Collateral Damage: The Impact of Shale Gas on Mortgage Lending*

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Abstract

We analyze mortgage lenders' behavior with respect to shale gas risk during the period of the U.S. shale gas boom, which coincided with the U.S. housing market rise, collapse and subsequent increase in lending industry scrutiny. Shale gas operations may place affected houses into technical default such that GSE’s (Fannie Mae and Freddie Mac) are unable to maintain them in their portfolios. We find that lenders did indeed increase the weight they place on shale risk relative to income risk in mortgage pricing behavior after 2010. This effect is particularly evident for groundwater dependent properties, indicating that lenders view shale activities as placing the residential value of these properties at greater risk. When we quantify the willingness to pay to avoid shale risk, we find that insurers, on average, lose around $9,496, or 5% of profit earned on an average mortgage.

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1 Introduction

The Energy Information Administration reports that between 2007 and 2014, annual natural gas withdrawals from shale deposits in the U.S. rose from 1.99 to 13.8 million cubic feet,\(^1\) growth that led to lower natural gas prices, higher natural gas demand, and substitution from other forms of fossil fuel consumption like coal. The recent rise of shale gas has been spurred by technological development that combines large-scale hydraulic fracturing and horizontal drilling techniques with other advances in three-dimensional surveying techniques. In addition to allowing for more efficient extraction from broad, tight-shale layers, the horizontal drilling technology increases access to large areas of shale from relatively confined surface areas, which allows firms to extract oil and natural gas stored in tight-shale formations located beneath densely populated neighborhoods.\(^2\) Unlike traditional oil and natural gas reservoirs, settlement patterns throughout the U.S. have evolved for decades with indifference to the location of shale. Consequently, we now often find residential properties located on top of those resources, bringing shale gas activity into homeowners’ backyards, and beginning in 2014, more than 15 million Americans live within one mile of an active oil or gas shale well.\(^3\) Our paper explores how the extraction technology changes have altered mortgage markets by evaluating how lenders internalize the increased risks born by leveraged houses located near to shale oil and natural gas development.

What are the chief risks born by proximity to these large-scale, hydraulically fractured wells? Current literature is evolving to identify and measure the health, economic, and geologic consequences of proximity to these industrial activities, especially in light of increased household exposure.\(^4\) Among the identified risks is air pollution from well production and transmission activities, which leads to increased methane emissions.\(^5\) Waste water disposal from fractured wells is linked to surface water contamination caused by radioactive salts and metals or by the chemicals used to the treat the fractured wells.\(^6\) Geologically, there is a growing literature linking the employment of large-scale fracturing technology and increased incidence of tremors.\(^7\) The

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\(^1\)Energy Information Administration, \url{https://www.eia.gov/dnav/ng/hist/ngm_epg0_fgs_nusmmcfa.htm}

\(^2\)In particular, many horizontal laterals can be drilled in different directions from a single wellpad, which may comprise less than an acre of land. Horizontal laterals can extend in mile-long segments beneath suburban and urban regions.


\(^4\)Coase (1960) noted that externalities are reciprocal in nature, and that the external costs of a production process will be low if no one is around to be affected by it. In contrast, the evolved shale gas extraction technology allows for many impacts on nearby landowners.

\(^5\)The literature measures the impact of increased methane emissions during the drilling, fracturing and production phases of well development, in particular, including Caulton et al. (2014), Brandt et al. (2014), and the increases in other particulate matter, including Colborn et al. (2011) and Roy et al. (2014), and volatile organic compounds, as in Gilman et al. (2013).

\(^6\)Olmstead et al. (2013), Warner et al. (2013), Fontenot et al. (2013), and Hill and Ma (2017)

\(^7\)Koster and van Ommeren (2015) and Cheung et al. (2016)
hedonic literature identifies community-wide costs and benefits of proximity to shale, whereby the costs are greatest for households accessing groundwater,\textsuperscript{8} and more broadly, communities experience degradation of amenities through increased noise, road damage, and traffic accidents.\textsuperscript{9} Conversely, the literature also identifies and measures economic benefits to having an active extraction industry, which includes higher wages, income, and municipal revenue.\textsuperscript{10} Added to the economic benefits are royalties and bonus payments earned by households that own (and lease) the rights to their sub-surface minerals. We contribute to this literature by identifying and measuring an additional cost of these technologies that is internalized by the mortgage lending industry and passed on to homeowners through higher lending rates.

The fact that shale development has the potential to impact property values means that it also has the potential to interact in a variety of ways with mortgage lending practices. First and foremost, a mortgage loan is commonly secured for both surface and subsurface rights. Mortgages do not generally allow homeowners to sell or lease parts of their property without prior approval from the lender; however, mortgage experts generally report that requests for approval are rare (NYT (2011) and Law (2011)).\textsuperscript{11} The situation is further complicated by the fact that most mortgages are not held in the portfolio of the primary lender, but are instead sold on the secondary mortgage market. Lenders participating in the secondary mortgage market, in effect, sell their loans to government sponsored enterprises (GSE’s)\textsuperscript{12} or investment banks that bundle the loans (securitize) and sell the resulting securitized assets (mortgage-backed securities) to individual investors. Participation in the secondary market increases lenders’ liquidity; however, GSE’s are prohibited from purchasing mortgages on properties engaged in industrial activities including transport or storage of toxic substances (chemicals, oil and gas products, or radioactive materials) (Law (2011)), and shale gas extraction involves a number of activities that have the potential to violate these rules, which would leave the borrower (without prior approval) in

\textsuperscript{8}Among other hedonic analyses, recent research has shown that shale development can negatively impact nearby housing prices (Throupe et al. (2013), Gopalakrishnan and Klaiber (2014), James and James (2014), and Muehlenbachs et al. (2015)), but other research has reached differing conclusions, as in Delgado et al. (2014) and Boslett et al. (2016).

\textsuperscript{9}Muehlenbachs and Krupnick (2013) and Graham et al. (2015) link increased incidence of traffic accidents to drilling activity.

\textsuperscript{10}Mason et al. (2015) and Bartik et al. (2016) measure some of the economic benefits, and Cesur et al. (2017) documents the benefits to infant mortality that are attributable to energy generation from natural gas as opposed to coal.

\textsuperscript{11}Nationwide, Bank of America reports receiving approximately 100 such requests per month, fewer than a dozen are sent to Fannie Mae and Freddie Mac each year. (NYT (2011)) Chesapeake Energy, one of the largest drilling companies, only seeks permission from lenders before a property is drilled, not before it is leased; this violates mortgage rules, which require approval to sign a lease. (NYT (2011))

\textsuperscript{12}Government sponsored enterprises (GSE’s) include Fannie Mae (Federal National Mortgage Association) and Freddie Mac (Federal Home Loan Mortgage Corporation).
technical default.\textsuperscript{13,14} Because the primary lender is responsible if the borrower defaults due to shale extraction activities, lenders have strong incentives to precisely evaluate the risks incurred by lending to homeowners in regions with high levels of shale development.

We use the mortgage market to learn about additional external costs associated with the “fracking boom” by quantifying the relationship between shale drilling risks and the frequency of subprime (higher priced) lending. In particular, we use loans issued to buy homes in Tarrant County, Texas, to test whether lenders mitigate shale development risks by issuing more subprime loans to households that are susceptible to drilling externalities. We measure the extent to which lenders’ preferences (distaste) for shale gas risk have shifted relative to other forms of risk since 2010 (i.e., during a period when the industry reconsidered its lending practices and shale gas activities rapidly expanded). In particular, the rise in shale extraction was coupled with the collapse of the housing market and, with it, an increased scrutiny of GSE activities leading lenders to worry they may have to assume control of the shale-exposed mortgages as a consequence of technical default. We find that it is important to estimate differential preferences for shale risk (relative to income risk) before and after this period in which technical default and lost property values become more significant concerns among lenders. Moreover, we consider how those preferences vary with water source, which is an important component of house value and may be particularly vulnerable to shale risk. In particular, we ask whether lenders treat properties that are already exposed to nearby drilling differently from those that have signed a lease to extract their minerals and are likely to be exposed hydraulic fracturing operations, compared to those that do not yet have a lease.

We use a combination of housing, lending, and drilling data to capture variation in subprime lending practices across households that are more or less susceptible to the negative externalities of nearby drilling behavior. Our data spans the growth of shale gas activity and the periods before and after the financial crisis, and we are able to estimate how lenders’ preferences differed

\textsuperscript{13} Further, there may be other discrepancies between the terms specified in an oil or natural gas lease agreement and those specified in a mortgage. For example, some states void title insurance if a property is used for any commercial venture (Law (2011)) and, without title insurance, participation in the secondary market is limited. Homeowner insurance policies will also be violated if industrial activities are present, leading to default (May (2011)).

\textsuperscript{14} Borrowers may even find themselves in violation of their mortgage agreement through no fault of their own, but rather simply because they happen to be located in close proximity to another property that engaged in shale development. For example, news reports described a couple in Washington County, PA who were denied a new mortgage for their property by Quicken Loans because of shale development on a neighboring plot. Quicken responded by saying – “While Quicken Loans makes every effort to help its clients reach their homeownership goals, like every lender we are ultimately bound by very specific underwriting guidelines. In some cases conditions exist, such as gas wells and other structures in nearby lots, that can significantly degrade a property’s value. In these cases, we are unable to extend financing due to the unknown future marketability of the property. (http://www.wtae.com/investigations/Couple-denied-mortgage-because-of-gas-drilling/12865512#.T6mu842bM44.facebook)”
across these two periods along several dimensions. In the data, we observe that lenders weight shale risk more heavily in their subprime decisions after the financial crisis. We then use a non-parametric estimator, first introduced by Fröhlich (2006), that captures firms’ preference heterogeneity for income and shale risks and allows us to estimate the trade-offs across our two types of risk without imposing functional form assumptions. Focusing on the comparisons across water sources, we estimate a distribution of lenders’ preferences for the two sources of risk and show how it differs by groundwater versus piped water. In particular, we find that after the financial crisis, lenders are willing to bear significantly more income risk to reduce their exposure to shale risk among houses accessing groundwater.

Our paper differs from the existing shale gas literature by focusing on lenders’ decisions who, we argue, are more likely to internalize risks especially in light of greater scrutiny in lending practices post-financial crisis, which evolved throughout 2008 and 2009. Lenders face trade-offs between different types of lending risk on a daily basis, allowing them for develop an expertise in weighing the risks of any particular loan. Conversely, households may only buy a few houses over the course of their lives, which tasks households with a large information burden when deciding how to internalize perceived shale risks. Using measures that capture shale and income risks, along with other household characteristics, we non-parametrically estimate lenders’ preferences to issue high risk loans, as indicated by a subprime mortgages. Following Bajari and Benkard (2005) and Fröhlich (2006), we characterize lenders’ heterogeneous preferences and trade-offs between two types of risks: income and shale.

The remainder of the paper proceeds as follows. We begin by fitting the analysis into additional literature that is relevant to our questions in Section 2. We then describe the secondary market for mortgages in Section 3 and our data set in Section 4, which combines information from the Home Mortgage Disclosure Act with data on property appraisals and well locations. Section 5 describes the theory used to describe lenders’ relative preferences for shale and income risk and Section 6 describes the empirical framework. Section 7 reports estimates and Section 9 concludes with a discussion of policy relevance.

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Our study is a complement to literature exploring whether there is an observed difference in mortgage default behavior for households living in active shale regions compared to non-shale. Upton (2016) studies the probability of default in the presence of shale gas development, examining the default behavior of landowners in a period when national default rates were rising. He finds that landowners living in shale gas regions are less likely to default on their mortgage loans, lending credence to the positive economic impacts of shale gas development. Shen et al. (2015) conduct a similar study with data describing default behavior of households in Pennsylvania located over the Marcellus Shale.
2 Other Relevant Literature

In addition to the already cited environmental economics literature identifying and measuring the costs and benefits of the growing shale extraction industry,\textsuperscript{16} this paper contributes to literature concerned with the effects of the financial crisis on lending practices, subprime lending practices, more generally, and the relationship between shale development and foreclosure incidence.

In addition to estimating preference heterogeneity for shale and income risk, we capture differences in lending practices before and after the financial crisis, which is a period of increased foreclosures and significant changes to industry behavior that emphasizes greater scrutiny for each considered mortgage. We add to the literatures linking increased foreclosure rates with subprime lending\textsuperscript{17} and impacts on property values and neighborhoods. Gerardi et al. (2007) study the relationship between subprime lending and the increased incidence of observed foreclosures in the context of the Massachusetts housing market. Focusing their analysis on years encompassing the U.S. financial crisis, the authors find that, along with house price depreciation, subprime lending is positively correlated with the incidence of foreclosure citing that subprime borrowers are 20% more likely to face foreclosure than prime borrowers. McCarthy et al. (2002) cite the longer-term effects of foreclosures on households’ access to stable and decent housing, and Simons et al. (2009) and Lin et al. (2009) capture the spillover effects of foreclosed properties on the surrounding blocks. Further, Immergluck and Smith (2006) find a positive correlation between foreclosures and violent crime levels at the neighborhood level.

In our analysis, we control for households’ race and ethnicity in order to capture the oft cited relationship between subprime lending and sociodemographic characteristics. Munnell et al. (1996) is the among the papers that use data assembled in compliance with the Home Mortgage Disclosure Act to study potential discriminatory lending practices toward minority households.\textsuperscript{18} Gerardi and Willen (2009) study Massachusetts property-level data and find that subprime lending leads to more turnover in minority neighborhoods.

The model we present captures the risk taking behavior of local and national lenders when assessing mortgages for properties that are potentially impacted by shale gas development. This is a complement to a new literature documenting the relationships between shale gas activity

\textsuperscript{16}Our review of this literature is by no means complete, as it is a growing literature.

\textsuperscript{17}Some of these studies also include those by Newman and Wyly (2004), Immergluck and Smith (2005), and Quercia et al. (2007).

\textsuperscript{18}The authors build on the literature analyzing discriminatory lending practices in mortgage markets that include Black et al. (1978), King (1981), Schafer and Ladd (1981) and Ladd (1998), among many others with additional literature summarized in Yinger (1996) and LaCour-Little (1999). A more descriptive analysis of the HMDA data and sociodemographic heterogeneity can be found in Avery et al. (2006).
and capital allocation and between shale gas activity and foreclosures. It also draws upon the literature analyzing other observable factors that lead to foreclosures, or more risky loans from the perspective of banks. A nascent literature explores the relationship between unconventional energy development and the local banking sectors. Plosser (2014) studies the deposit shock resulting from landowners profiting from nearby shale gas development and the banks’ resulting capital allocation decisions. Gilje (2012) uses shale discoveries as a natural experiment to study credit supply and its implication for economic outcomes especially in areas dominated by smaller banks. Gilje et al. (2016) explores whether banks export liquidity because they are exposed to positive shale gas shocks. The authors find that banks with branches located in non-shale areas increase lending in those branches, and the positive effect is more pronounced for ‘difficult-to-securitize’ loans. Finally, there is an older literature that employs models to capture the default probabilities for mortgages as a function of factors like loan-to-value ratios and house-price depreciation.\footnote{Refer to Quercia and Stegman (1992) and Kau et al. (1994) for other contributions to this literature.}

Gerardi et al. (2009) is a more recent paper exploring the effects of house price depreciation on foreclosure rates versus relaxed underwriting standards. They find that the two mechanisms are interrelated, but, without house-price depreciation, there would not have likely been such a dramatic increases in the foreclosure rate.

3 Secondary Markets for Mortgages

Lenders interested in selling their mortgages to government sponsored enterprises (GSE’s), like Fannie Mae (Federal National Mortgage Association) and Freddie Mac (Federal Home Loan Mortgage Corporation), in the secondary market must meet certain criteria that may be violated by proximity to shale development. These details are noted in the following section along with a short history of GSE’s and potential risks of shale gas development.

Among the largest actors in the secondary mortgage markets are the GSE’s. The government established these enterprises to introduce additional liquidity into the mortgage market and promote home ownership. In the U.S., 90% of all houses are purchased with mortgage financing. Lending institutions typically do not rely on their own capital to support most of the loans that they write over the long-term. Instead, loans are bundled (securitized) and sold to investors as mortgage-backed securities. The federal government, primarily through its GSE’s, is the largest of these investors, purchasing over 90% of mortgages in the US today.\footnote{https://smartasset.com/mortgage/everything-you-need-to-know-about-the-secondary-mortgage-market} Fannie Mae and Freddie Mac assume the credit risk for all mortgages that are re-sold as mortgage-backed securities. In
exchange for bearing this risk, the GSE’s keep the guaranty fees associated with the loans. As recently as 2008, Fannie Mae and Freddie Mac had owned or guaranteed about half of the U.S.’s $12 trillion mortgage market. This, along with the low rates at which the GSE’s are typically able to borrow, make them highly profitable enterprises. After the housing market crash of 2008, lenders began to evaluate the standards under which home mortgage loans were approved and credit was subsequently tightened. Both GSE’s underwent significant scrutiny, were restructured, and fell under the conservatorship of the Federal Housing Finance Agency. Since that time, both the House Financial Services and Senate Banking Committees passed reforms that would have reduced the government’s footprint in housing finance (although neither was passed into law), and a similar plan was proposed by the Obama administration. With this as a backdrop, the continued growth of shale gas in the U.S. led to congressional hearings beginning in 2009, continuing in 2010, and culminating in a flurry of activity in the fall of 2011 both inside and outside Washington.

3.1 The GSE’s and Shale Gas

The GSE’s specify a set of criteria to which lenders must adhere if they want to be able to sell mortgages on the secondary market, and there are many areas where these criteria may conflict with standard practice in shale gas or oil development. For example, Freddie Mac guidelines 39.4(i) specify that a mortgage can only be issued on a leased property if “exercise of the rights will not result in damage to the mortgaged premises or impair the use or marketability for residential purposes.” Furthermore, the guidelines prohibit “right of surface or subsurface entry within 200 feet of a residential structure,” and require “comprehensive endorsement to the title insurance company that affirmatively ensures the lender against damage or loss from exercise of such rights.” Practically, this requires “no structure erected on premises exceeding three stories or 35 feet,” that the premises “shall not be used for storage of any material machinery, equipment, or supplies,” and that the property will “not be used for commercial purposes.” Furthermore, the Freddie Mac guidelines 39.4(m) require that lenders must warrant that activities on the property:

21 https://www.nafcu.org/HousingFinanceReform/
22 https://www.treasury.gov/initiatives/Pages/housing.aspx
must not interfere with the use and enjoyment of any present or proposed improvements on the mortgaged premises or with the use and enjoyment of the balance of the mortgaged premises not occupied by improvements,

• must not affect the marketability of the mortgaged premises,

• must have no or minimal effect on the value of the mortgaged premises,

• must be commonly acceptable to private institutional mortgage investors in the area.

Fannie Mae and Freddie Mac have a fiduciary duty to establish rules that reduce the risk of lost house value or default. As noted in the introduction, the extraction of shale gas involves a number of activities that have the potential to violate these rules, which would leave the borrower (without prior approval) in technical default. Toxic chemicals are pumped, along with million gallons of water and sand, directly under mortgaged homes. “Produced” water, which is forced back out of the well, contains brine, fracking chemicals, and even radioactive substances. It is often stored on-site, sometimes in open holding ponds. Permanent easements for truck and pipeline transport, production platforms, and storage facilities (that can spill) are common on properties with or near drilling activity. Risks to home values can be a particular problem for homes where the water supply is threatened. Finally, without title insurance (see above), secondary lenders may not be able to hold mortgages.

The primary lender is responsible if the secondary lender does not know about the lease and the house goes into technical default as a result. Fannie Mae and Freddie Mac can demand that the originating lender buy back any loans that do not meet secondary market requirements (Carpenter 2011). There is not a good measure of how many mortgages may currently be in violation of secondary mortgage market rules.

In light of these growing concerns, a primary lender who believes that a property may soon be approached for shale development may worry that the property could default, would have to be foreclosed upon, or that shale development might hamper its ability to sell the mortgage on the secondary market. A related concern might be that noncompliant mortgages already sold to the secondary market would have to be bought back. As such, that lender may charge a premium to lend when there is concern over impending shale development. Alternatively, to the extent that they are able, lenders may simply exit markets where shale gas is prevalent.25

25NYT (2011) reports that, in 2011, at least eight local or national banks did not typically issue mortgages on properties exposed to shale gas development. In other instances, lenders began requiring drilling companies to indemnify property owners against any future losses to home value, or requiring home owners to expressly agree not to sign a lease as long as they hold the mortgage.
In the remainder of this paper, we quantify changes in lenders’ preferences by estimating the changes in trade-offs between income and shale risk before and after the financial crisis and concurrent increased interest in shale gas at the federal level. This provides us with a new perspective on the costs of the risks associated with shale gas development for nearby homeowners, specifically measured via the decisions made by housing professionals (i.e., mortgage lenders).

4 Data

4.1 Tarrant Co., Texas

Our analysis will focus on shale gas development and its impact on property markets in Tarrant County, Texas. Tarrant Co., located in north-central Texas, is the home to approximately 1.8 million residents. It is comprised of 41 incorporated areas, including Fort Worth, which is the county seat. The population of Tarrant is approximately 27% Hispanic or Latino (of any race). Tarrant Co. and the underlying Barnett shale are typically considered to be the birthplace of modern hydraulic fracturing because of innovations made there by Mitchell Energy.

4.2 HMDA

The goal of HMDA was to determine if lenders are serving communities’ financial needs, and to facilitate enforcement of fair lending laws. When buying a house, one typically fills out a form at closing that transmits information about the race, sex and income of the buyer along with the loan amount and terms. As of 2006, there were 8,850 lenders covered by the disclosure rules (approximately 80% of home lending nationwide). In 1989, the HMDA law was amended to require disclosure of loan-level information, and in 2004, it was further amended to require disclosure of information about loan pricing. Specifically, the lender is required to report the spread between the annual percentage rate and the applicable Treasury yield if it is greater than or equal to 3 percentage points for a first-lien loan. After 2009, the rule for first-lien loans was changed to require reporting if the difference between the annual percentage rate and the applicable average prime offer rate is greater than or equal to 1.5 percentage points (i.e., both the baseline and the cutoff rule changed, see Fed 2009). As such, HMDA does not specifically identify subprime loans, but rather “higher priced” loans; we use their reporting requirement in each period as the determinant of “higher priced”. We also follow the common practice in the literature and use these terms “higher priced” and “subprime” interchangeably. In 2004,

the first year when pricing information was provided, fewer than 20% of households had higher priced loans, and higher priced loans were more common amongst black and Hispanic borrowers.

The number of loans reported as higher priced depends upon many factors, some of which have nothing to do with the borrower’s riskiness. In particular, a narrowing of the difference between short and long-term interest rates can increase the number of loans overall exceeding the “higher priced” threshold; we account for this in our empirical model below.

We employ the debt-to-income ratio as our measure of “income” risk. *Ceteris paribus*, given two borrowers with the same income, the borrower with the larger debt will be more at risk of shocks that will prevent repayment of the loan, leading to default. HMDA reports both the size of the loan and the borrower’s (self-reported) income.

### 4.3 Dataquick

Data from the real estate data services company Dataquick, accessed through an a licensing agreement with the Duke University Department of Economics, are used in conjunction with information from the Tarrant Co. Assessor’s Office, to measure assessed value and other house attributes.

### 4.4 Exposure to Shale Development

We base our measure of “shale risk” on a duration model with censoring. In particular, shale risk is measured by the cumulative hazard function, which measures the total amount of risk that has been accumulated up to time $t$. In our duration model, a failure occurs when a drilling well appears within the 1-km radius of a particular house. For houses where a failure has not yet occurred, we consider the following attributes in calculating the cumulative hazard function: number of issued permits within 1-km radius and 2-km radius, number of producing wells within 2-km radius, and distance to the nearest producing well. In particular, we use the proportional hazard (PH) model to calculate shale risk. Let $\tau$ be a non-negative random variable denoting the time to a failure event, and denote $S(t)$ as $\tau$’s survivor function and $h(t)$ as its hazard function. Then we have

\[
S(t) = 1 - F(t) = Pr(\tau > t)
\]

\[
f(t) = \frac{dF(t)}{dt} = \frac{d}{dt}[1 - S(t)] = -S'(t)
\]

\[
h(t) = \lim_{\Delta t \to 0} \frac{Pr(t + \Delta t > \tau > t|\tau > t)}{\Delta t} = \frac{f(t)}{S(t)}
\]
In our PH model, we assume that the hazard function $h(t)$ has a Weibull distribution, with time-varying covariates:

$$h(t) = h_0(t) \exp(\beta_0 + x(t)'\beta) = pt^{p-1} \exp(\beta_0 + x(t)'\beta)$$

The time-varying covariates $x_{it}$ include: \{number of issued permits within 2-km radius, number of producing wells within 2-km radius, number of issued permits within 1-km radius, and distance to the nearest producing well\}. Start of exposure is defined as when the first permit is issued within 2-km radius, and length of exposure $t$ is defined as:

$$t = \text{current date} - \text{start of exposure}$$

We recover the parameters of this model through maximum likelihood estimation. The likelihood function is:

$$L(\beta|t, x, T) = \prod_{i=1}^{k_1} f(t_i|x_i(\cdot), \beta) \prod_{j=1}^{k_2} Pr(t_j > T|x_j(\cdot), \beta)$$

$$= \prod_{i=1}^{k_1} S(t_i|x_i(\cdot), \beta) \cdot h(t_i|x_i(t_i), \beta) \prod_{j=1}^{k_2} S(t_j|x_j(\cdot), \beta)$$

where $T$ is the termination period in our sample, $k_1$ is the set of houses that are exposed to shale risk during the sample period, and $k_2$ is the set of houses that are not yet exposed to shale risk.
by the end of our sample period. The cumulative hazard function is then given by:

\[ H(t) = \int_0^t h(u)du \]

Figure 2 describes the distributions of cumulative hazard rates across years. The mass point at 0 indicates the houses that have not yet been exposed to any shale risk (i.e., no permit issued inside a 2-km radius). The mass point at 2 indicates houses that have had a failure (a producing well appears within 1-km radius).

The results of the PH model estimation are reported in Table 1:
Table 1: Estimation Results of PH Model

<table>
<thead>
<tr>
<th>PH Coefficient</th>
<th>Number of Producing Wells within 2000m</th>
<th>Number of Issued Permits within 1000m</th>
<th>Number of Issued Permits within 2000m</th>
<th>Distance to the Nearest Producing Well</th>
<th>cons</th>
<th>log(p)</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>-0.00825***</td>
<td>0.0919***</td>
<td>0.00447***</td>
<td>-0.00427****</td>
<td>1.210***</td>
<td>-0.209***</td>
<td>1034756</td>
</tr>
<tr>
<td></td>
<td>(-5.86)</td>
<td>(79.87)</td>
<td>(5.85)</td>
<td>(-295.02)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

$t$ statistics in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

4.5 Summary Statistics

We conclude this section by describing the characteristics of households with and without subprime mortgage rates. In Table 2 we see that a subprime mortgage is more often associated with minority status (black or Hispanic, in particular) and with lower income.

Table 2: Characteristics of Households

<table>
<thead>
<tr>
<th></th>
<th>HHs with subprime</th>
<th>HHs without subprime</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total</td>
<td>10,892</td>
<td>84,360</td>
</tr>
<tr>
<td>American Indian / Alaska Native</td>
<td>0.59% (64)</td>
<td>0.51% (433)</td>
</tr>
<tr>
<td>Asian</td>
<td>2.58% (281)</td>
<td>4.25% (3583)</td>
</tr>
<tr>
<td>Black</td>
<td>14.88% (1621)</td>
<td>6.08% (5131)</td>
</tr>
<tr>
<td>White</td>
<td>69.97% (7621)</td>
<td>67.28% (56761)</td>
</tr>
<tr>
<td>Other</td>
<td>11.98% (1305)</td>
<td>21.88% (18452)</td>
</tr>
<tr>
<td>Hispanic</td>
<td>27.1% (2952)</td>
<td>11.43% (9640)</td>
</tr>
<tr>
<td>Female</td>
<td>63.56% (6923)</td>
<td>67.86% (57247)</td>
</tr>
<tr>
<td>Income (mean, in $1000)</td>
<td>77.14</td>
<td>93.18</td>
</tr>
</tbody>
</table>

In Table 3, we compare the number of different types of loans issued by different types of lenders. In particular, it has been suggested that “national” lenders may have simply pulled out of shale dependent areas as shale risk became an issue.27

27 We include the following lenders as “national”: Bank of America, Wells Fargo, Starkey Mortgage, Coun-
Table 3: Types of Loans Issued by Different Lenders

<table>
<thead>
<tr>
<th>tranyear</th>
<th>Subprime</th>
<th>Non-subprime</th>
<th>Subprime</th>
<th>Non-subprime</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Local</td>
<td>National</td>
<td>Local</td>
<td>National</td>
</tr>
<tr>
<td>2003</td>
<td>0</td>
<td>0</td>
<td>226 (72%)</td>
<td>86</td>
</tr>
<tr>
<td>2004</td>
<td>49 (91%)</td>
<td>5</td>
<td>448 (69%)</td>
<td>202</td>
</tr>
<tr>
<td>2005</td>
<td>290 (85%)</td>
<td>53</td>
<td>920 (64%)</td>
<td>527</td>
</tr>
<tr>
<td>2006</td>
<td>675 (79%)</td>
<td>180</td>
<td>1642 (55%)</td>
<td>1346</td>
</tr>
<tr>
<td>2007</td>
<td>312 (62%)</td>
<td>192</td>
<td>1864 (46%)</td>
<td>2155</td>
</tr>
<tr>
<td>2008</td>
<td>198 (49%)</td>
<td>209</td>
<td>2095 (51%)</td>
<td>2010</td>
</tr>
<tr>
<td>2009</td>
<td>160 (75%)</td>
<td>52</td>
<td>2065 (58%)</td>
<td>1485</td>
</tr>
<tr>
<td>2010</td>
<td>96 (88%)</td>
<td>13</td>
<td>2574 (62%)</td>
<td>1551</td>
</tr>
<tr>
<td>2011</td>
<td>144 (89%)</td>
<td>18</td>
<td>2467 (70%)</td>
<td>1043</td>
</tr>
<tr>
<td>2012</td>
<td>117 (91%)</td>
<td>12</td>
<td>1798 (75%)</td>
<td>596</td>
</tr>
</tbody>
</table>

In fact, we find that this was not the case. Figure 3 compares the difference in the percentage of all shale loans written by local and national lenders in Tarrant Co. each year to the same difference for non-shale loans, we find an almost identical time path. We do see that local lenders issued a much smaller percentage of all loans in the depths of the financial crisis, but that this reduction was proportional across shale and non-shale loans.

Figure 3: % Loans Local Lenders - % Loans National Lenders (Shale v. Non-Shale)

Figure 4 shows that, as shale development became more pervasive in Tarrant Co., loans with shale risk (unsurprisingly) outpaced those without.

For the remainder of the paper, we focus on loan pricing, but it is worthwhile to look briefly at whether shale gas exposure led to loan applications being rejected altogether. Figure 5 looks at the probability of loan denials. We see that, early in the period, a greater likelihood of denial

was associated with a non-shale risk loan, while later in the period, the opposite is true. We are not, however, able to disentangle whether this is the result of lenders becoming more wary of shale gas or simply that there were more homes with shale gas exposure in the pool of mortgage applications. We therefore focus our attention on loan pricing, where we are able to differentiate between these two effects.

Finally, many of our conclusions will focus on the house’s water source (i.e., piped v. private groundwater well). We demonstrate in Table 4 that groundwater dependent houses are larger, newer, on bigger plots, and are occupied by higher-income individuals on average.
<table>
<thead>
<tr>
<th>Household Characteristics w.r.t. House’s Water Source</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>HHs with Piped Water</strong></td>
</tr>
<tr>
<td><strong>Total</strong></td>
</tr>
<tr>
<td>American Indian / Alaska Native</td>
</tr>
<tr>
<td>Asian</td>
</tr>
<tr>
<td>Black</td>
</tr>
<tr>
<td>White</td>
</tr>
<tr>
<td>Other</td>
</tr>
<tr>
<td><strong>Income (mean, in $1000)</strong></td>
</tr>
<tr>
<td>Hispanic</td>
</tr>
<tr>
<td>Female</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Houses with Piped Water</th>
<th>Houses with Ground Water</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lot Size (mean)</td>
<td>9738.43</td>
</tr>
<tr>
<td>Year Built (mean)</td>
<td>1984</td>
</tr>
<tr>
<td>House Size (mean)</td>
<td>2020.58</td>
</tr>
<tr>
<td>Number of Bathroom (mean)</td>
<td>2.01</td>
</tr>
<tr>
<td>Number of Bedroom (mean)</td>
<td>3.26</td>
</tr>
</tbody>
</table>

5 Theory

The theory for analyzing the decisions made by lenders with respect to various risks and lending rates is similar to that used to describe the trade-offs made by workers choosing amongst risky jobs. This is the idea behind wage hedonics, where the compensating differentials associated with different job attributes are used to measure their values (Viscusi 1993). The idea even appears in Adam Smith’s seminal work (Smith 2015), where risky or unpleasant jobs are noted as commanding a premium. The problem has a simple graphical interpretation, which is described in the following figure. Firms are described by a series of iso-profit curves, denoted by $\pi_A$ and $\pi_B$. Along each curve, a lower risk must be accompanied by a lower wage in order for a constant level of profit to be maintained. Firm A is better at providing a low risk environment and is able to pay a higher wage when risks are low compared with firm B.

Workers face similar trade-offs, which are described by iso-expected utility curves $EU_0$ and $EU_1$. In particular, each worker is willing to accept a higher risk in exchange for a higher wage payment, with their willingness to do so, or marginal rate of substitution between risk and compensation, being summarized by the slope of the iso-expected utility curve.

As in other hedonic applications, the allocation of workers to jobs involves a sorting process whereby workers match with firms that yield them the highest expected utility, given firms’ wage offers described by their iso-profit curves. In particular, the best that worker #0 (who requires relatively little in terms of compensation in exchange for taking on more risk) can do is to match with firm A, which requires relatively small increases in wages in exchange for more risk in order to hold profits constant. Worker #0 ends up netting $(R_0, W_0)$ in hedonic
equilibrium. Worker #1 requires more compensation in exchange for taking on additional risk (i.e., a steeper indifference curve) and ends up choosing to match with firm B, yielding \((R_1, W_1)\). With continuously distributed workers and firms, the hedonic wage function describes the set of tangency points between the iso-profit functions of firms and iso-expected utility functions of workers.

We use \(U(W)\) to represent the utility from wage \(W\) if a worker is in the healthy state and \(V(W)\) to represent the utility from the same wage in an injured state; \(p\) represents the probability of that injury. \(EU\) represents expected utility:

\[
EU = (1 - p)U(W) + pV(W)
\]

Taking the total differential of expected utility yields:

\[
(1 - p)U'(W)dW + pV'(W)dW - U(W)dp + V(W)dp = 0
\]

With some re-arranging, we derive an expression for the slope of an iso-expected utility curve:
\[
dW[(1-p)U'(W) + pV'(W)] + dp[V(W) - U(W)] = 0
\]

\[
\left. \frac{dW}{dp} \right|_{dEU=0} = -\frac{EU_p}{EU_W} = \frac{U(W) - V(W)}{(1-p)U'(W) + pV'(W)}
\]

EU will be positively sloped if \(U(W) - V(W) > 0\) (i.e., utility from a given wage is greater in the non-injured state) and if \(U'(W), V'(W) > 0\) (i.e., utility is increasing in wages regardless of the injury state). 2nd order conditions require that curves have convexity as illustrated. Wage hedonic techniques use the slope of estimated relationship between risk and wage to recover \(\frac{dW}{dp} \mid_{dEU=0}\).

In the case of banks issuing mortgages, there are two forms of risk that we consider: (i) shale \((p_1)\) and (ii) income \((p_2)\). Shale risk refers to the list of reasons that shale development might put a house into technical default according to the standard mortgage guidelines described in the introduction. Foremost is the risk of any outcome that might detract from the house’s value as a residential structure. Income risk refers to the standard risk that liquidity constrained individuals will suffer income shocks that make it impossible for them to make timely mortgage payments.

Shale risk and income risk combine to create default risk, which is ultimately what concerns the lender. Perceived default risk faced by lender \(i\) is \(D_i = D_i(p_1, p_2)\), where \(0 \leq D_i \leq 1 \forall i\) and \(\frac{\partial D_i}{\partial p_j} \geq 0 \forall i\) and for \(j = 1, 2\). Assume that \(D_i(p_1, p_2)\) is quasi-convex.

Firm \(i\)’s expected profit from making a loan with risks \(p_1\) and \(p_2\) is given by:

\[
E\Pi_i = (1 - D_i(p_1, p_2))\pi_i^G(\sigma) + D_i(p_1, p_2)\pi_i^B
\]

where \(\pi_i^G(\sigma)\) measures the profit associated with the ‘good’ state (i.e., in which there is no foreclosure) when the mortgage rate is \(\sigma\) (i.e., the actual rate or an indicator for subprime or not). \(\pi_i^B(\sigma)\) measures expected profits in the bad state in which a foreclosure takes place and the lender takes control of the property, selling it for a loss. Taking the total derivative of expected profit with respect to the two forms of risk and rearranging, we can find the change in \(p_2\) associated with a small change in \(p_1\) that would hold \(E\Pi_i\) fixed.

\[
\left. \frac{d\sigma}{dp_j} \right|_{d\Pi_i=0} = \frac{\partial E\Pi_i}{\partial p_j} \left|_{d\Pi_i=0} \right. = \frac{\partial E\Pi_i}{\partial \sigma} \left[ \pi_i^G(\sigma) - \pi_i^B(\sigma) \right] \left(1 - D_i(p_1, p_2)\right) \pi_i^B(\sigma) \frac{\pi_i^G(\sigma)}{\pi_i^B(\sigma)} \quad j = 1, 2
\]
This final term represents the negative of the slope of the iso-expected profit curve. As such, we can recover firm \( i \)'s willingness to take on additional income risk in exchange for a one-unit reduction in shale risk by taking the ratio of the two hedonic gradients. Under our assumptions, this will be invariant to how we define the lending rate, and we can simply use a probability of subprime rather than an exact rate, which is useful because we do not see the exact rate unless it is a subprime.\(^{28}\) We therefore learn about the lender’s willingness to trade-off one type of risk for another holding the overall perceived default risk constant.

6 Empirical Model

6.1 Probability of Subprime — Local Logit Regression

Using a binary indicator variable for a high-priced mortgage as our proxy for subprime, we would like to measure the way in which the likelihood of a subprime mortgage varies with different risk variables without imposing a great deal of structure. Following on the model of Bajari and Benkard (2005), we allow the data to speak to the shape of this equilibrium hedonic function (in two different dimensions of risk) and recover a flexible representation of lender preferences that characterizes the distribution of heterogeneity. We then explore how these preferences vary over time and with water source.

Parametric regression models (such as the probit and logit) are commonly used to study binary dependent variables, but these models impose restrictive functional form assumptions. Semi-parametric binary choice estimators (single-index models) relax these restrictions but they effectively reduce the heterogeneity in the \( X \) characteristics to a single dimension. This restricts the interaction between covariates — specifically, the ratio of two marginal effects does not depend on \( X \) in the single-index model. Because we are interested in heterogeneity in the tradeoffs between two types of risk (income risk and shale risk), we instead follow Frölich (2006) to perform non-parametric regression for binary dependent variables (local likelihood logit estimation).

\(^{28}\)Since 2004, the Federal Reserve Board required lenders to collect and report the spread between the annual percentage rate (APR) on a loan and the yield on Treasury securities of comparable maturity if the spread is equal to or greater than 3.0 percentage points for a first-lien loan. In December 2008, the Board published an amendment to this rule which requires the lenders to report the spread if it is equal to or greater than 1.5 percentage points for a first-lien loan. In order to make our estimates consistent across years, we define the threshold to be 3.0 percentage points across all our sample periods.
The local likelihood logit estimator is:

\[ \hat{E}[Y \mid X = x] = \frac{1}{1 + e^{-x'\hat{\theta}_x}} \]

where

\[ \hat{\theta}_x = \arg \max_{\theta_x} \sum_{i=1}^{n} \left( Y_i \ln \left( \frac{1}{1 + e^{-X_i'\theta_x}} \right) + (1 - Y_i) \ln \left( \frac{1}{1 + e^{X_i'\theta_x}} \right) \right) K_H(X_i - x) \]

The regressors include shale risk (defined as cumulative hazard rate), income risk (defined as the debt-to-income ratio), housing characteristics (lot size, house size, number of bathroom, number of bedroom and year built of a property), loan attributes (yield of 30-year treasury bond, and whether the rate is fixed or not), and year dummies.

The kernel weight \( K_H(X_i - x) \) is computed as:

\[ K_{h,\delta,\lambda}(X_i - x) = \prod_{q=1}^{q_1} \kappa \left( \frac{X_{q,i} - x_q}{h} \right) \prod_{q=q_1+1}^{q_2} \delta_{|X_{q,i} - x_q|} \prod_{q=q_2+1}^{Q} \lambda_{X_{q,i} \neq x_q} \]

The kernel function measures the distance between \( X_i \) and \( x \) for each variable through one of three components, depending upon the particular type of variable: continuous regressors (the first term), ordered discrete regressors (the second term) and unordered discrete regressors (the third term). In our application, continuous regressors include cumulative hazard rate, debt-to-income ratio, lot size, house size, and yield of 30-year treasury bond; ordered regressors include number of bathrooms, number of bedrooms, and the age of each property; non-ordered regressors include year dummies, and whether the rate is fixed or not.

Values for the bandwidth and hyper-parameters \((h, \delta, \lambda)\) in the kernel function are obtained from cross-validation, and the cross-validation criterion is based on maximizing the leave-one-out fitted likelihood function:
\[ \text{CROSSVAL}(h, \lambda, \delta) = \sum_{i=1}^{n} Y_i \ln g(X_i, \hat{\theta}_{-X_i}|h,\delta,\lambda) + (1 - Y_i) \ln(1 - g(X_i, \hat{\theta}_{-X_i}|h,\delta,\lambda)) \]

where \( g(\cdot) \) is our local likelihood logit estimator.

We are interested in recovering banks’ willingness to trade income risk for shale risk, which is revealed in the ratio of the two hedonic gradients, where

\[
\rho(x) = \frac{\frac{\partial E[Y|X=x]}{\partial x_1}}{\frac{\partial E[Y|X=x]}{\partial x_2}} = \frac{\dot{\theta}_1(x)}{\dot{\theta}_2(x)}
\]

Unlike the single index models, this ratio is a flexible function of the regressors, \( x \). The negative of this ratio defines the negative of the slope of the expected iso-profit curve drawn in (shale risk, income risk)-space. Figure 7 illustrates a simple case with linear iso-profit curves.

Figure 7: Iso-profit Curves

A higher value suggests that iso-curve has become steeper, implying that lenders require a
larger reduction in income risk in order to accept another unit of shale risk, holding expected profits constant.

7 Results

7.1 Simple Logit Analysis

We begin with a simple logit specification to explore the determinants of subprime status. In particular, we model the likelihood of subprime as a function of housing attributes, loan attributes (fixed or variable rate, and yield curve of 30-yr treasure bond), year dummies, the debt-to-income ratio (i.e., income risk), and the cumulative hazard rate (shale risk). We estimate this model separately for two time periods – preceding and following the financial crisis (2003-2008 and 2010-2012). We expect that relative concerns of lenders over various sources of default risk might have changed after the financial crisis; in particular, the discussion in Section 3 suggests that policy makers placed increasing attention on shale risk beginning in 2010. While estimates based on the simple logit are generally insignificant, point estimates do behave as expected. Focusing on columns (1) and (2) of Table 5, the parameters on both forms of risk both increase in magnitude following the financial crisis, with that on shale risk becoming statistically significant. The lender’s average willingness to trade-off income risk for shale risk (i.e., the ratio of the shale risk to income risk parameters) rises substantially.

We next differentiate additionally by water source (i.e., groundwater v. piped water). Muehlenbachs et al. (2015) demonstrated that housing markets are particularly prone to capitalize the risks of nearby shale gas development when houses are dependent upon groundwater. We may therefore expect lenders to be particularly aware of shale risk when houses are groundwater dependent, and that this concern may have grown after the financial crisis. We find that this is indeed the case – we recover a large and statistically significant increase in the parameter on shale risk for groundwater dependent properties following the financial crisis, suggesting a large increase in lenders’ willingnesses to take on additional income risk to avoid shale risk on these properties. The same is not true of houses that rely on piped water, where the ratio of income risk to shale risk actually goes down (although all piped water house risk parameters are statistically insignificant).

While the simple logit results suggest that lenders may have become increasingly concerned about shale risk following the financial crisis, and that this was particularly true for groundwater dependent houses, the strong functional form restrictions placed on the model by the simple logit
### Table 5: Simple Logit Regression Results

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Debt-to-Income Ratio</td>
<td>0.0242</td>
<td>0.0717</td>
<td>0.0399*</td>
<td>0.0918</td>
<td>0.00291</td>
<td>0.0566</td>
</tr>
<tr>
<td></td>
<td>(1.67)</td>
<td>(1.55)</td>
<td>(2.15)</td>
<td>(1.55)</td>
<td>(0.12)</td>
<td>(0.76)</td>
</tr>
<tr>
<td>Shale Risk</td>
<td>0.0124</td>
<td>0.115*</td>
<td>-0.00140</td>
<td>0.174**</td>
<td>0.0356</td>
<td>0.0762</td>
</tr>
<tr>
<td></td>
<td>(0.60)</td>
<td>(2.39)</td>
<td>(-0.05)</td>
<td>(2.89)</td>
<td>(1.17)</td>
<td>(0.93)</td>
</tr>
<tr>
<td>Lot Size</td>
<td>-0.00000339**</td>
<td>0.00000470**</td>
<td>-0.00000637***</td>
<td>0.00000482*</td>
<td>0.00000942</td>
<td>0.00000295</td>
</tr>
<tr>
<td></td>
<td>(-2.94)</td>
<td>(2.94)</td>
<td>(-4.12)</td>
<td>(2.31)</td>
<td>(0.61)</td>
<td>(0.96)</td>
</tr>
<tr>
<td>Age of Houses</td>
<td>0.00915***</td>
<td>0.0153***</td>
<td>0.0134***</td>
<td>0.0280***</td>
<td>0.00631***</td>
<td>0.00874**</td>
</tr>
<tr>
<td></td>
<td>(12.37)</td>
<td>(7.19)</td>
<td>(11.44)</td>
<td>(7.64)</td>
<td>(6.28)</td>
<td>(2.93)</td>
</tr>
<tr>
<td>House Size</td>
<td>-0.000656***</td>
<td>-0.000238**</td>
<td>-0.000586***</td>
<td>-0.0000554</td>
<td>-0.000716***</td>
<td>-0.000475***</td>
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<td></td>
<td>(-23.98)</td>
<td>(-3.15)</td>
<td>(-16.86)</td>
<td>(-0.60)</td>
<td>(-15.24)</td>
<td>(-3.43)</td>
</tr>
<tr>
<td>Bathroom</td>
<td>-0.323***</td>
<td>0.137</td>
<td>-0.308***</td>
<td>0.124</td>
<td>-0.373***</td>
<td>0.0696</td>
</tr>
<tr>
<td></td>
<td>(-9.73)</td>
<td>(1.36)</td>
<td>(-7.50)</td>
<td>(1.03)</td>
<td>(-6.52)</td>
<td>(0.39)</td>
</tr>
<tr>
<td>Bedroom</td>
<td>0.332***</td>
<td>0.0795</td>
<td>0.287***</td>
<td>0.0715</td>
<td>0.392***</td>
<td>0.0845</td>
</tr>
<tr>
<td></td>
<td>(12.26)</td>
<td>(0.90)</td>
<td>(8.31)</td>
<td>(0.63)</td>
<td>(8.86)</td>
<td>(0.56)</td>
</tr>
<tr>
<td>Yield of 30-year TB</td>
<td>-0.162**</td>
<td>-0.165</td>
<td>-0.0311</td>
<td>-0.187</td>
<td>-0.367***</td>
<td>-0.0941</td>
</tr>
<tr>
<td></td>
<td>(-3.17)</td>
<td>(-1.90)</td>
<td>(-0.47)</td>
<td>(-1.73)</td>
<td>(-4.52)</td>
<td>(-0.65)</td>
</tr>
<tr>
<td>Dummy of Loan being Fixed</td>
<td>-2.414***</td>
<td>-0.703***</td>
<td>-2.450***</td>
<td>-0.582*</td>
<td>-2.357***</td>
<td>-0.927**</td>
</tr>
<tr>
<td></td>
<td>(-81.56)</td>
<td>(-3.80)</td>
<td>(-66.42)</td>
<td>(-2.51)</td>
<td>(-47.22)</td>
<td>(-3.02)</td>
</tr>
<tr>
<td>Year Dummies</td>
<td>YES</td>
<td>YES</td>
<td>YES</td>
<td>YES</td>
<td>YES</td>
<td>YES</td>
</tr>
<tr>
<td>_cons</td>
<td>1.483***</td>
<td>-3.349***</td>
<td>0.811*</td>
<td>-4.127***</td>
<td>2.437***</td>
<td>-2.543***</td>
</tr>
<tr>
<td></td>
<td>(6.02)</td>
<td>(-9.07)</td>
<td>(2.55)</td>
<td>(-8.72)</td>
<td>(6.23)</td>
<td>(-4.06)</td>
</tr>
<tr>
<td>N</td>
<td>60490</td>
<td>15961</td>
<td>40083</td>
<td>10613</td>
<td>20407</td>
<td>5348</td>
</tr>
</tbody>
</table>

* $t$ statistics in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$
may constrain our ability to learn about these relationships. In the following subsection, we relax these constraints by employing a flexible local logit specification.

### 7.2 Local Logit Estimation

We estimate the flexible local logit specification paying particular attention to the role of water source. In particular, we use estimated preference ratios \( \rho(x) \) to illustrate lenders’ indifference curves in (shale risk, income risk) space. We begin by taking the distribution of the ratios of debt-to-income (i.e., income risk) and shale risk, as it was defined earlier in the paper. These ratios represent the slopes of lenders’ “indifference curves”, drawn in the space of the two types of risks. We then calculate the kernel weighted averages of all estimated preference ratios at each point in risk space, where the smoothing parameter is set using Silverman’s rule. The subsequent mapping describes the indifference curve map at every point in risk space. In Figure 8 Black arrows are used to describe mean estimates, while the gray arrows describe the 5% and 95% confidence bands calculated from 500 bootstrap iterations.

**Figure 8: Indifference Curve Map**

![Figure 8: Indifference Curve Map](image)
A number of interesting features appear immediately. First, in the period before the financial crisis, the slope of the indifference curves are indistinguishable from zero for both piped and groundwater homes (particularly for the former). Assuming income risk is always considered by lenders to be a “bad,” this would imply that shale risk was not taken into consideration by banks in the earlier period. The same is true for piped water houses as we move into the post-crisis period, but we see a dramatic shift in the indifference curve mapping for ground water homes. In particular, lenders’ indifference curves are now negatively sloped for those houses, and those slopes are statistically significant. Income risk substitutes for shale risk in the post-crisis period for ground water houses – i.e., lenders are willing to take on significantly more income risk to avoid additional shale gas risk.

7.3 Robustness Check

Here we address some concerns that may generate alternative interpretations of our results.

One alternative story is that home buyers are likely to overstate their incomes in the period prior to the financial crisis, leading to an under-reporting of debt-to-income ratios, so the debt-to-income ratio in the first period is downward biased in our data set. To address this concern, below is a comparison between the distribution of debt-to-income ratio between the first and second period. Although the distribution in the pre-crisis period does have a bit more mass at the lower values of the debt-to-income ratio, the differences are not large.

8 Valuing Shale Risk

To value the willingness to pay to avoid shale risk, through taking on additional income risk by lending to individuals with higher debt-to-income ratios, we use transaction and foreclosure
information from Dallas County. Dallas neighbors Tarrant County and has comparable land and population sizes and population density, yet Dallas does not have active oil and natural gas drilling.\textsuperscript{29} Our “back-of-the-envelope” calculation combines the estimated relationships between the debt-to-income ratio and rate of foreclosure, as described in the upper panel of Table 6,\textsuperscript{30} and between the foreclosure rate and transaction value, as described in the lower panel of Table 6. In particular, we observe that a standard deviation increase in the debt-to-income ratio increases the likelihood that a home is foreclosed in the future by 0.255 in Dallas, and a foreclosed home sells for 8\% percent less in Dallas (using the -0.330 coefficient in the lower panel of Table 6), controlling for other house observable characteristics and city and year fixed effects. Using the logarithm of the mean sales value for single family, residential homes located in Dallas County, we calculate the willingness to pay to avoid shale risk using the following formula:

\[
WTP = \frac{\partial V_0}{\partial \text{foreclosure}} P(\text{foreclosure}|X) = \exp(11.736) - \exp(11.736 - 0.255 \times 0.311) = 9,496
\]

where \( V_0 \) is the mean value of a property sold in Dallas County within the sample period. The estimates from Dallas County suggest that an additional unit of shale risk is valued at $9,496, which is roughly 8\% of the mean sales value among all homes sold in Dallas County during the sample period spanning 2003 to 2012.

In order to scale the willingness to pay, we compare that value to mean value earned on an average mortgage issued in Dallas. We approximate that the average mortgage interest rate is roughly 6.7\% and the average loan amount is $151,476. Further, most of the loans in Dallas have 30 year term lengths and lenders earn around $190,513 on the average loan in Dallas. Our willingness to pay comprises roughly 5\% of the profit earned on an average Dallas mortgage.

9 Conclusion

This paper explores the housing market impacts of shale gas development. Previous work has done so using data on the capitalization of shale gas activities into housing prices and hedonic theory to give those estimates a welfare interpretation. In this paper, we come at the question

\textsuperscript{29}In particular, Dallas county is comprised of 909 square miles, 2.5 million people, and a population density of 2,950 individuals per square mile. Tarrant county, on the other hand, is comprised of 902 square miles, 1.8 million people, and a population density of 2,095 individuals per square mile.

\textsuperscript{30}Table 6 also includes estimates using data from Tarrant County to demonstrate similarity across counties. However, note that debt-to-income has a smaller effect on the likelihood of foreclosure and foreclosed homes sell for more, on average, than those located in Dallas.
Table 6: Simple Foreclosure Logit & Hedonic

<table>
<thead>
<tr>
<th>Dependent Variable: Loan Foreclosed Dummy</th>
<th>Dallas</th>
<th>Tarrant</th>
</tr>
</thead>
<tbody>
<tr>
<td>Debt-to-Income</td>
<td>0.25512***</td>
<td>0.22390***</td>
</tr>
<tr>
<td></td>
<td>(0.01860)</td>
<td>(0.01853)</td>
</tr>
<tr>
<td>Gov’t Purchased</td>
<td>-0.12976***</td>
<td>-0.11854***</td>
</tr>
<tr>
<td></td>
<td>(0.03110)</td>
<td>(0.02784)</td>
</tr>
<tr>
<td>Adjustable Rate</td>
<td>0.07815**</td>
<td>0.11530***</td>
</tr>
<tr>
<td></td>
<td>(0.03199)</td>
<td>(0.03257)</td>
</tr>
<tr>
<td>Constant</td>
<td>-4.44095***</td>
<td>-3.33334***</td>
</tr>
<tr>
<td></td>
<td>(0.36005)</td>
<td>(0.42872)</td>
</tr>
<tr>
<td>Observations</td>
<td>61,962</td>
<td>70,399</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Dependent Variable: Log Sales Values</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Beds</td>
<td>-0.050***</td>
<td>-0.065***</td>
</tr>
<tr>
<td></td>
<td>(0.002)</td>
<td>(0.002)</td>
</tr>
<tr>
<td>Baths</td>
<td>0.416***</td>
<td>0.114***</td>
</tr>
<tr>
<td></td>
<td>(0.002)</td>
<td>(0.002)</td>
</tr>
<tr>
<td>House Size</td>
<td>0.000***</td>
<td>0.000***</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
<tr>
<td>Lot Size</td>
<td>-0.000***</td>
<td>0.000***</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
<tr>
<td>Age of House</td>
<td>0.004***</td>
<td>-0.000***</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
<tr>
<td>Foreclosure</td>
<td>-0.333***</td>
<td>-0.232***</td>
</tr>
<tr>
<td></td>
<td>(0.003)</td>
<td>(0.002)</td>
</tr>
<tr>
<td>Constant</td>
<td>10.534***</td>
<td>10.568***</td>
</tr>
<tr>
<td></td>
<td>(0.013)</td>
<td>(0.019)</td>
</tr>
<tr>
<td>Observations</td>
<td>340,628</td>
<td>131,526</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.410</td>
<td>0.674</td>
</tr>
</tbody>
</table>

City Dummies x x
Year Dummies x x

Standard errors in parentheses
* p < 0.05, ** p < 0.01, *** p < 0.001
from a different perspective, focusing instead on the pricing decisions of mortgage lenders. Mortgage lenders bear the risk that borrowers will default – a fact highlighted by a housing market bust and financial collapse less than a decade ago. These swings in the housing market coincided with the U.S. shale boom. By 2010, lenders and policy-makers had become much more mindful of these risks and began to recognize the role that shale gas development might play in default.

We estimate lenders relative preferences for two different types of risk – shale and income. We find strong evidence that lenders became more concerned with shale risk by the period following the financial crisis, and that this is particularly true when we differentiate by the source of houses’ water supply. We might suspect that shale development would place the residential value of a groundwater-dependent property at greater risk. This indeed appears to be the case, with lenders showing a relative distaste for shale risk on loans to these properties.

These results confirm the conclusions of previous research with respect to the expected house price impacts of shale gas development for groundwater homes (Muehlenbachs et al. 2015). Importantly, they do so based on the decisions of agents (i.e., lenders) who make hundreds (possibly thousands) of lending decisions each year. One might worry that homebuyers in hedonic models might lack expertise – the typical individual may only buy a couple of houses over the course of her lifetime, and the cost of information acquisition is high. The same is not true for professional lenders.
References


King, Alvin Thomas, *Discrimination in mortgage lending: A study of three cities*, New York University, Graduate School of Business Administration, Salomon Brothers Center for the Study of Financial Institutions, 1981.


