012; Gu and Huang, 2014; Arestis and Gonzalez, 2014; Perugini et al., 2015). Although these studies provide important insights on the drivers

of private credit, we do not discuss them because as we are interested in household debt specifically, the nature of which is different from private credit. Private credit consists not only of household credit, but of business credit as well. As a result, the arguments and subsequent empirical analyses on the drivers of private credit differ substantially from those on household indebtedness. Consequently, the empirical studies on determinants of household debt is more closely related to our study, and as such, we focus our empirical review on the empirical studies concentrating on household debt. We review both time-series and panel empirical studies, with emphasis placed on the panel studies as they are directly comparable to our study. Table 2 summarises the existing econometric studies on the drivers of household debt.³

Table 2: Summary of econometric studies on the macroeconomic drivers of household debt

Author(s)	Explanatory variables	Sample; Estimation method	Hypotheses tested
Time series empirical studies on the drives of household debt			
Kohn and Dynan (2007)	HP, DEMOG, Y	US, SCF data in waves, 1983, 1989, 1992, 1995, 1998, 2001, 2004; OLS	HPH (+)
Oikarinen (2009)	HP, GDP, i, SP	Finland, 1975-2006; VECM, Granger causalities	HPH (+) FAH (0)
Gimeno and Martinez-Carrascal (2010)	HP, WS,i, iN	Spain, 1984 – 2009; CVECM	HPH (+)
Valverde and Fernandez (2010)	HP, HP/rentY, iM, GROSSWAGE, i, GDP PC, DEFAULT, IBEX-35	Spain, 1988 – 2008; VECM	HPH (+) FWH (+)
Meng et al. (2013)	DWELLINGS, HP, i, UE, GDP, POP	Australia, 1988 – 2011; CVAR	HPH (+) LIH (-)
Anundsen and Jansen (2013)	HP, Yd, STOCK, HTURNOVER, e, iL, iPT, HSTARTS, HI, CC, DEP	Norway, 1986 – 2008; CVAR, SVECM	HPH (+)
Panel empirical studies on the drives of household debt			
Malinen (2014)	TOP1, I, GDP PC, M2, i	8 OECD countries, 1960-2008; DSUR panel co-integration, FD estimations, Granger causalities	ECH (0) LIH (-)

³ Expanded summaries the empirical studies on the macroeconomic drivers of household indebtedness are provided in table A1 in the appendix.

Rubaszek and Serwa (2014)	HP, IC, Yunc, YPC, i, UE	36 high- and middle-income countries, 1995-2009; Panel co-integration	HPH (+)
Klein (2015)	TOP1, Pareto-Lorenz, WS, GINI	9 OECD countries, 1953-2008; Panel co-integration	ECH (+)
Stockhammer and Wildauer (2017)	HP, Yd, i, AGE, TOP1, GINI,	11 OECD countries, 1980-2011; ECMs with DFE and PMG	HPH (+) ECH (0) ASH (+) CSH (-)

Notes: **Hypothesis abbreviations:** HPH refers to the house price hypothesis, FAH refers to the financial asset hypothesis, ECH refers to the expenditure cascades hypothesis, FWH refers to the falling wages hypothesis, LIH refers to the low interest rate hypothesis, ASH refers to the age structure hypothesis and CSH refers the credit supply hypothesis. (+) indicates a statistically significant and positive effect on household debt. (-) indicates a statistically significant and negative effect on household debt. (0) indicates no statistically significant effect on household debt. **Estimation method abbreviations:** CVAR refers to co-integrated vector auto-regressive approach; OLS refers to ordinary least squares regression approach; VECM refers to vector error-correction models; SVECM refers to structural vector equilibrium correcting model; FD estimations refer to first-differenced estimations; ECM refers to error correction model; DFE refers to dynamic fixed effects estimator; PMG refers to pooled mean group estimator. **Variable abbreviations:** WS refers to labour income; iN refers to nominal interest rate; iS refers to interest rate on savings deposits less money market rates, iSPREAD refers to the difference between the lending and deposit rate; HP refers to house prices, DEMOG refers to demographic variables (age of head, age of head squared, age of head cubed, head has high school degree, head has college degree, where head refers to the head of the household), Y refers to income; Yd refers to household disposable income; GDP refers to real GDP, i refers to the real interest rate, SP refers to real stock prices; HP/rentY refers to house prices/rental income; iM refers to nominal mortgage credit interest rate; GROSSWAGE refers to real gross salary per employee; GDP PC refer to GDP per capita; DEFAULT refers to mortgage credit default rate; IBEX-35 refers to the rate of variation in IBEX-35 in the Spanish stock exchange; DWELLINGS refers to no. of new dwellings approved; UE refers to the unemployment rate; POP refers to population; STOCK refers to housing stock; HTURNOVER refers to housing turnover; e refers to expectations; iL refers to nominal interest rate on loans; iPT refers to real post-tax interest rate; HSTARTS refers to housing starts; HI refers to investments in housing; CC refers to construction costs; DEP refers to rate of depreciation of housing stock; TOP1 refers to the income share of the top 1%; I refers to investment/GDP; M2 refers to M2/GDP; Yunc refers to income uncertainty; YPC refers to income per capita; PL refers to the inverted Pareto-Lorenz coefficient; AGE refers to fraction of population aged 65 and older; GINI refers to the Gini coefficient and CRED refers to a credit regulation index.

Of our hypotheses on the macroeconomic drivers of household debt, those that have been tested econometrically are the house price hypothesis, the financial asset hypothesis, the expenditure cascades hypothesis, the falling wages hypothesis, the age structure hypothesis, the credit supply hypothesis and the low interest rate hypothesis, albeit not together in one study. The welfare retrenchment hypothesis has not yet been tested. Except for Stockhammer and Wildauer (2017), who test the house price hypothesis, expenditure cascades hypothesis, age structure hypothesis and credit supply hypothesis together in one study, the existing econometric studies on household debt tend to test only one or two of our hypotheses on the macroeconomic drivers of household debt.

The house price hypothesis is tested by using house prices as an explanatory variable for household debt both in panel (Rubaszek and Serwa, 2014; Stockhammer and Wildauer, 2017) and time series work (Kohn and Dynan, 2007; Oikarinen, 2009; Gimeno and Martinez- Carrascal, 2010;

Valverde and Fernandez, 2010; Anundsen and Jansen, 2013; Meng et al., 2013). The econometric approaches used are panel co-integration (Rubaszek and Serwa, 2014), error correction models (ECMs) (Stockhammer and Wildauer, 2017), in the panel econometric studies, and a vector error correction model (VECM) and Granger causality tests (Oikarinen, 2009), co-integrated vector autoregressive models (CVAR) (Gimeno and Martinez- Carrascal, 2010; Anundsen and Jansen, 2013; Meng et al., 2013) and OLS regressions (Kohn and Dynan, 2007), in the time-series studies. Despite the differences in sample and method used, all the papers find a positive and significant impact of house prices on household indebtedness. This is critical as it suggests that house prices are robust drivers of household indebtedness.

Regarding the financial asset hypothesis, Oikarinen (2009) applies a VECM and Granger causality tests to Finnish time-series data from 1975 – 2006 in the search for evidence for a co-integrating relationship and direction of causality between housing loans and stock prices. He does not find any evidence for a long-run relationship between housing loans and stock prices, and the Granger causality tests do not provide information on the direction of causality between housing loans and stock prices. The financial asset hypothesis is not supported by Oikarinen's work.

To test the expenditure cascades hypothesis, Malinen (2014), Klein (2015) and Stockhammer and Wildauer (2017) use the income share of the top 1% as an explanatory variable for household debt accumulation in panel studies. Malinen (2014) uses panel co-integration, first difference estimations and Granger causality tests for 8 OECD countries to determine the impact of the top 1% income share on household debt. He does not find evidence for an effect of the top 1% income share on household debt. Klein (2015) uses panel co-integration to test the bivariate relationship between the top 1% income share and household debt for 9 OECD countries. He finds a statistically significant positive effect of the top 1% income share on household debt. Stockhammer and Wildauer (2017) apply ECMs to data from 11 OECD countries. They fail to find a robust statistically significant relationship between the top 1% income share and household debt. Notably, Klein (2015) applies bivariate co-integration testing, which may explain why he finds evidence of a co-integrating relationship, whereas Malinen (2014) and Stockhammer and Wildauer (2017) do not. Determining the drivers of household debt involves more than examining the relationship between household debt and just one other variable. This contradiction in the findings of Malinen (2014) and Stockhammer and Wildauer (2017) on the one hand, and Klein (2015) on the other hand, emphasises the need to include other variables in the empirical determination of household debt, which is the purpose of this paper.

Akin to the papers referred to above, we also use the income share of the top 1% to capture the expenditure cascades hypothesis when testing the competing determinants of household indebtedness. The share of total income which is received by the richest 1% of households communicates the dynamics at the top of the distribution, and the expenditure cascades hypothesis predicts that household debt accumulation is sparked by concentration of income at the top. Thus, focus on incomes at the top end of the distribution is appropriate for the expenditure cascades hypothesis as households are looking up to the next income group, and thus indirectly to the very top. The other standard measures of income inequality, the Gini coefficient is less useful in this context because it is more sensitive to the middle rather than the upper tail of the distribution, which is key for the expenditure cascades hypothesis. Therefore, the income-share of the top 1% is standard in the empirical literature as a proxy for the expenditure cascades hypothesis (see, for example, Malinen, 2014; Klein, 2015 and Stockhammer and Wildauer, 2017). In addition to this norm in the related empirical literature, the “trickle-down consumption” literature has shown that rising incomes and consumption at the top of the income distribution induce households in the lower tiers of the distribution to increase their consumption (see, for example, Bertrand and Morse, 2013 and Alvarez-Cuadrado and Japaridze, 2017).⁴ This provides empirical support for our use of the top 1% income share.

Meng et al. (2013) and Malinen (2014) econometrically test the low interest rate hypothesis, using time series and panel analyses respectively. Meng et al. (2013) include the official interest rate of the Reserve Bank of Australia, which is the interest rate used for monetary policy. Meng et al. 's (2013) application of a CVAR model to test the determinants of household indebtedness in Australia reveals that the official interest rate has a statistically significant, robust and negative long-run effect, which is in line with the low interest rate hypothesis. Malinen (2014) uses the short-term interest rate obtained from the dataset of Schularik and Taylor (2012). He also finds a negative relationship between the short-term interest rate and the household debt. However, the robustness of this statistical relationship is not clear, as out of four specifications presented, the short-term interest rate is included in three, and out of these three specifications, it is statistically significant in two. Nonetheless, one can interpret this as evidence in support of the low interest rate hypothesis. Chrystal

⁴ In practice, testing the expenditure cascades hypothesis as a driver of household debt, and using the income share of the top 1% as a proxy for this hypothesis, is warranted if income growth and consumption growth are closely linked at the top. If the rich got richer, but don't spend more, there would be no cascade. The empirical support for expenditure cascades is relatively recent, and the results are mixed (see, for example, Christen and Morgan, 2005; Leigh and Possi, 2009 and Drechsel-Grau and Schmid, 2014). That the results of the effects of expenditure cascades on household debt accumulation at present are also mixed and limited calls for further testing of the expenditure cascades hypothesis.

and Mizen (2005), Oikarinen (2009), Gimeno and Martinez- Carrascal (2010), Valverde and Fernandez (2010), Anundsen and Jansen (2013), Rubaszek and Serwa (2014) and Stockhammer and Wildauer (2017) also control for interest rates in their econometric testing of household indebtedness.⁵ However, the interest rates used in these studies are not the interest rate required for the low interest rate hypothesis, which is the Central Bank (CB) policy rate, typically used to target interbank rates. Although the interest rates used in the abovementioned studies tend to be highly correlated with the CB policy rate, implying that they may have a comparable effect on household debt, using the closest proxy to the CB policy would best captures the explanatory power of the low interest rate hypothesis. The closest proxies are the overnight interest rate and the short-term interest rate. The studies by Chrystal and Mizen (2005), Oikarinen (2009), Gimeno and Martinez- Carrascal (2010), Valverde and Fernandez (2010), Anundsen and Jansen (2013), Rubaszek and Serwa (2014) and Stockhammer and Wildauer (2017) do not include such interest rates.

Regardless of the hypotheses they test, the panel studies on the determinants of household debt use panel co-integration estimations, which allow for both the long-run and short-run determinants of household debt. This study also uses a panel co-integration estimation technique. The difference between this paper and the existing panel studies on household debt is twofold. Firstly, we test seven hypotheses on the drivers of household debt in one regression framework. Given data availability, we control for the maximum number of arguments possible in the debate on the causes of household indebtedness. Secondly, we explicitly allow for the cycle-dependent asymmetric effects of household debt accumulation. That is, we allow for the building up and deleveraging of debt that may be influenced by the boom and bust periods of household debt and house price cycles. Only Stockhammer and Wildauer (2017) have explicitly catered for asymmetric behaviour in household debt accumulation thus far.

Thus, to synthesise the previously discussed hypotheses about household debt accumulation and the existing empirical literature, we use the following household debt equation:

$$HHD = f(HP, SP, TOP1, WAGES, WELFARE, AGE, i,) \quad (1)$$

(+)
(+)
(+)
(-)
(-)
(-)
(-)

We estimate household debt (households' loans and debt securities) as a percentage of GDP (*HHD*) as a function of the natural logarithm of real house prices (*HP*), the natural logarithm of real stock prices (*SP*), the top 1% share of income (*TOP1*), the natural logarithm of real average wages

⁵ A comparison of the interest rates used in the studies is presented in Table A2 in the appendix.

(*WAGES*), state welfare spending on housing, health and education as share of GDP (*WELFARE*), the fraction of population aged 65 and older (*AGE*) and the real short-term interest rate (*i*). Data used, and the full econometric details are presented in the following section.

4. Data and econometric results

4.1 Data

The dataset consists of annual data in an unbalanced panel of 13 OECD countries (Australia, Belgium, Canada, Germany, Spain, Finland, France, United Kingdom, Italy, Japan, Norway, Sweden and the US) over the period 1993-2011. The period is selected due to data availability. *HHD* and *HP* are from the Bank of International Settlements (BIS). *TOP1* is from the World Wealth and Income Database. *AGE* is from the World Bank database. *SP*, *WAGES*, *i* and *WELFARE* are from OECD data bases. All variables are in real terms. Definitions, links to data sources and descriptive statistics of variables are provided in Tables A3 and A4 in the Appendix.

Data on *WELFARE* are available for most of the countries in our sample from the late 1990's, and for some it is available only from the 2000's (e.g. for Belgium it is available from 2000 and for Japan it is available from 2005). Including *WELFARE* in our estimations restricts our sample to 179 observations, while excluding it allows for 234 observations. Consequently, we include *WELFARE* only in one specification reported here. We use *i*, the real short-term interest rate as proxy variable for the low interest rate hypothesis. Real overnight interest rates would also be an appropriate variable, but, as it is only available for 7 of our 13 countries, we do not utilise it. The credit supply hypothesis requires data on securitisation, market-based financial intermediation, and changes in financial regulation. Although we have an index of financial reforms, from Abiad et al. (2008) in the IMF's Database of Financial Reforms, and a credit regulation index, from the Fraser Institute, we cannot use these variables for two reasons. Firstly, the index of financial reforms and the credit regulation index reduce the number of observations and cross sections substantially as the index of financial reforms is available only until 2005, and the credit regulation index is available only in 5-year intervals from 1970 until 2000, after which it is available until 2014 on an annual basis. Secondly, these variables do not capture bank activities which reflect bank-side drivers of household debt, such as the use of off-balance sheet vehicles and securitisation.

4.2 Econometric results

The panel approach was selected due to limited length of data series which prevents single-country time series analysis. In its longest period, 1993 – 2011, our panel has a small N and a somewhat larger T (N=13, T=18). We tested all variables for non-stationarity using panel unit root tests which encompass both common and individual unit root tests. After first differencing, all panel unit root tests reject the null that unit roots are present at the 1 % level for all variables. Full details of the unit root tests for all variables are in Table A5 in the appendix. We test the baseline for co-integration using single equation error correction models (ECMs), as they have been shown to have more power than Engle-Granger residual-based tests (e.g. Kremers et al., 1992; Banerjee et al., 1998) and because existing panel studies on the drivers of household debt prefer the use of panel co-integration estimation techniques (Malinen, 2014; Rubaszek and Serwa, 2014; Klein, 2015; Stockhammer and Wildauer, 2017). The ECM test for co-integration is based upon the ordinary OLS coefficient (adjustment speed) of the lagged dependent variable in an autoregressive distributed lag model augmented with leads of the regressors. Our ECM takes the following form:

$$\begin{aligned} \Delta HHD_{it} = & \beta_1 HHD_{it-1} + \beta_2 HP_{it-1} + \beta_3 SP_{it-1} + \beta_4 TOP1_{it-1} + \beta_5 WAGES_{it-1} \\ & + \beta_6 WELFARE_{it-1} + \beta_7 AGE_{it-1} + \beta_8 i_{it-1} + \beta_9 \Delta HP_{it} + \beta_{10} \Delta SP_{it} + \beta_{11} \Delta TOP1_{it} + \\ & \beta_{12} \Delta WAGES_{it} + \beta_{13} \Delta WELFARE_{it} + \beta_{14} \Delta AGE_{it} + \beta_{15} \Delta i_{it} + \alpha_{it} + \mu_{it} \end{aligned} \quad (2)$$

The parameter of interest for testing co-integration is β_1 , which is the adjustment speed at which the dependent variable, *HHD*, returns to its equilibrium. If the adjustment speed is between -1 and 0, and statistically significant, then there is a genuine long-run, co-integrating relationship between the independent variables and the dependent variable. For most of our specifications the co-efficient and t-statistic of β_1 indicate co-integration. Consequently, we use the single equation panel ECM to test the relative explanatory power of our hypotheses on the drivers of household debt. Fixed effects (FE) are included as they are significant to the estimation. Panel-corrected standard errors (PCSE), which account for autocorrelation and heteroskedasticity, are applied in all specifications.

Table 3 below presents the results of the panel ECM. We report 8 specifications. Specification 1 is our baseline ECM. It tests the relative long-run and short-run explanatory power of the house price hypothesis, financial asset hypothesis, expenditure cascades hypothesis, falling wages hypothesis, age structure hypothesis and low interest rate hypothesis. In specification 2 a time dummy

for the year of the global financial crisis is applied to capture the effect of the crisis on household debt accumulation. The time dummy is added for the following reason: initially we included both country and period fixed effects and tests indicated that both were statistically significant. Inspection of the time effects made apparent that the relevance of the period effects was driven by the dummy for the year 2009, which is intuitively plausible given the global financial crisis. Thus, instead of applying period effects for every year of the estimation, specification 2 includes a time dummy for the year 2009 only. Specification 3 includes *WELFARE*, and thus has a smaller number of observations than in the baseline. Specification 4 applies up to two lags to all the first differenced explanatory variables, to determine if their effect on household debt is driven by past years, i.e., if their effect has a lagged component. Only the variables with statistically significant lag effects are included in the table below. In specifications 5 and 6 we restrict the sample to observations characterized by $\Delta HHD > 0$ and $\Delta HHD < 0$, respectively. The restriction is done to allow for asymmetric effects of the explanatory variables on household debt under different phases in the household debt cycle. Similarly, given that most countries in our sample experienced house price bubbles under our period of study, in specifications 7 and 8 we restrict the sample to observations characterized by $\Delta HHP > 0$ and $\Delta HHP < 0$ respectively, to capture the effects of the explanatory variables on household debt under the boom and bust phases in the house price cycle. In the specifications presented here, we find a co-integrating relationship in all specifications except in specification 6, which captures the downturn in the debt cycle. This suggests that there is a genuine long-run relationship between the explanatory variables and household debt when households are building up debt, but this relationship breaks down when households are de-leveraging. The lagged dependent variable is included in all the specifications presented here, as it robustly statistically significant, and including it has the benefit of further catering for autocorrelation and improving the adjusted R^2 greatly.

Table 3: Econometric results

specification	1 baseline	2 dummy 2009	3 WELFARE	4 lagged effects	5 Δ HHD>0	6 Δ HHD <0	7 Δ HP>0	8 Δ HP<0
sample period	1993-2011	1993-2011	1994-2011	1995-2011	1993-2011	1993-2011	1993-2011	1993-2011
adjustment speed	-0.0745*** (-4.6829)	-0.0817*** (-5.3228)	-0.0887*** (-3.4359)	-0.0772*** (-3.2715)	-0.0600*** (-3.4357)	-0.0483 (-1.3672)	-0.0798*** (-3.9482)	-0.0557* (-1.8907)
long-run coefficients								
HP (-1)	0.0439*** (4.5226)	0.0418*** (4.6916)	0.0447*** (3.9558)	0.0428*** (2.8456)	0.0339*** (3.1296)	0.0214 (1.0548)	0.0407*** (3.1586)	0.0322* (1.6811)
SP (-1)	-0.0028 (-0.7279)	0.0009 (0.2535)	-0.0059 (-1.3103)	-0.0059 (-0.8676)	-0.0063 (-1.4403)	0.0142** (2.0834)	-0.0021 (-0.4157)	-0.0073 (-0.8321)
TOP1(-1)	-0.0456 (-0.3087)	-0.0317 (-0.2313)	-0.0736 (-0.6298)	-0.2327 (-1.3491)	0.0191 (0.1173)	-0.2236 (-0.8531)	0.0133 (0.0737)	-0.2879 (-0.7935)
WAGES (-1)	-0.0157 (-0.5431)	-0.0364 (-1.3719)	0.0057 (0.1887)	0.0203 (0.4767)	0.0145 (0.4349)	-0.1391* (-2.0262)	0.008386 (0.2119)	0.0384 (0.4929)
WELFARE (-1)			-0.0832 (-0.4566)					
AGE (-1)	-0.0010 (-1.4839)	-0.0012** (-2.0486)	-0.0000 (-0.1083)	-0.0014 (-1.6479)	-0.0015* (-1.8269)	-0.0021 (-1.5036)	-0.0013 (-1.155)	0.0009 (0.7219)
i (-1)	0.0089 (0.1095)	0.0281 (0.3793)	-0.0497 (-0.5112)	-0.0795 (-0.5606)	-0.0308 (-0.3376)	-0.1319 (-0.6963)	0.0057 (0.0511)	-0.0022 (-0.0119)
short-run coefficients								
Δ HP	0.1090*** (4.7951)	0.1267*** (5.9814)	0.1195*** (5.6626)	0.1076*** (3.4947)	0.1191*** (4.6141)	0.0247 (0.6143)	0.1065*** (2.6212)	0.0600 (0.8293)
Δ SP	-0.0178*** (-3.0497)	-0.0064 (-1.1838)	0.0064 (1.0321)	-0.0144 (-1.7158)	-0.0172*** (-2.8128)	0.0022 (0.2119)	-0.0209*** (-2.8492)	-0.0196 (-1.4057)
Δ TOP1	-0.0092 (-0.0497)	0.0153 (0.0879)	0.0326 (0.2574)	-0.036 (-0.2259)	0.1057 (0.5522)	-0.5419* (-1.8099)	0.0029 (0.0161)	-0.0840 (-0.1482)
Δ WAGES	0.1441**	0.0920	0.0701	0.1548	0.0849	0.0495	0.2505**	0.2053

	1.9986	(1.4196)	(0.9812)	(1.9492)	(1.1028)	(0.5205)	(2.4892)	(1.4536)
Δ WELFARE			1.8505*** (7.3558)					
Δ AGE	-0.0034 (-0.7303)	-0.0023 (-0.5599)	-0.0000 (-0.0192)	-0.0016 (-0.2156)	0.0039 (0.7557)	-0.0038 (-0.3686)	0.0030 (0.4733)	-0.0107 (-1.1556)
Δ i	0.0135 (0.1343)	-0.0365 (-0.4007)	0.1102 (1.1424)	0.0095 (0.0848)	0.0329 (0.2837)	-0.3454** (-2.1822)	0.1521 (1.0991)	-0.0962 (-0.5306)
dummy2009		0.0289*** (6.8893)						
Δ HHP (-2)				0.0751*** (2.4798)				
Δ TOP1 (-2)				0.4114** (2.4683)				
Δ HHD (-1)	0.5117*** (7.6089)	0.5696*** (9.2791)	0.4759*** (6.6594)	0.5599*** (5.2693)	0.33785*** (4.1639)	0.3477*** (2.9974)	0.4695*** (5.4803)	0.5078*** (3.352)
DW	1.9552	1.8955	1.9684	1.9562	1.7565	2.6416	1.7638	2.2036
r ² adj	0.6503	0.7155	0.7753	0.6982	0.5009	0.4733	0.6884	0.4675
observations	234	234	179	208	171	63	151	83
cross-sections	13	13	12	13	13	13	13	13

Notes: All specifications include FE and PCSE. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Results are written as: coefficient (t-value). Adjustment speed refers to the coefficient of HHD (-1). DW refers to the Durbin-Watson statistic, while r²adj refers to the adjusted R-squared.

In specification 1, the baseline specification, *HP*, which captures the house price hypothesis, is statistically significant in the long-run and in the short-run at the 1% level. All other variables are statistically insignificant regarding long-run effects. *SP*, which captures the financial asset hypothesis, and *WAGES*, which captures the falling wages hypothesis, have short-run effects that are statistically significant at the 1% and 5% levels respectively, but the signs are perverse. In specification 2, the dummy variable for the year 2009 which captures the effect of the global financial crisis, is positive and statistically significant at the 1% level. *HP* has the expected positive sign and statistically significant in the long-run and in the short-run at the 1% level in specification 2. *AGE* is also statistically significant in the long-run, and it has the negative sign as predicted by the age structure hypothesis. Specification 3 includes *WELFARE*. Compared to specifications 1 and 2, which both have 234 observations and 13 cross-sections, specification 3 has 179 observations and 12 cross-sections. In specification 3 *HP* is positive and statistically significant in the long-run and in the short-run at the 1% level. *SP* has statistically significant effects in the short-run at the 10% level, carrying the positive sign in line with the financial asset hypothesis. *WELFARE* is statistically significant at the 1% level, in the short-run, but with positive sign, which contradicts the expected sign of the welfare retrenchment hypothesis. In specification 4, where we capture the effect of the explanatory variables from previous years on household debt, *HP* is again statistically significant at the 1%, both in the long- and short-run. ΔHP (-2) and $\Delta TOP1$ (-2), i.e., the values of ΔHP and $\Delta TOP1$ from two years ago, are statistically significant at the 1% and 5% levels respectively, and with the expected signs. This suggests that both *HP* and *TOP1* have lagged effects on household debt. Nothing else is statistically significant in specification 4.

Specifications 5 and 6 allow coefficients to differ for periods of increasing and decreasing household debt to capture the effects of the debt cycle. Specification 5 captures the effects during the boom phase of the debt cycle. Again, *HP* is positive and statistically significant in the long-run and in the short-run at the 1% level. *AGE* is statistically significant in the long-run, with the expected negative sign, and *SP* is statistically significant in the short-run, albeit with a perverse sign. Specification 6 covers the bust phase of the debt cycle, i.e. periods of household de-leveraging, and has 171 observations. For this sample, we fail to find evidence for co-integration between the dependent variable and the explanatory variables. This indicates that our results are driven by the boom period of the debt cycle. Specifications 7 and 8 isolate for periods of the housing boom and bust respectively, to obtain the differential effects of the explanatory variables on household indebtedness during the two phases of the house price cycle. Specification 7 covers the upswing of the house price cycle and has 151

observations. *HP* is positive and statistically significant in the long-run and in the short-run at the 1% level. The results of specification 7 reveal that in the short-run *SP* and *WAGES* are statistically significant at the 1% and 5% levels respectively. However, the signs of *SP* and *WAGES* in specification 7 are not in line with the signs of their respective hypotheses. In specification 8, which represents the downturn of the house price cycle and has 83 observations, the co-integrating relationship is relatively weak. *HP* is positive and statistically significant in the long run only at the 10% level. No other explanatory variable is statistically significant in specification 8.

Importantly, our results reveal that accounting for asymmetric debt accumulation and the cyclical dimensions of the debt and housing cycles is relevant. In specification 6, which captures the downturn of the debt cycle and thus a period in which households are deleveraging, the co-integrating relationship between the explanatory variables and the dependent variable is not statistically significant. Similarly, in specification 8, which represents the house price bust, the co-integrating relationship between the explanatory variables and the dependent variable becomes weak. These results indicate that the statistically significant findings on the drivers of household debt accumulation are primarily due to behaviour during in upswings. Although our results are consistent with asymmetric effects over of debt and housing cycles, we are hesitant to draw strong conclusions. The lack of a statistically significant relationship between household debt and house prices in bust phases may simply be due to the fact that the downswing subsample is smaller than the upswing subsample. This, in turn, may imply that our sample period is unable to identify the exact relationship between household debt and house prices when both variables are in decline. If, for example, during house prices and household debt busts, the impacts of house prices on household deleveraging is particularly sluggish, more years of the downswing in our sample would be required to detect the type of relationship between household debt and house prices. In addition, the deleveraging process could be sluggish relative to the process of debt build-up due to institutional reasons, and this may be reinforced by policy responses as deleveraging episodes will typically be associated with a crisis.

Despite these qualifications, the finding that the strength of the relationship between household debt and house prices is affected by the phases of the household debt and house price cycle is meaningful. As discussed in section 3, in the empirical literature on the determinants of household debt accumulation, allowing for the building up and deleveraging of debt that may be influenced by the boom and bust periods of household debt and house price cycles have only been accounted for by Stockhammer and Wildauer (2017) thus far. Thus, our finding is a valuable addition to the empirical work on household debt accumulation.

Overall, our ECM results show that the most robust driver of household debt in the long-run and the short-run is *HP*. Thus, our statistical results robustly support the house price hypothesis. We find weak support for the age structure hypothesis in the long run, as in the long-run, *AGE* is statistically significant at the 5% and 10% levels and carries the expected sign. Our results also provide weak evidence against both the financial asset hypothesis and the falling wages hypothesis in the short-run. Additionally, testing short-run effects using lagged values revealed that *HP* and *TOP1* have lagged effects on household debt. For *TOP1*, which is otherwise consistently statistically insignificant, this finding suggests that its relationship with household debt is based on its past values. This warrants further research which focuses on the lagged effects of *TOP1* on household debt accumulation.

As *HP* is the only robust variable in the long and short-run, we test for inverse causality between it and the dependent variable, *HHD*, using Granger causality tests. As shown in Table 4 below, the Granger causality tests show that in first differences with two lags, *HP* is a useful predictor of *HHD*, while the reverse does not hold. The results of the Granger causality tests point to the causality going from *HP* to *HHD*. This is consistent with our interpretation of the ECM regression results that real residential house prices are robust drivers of household indebtedness.

Table 4: Results of Granger causality tests

Null hypothesis	Period	OBS	F-Stat
HHD does not cause HP	1993 -2011	247	0.2362
HP does not cause HHD			26.6961***

Notes: * p<0.1, ** p<0.05, *** p<0.01

Summing up, the results from the ECM estimations, supported by Granger causality tests, show that real residential house prices are the most robust determinants of household debt, and support the house price hypothesis. Our results lend some evidence to the age structure hypothesis in the long-run. Additionally, our statistical results point to some evidence against the financial asset hypothesis and the falling wages hypothesis in the short-run, and but no evidence for these hypotheses in the long-run. For the expenditure cascades hypothesis, we find weak evidence for short-run effects when longer lags are considered. This suggests that future testing of the expenditure cascades hypothesis requires an approach that emphasises lag structure. We do not find support for the welfare retrenchment hypothesis nor the low interest rate hypothesis in the long or short-run. Our results also uncover that accounting for cyclicalities is relevant. Our specifications which capture the booms and busts of the debt and house price cycles respectively show that the co-integrating

relationships between the explanatory variables and the dependent variables holds in the boom phases, but not in the bust phase of the debt cycle and is weak in the bust phase of the house price cycle. We also find the existence of asymmetric effects on household debt accumulation behaviour which are dependent on the phases of debt and house price cycles.

It is important to situate our results within the related literature. Our results in relation to the existing empirical literature on the macroeconomic determinants of household debt accumulation are summarised in Table 5. We have tested three hypotheses, the financial asset hypothesis, the falling wages hypothesis, and the welfare retrenchment hypothesis, which have not been considered in existing studies. Our finding of real residential house prices being most robust supports the findings of previous time-series and panel econometric work of the relationship between house prices and household debt (Kohn and Dynan, 2007; Oikarinen, 2009; Gimeno and Martinez- Carrascal, 2010; Anundsen and Jansen, 2013; Meng et al., 2013; Rubaszek and Serwa, 2014; Stockhammer and Wildauer, 2017), suggesting that research on the determinants of household debt which exclude house prices have omitted a central variable. This calls for inclusion of real residential house prices in empirical work related to the drivers of household indebtedness. Our results also highlight that the relationship between real residential house prices and household indebtedness, as well as debt and house price cycles are key areas for further understanding of household debt accumulation behaviour.

Table 5: Comparison of our econometric findings with findings of existing panel econometric studies

Hypotheses tested	Finding	Consistency with the findings of existing panel econometric studies on the drivers of household indebtedness
1. House price hypothesis	Robust evidence for long- and short-run effects. Hypothesis supported	Consistent with Rubaszek and Serwa (2014), and Stockhammer and Wildauer (2017)
2. Financial asset hypothesis	Some evidence for rejection of the hypothesis in the short-run; no evidence for long-run effects	Not tested by other authors using panel econometrics
3. Expenditure cascades hypothesis	No evidence for long- or short-run effects. Hypothesis rejected	Consistent with Malinen (2014) and Stockhammer and Wildauer (2017), inconsistent with Klein (2015)
4. Falling wages hypothesis	Some evidence for rejection of the hypothesis in the short-run; no evidence for long-run effects	Not tested by other authors using panel econometrics

5. Welfare retrenchment hypothesis	No evidence for long- or short-run effects. Hypothesis rejected	Not tested by other authors using panel econometrics
6. Age structure hypothesis	Some evidence for support of the hypothesis in the long-run; no evidence for short-run effects	Inconsistent with Stockhammer and Wildauer (2017)
7. Low interest rate hypothesis	No evidence for long- or short-run effects. Hypothesis rejected	Inconsistent with Malinen (2014)

5. Conclusion

The objective of this paper has been to provide an empirical clarification on the macroeconomic determinants of household debt by econometrically testing the house price hypothesis, the financial asset hypothesis, the expenditure cascades hypothesis, the falling wages hypothesis, the welfare retrenchment hypothesis, the age structure hypothesis and the low interest rate hypothesis respectively together in a comprehensive framework. We used error correction models to test both long-run and short-run effects for 13 OECD countries over the period 1993 – 2011. We also investigated whether the determinants of household debt depend on phases of debt and house price cycles. The results show that real residential house prices is the most robust determinant of household indebtedness in the long-run and the short-run, and that the explanatory variables have cycle-dependent asymmetric effects on household debt accumulation. Our results indicate that household debt accumulation is primarily an outcome of residential real estate transactions, that the phase of the debt and house price cycles matters, and that our results are driven by the boom periods. Additionally, Granger causality tests suggest causality going from real residential house prices to household debt.

These findings have important implications both for economic research and for economic policy. In terms of research it highlights the need for macroeconomic models that explicitly model the real estate market and household debt (e.g. Ryo 2016). Our study has used a panel approach with a relatively recent sample. Future research should investigate whether our results hold over longer time periods and whether they hold at the country level. The finding of asymmetric behaviour in the boom and bust raises interesting questions, but it is doubtful whether the changes in the behavioural functions can be identified with macroeconomic data. For future research, using household-level data is a more promising route here. In terms of economic policy our findings reinforce ongoing debates

on macroprudential policies. Since unsustainable levels of household debt are known to threaten financial and macroeconomic stability (Mian and Sufi, 2014), central banks should act against debt and real estate booms. To do so they will have to change policies and actively lean against markets and they will have to broaden their economic policy repertoire.

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7. Appendix

Table A1: Summary of existing empirical studies which test some of the competing hypotheses on the macroeconomic determinations of household debt

Author(s)	Sample	Estimation method(s)	Dep. Var(s)	Indep. vars	Hypotheses tested
Chrystal & Mizen (2005)	UK data, 1979Q1 to 1998Q4	CVAR	unsecured household sector credit (excludes mortgages)	real consumer expenditure by households, real M4 balances held by households, real net labour income, household real net total wealth (wt) (defined as housing wealth plus financial assets minus total debt), interest rate on savings deposits less money market rates, (rdt), which measures the return to bank deposits held as savings; and spread of the consumer credit rate over base rate, (rct), which measures the relative cost of unsecured household borrowing.	HPH (+)
Kohn & Dynan (2007)	US	OLS regressions	household debt/income	house prices , age (demographic) variables and log of income	HPH (+)
Oikarinen (2009)	Finish data, quarterly 1975-2006	VECM, Granger causalities	household liabilities/GDP	real house prices , real GDP, the real interest rate, real stock prices	HPH (+) FAH (0)
Gimeno & Martinez-Carrascal (2010)	Spanish data, 1984Q1 to 2009Q1	CVAR	mortgage loans	house prices , labour income (wage share), nominal and real interest rates	HPH (+)
Valverde & Fernandez (2010)	Spanish data, 1988Q4 to 2008Q4	VECM	real mortgage credit per household	house prices , house prices/rental income (housing rental revenues index), nominal mortgage credit interest rate, real salary per employee, real interest rate, GDP per capita, mortgage credit default rate, rate of variation in the stock exchange IBEX-35	HPH (+)

Meng et al. (2013)	Australian data, 1988Q2 to 2011Q2	CVAR	nominal household liabilities	Number of new dwellings approved, housing price , Interest rate, Unemployment rate, GDP – Gross domestic product, Population, CPI	HPH (+)
Anundsen & Jansen (2013)	Norwegian data, 1986Q2 to 2008Q4	cointegrated VAR (CVAR) for long-term dynamics structural vector equilibrium correcting model (SVEC) for short-term dynamics	real household liabilities	House prices , household disposable income (excluding equity income), housing stock, housing turnover, expectations, nominal interest rate on loans, real after-tax interest rate, housing starts, investments in housing, construction costs, rate of depreciation of housing stock	HPH (+)
Malinen (2014)	Panel of 8 OECD countries, 1960-2008	Panel co-integration, FD estimations, Granger causalities	household loans/GDP	the top 1 % income share , investment/GDP, real GDP per capita, M2/GDP, short-term interest rate	ECH (0) LIH (0)
Rubaszek & Serwa (2014)	Panel of 36 countries, 1995-2009	Panel co-integration	household debt/GDP	spread (difference between the lending and deposit rates), individual income uncertainty, disposable income per capita , real interest rate, unemployment rate, house prices .	HPH (+)
Klein (2015)	Panel of 9 OECD countries, 1953-2008	Panel co-integration	real credit to private households	top 1% income share , inverted Pareto-Lorenz coefficient, labour share of income (the wage share), the Gini index.	FWH (-) ECH (+)
Stockhammer & Wildauer (2017)	Panel of 11 OECD countries, 1980-2011	ECMs, with DFE and PMG	total credit to the household sector	household disposable real gross income, real long-term interest rate , fraction of population aged 65 and older , top 1% income share , Gini coefficient (pre-tax and post transfer), real property prices , Fraser Credit Regulation index	HPH (+) ECH (0) ASH (+) CSH (-)

Notes: (+) indicates a statistically significant and positive effect on household debt. (-) indicates a statistically significant and negative effect on household debt. (0) indicates no statistically significant effect on household debt: **HPH** refers to the house price hypothesis, **FAH** refers to the financial asset hypothesis, **ECH** refers to the expenditure cascades hypothesis, **FWH** refers to the falling wages hypothesis, **LIH** refers to the low interest rate hypothesis, **ASH** refers to the age structure hypothesis and **CSH** refers the credit supply hypothesis.

Table A2: Summary of interest rates used in time-series and panel papers on the determinants of household indebtedness

Paper	Interest rate (s) used	Best available interest rate proxy for low interest rate hypothesis? Y or N
Chrystal and Mizen (2005)	Interest rate on savings deposits less money market rates. This measures the return to bank deposits held as savings.	N
	Spread of the consumer credit rate over base rate. This measures the relative cost of unsecured household borrowing.	N
Oikarinen (2009)	Real interest rate. Refers to the real after-tax lending rate and the real before tax lending rate.	N
Gimeno and Martinze-Carrasca (2010)	Nominal and real interest rate	N
Valverde and Fernandez (2010)	Nominal mortgage credit interest rate. Refers to the average mortgage credit interest rate.	N
	Real interest rate	N
Meng et al. (2013)	The "official interest rate" of the reserve bank of Australia. This is the Australian Central Bank's base rate. Banks pay this interest rate when they take out a loan with a maturity of 1 day from another bank.	Y
Anundsen and Jansen (2013)	Nominal interest rate on loans. This is the nominal interest rate paid by households on loans in private financial institutions.	N
	Real post-tax interest rate. This is used to determine how much mortgage interest a household pays after tax has been deducted from the interest rate on mortgages.	N
Malinen (2014)	Short-term interest rate. This is obtained from the dataset of Schularik and Taylor (2012).	Y
Rubaszek and Serwa (2014)	Spread. This is the difference between the lending and deposit rates offered to households.	N
	Real interest rate. This is the real market lending rate obtained from the World Bank.	N
Stockhammer and Wildauer (2017)	Real long-term interest rate	N

Table A3: Full details of variables used in estimations and discussed in this paper

Variable	Measure	Unit	Additional details	Source
Variables used in estimations reported in this paper				
HHD	Real household debt/GDP	%GDP	Households' loans and debt securities/GDP See http://www.bis.org/statistics/totcredit.htm	BIS
HP	Real residential house prices	Index	See http://www.bis.org/statistics/pp_long.htm	BIS
SP	Real stock prices	Index	Prices of common shares of companies traded on national or foreign stock exchanges. See http://stats.oecd.org/Index.aspx?DataSetCode=MEI_FIN	OECD's MEI database
TOP1	Top-1% income share	% income	Exact definition may vary between countries. See http://wid.world/	World Wealth and Income Database
WAGES	Average annual wages per full-time and full-year equivalent employee in the total economy.	Thousands, National currency	See http://stats.oecd.org/	OECD
WELFARE	Government spending on welfare/GDP	%GDP	Combines government spending on health, education and housing. See https://data.oecd.org/gga/general-government-spending.htm	OECD
AGE	Fraction of population aged 65 and older	% of population	The ratio of older dependents- people older than 64-to the working-age population- those aged 15-64. Data are shown as the proportion of dependents per 100 working-age population. See http://databank.worldbank.org/data/reports.aspx?source=2&series=SP.POP.D.PND.OL&country=	World Bank
i	Real short-term interest rate	% per annum	Short term rates are usually either the three-month interbank offer rate attaching to loans given and taken amongst banks for any excess or shortage of liquidity over several months or the rate associated with Treasury bills, Certificates of Deposit or comparable instruments, each of three-month maturity. See http://stats.oecd.org/Index.aspx?DataSetCode=MEI_FIN	OECD's Main Economic Indicator (MEI) database

Variables discussed, but not used in estimations reported in this paper.

Credit Regulation Index	Index between [0,10]	index	Fraser Index, Subcategory 5A Credit Regulation Index, which consists of: percentage of privately held deposits, interest rate controls, private sector credit. A higher score indicates less regulation. See https://www.fraserinstitute.org/resource-file?nid=10159&fid=4820	Fraser Institute
Index of financial reforms	Index between [1,21]	index	Index of financial reforms measuring: credit controls, interest rate controls, entry barriers, state ownership in banking, capital account restrictions, supervision of the banking sector and securities market policy. Higher numbers represent liberal policies. See https://www.imf.org/external/pubs/cat/longres.aspx?sk=22485.0	IMF (Abiad et al. 2008 - A New Database of Financial Reforms)
Overnight interest rate	Overnight interest rate	% per annum	Overnight (or immediate rate) is a term used to describe the official discount rates at which central banks make advances to, or discount eligible bills of exchange for, selected banks and other financial intermediaries. See http://stats.oecd.org/Index.aspx?DataSetCode=MEI_FIN	OECD's Main Economic Indicator (MEI) database
Variables used to transform other variables				
GDP	Real GDP	Millions, national currency	See http://ec.europa.eu/economy_finance/ameco/user/serie/ResultSerie.cfm	AMECO
Consumer Price Index (CPI)	Index	Index	CPIs measure the average changes in the prices of consumer goods and services purchased by households, where the year 2010 is the base year. For more information on the full basket of goods and services contained in this CPI, see here http://unstats.un.org/unsd/cr/registry/regcst.asp?Cl=5 .	OECD

Table A4: Descriptive statistics of all variables, for the main estimation period, 1993 – 2011. All variables are in real terms.

Variable	Units	Mean	Median	Maximum	Minimum	Std. Dev.	Observations
HHD	%GDP, in decimal form	0.587659	0.601500	1.120500	0.173250	0.205143	247
HP	index	181.5991	165.3602	354.1958	60.73982	67.53085	247
SP	index	99.62329	98.00980	250.6961	22.82940	38.13139	247
TOP1	% income	0.094491	0.087215	0.183300	0.046983	0.030155	245
WAGES	thousands, national currency	397611.0	41499.00	4131088	25774.00	1064793.	247
WELFARE	% GDP, in decimal form	0.126898	0.126522	0.168955	0.097031	0.015068	193
AGE	% population	50.28702	50.54972	58.24497	43.17681	3.337744	247
i	%, in decimal form	0.019006	0.018030	0.068430	-0.033659	0.018654	237

Table A5: Description of outcome of panel unit root tests. All unit roots were tested using Levin-Lin-Chu test, Im-Pesaran-Shin test, ADF- Fisher Chi-square test and PP Fisher Chi-square test. All variables are in real terms.

Variable	Specification	Levin-Lin-Chu test	Im-Pesaran-Shin test	ADF- Fisher Chi-square test	PP - Fisher Chi-square test	Notes and required transformation
HHD	Intercept and trend	Not stationary in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Stationary in levels at the 10%. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Stationary only in 1 st differences according to all unit root tests. Must transform to 1 st differences.
HP	Intercept and trend	Not stationary in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Stationary only in 1 st differences according to all unit root tests. Must transform to 1 st differences.
SP	None	Not stationary in levels. Stationary at the 1% in 1 st differences	Not stationary in levels. Stationary at the 1% in 1 st differences	Not stationary in levels. Stationary at the 1% in 1 st differences	Not stationary in levels. Stationary at the 1% in 1 st differences	Stationary only in 1 st differences according to all unit root tests. Must transform to 1 st differences.
TOP1	Intercept and trend	Stationary at the 1% in levels.	Stationary at the 1% in levels.	Stationary at the 1% in levels.	Not stationary in levels. Stationary at the 1% in 1 st difference.	Mostly stationary in levels. Transform to 1 st differences for consistency with other variables.
WAGES	Intercept and trend	Not stationary in levels. Stationary at the 1% in 1 st differences.	Stationary at the 5% in levels. Stationary at the 1% in 1 st differences.	Stationary at the 5% in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences	Stationarity varies in levels and in 1 st differences. Transform to 1 st differences for consistency with other variables.
WELFARE	Intercept and trend	Stationary at the 1% in levels.	Stationary at the 10% in levels. Stationary at the 1% in 1 st differences.	Stationary at the 5% in levels. Stationary at the 1% in 1 st differences.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Stationarity varies in levels and in 1 st differences. Transform to 1 st differences for

						consistency with other variables.
AGE	None	Not stationary in levels. Stationary at the 1% in 1 st differences	Stationary at the 10% in levels Stationary at the 1% in 1 st differences	Not stationary in levels. Stationary at the 1% in 1 st differences	Stationary at the 10% in levels Stationary at the 1% in 1 st differences	Must transform to 1 st differences.
i	Intercept and trend	Stationary at the 1% in levels.	Stationary at the 1% in levels.	Stationary at the 1% in levels.	Not stationary in levels. Stationary at the 1% in 1 st differences.	Mostly stationary in levels. Transform to 1 st differences for consistency with other variables.