Housing Price and Population Changes: Growing vs Shrinking Cities

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Abstract

Housing prices in the United States experienced a significant meltdown during the Great Recession. Since then, the housing market has seen a nationwide recovery, even over-heating in some regions. In this paper, we look back at that era and study the possible impact of population changes on housing markets in 2010-2017. Our motivation stems from an important earlier work on the same relationship between population changes and housing values by Glaeser and Gyourko (2005). Their main finding is that there is an asymmetrical reaction in housing prices during the period of a population increase and a population decrease. Our finding is similar to this existing literature even during this very different market situation; housing prices go up at a slower rate when population increases compared to the rate of price decrease when population decreases. That is, housing price changes are more elastic when population increases, due to the durability of housing stock. Our focus here is not to look at how the housing market recovered from the recession per se. Rather, our study is to look at the robustness of the relationship between population changes and their asymmetrical impact on housing prices, through the recovery period after the crash. In addition to the population variable, our model employs a few other factors known to affect housing prices, such as the unemployment rate, the GDP per capita, mortgage rates and the housing supply.

Keywords: asymmetrical housing price reaction, elasticity of price change, housing price, population change

1. Introduction

There have been a number of studies on housing price changes. They find various socioeconomic factors that influence house price changes, such as population, employment growth, income, and interest rates (Glaeser et al. (1992), Black and Henderson (1999), Abraham and Hendershott (1996), Potepan (1996), Jun and Winkler (2002), Fontenla et al. (2009), and Feng et al. (2017)). Population, a main driver for housing demand, is the focus of this study.

Glaeser and Gyourko (2005) analyzed the asymmetric impact of population gains and losses on housing prices using the decadal data from the 1970s until the 1990s. Their premise is that homes can be built rather quickly, but disappear slowly, therefore urban decline is not the mirror image of urban growth. When a city grows, it attracts more population, but since housing can be built reasonably quickly, housing prices do not go up as fast as the population growth rate. On the other hand, when a city’s economy deteriorates, housing stock endures, which means an oversupply of housing stock, and so housing prices drop faster than the reduction in population. They verify that positive economic shocks increase population more than they increase housing prices while negative economic shocks decrease housing prices more than they decrease population. They also suggest that the remaining population in the decayed urban landscape comprises a large portion of the poor and the unskilled labor.

We apply their hypothesis on the relationship between population changes and housing prices using more recent (2010-2017) panel data on Metropolitan Statistical Areas (MSA) that includes cities with growing population as well as the cities with shrinking population. This period happens to coincide with a rapid economic and housing market recovery.

As pointed out by Quigley (1995) and Leung (2003), economic fundamentals are important in understanding housing markets. Although using only year 2010’s data, Lin et al. (2014) were able to analyze a group of macro-economic and demographic factors affecting on housing prices, focusing on differences in various regions (i.e., the 363 MSAs in the US) while assuming that supply is constant.

Moreover, several studies focus on the reasons why the demand for housing stock can be deterred by some
determinants. For examples, Stein (1995), who coined “housing lock-in effect”, pointed out that homeowners who experience a negative house price shock may be unable to move because they may not have sufficient existing equity to finance a new home even they could have found a better job somewhere else. As a result, negative house price shock restricts the job mobility of homeowners and reduce both the demand and supply sides of the housing stock. Glaeser and Gyourko (2005) observe that exogenous forces predicting urban growth will have larger effects on population and smaller effects on housing prices. Conversely, exogenous forces that predict urban decline will have smaller effects on population and bigger effects on housing prices. They also pointed out that the remaining population in the decayed urban landscape comprises a large portion of the poor and the unskilled labor population. It is clear that this group of population does not have financial power to constitute an effective demand for the housing stock. Since the housing stock disappears slowly, the only way for the market to restore the market equilibrium is to drop the price.

Melzer (2017) also shows that homeowners at risk of default face a debt overhang that reduces their incentive to invest in their property. They tend to cut back substantially on home improvements and to reduce their housing consumption in terms of the quality of housing services, even when they appear financially unconstrained. These findings point out that the stifled housing investment could further make negative house price shock even worse due to the lack of housing maintenance, which should further drag down the house price.

When the price has dropped to a certain point, we start to observe some negative equity effects that trigger default and then foreclosure events in the mortgage industry. Economists have developed two theories to explain default risk. The payment-ability theory argues that the risk of default increases when the borrower cannot make the payments on the loan. The equity theory, on the other hand, claims that no borrower would ever activate default option embedded in the mortgage, even if the borrower is unable to make the payments, as long as there is positive equity in the property. Individually, the equity theory makes more intuitive sense. However, the two theories should not be viewed as conflicting but rather be the supplemental of each other. In other words, if the house owner gets stuck in both situations, it is most likely that the default will happen and force the lender to take foreclosure action. The foreclosure auction creates distress sales and decreases not only the price of the subject property but also the properties in the neighborhood. Carter and Gottschalck (2009) also show that homeowners in negative equity were significantly more likely to change their tenure choice from homeownership to renter status in the following period than homeowners that had positive equity during the recent great recession. This change of the tenure choice by the ex-houseowner should further reduce the effective demand for the housing stock.

It has been proven that foreclosures could be contagious in the neighborhood. Lin and Yao (2009) and Towe and Lawley (2013) observe that if one household enters foreclosure, it increases the probability that neighboring households will also foreclose. The contagion could be explained partially by the findings of Currie and Tekin (2011) that foreclosures have negative effects for those living in the same area due to stress and diminished neighborhood quality. The diminished neighborhood quality clearly would downgrade the locational value of a property; hence further damage the overall properties’ price in that area. If this kind of downward spiral could not be stopped, a vicious circle for the housing price drop would trigger even more default of mortgages and foreclosures.

As an inspiration from those literature and unlike Glaeser and Gyourko (2005) however, who concentrated on the dynamics between just the population changes and housing prices, we also look at this relationship using alternative models that include a few additional variables, namely, the unemployment and mortgage rates, which are negatively associated with house price appreciation, the GDP per capita growth rate and the housing supply (represented by building permits) which are known to be positively associated with housing appreciation. By including these additional economic variables in the model, we hope to enhance our understanding of the relationship between some economic indicators for the macroeconomic condition and the changes in house prices.

2. Methodology and Results
2.1 Data
Our data come from several different sources. The Federal Housing Finance Agency (FHFA) publishes quarterly house price indices (HPI) for MSAs using sales prices and appraisal data. We use their fourth quarter housing price index as our annual price level to calculate the annual house price growth rate (HRet). Our population estimates from 2010 to 2017 for each MSA come from the Census Bureau. Our GDP per capita and its growth rates are provided by the Bureau of Economic Analysis (BEA). This GDP data are only available until 2016. Our labor market unemployment rates come from the Bureau of Labor Statistics (BLS). We use December unemployment rates as our annual estimates. We obtained the annual average 30-Year fixed rate mortgage rates in the United States from the St. Louis Federal Reserve’s FRED. Finally, our data on total new building permits issued every year at MSA level are
from the Building Permits Survey provided by the Census Bureau. The merged data have information on 357 MSAs. All these variables and their data sources are summarized below in Table 1.

Table 1. Data source

| Variable            | Source                                                   |
|---------------------|----------------------------------------------------------|
| Real estate prices  | Federal Housing Finance Agency Housing Price Index       |
| Population          | US Census Bureau population estimates                    |
| GDP per capita      | Bureau of Economic Analysis                              |
| Unemployment rates  | Bureau of Labor Statistics                               |
| Mortgage rates      | St. Louis Fed FRED                                       |
| Building Permits    | US Census Bureau building permits survey                 |

2.2 Cross-Sectional Analysis of the House Price and Population Changes

We calculate the house appreciation rate and the population growth rate for each MSA over the sample period as:

\[ \text{CumHRet}_i = \frac{HPI_{i,2017}}{HPI_{i,2010}} - 1, \]  

and

\[ \text{CumPop}_i = \frac{\text{Population}_{i,2017}}{\text{Population}_{i,2010}} - 1. \]

The top and bottom 10 MSAs in housing price appreciation are displayed in Table 2. The top and bottom 10 MSAs in population growth are displayed in table 3.

Table 2. Top and Bottom MSAs in Cumulative Housing Price Appreciation over 2010-2017

|   | MSA                     | CumHR  | CumPop  | MSA                     | CumHR  | CumPop  |
|---|-------------------------|--------|---------|-------------------------|--------|---------|
| 1 | Bend-Redmond, OR        | 90.7%  | 18.5    | City-Hammonton, NJ      | -12.1% | -1.8%   |
| 2 | Merced, CA              | 84.6%  | 6.2%    | Sierra Vista-Douglas, AZ| -10.4% | -5.3%   |
| 3 | Stockton-Lodi, CA       | 81.3%  | 8.5%    | Jacksonville, NC        | -10.4% | 3.7%    |
| 4 | Modesto, CA             | 80.8%  | 6.3%    | East Stroudsburg, PA    | -10.4% | -1.1%   |
| 5 | Reno, NV                | 78.9%  | 9.1%    | Valdosta, GA            | -6.6%  | 3.8%    |
| 6 | Vallejo-Fairfield, CA   | 77.8%  | 7.6%    | Columbus, GA-AL         | -6.6%  | 2.5%    |
| 7 | Port St. Lucie, FL      | 75.6%  |         | Goldsboro, NC           | -6.4%  | 1.1%    |
| 8 | San Jose-Sunnyvale-     | 73.0%  | 8.5%    | Vineland-Bridgeton, NJ  | -5.9%  | -2.7%   |
|   | Santa Clara, CA         |        |         |                         |        |         |
| 9 | Las Vegas-Henderson     | 72.0%  |         | Fayetteville, NC        | -4.0%  | 3.2%    |
|   | -Paradise, NV           |        |         |                         |        |         |
|   | Denver-Aurora-Lakewood, | 12.9%  |         |                         |        |         |
|   |                         | 13.1%  |         |                         |        |         |
| 10| CO                      | 70.9%  |         | Rockford, IL            | -3.8%  | -3.1%   |
Table 3. Top and Bottom MSAs in Cumulative Population Growth over 2010-2017

|   | MSA                                      | CumPo | CumHR | Botto MSA                          | CumPo | CumHR |
|---|------------------------------------------|-------|-------|------------------------------------|-------|-------|
| 1 | The Villages, FL                         | 32.8% | 29.9% | Pine Bluff, AR                     | -9.1% | 7.2%  |
|   | Myrtle Beach-Conway                      |       |       |                                     |       |       |
| 2 | Pine Bluff, AR                           |       |       |                                     |       |       |
|   | Myrtle Beach, SC-NC                      | 22.6% | 8.9%  | Johnstown, PA                      | -7.2% | 7.8%  |
| 3 | Austin-Round Rock, TX                    | 22.5% | 61.8% | Charleston, WV                     | -5.5% | 7.5%  |
| 4 | Midland, TX                              | 20.4% | 46.5% | Sierra Vista-Douglas, AZ           | -5.3% | -10.4%|
| 5 | Greeley, CO                              | 19.8% | 69.8% | Beckley, WV                        | -5.1% | 6.6%  |
| 6 | St. George, UT                           | 19.7% | 41.6% | WV-OH                             | -4.9% | 11.6% |
|   | Cape Coral-Fort Myers                    |       |       |                                     |       |       |
| 7 | Bend-Redmond, OR                         | 19.1% | 68.6% | Danville, IL                       | -4.6% | 7.9%  |
| 8 | Raleigh, NC                              | 17.4% | 26.7% | Wheeling, WV-OH                    | -4.5% | 20.7% |
| 9 | Orlando-Kissimmee                        | 17.3% | 50.0% | Decatur, IL                        | -4.5% | 3.5%  |
| 10| Sanford-Kissimmee                        |       |       |                                     |       |       |

Table 4. Summary Statistics of 2010-2017 Cumulative Changes

| Variable | Obs | Mean | Std. Dev. | Min | Max |
|----------|-----|------|-----------|-----|-----|
| CumHRet  | 357 | 21.6%| 19.3%     | -12.1% | 90.7% |
| CumPop   | 357 | 4.7% | 5.7%      | -9.1% | 32.8% |

The average house price appreciation rate for all MSAs over this seven-year period is 21.6%. The average population growth rate is 4.7%. The distribution of house price growth and population growth are displayed in figures 1 and 2.
Neither the distribution of house price change nor the distribution of population growth is symmetric. 337 out of 357 MSAs experience house price increase over the period while 270 out of 357 MSAs experience population growth. We do not run a cross-sectional regression due to the limited number of observations for MSAs experiencing house price depreciation over the period. However, we can perform regressional analysis with MSA-year level data in the next section.

2.3 Panel Data Analysis

Our base model is displayed in equation (1) controlling for a city’s major economic and demographic factors:

\[ HRet_{i,t} = \beta_0 + \beta_1 \text{PopGain}_{i,t} + \beta_2 \text{PopLoss}_{i,t} + \beta_3 \text{UnEmploy}_{i,t} + \beta_4 \text{GdpGrowth}_{i,t} + \beta_5 \text{Mort}_{t} + u_i + \epsilon_{i,t} \]

(1)

where \( HRet_{i,t} \) is the house price appreciation rate for city \( i \) in year \( t \), \( \text{PopGain} \) equals the population growth rate if the rate is positive, otherwise, it takes the value of 0, \( \text{PopLoss} \) equals the opposite of population growth rate if the rate is negative, otherwise, it equals 0. We also control for unemployment rates, the GDP per capita growth rate, as well as the nation-wide mortgage rates.

As in the general economic theory, we expect to see \( \beta_2 < 0 \) and \( |\beta_2| > |\beta_1| > 0 \), meaning the elasticity of price changes with respect to population changes is higher when population is declining than when population is growing. We would also expect \( \beta_3 < 0 \), \( \beta_4 > 0 \) and \( \beta_5 < 0 \) since lower unemployment rate, higher GDP, and lower mortgage rates are expected to positively affect housing demand. \( u_i \) is the unknown fixed effect intercept for each MSA and the \( \epsilon_{i,t} \) is the error term.

We test these hypotheses along with alternative regression models, such as controlling for new housing supply represented by the number of building permits. Table 5 presents the regression results. They all indicate a stronger effect of the population loss on house prices than of the population gain. One percent increase in population, on average, increases the house price by 0.896% while a 1% loss of population, on average, decreases the house price by 1.515% (model 1). The results are similar to but different in magnitude from Glaeser and Gyourko (2005) where they find a 0.23 and -1.8 elasticity of price change with respect to population gains and to population losses. They use a pool regression on decadal prices. Unlike our models here, their model does not include other variables.
Table 5. Regression results of house price appreciation rate on population change

| Variable         | (1)          | (2)          | (3)          | (4)          |
|------------------|--------------|--------------|--------------|--------------|
| Popgain_{it}     | 0.896(5.44)  | 1.214(8.1)   | 0.717(4.1)   | 0.998(6.27)  |
| Poploss_{it}     | -1.515(-3.77)| -1.384(-3.78)| -1.528(-3.39)| -1.107(-2.77)|
| Unemployment_{it}| -0.016(-30.09)| -0.015(-34.66)| -0.015(-25.68)| -0.014(-28.41)|
| GDPgrowth_{it}   | 0.114(5.45)  |              | 0.121(5.63)  |              |
| Mort_{i}         | -0.008(-3.36)| -0.010(-4.32)| -0.009(-3.7) | -0.011(-4.83)|
| L1.Build_{it}    |              |              | 0.005(2.3)   | 0.008(3.91)  |
| Const_{i}        | 0.152(17.31) | 0.150(17.62) | 0.120(6.83)  | 0.100(6.32)  |

MSA dummy Y Y Y Y
Adjusted $R^2$ 0.51 0.53 0.51 0.53
# of Obs. 2499 2856 2351 2708

Unemployment rates and mortgage rates are negatively associated with house price appreciation. A GDP per capita growth rate is positively associated with housing appreciation. Note that when the GDP per capita growth is included in the regression, we have one fewer annual observation due to data availability. Finally, we include a building permit variable to control for housing supply (models 3 and 4). We use a one-period lag variable since it usually takes one year or longer to build new houses. Surprisingly, the results show a positive relationship between the housing supply and the house price appreciation. The positive effect could be driven by data asymmetry given the fact that our sample period covers mostly a boom market in real estate and builders are simultaneously building more houses.

3. Conclusion

We looked at the impact of population changes on housing prices in 357 U.S. Metropolitan areas in 2010-2017. Our motive was to test the well-known thesis of Glaeser and Gyourko (2005) that when population decreases, housing prices drop faster than the rate of housing appreciation in case of population increases. Their premise is that since housing is a durable good, when the demand goes down, as in the case of an urban decline, its value will drop accordingly; on the other hand, in the case of an urban growth, the supply of housing can catch up more easily, therefore its value will not go up as fast.

Our regression results all indicate similar asymmetrical housing price reactions to the changes in population; stronger effect of population loss on house prices than of population gain. One percent increase (+1%) in population, on average, increases the house price by 0.896% while one percent loss (-1%) of population, on average, decreases the house price by 1.515%. The results are similar to but different in magnitude from Glaeser and Gyourko (2005) where they find a 0.23 and -1.8 elasticity of price change with respect to population gains and to population losses. Note that they use a pool regression on decadal prices whereas we use annual price data.

We tested alternative regression models as well, adding the variables of unemployment rates, GDP growth, and mortgage rates in addition to new housing supply represented by the number of building permits. As before, the elasticity of price changes with respect to population changes is higher when population is declining than when population is growing. In addition, as the general economic theory would predict, our analysis also shows that lower unemployment rate, higher GDP and lower mortgage rates positively affect housing demand, hence housing prices.

Finally, we include a building permit variable to control for housing supply. We use a one-period lag variable since it usually takes one year or longer to build new houses. Surprisingly, the results show a positive relationship between the housing supply and the house price appreciation.

Our best explanation of this unexpected positive effect is that our sample period covers mostly a boom market in real estate right after the Great Recession, a period when recovery was taking hold fast even as builders were simultaneously building more houses.

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