

Immigrant Residency and Happiness in New York City

Alison Tuiyott* Robert Garrett* Lydia Carter* Ben Schweitzer*
Thomas Fisher* Karsten Maurer*

Abstract

We explore the quality of life of immigrants in New York City through housing and neighborhood conditions by creating a happiness metric to measure a households' quality of life. Utilizing data provided by: the New York City Housing and Vacancy Survey (NYCHVS), the New York City Police Department, the New York City Department of Education, the New York City Department of Health and Mental Hygiene, along with reports from Happy City and the New Economics Foundation (HCNEF), a happiness score was assigned to each sub-borough in New York City. This happiness score evaluated five main domains: work, place, community, education, and health. As a result of this analysis, we discovered higher happiness scores were associated with lower percentages of immigrant households.

Key Words: Data Expo, Happiness Score, Immigrant Residency, New York City Housing and Vacancy Survey

1. Introduction

Many people around the globe venture to the United States seeking the American Dream. Some immigrants travel with hopes of a better quality of life. In this analysis, we use the data provided by the New York City Housing and Vacancy Survey (NYCHVS) to create a happiness metric to measure quality of life. According to the New York City Department of Housing Preservation and Development (HPD) [5]:

The NYCHVS is a representative survey of the New York City housing stock and population. It is the longest running housing survey in the country and is statutorily required. The Census Bureau has conducted the survey for the City since 1965. HPD is the only non-federal agency that sponsors a Census product. The HVS is a triennial survey with data collected about every three years. Each decade, a representative sample of housing units is selected, which represents the core sample.

Our goal is to use this data, in combination with other sources of New York City related to wellbeing, to explore the quality of life of immigrants in New York City through housing and neighborhood conditions. Our results help guide individuals to understand more about immigrant households and how being in an immigrant household relates to their quality of life.

To measure quality of life for New York City neighborhoods, we utilize a happiness metric. Happiness, according to HCNEF [1], is a city's success in providing the conditions that create 'sustainable wellbeing'. Sustainable wellbeing is made up of five main domains: work, place, community, education, and health. According to the Happy City Index 2016 Report, the happiness metric, "aims to be a practical tool that can help local policymakers understand how well their city is doing in comparison to the other cities and prioritize key policy areas" [3]. Using the data from

*Miami University Department of Statistics, 501 E High St, Oxford, OH 45056

the housing surveys about immigrant residency, we attempt to find a connection between happiness and immigrant residency.

2. Data

In New York City, there are five *boroughs*: Brooklyn, Bronx, Manhattan, Queens, and Staten Island. Each borough is further partitioned for local governance into roughly equivalent sized regions called *sub-boroughs*. In order to compare immigrant residency to happiness, we extract the immigrant residency information from the data provided by the NYCHVS. To create the happiness metric, we need a measure for the five main domains described by the HCNEF: work, place, community, education, and health. For work and place, we utilize the NYCHVS data. Community, education and health require data from external resources: the New York City Police Department, the New York City Department of Education, and the New York City Department of Health and Mental Hygiene, respectively.

2.1 Immigrant Residency: Place of Birth

In the NYCHVS data, there is a field that identifies the place of the householder's birth, and thus we are able to determine immigrant head of household. We apply the sample weights provided to the number of immigrant households and calculate the percentage of immigrant households within each sub-borough. **Figure 1** is a map of New York City illustrating immigrant residency by sub-borough.

2.2 Work: Income

In the NYCHVS data, there is a field that identifies total household income. We utilize this as a measure of *work*. We use the total household income because higher household income can be associated with well paying jobs, which can be a sign of a thriving household. Using this field, we adjust for inflation and calculate the average total household income per sub-borough. **Figure 2** is a map of New York City illustrating the average total household income by sub-borough.

2.3 Place: Rent

In the NYCHVS data, there is a field that identifies monthly contract rent. Monthly contract rent is the total rent scheduled to be paid to the landlord [4]. We utilize this as a measure of *place*. Using this field, we adjust for inflation and calculate the average monthly contract rent per sub-borough. **Figure 3** is a map of New York City illustrating the average monthly contract rent by sub-borough.

2.4 Community: Crime

For a measure of the *community*, we chose crime data provided by the New York City Police Department. Our assumption is a community with high crime might negatively affect the happiness of the households in that area. This data is provided at the precinct level. A precinct is an area of the city as defined for police purposes. Every precinct region in New York City is unique and the boundaries do not match the sub-borough regions. The crime data provided lists the total number of seven major felony offenses per precinct per year from 2000 to 2018. Since the NYCHVS data is collected every three years, we only selected years that overlapped with the

NYCHVS data. **Figure 4** is a map of New York City illustrating the total number of seven major felony offenses by precinct.

2.5 Education: High School Achievement Rates

For a measure of the *education*, we chose data on high school achievement in graduates provided by the New York City Department of Education. Our assumption is a community with many achieving high school graduates might positively affect the happiness of the households in that area. This data is provided at the school district level. Every school district region in New York City is defined by boundaries that do not match either the sub-borough or precinct regions. The education data provided lists each school within a school district, their cohort, and the number of achieved graduates in each cohort. A cohort is all students who first entered ninth grade in a given school year (the cohort of 2006 entered ninth grade in the 2006-2007 school year). Graduates are defined as those students earning either a Local or Regents diploma and exclude those earning either a special education (IEP) diploma or GED. Achieved graduates are the number of graduates achieving Aspirational Performance Measure (APM). The New York State Education Department (NYSED) defined English/Math Aspirational Performance Measure (APM) as the percentage of students that, after their fourth year in high school, have met NYSED standards:

- Graduated by August with a Regents or Local diploma, AND
- Earned a 75 or higher on the English Regents, AND
- Earned an 80 or higher on one Math Regents

We utilize the number of high school graduates who achieved APM. This metric measures the quality of the education.

The education data is only available for cohorts that graduated from 2010 to 2015. We extrapolate the data for 2017 using the 2015 data. We bin the remaining data into 2011 and 2014. This means 2010-2011 data was grouped together to 2011 data and 2012-2014 data was grouped together to 2014 data. **Figure 5** is a map of New York City illustrating the rate of the number of achieved high school graduates and the number of total graduates in the cohort by school district.

2.6 Health: Mortality Age Range

For a measure of the *health*, we chose mortality data provided by the New York City Department of Health and Mental Hygiene. Our assumption is a community with a lower death age will negatively affect the happiness of the households in that area. This data is provided at the community district level. Every community district region is defined by boundaries that do not match the sub-borough, precinct or school district regions. Also, community districts may not align with the precincts or school districts either. The mortality data provided lists the total number of deaths and the number of deaths in each 5 year age range per community district per year.

To calculate the median death age range, we first calculate the median number of deaths. Working backwards from the oldest age range, we find the age range containing this median number of deaths. For example, if there were 15 deaths in the 85+ range, 20 deaths in the 80-84 age range, 5 deaths in the 75-79 age range,

and no deaths in the remaining age ranges, there would be 40 total deaths. We start from the oldest age range (85+) and cumulatively sum up the total number of deaths. If we start from the oldest age range and cumulatively sum the total number of deaths, we would get: 15 cumulative deaths in 85+ range, 35 cumulative deaths in the 80-84 range, and 40 cumulative deaths in all of the subsequent ranges. From here, we select the oldest age range that contains our median number of deaths. In our case, we are looking for the age range that has more than or equal to 20 cumulative deaths because that is the age range where the median lies. This means the 85+ range is ruled out because it only has 15 cumulative deaths. The 80-84 range has 35 cumulative deaths and it is the oldest age range with the median, therefore, we stop. We repeat this process for each community district for each year.

The data ranges from 1999 to 2014. Since the NYCHVS data is collected every three years, we only selected years that overlapped with the NYCHVS data. **Figure 6** is a map of New York City illustrating the median death age range by community district.

3. Methods

3.1 Aggregating the Data

The shape file data is provided by the New York City Department of City Planning. This website provides the shape file data for the Census blocks or sub-borough regions which is what is used in the NYCHVS data, school district regions for education and community district regions for health. The precinct shape files for the crime data are provided by New York City OpenData.

Starting with the NYCHVS data, we use the `pumas` function in the `tigris` package in R to extract New York specific data [2]. We use the geoid's provided by the NYCHVS data to extract the necessary geoid's from the Public Use Microdata Area (PUMA) data. The data is read in and projected to the Long Island State Plane coordinate reference system. From there we can create an outline of New York City and the unique sub-boroughs. For the purposes of explaining this process in greater detail, we demonstrate with Brooklyn. **Figure 7a** shows Brooklyn with its sub-boroughs outlined in red. We refer to these unique sub-borough regions as R_i , where i represents the region for the i^{th} sub-borough.

In order to aggregate all of the different regions (sub-boroughs, police precincts, school districts, and community districts), we first remove all self intersections. To do this, we use the `gBuffer` function from the `rgeos` package in R to make the widths equal to zero [7]. We use the `gIntersection` function from the same package to first intersect the sub-borough regions and the precinct regions. **Figure 7b** shows Brooklyn with its sub-boroughs outlined in red and precinct regions outlined in blue. We refer to these unique sub-borough and precinct regions as R_{ij} , where j represents the region for the j^{th} precinct.

Building off this combination, we follow the same process to intersect the sub-borough and precinct regions and the school district regions. **Figure 7c** shows Brooklyn with its sub-boroughs outlined in red, precinct regions outlined in blue, and school district regions outlined in orange. We refer to these unique sub-borough, precinct and school district regions as R_{ijk} , where k represents the region for the k^{th} school district.

Continuing the previous combination, we follow the same process to intersect the sub-borough, precinct, and school district regions and the community district

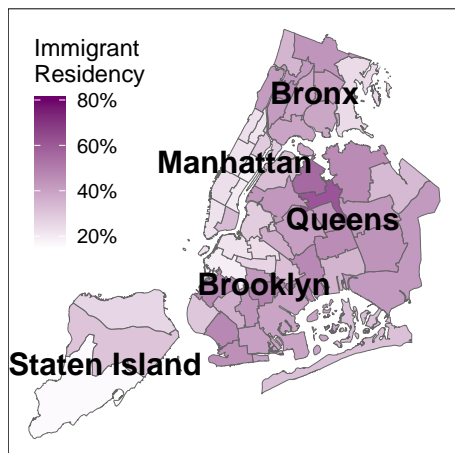


Figure 1: The percentage of immigrant households by sub-borough region.

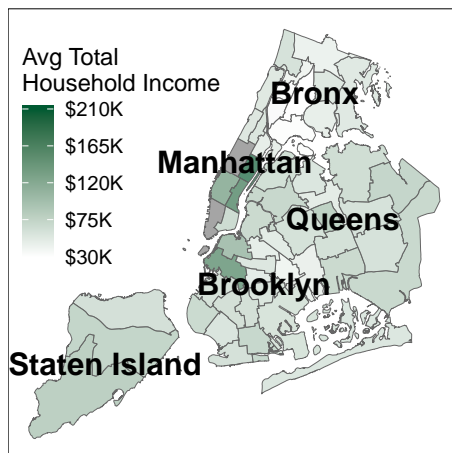


Figure 2: The average total household income by sub-borough region.

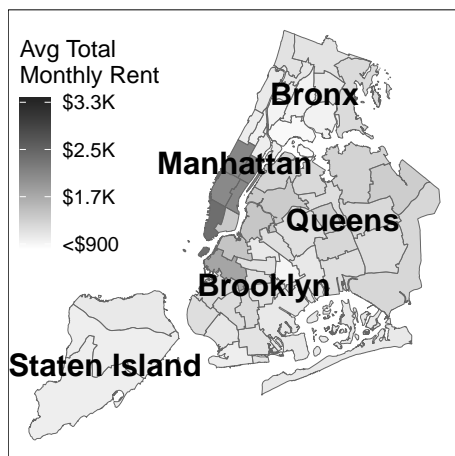


Figure 3: The average total monthly contract rent by sub-borough region.

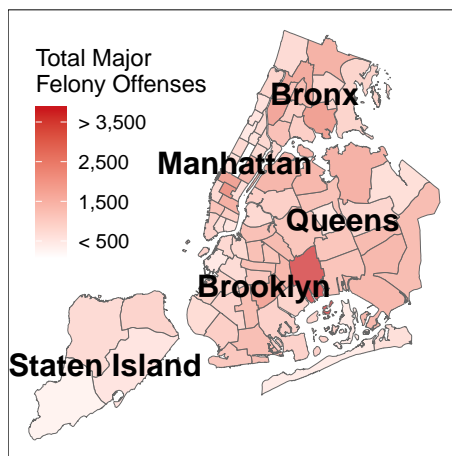


Figure 4: The total number of major felony offenses by precinct region.

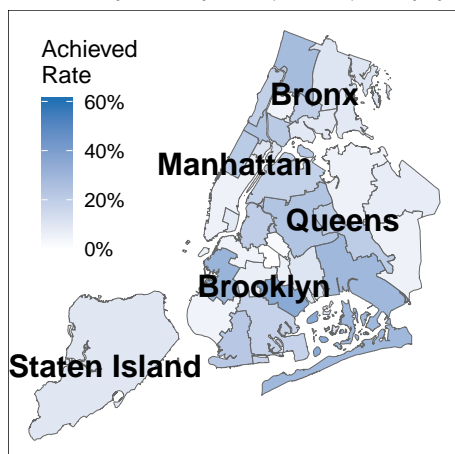


Figure 5: The rate of achievement of high school graduates by school district region.

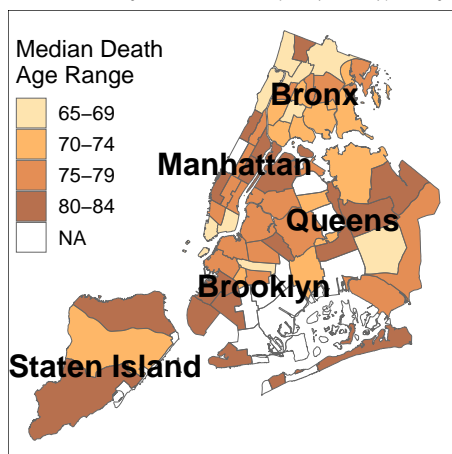


Figure 6: The median death age range by community district region.

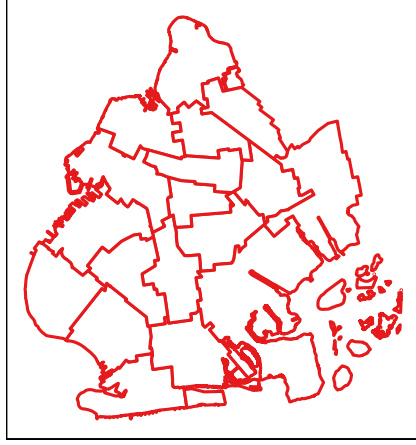


Figure 7a: Region R_i of sub-borough regions in red.

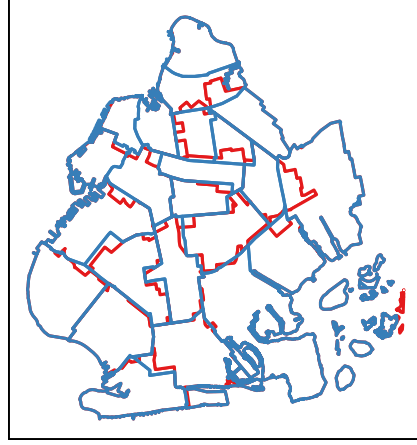


Figure 7b: Region R_i of sub-borough and precinct regions in red and blue, respectively.

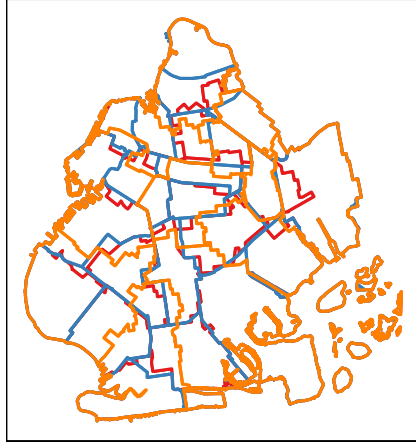


Figure 7c: Region R_{ij} of sub-borough, precinct and school, district regions in red, blue and orange, respectively.

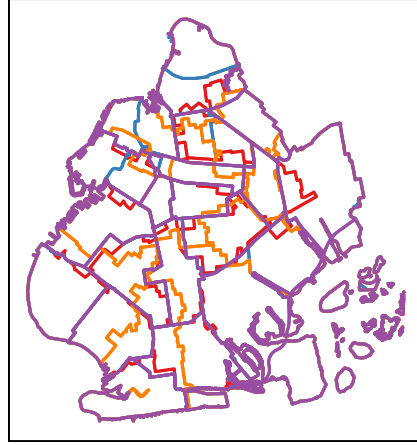


Figure 7d: Region R_{ijkl} of sub-borough, precinct, school district, and, community district regions in red, blue, orange, and purple, respectively.

regions. **Figure 7d** shows Brooklyn with its sub-boroughs outlined in red, precinct regions outlined in blue, school district regions outlined in orange, and community districts outlined in purple. We refer to these unique sub-borough, precinct, school district and community district regions as R_{ijkl} , for the intersection of the i^{th} sub-borough, j^{th} precinct, k^{th} school district, and the l^{th} community district.

After the intersections are complete, we project the data back to the geographic coordinate system of longitudes and latitudes. We return to the original four regions and extract the ID's for each dataset type by using the `sapply` function from base R. Every ID is unique for each polygon in the different regions (sub-boroughs, police precincts, etc.). We concatenate all four ID's to create a single ID (R_{ijkl}) for each unique polygon in the intersected data. We add this ID to the spatial polygon data frame. Using these ID's we can utilize the `SpatialPolygons` function from the `sp` package in R to left join all of the necessary data from the different regions to their respective polygons [6,8].

Figure 8 zooms in on a specific sub-borough in Brooklyn where we can notice all the different geographic regions that are necessary to calculate the happiness score index. Each region has a unique combination of sub-borough, precinct, school district and community district regions, R_{ijkl} .

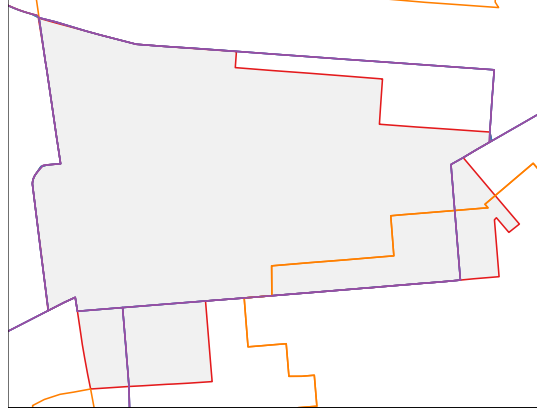


Figure 8: Region R_{1234} of sub-borough region 1 in red, precinct region 2 in blue, school district region 3 in orange, and community district region 4 in purple.

3.2 Calculating Happiness Score Index

When calculating the happiness score index, we first take all of the intersected regions and combine them with the data from their respective regions. For example, the crime measure will be combined with its respective precinct region. We extract the area in squared meters of each unique intersecting region. We call this area, a_{ijkl} , i^{th} sub-borough, j^{th} precinct, k^{th} school district, and the l^{th} community district.

Using these areas as weights, we calculate the happiness score for each sub-borough region, H_i .

$$H_i = a_{i...}(income_i) + a_{i...}(place_i) + a_{ijkl}(crime_j) + a_{ijkl}(education_k) + a_{ijkl}(health_l)$$

where $income_i$ represents the average total household income for the i^{th} sub-borough region, $place_i$ represents the proportion of average monthly contract rent to average total household income for the i^{th} sub-borough region, $crime_j$ represents the total number of major felony offenses for the j^{th} precinct region, $education_k$ represents the achievement rate of high school graduates for the k^{th} school district region, and $health_l$ represents the lowest age in the median death age range for the l^{th} community district region. Note: The value of each weighted average is scaled from 0 (Worst) to 1 (Best) before taking the sum, and therefore the happiness scores have a maximum range between 0 and 5. This formula is how to calculate the happiness score index for each sub-borough region in New York City.

4. Results

Figure 9 shows the happiness score index for each sub-borough in New York City. **Figure 10** is another visualization to better understand the association between immigrant residency and happiness scores. From **Figure 10**, we see that Manhattan and Staten Island tend to have lower percentages of immigrant households and higher happiness score indexes. The Bronx tends to have the reverse effect with higher percentages of immigrant households and lower happiness score indexes. Then, there are also those boroughs, Queens and Brooklyn, that are relatively average across.

Happiness Score Index

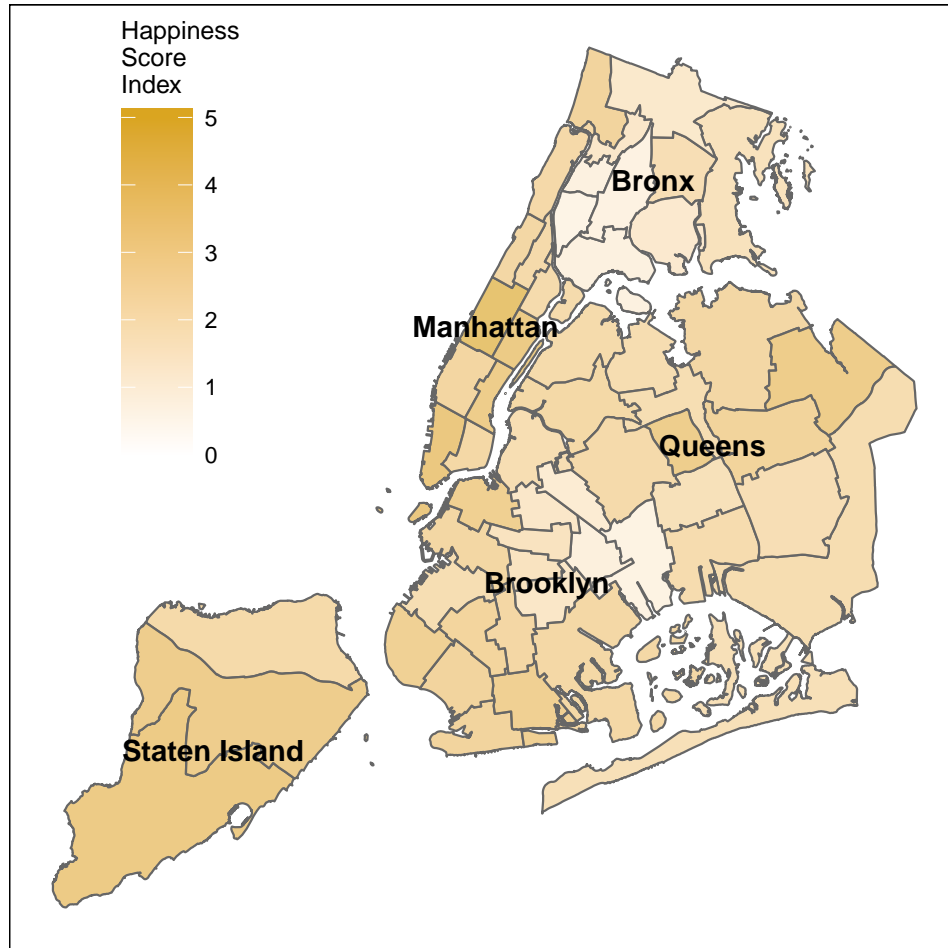


Figure 9: A map of New York City where darker yellow represents happier sub-borough regions.

5. Discussion

In this analysis, we utilized the NYCHVS data and other external resources to create a happiness metric to evaluate the quality of life for immigrant households in New York City. We took advantage of the happiness score index created by HCNEF to create a metric for each sub-borough in New York City. From the happiness score index map visualization, it is clear that most of Manhattan and Staten Island tend to be happier than some areas of Brooklyn/Queens and the Bronx. It would be interesting to see if over time, these happiness scores change within a given sub-borough. In order to achieve this task, we would need current education data which is not available at the moment. When analyzing the scatter plot visualization with all of the sub-boroughs, it would be interesting to delve deeper into the original countries of some of these immigrant households. Also, if length of residency in the United States is associated with higher happiness scores. Overall, we know that happiness, in our calculations, is heavily skewed in favor of wealth. However, we know that being wealthier may not equate to being happier, but more wealth can alleviate other real life problems.

Immigrant Residency and Happiness Score Index



Figure 10: A scatter plot where the color represents each sub-borough and the size represents number of households.

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