### Excel Step-by-Step Data Manipulation

## **Opportunity Atlas**

## Filtering and Manipulating Data

- 1. Downloaded Baltimore Median Household Income data from the Opportunity Atlas for All Income and All Race subgroups
- 2. Used Text-to-Columns tool to focus only on neighborhoods within Baltimore City

### **Creating Graphs**

- 1. Created a Pivot Table in order to find the median household income for each neighborhood and inserted a bar graph (*Baltimore City Median Household Income by Neighborhood*) to visually interpret the results
- 2. Checked neighborhoods against the official Baltimore City neighborhood list to ensure that all neighborhoods included were a part of Baltimore City

### **Open Baltimore**

### Filtering and Manipulating Data

- 1. Downloaded the Vacant Buildings data from Open Baltimore, last updated on October 26, 2020
- 2. Added a new column, Number of Vacant Buildings, to each row as data was presented by address and neighborhood

# Creating Graphs

- 1. Created a Pivot Table in order to sum the total number of vacant buildings per each neighborhood
- 2. Inserted a bar graph to represent this visually (*Total Number of Vacant Buildings in Baltimore City by Neighborhood*)
- 3. Used Pivot Table again to look at the vacant buildings total for each police district
- 4. Inserted a bar graph to represent this visually (*Total Number of Vacant Buildings in Baltimore City by Neighborhood*)

### U.S. Census

## Filtering and Manipulating Data

1. Downloaded the U.S. Census data for Baltimore City and focused on pct\_black (percent Black residents per neighborhood), pct\_white (percent White residents per neighborhood), racial\_diversity\_index (the diversity score for each neighborhood from 0-100), pct\_households\_belowpovertyline (percent of households below the poverty line)

### **Combining Data Sets**

- 1. Used =VLOOKUP to match the median household income data and vacant buildings data to each neighborhood row in the U.S. Census datasheet
- 2. The U.S. Census data combined several neighborhoods together along with having individual neighborhoods for their rows, unlike the Opportunity Atlas and Open Baltimore data sets which had individual neighborhoods

- 3. Created a combined data set that encompassed vacant buildings data from Open Baltimore, median household income data from Opportunity Atlas, and demographic data from the U.S. census by:
  - 1. Averaged the median household income data for the separate neighborhoods that were grouped together by the U.S. Census using =AVERAGE
  - 2. Found the total sum of vacant buildings in the neighborhoods grouped together by the U.S. Census using = SUM

# **Creating Visuals: Combined Data Sets**

### Bar Graphs

- 1. Used the combined dataset to focus on the race breakdown for each neighborhood
- 2. Inserted a bar graph showing Total Vacant Buildings by Percent of White Residents per Baltimore City Neighborhood
- 3. Inserted a bar graph showing *Total Vacant Buildings by Percent of Black Residents per Baltimore City Neighborhood*

### Scatter Graph

- 1. Created a simple linear regression for median household income using =SLOPE, =INTERCEPT, =RSQ, and =STDEV for the table
- 2. Calculated Predicted Vacant Buildings using the y = mx + b formula and the above values
- 3. Calculated the error using a subtraction formula and then created an Outliers column finding outliers greater than 2\*STDEV to find percentage of total outliers for the neighborhoods
- 4. Inserted a scatter graph with linear trendline, *Vacant Buildings vs. Median Household Income* in Baltimore City Neighborhoods

### **Clustering: Combined Data Sets**

### Cluster Analysis

- 1. Numbered the neighborhood groups from 1 to 33
- 2. Calculated the Mean and Standard Deviation using AVERAGE and STDEV functions for the five variables (pct\_black, pct\_white, pct\_householdsbelowpovertyline, median\_income, vacant\_buildings)
- 3. Calculated the z-eligible columns using the STANDARDIZE function for each of the three variables
- 4. Highlighted the whole data array to name it "Cluster" and numbered the z-scores columns in the table above
- 5. Used the VLOOKUP function to set up our anchors in finding the names for our random cluster numbers in the table above as well as respective z-values for each of the variables
- 6. Used the SUMXMY2 function to calculate the distance between the variable z-scores for each agency and the z-values of the anchors
- 7. Used the MIN function to find the smallest distance of the four distances in each agency row
- 8. Used the MATCH function to assign the smallest distance to an anchor
- 9. Calculated the sum of all the minimum distances by using the SUM function
- 10. Applied the Excel Solver operations for a cluster analysis on the dataset
  - a. Set objective to our sum of the minimum distances cell
  - b. Selected min

- c. By changing variable cells to the cluster numbers in the table
- d. Subject to the constraints with the cluster cells > 1, < 56, and = 1 integer
- e. Selected Evolutionary as the solving method

# Table with Grouped Neighborhoods

- 1. Filtered the cluster analysis results to see which neighborhoods were grouped into cluster 1, 2, 3, or 4
- 2. Copied over the neighborhoods to create a table showing these cluster groupings