Collaborators: None

Project Overview

 Goal: How can we find natural groups among subsidized housing based on their key features?

• **Dataset**: Subsidized Housing - Six Metro Areas - 2017, 347 KB (1,945 lines)

Data Processing

- **Loading**: The function *load_cleaned_data()* opens the CSV via *csv::Reader* and *serde::Deserialize*
 - Skips invalid rows (logging errors) and returns a Vec<HousingProperty> with fields total_units: u32, subsidy_count: u32, and owner_type: String
- **Cleaning/transformations**: Removed rows with missing entries (and referenced the CSV in SPSS to double-check), selected only the variables Latitude, Longitude, TotalUnits, ActiveSubs, and OwnerType (but finalized with the last three), and wrote out to "cleaned_subsidized_housing.csv" from "data_cleaning.py"

Code Structure

Modules

- "data.rs" loads and deserializes the cleaned CSV into structs
- "clustering.rs" converts structs into a scaled 2D dataset, runs k-means (k =
 4), denormalizes centroids, and prints summary statistics
- "plot.rs" creates a 1600x1200 PNG scatter plot of TotalUnits and ActiveSubs, colored by cluster, with a legend and 'x' markers on each centroid
- o "utils.rs" provides a function used to match cluster IDs to colors consistently
- o "main.rs" runs all of the above where the project loads, clusters, and plots

• Key functions & types

- o *struct HousingProperty* ("data.rs") → represents one housing record with fields *total units, subsidy count, owner type*
- load_cleaned_data(path) ("") → inputs file path to cleaned CSV and outputs Vec<HousingProperty> or error; it opens a file, iterates, and pushes valid entries
- o to_ndarray_with_scales(properties) ("clustering.rs") → inputs slice of HousingProperty and outputs (Array2<f64>, min[0, 1], range[0, 1]); it fills raw values, computes the minimum and maximum per column, and applies a normalization formula
- o $cluster_properties(properties, k)$ ("") \rightarrow inputs data slice and cluster count and outputs Vec < usize > of labels; it calls the scaler, fits k-means, predicts labels,

- and computes denormalized centroids, sizes, and owner-type distributions before printing
- plot_clusters(properties, labels) ("plot.rs") → inputs slice of properties and labels and outputs () or error; it automatically increments a new filename, builds BitMapBackend, draws points by cluster with legend entries, computes centroids and overlays markers, renders a legend, and saves
- o $get_cluster_color(cluster_id)$ ("utils.rs") \rightarrow inputs cluster index and outputs RBGColor for plotting

• Main workflow

- \circ main.rs calls load cleaned data \rightarrow cluster properties \rightarrow plot clusters
- Each stage uses only the public functions above, keeps data in memory, and communicates via return values and printed logs

Tests

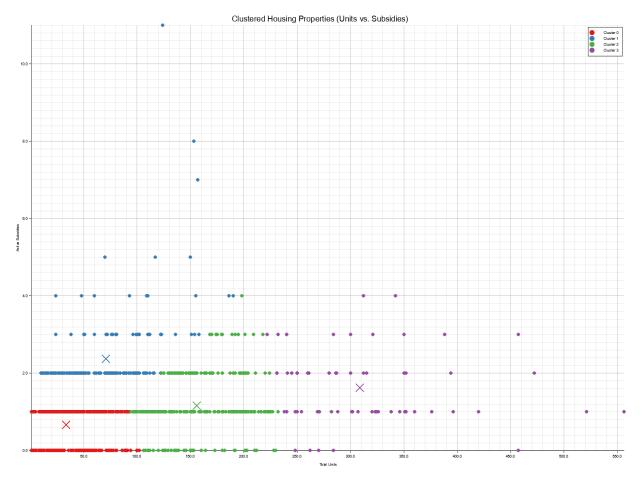
- For "utils.rs": test_get_cluster_color_diff() checks whether different cluster IDs map to different colors on the RGB scale
- For "clustering.rs": test_two_clear_clusters() checks if cluster_properties groups clearly distinct sample points into exactly two clusters
- For "data.rs": test_load_cleaned_data_simple() checks that CSV parsing works on a basic example as a behavior scan
- For "plot.rs": test_plot_clusters_runs() checks for plot_clusters execution doesn't have errors and writes a PNG output file

Results

```
Running 'target\release\section=8-rust.exe'
Loaded 1145 cleaned housing entries.

Cluster centroids:

Cluster 8:
Total Units: 34
Active Subsidies: 1
Cluster 2:
Total Units: 571
Active Subsidies: 2
Cluster 2:
Total Units: 156
Active Subsidies: 1
Cluster 3:
Total Units: 389
Active Subsidies: 2
Cluster 3:
Cluster 3:
Cluster 8:
Cluster 8: 583 properties
Cluster 8: 583 properties
Cluster 1: 204 properties
Cluster 1: 294 properties
Cluster 2: 294 properties
Cluster 3: 68 properties
Cluster 3: 69 properties
Cluster 3: 69 properties
Cluster 3: 69 properties
Cluster 3: 60 properties 30
Multiple: 19
Non-Profit: 103
Non-Profit: 70
Cluster 2:
For Profit: 52
Multiple: 4
Non-Profit: 52
Multiple: 4
Non-Profit: 12
Saved plot to output/clusters_11.png
PS C:\Users\jiay\lambda(section=8-rust>)
```



Interpretation

- Cluster 0 grouped a large number of <u>small</u>, <u>least subsidized buildings</u> where for-profits operate a majority and nonprofits fill in the rest; this suggests a widespread baseline assistance across both sectors
- Cluster 1 grouped <u>mid-sized developments with slightly more aid</u> that are controlled mostly by for-profit companies; may suggest that gentrification is increasing
- Cluster 2 grouped <u>larger buildings receiving minimal aid</u> where for-profits hold a three-quarters supermajority ownership; this suggests more housing scarcity and inequity over the distribution of resources
- Cluster 3 grouped a small amount of the <u>largest buildings with higher total</u> <u>subsidies</u>, also controlled primarily by for-profits; may suggest housing displacement if the owners choose profit over people

Usage Instructions

1. **Clone**

- a. git clone https://github.com/ username /section-8-rust.git
- b. cd section-8-rust

2. Build/Test

- a. cargo test
- b. cargo build

3. **Run**

a. cargo run --release

4. Runtime

a. Data loading + clustering + plotting should finish in about 3-5 seconds

AI-Assistance Disclosure and Other Citations

• I used ChatGPT to help me navigate file I/O when writing in Python (using Pandas) to clean my dataset. I am used to using absolute paths, but since the context of this assignment was to build a publicly accessible project, that was not viable. For this, I asked the following: "How do I use a relative path? In the grand scheme of things, I need my entire project to be accessible through GitHub (I want to do a push-pull request to update the work I've done today) so I don't think using absolute paths is

convenient." It then produced this:

Keep a folder structure like this in your GitHub repo:

When you cd your-project in your shell (or open that folder in Sublime), all paths below are relative to that

- Among other things, ChatGPT helped me debug cases where my "NaN" filters were having errors and helped me track input and output locations (very useful since I started from C:\Users\jiayi and was not doing it in my dedicated DS 210 folder)
- Found this blog post to be particularly useful when thinking of how to visualize k-means clustering with my specific dataset, since my proposal was rejected for lack of graphability (and this was done in R, so no, I did not use any of their code)