

The Consequences of Mortgage Credit Expansion: Evidence from the 2007 Mortgage Default Crisis*

Atif Mian

University of Chicago Graduate School of Business and NBER

Amir Sufi

University of Chicago Graduate School of Business

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Abstract

We demonstrate that a rapid expansion in the supply of mortgages driven by disintermediation explains a large fraction of recent U.S. house price appreciation and subsequent mortgage defaults. We identify the effect of shifts in the supply of mortgage credit by exploiting within-county variation across zip codes that differed in latent demand for mortgages in the mid 1990s. From 2001 to 2005, high latent demand zip codes experienced large relative decreases in denial rates, increases in mortgages originated, and increases in house price appreciation, despite the fact that these zip codes experienced significantly negative relative income and employment growth over this time period. These patterns for high latent demand zip codes were driven by a sharp relative increase in the fraction of loans sold by originators shortly after origination, a process which we refer to as “disintermediation.” The increase in disintermediation-driven mortgage supply to high latent demand zip codes from 2001 to 2005 led to subsequent large increases in mortgage defaults from 2005 to 2007. Our results suggest that moral hazard on behalf of originators selling mortgages is a main culprit for the U.S. mortgage default crisis.

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Recent developments in the U.S. housing market are the focus of increased anxiety among policy-makers, investors, and financial markets. After experiencing a dramatic rise in house prices and outstanding mortgage debt, the U.S. has experienced a sharp increase in mortgage defaults. Figure 1A shows that the average house price in the U.S increased by almost 200% from 1996 to 2005. Figure 1B demonstrates a similar rise in outstanding mortgage debt, and also that the rise in mortgage debt has been significantly larger than the rise in other types of consumer debt. Figure 1C shows that the total number of accepted mortgage applications for new home purchase increased rapidly from 2003 to 2005, suggesting the entry of new consumers in this market during this time period.

The rapid growth in mortgage credit and house prices has given way to grave concerns as mortgage defaults continue to mount. For example, default rates rose by over 50% between the fourth quarter of 2005 and the second quarter of 2007 (Figure 1D). The market value of mortgage securities has fallen precipitously as well, with some tranches losing up to 70 to 80% of their value in less than a year. Many believe that weakness in the U.S. housing market poses a serious threat to financial markets and economic growth.¹ The January 22nd, 2008 FOMC statement justified a 75 basis point reduction in the federal funds rate in part because "... incoming information indicates a deepening of the housing contraction ..."

The goal of this paper is to investigate the cause of the sharp rise in house prices and subsequent spike in mortgage default rates. Our central finding is that a rapid expansion in the *supply* of credit to zip codes with high latent demand for mortgages is a main cause of both house price appreciation from 2001 to 2005 and the subsequent sharp increase in defaults from

¹ See Ng and Lauricella (WSJ, July 13th, 2007), Wessel (WSJ, September 13th, 2007), Lahart (August 27th, 2007), and Bater (WSJ, November 20th, 2007). Press articles refer to the current mortgage environment as a "mortgage crisis" (Creswell and Bajaj, NYT, March 5th, 2007) that "is comparable to some of the biggest financial disasters of the past half-century" (Ip, Whitehouse, and Lucchetti, WSJ, December 10th, 2007).

2005 to 2007. The expansion in credit supply was driven by a shift in the mortgage industry towards “disintermediation”, which we define as the process in which originators sell mortgages in the secondary market shortly after origination.

Our analysis is based on a new data set constructed from a number of proprietary and public data sources. It represents one of the most comprehensive and disaggregated data sets in the real estate and consumer credit literature. More specifically, our zip code-year level data set covers 1996 through 2007 and includes a number of key variables of interest including outstanding consumer debt of different types, defaults, house prices, mortgage loan application characteristics, mortgage terms, and demographic variables such as income and crime. Given the large number of zip codes in our sample, we are able to exploit cross-sectional variation over time in zip code outcomes to empirically isolate our coefficients of interest.

We identify the causal effect of expansion in the supply of mortgage credit on subsequent changes in loan originations, house prices, and defaults by exploiting within-county across-zip code variation in initial *latent demand* for mortgages. Our primary measure of initial latent demand is the fraction of mortgage applications denied by originating institutions in the zip code in 1996.

The intuition behind our identification strategy can be understood through a simple example. Consider two zip codes A and B that lie in the same county, but differ in the credit worthiness of its households. In particular, suppose that all households in zip code A are creditworthy enough to be given mortgages at the risk free rate. Everyone in zip code B also applies for a mortgage loan, but only 50% of households have their applications accepted. The remaining 50% are rejected due to a poor credit record. Then suppose there is an outward shift in the supply of capital such that lenders are willing to lend to riskier households. Then, *ceteris*

paribus, zip code B that has high initial latent demand (i.e., unfulfilled demand) will experience a sharp growth in credit, while zip code A will experience no change.²

The key identifying assumption in the example above is that initial 1996 latent demand for loans in a zip code is not positively correlated with subsequent improvements in credit quality. For example, if households in zip code B experience a greater increase in economic opportunities than those in A, then credit growth in B may be attributed to the increase in credit quality rather than an increase in supply. However, we show that in our context zip codes with higher initial latent demand for mortgages experienced *negative* relative income and employment growth from 2001 to 2005.

We implement this empirical strategy by first showing evidence of a shift in the supply of mortgage credit toward high 1996 latent demand zip codes from 2001 to 2005. We demonstrate that high 1996 latent demand zip codes experience a dramatic relative reduction in mortgage denial rates during this time period. Simultaneous with the reduction in denial rates, high 1996 latent demand zip codes experience much larger increases in the debt to income ratio of accepted mortgage applications. The source of this increase in credit availability is disintermediation: the fraction of mortgages sold by originators in the secondary market experiences a sharp relative increase for high latent demand zip codes from 2001 to 2005. Finally, the interest spread between mortgages to low credit quality borrowers and high quality borrowers narrows to historical lows during this period. Taken together, these facts demonstrate a sharp relative increase in the supply of mortgages to high 1996 latent demand zip codes from 2001 to 2005.

We then show the effects of the credit expansion on originated mortgage amounts, house price appreciation, and subsequent defaults. We find that high 1996 latent demand zip codes experience a large relative increase in both mortgage debt and house prices from 2001 to 2005.

² In Section III, we formalize this intuition in a simple Stiglitz and Weiss (1981) inspired model.

The relative increase in house prices in high 1996 latent demand zip codes occurs despite the fact that these zip codes experience *negative* relative income and employment growth over this time period. Our findings suggest that expansion in the supply of mortgage credit is a primary cause of house price appreciation in high latent demand zip codes from 2001 to 2005.

The expansion in credit to high latent demand zip codes is followed by a large increase in default rates. In terms of magnitudes, a one standard deviation increase in “supply-driven” mortgage debt from 2001 to 2005 leads to a one-half standard deviation increase in mortgage default rates from 2005 to 2007. Furthermore, a one standard deviation increase in supply-driven house price appreciation leads to a two standard deviation increase in mortgage default rates.

Our findings demonstrate that the expansion in the supply of credit driven by disintermediation is responsible for the rapid increase in new loan originations, house price appreciation, and subsequent large increase in default rates. By allowing mortgage originators to shed credit risk by selling loans, disintermediation significantly increased the amount of lending to riskier borrowers. We directly link the disintermediation process to credit expansion, house price appreciation, and ultimate defaults by showing that these changes take place in precisely those zip codes that experienced the greatest increase in disintermediation. For example, credit growth from 2001 to 2005 and growth in default rates from 2005 to 2007 is significantly higher for zip codes with larger increases in disintermediation.

Furthermore, the positive relation between disintermediation and subsequent defaults is concentrated in zip codes in which a larger fraction of loans were sold by originators to unaffiliated, non-commercial bank institutions. In other words, disintermediation only leads to higher default rates when originator incentives are less aligned with buyers and when the buying

institution has no specialized screening skills. Taken together, these findings suggest that moral hazard on behalf of originators is a primary culprit for the default crisis.

Research presented here is related to recent working papers examining the rise in default rates on subprime mortgages (Keys, Mukherjee, Seru and Vig (2008), Demyanyk and Van Hemert (2007), and Doms, Furlong, and Krainer (2007)). Among these papers, the closest to ours is Keys, Mukherjee, Seru and Vig (2008), who exploit credit score threshold rules used for securitization to show that securitization leads to more defaults. Our work is also related to an earlier strand of literature that examines the relation between housing price changes and consumer borrowing (Poterba (1984), Case and Shiller (1989), Stein (1995), Genesove and Mayer (1997, 2001), Hurst and Stafford (2004), Glaeser and Gyourko (2005), Himmelberg, Mayer, and Sinai (2005), Brunnermeier and Julliard (2007)).

This paper makes a novel contribution to this literature on several dimensions. First, and most important, we believe that our analysis is the first to demonstrate the causal effect of mortgage credit expansion driven by disintermediation from 2001 to 2005 on house price appreciation and subsequent defaults. We are able to demonstrate causality given our unique empirical strategy that exploits within-county variation across zip codes in initial latent demand for mortgages. In addition, our zip code data on income, employment, and crime allow us to perform tests that mitigate concerns that omitted credit quality variables are polluting the estimates of the supply effect.

Second, the data set we employ covers all major geographic areas of the United States. As a result, we can utilize the microeconomic estimates to examine the macroeconomic effects of supply expansion. Our conservative estimates based on the difference in difference within-county estimator suggests that at least 15% of home purchase originations, and 10% of house

price appreciation from 2001 to 2005 can be attributed to credit supply expansion (see Section V for details).

Third, the causal link we establish between credit supply, house prices, and subsequent financial crisis is not an isolated incident. Reinhart and Rogoff (2008) document that a liquidity and asset price boom followed by financial collapse and economic slowdown is a trademark of crises in many developed countries, including Japan, Spain, UK, and Norway. Our findings provide important microeconomic evidence on this broader phenomenon by documenting the precise channel through which mortgage credit expansion in the U.S. caused a rapid appreciation in house prices and a subsequent increase in mortgage defaults.

The rest of the paper proceeds as follows. The next section describes the data and presents the summary statistics. Section II presents evidence of an aggregate shift in the supply of mortgage credit from 2001 to 2005. Section III presents the empirical methodology, Section IV presents the core results of our analysis, and Section V concludes.

I. Data and Summary Statistics

Our empirical analysis employs a unique zip code-year level panel data set with information on outstanding consumer debt, consumer debt defaults, house prices, mortgage terms, and demographic variables. There are three main data sources that comprise our final data set: Equifax data on consumer credit, Fiserv Case Shiller Weiss data on house prices, and Home Mortgage Disclosure Act (HMDA) data on loan terms. In addition, we obtain demographic information from the Decennial 2000 Census, the Census Statistics of U.S. Businesses, the Internal Revenue Service, and CAP Index Inc. Our final data set includes 2,920 zip codes from 1996 through 2007, and these zip codes represent more than 40% of outstanding consumer debt

in the United States. In this section, we explain the construction of the final data set and present summary statistics.

A. Equifax Predictive Services Data

We collect data on outstanding consumer credit amounts and defaults from Equifax Predictive Services. Equifax is a consumer credit rating agency that collects, organizes, and manages credit information for U.S. consumers. The original Equifax data have credit information for almost 170 million individuals going as far back as 1990. However, due to cost and confidentiality concerns, Equifax aggregated the provided data at the zip code level at a quarterly frequency from 1998Q1 to 2007Q2, and at an annual frequency from 1991 to 1997.

We therefore have aggregate debt composition and defaults of every U.S. zip code at a quarterly frequency from 1998 through the second quarter of 2007, and at an annual frequency from 1991 through 1997. The outstanding consumer credit and delinquency data at the zip code level are broken down by type of consumer loans (credit cards, mortgages, home equity lines, auto loans, etc.), as well as consumer credit score.

We construct aggregate measures of mortgage debt, home equity debt, and non-home consumer debt. The latter category aggregates credit card debt, consumer loans, student loans, and auto loans. The Equifax data record default amounts for the following varying degrees of default: 30 days late, 60 days late, 90 days late, 120 days late or collections, severe derogatory, and bankruptcy. In the analysis below, we define defaults as broadly as possible: default amounts include any amounts 30 days late or more. The main reason for this choice is that many defaults are recent. All of our specifications are run in first-differences, which mitigates concerns about the average level of default for different default categories. Our results are materially unchanged if we use an alternative definition of default such as 60 plus days late or more.

B. Home Mortgage Disclosure Act (HMDA) Data

While the Equifax data provide us a comprehensive picture of the stock of consumer credit at the zip code level, they do not provide information on the flow of new mortgage and home equity loans being originated. We therefore collect this information from loan origination data sets collected under the “Home Mortgage Disclosure Act” (HMDA).

In order to supervise and enforce fair lending practices across that U.S., the U.S. Congress mandates that all loans applications related to home purchase, refinancing, and home improvement be reported to the federal government. The loan application information is publicly available through HMDA from 1996 through 2006. For every loan application, the public data record its status (denied / approved / originated), purpose (home purchase / refinancing / home improvement), loan amount, and applicant characteristics including race, sex, income and home ownership status. It also reports lender information, including the lender’s reasons for applicant denial, type of lender, and whether the loan originator sold the loan to the secondary market within a year. Since 2004, HMDA has also recorded the initial interest rate spread of loan originations, and lien status. HMDA does not provide information on the maturity structure of a loan, or whether the loan has a fixed rate mortgage or ARM. Nonetheless, with millions of loan applications recorded every year, HMDA remains one of the best sources for understanding loan origination patterns.

Since our unit of analysis is a zip code, we aggregate the application-level HMDA data to census tracts, which are the smallest available geographical identifiers in the data. The census tract level HMDA data are then aggregated into zip codes using the census tract to zip code match provided by *Geolytics*. Census tracts are smaller than zip codes on average, with about 60,000 census tracts for approximately 40,000 zip codes. Consequently the quality of the match

from census tract to zip code is excellent. For example, 85% of matched census tracts in our final sample have over 90% of their population living in the zip code to which they are matched. The intersection of HMDA and Equifax data contains 19,368 zip codes.

C. Fiserv Case Shiller Weiss Data

Our primary data source for zip code level house price indices is Fiserv Case Shiller Weiss. FCSW uses same house repeat sales data to construct house price indices at the zip code level. The zip code level house price data we utilize in this study underlies the MSA level S&P/Case Shiller indices, upon which futures are traded on the Chicago Mercantile Exchange. The data set includes house price indices through the second quarter of 2007.

One limitation of the data is that FCSW require a significant number of transactions in a given zip code to obtain reliable estimates of changes in house prices over time. As a result, FCSW has house prices for only 3,056 of the zip codes in Equifax-HMDA sample. While FCSW covers only 15% of the number of zip codes in the Equifax-Census sample, their indices are constructed for all major metropolitan and highly populated zip codes in the United States. As a result, our final sample includes over 40% of the aggregate outstanding amounts of consumer debt, and almost 45% of the aggregate home debt outstanding.

The Appendix Table compares the sub-samples based on the house price index restriction, and shows that the primary difference is the fraction of households in urban areas. The average fraction of households in an urban area for our 3,056 zip codes is 92%, whereas the average fraction of households in an urban for the other 16,312 zip codes for which we do not have house price indices is only 46%. Our analysis is therefore concentrated on densely populated urban zip codes in the United States. While there are other differences between the zip codes included in our final sample and those excluded given the lack of house price data, they

are minor. In addition, all of our core results that do not require house price data are similar in direction and magnitude if we use the full sample of 19,368 zip codes.

As a robustness check of FCSW price indices, we also collect zip code level price indices for 2,248 zip codes from Zillow.com, an online firm that provides house price data for potential buyers and sellers. House price changes from FCSW and Zillow have a correlation coefficient of 0.91, and all of our core results are robust to the use of price indices from Zillow instead of FCSW.

D. Census, Business Statistics, IRS Data, and CapIndex crime statistics

Zip code level demographic attributes such as population, race, poverty, mobility, unemployment and education come from the Decennial 2000 Census. We also collect annual measures of business opportunities available in a given zip code through the Business Statistics published by the U.S. Census Bureau. These statistics provide data on wages, employment, and number of establishments at the zip code level. Given a three year lag in the reporting of information, the business opportunity data are available from 1996 through 2004. Finally, we collect zip code level average “adjusted gross income” as reported by the IRS. The IRS currently provides these data for 1998, 2001, 2002, 2004 and 2005. The income variable from the IRS is important because it tracks the income of consumers living inside a given zip code, as opposed to Business Statistics which provide wage and employment statistics for individuals working, but not necessarily living, in a zip code.

Since a potentially important neighborhood determinant of house prices and credit market conditions is crime, we also collect zip level statistics on total crime from 2000 to 2007. These data are from CAP Index, Inc., a firm specializing in providing crime data.³

E. Summary Statistics

³ See Garmaise and Moskowitz (2006) for more information on CAP Index crime data.

Table 1 presents summary statistics for the final sample of 2,920 zip codes, for which we have data available at an annual frequency from 1996 to 2007. All variables are measured as of the fourth quarter of each year, except for 2007 for which variables are measured as of the end of the second quarter. Mortgage debt represents over 74% of consumer debt in our sample, and the annualized growth in mortgage debt outstanding is 10.2% from 2001 to 2005. In contrast, the annualized growth in non-home debt outstanding is only 4.6%. The average default rate on mortgages in 1996 is 3%. While there is little change in the default rate from 1996 to 2005, there is an increase of 1.7% in the mortgage default rate from 2005 to 2007, which represents almost a three-quarter standard deviation and 50% of the mean. While the default rate on non-home debt also increases from 2005 to 2007, the increase is smaller.

House price growth is strong in our sample period, with house prices growing by an annualized rate of 7.3% from 1996 to 2001 and 11.3% from 2001 to 2005. HMDA data on the growth in amounts originated are consistent with the growth in outstanding mortgage debt from Equifax: originations for home purchase grow at an annualized rate of 13.4% from 2001 to 2005. The denial rate for mortgages in 1996 is 22%. There is a dramatic rise in the fraction of mortgages sold to non-mortgage agency investors from 2001 to 2005: the fraction of these mortgages increases by 26 percentage points.⁴

II. Evidence of an Aggregate Supply Shift in Mortgage Credit

Figure 1B and Table 1 demonstrate an increase in outstanding mortgage debt from 2001 to 2005 that is almost 30% larger than the expansion in non-home debt over the same time period. Figure 1C shows suggest that the increase in mortgage debt is not entirely driven by

⁴ By “non-mortgage agency investors”, we mean investors *other than* Freddie Mac, Fannie Mae, Federal Farmers Home Administration, and Ginnie Mae.

larger mortgages: it shows a larger increase in number of accepted mortgage applications for home purchase from 2003 to 2005 than in any other period in our sample. Taken together, these figures demonstrate a sharp rise in mortgage credit on both intensive and extensive margins from 2001 to 2005.

The increase in the quantity of credit is also associated with an increase in the observable riskiness of mortgage credit. Figure 2A maps the median, 75th percentile, and 90th percentile of mortgage debt to income ratios of accepted mortgage applications.⁵ There is a slight upward trend in the ratios from 1996 through 2000. However, the increase in mortgage debt to income ratios from 2001 to 2005 is much larger. The mortgage debt to income ratio of borrowers in the 90th percentile increases by 1 unit over this time period. This change represents a remarkable two standard deviation increase in the mortgage debt to income ratio at the 90th percentile.

In Figure 2B, we examine the aggregate mortgage debt to income ratio of the entire zip code, as opposed to the mortgage debt to income ratio of accepted mortgage applications examined in Figure 2A. Aggregate mortgage debt originated for home purchase comes from the HMDA data in a given year, and it is scaled by the aggregate zip code income reported to the IRS. We utilize this measure as an alternative measure of the risk profile of zip codes given potential fraud associated with income reporting on mortgage applications. The drawback of this measure is that income data from the IRS is available only for 1998, 2001, 2002, 2004, and 2005. Figure 2B shows a very similar pattern to Figure 2A. The total increase in mortgage debt to income ratios from 2001 through 2005 is 0.12, which represents a two-thirds standard deviation increase in the mortgage debt to income ratio.

The sharp rise in debt to income ratios during the credit expansion highlights the willingness of mortgage originators to take on increasing levels of risk. One concern is that the

⁵ The percentiles are computed from accepted mortgage applications in a zip code using the HMDA data set.

increase in debt to income ratios is compensated by a concurrent decrease in debt to value ratios as house prices increase. However, this is not the case. While we do not have mortgage level home value data, Demyanyk and Van Hemert (2007) utilize such data and show that debt to value ratios also increased from 2001 to 2005.

A key remaining question is whether investors holding the new riskier mortgage securities were compensated for the greater risk through higher interest rates. While we do not have data on mortgage level interest rates, Chomsisengphet and Pennington-Cross (2006) show that the subprime-prime mortgage spread for 30-year fixed mortgages dropped from 225 basis points to historical lows of 175 basis points from 2001 to 2004. Their calculation uses 30-year fixed mortgage rates, which mitigates concern that the decline in subprime-prime mortgage spreads is due to low teaser rates on subprime loans at origination. Demyanyk and Van Hemert (2007) reach a similar conclusion using a different data set (see their Figure 8).

What precipitated such a dramatic increase in the riskiness and quantity of mortgages during this time period? It is difficult to reconcile the evidence solely with changes in demand for credit, particularly given that the subprime mortgage interest rates declined sharply despite increases in quantity and risk. Figure 3 demonstrates one potential source of these trends. It displays the dramatic rise in disintermediation, or the fraction of mortgages originated that are sold to non-mortgage agency investors in the secondary market from 2001 to 2005. The fraction of these mortgages is relatively constant up to 2001, and then increases by 25 percentage points from 2001 to 2005.

Two fundamental macroeconomic factors are likely to have played an important role in pushing the securitization wave. First, 2001-2005 was a period of very low interest rates and high liquidity in the United States. Despite the low interest rates, there was a flood of

international liquidity coming into the U.S. from the middle east, China, and India – a phenomena referred to as “macro imbalances” in the international finance literature. Second, innovations in the financial sector led to the creation of the collateralized debt obligation (CDO) market that allowed risky mortgages to be pooled together and sold off in tranches of varying seniority. Since the senior tranches were often given a very highly rated (e.g. AAA), these new mortgage backed securities satisfied the rating criteria of institutional capital.

III. Empirical Methodology

Section II demonstrates an outward shift in the supply of mortgage credit between 2001 and 2005 driven by disintermediation. In this section we develop an empirical methodology that permits us to isolate the causal impact of the supply shift on mortgage originations, house prices, and subsequent defaults.

A. Empirical Model

Our empirical methodology for isolating the supply channel is based on the premise that if one were to observe the latent (i.e. unfulfilled) loan demand, then *ceteris paribus* an expansion in credit supply should lead to higher credit uptake in areas with greater initial latent demand. The key identifying assumption, and one we test rigorously, is that areas with greater initial unfulfilled credit demand do not experience larger positive income, credit quality, or economic opportunity shocks in the future.

We illustrate our empirical methodology in the following parsimonious model of mortgage originations for home purchases. Consider customers living in zip code z in county c at time t . In every period customers of measure one are interested in purchasing a new home that requires one unit of capital. For simplicity, we assume that a qualified customer takes the

mortgage this period, and promises to completely pay off principal and interest next period. We define customers as “Prime” if their income profile exceeds a certain threshold such that there is no possibility of default next period. As a result, all lenders are willing to lend to Prime customers at the risk free rate normalized to 1. We denote the fraction of prime customers in a zip code by $f_{zt}(I_{zt})$, with the argument I_{zt} reminding us that f_{zt} depends on the overall income distribution within a zip code.

We define customers with income profiles below the Prime threshold as “Sub-Prime”. What distinguishes Sub-Prime customers is that they have a positive probability, p , of default if their realized income next period is sufficiently low. Sub-Prime customers have different individual income profiles, and can therefore differ in their probability of default, p . We assume the mortgage market is competitive at the national level, and that lenders recover nothing in case of default. At each t , the interest rate offered to a Sub-Prime customer is given by:

$$r = \frac{1}{1-p} + \theta \text{ if } \frac{1}{1-p} + \theta \leq \bar{r} \quad (1)$$

$$r = \infty \text{ otherwise}$$

In (1), θ reflects the “risk premium” that the market charges for bearing the probability of default, and \bar{r} is an interest rate ceiling above which no lender is willing to lend. We do not model explicitly the underlying friction that leads to an interest rate ceiling above which originators are unwilling to lend—borrower moral hazard (Diamond (1991), Holmstrom and Tirole (1997)) or adverse selection (Stiglitz and Weiss (1981)) are potential reasons.⁶

The net result of equation (1) is that only a fraction g_{zt} of Sub-Prime customers in each period t obtain mortgages. The fraction g_{zt} depends on the market risk premium (θ_t) and distribution of p among Sub-Prime customers, which in turn is a function of the overall income

⁶ Gabriel and Rosenthal (2007) explicitly model how a supply expansion affects borrowers with a Stiglitz and Weiss (1981) adverse selection problem. Their conclusions are similar to ours.

distribution I_{zt} in the population.⁷ We can therefore write g_{zt} as $g_{zt}(\theta_t, I_{zt})$, with $g_\theta < 0$ and $g_I > 0$.

The preceding discussion gives us the equilibrium determination of mortgage originations in zip code z at time t (L_{zt}) as:

$$L_{zt} = f_{zt} + (1 - f_{zt}) * g_{zt} \quad (2)$$

We have suppressed arguments of f and g for notational simplicity. Allowing for other possible factors affecting L_{zt} , yields:

$$L_{zt} = f_{zt} + (1 - f_{zt}) * g_{zt} + \alpha_z + \alpha_{ct} + \varepsilon_{zt} \quad (3)$$

In (3), α_z reflects time-invariant determinants of loan origination for a given zip code, α_{ct} reflects time-varying county-level factors affecting loan originations in a given zip code, and ε_{zt} is an unobserved error term.

The fundamental economic drivers of equilibrium loan originations in equation (3) are income factors, which are summarized by income distribution I_{zt} , and credit supply factors, which are summarized by the mortgage risk premium θ_t . The challenge of our empirical methodology is to isolate the effect of changes in supply factors on loan originations while controlling for income factors. Since equation (3) includes county interacted with time fixed effects, any changes in income that are common across zip codes in the same county are non-parametrically removed. For example, an economic boom in a given county is absorbed by α_{ct} .

In order to clarify the identifying assumption we make to isolate the supply channel, we first make the (strong) assumption that *all* variation in income factors occurs at the county level. Given that there is no residual time variation left in f_{zt} (which does not depend on the risk premium), we can replace it by the initial fraction of Prime customers in a zip code, f_{z0} . Since we are interested in shocks to loan originations, we first-difference equation (3) and suppress time

⁷ Solving explicitly, g_{zt} is the subset of Sub-Prime customers with $p \leq \frac{\bar{r} - \theta_t - 1}{\bar{r} - \theta_t}$.

subscripts for simplicity. Therefore, under the assumption that income factors only vary at the level of the county, first-differencing equation (3) gives us:

$$\Delta L_{zt} = \beta * (1 - f_{z0}) + \alpha_c + \Delta \varepsilon_z \quad (4)$$

where $\beta = \Delta g_z$, which depends only on the credit supply shock θ . A negative θ reflects a reduction in market risk premium and hence a positive credit supply shock. A positive credit supply shock would lead to more Sub-Prime consumers obtaining mortgages and hence a positive β . In other words, $\hat{\beta}$ identifies the impact of a credit supply shock on L_{zt} under the identifying assumption that all income shocks occur at the county level.

We are now in a position to relax our identifying assumption further. Income shocks may be zip code specific, but as long as they are orthogonal to the initial latent demand conditions ($I-f_{z0}$), $\hat{\beta}$ retains its interpretation. A natural corollary is that if zip code specific income shocks are *negatively* correlated with the initial fraction of Sub-Prime customers, then our interpretation of $\hat{\beta}$ as a credit supply coefficient is still accurate, but the magnitude is an under-estimate of the true supply effect. More specifically, if areas with a higher fraction of Sub-Prime customers have negative future income shocks, then $\hat{\beta}$ will understate the effect of credit supply on originations. As we show below, the fraction of Sub-Prime customers at the beginning of our sample is negatively correlated with observable measures of future income shocks. This negative correlation strengthens our identifying assumption that future income shocks are not positively correlated with the initial fraction of Sub-Prime customers, and further suggests that our estimates may *understate* the effect of credit expansion on outcomes.

Equation (4) represents our primary regression specification. In order to estimate this equation, our data provides us with many possible measures of initial latent demand conditions, or equivalently “Sub-Prime” customers ($I-f_{z0}$). For example, through the HMDA data we observe

the fraction of home loan applications in a given zip code that are denied in 1996. We also observe the fraction of customers who have a credit score of under 660 or under 620 in 1996, the fraction of home loans that are guaranteed by FHA under a “low income housing” initiative in 1996, and the fraction of households who are renters as of 2000. We use fraction of loan applications denied in 1996 as our main measure of high latent demand zip codes, and show that all of our core results are robust to alternative definitions as well.

B. Characteristics of High 1996 Latent Demand Zip Codes

As equation (4) demonstrates, initial latent demand ($1-f_{z0}$) is of primary importance in our empirical analysis. Table 2 presents the correlation of our main measure of latent demand in a zip code, the fraction of mortgage applications denied in 1996, with other variables. This measure is strongly correlated with alternative measures of high latent demand/high credit risk, such as the fraction of subprime borrowers, the fraction of loans backed by FHA, or the fraction of housing units rented. It is also strongly correlated with poverty and unemployment, and negatively correlated with household income measures.

The bottom panel of Table 2 demonstrates that measures of future growth in economic opportunities are *negatively* correlated with the fraction of 1996 mortgage applications denied. As mentioned above, the critical identifying assumption of our empirical methodology is that areas with high initial latent demand do not experience subsequent increases in income, credit quality, or economic opportunity. The correlations in Table 2 strongly support the identifying assumption, given that observable measures of future growth in economic opportunity are in fact *negatively* correlated with our primary measure of high 1996 latent demand. This also suggests that our estimated $\hat{\beta}$ is an under-estimate of the true coefficient on supply shock β .

IV. Credit Expansion, Mortgage Debt, Housing Prices, and Defaults

In this section, we implement the empirical methodology discussed above. Our goal is to estimate the causal effect of mortgage credit expansion (due to disintermediation) on growth in mortgage debt, house prices, and mortgage defaults.

A. Credit Expansion to High 1996 Latent Demand Zip Codes

We first demonstrate that high 1996 latent demand zip codes experience a relative increase in credit supply to riskier borrowers from 2001 to 2005. In Figures 4 through 7, we plot coefficient estimates from a year-by-year set of county fixed effects regressions of the following general specification:

$$y_{zct} - y_{zc,1996} = \alpha_c + \beta_t * HighLatentDemand_{zc,1996} + \varepsilon_{zct}, \quad (5)$$

$$for\ t = 1997, 1998, \dots, 2007$$

In other words, for each year t from 1997 to 2007, we estimate a first-difference county fixed effects specification relating the change in outcome y for zip code z in county c from year 1996 to year t to our primary measure of high 1996 latent demand, which is the fraction of 1996 mortgage applications denied in the zip code. We plot the coefficient estimates of β_t for each year t , along with the corresponding 95% confidence interval. The plotted coefficient estimates represent the differential effect on the change in outcome y from 1996 to t for high latent demand zip codes, after controlling for county fixed effects (α_c). This methodology allows us to show the differential effect of mortgage credit expansion on high latent demand zip codes throughout our sample period, instead of taking a particular stand on exactly the time period over which expansion occurs. The inclusion of county fixed effects in the first-differenced specification also takes out all possible shocks affecting the outcome of interest at the county level.

Figure 4 examines the differential pattern of denial rates, debt to income ratios, and disintermediation for high 1996 latent demand zip codes. There is a dramatic differential decrease in denial rates for high 1996 latent demand zip codes beginning in 2001 and lasting through 2006. The coefficient estimate for 2004 implies that a one standard deviation increase in 1996 latent demand (0.08) leads to a reduction in the denial rate of 2 percentage points from 1996 to 2004, which is a one-third standard deviation of the left hand side variable. Figure 4B shows a corresponding increase in the average debt to income ratio of high 1996 latent demand zip codes. Beginning in 2002, there is a sharp relative increase in the mortgage debt to income ratio of high 1996 latent demand zip codes. The coefficient estimate for 2005 implies that a one standard deviation increase in 1996 latent demand leads to a one-third standard deviation increase in mortgage debt to income ratios from 1998 to 2005.

The reduction in denial rates and increases in mortgage debt to income ratios for high 1996 latent demand zip codes demonstrates a strong shift in the supply of mortgage credit to a riskier consumer base from 2001 to 2005. Figure 4C shows the source of this credit expansion. Beginning in 2001, there is a sharp relative rise in disintermediation for high 1996 latent demand zip codes. The coefficient estimate for 2006 implies that a one standard deviation increase in 1996 latent demand leads to a 2.4 percentage point increase in the fraction of mortgages disintermediated from 1996 to 2006, which is more than a one-third standard deviation of the left hand side variable⁸.

B. The Effect of Credit Expansion on Mortgage Debt and Housing Prices

Figure 4 demonstrates that zip codes for which there is a high fraction of mortgage applications denied in 1996 experience a large relative increase originators' willingness to supply

⁸ There is a "bump" in Figure 4C prior to 2001. However, since the aggregate level of disintermediation does not increase during the pre-2001 period (Figure 3), this bump is unlikely to have much of an effect in the aggregate.

credit from 2001 to 2005. This trend appears to be driven in large part due to an increase in the disintermediation of mortgages and a consequent increase in the risk tolerance of originators. In this section, we examine the differential effect of the increase in mortgage supply on mortgage debt and housing prices in high 1996 latent demand zip codes.

Figure 5A plots coefficient estimates of the differential growth in the number of accepted mortgages for home purchase for high 1996 latent demand zip codes. From 2002 to 2005, the coefficient estimate on high 1996 latent demand increases by 300%. The coefficient estimate for 2005 implies that a one standard deviation increase in 1996 latent demand (0.08) leads to a relative increase in the growth rate of accepted mortgage applications from 1996 to 2005 of 24%, which is a one-half standard deviation of the left hand side variable. There is a slight increase from 1998 to 2000, but this increase is less than half the increase from 2002 to 2005. The evidence suggests that high 1996 latent demand zip codes experience dramatic growth in the number of new homeowners from 2002 to 2005 as a result of the expansion in mortgage supply.

Figure 5B examines the relative growth in mortgage debt outstanding of high 1996 latent demand zip codes. The figure demonstrates that the sensitivity of mortgage debt growth in a zip code to high 1996 latent demand increases from 1999 through 2007. The coefficient estimate for 2007 implies that a one standard deviation increase in 1996 latent demand leads to a relative increase in the growth rate of mortgage debt outstanding from 1996 through 2007 of 4 percentage points, which is one-eighth of a standard deviation of the left hand side variable.⁹

⁹ The estimates in Figure 5B are relatively imprecise and small in magnitude compared to other estimates of mortgage growth because the Equifax measure of mortgage debt used in the figure does not differentiate mortgage debt for new home purchase versus mortgage debt obtained through refinancing. This is important because high 1996 latent demand zip codes do not refinance as aggressively in response to declining interest rates as low 1996 latent demand zip codes (something we confirm in the HMDA data that separates originations for refinancing versus home purchase). Interest rates declined sharply from 2001 to 2003, and so there was a relative *decrease* in mortgage debt due to refinancing for high 1996 latent demand zip codes from 2001 to 2003 even though there was a strong relative *increase* in mortgage debt for new home purchase for high latent demand zip codes over this time period.

Figure 5C shows a sharp relative increase from 2002 to 2006 in the volume of home purchase loan originations for high 1996 latent demand zip codes. The coefficient estimate for 2006 implies that a one standard deviation change in 1996 latent demand leads to a relative increase in the growth rate of originated mortgage amounts for home purchase of 28%, which is one-half standard deviation of the left hand side variable. As in the number of mortgage applications, there is a relative increase in the growth rate for high 1996 latent demand zip codes from 1998 to 2000, but the increase from 2002 to 2006 is significantly larger.

Figure 6 demonstrates the effect of increased supply on house price appreciation. Zip codes with high 1996 latent demand do not experience any higher growth in house prices from 1996 to 1998. However, as credit supply starts to expand disproportionately more in high latent demand zip codes in 1999 (Figure 5A), they start to experience a relative increase in house price appreciation. The relative growth in house price appreciation accelerates from 2001 onward. The coefficient estimate for 2000 implies that a one standard deviation increase in 1996 latent demand leads to a relative increase in house price appreciation from 1996 to 2000 of 0.8%, which is less than a one-fifteenth standard deviation in house price appreciation. The coefficient estimate for 2006 implies that a one standard deviation increase in 1996 latent demand leads to a relative increase in house price appreciation from 1996 to 2006 of almost 6%, which is one-third of a standard deviation. It is important to emphasize that the relative increase in housing prices for high latent demand zip codes occurs despite relative negative income and employment growth during this time period. The evidence suggests that a significant fraction of relative house price appreciation from 2001 to 2006 in high 1996 latent demand zip codes is a direct result of the expansion of mortgage credit supply.

As discussed in Section III, the key identifying assumption of our empirical methodology is that high 1996 latent demand zip codes do not experience future increases in economic opportunities or income shocks. Table 2 strongly supports this identifying assumption, as it documents that observable measures of future growth in economic opportunities are negatively correlated with our measure of high 1996 latent demand. Nonetheless, in Table 3, we control directly for changes in economic opportunity. More specifically, Table 3 presents coefficient estimates from the following first difference county fixed effects specification:

$$y_{zct2005} - y_{zct2001} = \alpha_c + \beta * HighLatentDemand_{zc,1996} + \gamma * (X_{zct2005} - X_{zct2001}) + \varepsilon_{zct} \quad (6)$$

In (6), X represents a matrix of control variables for economic opportunities including income growth, wage growth, establishment growth, and employment growth. We also include zip code level growth in crime rate as measured by the CAP Index crime index to account for variation in neighborhood safety. We choose the period 2001 to 2005 for the regressions given the evidence from Figures 4 through 6 that this is the main period over which supply expansion occurs. Minor variations of this time frame do not affect the results.

Panel A of Table 3 presents the estimates. The coefficient estimate in column 1 implies that a one standard deviation change in 1996 latent demand leads to a 17% higher growth rate in originated mortgage amounts for home purchase from 2001 to 2005, which is a one-third standard deviation of the left hand side variable. The coefficient estimate in column 2 implies that a one standard deviation change in 1996 latent demand leads to a 3% higher growth rate in mortgage debt outstanding, which is a one-eighth standard deviation of the left hand side variable. Finally, the coefficient estimate in column 3 implies that a one standard deviation change in 1996 latent demand leads to a 3% higher growth rate in house prices, which is a one-eighth standard deviation of the left hand side variable. The results in Table 3 suggest that magnitudes are similar after controlling for changes in economic opportunities.

In Panel B of Table 3, we examine the differential effect of credit expansion on mortgage debt and house prices using a more direct measure of relative supply shifts. More specifically, Panel B presents coefficient estimates from the following specification:

$$y_{zc2005} - y_{zc2001} = \alpha_c + \beta(\text{Denial Rate}_{zc2005} - \text{Denial Rate}_{zc2001}) + \gamma(X_{zc2005} - X_{zc2001}) + \varepsilon_{zc} \quad (7)$$

In other words, we estimate a first difference county fixed effects specification relating the change in outcome y for zip code z in county c from 2001 to 2005 to the *change* in denial rates from 2001 to 2005. We utilize the change in denial rates over this time period to measure the change in the willingness of lenders to supply mortgage credit to a given zip code. The estimates in Panel B are largely consistent with the estimates in Panel A. Zip codes that experience a reduction in denial rates from 2001 to 2005 experience an increase in mortgage amounts originated for home purchase, mortgage debt outstanding, and house price appreciation.

The evidence in this sub-section demonstrates that increased risk tolerance on behalf of originators led to a sharp increase in the supply of mortgage credit high 1996 latent demand zip codes. As a result, these zip codes experienced large increases in originated mortgage amounts and house prices from 2001 to 2005. Our results suggest that a substantial fraction of relative house price appreciation and originated mortgage amounts in high 1996 latent demand zip codes *would not have occurred* in the absence of the expansion in mortgage credit availability.

C. The Effect of Credit Expansion on Default Rates

The results above suggest that a shift in the supply of mortgage credit caused a significant increase in mortgage debt outstanding and house prices for zip codes with high 1996 latent demand for mortgages. In this section, we explore the effect of credit expansion on mortgage default rates.

Figure 7A replicates the methodology of Figures 4 through 6 with the change in default rates as the left hand side variable. Figure 7A shows that there is a relative increase in mortgage

defaults in 2001 for high 1996 latent demand zip codes. This relative increase during a recession year is consistent with the results in Table 2 that show that high 1996 latent demand zip codes are lower income, higher unemployment areas. However, it is important to emphasize that the 2001 increase in defaults is not preceded by a relative decrease in denial rates (Figure 4A) or an increase in aggregate mortgage debt to income ratios (Figure 4B). In other words, there is less evidence that the relative increase in default rates for high 1996 latent demand zip codes in 2001 is due to an expansion of credit supply. Instead, it appears to be driven primarily by the economic slowdown.

Figure 7A also shows a sharp increase in default rates for high 1996 latent demand zip codes from 2005 to the second quarter of 2007. In contrast to the relative increase in defaults in 2001, this increase is not preceded by or concurrent with a recession. Instead, the increase in defaults for high 1996 latent demand zip codes is preceded by a decline in denial rates (Figure 4A) and an increase in aggregate mortgage debt to income ratios (Figure 4B). These facts suggest that the relative increase in default rates from 2005 to 2007 is due to the rapid expansion in the supply of credit from 2001 to 2005. In terms of magnitudes, a one standard deviation increase in 1996 latent demand for a zip code leads to a 1 percentage point increase in default rates from 1996 to 2007, which is one-third standard deviation of the left hand side variable.

In Figure 7B, we examine the differential effect of credit expansion on mortgage default rates by using a time-varying measure of relative supply shifts. More specifically, Figure 7B plots coefficient estimates from a year-by-year set of county fixed effects regressions of the following general specification:

$$Default Rate_{zct} - Default Rate_{zct-2} = \alpha_c + \beta_t * (Denial Rate_{zct-3} - Denial Rate_{zct-5}) + \varepsilon_{zct} \quad (8)$$

for t = 2001, 2002, ..., 2007

In other words, for each year t from 2001 to 2007, we estimate a first difference county fixed effects specification relating the change in default rates for zip code z in county c from year $t-2$ to year t to the *change* in denial rates from $t-5$ to $t-3$. For example, for $t = 2007$, the specification estimates whether default rates from 2005 to 2007 are correlated with changes in denial rates from 2002 to 2004. We utilize the change in denial rates as a measure of the change in the willingness of lenders to supply mortgage credit to a given zip code.

Figure 7B shows the year-by-year estimates. It demonstrates that the *only* year in our sample for which changes in default rates are *negatively* correlated with lagged changes in denial rates is 2007. In other words, zip codes that experienced a *reduction* in denial rates from 2002 to 2004 experience an *increase* in default rates from 2005 to 2007.

In columns 1 and 2 of Table 4, we examine a reduced form specification relating the change in default rates from 2005 to 2007 to 1996 latent demand and the change in denial rates from 2001 to 2005, respectively. The two right hand side variables measure zip codes that experience a relative increase in the supply of credit from 2001 to 2005; therefore, the coefficient estimates represent the reduced form effect of credit expansion on default rates. The estimate in column 1 implies that a one standard deviation increase in 1996 latent demand leads to a 0.52% increase in default rates from 2005 to 2007, which is a one-fifth standard deviation increase in the left hand side variable.

In columns 3 through 6 of Table 4, we examine the effect of supply-driven mortgage growth and house price growth from 2001 to 2005 on the change in default rates from 2005 to 2007. We define supply driven mortgage growth and house price growth as the predicted values from first stage regressions relating mortgage growth and house price growth from 2001 to 2005 to either the fraction of 1996 mortgage applications denied or the change in denial rates from

2001 to 2005. The first stage estimates used to predict supply driven mortgage growth and house price growth are in columns 1 and 3 of Table 3.

The estimates in columns 3 through 6 of Table 4 demonstrate that supply driven mortgage growth and house price growth from 2001 to 2005 have a strong effect on the increase in default rates from 2005 to 2007. The estimate in column 3 implies that a one standard deviation increase in supply driven mortgage growth from 2001 to 2005 leads to a 1.6 percentage point increase in default rates, which is more than a one-half standard deviation increase of the left hand side variable. A one standard deviation increase in supply driven house price appreciation leads to a 5 percentage point increase in default rates, which is almost a two standard deviation change in the left hand side variable.

D. Disintermediation and Moral Hazard

The evidence in the section above demonstrates that house price appreciation and debt growth driven by an increase in the supply of mortgage credit from 2001 to 2005 caused substantial increases in default rates. The evidence also suggests that investors that entered this market experienced large losses as a result of the expansion in credit. For example, our estimates imply that a 10% expansion in mortgage credit from 2001 to 2005 to high 1996 latent demand zip codes caused a 3 percentage point increase in default rates, which is a doubling of the sample average. Given that foreclosure recovery rates are typically between 40 and 70% (Pence (2006)), the increase in default rates represented significant losses for mortgage investors. In addition, the subprime-prime mortgage spread fell to historical lows during this period, which suggests that investors were not compensated for the additional ex post risk. These losses beg the question: Why did investors agree to fund these mortgages?

Figure 3 above shows that the sharp rise in mortgage growth coincides with the increase in disintermediation of mortgages. Moreover, the disintermediation is significantly stronger in zip codes with high 1996 latent demand for loans (Figure 4C). Models of financial disintermediation and moral hazard argue that financial intermediaries exerting unobservable monitoring effort must hold a significant fraction of credit risk to ensure proper incentives (Holmstrom and Tirole (1997), He and Krishnamurthy (2006)). The findings above suggest that moral hazard on behalf of loan originators that no longer hold a stake in risky loans is a potential cause of the increase in default rates.

In Table 5, we present further evidence of originator moral hazard. Column 1 in Panel A reaffirms the result shown earlier in Figure 4C: High initial latent demand zip codes experience a larger increase in fraction of loans sold off to investors within a year. Columns 2 through 6 of Panel A break down the fraction sold by the identity of the party buying the mortgage from the originating institution. The correlation between high latent demand and disintermediation is being driven by mortgages sold in private securitizations to unaffiliated investors, and to non-bank financial firms. In other words, high latent demand zip codes did not experience an increase in the fraction of mortgages sold to affiliated institutions or banks.

Column 1 of Panel B demonstrates that zip codes experiencing a relative increase in disintermediation also experience increases in mortgage debt to income ratios, which is consistent with originators shedding credit risk during the 2001 to 2005 expansion. Columns 2 through 6 show the correlation of the fraction of loans sold in a zip code from 2001 to 2005 by the type of investor buying the mortgage with subsequent default rates from 2005 to 2007. The estimates demonstrate that zip codes in which a larger fraction of mortgages are sold in private

securitizations and to non-commercial bank financial firms experience relatively larger increases in default rates from 2005 to 2007.

In contrast, column 2 of Panel B shows that zip codes in which originators sell more mortgages to affiliated investors do not experience an increase in default rates. Under the assumption that originators' incentives are more closely aligned with affiliated versus non-affiliated investors, these results suggest that undetected moral hazard is a primary cause for the higher default rates on mortgages sold to non-affiliated investors. In addition, column 5 of Table 5 demonstrates that zip codes in which originators sell more mortgages to other commercial banks do not experience an increase in default rates. Given that commercial banks have specialized skill in screening loans, these results suggest that originators only sold bad loans to unaffiliated investors lacking the skills to judge loan quality. Together with the findings in Panel A, these findings support the hypothesis that moral hazard on behalf of originators is a main culprit for the rise in default rates.

A caveat is in order. Unlike the results above that demonstrate a causal effect of mortgage credit expansion on house price appreciation and subsequent defaults, we view our evidence on moral hazard as suggestive. In other words, it is more difficult to assert that undetected moral hazard on behalf of originators caused the spike in mortgage defaults for two reasons. First, there is the lack of exogenous within-county variation across zip codes in the ability of originators to sell mortgages. Without such variation, it is difficult to rule out alternative explanations. Second, we do not have loan-level interest rate data, which makes it difficult to examine whether moral hazard is priced. Nonetheless, our suggestive findings are supported by both academic and anecdotal evidence. Keys, Mukherjee, Seru and Vig (2008) exploit cutoffs in FICO scores required for securitization of mortgages to show that securitization causes an increase in default

rates, which they interpret as evidence of bank moral hazard. They show that interest rate differentials between mortgages above and below the cutoffs are similar. In addition, as mentioned above, interest rates on subprime mortgage declined to historical lows during this time period, which suggests that investors did not price increased moral hazard concerns. Finally, anecdotal evidence on moral hazard of originators is plentiful (see Zuckerman, WSJ, January 15th, 2008; Ng and Mollenkamp, WSJ, January 14th, 2008; and Plevin and Craig, WSJ, January 17th, 2008).

E. Robustness

In this section, we present two sets of robustness tests to ensure that the estimates presented above reflect our credit expansion interpretation. The specifications reported in Table 6 examine non-home consumer debt, which includes credit card, auto, student, and consumer loans. Given that our supply interpretation is unique to the expansion in mortgage debt availability in high latent demand zip codes, we do not expect to find the same pattern in debt growth and defaults for non-home consumer debt. As Table 6 demonstrates, high 1996 latent demand zip codes experience a relative *reduction* in non-home debt from 2001 to 2005. This result also mitigates the concern that high 1996 latent demand zip codes are experiencing increases in economic opportunity over this time period, and it suggests that households may respond to mortgage credit expansion by reducing other consumer debt. Columns 3 and 4 demonstrate that high 1996 latent demand zip codes do not experience any differential increase in default rates on non-home debt, which helps rule out alternative stories for our results.

In Table 7, we examine alternative measures of high 1996 latent demand. As we discussed in Section III, our empirical methodology relies on variation across zip codes in the initial fraction of borrowers that cannot obtain mortgages due to excessive credit risk. In Table 7,

we examine the fraction of loans backed by the Federal Housing Authority in a zip code in 1996, the fraction of subprime borrowers in the zip code in 1996, and the fraction of housing units occupied by renters in a zip code in 2000 as alternative measures of initial high latent demand. As the estimates in Table 7 demonstrate, our core results are robust to these alternative measures of high latent demand. The magnitudes of the second stage coefficients are larger for the FHA and subprime shares, and slightly smaller for the fraction of units rented.

V. Concluding Remarks: What are the Macro-Economic Magnitudes of the Supply Shift?

The process of disintermediation in the mortgage industry led to a sharp shift in the supply of mortgage credit from 2001 to 2005. The expansion in supply was targeted at subprime customers who were traditionally marginal borrowers unable to access the mortgage market. The shift in mortgage supply consequently led to a rapid rise in the risk profile of borrowers, and a surge in supply-induced house price and mortgage credit growth. These changes caused a subsequent spike in default rates, which have in turn depressed the housing market and caused financial market turmoil.

The main contribution of our work is to empirically isolate the mechanism, magnitude, and consequences of the historic shift in mortgage supply. Given that we have identified the expansion in credit and increase in house price due to the shift in the supply of credit, we can use our microeconomic estimates to answer an important macroeconomic counter-factual: How would mortgage lending and house prices have evolved if the shift in supply in the mortgage industry had *not* occurred?

To answer this question, we sort zip codes by 1996 denial rates and categorize them into 20 equal bins with 5% of zip codes in each bin. Let i index each bin, and denote by d_i the median

denial rate inside a 5% bin. Given a coefficient of 2.11 (Table 3, column 1) for the marginal effect of initial denial rate on mortgage growth from 2001 to 2005, the incremental supply-induced loan origination in bin i is equal to $2.11 * L_{i,2001} * (d_i - d_1)$, where $L_{i,2001}$ is aggregate loan origination in bin i in 2001. The total supply-induced loan origination in 2005 is thus equal to:

$$\sum_{i=2}^{20} (2.11 * L_{i,2001} * (d_i - d_1))$$

The calculation above assumes that our estimated effect of 2.12 is linear over the entire data distribution. We estimate a non-parametric plot of this coefficient and find that linearity is a reasonable approximation.

A similar calculation for calculating the average increase in house prices due to supply-shift between 2001 and 2005 can be written as:

$$\sum_{i=2}^{20} (0.34 * w_{i,2001} * (d_i - d_1))$$

where 0.34 is the house price growth coefficient from column 3 of Table 3, and $w_{i,2001}$ is the weight of bin i according to the share of loan originations belonging to that zip code.¹⁰ Solving the above expressions in our data gives us \$83 billion of additional home purchase loan originations in 2005 due to supply shift, or 15% of total home purchase originations in 2005¹¹. Similarly we get a 4.3% increase in house prices between 2001 and 2005 due to supply shift, or almost 10% of aggregate house price appreciation in the US between 2001 and 2005.

It is important to emphasize that the calculations described above are an underestimate of the true impact of supply shift for two reasons. First, changes in borrower credit quality are

¹⁰ Alternatively we could have given each bin an equal weight of 1/20, which essentially gives the same result. Also our exact calculation is slightly different in that we convert log changes to percentage changes by taking their exponent and subtracting 1.

¹¹ To remain consistent with the calculation, we only include the zip codes that are part of our study in computing the total home purchase originations.

negatively correlated with initial latent demand for mortgages which biases downward our regression estimates (as described in Section III). Second, since our empirical methodology is based on a difference-in-differences estimator, we can only estimate the *relative* impact of the shift in supply. In other words, we estimate the differential effect of the supply shift on high 1996 latent demand zip codes relative to low 1996 latent demand zip codes. Consequently, our calculation above disregards any *level* impact of the supply shift which impact all zip codes. There is evidence to suggest that this may be a significant omission. For example, even zip codes with low denial rates in 1996 took advantage of the lower lending rates by taking out home equity loans and refinancing in large amounts. The effect of the supply shift on the intensive margin of higher credit quality home-owners is material for future research.

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Figure 1A
Housing Prices, Indexed to 1996

This figure presents the house price index for the U.S. from 1992 to 2007, indexed to 1996. The house price index is constructed from the equal-weighted average of the 2,920 zip codes in our final sample. Data are from Fiserv Case Shiller Weiss

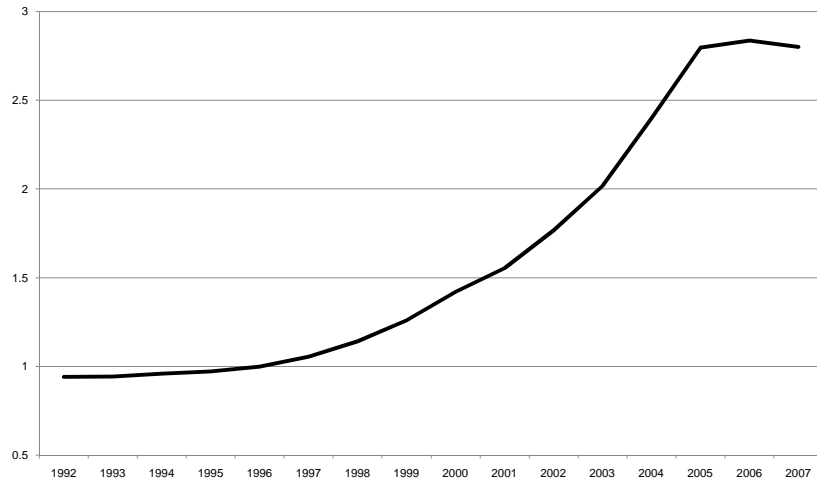


Figure 1B
Mortgage and non-Mortgage Debt Outstanding, Indexed to 1996

This figure presents total mortgage and non-mortgage consumer debt outstanding for the U.S. from 1992 to 2007, indexed to 1996. Total non-mortgage consumer debt includes student loans, auto loans, consumer loans, and outstanding credit card balances. Data are from Equifax Predictive Services.

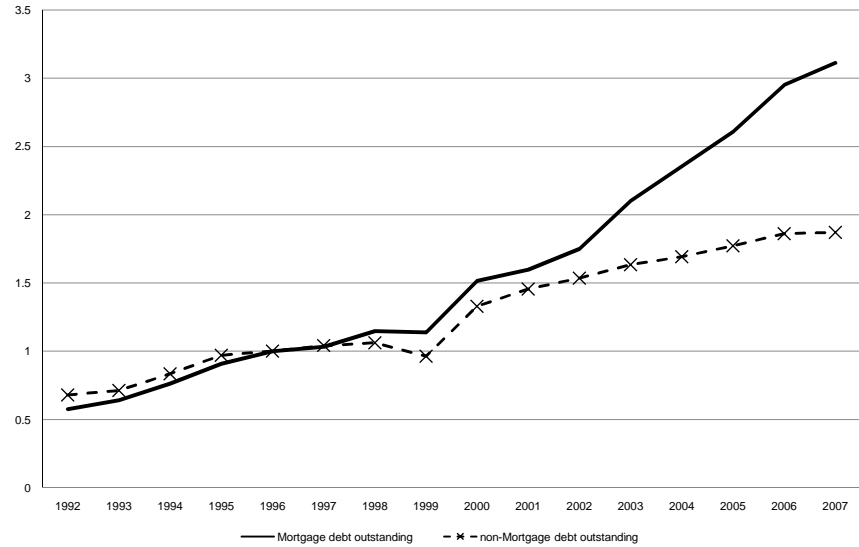


Figure 1C
Accepted Mortgage Applications for Home Purchase, Indexed to 1996

This figure presents the number of accepted mortgage applications for home purchase from 1996 to 2006, indexed to 1996. Data are from HMDA.

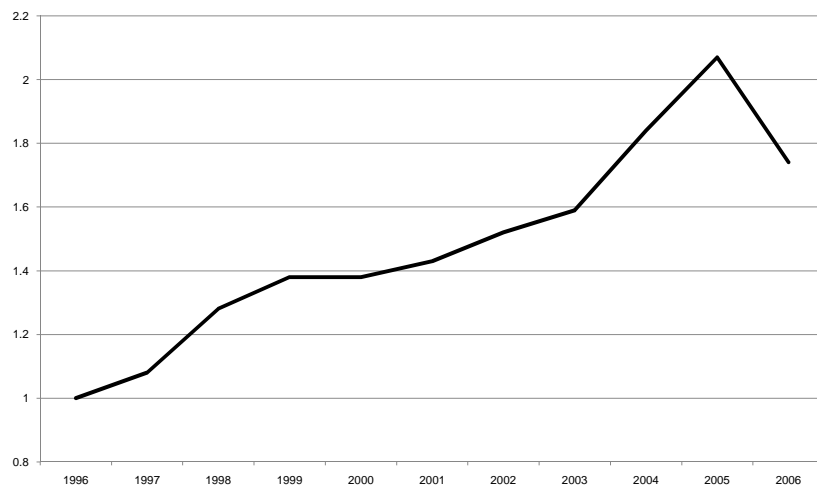


Figure 1D
Default Rates for Mortgage and non-Mortgage Debt, Indexed to 1996

This figure presents the default rate for consumer debt outstanding for the U.S. from 1992 to 2007, indexed to 1996. The total non-mortgage default rate is calculated using non-mortgage debt which includes student loans, auto loans, consumer loans, and outstanding credit card balances. Data are from Equifax Predictive Services.

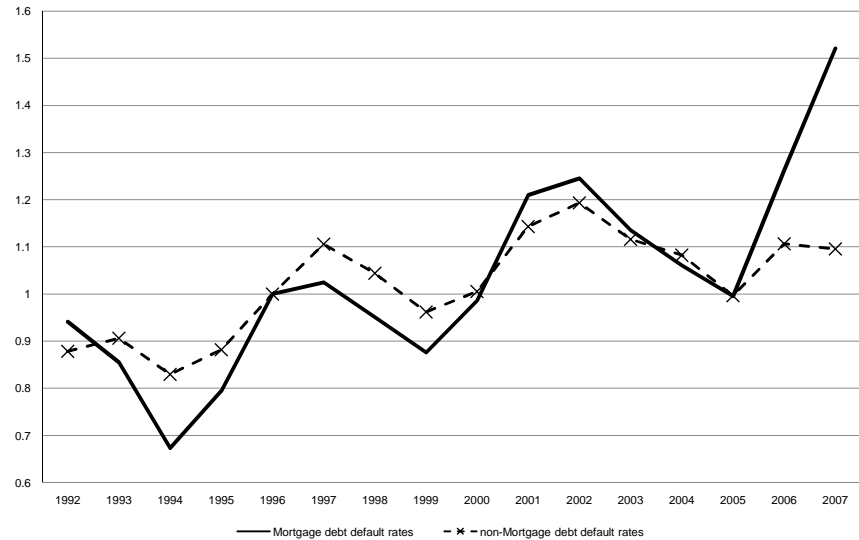


Figure 2A
Debt to Income Ratios for Accepted Mortgage Applications, Relative to 1996

This figure presents the mortgage debt to income ratios of accepted mortgage applications at the median, 75th, and 90th percentiles from 1996 to 2006. The 1996 level is subtracted from each series. Data are from HMDA.

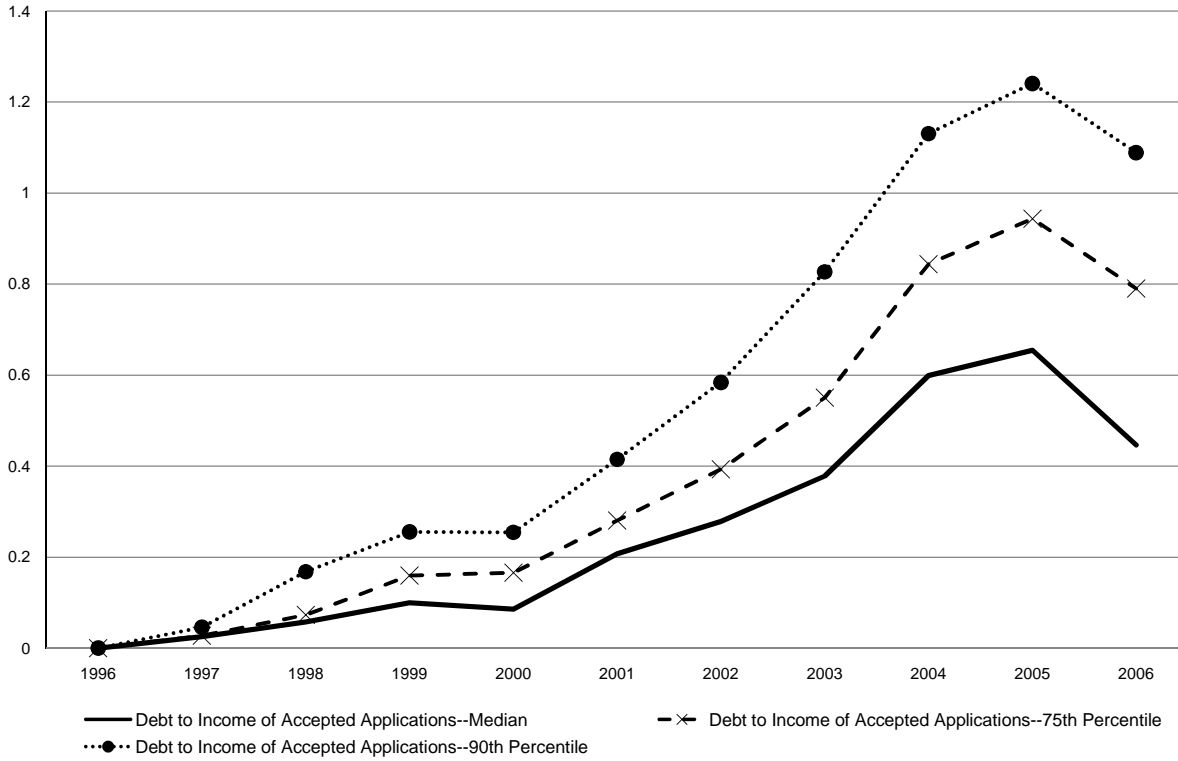


Figure 2B
Originated Mortgage Debt for Home Purchase to Income Ratios, Relative to 1996

This figure presents the average originated mortgage debt to aggregate income ratio across zip codes from 1998 to 2005. The 1998 level is subtracted from the series. Originated mortgage debt is from HMDA and aggregate income is from the IRS. * indicates data missing for the year in question.

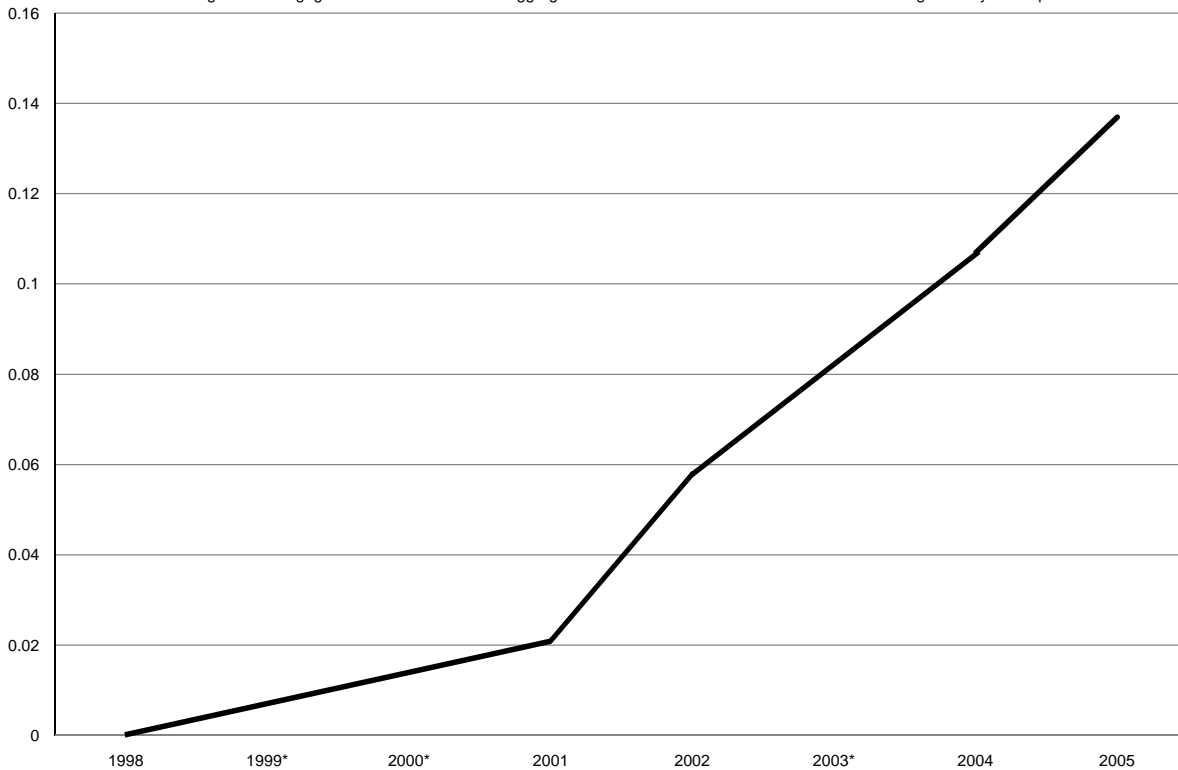


Figure 3
Fraction of Mortgages Sold to Non-Mortgage Agency Institutions

This figure presents the fraction of originated mortgages that are sold to non-mortgage agency institutions within one year of origination. Non-mortgage agency institutions include all third parties except for Fannie Mae, Freddie Mac, Ginnie Mae, and Farmer Mac. Data are from HMDA.

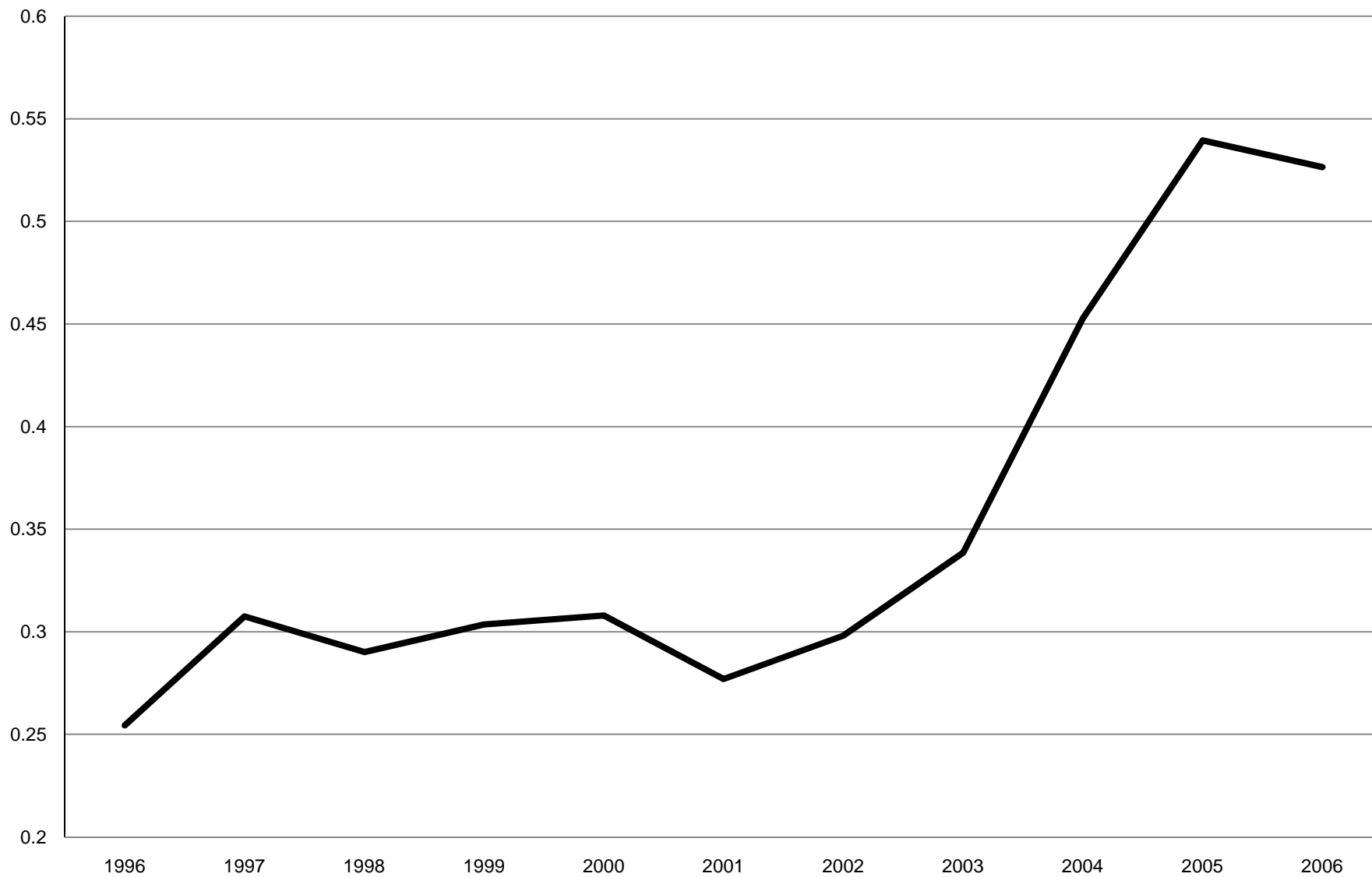


Figure 4A
Mortgage Denial Rates For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$Denied_{zct} - Denied_{zct,1996} = \alpha_c + \beta_t * HighLatentDemand_{zct,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

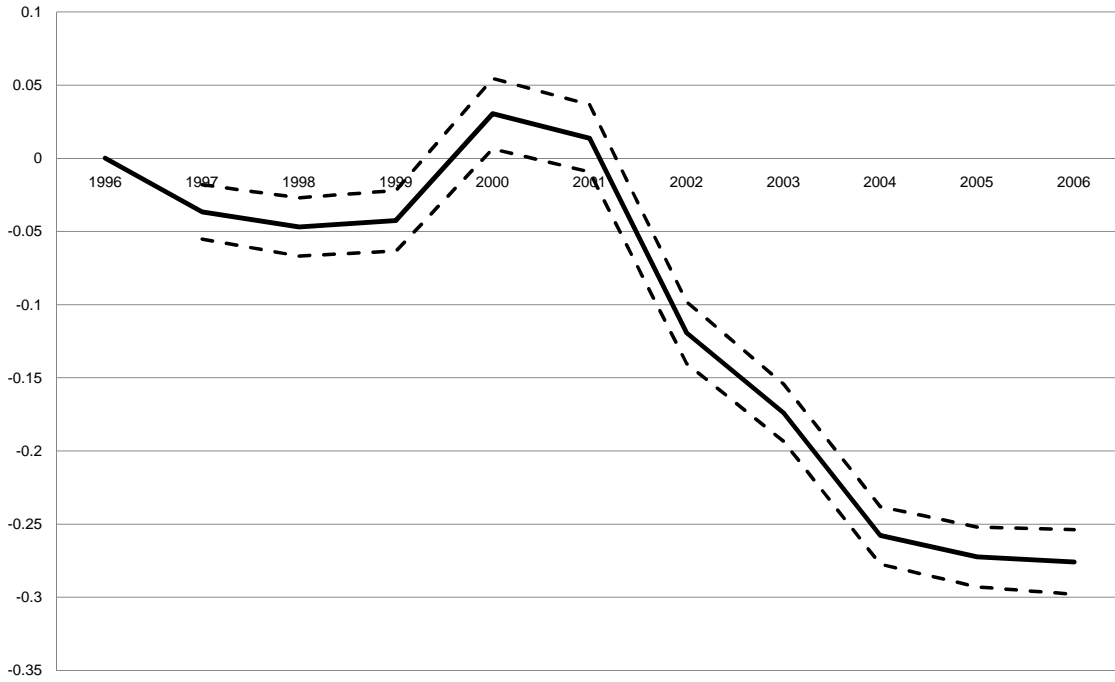


Figure 4B
Originated Mortgage Debt to Income Ratio For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$D2I_{zct} - D2I_{zct,1996} = \alpha_c + \beta_t * HighLatentDemand_{zct,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

* indicates data missing for the year in question.

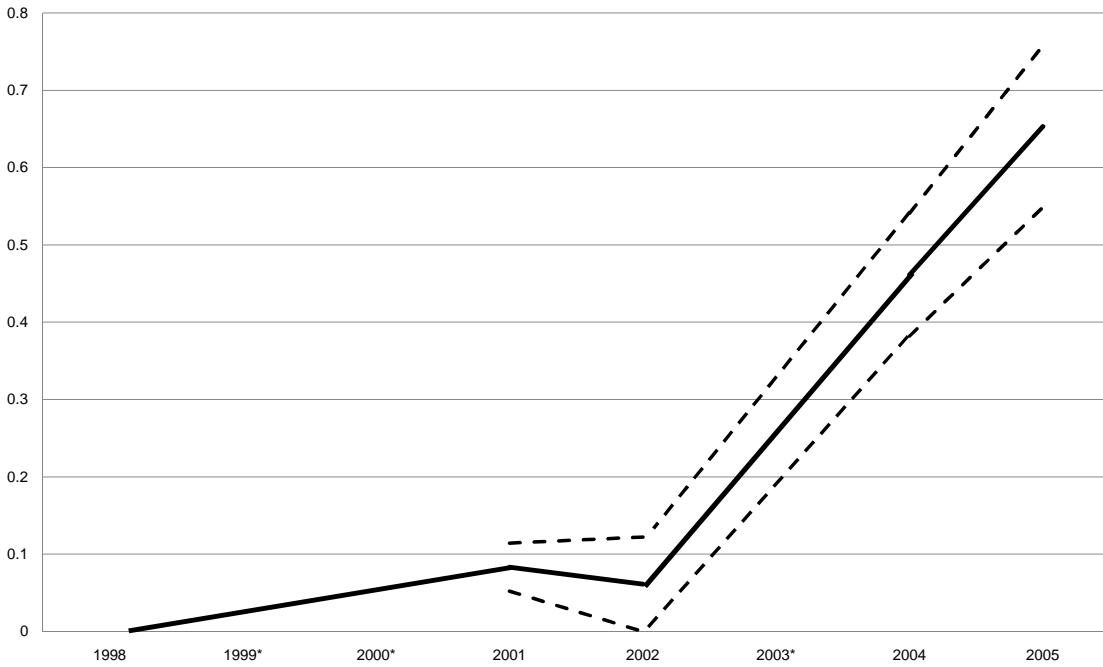


Figure 4C
Disintermediation For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$Sold_{zct} - Sold_{zc,1996} = \alpha_c + \beta_t * HighLatentDemand_{zc,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

Disintermediated loans are loans sold to any third party except for Fannie Mae, Freddie Mac, Ginnie Mae, and Farmer Mac within 1 year of origination.



Figure 5A
Number of Originated Mortgages for Home Purchase For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$\ln(\text{Orig})_{zct} - \ln(\text{Orig})_{zct,1996} = \alpha_c + \beta_t * \text{HighLatentDemand}_{zct,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

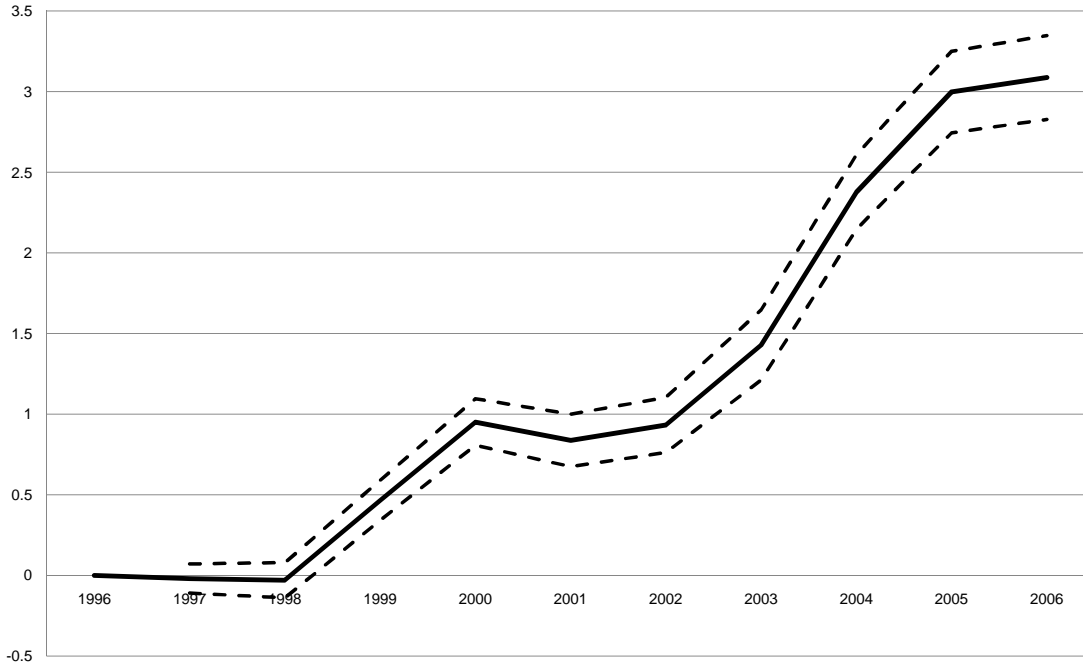


Figure 5B
Outstanding Mortgage Debt For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$\ln(\text{MortgageDebt})_{zct} - \ln(\text{MortgageDebt})_{zct,1996} = \alpha_c + \beta_t * \text{HighLatentDemand}_{zct,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

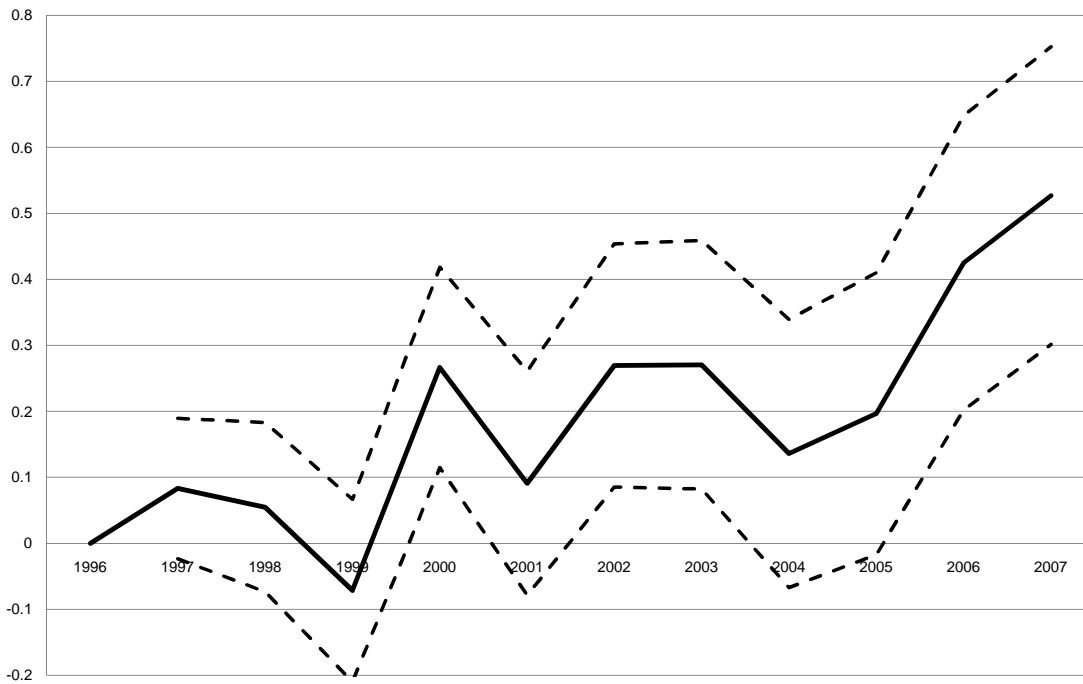


Figure 5C

Amount of Originated Mortgages for Home Purchase For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$\ln(Amt)_{zct} - \ln(Amt)_{zc,1996} = \alpha_c + \beta_t * HighLatentDemand_{zc,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

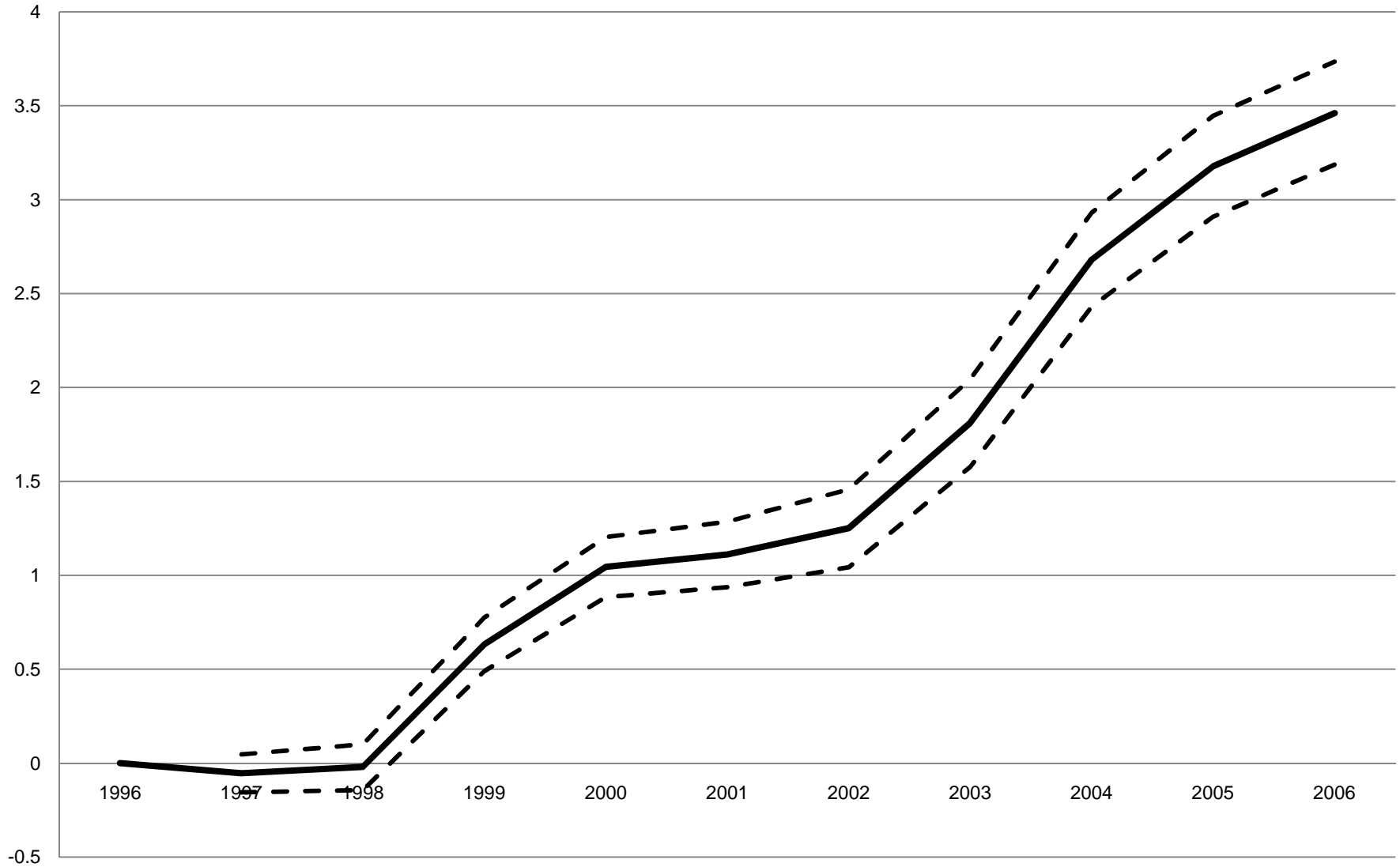


Figure 6
Relative House Price Appreciation For High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$\ln(HP)_{zct} - \ln(HP)_{zc,1996} = \alpha_c + \beta_t * HighLatentDemand_{zc,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$

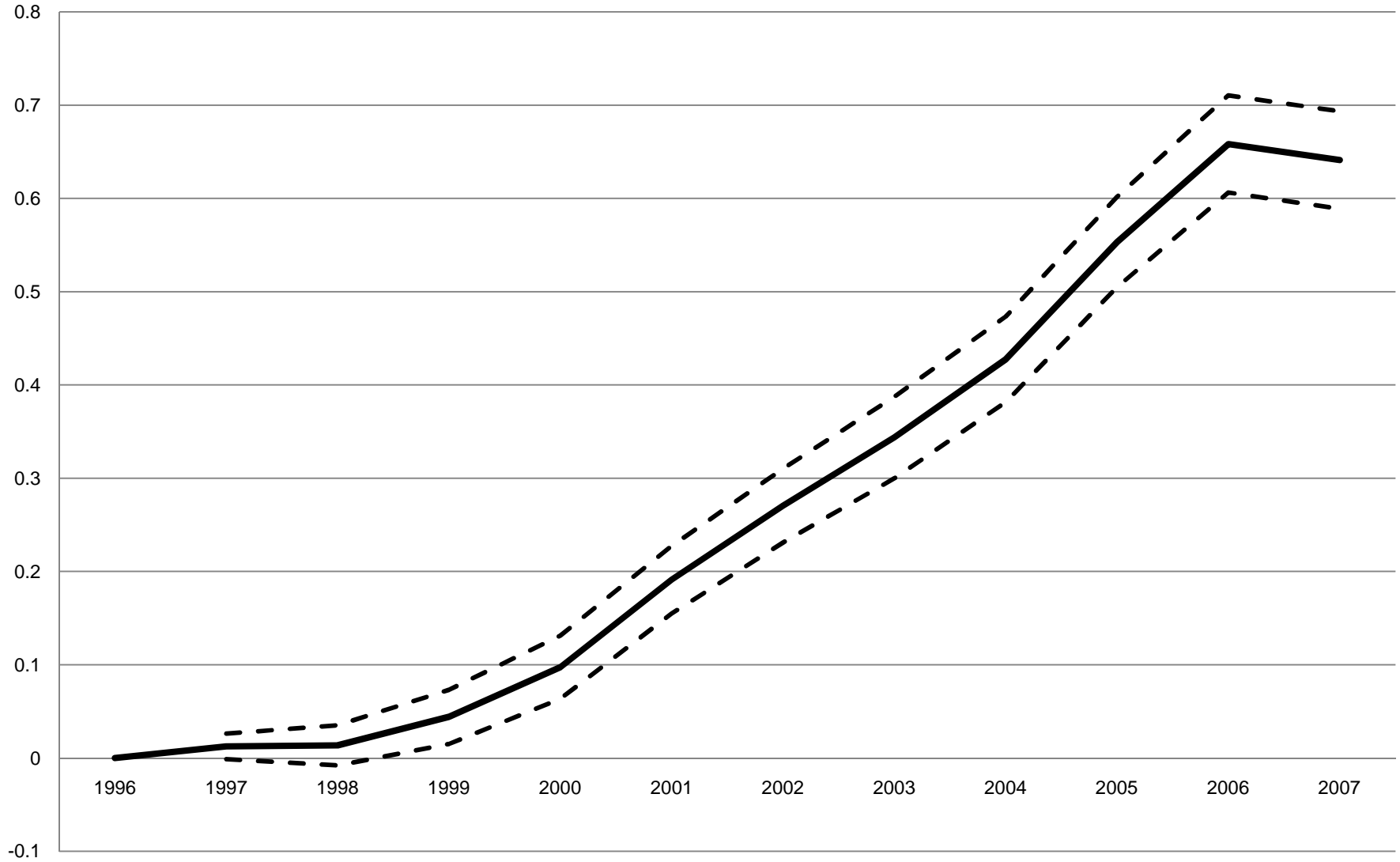


Figure 7A
Mortgage Default Rates for High 1996 Denial Zip Codes

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$DefRate_{zct} - DefRate_{zct,1996} = \alpha_c + \beta_t * HighLatentDemand_{zct,1996} + \varepsilon_{zct} \text{ for } t = 1997, 1998, \dots, 2007$$



Figure 7B
Mortgage Default Rates by Lagged Change in Denial Rates

This figure plots the estimated coefficients of β and 95% confidence intervals for each year for the following first difference county fixed effects specifications:

$$DefRate_{zct} - DefRate_{zct,t-2} = \alpha_c + \beta_t * (Denied_{zct,t-3} - Denied_{zct,t-5}) + \varepsilon_{zct} \text{ for } t = 2001, \dots, 2007$$

For example, for $t = 2007$, the coefficient estimate shows how default rates change from 2005 to 2007 for zip codes experiencing a change in denial rates from 2002 to 2004.

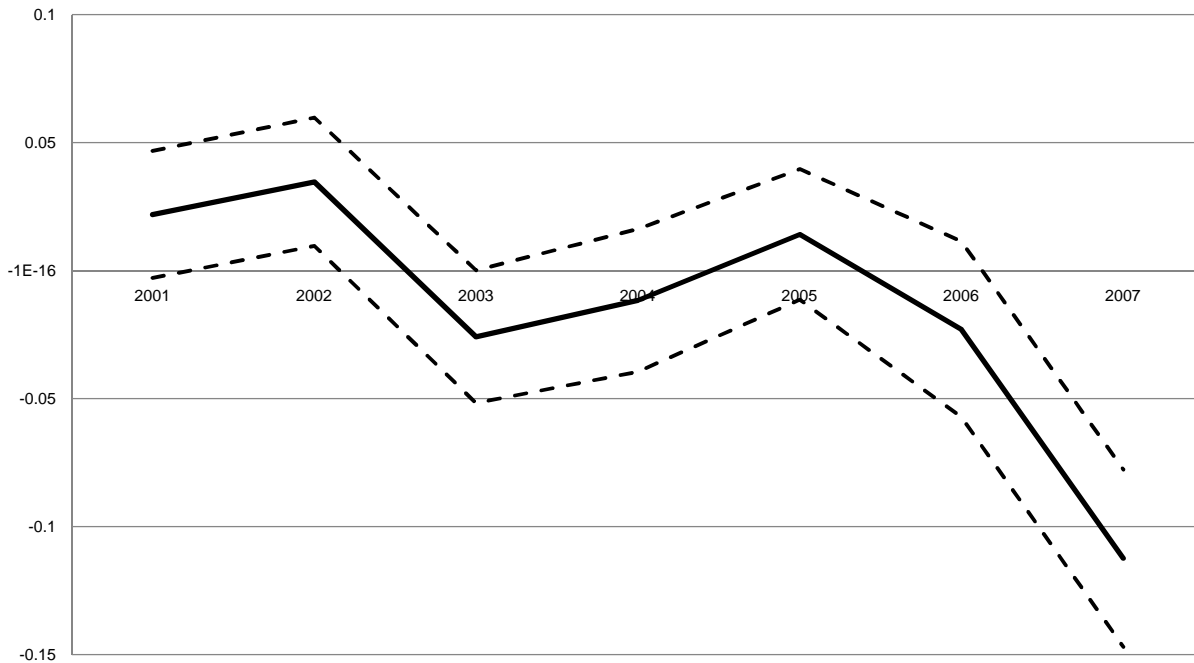


Table 1
Summary Statistics

This table presents summary statistics for the 2,920 zip codes in our sample.

	Mean	SD	10 th	50 th	90 th
<i>Equifax Data</i>					
Mortgage debt as fraction of total debt, 1996	0.741	0.106	0.600	0.761	0.854
Home equity debt as fraction of total debt, 1996	0.044	0.028	0.018	0.037	0.075
Non-home debt as fraction of total debt, 1996	0.215	0.085	0.121	0.201	0.330
Mortgage debt annualized growth, 1996 to 2001	0.069	0.042	0.023	0.068	0.115
Non-home debt annualized growth, 1996 to 2001	0.061	0.042	0.019	0.061	0.103
Mortgage debt annualized growth, 2001 to 2005	0.102	0.046	0.054	0.103	0.149
Non-home debt annualized growth, 2001 to 2005	0.046	0.046	-0.007	0.047	0.098
Mortgage default rate, 1996	0.030	0.024	0.005	0.024	0.060
Non-home default rate, 1996	0.069	0.035	0.029	0.063	0.114
Mortgage default rate change, 1996 to 2005	-0.003	0.028	-0.032	-0.004	0.025
Non-home default rate change, 1996 to 2005	-0.008	0.030	-0.041	-0.007	0.024
Mortgage default rate change, 2005 to 2007	0.017	0.027	-0.009	0.014	0.051
Non-home default rate change, 2005 to 2007	0.009	0.024	-0.016	0.009	0.034
Sub-prime consumer fraction (under 659)	0.288	0.112	0.156	0.270	0.447
<i>Fiserve Case Shiller Weiss Data</i>					
Home price annualized growth, 1996 to 2001	0.073	0.023	0.041	0.075	0.102
Home price annualized growth, 2001 to 2005	0.113	0.044	0.039	0.117	0.166
<i>HMDA Data</i>					
Mortgages for home purchase annualized growth, 1996 to 2001	0.095	0.046	0.039	0.097	0.152
Mortgages for home purchase annualized growth, 2001 to 2005	0.134	0.078	0.048	0.136	0.220
Fraction of applications denied, 1996	0.219	0.079	0.122	0.211	0.329
Fraction of mortgages sold to non-agency investors, 2001	0.277	0.074	0.184	0.275	0.374
Change in fraction sold to non-agency investors, 2001 to 2005	0.263	0.057	0.191	0.262	0.334
Fraction of mortgages backed by FHA, 1996	0.129	0.111	0.014	0.097	0.283
<i>Decennial Census, IRS, and Census Statistics of U.S. Business</i>					
Median household income, thousands, 2000	55.7	20.0	33.4	52.5	81.8
Fraction non-white, 2000	0.202	0.197	0.028	0.132	0.496
Fraction with education less than high school, 2000	0.166	0.117	0.051	0.135	0.320
Fraction unemployed, 2000	0.051	0.032	0.023	0.042	0.091
Fraction of housing units rented, 2000	0.300	0.160	0.105	0.277	0.529
Income growth (from IRS), 2001 to 2005	0.127	0.094	0.033	0.114	0.241
Wage growth (from Business Conditions), 2001 to 2004	0.072	0.121	-0.050	0.072	0.196
Employment growth, 2001 to 2004	0.025	0.192	-0.172	0.017	0.217
Establishment growth, 2001 to 2004	0.057	0.100	-0.039	0.046	0.170
Crime index growth, 2001 to 2005	0.024	0.050	-0.027	0.016	0.087

Table 2**Correlation of Fraction of 1996 Mortgage Applications Denied with Other Variables**

This table presents correlations of our main measure of latent demand for mortgages in a zip code, the fraction of 1996 mortgage applications denied, with other variables, after controlling for county fixed effects. The sample includes 2,920 zip codes.

	<u>Correlation with fraction of 1996 applications denied</u>
<i>Alternative measures of latent demand</i>	
Fraction of subprime borrowers in 1996 (under 619)	0.795**
Fraction of subprime borrowers in 1996 (under 659)	0.789**
Fraction of loans backed by FHA in 1996	0.516**
Fraction of housing units rented in 2000	0.399**
<i>Demographic variables from Census 2000</i>	
Fraction non-white	0.680**
Median household income	-0.585**
Per capital income	-0.568**
Poverty rate	0.643**
Fraction with less than high school education	0.662**
Fraction unemployed	0.567**
<i>Measures of future growth in economic opportunity</i>	
Income growth, 2001 to 2005	-0.209**
Wage growth, 2001 to 2004	-0.028
Employment growth, 2001 to 2004	-0.078**
Establishment growth, 2001 to 2004	-0.122**

**,* Correlation statistically distinct from 0 at the 1% and 5% levels, respectively

Table 3**The Effect of Mortgage Credit Expansion on Mortgage Amounts and House Prices**

Columns 1 and 2 of Panel A present estimates of the effect of high 1996 latent demand in a zip code on growth in originated mortgage amount for home purchase and growth in mortgage debt outstanding from 2001 to 2005, respectively. Column 3 presents estimates of the effect of high 1996 latent demand in a zip code on growth in house prices from 2001 to 2005. Panel B examines how outcomes vary with an alternative measure of zip codes receiving a relative increase in supply: the change in denial rates from 2001 to 2005. All specifications include county fixed effects.

Panel A: Fraction of 1996 applications denied			
	(1)	(2)	(3)
	Originated mortgage amount for home purchase growth 2001 to 2005	Mortgage debt outstanding growth 2001 to 2005	House price growth 2001 to 2005
Fraction of 1996 applications denied	2.109** (0.115)	0.444** (0.064)	0.339** (0.016)
Income growth, 2001 to 2005	0.103 (0.083)	0.500** (0.046)	-0.031** (0.011)
Wage growth, 2001 to 2004	-0.036 (0.057)	0.044 (0.031)	-0.004 (0.008)
Establishment growth, 2001 to 2004	0.008 (0.042)	-0.015 (0.023)	0.009 (0.006)
Employment growth, 2001 to 2004	0.011 (0.090)	0.669** (0.050)	-0.032** (0.012)
Crime growth, 2001 to 2005	-0.050 (0.162)	0.408** (0.090)	-0.079** (0.022)
N	2897	2897	2897
R ²	0.41	0.30	0.96
Panel B: Change in denial rates from 2001 to 2005			
	(1)	(2)	(3)
	Originated mortgage amount for home purchase growth 2001 to 2005	Mortgage debt outstanding growth 2001 to 2005	House price growth 2001 to 2005
Change in denial rates, 2001 to 2005	-3.868** (0.198)	-0.477** (0.111)	-0.557** (0.027)
Income growth, 2001 to 2005	-0.068 (0.081)	0.448** (0.045)	-0.062** (0.011)
Wage growth, 2001 to 2004	-0.065 (0.056)	0.038 (0.031)	-0.009 (0.008)
Establishment growth, 2001 to 2004	0.015 (0.041)	-0.016 (0.023)	0.010** (0.006)
Employment growth, 2001 to 2004	-0.079 (0.089)	0.650** (0.050)	-0.046** (0.012)
Crime growth, 2001 to 2005	-0.183 (0.159)	0.335** (0.089)	-0.109** (0.022)
N	2897	2897	2897
R ²	0.42	0.29	0.96

**,* Coefficient estimate statistically distinct from 0 at the 1% and 5% levels, respectively

Table 4
The Effect of Mortgage Credit Expansion on Mortgage Default Rates

This table presents estimates of how the shift in mortgage supply from 2001 to 2005 affects default rates from 2005 to 2007. Columns 1 and 2 report estimates from the reduced form relating the increase in defaults from 2005 to 2007 to measures of zip codes experiencing relative supply increases from 2001 to 2005. Columns 3 through 6 relate the increase in default rates from 2005 to 2007 to measures of supply driven house price growth and mortgage origination growth, which are obtained from a first stage that relates these variables to two different supply shifters: the fraction of denied applications in 1996 and the change in denial rates from 2001 to 2005. The first stage estimates are in Table 3. All specifications include county fixed effects.

	(1)	(2)	(3)	(4)	(5)	(6)
	Change in mortgage default rates from 2005 to 2007					
Fraction of 1996 applications denied	0.066** (0.008)					
Change in denial rates, 2001 to 2005		-0.092** (0.014)				
Supply driven originated mortgage amount for home purchase growth, 2001 to 2005			0.031** (0.004)		0.024** (0.004)	
Supply driven house price growth, 2001 to 2005				0.194** (0.024)		0.165** (0.025)
Income growth, 2001 to 2005	-0.033** (0.006)	-0.040** (0.006)	-0.036** (0.006)	-0.027** (0.006)	-0.038** (0.006)	-0.030** (0.006)
Wage growth, 2001 to 2004	0.003 (0.004)	0.003 (0.004)	0.005 (0.004)	0.004 (0.004)	0.004 (0.004)	0.004 (0.004)
Establishment growth, 2001 to 2004	-0.001 (0.003)	-0.001 (0.003)	-0.001 (0.003)	-0.003 (0.003)	-0.002 (0.003)	-0.003 (0.003)
Employment growth, 2001 to 2004	0.011 (0.006)	0.008 (0.006)	0.010 (0.006)	0.017** (0.006)	0.010 (0.006)	0.015* (0.006)
Crime growth, 2001 to 2005	0.004 (0.011)	-0.004 (0.011)	0.005 (0.012)	0.019 (0.012)	0.000 (0.011)	0.014 (0.012)
Supply shifter?	Reduced form	Reduced form	1996 Denied applications	1996 Denied applications	Change in denial rates, 2001 to 2005	Change in denial rates, 2001 to 2005
N	2897	2897	2897	2897	2897	2897
R ²	0.20	0.19	0.11	0.16	0.16	0.18

**,* Coefficient estimate statistically distinct from 0 at the 1% and 5% levels, respectively

Table 5
Evidence of a Disintermediation Channel

Panel A presents coefficient estimates relating the change in the fraction of originated mortgages sold in a zip code to latent demand as of 1996. Column 1 of Panel B relates the change in median debt to income ratios of accepted applications from 2001 to 2005 to the change in the fraction of loans sold to investors. Columns 2 through 5 of Panel B presents estimates relating default rates from 2005 to 2007 to the fraction of loans sold by originators to investors from 2001 to 2005. All specifications include county fixed effects and control variables for income, wage, employment, establishment, and crime growth.

Panel A: High latent demand zip codes and mortgage sales					
	(1)	(2)	(3)	(4)	(5)
	Change in fraction sold to all investors, 2001 to 2005	Change in fraction sold to affiliates, 2001 to 2005	Change in fraction sold in private securitizations, 2001 to 2005	Change in fraction sold to banks, 2001 to 2005	Change in fraction sold to non-bank financial firms, 2001 to 2005
Fraction of 1996 applications denied	0.161** (0.014)	-0.072** (0.008)	0.177** (0.006)	0.004 (0.004)	0.125** (0.006)
N	2897	2897	2897	2897	2897
R ²	0.47	0.54	0.63	0.45	0.58

Panel B: Mortgage sales and changes in default rates						
	(1)	(2)	(3)	(4)	(5)	(6)
	Change in Debt to Income Ratio 2001 to 2005	Change in mortgage default rates from 2005 to 2007				
Change in fraction sold to all investors, 2001 to 2005	0.314** (0.056)	0.042** (0.011)				
Change in fraction sold to affiliates, 2001 to 2005			-0.105** (0.019)			
Change in fraction sold in private securitizations, 2001 to 2005				0.147** (0.023)		
Change in fraction sold to banks , 2001 to 2005					-0.023 (0.036)	
Change in fraction sold to non-bank financial firms , 2001 to 2005						0.136** (0.023)
N	2919	2897	2897	2897	2897	2897
R ²	0.77	0.18	0.18	0.19	0.17	0.19

**,* Coefficient estimate statistically distinct from 0 at the 1% and 5% levels, respectively

Table 6
Non-home Debt Growth and Defaults

This table presents estimates of specifications that replicate specifications reported in Table 3 but with non-home debt and non-home default rates as the dependent variables. All specifications include county fixed effects.

	Non-home debt growth and defaults			
	(1) Non-home debt growth, 2001 to 2005	(2)	(3) Change in non-home default rates, 2005 to 2007	(4)
Fraction of 1996 applications denied	-0.117* (0.059)		0.007 (0.008)	
Change in denial rates, same years as dep. var.		0.012 (0.102)		-0.002 (0.014)
Income growth, 2001 to 2005	0.400** (0.043)	0.419** (0.042)	0.000 (0.006)	-0.001 (0.006)
Wage growth, 2001 to 2004	0.013 (0.029)	0.014 (0.029)	0.000 (0.004)	0.000 (0.004)
Establishment growth, 2001 to 2004	-0.012 (0.021)	-0.011 (0.021)	-0.001 (0.003)	-0.001 (0.003)
Employment growth, 2001 to 2004	0.585** (0.046)	0.590** (0.046)	-0.001 (0.006)	-0.001 (0.006)
Crime growth, 2001 to 2005	0.386** (0.083)	0.420** (0.082)	-0.002 (0.011)	-0.004 (0.011)
N	2897	2897	2897	2897
R ²	0.24	0.24	0.02	0.02

**,* Coefficient estimate statistically distinct from 0 at the 1% and 5% levels, respectively

Table 7**Alternative Measures of High Latent Demand for Mortgages**

This table presents estimates using alternative measures of latent demand. Panel A examines the fraction of 1996 loans originated that are backed by the Federal Housing Administration. Panel B examines the fraction of 1996 population with subprime credit scores (below 659). Panel C examines the fraction of housing units in the zip code occupied by renters measured as of 2000. All specifications include county fixed effects and control variables for income, wage, employment, and establishment growth.

	(1) Reduced form Change in default rates 2005 to 2007	(2) First stage Amount growth for purchase 2001 to 2005	(3) Second stage Change in default rates 2005 to 2007
Panel A			
Fraction of 1996 loans backed by FHA			
Fraction of 1996 loans backed by FHA	0.072** (0.006)	1.281** (0.092)	
Amount growth for home purchase, 2001 to 2005			0.056** (0.006)
Panel B			
Fraction of 1996 loans to subprime borrowers			
Fraction of 1996 loans to subprime borrowers (<659)	0.059** (0.005)	1.330** (0.077)	
Amount growth for home purchase, 2001 to 2005			0.044** (0.004)
Panel C			
Fraction of 2000 housing units occupied by renters			
Fraction of 2000 housing units occupied by renters	0.019** (0.003)	0.654** (0.054)	
Amount growth for home purchase, 2001 to 2005			0.029** (0.005)

**,* Coefficient estimate statistically distinct from 0 at the 1% and 5% levels, respectively

Appendix Table

Characteristics of Zip Codes With and Without House Price Data

This table compares the 3,056 zip codes for which FCSW collects house price data to the 16,312 zip codes for which house price data are unavailable.

	House price data available	House price data not available
Fraction urban, 2000	0.919	0.464
Population, 2000	6,630	4,763
Median household income (thousands), 2000	55.7	39.8
Poverty rate, 2000	0.092	0.129
Fraction of housing units built in last 5 years, 2000	0.084	0.116
Fraction of households that have moved in last 5 years, 2000	0.352	0.337
Mortgage default rate, 1996	0.030	0.028
