

The Financial Accelerator in Household Spending: Evidence from International Housing Markets*

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Abstract

This paper explores contractual features of housing finance and uses data from international housing markets to identify and test the “financial accelerator” (Bernanke et al., 1996). Among households whose housing demand is constrained by the availability of collateral, those who can borrow against a larger fraction of the housing value (achieve a higher loan-to-value, or LTV ratio) have more procyclical debt capacity. This procyclicality in borrowing capacity is at the heart of the mechanism underlying the financial accelerator. Our empirical strategy, which borrows from Lamont and Stein (1999), uses international variation in maximum LTV ratios to show that housing prices as well as demand for new mortgages are more sensitive to income shocks in countries with higher LTV ratios. These results are consistent with the dynamics of a collateral-based financial accelerator in household spending. We also find that the empirical relationship between LTV ratios and income sensitivities is stronger in countries where housing prices are low relative to household income. Because collateral constraints are less likely to bind when housing is more expensive (an income constraint may bind instead), these latter results further indicate that a collateral-based accelerator is indeed behind the observed cross-country differences in income sensitivities.

JEL classification: E32, E44, G15.

Key words: Financial accelerator, household spending, housing prices, collateral constraint, income constraint.

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I Introduction

Recent theoretical research proposes that endogenous developments in financial markets can greatly amplify and propagate small income or interest rate shocks throughout the economy (Kiyotaki and Moore, 1997; and Bernanke et al., 1996, 1999). Bernanke et al. (1996) call this amplification mechanism the “financial accelerator” or “credit multiplier.” The key idea behind the financial accelerator is the notion that shocks to the net worth of firms and households have a procyclical effect on their borrowing capacity. This could happen either because the information cost wedge between external and internal finance moves countercyclically (Bernanke and Gertler, 1989), or because a procyclical change in the value of collateralizable assets changes the amount of collateralized external finance in the same direction (Kiyotaki and Moore, 1997). Following a positive income shock, agents should be able to raise more external finance and the increase in borrowing capacity would further boost investment spending. According to this view, financial mechanisms such as the endogenous procyclicality of external financing capacity can help explain important features of the business cycle and the transmission of monetary policy.

There is little direct evidence on the amplification mechanism that underlies the financial accelerator. Most empirical studies use firm data to explore one insight behind the accelerator: income shocks should affect corporate spending only when firms have imperfect (constrained) access to external finance. Empirically, the investment spending of financially constrained firms should be more sensitive to changes in net worth than the investment spending of unconstrained firms (Fazzari et al., 1988).¹ In the same vein, constrained firms’ spending and borrowing should fluctuate relatively more in the aftermath of monetary and other macroeconomic shocks (Gertler and Gilchrist, 1993, 1994). Unfortunately, while comparisons between constrained and unconstrained firms may indicate whether one group’s spending is more dependent on current income following an economic shock, they will *not* identify whether differences in spending stem from an endogenous financial amplification mechanism: because constrained firms are more dependent on current income for investment funding, they should be more sensitive to a shock that affects income even when the shock has no cyclical effect on their borrowing capacity.

So how can one identify whether there is an independent spending effect coming from an endogenous change in borrowing capacity following an income shock? The theory suggests that the quantitative effect of an aggregate income shock on constrained agents’ spending would be greater

¹See Hubbard (1998) for a survey of the literature on financial constraints and investment-cash flow sensitivities.

when debt capacity is procyclical. Hence, one testing approach consists of directly quantifying the *overall magnitude* of the amplification effect for constrained firms with procyclical net worth; this is the spirit of the simulations in Bernanke et al. (1999).² An alternative approach is to gauge the degree of procyclicality in agents' borrowing capacity and then pin down the dynamics of the financial accelerator by looking at *cross-sectional differences* in the spending responses to economic shocks among *strictly constrained, cyclical* agents. We pursue such an approach in this paper.

The housing sector is an ideal laboratory for conducting our proposed test of the theory. As suggested by Bernanke et al. (1996), households fit models of the financial accelerator particularly well; especially collateral-based versions of the accelerator. The crucial feature of housing finance contracts we explore in this paper is that the availability of mortgage credit to households is typically limited to a specific proportion of the value of the home they own or are about to purchase (the maximum loan-to-value, or LTV ratio). The maximum LTV ratio is, in effect, a credit *quantity constraint*. To see how the financial accelerator works in the housing market, suppose households receive a positive income shock that boosts their demand for housing. Clearly, the higher the LTV ratio that households can achieve the higher the increase in their borrowing capacity. Housing values should thus respond more to the initial income shock when the LTV is high. The procyclical increase in the housing value will itself allow households to further increase borrowing and the collateral-based spending cycle gets amplified. The relationship between LTV ratios and the income sensitivity of housing demand therefore provides for a direct test of the endogenous mechanism underlying the financial accelerator: the impact of shocks to household income on housing spending is amplified by the higher marginal opportunity to borrow associated with a high loan-to-value ratio.

Testing this prediction requires some degree of (exogenous) variation in borrowing constraints (i.e., LTV ratios). Fortunately, because LTV ratios vary widely around the world, data from international housing markets can be used to test our accelerator story.³ To give a concrete example of what we have in mind, consider a country in which housing finance is not well-developed, such as Italy, where historical maximum LTV ratios do not exceed 60%. Take, on the other hand, a country such as the UK, where LTV ratios averaged 90% in the last two decades. The accelerator argument would suggest that so long as the collateral constraint is binding in both countries the housing

²Carlstrom and Fuerst (1997) also calibrate the Bernanke and Gertler (1989) model to directly gauge the quantitative effect of the accelerator.

³We are not the first to explore the contrasts provided by international housing markets to make inferences about aggregate economic behavior. Jappelli and Pagano (1989, 1994), for example, have used international differences in LTV ratios to study consumption and savings behavior.

credit multiplier should be much stronger in the UK than in Italy. Simply put, the collateral-based accelerator story we test predicts that because households in high LTV countries have more procyclical debt capacity, their demand for housing will be more sensitive to income shocks.

Of course, a direct relationship between maximum LTV ratios and housing price fluctuations should only hold if housing demand is effectively constrained by the availability of collateral. Indeed, it is possible that housing demand is limited by an alternative borrowing constraint: the *income constraint*. The income constraint stems from real-world features of mortgage contracts that limit the yearly amount of housing expenditures associated with the loan (loan payments plus property taxes) to a certain fraction of the household’s yearly income. For our purposes, the key difference between the collateral and the income constraints is that only the former gives rise to a credit multiplier.⁴ If the income constraint binds, a household’s marginal opportunity to borrow depends on its future income stream rather than on the current value of housing. The upshot of integrating both types of constraints on household spending in a financial accelerator model is the observation that whenever the income constraint binds the positive relationship between LTV ratios and the sensitivity of housing prices to income vanishes.⁵ Empirically, recognizing this additional constraint provides for yet another layer of contrasts for our panel data tests of the accelerator.

Disentangling the effects of the income and collateral constraints is not a trivial task. Our empirical strategy, however, explores well-known characteristics of international housing markets to identify situations in which the income constraint is more likely to bind in the first place. In particular, we conjecture that the income constraint is more likely to bind when the price of a typical housing unit is relatively high vis-à-vis household disposable income (high “price-to-income ratio”). Cross-country differences in price-to-income ratios — engendered, for example, by demographic and geographical factors — introduce variation in housing affordability. We predict that if the relationship between price sensitivities and the LTV ratio is driven by the collateral constraint, then it should be especially strong in countries with more affordable housing.

The evidence of this paper supports the financial accelerator. Our tests show that housing prices are indeed more sensitive to income shocks in countries with higher maximum LTV ratios.

⁴The idea that not all types of credit constraints have a natural amplification mechanism is also explored by Kocherlakota (2000). He argues that a credit constraint will only generate an amplification mechanism when the borrowing limit is directly associated with asset values (as in a maximum LTV-type constraint).

⁵Likewise, if households were to become financially unconstrained as a result of increases in maximum LTV ratios, then the amplification mechanism would die out. We discuss the extent to which this possibility is consistent with our data in section IV.B.

Our estimates indicate that in countries like the UK, where the LTV ratio is around 90%, housing prices decrease by more than 1.2% in the first year following a 1% decrease in per capita GDP. On the other hand, in countries such as Italy, where the LTV ratio is around 60%, housing prices decrease by only some 0.8% following a 1% decrease in per capita GDP. These results indicate that the credit multiplier has a considerably greater impact on household spending in countries where the maximum LTV is high. Consistent with our conjectures about the joint role of income and collateral constraints, we also find that the relationship between LTV ratios and income sensitivities is stronger in countries where housing is cheaper relative to household income.

Our empirical analysis explicitly recognizes a number of alternative factors that could potentially influence the results we obtain. For example, we control for variables which are likely to be correlated with maximum LTV ratios and which could also explain the cross-country differences in income sensitivities, such as economic development and the propensity for homeownership. In particular, we find that the effect of the LTV ratio remains after expunging the component that is explained by economic development and homeownership. At the same time, the relationship between the LTV ratio and income sensitivities remains after instrumenting the LTV ratio with variables that absorb cross-country differences in overall financial development, such as the development of the judicial system and the quality of accounting standards. Our evidence suggests that financial development is a contributing factor to the real-side effects of the financial accelerator. As a final robustness check, we avoid looking at housing price responses altogether, focusing instead on the demand for new mortgages. Although we only have limited data on mortgages, we again find evidence that is consistent with the multiplier: new mortgages respond positively to household income shocks, and those responses are increasing in LTV ratios.

Our paper is related to several different strands of literature. Our empirical approach borrows from Jappelli and Pagano (1989, 1994), who study the relationship between financial development and macroeconomic variables such as savings growth and the sensitivity of the consumption to changes in income. Jappelli and Pagano use maximum LTV ratios as a measure of financial constraints on households exactly as we do — higher maximum LTV ratios are associated with higher debt capacity and less financial constraints on households. They find evidence that maximum LTV ratios help explain cross-country differences in key macroeconomic variables, such as savings.

The sensitivity of housing prices to household income has been examined by Lamont and Stein (1999). Using data from the US, they find that housing prices are more sensitive to changes in

city-level GDP in years when homeowners in a particular city have very high leverage. They interpret these results as evidence that households are likely to be liquidity-constrained when their leverage is high, which is consistent with the idea that constrained agents are more sensitive to income shocks than unconstrained ones. Our analysis, in contrast, uses international variation in maximum LTV ratios and in price-to-income ratios to identify procyclicality in debt capacity among strictly constrained agents.⁶ While Lamont and Stein's goal is to examine the effects of liquidity constraints on asset pricing (with an application to housing), our paper aims at shedding some new light at the amplification mechanism implied by the financial accelerator.

There is also a large literature that uses micro-level data to test the effects of financial constraints on the housing market. This literature suggests that financial constraints help explain several housing variables, such as the propensity for homeownership (Linneman and Wachter, 1989; and Haurin et al., 1997), the type of mortgage chosen (Hendershott et al., 1997), mortgage refinancing and prepayments (Archer et al., 1996; and Caplin et al., 1997), and owner-occupants' selling behavior (Genesove and Mayer, 1997). Most evidence pertains to the US market, with several of the papers analyzing the joint effect of the income and collateral constraints.

There are very few international studies on housing markets, with most of them looking at small sets of developed countries (e.g., Cutler et al., 1991). Englund and Ioannides (1997) characterize housing price dynamics for a panel of 15 OECD countries, but do not look at financial constraints. Malpezzi (1990) discusses some interactions between financial development and housing markets, but does not focus on housing price dynamics. The closest to our paper is the literature that looks at episodes of financial market liberalization in some European countries (Stephens, 1995; MacLennan et al., 1998; and Iacoviello and Minetti, 2003). Consistent with the argument that we develop in this paper, that literature suggests that financial market liberalization have contributed to increases in housing price volatility and to the sensitivity of housing prices to interest rates.

The role of financial constraints in housing markets has also been studied by a few theoretical papers. Stein (1995) analyzes the impact of a down payment constraint on the equilibrium of the housing market and finds that the sensitivity of prices to shocks to fundamentals (such as income) is higher in the constrained equilibrium. Ortalo-Magne and Rady (1998, 1999) consider the effect of an interaction between household heterogeneity and a collateral-type constraint for housing price

⁶Notice also that the key housing finance variable we use (the maximum LTV ratio) is conceptually different from household's existing leverage. The maximum LTV ratio represents the marginal opportunity to borrow as a function of the value of housing, while household's leverage is an endogenous variable determined by past borrowing decisions.

fluctuations. Their model features an amplification mechanism that relates to the one we emphasize in this paper: an income shock gets amplified through its effect on the ability of young households to afford down payments.⁷ Neither of those papers analyzes the effect of changes in the down payment requirement (the LTV ratio), nor the independent effect of the income constraint. Even though our focus is empirical, this gap motivates us to start the paper by laying out a very simple model where we introduce these considerations. We do this in the next section (Section II). In Section III we provide a detailed description of the international housing markets data we use in the study. In Section IV we present our empirical results. Section V concludes the paper.

II Collateral and income constraints in housing demand

We formalize our arguments about (constrained) housing demand using a simple model framework; based on Stein (1995). There are two goods in the economy, housing (H) and food (Z). The price of housing (P) is measured in units of food. There is a representative household endowed with total lifetime income equal to $W_1 + W_2$. There are two time periods in the model. The household only consumes in the final period (t_2), but it must choose at time t_1 how much to spend on housing. In contrast, food is purchased at the time it is consumed.⁸ We assume throughout that the gross, riskless rate of interest in the economy is equal to 1.

At time t_1 , the household only has access to current income W_1 . It cannot borrow directly against future income W_2 because future income cannot be pledged to creditors, and thus the household might be constrained in its choice of housing. The household can raise mortgage debt against the value of its housing wealth.⁹ The value of the mortgage loan (call it B_1) that can be raised is subject to two constraints. First, the loan cannot be higher than a certain fraction $\lambda \in [0, 1]$ of the value of the home, that is

$$B_1 \leq \lambda PH. \tag{1}$$

The parameter λ can be interpreted as the maximum loan-to-value ratio. The higher the λ , the easier it is for a household to borrow in order to finance spending. In the real-world, this parameter

⁷One of their main implications is that since housing prices depend particularly on the income of young households, the income of young households might be a more appropriate aggregate variable to include in housing price regressions. Unfortunately, the international data we have do not allow us to account for such heterogeneity in our tests.

⁸These assumptions eliminate intertemporal effects other than the fact that the household must purchase housing before consuming it fully.

⁹We are not explicitly considering the role of inherited leverage from past mortgages and other borrowings. However, existing leverage can be thought of as a reduction in the income that is available to finance the housing down payment, W_1 .

depends on variables such as the costs of enforcing and disposing of collateral, regulations about housing finance, and the amount of information creditors have about borrowers.¹⁰ The fact that the λ can be lower than 1 represents in effect a credit quantity constraint on households. An alternative approach would be to focus on the relative cost of funds, or more specifically, on the wedge between the borrowing rate in the mortgage market and an appropriate lending rate. As discussed by Jappelli and Pagano (1989, 1994), however, this wedge does not appear to be a viable explanation of the cross-country differences in the financial liabilities of households. Differences in interest rate wedges across countries seem negligible and there is no clear relation between lending volumes and wedges. For simplicity, we thus assume that the household pays no interest rate premium when it borrows up to the limit λPH . We call Eq. (1) the household’s “collateral constraint.”

The other constraint faced by households in real-world mortgage contracts is the “income constraint.” The income constraint essentially limits the yearly amount of housing expenditures associated with the mortgage contract (loan payments plus property taxes and insurance) to a certain fraction of the household’s yearly income, which in the US is around 28%.¹¹ Stein’s model does not incorporate the idea of an income constraint. In order to accommodate that constraint in the present model — which also lacks an explicit intertemporal component — we assume that the total value that must be repaid to creditors at t_2 (that is, B_1) must be lower than a certain fraction $k \in [0, 1]$ of the household’s future income W_2 , plus any amount saved from t_1 to t_2 (call it s_1):

$$B_1 \leq k(W_2 + s_1). \quad (2)$$

For $\alpha \in (0, 1)$, we assume that the household’s utility function is given by (Stein, 1995):

$$U = \alpha \ln(H) + (1 - \alpha) \ln(Z). \quad (3)$$

A Analysis

The household whose borrowing is not constrained will maximize the utility function above (Eq. (3)) subject to a lifetime budget constraint:

$$PH + Z = W_1 + W_2. \quad (4)$$

The optimal unconstrained housing demand (as a function of housing price) is then given by

$$H^U(P) = \alpha \frac{W_1 + W_2}{P}. \quad (5)$$

¹⁰See Jappelli and Pagano (1994) for a detailed discussion. Spiegel (2001) endogenizes down payment requirements, and argues that LTV ratios can be used to forecast future housing returns.

¹¹Unfortunately, we do not have data on the income limits for other countries.

The household will be constrained when the maximum amount that it can borrow is not enough to finance the unconstrained demand $H^U(P)$. Let B_1^{\max} be defined by¹²

$$B_1^{\max} = \min[\lambda PH^U(P), kW_2]. \quad (6)$$

The household will be constrained when

$$PH^U(P) > W_1 + B_1^{\max}. \quad (7)$$

In this case, the optimal housing demand is determined directly from the constraints.¹³ There are two possibilities to consider, depending on which constraint (collateral or income) is binding:

i) If the income constraint is binding, the maximum housing demand that the household can finance is given by

$$H^I(P) = \frac{W_1 + kW_2}{P}. \quad (8)$$

ii) If the collateral constraint is binding, then we have

$$H^C(P) = \frac{W_1 + \lambda PH^C(P)}{P}. \quad (9)$$

In either of these cases, the household consumes the rest of its lifetime income in food, that is, $Z = W_1 + W_2 - PH$.

Eqs. (8) and (9) reveal the key difference between the collateral and the income constraints and clarify the role these constraints play in the financial accelerator. The collateral constraint is endogenous: the household's capacity for external finance depends on the value of housing. Hence a shock to current income W_1 that changes housing demand will be amplified by the endogeneity of debt (as in Kiyotaki and Moore, 1997). When the income constraint binds, on the other hand, debt capacity only depends on future income and there is no natural amplification mechanism.

Considering the effect of a change in current income on housing demand (while taking the housing price as fixed) it is easy to see that

$$\frac{\partial H^U}{\partial W_1} = \frac{\alpha}{P} \leq \frac{\partial H^I}{\partial W_1} = \frac{1}{P} \leq \frac{\partial H^C}{\partial W_1} = \frac{1}{(1-\lambda)P}. \quad (10)$$

¹²Notice that if the constraint in Eq. (6) is binding, then the household cannot bring enough income from the future to the present in order to finance its optimal housing expenditures, which implies $s_1 = 0$. This is true even when the income constraint is binding, since $k \leq 1$.

¹³Notice that if W_1 is low the household is more likely to be constrained. If we think of past leverage as a reduction in W_1 , this effect is consistent with the main hypothesis tested in Lamont and Stein (1999): when leverage is high households are more likely to be constrained and the effect of an income shock on housing prices is larger than in the benchmark unconstrained case.

The reason for the first inequality is that the constrained household spends a greater fraction of current income on housing. In terms of the model, the optimal fraction to spend on housing is given by the parameter α . The constrained household is underinvesting in housing, and thus will direct the entire change in current income to housing.¹⁴ The second inequality shows that the sensitivity of demand to income will be even higher when the collateral constraint binds, because of the amplification effect associated with the endogenous change in debt capacity.

B Empirical implications

Our simple analysis suggests a number of interesting implications. First, notice that the sensitivity of demand to income will depend on the loan-to-value ratio *if and only if* the collateral constraint binds. So long as the household is collateral-constrained, an increase in the loan-to-value ratio will lead to an increase in the sensitivity of demand to income. Thus, the *less* constrained the household is, the *higher* the sensitivity of demand to current income will be. If the LTV ratio is so high that the borrowing constraint no longer binds, then the sensitivity disappears either because the household becomes unconstrained or because the income constraint will bind instead. To see this effect, notice from Eq. (6) that the necessary condition for the income constraint to bind is

$$\frac{k}{\lambda} < \frac{PH}{W_2}. \quad (11)$$

This condition indicates that the income constraint will bind when the value of housing is high relative to household income. By Eq. (8), when the income constraint binds the relationship between constrained housing demand and maximum LTV ratios vanishes.

One difficulty that we face when taking this model to the data is that changes in housing demand for a particular country are not directly observable. In particular, the available international data consist primarily of housing price indices for different countries. One would need implications for the housing market equilibrium, and in particular for the sensitivity of housing prices to current income, in order to utilize those data. In our case, we need to assume some degree of rigidity in housing supply in order for our model implications to carry to price data. To see this in the simplest possible way, suppose that housing supply is perfectly inelastic. Then we can show that

$$\frac{\partial P^U}{\partial W_1} = \alpha \leq \frac{\partial P^I}{\partial W_1} = 1 \leq \frac{\partial P^C}{\partial W_1} = \frac{1}{1 - \lambda}, \quad (12)$$

¹⁴If there was intermediate consumption in the model, then the constrained fraction spent on housing would also be higher than the unconstrained one, but it would no longer be optimal to spend all income in housing. The gist of our results, however, would be the same.

where P^U is the unconstrained price level, and P^I and P^C are, respectively, the price levels when the income or the collateral constraint binds. Clearly, the constrained sensitivities are higher than the unconstrained one, and the sensitivity is highest when the collateral constraint binds. Furthermore, if the collateral constraint binds the sensitivity is increasing in the loan-to-value ratio. If one can assume that housing supply is sufficiently inelastic in the short-run, then the properties we derived for housing demand should translate into housing prices. This particular supply-rigidity assumption is standard in the housing literature (see, e.g., Meen, 1996; Voith, 1996; Malpezzi and Mayo, 1997; and Mayer and Sommerville, 2000) and it is well-supported in the limited data we have on housing starts.¹⁵ Should housing supply be very elastic, on the other hand, then we should fail to uncover any evidence of the accelerator.

Another issue for empirical our tests is that when we compute the effect of changes in W_1 we are effectively assuming that there is little correlation between changes in current income and changes in future income (W_2). If there is a strong correlation between W_1 and W_2 , the comparison between the sensitivities across cases (for example, constrained versus unconstrained) might not be as clean as what we had above. In order to see this, suppose that

$$W_2 = \widetilde{W}_2 + \rho W_1, \quad (13)$$

where ρ is a measure of the correlation between W_1 and W_2 . It is easy to see that the sensitivities of price to current income will become

$$\frac{\partial P^U}{\partial W_1} = \alpha(1 + \rho), \quad \frac{\partial P^I}{\partial W_1} = (1 + k\rho), \quad \text{and} \quad \frac{\partial P^C}{\partial W_1} = \frac{1}{1 - \lambda}. \quad (14)$$

It is no longer clear that the sensitivity is highest when the collateral constraint is binding. It could also be the case that the unconstrained sensitivity is higher than the constrained ones. Yet, the implication that sensitivity increases in the loan-to-value ratio if and only if the collateral constraint is binding remains unchanged. Note that, this is the only theoretical result we need to support our claims about testing for financial accelerator effects *within* a set of *financially constrained* agents.

Finally, as in most papers dealing with measures of financial development, we need to ensure that our findings are not simply driven by economic development. In our case, it is possible that the sensitivity of prices to income increases with economic development even when households are unconstrained. This could happen if the fraction of income spent on housing, α , increases with

¹⁵Regressing a measure of new dwellings (and, alternatively, a measure of the change in housing stock) on various lags of GDP, we find no evidence of a significant response of housing starts to current and recent lags of income.

economic development. Since economic and financial development are related, it could be difficult to disentangle this story from our explanation based on collateral constraints. In the empirical section, we shall verify that our results cannot be ascribed to economic development alone.

We can summarize the testable implications of our model as follows:

1. *If the collateral constraint is binding, then the sensitivity of housing prices to shocks to current income should be increasing in the maximum LTV ratio available to households.*
2. *If the relationship between price-income sensitivities and the LTV ratio is driven by the collateral constraint, then it should be especially strong in countries where the income constraint is less likely to bind.*

We empirically test both of these predictions in turn.

III Data description

We use data on housing price indices for the 26 countries listed in Table 2 for the period 1970-1999. The housing price data are summarized in Table 1 together with the data on per capita GDP (the main driving variable in the empirical specification) and annual new mortgages (which we use as an alternative endogenous variable). We use yearly changes in the logs of GDP and housing prices, deflating the data with consumer price index series taken from the IMF's *International Finance Statistics* database. New mortgages are expressed as a fraction of nominal GDP. The data on housing prices and new mortgages are hand-collected from a number of different sources, while the GDP data are taken from the IMF financial statistics. We list all of our data sources as well as provide detailed information about the different indices used in the Appendix.¹⁶ Our sample has 754 country-year observations.

– insert Table 1 here –

Table 2 displays country-level data on maximum LTV, price-income, and homeownership ratios. Each of the series is relevant to our analysis. The maximum LTV ratio is the empirical counterpart of the parameter λ in the model of Section II. Most of the LTV data are taken from Jappelli and Pagano (1989, 1994), who also use the maximum LTV ratio as a measure of the availability of credit to households in an international context. Those authors argue that the maximum LTV ratio

¹⁶The data used in this paper are available from the authors upon request.

is a direct and objective measure of liquidity constraints on households that is comparable across countries. We were able to augment the Jappelli and Pagano dataset using data from Chiuri and Jappelli (2000), and by looking into the sources cited therein. We also use various other sources for LTV ratios (see the Appendix). Table 2 shows that maximum LTV ratios vary significantly around the world. Developing countries, such as Korea and Taiwan, generally have lower LTV ratios (as low as 30%). However, there is variability even among developed economies, as evidenced by the case of Italy, where LTV is 60% during the 1990's, versus 95% for the UK in that same period. Maximum LTV ratios vary less over time within a country, with a few exceptions, such as Sweden and Spain.

– insert Table 2 here –

Our empirical analysis uses homeownership to help isolate the effects of economic development in the estimations performed. The homeownership ratio represents the proportion of home owners as a fraction of total households in a country. Country-level homeownership data were hand-collected from several sources. Table 2 suggests that these country series remain stable over time.

In Section II we argue that the effect of the maximum LTV ratio on income sensitivities should only be significant when the collateral constraint is binding. Of course, it is also possible that the income constraint binds, which could dampen the positive association between LTVs and income sensitivities. Identifying which of those constraints will bind first in each of the countries studied is not an easy task. Arguably, though, the income constraint is more likely to bind in countries where the value of a typical home is high when compared to household disposable income. Empirically, if housing prices vary across countries due to factors such as geography and demography, then we should expect the income constraint to be more likely to bind in countries where housing is relatively expensive.¹⁷ In this vein, we use the ratio of the price of a typical dwelling unit to yearly median household disposable income (price-income ratio) in order to gauge the likelihood that the income constraint binds in a given country. Intuitively, this measure represents the number of years it takes for the median household to “earn” the value of a typical home.

Table 2 shows that there is large cross-country variability in the ratio of the value of a typical housing unit to disposable income. In countries such as Switzerland and Singapore, typical housing units are substantially more “expensive” than in other countries, such as the US and Canada. These

¹⁷This condition is intuitive. In some countries, housing can be so expensive in comparison to household income that it does not matter whether the maximum loan-to-value a household can achieve is 10% or 90%.

differences seem to be driven by factors such as country geography and size as well as population growth and density; and they seem to capture information about the relative affordability of housing across different countries.

IV Empirical tests

Our main goal is to examine the empirical relationship between the sensitivity of housing prices to income shocks and maximum LTV ratios across countries. According to the financial accelerator hypothesis, that sensitivity should be especially strong if the maximum LTV ratio is high, because of the endogenous change in debt capacity following a positive shock to income. Since the collateral constraint is more likely to bind in countries with more affordable housing, the relationship between LTV ratios and income sensitivities should be stronger in countries with relatively cheap housing (i.e., low price-income ratios). Finding that these patterns are present in the data is consistent with evidence in favor of the financial accelerator in housing spending.

A Housing price dynamics

In order to test our hypotheses we need a benchmark empirical model of housing price dynamics. Following Lamont and Stein (1999) and other papers in the housing literature, we use the log change in the housing price index as the endogenous regressor in our housing pricing model. The literature also suggests a set of determinants (other than current income) to include in that model. For instance, there is ample evidence of a consistent autoregressive pattern in housing prices. There is positive autocorrelation at short lags (Case and Shiller, 1989; Poterba, 1991; and Lamont and Stein, 1999), but negative serial correlation at longer lags (Case and Shiller, 1990; and Lamont and Stein, 1999). This pattern has been shown to hold in international data as well (Englund and Ioannides, 1997). We experiment with the use of these lag structures in turn.

In Table 3 we pool the sample in a panel regression and search for an appropriate empirical model to fit the data on housing prices. All regressions include year effects. Column (1) shows that real housing prices are indeed correlated with real current income (proxied by real per capita GDP). Two additional lags of per capita GDP are also significant when no other variables are included in the regression, as shown in column (2). Columns (3) and (4) show that there is positive price autocorrelation at short lags, but negative autocorrelation at longer lags (long-term reversal). This

is true both with and without the inclusion of country effects.¹⁸ Our international data shows most of the same patterns of previous studies focusing on the US housing market.

– insert Table 3 here –

Column (5) adds other macroeconomic variables to the model of column (3). Both the real interest rate and the inflation rate have negative effects on housing prices, but their effects are not always significant. Finally, in columns (6) and (7) we use the empirical model proposed by Lamont and Stein (1999) in their study of housing price dynamics in US cities. Essentially, they replace longer lags of price and income changes with the start-of-period ratio of price to per capita income ($Price_{t-1}/Income_{t-1}$). As in Lamont and Stein, column (6) shows that longer lags of price and income become insignificant when we include the lagged ratio of price to per capita income. The more parsimonious specification of column (7) seems to capture well the effects of the longer lags.

In the next section we introduce the LTV ratio and the income constraint in the analysis, using, alternatively, the specifications in columns (1), (3), (4), (5), and (7) of Table 3. This verifies that our findings do not hinge on the selection of a particular specification for housing price dynamics.

B Financial constraints and the income sensitivity of housing prices

We introduce credit constraint effects in our analysis by allowing the price-sensitivity of income to vary according to the maximum LTV ratio. This amounts to augmenting our baseline empirical models by adding an intercept term for the LTV ratio and another term capturing the interaction between LTV and per capita GDP growth. When we use lags of GDP growth, we interact the LTV ratio with all of the lags of GDP change, besides the current change (lag 0). This approach will capture the effect of the accelerator even if it takes longer for it to feed through the economy. We then test whether an increase in LTV increases sensitivities by testing whether the parameters on those interaction terms are significantly greater than zero.

Table 4 presents one of the main set of results of the paper. Column (1) shows that the correlation between changes in housing prices and changes in household income is indeed higher in countries with higher maximum LTVs. The positive effect of the LTV ratio remains after we include further lags of price and income in the specification, as shown in column (2). The sum of the interaction terms of the LTV with the current and past lags of the change in income is positive and

¹⁸Following the standard approach in the literature, most of our models are estimated via OLS and include both lagged dependent variables and fixed effects (see, e.g., Lamont and Stein, 1999). We, however, recognize the potential for biases in this procedure, and later emphasize results from the Arellano-Bond dynamic panel GMM estimator.

significant at the 1% level. When we include country effects in the model the sum of the interaction terms increases (see column (3)). Column (4) shows that the inclusion of inflation and interest rates in the specification reduces the effect of the LTV ratio, but the sum of the interaction terms is still positive and significant. Finally, the interaction of the LTV ratio with the current change in income is also significant (at the 10% level) when we use the Lamont and Stein specification. This last specification makes it convenient to assess the implied magnitude of the effect of the LTV ratio on income sensitivities. The coefficient returned for $\Delta \text{Log}(\text{Income})_t \times \text{LTV}_t$ suggests that if the LTV goes from 0.60 to 0.90, the income sensitivity increases nearly 50%, from 0.84 to 1.23. These estimates imply, for example, that a 2% drop in per capita GDP will depress housing prices by some 1% *more* in the UK than in Italy.

— insert Table 4 here —

Table 5 reports the results we obtain after imposing several modifications to our basic empirical models. For brevity, we use the specification with three lags of income and prices (columns (2) and (3) of Table 4) as a benchmark.¹⁹

— insert Table 5 here —

Our interpretation of the positive correlation between the LTV ratio and the income sensitivity of housing prices is that this effect is driven by differences in the availability of mortgage finance to households in different countries. To provide further evidence that our results are indeed driven by differences in financial constraints (as opposed to some trivial simultaneity story), we instrument the LTV ratio with variables which we expect to be related to the overall level of financial development in different countries. In countries with higher financial development it should be easier for both firms and households to raise outside finance. In the context of mortgage finance, a higher level of financial development should be reflected in the availability of higher LTV ratios for households.²⁰ Our set of instruments includes the index of accounting standards computed by the Center for International Financial Analysis and Research. Accounting standards have been used as an instrument for financial development in Rajan and Zingales (1998), among others. The second variable included in our instrument set is a proxy for the effectiveness of the country's judicial system. This proxy is

¹⁹Our conclusions are similar when we choose other specifications featured in Table 4 for this new series of tests.

²⁰Notice that *financial* development is in principle distinct from overall *economic* development; even though they are correlated.

taken from LaPorta et al. (1998). The idea is that, the higher the standards of financial disclosure and the more advanced the judicial system in a country, the easier it is for firms to raise funds from a wider circle of investors. Assuming that similar variables influence the availability of finance to firms and households, accounting standards and judicial efficiency are appropriate instruments for the maximum LTV ratio.²¹ The results on the first two columns of Table 5 show that the effect of the LTV ratio on income sensitivities actually increases after instrumenting for overall financial development. This is true both with and without the inclusion of country effects. These results suggest that our previous findings are indeed driven by variables affecting the availability of finance.

To the extent that maximum LTVs and economic development might be correlated, one could argue that the results in Table 4 are not primarily driven by financial development, but simply by cross-country differences in *economic* development. The theoretical section suggests that if the fraction of wealth spent in housing increases with wealth, then it could be the case that richer countries have larger income sensitivities, even if financial constraints are never binding. This provides for an “unconstrained explanation” for the observed pattern in sensitivities. Another possible explanation for our results is that the relationship between maximum LTV ratios and income sensitivities is driven by cross-country differences in homeownership. One could argue, for example, that countries with large rental markets have lower sensitivities and lower LTV ratios because the rental market helps absorb the effect of an income shock, or because only the wealthiest households own homes in countries with low LTV ratios. In particular, in economies with high LTV ratios the demand for housing could be more cyclical simply because the marginal borrower in these economies is poorer, and thus more sensitive to current economic conditions. If this argument explains our results, then the cross-country differences in income sensitivities we observe should be absorbed by variations in the homeownership ratio.

In columns (3) through (6) we address the relevance of these competing stories by adding proxies for economic development (ten-year average values of per capita GDP in constant international prices)²² and homeownership to our specification. In columns (3) and (4) we add the economic development proxy together with all of its interactions with lags of log income change (lags 0 through 2).²³ In columns (5) and (6) a similar approach is used to control for homeownership.²⁴ The results

²¹The first-stage regressions indeed show that our instruments and the maximum LTV are strongly positively correlated. The R^2 of the first-stage regression is 0.39.

²²The averaging is intended to match the frequency of the LTV series.

²³The coefficients returned for these controls are mostly insignificant and are thus omitted from Table 5.

²⁴Results are similar if we use both of these variables and all of their interaction terms together in one specification.

from these tests suggest that neither economic development nor homeownership are robustly related to income sensitivities after controlling for the LTV ratio. The sum of the interactions with economic development are positive as expected, but never significant. More importantly, the positive effect of LTV on sensitivities remains mostly unchanged after controlling for homeownership and economic development. The sum of the interaction terms of the income changes with the LTV ratio is positive and significant at better than 5% test level in 3 of the 4 added specifications; in the remaining specification (column (4)) the sum of the interaction terms is only marginally significant (p -value of 11.9%).

In column (7) we estimate our baseline model using the GMM estimator for dynamic panel data proposed by Arellano and Bond (1991). More precisely, we implement the one-step Arellano-Bond estimator with each of the base model variables instrumented by two of their own lags (in levels). The Arellano-Bond estimator returns coefficients that are smaller than those from the OLS regression. Yet, the effect of the maximum LTV ratio on income sensitivities is still positive and statistically significant.²⁵ Recognizing the limitation of our sample size, we provide for a direct check of the argument that our results could be driven by observations from one particular country. We do this check by performing a series of GMM estimations of our baseline model in which we disregard data from one of the sample countries at each run. The lowest point estimate we obtain for $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j} \times \text{LTV}_t$ equals 0.56, which is returned when we exclude Japan from the sample. That estimate is statistically significant at the 3% level. Eliminating any of the other sample countries returns coefficients that are significant at better than the 1% level.

Our model suggests a specific economic mechanism behind the relationship between financial constraints and price-income sensitivities. If the household is financially constrained, the effect of a change in income is amplified by the associated increase in borrowing capacity; and this amplification effect is larger the higher is the LTV ratio. If this argument is correct, the income sensitivity of new *borrowings* by households should also be higher in countries with high LTV ratios. In columns (8) and (9) of Table 5 we use total annual new mortgages as a fraction of GDP as an alternative dependent variable in the empirical model. In the absence of priors for the dynamics of

The same holds under a more parsimonious approach where we only use the LTV ratio and its interactions with income change in the specification after expunging economic development and homeownership main effects from LTVs (i.e., using a “residual LTV”).

²⁵The Sargan test statistic associated with the Arellano-Bond estimator of Table 5 ($\chi^2_{(403)}=392.1$, p -value=64.2%) reveals that the null of instrument validity cannot be rejected. Also supporting the adequacy of the estimator is the high p -value (=83.5%) associated with the test of the null of no second-order autocorrelation in the residuals.

new mortgages, we use a more parsimonious specification which includes two lags of income changes besides the current change. Even though the sample is considerably smaller, we find evidence that new mortgages respond more to changes in income when the LTV is high. The interaction between income and the LTV ratio is positive and statistically significant when we do not add country dummies.

It might be useful to illustrate the differences in responses of new mortgages to income shocks across high and low LTV countries using a structural estimation approach. We do this by estimating a four-equation, two-lag VAR system that includes the change in log per capita GDP, the change in log CPI, the change in log interest rates, and the log of new mortgages (orthogonalized in this order). Unfortunately, only one low LTV country (which is Italy) has enough time series data on mortgages. We thus graph the impulse-response functions of new mortgages to one-standard-deviation shocks to GDP for Italy and the US. Comparisons between the two panels in Figure 1 make it clear that the demand for housing financing increases far more strongly in the US than it does in Italy following a similar positive shock to per capita GDP. Evidence from new mortgages, too, agrees with the accelerator mechanism underlying the dynamics of the relationship between income shocks and constrained borrowing in housing markets.

Finally, one might wonder whether households in countries with high maximum LTV ratios such as the UK are effectively financially unconstrained. If households become unconstrained, the amplification mechanism associated with the financial accelerator will die out. More weakly, in countries with high maximum LTV ratios the fraction of people whose LTV ratio is close to the maximum might decrease, thereby decreasing the impact of an income shock on the housing market. The evidence in Tables 4 and 5 is inconsistent with this hypothesis.²⁶ However, we can also perform a more direct check of this challenge to our identification strategy. One measure of the ratio of households that are subject to binding collateral constraints is the ratio between the average and the maximum LTVs. We have some limited data on total outstanding mortgages that we can use to estimate average LTV ratios.²⁷ If higher maximum LTVs increase the fraction of households that are unconstrained, then we would expect the ratio of average-to-maximum LTV

²⁶Our findings are consistent with previous research on developed housing markets where maximum LTV ratios are high. This research has found evidence of direct effects of financial constraints on the aggregate housing market of developed countries such as the US (see, e.g., Linneman and Wachter, 1989 and Lamont and Stein, 1999).

²⁷Our estimate of the average LTV ratio is constructed as the ratio of mortgage debt outstanding to the value of owner occupied housing, with the latter computed as the stock of housing times the homeownership ratio times the housing price level.

to be *negatively* related to maximum LTVs. As it turns out, we find that countries with *higher* maximum LTVs have *higher* ratios of average-to-maximum LTVs.²⁸ This finding is inconsistent with the hypothesis that higher maximum LTVs allow a larger fraction of households to become financially unconstrained.

C The income constraint

Our hypothesis is that if the relationship between price-income sensitivities and the maximum LTV ratio is driven by the collateral constraint then it should be especially strong in countries where the income constraint is less likely to bind. The ratio of the value of a typical housing unit to disposable income is a natural proxy for the income constraint. In countries where housing is more expensive, it is more likely that the income constraint will bind. If our hypothesis is correct, the relationship between sensitivities and maximum LTV ratios should be driven by countries with inexpensive housing.

In the final set of tests of the paper, we rank countries according to the distribution of the price-disposable income ratio and classify as “expensive” (“cheap”) those in the top (bottom) third of this ranking.²⁹ We then run separate regressions for the two subsamples. To demonstrate the robustness of our findings, we report outputs from OLS and GMM estimations of the model with three lags of income and prices as well as the results pertaining to the Lamont and Stein (1999) specification.

The results from the subsample regressions are reported in Table 6. Consistent with our predictions, there is a positive association between the LTV ratio and income sensitivities, but this relationship is only significant in countries with relatively cheaper housing. Such pattern holds steady across the various estimation procedures and empirical specifications we examine.³⁰ We interpret this last set of results as further evidence that increases in the maximum LTV ratio increase the sensitivity of housing prices to income because the financial accelerator is stronger when the

²⁸The correlation between the ratio of average-to-maximum LTV and the maximum LTV is 0.29. Furthermore, a regression of this ratio on maximum LTVs (plus time dummies) returns a positive and significant coefficient for maximum LTVs.

²⁹Our main conclusions are insensitive to whether we partition the data according to the median income-price ratio or, alternatively, according to quartiles. As should be expected, the latter partition produces stronger but noisier coefficients.

³⁰Noteworthy, the ratio of the average and the maximum LTV ratios is considerably lower in countries with more expensive housing (results available upon request). This is further evidence that in countries with expensive housing households borrow less than the limit allowed by the maximum LTV because the income constraint binds.

LTV is higher and households are collateral-constrained.

– insert Table 6 here –

V Concluding Remarks

In this paper we use the specific features that characterize housing finance contracts and international housing markets to provide fresh evidence supporting the “financial accelerator” introduced by Bernanke et al. (1996). Specifically, we use international variation in maximum loan-to-value (LTV) ratios to identify within a group of constrained agents those with more procyclical borrowing capacity. Since the procyclicality in the borrowing capacity of constrained agents is the amplification mechanism at the heart of the financial accelerator, our empirical strategy allows us to provide a direct test of the endogenous mechanism that underlies the accelerator. Our results show that housing prices are more sensitive to aggregate income shocks in countries with higher maximum LTV ratios, indicating that debt capacity is more strongly procyclical in such countries. Furthermore, the empirical relationship between LTV ratios and income sensitivities is stronger in countries where housing is cheaper relative to household income. Because the collateral constraint is more likely to bind in such countries, this result is consistent with the idea that a collateral-based financial accelerator is behind the cross-country differences we observe in income sensitivities. Our empirical analysis explicitly addresses a number of factors that could potentially influence the results we obtain.

Besides being a nice laboratory to study the economic effects of the financial accelerator, the housing market is also one of the markets where the significance of such effects is likely to be high. Previous literature has shown that consumer spending is intimately linked to housing wealth (see, e.g., Case et al., 2001; and Engelhardt, 1996), and that housing investment plays a major role in the business cycle (Mishkin, 1977, 1978; and Bernanke and Gertler, 1995). This paper shows that the effect of the financial accelerator in household spending and housing prices may help characterize the mechanism through which what seem to be small, localized shocks get amplified and transmitted throughout the economy.

Finally, the results in this paper may have interesting implications for the welfare effects of financial development. Previous research has identified excessive volatility in housing prices (Poterba, 1991), and has argued that, within OECD countries, those with more liberal financial markets experienced undesirably high levels of housing price volatility during the 1980’s and 1990’s

(Stephens, 1995). Our results suggest a mechanism through which financial development and liberalization could magnify fluctuations in housing prices. When financial development is associated with higher maximum LTV ratios collateral constraints are relaxed and the financial accelerator becomes stronger. Whether the financial accelerator and other theories stemming from financial imperfections can account for the excess volatility of housing prices is an important matter for public policy and for future research.

References

- Arellano, Manuel and Stephen Bond** “Some Tests of Specification for Panel Data: Monte Carlo Evidence and an Application to Employment Equations.” *Review of Economic Studies* 58(2), April 1991, pages 277-97.
- Archer, Wayne R., David C. Ling and Gary A. McGill** “The Effect of Income and Collateral Constraints on Residential Mortgage Terminations.” *Regional Science and Urban Economics*, 26(3-4), June 1996, pages 235-61.
- Barro, Robert and Jong-Wha Lee** “Sources of Economic Growth.” *Carnegie-Rochester Conference Series on Public Policy* 40, June 1994, 1-46.
- Bernanke, Ben and Mark Gertler** “Agency Costs, Net Worth, and Business Fluctuations.” *American Economic Review* 79(1), March 1989, pages 14-31.
- Bernanke, Ben and Mark Gertler** “Inside the Black Box: The Credit Channel of Monetary Policy Transmission.” *Journal of Economic Perspectives* 9(4), Fall 1995, pages 27-48.
- Bernanke, Ben, Mark Gertler and Stephen Gilchrist** “The Financial Accelerator and the Flight to Quality.” *Review of Economics and Statistics* 48, February 1996, pages 1-15.
- Bernanke, Ben, Mark Gertler and Stephen Gilchrist** “The Financial Accelerator in a Quantitative Business Cycle Framework,” in Taylor, John B. and Michael Woodford, eds: *Handbook of macroeconomics. Volume 1C. Handbooks in Economics, vol. 15.* Amsterdam, New York and Oxford: Elsevier Science, North-Holland, 1999, pages 1341-93.
- Carlstrom, Charles and Timothy Fuerst** “Agency Costs, Net Worth and Business Fluctuations: A Computable General Equilibrium Analysis.” *American Economic Review* 87, December 1997, pages 893-910.
- Caplin, Andrew, Charles Freeman and Joseph Tracy** “Collateral Damage: Refinancing Constraints and Regional Recessions.” *Journal of Money, Credit, and Banking* 29(4), Part 1, November 1997, pages 496-516.
- Case, Karl and Robert Shiller** “The Efficiency of the Market for Single-Family Homes.” *American Economic Review* 79, March 1989, pages 125-37.
- Case, Karl E., John M. Quigley and Robert J. Shiller** “Comparing Wealth Effects: The Stock Market Versus the Housing Market.” *NBER Working Paper* No.w8606, issued in November 2001.
- Chiuri, Maria C. and Tullio Jappelli** “Financial Market Imperfections and Homeownership: A Comparative Study.” *mimeo*, 2000.
- Cutler, David, James Poterba and Lawrence Summers** “Speculative Dynamics”, *Review of Economic Studies* 58, May 1991, pages 529-46.
- Engelhardt, Gary V.** “Consumption, Down Payments, and Liquidity Constraints” *Journal of Money, Credit, and Banking* 28(2), May 1996, pages 255-71.

- Englund, Peter and Yannis Ioannides** “Housing Price Dynamics: An International Empirical Perspective.” *Journal of Housing Economics* 6, June 1997, pages 119-36.
- Fazzari, Steven, R. Glenn Hubbard and Bruce Petersen** “Financing Constraints and Corporate Investment.” *Brooking Papers on Economic Activity* 1, 1988, pages 141-95.
- Genesove, David and Christopher J. Mayer** “Equity and Time to Sale in the Real Estate Market.” *American Economic Review* 87(3), June 1997, pages 255-69.
- Gertler, Mark and Stephen Gilchrist** “The Role of Credit Market Imperfections in the Monetary Transmission Mechanism: Arguments and Evidence.” *Scandinavian Journal of Economics* 95(1), 1993, pages 43-64.
- Gertler, Mark and Stephen Gilchrist** “Monetary Policy, Business Cycles, and the Behavior of Small Manufacturing Firms.” *Quarterly Journal of Economics* 109(2), May 1994, pages 309-40.
- Haurin, Donald R., Patric H. Hendershott and Susan M. Wachter** “Borrowing Constraints and the Tenure Choice of Young Households.” *Journal of Housing Research* 8(2), 1997, pages 137-54.
- Hendershott, Patric H., William C. LaFayette and Donald R. Haurin** “Debt Usage and Mortgage Choice: The FHA-Conventional Decision.” *Journal of Urban Economics* 41(2), March 1997, pages 202-17.
- Hubbard, R. Glenn** “Capital Market Imperfections and Investment.” *The Journal of Economic Literature* 36, March 1998, pages 193-227.
- Jappelli, Tullio and Marco Pagano** “Consumption and Capital Market Imperfections: An International Comparison.” *American Economic Review* 79, December 1989, pages 1088-1105.
- Jappelli, Tullio and Marco Pagano** “Savings, Growth and Liquidity Constraints.” *Quarterly Journal of Economics* 109, February 1994, pages 83-109.
- Kiyotaki, Nobuhiro and John Moore** “Credit Cycles.” *Journal of Political Economy* 105, April 1997, pages 211-48.
- Kocherlakota, Narayanan** “Creating Business Cycles Through Credit Constraints.” *Federal Reserve Bank of Minneapolis Quarterly Review*, Summer 2000, pages 2-10.
- Iacoviello, Matteo and Raoul Minetti** “Financial Liberalisation and the Sensitivity of Housing Prices to Monetary Policy: Theory and Evidence.” *The Manchester School* 71, January 2003 pages 20-34.
- Lamont, Owen and Jeremy Stein** “Leverage and Housing Price Dynamics in US Cities.” *RAND Journal of Economics* 30, Autumn 1999, pages 498-514.
- LaPorta, Raphael, Florencio Lopez-De-Silanes, Andrei Shleifer and Robert Vishny** “Law and Finance.” *Journal of Political Economy* 106, pages 1113-55.

- Linneman, P. and Susan Wachter** “The Impacts of Borrowing Constraints on Homeownership.” *AREUEA Journal* 27, Winter 1989, pages 389-402.
- Malpezzi, Stephen** “Urban Housing and Financial Markets. Some International Comparisons.” *Urban Studies* 27, December 1990, pages 71-102.
- Malpezzi, Stephen and Stephen Mayo** “Getting Housing Incentives Right: A Case Study of the Effects of Regulation, Taxes and Subsidies on Housing Supply in Malaysia”, *Land Economics* 73, 1997, pages 372-91.
- Mayer, Christopher and C. Tsur Sommerville** “Residential Construction: Using the Urban Growth Model to Estimate Housing Supply.” *Journal of Urban Economics* 48(1), 2000, pages 85-109.
- MacLennan Duncan, John Muellbauer and Mark Stephens** “Asymmetries in Housing and Financial Market Institutions and EMU.” *Oxford Review of Economic Policy* 14, Autumn 1998, pages 54-80.
- Meen, G.** “Ten Propositions in UK Housing Macroeconomics: An Overview of the 1980s and Early 1990s.” *Urban Studies* 33, 1996.
- Mishkin, Frederic S.** “A Note on Short-Run Asset Effects on Household Saving and Consumption.” *American Economic Review* 67(2), March 1977, pages 246-48.
- Mishkin, Frederic S.** “The Household Balance Sheet and the Great Depression.” *Journal of Economic History* 38(4), December 1978, pages 918-37.
- Ortalo-Magne, Francois and Sven Rady** “Housing Market Fluctuations in a Life-Cycle Economy with Credit Constraints.” 1998, mimeo.
- Ortalo-Magne, Francois and Sven Rady** “Why Are Housing Prices So Volatile? Income Shocks In a Stochastic Heterogeneous-Agents Model.” 1999, mimeo.
- Poterba, James M.** “House Price Dynamics: The Role of Tax Policy and Demography.” *Brook- ing Papers on Economic Activity* 4(2), 1991, pages 143-83.
- Rajan, Raghuram G. and Luigi Zingales** “Financial Dependence and Growth.” *American Economic Review* 88(3), June 1998, pages 559-86.
- Spiegel, Matthew** “Housing Return and Construction Cycles.” *Real Estate Economics* 29, Winter 2001, pages 521-51.
- Stein, Jeremy** “Prices and Trading Volume in the Housing Market: A Model with Down-Payment Effects.” *Quarterly Journal of Economics* 110, May 1995, pages 379-406.
- Stephens, Mark** “Monetary Policy and Housing Price Volatility in Western Europe.” *Housing Studies* 10, 1995, pages 105-20.
- Voith, Richard** “The Suburban Housing Market: The Effects of City and Suburban Job Growth.” *Business Review*, Federal Reserve Bank of Philadelphia, Nov./Dec. 1996, pages 13-25.

A Data Appendix

This appendix describes in detail several of the data items we use in the paper.

A Housing Price Indices

Most of the data for developed countries are supplied by Peter Englund, which is the same data used in Englund and Ioannides (1997). Below we refer to this source as “EIO”. Their data covers the period 1970-1992. We update their data set using the Annual Reports from the Bank of International Settlements (BIS), which give information on the same indices used by Englund and Ioannides. For countries not included in the Englund and Ioannides data set, we use other sources described below. We list all the specific sources for each country, and the information we have about the respective indices.

Australia. EIO, and BIS. Weighted average index of prices for all capital cities and other areas; obtained from quarterly national census of home loan approvals, available annually. Updated using the AUEHPI index from the Australian Bureau of Statistics.

Belgium. EIO, and BIS. Index based on annual transactions reports on small and medium sized dwellings from entire country, with outliers excluded, available annually.

Canada. EIO, and BIS. Average annual transaction prices reported by multiple listing services for entire country, covering 70% of all transactions. Updated using the New House Price Index from the Statistics Canada, available at <http://cansim2.statcan.ca>.

Chile. Data provided by Felipe Morande, from Morande, F. and R. Soto (1992) updated by R. Soto. Based on standardized dwellings in the area of Santiago, annual average.

Denmark. EIO, and BIS. Average value of single-family homes, including only arms' length sales, available annually.

Finland. EIO and BIS. Average price per apartment and terraced houses, obtained per square meter, as recorded by realtors (including 30% of all transactions), weighted by region, available quarterly.

France. EIO and BIS. Index based on BIS' own estimate, based on annual values for the Paris region, adjusted by four-year survey for entire country.

Germany. EIO and BIS. Transaction prices per square meter, obtained from realtors for the four largest cities, available annually.

Hong Kong. Index constructed by the Rating and Valuation Department, from the Hong Kong Property Review, data from Chou and Shih (1995), updated using data on the same index available at <http://www.info.gov.hk>.

Ireland. EIO and BIS. Average transactions price for existing homes, based on all loan approvals, available annually.

Israel. Property price index representative of the entire country, from the Social Sciences Data Archive (data used in Bar Nathan et al., 1998), updated using data from the Israel Central Bureau of Statistics (www.cbs.gov.il/srcer.cgi)

Italy. EIO and BIS. Average price for new and completely refurbished dwellings in large and middle-sized cities and tourist areas, reported by realtors, available annually.

Japan. Urban Residential Land Price Index, from the Japanese Real Estate Institute, available at www.reinet.or.jp.

Korea. Land Price of Housing, from the Korea Appraisal Board, Appraisal Research & Development Center, available at www.kreic.com.

Malaysia. IHRM (Malaysian House Price Index % change from previous year). Data provided by Steve Malpezzi and used in Malpezzi and Mayo (1997), updated using data from the Countrywide's Sourcebook 2000.

Netherlands. EIO and BIS. Weighted average sales price for existing single and multi-family homes, reported by realtors, including 50-60% of all transactions, available annually.

New Zealand. Data from Dalziel and Lattimore (1999), Valuation New Zealand Housing Price Series, average prices of free-hold house sales, adjusted for quality, updated using BIS data.

Norway. EIO and BIS. Average sales price of existing homes, weighted by type of dwelling, reported by Property Owner's Association, covering about 50% of all transactions.

Singapore. Data from Phang and Wong (1997). Value weighted average of current prices of five types of property in five planning districts. Excludes public housing. Updated using the SIPRIRES index of the Singapore Department of Statistics.

Spain. Data provided by O. Bover. Prices per square meter of new dwellings in Madrid, used in Bover (1993). Updated with the Price Index for Existing Dwellings, from Hypostat 1999.

Sweden. EIO and BIS. Index based on owner-occupied one- and two-dwelling buildings, based on reports of title registrations for arm's length transactions, weighted by type of dwelling, available annually.

Switzerland. Real estate price index for 3-5 bedroom single family homes, from the Swiss National Bank (<http://www.snb.ch/e/search/index.html>).

Taiwan. Median of Housing Prices in Taipei, provided by Shiawee Yang.

Thailand. Real housing price index used in Malpezzi and Mayo (1997). Data provided by S. Malpezzi covering the period from 1970-1986. Updated using the series on Land Price Increases in Bangkok, from the Agency for Real Estate Affairs.

UK. EIO and BIS. Index based on survey of all dwellings with building societies mortgages, weighted by type of dwelling, available annually.

US. EIO and BIS. Index based on sales price of existing single-family homes, based on realtor reports, adjusted by regional availability of single-family homes and homeowner mobility, available annually.

B New Mortgages

Data for net new mortgage lending for Belgium, Denmark, Germany, Ireland, Finland, Netherlands and Spain is from Hypostat *1989-1999*, and data for Canada, France, Italy, Japan, Norway, Sweden, UK and US is from the OECD, also used by Girouard and Blondal (2001), and kindly provided to us by Nathalie Girouard.

C Maximum LTV Ratios

Data is from Jappelli and Pagano (1994), updated with data from Chiuri and Jappelli (2000). The data is given in 10-year averages. We extended this data as follows: for Denmark, Japan, New Zealand and Norway we assumed the Jappelli and Pagano 1980-1987 data extends to 1990. We took 1991-1999 data for Denmark and Norway from MacLennan, Muellbauer and Stephens (1998). Singapore 1991-1999 data is from Phang and Wong (1997). The data for Chile, Hong Kong, Korea (1980-1999), Japan (1991-1999), New Zealand (1991-1999) and Switzerland is from the Countrywide's Sourcebook, 1995 and 2000. Malaysia and Thailand 1991-1999 data is from the Asian Development Bank, 1999.

D Homeownership Ratios

Data for Australia, Belgium, Canada (1970-1989), France (1970-1980), Germany (1970-1980), Italy (1970-1980), Netherlands (1970-1980), Spain (1970-1980), and Taiwan is from Chiuri and Jappelli (2000). Data for Chile, Denmark, Finland, France (1981-1999), Germany (1981-1999), Hong Kong, Ireland, Italy (1981-1999), Japan, Netherlands (1981-1999), Norway, Sweden and Spain (1981-1999) is from the Countrywide's Sourcebook *2000*. Data for Korea and Malaysia is from the Asian Development Bank, 1999. Data for Canada (1991-1999), New Zealand, Singapore, Switzerland, Thailand, UK and US is from the Euromonitor (available at www.euromonitor.com).

E Price-Income Ratios

The data on personal disposable income is from the Economic Outlook No 70: Annual and Semi-annual data (Source: OECD), with the following exceptions: the data for Denmark and Thailand is from DRI-Wefa (<http://www.dri-wefa.com/>), the data for Taiwan is taken from the Government statistics at <http://www.stat.gov.tw>. We collected the nominal housing price for a particular year, and then we used the housing price index described above to extrapolate the series for all years. The data for Belgium, Denmark, Finland, France, Italy, Netherlands, Spain and Sweden represents the typical price for a flat of 150 square meters in 1999, and is taken from the Countrywide's Sourcebook 2000. The data for Canada (average price of all dwellings, 1995-1999), Ireland (average new house price for the whole country, 1996-1998), Korea (median price of typical 710 square feet apartment in Seoul in 1990), New Zealand (median price of a home, 1999), UK (Mix-adjusted average house price in 1999), and the US (average existing single family house price from 1990-1999), are also taken from the Countrywide's Sourcebook 2000. Below we list the sources and definitions for the remaining countries:

Australia - typical house price in 1999, from <http://www.amp.com.au/au/ampweb.nsf/Content>.
 Chile - price of an standardized dwelling in selected areas of Santiago, 1975-1998, from Morande and Soto (1992).
 Germany - price of existing detached houses, 1970-1993, from Muelder and Wagner (1998).
 Hong Kong - price of a 100 square meter flat, 1982-1992, from Chou and Shih (1995).
 Israel - typical apartment price in 1999, from www.jpost.com
 Japan - typical apartment price in 1999, from www.pricechecktokyo.com
 Malaysia- typical price of a single-story detached home in 1998, from www.jp-ph.gov.my
 Norway - average price of a 150 square meter flat, from Statistics Norway (www.ssb.no).
 Singapore - 1999, average house price from Asia Week, www.asiaweek.com
 Switzerland - price of an average 4 bedroom semi-detached house with parking in 1999, from www.expatacess.com
 Taiwan - actual average housing purchase price, 1981-1989, from Lin (1993).
 Thailand - 1994-97 average house price, from the Asian Development Bank.

References for the Appendix

- Asian Development Bank** *Mortgage-Backed Securities Markets in Asia*, 1999, online publication available at <http://www.adb.org>.
- Bank of International Settlements** *Annual Reports* no. 64, 65, 66, 67, 68, 69, 70.
- Bar Nathan, M., M. Beenstock and Y. Haitovsky** "The Market for Housing in Israel." *Regional Science and Urban Economics* 28, 1998, pages 21-49.
- Bover, Olympia** "Un Modelo Empirico de la Evolucion de los Precios de la Vivienda en Espana." *Investigaciones Economicas* XVII, 1993, pages 65-86.
- Chou, W.L. and Y.C. Shih** "Hong Kong Housing Markets: Overview, Tenure Choice and Housing Demand." *Journal of Real Estate Finance and Economics* 10, 1995, pages 7-21.
- Countrywide International** *International Union for Housing Finance (IUHF) International Data and Analysis: Countrywide's Sourcebook*, 1995, 2000.
- Dalziel, Paul and Ralph Lattimore**, *The New Zealand Macroeconomy: A Briefing on the Reforms*. Second edition. Melbourne; Oxford and New York: Oxford University Press, 1996.
- European Mortgage Federation**, *Hypostat 1989-1999: Mortgage and Property Markets in the European Union and Norway*, 2000, available online at www.hypo.org.
- Girouard, Nathalie and Sveinbjorn Blondal** "House Prices and Economic Activity." OECD Economics Department Working Papers no. 279, 1999.
- Lin, C. S.**, "The Relationship Between Rents and Prices of Owner-Occupied Housing in Taiwan", *Journal of Real Estate Finance and Economics* 6, 1993, pages 25-54.
- Malpezzi, Stephen and Stephen Mayo** "Getting Housing Incentives Right: A Case Study of the Effects of Regulation, Taxes, and Subsidies of Housing Supply in Malaysia." *Land Economics* 73, 1997, pages 372-91.
- Morande, Felipe and Raimundo Soto** "Una nota sobre la construcción de series de precios de activos reales: Tierra y Casas en Chile." *Revista de Análisis Económico* 7, 1992, pages 169-78.
- Muelder C. and M. Wagner** "First Time Home-Ownership in the Family Life Course: A West German-Dutch Comparison." *Urban Studies* 35, 1997, pages 687-713.
- OECD**, *Economic Outlook No 70: Annual and Semiannual Data*.
- Phang, Sock-Yang and Wing-Keung Wong** "Government Policies and Private Housing Prices in Singapore." *Urban Studies* 34, 1997, pages 1819-29.

Table 1: Summary Statistics of Housing Price Changes, Income Growth, and New Mortgages

This table displays summary statistics for housing prices changes, income growth, and new mortgages for 26 countries over the 1970-1999 period. $\Delta\text{Log}(\text{Price})$ is the log change in the real housing price index. $\Delta\text{Log}(\text{Income})$ is the log change in real per capita GDP. New mortgages are net new lending against mortgage in residential property divided by nominal GDP. GDP, population, and inflation data are from the IMF's *International Financial Statistics*. The housing price and new mortgage data are described in the Appendix.

| | Mean | Std. Dev. | Pct 5 | Pct 25 | Median | Pct 75 | Pct 95 | N. Obs |
|-----------------------------------|-------|-----------|--------|--------|--------|--------|--------|--------|
| $\Delta\text{Log}(\text{Price})$ | 0.020 | 0.116 | -0.150 | -0.034 | 0.015 | 0.072 | 0.210 | 718 |
| $\Delta\text{Log}(\text{Income})$ | 0.030 | 0.045 | -0.033 | 0.007 | 0.027 | 0.051 | 0.102 | 754 |
| <i>New Mortgages</i> | 0.030 | 0.022 | 0.002 | 0.015 | 0.027 | 0.040 | 0.069 | 278 |

Table 2: Maximum Loan-to-Value (LTV), Homeownership, and Price-Income Ratios by Country-Decade, 1970-1999

Maximum LTV ratios represent the highest mortgage loan that households can get from lenders as a fraction of the value of the property owned. The homeownership ratio is the proportion of homeowners as a fraction of total households. The price-income ratio is the nominal price of a typical home divided by personal disposable income per capita. All data items are described in the Appendix.

| Country | LTV Ratio | | | Homeown. Ratio | | | Price/Income Ratio | | |
|-------------|-----------|------|------|----------------|------|------|--------------------|------|------|
| | 70's | 80's | 90's | 70's | 80's | 90's | 70's | 80's | 90's |
| Australia | 0.70 | 0.80 | 0.80 | 0.70 | 0.70 | 0.70 | 10.5 | 8.5 | 9.5 |
| Belgium | 0.65 | 0.75 | 0.80 | 0.66 | 0.66 | 0.66 | 9.4 | 7.5 | 8.4 |
| Canada | 0.75 | 0.80 | 0.80 | 0.63 | 0.63 | 0.64 | 8.9 | 7.8 | 8.6 |
| Chile | N/A | N/A | 0.78 | N/A | 0.63 | N/A | N/A | 9.0 | 13.0 |
| Denmark | 0.85 | 0.95 | 0.80 | N/A | 0.55 | 0.52 | N/A | 8.8 | 7.6 |
| Finland | 0.80 | 0.85 | 0.80 | 0.61 | 0.65 | 0.62 | 16.4 | 15.3 | 10.1 |
| France | 0.80 | 0.80 | 0.80 | 0.57 | 0.53 | 0.54 | 8.9 | 8.4 | 9.8 |
| Germany | 0.65 | 0.65 | 0.80 | 0.43 | 0.43 | 0.41 | 22.3 | 18.0 | 15.7 |
| Hong Kong | N/A | 0.90 | 0.70 | 0.18 | 0.33 | 0.47 | N/A | 21.8 | 34.0 |
| Ireland | 0.80 | 0.90 | 0.80 | N/A | 0.77 | 0.79 | 9.6 | 9.0 | 9.1 |
| Israel | 0.50 | 0.70 | N/A | 0.70 | N/A | 0.80 | N/A | 19.0 | 25.1 |
| Italy | 0.50 | 0.56 | 0.60 | 0.63 | 0.67 | 0.73 | 17.2 | 14.4 | 10.7 |
| Japan | N/A | 0.60 | 0.55 | 0.60 | 0.62 | 0.60 | 22.2 | 22.1 | 20.4 |
| Korea | 0.30 | 0.40 | 0.40 | 0.59 | 0.52 | 0.52 | 41.6 | 42.4 | 32.3 |
| Malaysia | 0.65 | N/A | 0.85 | N/A | N/A | 0.67 | N/A | 21.9 | 24.2 |
| Netherlands | 0.75 | 0.75 | 0.75 | 0.48 | 0.46 | 0.51 | 12.2 | 9.5 | 11.0 |
| New Zealand | 0.66 | 0.80 | 0.80 | N/A | 0.71 | 0.73 | 7.4 | 7.0 | 8.8 |
| Norway | 0.75 | 0.80 | 0.80 | 0.74 | 0.78 | 0.76 | 13.8 | 13.6 | 9.6 |
| Singapore | N/A | N/A | 0.85 | N/A | 0.90 | 0.88 | N/A | 32.4 | 43.2 |
| Spain | 0.60 | 0.80 | 0.80 | 0.74 | 0.73 | 0.78 | 10.3 | 10.6 | 13.2 |
| Sweden | 0.90 | 0.95 | 0.75 | 0.50 | 0.54 | 0.60 | 15.1 | 11.2 | 9.6 |
| Switzerland | N/A | N/A | 0.90 | 0.30 | 0.31 | N/A | N/A | 36.5 | 27.1 |
| Taiwan | 0.40 | N/A | N/A | 0.77 | 0.78 | 0.84 | 7.2 | 7.6 | 4.5 |
| Thailand | 0.65 | N/A | 0.75 | 0.89 | 0.86 | 0.82 | 17.4 | 16.5 | 29.4 |
| UK | 0.81 | 0.87 | 0.95 | 0.56 | 0.61 | 0.67 | 10.4 | 10.6 | 8.6 |
| US | 0.80 | 0.89 | 0.80 | 0.66 | 0.64 | 0.65 | 7.4 | 7.3 | 6.9 |

Table 3: Housing Price Dynamics

The dependent variable is $\Delta \text{Log}(\text{Price})$, the log change in the real housing price index. $\Delta \text{Log}(\text{Income})$ is the log change in real per capita GDP. $\text{Price}_{t-1}/\text{Income}_{t-1}$ is the start-of-period ratio of the real housing price index to real per capita GDP. Real interest rate is the nominal long-term interest rate on a government bond (usually 10-year benchmark government bond yield), from the IMF's *International Financial Statistics* or from the OECD's *Economic Outlook*, minus the inflation rate in the same year. Inflation rate is the change in the consumer price index for the current year, taken from the IMF's *International Financial Statistics*. The estimation period is 1970-1999. The estimations correct the error structure for heteroskedasticity using the White-Huber estimator. t -stats (in parentheses).

| Indep. Variables | (1) | (2) | (3) | (4) | (5) | (6) | (7) |
|---|--------------------|--------------------|--------------------|--------------------|----------------------|----------------------|----------------------|
| $\Delta \text{Log}(\text{Income})_t$ | 1.187 (9.07)*** | 0.942 (6.61)*** | 0.942 (6.36)*** | 1.125 (7.17)*** | 1.061 (6.94)*** | 1.009 (7.13)*** | 1.022 (7.49)*** |
| $\Delta \text{Log}(\text{Income})_{t-1}$ | | 0.510 (3.38)*** | 0.409 (2.52)*** | 0.555 (3.47)*** | 0.214 (1.44) | 0.356 (2.40)** | |
| $\Delta \text{Log}(\text{Income})_{t-2}$ | | 0.248 (2.23)** | 0.083 (0.58) | 0.342 (2.15)** | 0.171 (1.13) | 0.176 (1.17) | |
| $\Delta \text{Log}(\text{Price})_{t-1}$ | | | 0.241 (3.33)*** | 0.193 (2.61)*** | 0.347 (4.84)*** | 0.278 (3.76)*** | 0.348 (5.34)*** |
| $\Delta \text{Log}(\text{Price})_{t-2}$ | | | -0.099 (-1.62) | -0.111 (-1.80)* | -0.169 (-2.46)*** | 0.045 (0.85) | |
| <i>Interest Rate</i> | | | | | -0.289 (-2.41)** | | |
| <i>Inflation Rate</i> | | | | | -0.109 (-1.06) | | |
| $\text{Price}_{t-1}/\text{Income}_{t-1}$ | | | | | | -0.253 (-7.67)*** | -0.246 (-7.50)*** |
| $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j}$ | | 1.700 | 1.434 | 2.022 | 1.446 | 1.541 | |
| Summation Test p -value | | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | |
| Exclusion Test p -value | | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | |
| Country Effects? | No | No | No | Yes | No | Yes | Yes |
| Year Effects? | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| Observations | 718 | 679 | 666 | 666 | 616 | 666 | 692 |
| Adj- R^2 | 0.226 | 0.265 | 0.310 | 0.317 | 0.381 | 0.408 | 0.378 |

***, **, * indicate statistical significance at 1%, 5%, and 10% (two-tail) test levels, respectively.

Table 4: House Prices and the Multiplier Effect: Baseline Regressions

The dependent variable is $\Delta \text{Log}(\text{Price})_t$, the log change in the real housing price index. $\Delta \text{Log}(\text{Income})_t$ is the log change in real per capita GDP. $\text{Price}_{t-1}/\text{Income}_{t-1}$ is the start-of-period ratio of the real housing price index to real per capita GDP. Real interest rate is the nominal long-term interest rate on a government bond (usually 10-year benchmark government bond yield), from the IMF's *International Financial Statistics* or from the OECD's *Economic Outlook*, minus the inflation rate in the same year. Inflation rate is the change in the consumer price index for the current year, taken from the IMF's *International Financial Statistics*. LTV_t is the maximum LTV ratio for year t . The estimation period is 1970-1999. The estimations correct the error structure for heteroskedasticity using the White-Huber estimator. t -stats (in parentheses).

| Indep. Variables | (1) | (2) | (3) | (4) | (5) |
|---|--------------------|--------------------|---------------------|--------------------|----------------------|
| $\Delta \text{Log}(\text{Income})_t$ | -0.437 (-0.96) | -0.787 (-1.31) | -0.622 (-1.00) | -0.273 (-0.42) | 0.051 (0.10) |
| $\Delta \text{Log}(\text{Income})_{t-1}$ | | 1.174 (1.11) | 1.029 (1.03) | 0.132 (0.17) | |
| $\Delta \text{Log}(\text{Income})_{t-2}$ | | -0.470 (-0.62) | -0.199 (-0.24) | 0.504 (0.91) | |
| $\Delta \text{Log}(\text{Price})_{t-1}$ | | 0.228 (3.01)*** | 0.174 (2.08)** | 0.299 (4.21)*** | 0.332 (4.88)*** |
| $\Delta \text{Log}(\text{Price})_{t-2}$ | | -0.070 (-1.35) | -0.081 (-1.47) | -0.089 (-1.71)* | |
| <i>Interest Rate</i> | | | | -0.287 (-1.39) | |
| <i>Inflation Rate</i> | | | | -0.092 (-0.69) | |
| $\text{Price}_{t-1}/\text{Income}_{t-1}$ | | | | | -0.231 (-8.71)*** |
| LTV_t | -0.065 (-1.52) | -0.037 (-0.81) | -0.214 (-2.40)** | 0.007 (0.16) | -0.068 (-0.82) |
| $\Delta \text{Log}(\text{Income})_t \times \text{LTV}_t$ | 2.276 (3.58)*** | | | | 1.315 (1.80)* |
| $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j} \times \text{LTV}_t$ | | 2.152 (2.45)*** | 2.414 (1.95)** | 1.420 (1.75)* | |
| Country Effects? | No | No | Yes | No | Yes |
| Year Effects? | Yes | Yes | Yes | Yes | Yes |
| Observations | 611 | 567 | 567 | 531 | 589 |
| Adj- R^2 | 0.220 | 0.297 | 0.316 | 0.342 | 0.362 |

***, **, * indicate statistical significance at 1%, 5%, and 10% (two-tail) test levels, respectively.

Table 5: House Prices and the Multiplier Effect: Alternative Specifications

The dependent variable in columns (1) through (7) is $\Delta \text{Log}(\text{Price})$, the log change in real housing price index. The dependent variable in columns (8) and (9), New Mortgages, is the net new lending against mortgage in residential property divided by GDP. $\Delta \text{Log}(\text{Income})$ is the log change in real per capita GDP. $\text{Price}_{t-1}/\text{Income}_{t-1}$ is the start-of-period ratio of the real housing price index to real per capita GDP. In columns (1) and (2) we instrument LTV_t with proxies for the quality of accounting standards and judicial efficiency. Judicial efficiency is an assessment of the efficiency and integrity of the legal environment as it affects business, compiled by the Business International Corporation, taken from LaPorta et al. (1998). Accounting standards is the index of accounting standards computed by the Center for International Financial Analysis and Research, data from LaPorta et al. (1998). In columns (3) and (4) we control for the level of economic development (PPP-adjusted per capita GDP) by including the intercept variable as well as its interactions with each of the lags of $\Delta \text{Log}(\text{Income})$ (coefficients omitted). Likewise, in columns (5) and (6) we add intercept and interaction terms for homeownership and $\Delta \text{Log}(\text{Income})$. The data for per capita GDP in constant prices is from *Penn World Tables*, taken from the Barro and Lee (1994) dataset, and augmented with data from the Global Development Finance & World Development Indicators. The homeownership ratio is the proportion of homeowners as a fraction of total households. In column (7) we use the GMM estimator for dynamic panel data proposed by Arellano and Bond (1991). The estimation period is 1970-1999. The OLS estimations correct the error structure for heteroskedasticity using the White-Huber estimator. t -stats (in parentheses).

| Indep. Variables | IV | | Added Controls for | | Added Controls for | | GMM | Dep. Variable: | |
|---|----------------------|---------------------|-----------------------|---------------------|----------------------|---------------------|----------------------|----------------------|--------------------|
| | Fin. Develop. (1) | (2) | Econ. Develop. (3) | (4) | Homeownership (5) | (6) | (7) | New Mortgages (8) | (9) |
| $\Delta \text{Log}(\text{Income})_t$ | -2.940 (-2.51)** | -2.717 (-2.36)** | -0.671 (-1.17) | -0.512 (-0.85) | -0.971 (-0.88) | -0.900 (-0.79) | -0.471 (-0.94) | -0.632 (-1.82)* | -0.584 (-1.75)* |
| $\Delta \text{Log}(\text{Income})_{t-1}$ | 1.802 (1.04) | 1.884 (1.13) | 1.194 (1.13) | 1.054 (1.07) | 1.044 (0.86) | 0.828 (0.68) | 1.968 (2.75)*** | 0.266 (0.86) | 0.195 (0.82) |
| $\Delta \text{Log}(\text{Income})_{t-2}$ | -1.583 (-1.35) | -1.357 (-1.09) | -0.411 (-0.53) | -0.187 (-0.23) | -0.113 (-0.13) | 0.001 (0.01) | -1.766 (-3.73)*** | -0.219 (-0.64) | -0.176 (-0.63) |
| $\Delta \text{Log}(\text{Price})_{t-1}$ | 0.227 (2.86)*** | 0.188 (2.14)** | 0.230 (3.05)*** | 0.184 (2.21)** | 0.233 (2.90)*** | 0.164 (1.77)* | 1.051 (23.80)*** | | |
| $\Delta \text{Log}(\text{Price})_{t-2}$ | -0.088 (-1.64) | -0.099 (-1.49) | -0.076 (-1.45) | -0.091 (-1.64) | -0.049 (-0.83) | -0.068 (-1.08) | -0.315 (-7.45)*** | | |
| LTV_t | -0.077 (-0.97) | -0.079 (-1.01) | -0.056 (-1.17) | -0.175 (-2.10)** | -0.038 (-0.79) | -0.217 (-2.40)** | -0.025 (-2.23)** | 0.052 (3.88)*** | 0.059 (2.62)*** |
| $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j}$ $\times LTV_t$ | 6.078 (3.66)*** | 5.924 (2.83)*** | 2.085 (2.04)** | 2.275 (1.56) | 2.048 (2.15)** | 3.214 (2.60)*** | 0.754 (3.37)*** | 0.947 (1.85)* | 0.765 (1.48) |
| Country Effects? | No | Yes | No | Yes | No | Yes | Yes | No | Yes |
| Year Effects? | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| Observations | 540 | 540 | 567 | 567 | 510 | 510 | 567 | 265 | 265 |
| Adj- R^2 | 0.313 | 0.322 | 0.307 | 0.323 | 0.289 | 0.316 | 52.05 ^(a) | 0.231 | 0.512 |

Table Notes: ^(a) F -statistic. ***, **, * indicate statistical significance at 1%, 5%, and 10% (two-tail) test levels, respectively.

Table 6: The Income Constraint Effect

For each country, we use the average price-income ratio for the period 1970-1999 (subject to data availability) to classify countries in the “cheap” and “expensive” categories. The price-income ratio is the nominal price of a typical home divided by personal disposable income percapita. Cheap (expensive) housing countries are those ranked in the bottom (top) third of the cross-country distribution of the ratio of house prices to per capita GDP. The countries in the cheap housing category are: Australia, Belgium, Canada, Denmark, France, Netherlands, New Zealand, Tawain, and the US. The expensive housing countries are: Hong Kong, Israel, Japan, Korea, Malaysia, Singapore, Switzerland, and Thailand. The estimation period is 1970-1999. The baseline-OLS specification is the one in column (3) of Table 4, including three lags of income and housing price changes, as well as the interactions of the LTV ratio with the income changes. The baseline-GMM specification uses the GMM estimator for dynamic panel data proposed by Arellano and Bond (1991), as in column (7) of Table 4. The Lamont and Stein specification is the one in column (5) of Table 4, which includes the current change in per capita GDP and its interaction with the LTV ratio. The OLS estimations correct the error structure for heteroskedasticity using the White-Huber estimator. t -stats (in parentheses).

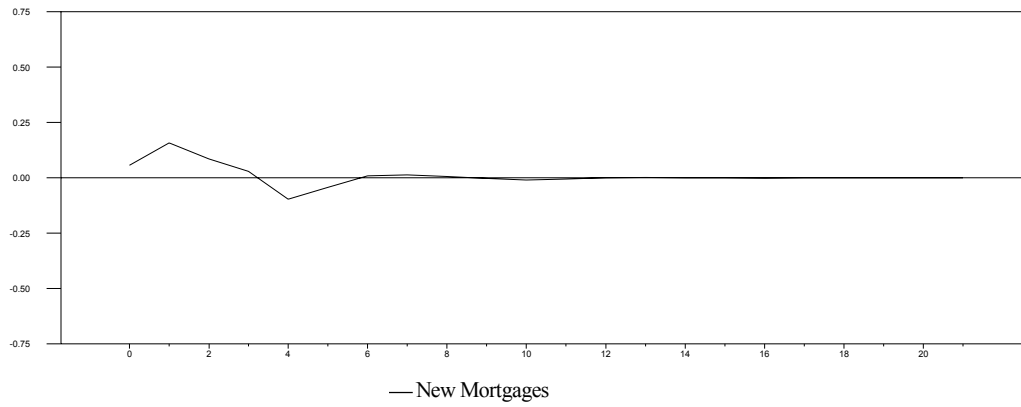
| | Baseline-OLS $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j} \times \text{LTV}_t$ | Baseline-GMM $\sum_{j=0}^2 \Delta \text{Log}(\text{Income})_{t-j} \times \text{LTV}_t$ | Lamont-Stein $\Delta \text{Log}(\text{Income})_t \times \text{LTV}_t$ |
|-----------------------------|---|---|--|
| Cheap Housing Countries | 8.497 (1.64)* | 2.808 (2.59)*** | 9.742 (3.92)*** |
| Expensive Housing Countries | 2.508 (1.38) | 0.227 (0.45) | 1.001 (0.93) |

***, **, * indicate statistical significance at 1%, 5%, and 10% (two-tail) test levels, respectively.

Figure 1: Impulse-Response Functions

This figure shows the response of new mortgages to a one-standard-deviation shock to per capita GDP. The estimates come from a two-lag VAR system that also includes changes in log CPI and interest rates. Panel A plots results for data from Italy and Panel B uses US data. The system is estimated with yearly series from 1975 through 1999.

Panel A: Italy



Panel B: US

