Data Science for All (DS4A) Investigating Technology-Driven Gentrification in Miami, Austin, and Atlanta Team 28 Kiersa Sanders, Adrian Ricketts, Natasha Urbany, Ericka Howard

<u>Investigating Technology-Driven Gentrification in Atlanta, Austin, and Miami</u>

Overview

Cities across the United States are continuously evolving, with some cities undergoing dramatic changes due to the process of gentrification. Generally speaking, gentrification is a "process where wealthy college-educated individuals begin to move into poor or working-class communities, often originally occupied by communities of color." Research has shown that "gentrification usually leads to negative impacts such as forced displacement, a fostering of discriminatory behavior by people in power, and a focus on spaces that exclude low-income individuals and people of color."

In order to combat these negative impacts of gentrification, it is critical that local law and policymakers understand the factors that contribute to gentrification and the socio-economic. This report aims to equip those groups with insights into how the advent of the digital age and the movement of technology workers around the United States has led to gentrification in emerging technology cities such as Miami, Austin, and Atlanta.

Data Overview

To examine the impacts of technology-driven gentrification, this report uses time-series data on Median Home Value Growth, Technology Worker Population Growth and Salary, Technology Worker Education and Salary Levels, and Luxury Coffee Shop Growth. A linear regression model was used to both explore the relationship between variables and also generate a "gentrification score" that enabled the team to explore the following hypotheses:

- Growth in the population of affluent tech workers in cities like Miami, Austin, and Atlanta contributes to increases in housing prices making neighborhoods inaccessible and displacing pre-existing communities
- Growth in the population of tech workers leads to a rise in the presence of luxury coffee shops; this may be because there is a new demand for goods that are at a higher price point than previously experience

The team collected, cleaned, and modified data where appropriate to create a cohesive data frame. The final dataframe relied on zip code as the basic unit for data merging. This data frame was then used as the input for regression analysis. Based on the outputs of the regression analysis, a gentrification equation where select variables are weighted based on their impact was created. See below for more information on the synthesized data frame:

¹ National Geographic Society. "Gentrification." National Geographic Society, September 9, 2019. https://www.nationalgeographic.org/encyclopedia/gentrification/.

² Chong, Emily. "Examining the Negative Impacts of Gentrification." *Georgetown Law*, Georgetown Journal on Poverty Law & Policy, 17 Sept. 2017, https://www.law.georgetown.edu/poverty-journal/blog/examining-the-negative-impacts-of-gentrification/.

Home Value Growth Data

Source: Redfin

Description: Data includes data on the number of new listings, number of sales pending sales, number of homes sold, listing price, median sale price, and other information.

This data provides a weekly view of housing figures and encompasses data from Q1 2012-Q3 2021. For the purposes of our analysis, the team chose to pull information for Miami, Florida, Austin, Texas, and Atlanta, Georgia as these cities have experienced shifts in people movement and growth in their technology sectors.

Field	Data Type	Description	
zip	integer	5-digit zip code	
latitude	float	Geocoordinate assigned to select zip code, original latitude information obtained from US Census data	
longitude	float	Geocoordinate assigned to select zip code, original longitude information obtained from US Census data	
parent_metro_region	object	Region name (e.g., Los Angeles, CA, metro area)	
median_sale_price	integer	Sales price data for single-family residential property types in US dollars (e.g., \$495K)	
percent_change_home_pri ce	integer	Percent change in median home price for single-family residential properties between Q2-Q3 2021 This field was created by the tame based on median_sale_price and time-series data	
period_begin	datetime	Month and year values (the period used for analysis account for 09/01/2021-11/30/2021)	

Technology Worker Population Growth Data

Source: 2019 Census Data

Description: Data includes a comprehensive view of key demographic indicators such as the total population over the age of 16, the number of individuals available to participate in the workforce

The data provides a view of key demographic indicators

total_population_over_16	integer Total population over 16 years of age		
labor_force	integer	Number of individuals available to participate in workforce	
percent_in_labor_vs_popu lation	integer	Percentage comparison of Labor Force vs the Total Population over 16	

unemployment	integer	Percentage of individuals who are unemployed
information_industry_pop ulation	integer	Number of workforce people in the Professional, Scientific, and Tech industry
information_as_percent_o f_labor	integer	Percentage comparison of the number of workforce people in the Professional, Scientific, and Tech industry vs Labor Force

Technology Worker Education and Salary Levels

Source: Miami-Fort Lauderdale-West Palm Beach, FL - May 2020 OEWS Metropolitan and Nonmetropolitan Area Occupational Employment and Wage Estimates (as a narrowed example)

Description: Data includes a comprehensive view of industry workers and their wages based on target cities/locations. The data provides an overall view and comparison of tech industry wages as it correlates to location and other key demographic indicators.

occupation	string	Category of occupation, the data collected was narrowed down using the Computer and Mathematical Occupations field	
total_employment	float	Total number of technology industry workers	
jobs_per_1000	float	Total number of technology industry jobs per 1,000	
annual_mean_salary	float	Mean salary amount for technology industry workers (Miami, Austin, Atlanta)	
annual_median_salary	float	Median salary amount for technology industry workers for select locations (Miami, Austin, Atlanta)	

Luxury Coffee Shop Store Growth

Source: Starbucks Locations

Description: Data includes comprehensive data that provides insight into the number of starbucks stores active within certain zip codes in 2021

total_stores	Sum of the number of stores present within a zip code (the sum was created by grouping by zip, latitude,
	longitude, etc.)

Model and Analysis

To complete the analysis on combined data, we joined each dataset on zip code and selected the most recent data. Using the combined data, we conducted multiple linear regressions to understand the relationships between the variables we hypothesized would have the most impact on gentrification. The regression was used to inform the weights used to create the gentrification score.

Regression Modeling

We selected median home price as the independent variable. Percentage change in home price, information jobs as a percentage of the total labor market, total Starbucks stores, and annual median salary were the dependent variables. The objective of our regression was to determine the relationship between our gentrification indicators and home prices.

	OLS	Regres	sion R	esults				
Dep. Variable:	median sale	price	R-sq	uared:		0.36	 52	
Model:		OLS	_	R-squared:		0.34	16	
Method:	Least Sq	uares	F-st	atistic:		22.9	98	
Date:	Tue, 15 Mar	2022	Prob	(F-statist	ic):	4.71e-1	15	
Time:	13:	09:55	Log-	Likelihood:	•	-2503	. 0	
No. Observations:		167	AIC:			5016	5.	
Df Residuals:		162	BIC:			5032	2.	
Df Model:		4						
Covariance Type:	nonr	obust						
0.975] 								
Intercept 69e+06		2.01	e+06	8.54e+05	2.354	0.020	3.24e+05	3.
percent_change_home 04e+05	e_price	6.601	e+05	7.27e+04	9.081	0.000	5.17e+05	8.
annual_median_salar 2.620	Ту	-18.	3810	10.635	-1.728	0.086	-39.383	
information_as_pero 25e+05	cent_of_labor	2.36	e+04	5.14e+04	0.459	0.647	-7.79e+04	1.
total_stores 02e+04		1.819	e+04	3.65e+04	0.499	0.619	-5.38e+04	9.

Weighted Sum Application

We applied the weighted to the top 5 gentrification indicators we identified as follows:

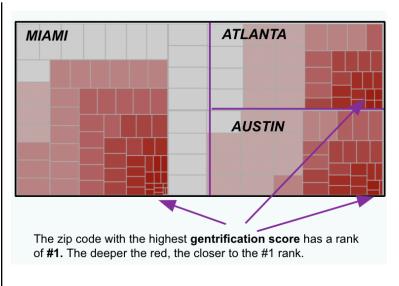
Variable	Weight	Rationale
Median Home Sale Price	50%	Applied 50% weight because it's the primary indicator we are using to observe change.

Percentage Change in Home Price	20%	P value is 0 indicating the observed difference is unlikely due to chance.
Information as a Percentage of the Total Labor Market	10%	Selected 10% weight due to a high p value.
Total Starbucks Stores	10%	Selected 10% weight because data was less complete and p value is high.
Annual Median Salary	10%	Selected 10% weight due to a high p value.

Gentrification Score Output

We created a new column to derive the gentrification score based on the weight sum model percentages above. Gentrification scores vary from 1 (lowest gentrification) to 146 (highest gentrification) across zip codes in Miami, Atlanta, and Austin.

	zip	gentrification_score
38	30350	146
106	78610	145
11	30291	144
24	30317	143
135	78742	142
68	33140	5
66	33138	4
67	33139	3
75	33149	2
55	33109	1

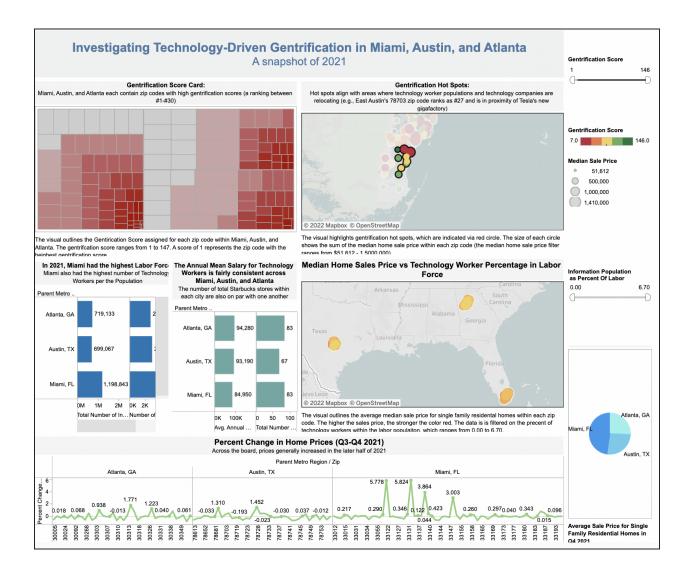


Miami's zip code 33109 (Fisher Island) yielded a score of #1

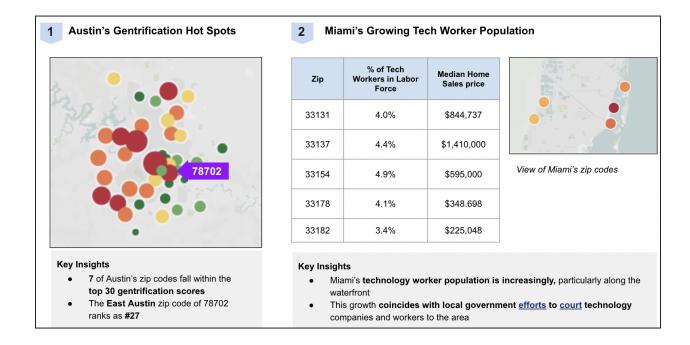
<u>Outputs</u>

The team created an <u>interactive dashboard</u> via Tableau to visualize key findings. The dashboard includes the following 6 components from left to right:

Component	Description	Interactive Feature
Gentrification Score Card	View of the gentrification score for each zip code within Miami, Austin, and Atlanta A rank of #1 represents the zip code with the highest score The darkest color red also indicates the zip codes with the highest scores	Filter by parent metro region
Gentrification Hot Spots	Map view of gentrification hot spots. Those marked with deeper red represent the highest gentrification scores while those marked with green have the lowest scores. The size of the circle is indicative of the medium sales price for single family residential homes within that zip code	Filter by gentrification score and zip code Zoom in and zoom out to visualize geographies
Total Labor Force and Technology Labor Force	Bar chart view of the current Total Labor Force in Miami, Austin, and Atlanta Highlights the number of workers who make up the Technology Labor Force	Filter by parent metro region
Annual Mean Salary for Technology Workers and Total Number of Starbucks Stores	Bar chart view of the average salary for the Technology Labor force in Miami, Austin, and Atlanta Includes a view of the number of total Starbucks stores within the city	Filter by parent metro region
Median Home Sales Price vs Technology Worker Percentage in Labor Force	Map view of the median sales prices for single family homes within Miami, Austin, and Atlanta	Filter by the percentage of Technology Workers in the Labor Force
Percent Change in Home Prices	Line chart of the percent change in home prices experienced by each zip code between Q3-Q4 2021	Filter by zip code

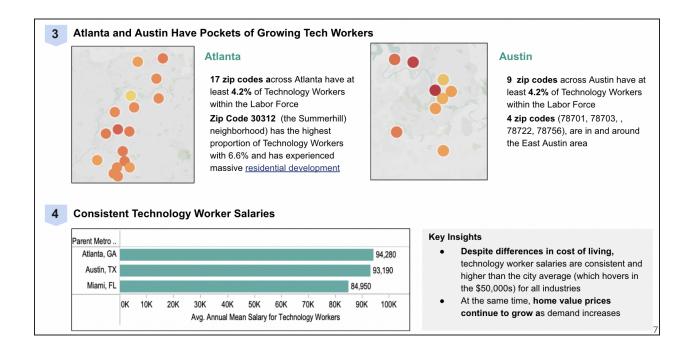


Key Insights



The ranking of the East Austin zip code 78702 reflects the surge in home values in addition to the growing presence of technology workers and the luxury commodities that begin to appear with a more affluent demographic. As of 2021 the mean sales price of a single family residential home in the 78702 zip code was \$695, 664 while at the beginning of 2019 homes in the same region sold for an average of \$472,090.

This shift in East Austin's cost of living, may be related to the emergence of new technology companies in the area. Tesla, which initially announced its intention to move to the outskirts of East Austin in 2019, began constructing its new gigafactory on the outskirts in 2020 opened their new facility in the nearby zip code of 78725. The new factory is also Tesla's new headquarters. The factory is reported to have created 20,000 direct and 100,000 indirect jobs across the technology and other industries



Conclusion & Recommendations

Our investigation yielded three key insights: 1) technology workers are embracing new hubs, 2) home prices are increasing in affordable areas, and 3) technology worker salaries are consistent and higher than non-technology workers.

Technology workers are embracing new hubs

Technology workers are gravitating towards cities outside of Silicon Valley and their movement may be attributed to both the emergence of technology companies in new areas, attractive local initiatives, and flexible remote work policies. The higher volume of technology workers may continue to attract growing numbers of tech workers as companies and people lay their roots,

Home prices are increasing in affordable areas

Across the board, prices are growing, especially in areas like East Austin, which are experiencing dramatic shifts in population and demographics Notably, areas with low median sales price have the highest gentrification score. Price hikes in areas with low home prices may continue to displace populations that are <u>reported</u> to be already vulnerable.

Technology worker salaries are consistent and higher than non-technology workers

Salaries are consistent across Miami, Austin, and Atlanta, making all three attractive options for technology workers. However, the average salary for tech workers is \$30,00 - \$40,000 higher than the average for all industries. While tech salaries are robust, lower salaries for non-tech workers may not be sufficient to support increasing housing and living costs

In order to combat the negative impact of tech-driven gentrification we recommend that local and state policy makers:

- 1) Assess the rate of displacement of existing communities within identified gentrification hot-spots to determine priority areas/groups to support
- 2) Invest in impacted communities (e.g., home buyer and business owner programs targeting marginalized communities), empowering and enabling people to remain in place
- 3) Incentivize technology companies (and other emerging industries) to engage with and support their communities by requiring them to invest resources in order to receive local tax breaks or other benefits