

Visualization of Water Usage and Affordability Clusters

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Summary

- **Motivation:** Water access, affordability, and quality have re-emerged as serious issues in the US and continue to be major problems in developing countries
- **Objective:** Visualize the extent of the this issue in the US using water price data, demographic information using cluster analyses
- **Analyses:**
 - **Univariate clustering** on water usage, price, and income to identify municipalities and and regions of where water may be less affordable
 - **Multivariate clustering** to see how other factors, including demographic variables, may correlate with income and water usage and price

Methodology: Data pipeline

Sources

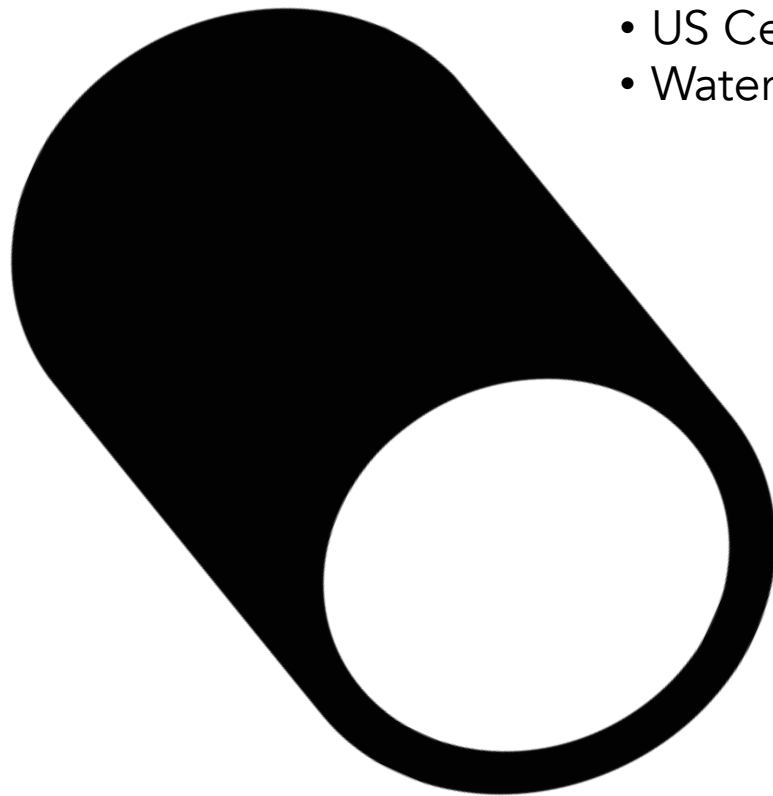
- US Census American Consumer Survey (2014)
- Water utilities price data*

Wrangling

- Clean, normalize, and join geospatial and non-geospatial data

Analysis and visualization

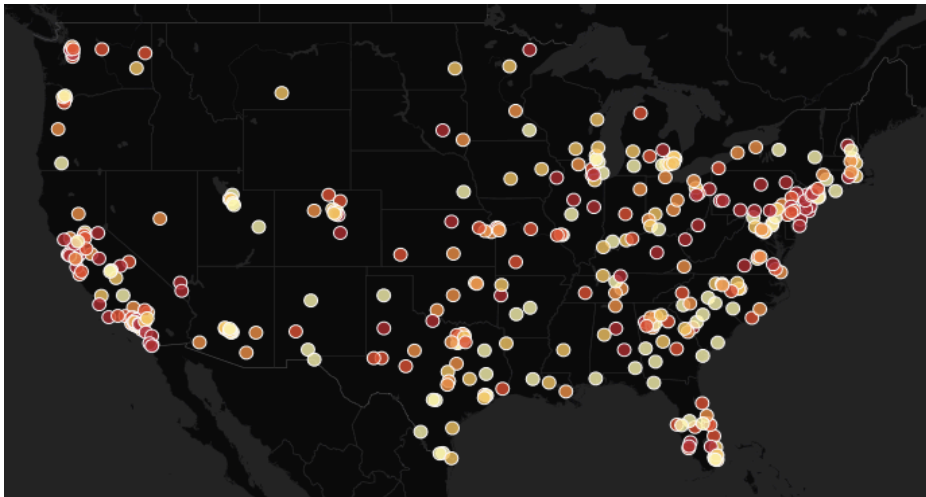
- Univariate modeling and scoring
- Multivariate k means feature & k selection
- Worked with training data only
- Visualize univariate outputs on maps, multivariate outputs on scatterplot matrices



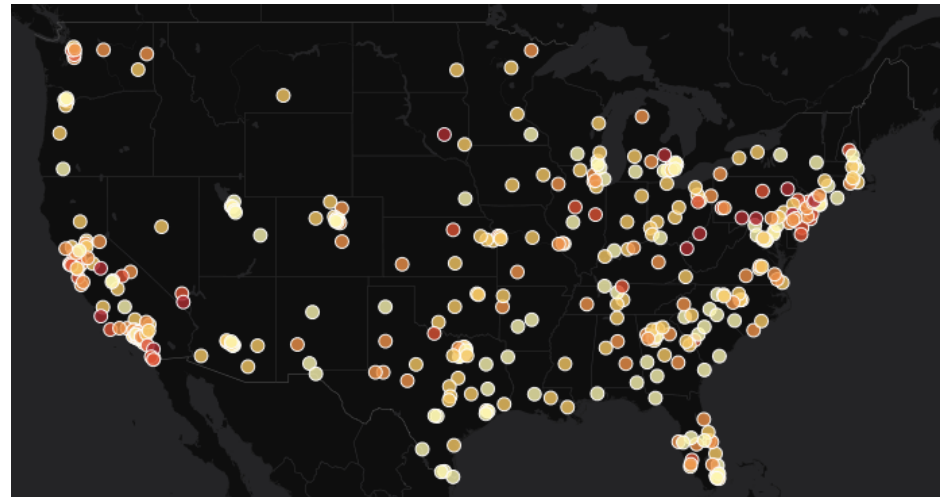
*Pipeline icon created by misirlou from Noun Project

Univariate clustering: Quintiles v. Jenks

Quintiles



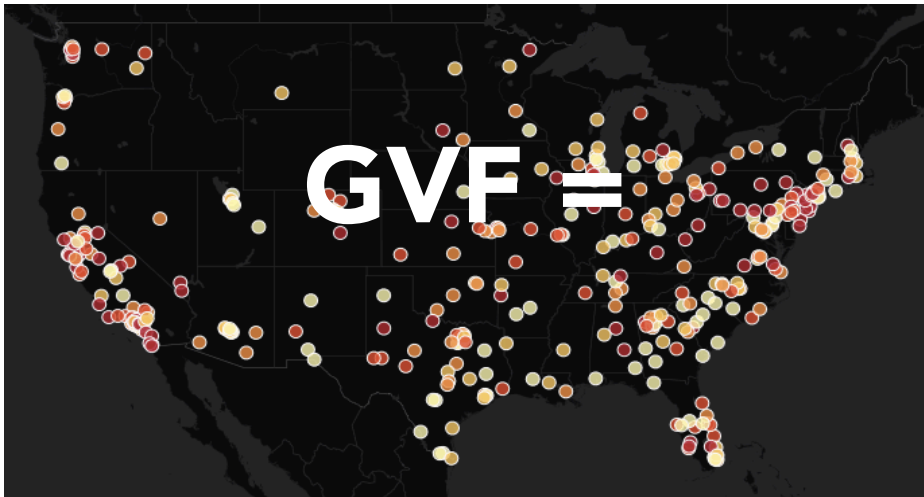
Jenks Breaks



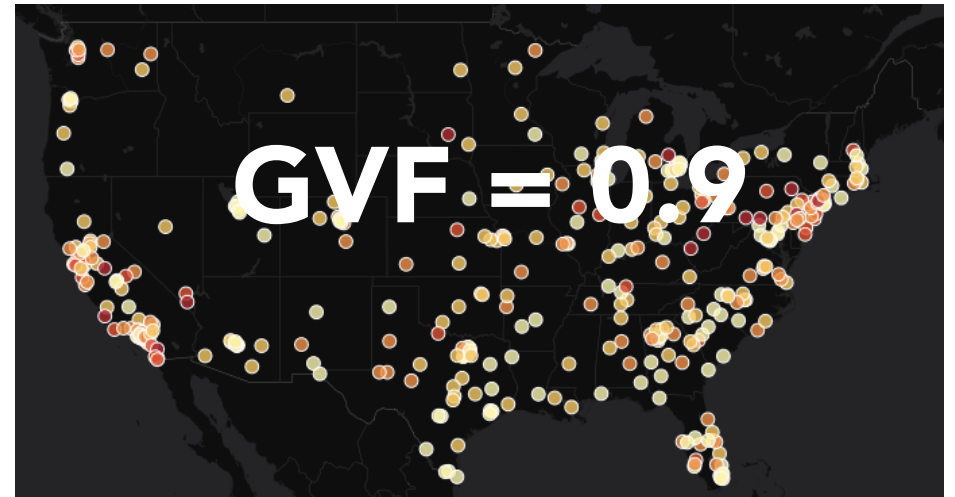
Both look similar above, but goodness of variance fits (GVF) are different

Univariate clustering: Quintiles v. Jenks

Quintiles

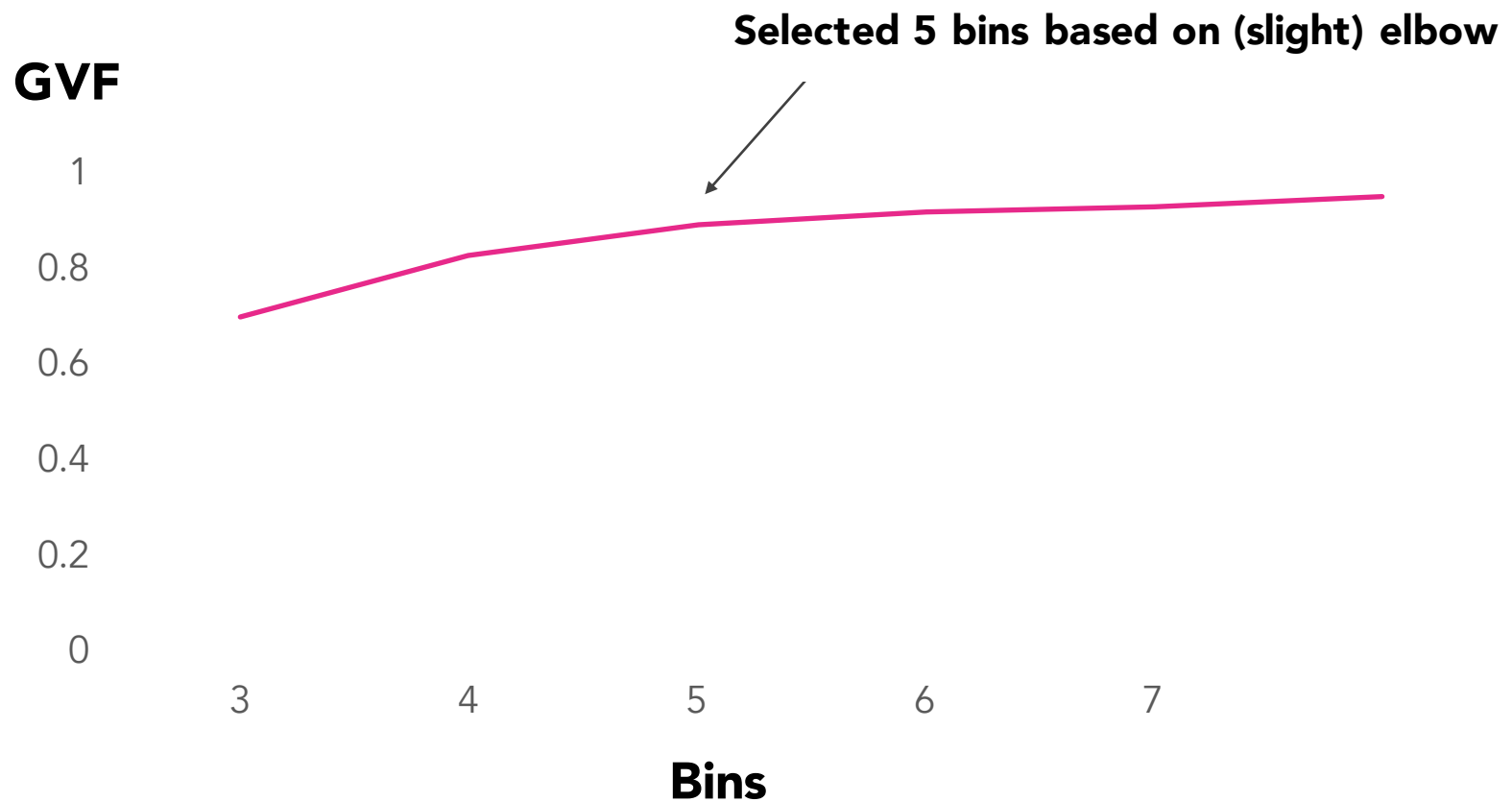


Jenks Breaks

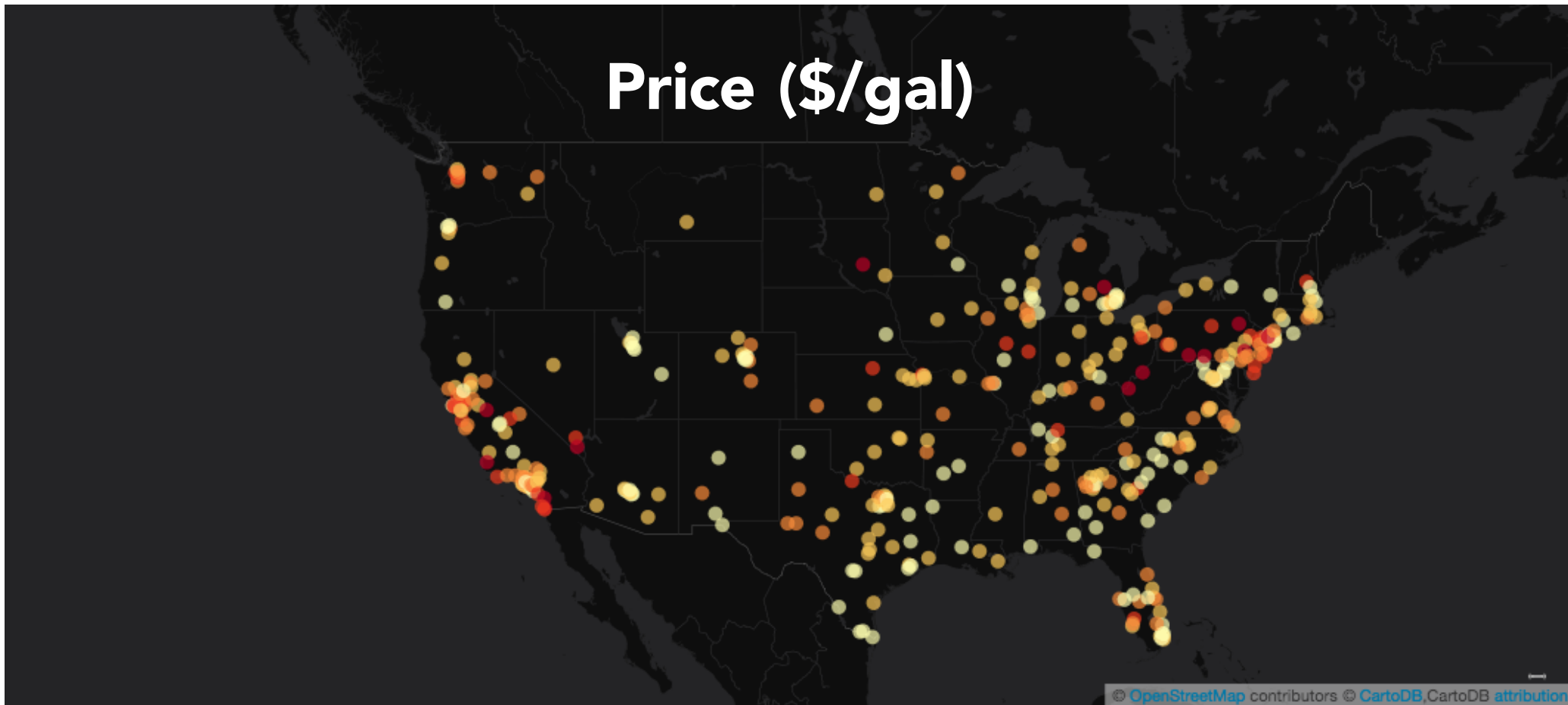


Based on GVF, Jenks wins

Univariate clustering: Goodness of variance fit (GVF)

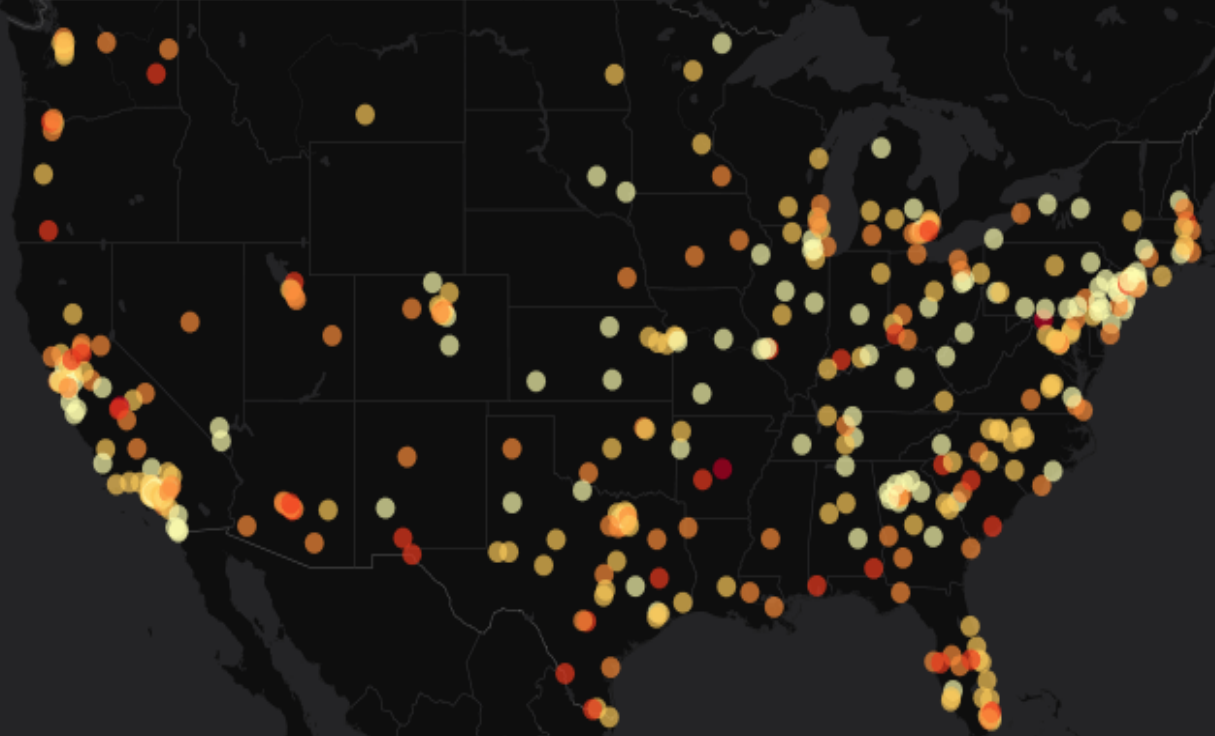


Univariate clustering: Results



Univariate clustering: Results

Usage (gal/person/year)

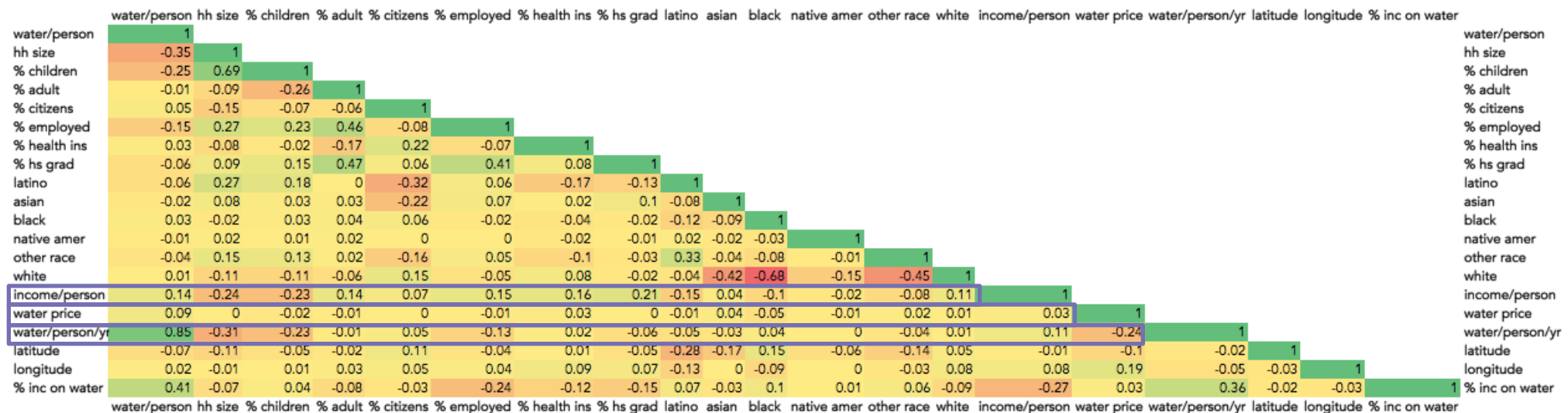


Multivariate clustering: k means

“k means clustering aims to partition n observations into k clusters in which each observation belongs to cluster with nearest mean”
(Wikipedia)

Multivariate clustering: Insights from the correlation matrix

Feature set in a correlation matrix

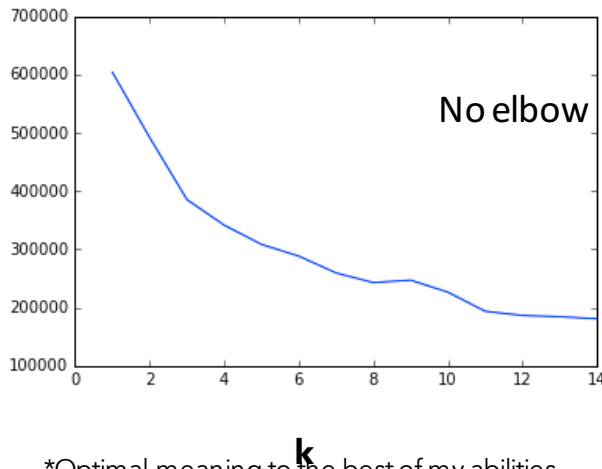


Multivariate clustering: iterate with different feature sets across various k's

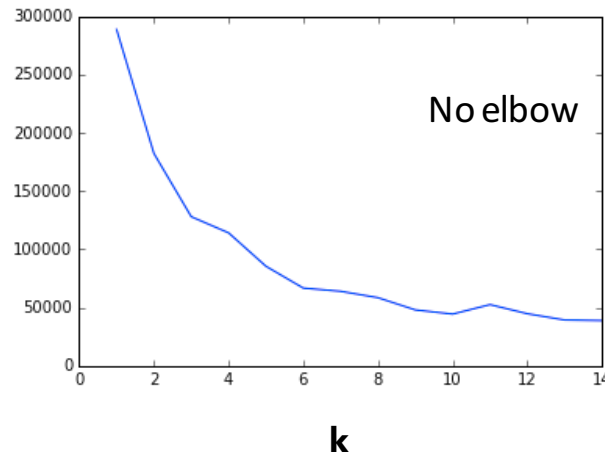
Get optimal* feature set and k clusters

- Remove features that are irrelevant to clusters to avoid curse of dimensionality
- Find the elbow across various feature sets and k's to get optimal feature set and k

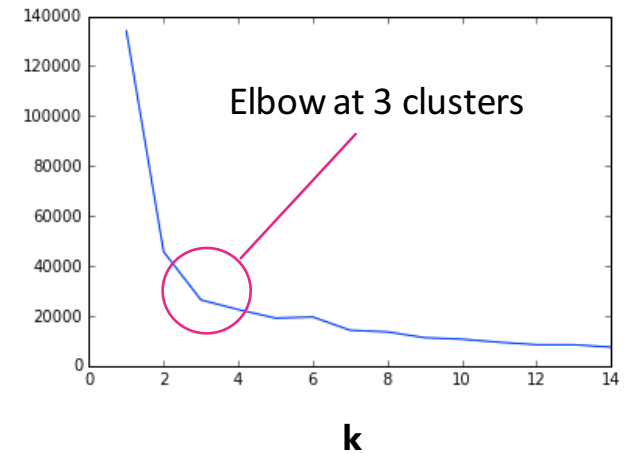
Features = 18



Features = 9

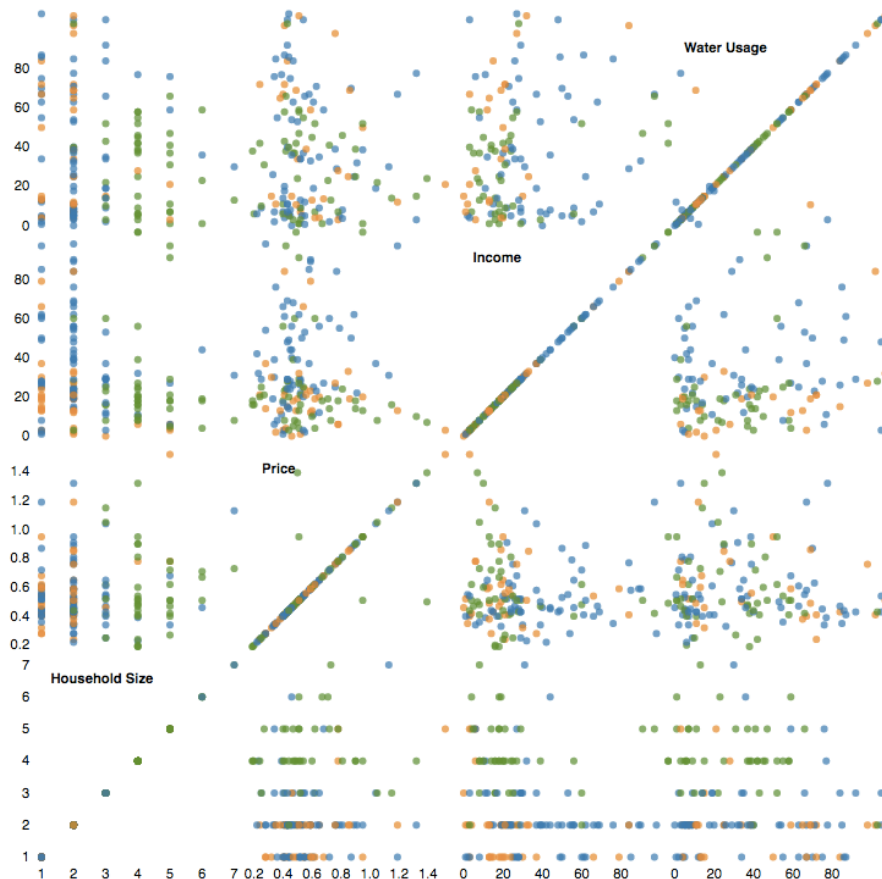


Features = 5



*Optimal meaning to the best of my abilities

Multivariate clustering: visualize outputs



- Visualize facets of higher dimensional space through scatterplot
- Here, the 3 colors each represent a class
- However, classes do not seem coherent based on visual inspection
- Therefore, preliminary multivariate cluster results inconclusive

Conclusions

- **Overall:** Univariate results were more intelligible and conclusive than multivariate ones
- **Univariate clustering:**
 - Jenks often preferable to quintiles → better analyses and visualizations
 - Method good for getting quick, high-level insights
- **Multivariate clustering:**
 - Initial results inconclusive despite good-looking elbow plots
 - May not be the appropriate method for this problem
- **Other approaches v multi-variate clustering:** Treating this as a supervised problem may be more appropriate for this analysis

Appendix

Univariate analysis: Results

Income (\$/person/year)

