

Motivation and Contribution

- Recent advances in shale development have produced both positive and negative outcomes for local communities, with higher employment and income known to be the most significant effects.
- Despite the stated importance of the distribution of economic gains among local populations in previous literature, adequate research on the shale boom's impact on inequality and affordability does not exist.
- I employ the difference-in-difference (DiD) method to study the unintended social consequences of the hydraulic fracturing boom in Oklahoma, the second-largest producer of oil and gas in the country, over the period of 2004-2017.

Methodology

- The difference-in-difference (DiD) method: identifying a specific intervention or treatment and then comparing the difference in outcomes after and before the intervention for groups affected by it to the same difference for unaffected groups.
- I define the treatment group to be counties that experienced shale extraction and the treatment to be the shale development boom that happened in 2008.
- To avoid endogeneity, shale counties were selected based on their geographic location, as it is common in the literature.
- To avoid spillover effect, I construct and report results for a second sample that excludes non-shale counties that share a border with a shale county.
- Moreover, I apply same estimation to a third data set with all the counties except the only two metropolitan counties, Oklahoma county and Tulsa county, to avoid their excessive influence on the regression results.
- The DiD estimator is based on a strong identifying assumption: the availability of a treatment and a control group that would have had a similar trend without the treatment. To address this issue, I include a set of covariates; these covariates are used to describe how the average effect of the treatment varies with changes in observed characteristics. I also follow the literature (e.g., Linden and Rockoff, 2008; Caselli and Michaels, 2013) to examine preexisting differences in counties' characteristics using a cross-sectional estimator.
- Given the similarity of pre-treatment trends in shale and non-shale counties, I then use a linear DiD model to estimate shale boom effect on local communities:

$$\ln(Outcome_{ct}) = \beta_0 + \beta_1 Shale_c + \beta_2 Post2008_t + \beta_3 Shale_c * Post2008_t + \mu_c + v_t + \epsilon_{ct}$$

The dependent variable, $\ln(Outcome_{ct})$, represents outcomes of interest: housing price index, Gini coefficient, and housing affordability index. β_3 is the coefficient of interest which measures the average shale development effect on shale counties by differencing the changes in outcomes in shale counties after 2008 with non-shale counties.

- I included income per capita, housing density and population density using the Census 2010 data as a set of control variables that allows counties with different characteristics to have different outcome. To make the specification even more robust, I follow the literature to include control variables for neighboring counties as well:

$$\ln(Outcome_{ct}) = \beta_0 + \beta_1 Shale_c + \beta_2 Post2008_t + \beta_3 Shale_c * Post2008_t + \delta_1 X_c + \delta_2 C_c + v_t + \epsilon_{ct}$$

where X_c is a set of control variables capturing counties' observable characteristics and C_c is neighboring counties' observable characteristics.

Results

	Dependent Variable								
	Housing Price Index			Gini Coefficient			Housing Affordability Index		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Post 2008	0.00329 (0.01)	0.00479 (0.01)	0.00291 (0.01)	0.382*** (0.05)	0.382*** (0.05)	0.313*** (0.04)	0.158*** (0.02)	0.157*** (0.02)	0.160*** (0.02)
Shale County	0.103*** (0.01)	0.0243 (0.02)	0.0340 (0.02)	0.363*** (0.04)	0.00889 (0.06)	0.00652 (0.06)	-0.0248 (0.02)	-0.0139 (0.02)	-0.0257 (0.02)
Shale x Post 2008	0.0554*** (0.01)	0.0553*** (0.01)	0.0580*** (0.01)	0.0549 (0.05)	0.0549 (0.05)	0.0481 (0.05)	-0.0546** (0.02)	-0.0547** (0.02)	-0.0633*** (0.02)
Constant	4.178*** (0.01)	4.126*** (0.02)	4.211*** (0.13)	-2.627*** (0.04)	-2.184*** (0.05)	-2.460*** (0.36)	6.636*** (0.01)	6.673*** (0.02)	6.390*** (0.17)
County Fixed Effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Year Fixed Effect	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
County Set of Control Variables	No	Yes	Yes	No	Yes	Yes	No	Yes	Yes
Neighboring County Set of Control Variables	No	No	Yes	No	No	Yes	No	No	Yes
Number of Observations	684	684	628	686	686	630	684	684	628
Adj. R ²	0.834	0.323	0.352	0.682	0.562	0.575	0.392	0.434	0.477

Standard errors are in parentheses
*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Data

- For housing values: county-level annual housing price index (HPI) from the Federal Housing Finance Agency (FHFA)
- For inequality: Internal Revenue Service (IRS) data for the Gini index
- For affordability: the median household income from the IRS data and the real housing price index from FHFA for the frequently used housing affordability index, the income-to-housing price ratio
- For counties' characteristics: Census 2010 data (income per capita, population density, and housing density)

Conclusion

- The results suggest that the shale boom was associated with appreciation in housing values for 5.5% and a decrease of %6 in affordability in the state of Oklahoma for shale counties compared to non-shale counties.
- Although previous literature provides evidence for higher employment and income in shale counties due to the boom, the estimation fails to find any statistically significant effect on inequality.
- The results are consistent across three different samples, with or without covariates for shale counties or their neighbors.

References

- Francesco Caselli and Guy Michaels. "oil windfalls improve living standards? Evidence from Brazil". In: American Economic Journal: Applied Economics 5.1 (2013), pp. 208-38.
- Leigh Linden and Jonah E Rockoff. "Estimates of the impact of crime risk on property values from Megan's laws". In: American Economic Review 98.3 (2008), pp. 1103-27.