

Seattle Food Inspections

Technology Review

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Background

Starting 2012, jurisdictions across the country including King County have begun publishing health inspection scores using a standardized scoring system called LIVES. This open data allowed restaurant consumers to make informed decisions based on where they want to eat and motivated a lot of restaurant establishments to improve their inspection score in the hopes of attracting a bigger customer base.



Brief overview

1. **Trend Analysis:** What is the overall trend for the aggregate number of different risk violations from 2016 to 2019?
2. **Location Specific Contributions:** Where are the specific regions in Seattle that have the most high risk violations? How much do those regions contribute in terms of total high risk violations in Seattle?
3. **Interactive Map:** showing health inspection ratings in Seattle.
4. Then overlay these ratings with demographic information in the areas around restaurants.
5. Users will be able to explore **correlations** between the demographics and health inspection ratings to help make decisions about where to eat.

Dataset

1. Food Establishment Inspection Data - King County Open Data:

<https://data.kingcounty.gov/Health-Wellness/Food-Establishment-Inspection-Data/f29f-zza5>

2. American Community Survey, 5-Year Estimates:

<https://data.census.gov/cedsci/table?q=S1903&table=S1903&tid=ACSST1Y2018.S1903&lastDisplayedRow=0>

<https://data.census.gov/cedsci/table?q=S1201&table=S1201&tid=ACSST1Y2018.S1201&lastDisplayedRow=0>

Seattle Restaurants Inspection

Select Year:

2014

Select Month:

Jan

Select Day:

01

Violation Type:

No Risk

☐ Display High Risk Violations Heatmap by Postcode

Plot Restaurants of HR Violations for Top 7 Postcode:

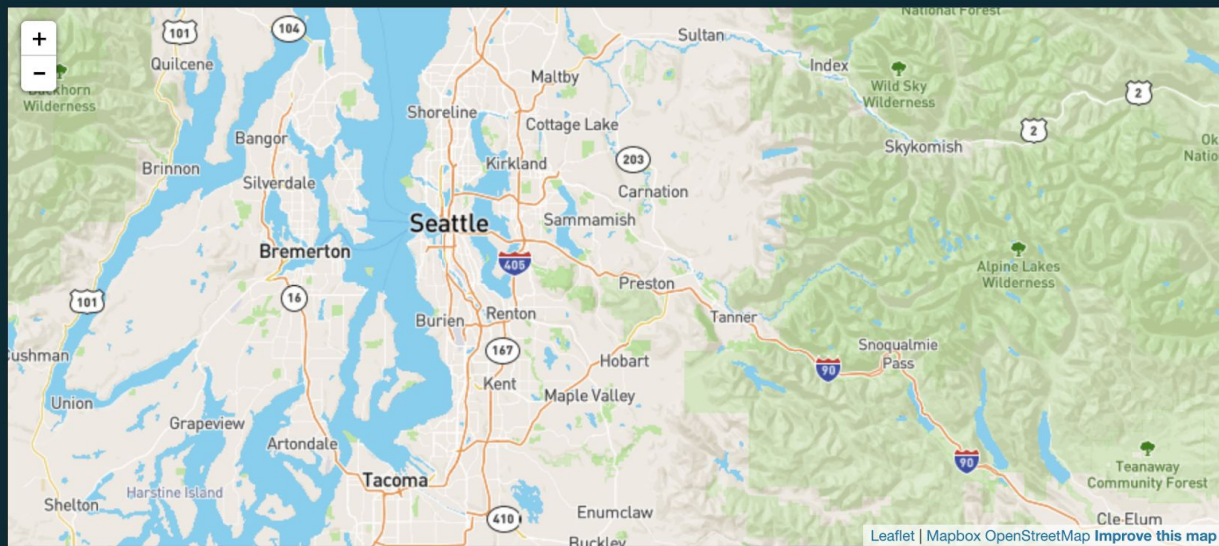
98105

SF Restaurant Location Map

Time Series: Number of Risk Violations

Bar Chart: Risk Violations by SF Postcodes

Bar Chart: Top 20 Restaurants Postcode



Use Case



[BMC Public Health](#), 2014; 14: 571.

Published online 2014 Jun 7. doi: [10.1186/1471-2458-14-571](https://doi.org/10.1186/1471-2458-14-571)

PMCID: PMC4057591

PMID: [24908104](https://pubmed.ncbi.nlm.nih.gov/24908104/)

Factors affecting food handling Practices among food handlers of Dangila town food and drink establishments, North West Ethiopia

[Ayehu Gashe Tessema](#),¹ [Kassahun Alemu Gelaye](#),¹ and [Daniel Haile Chercos](#)^{✉1}

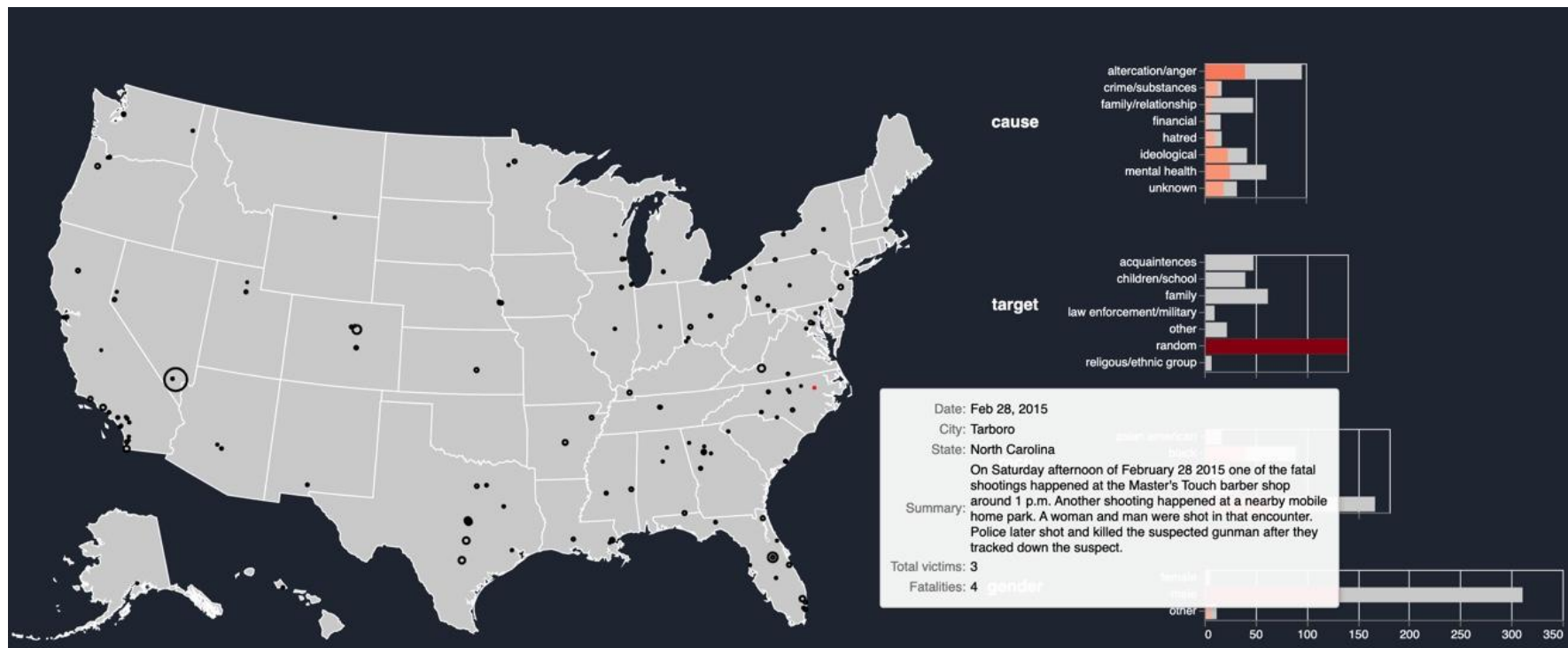
At least in NW Ethiopia, certain demographic information of the workers is associated with food handling practices.

Users can see visually if such correlation between restaurant locations and some of the same demographic information in Seattle.

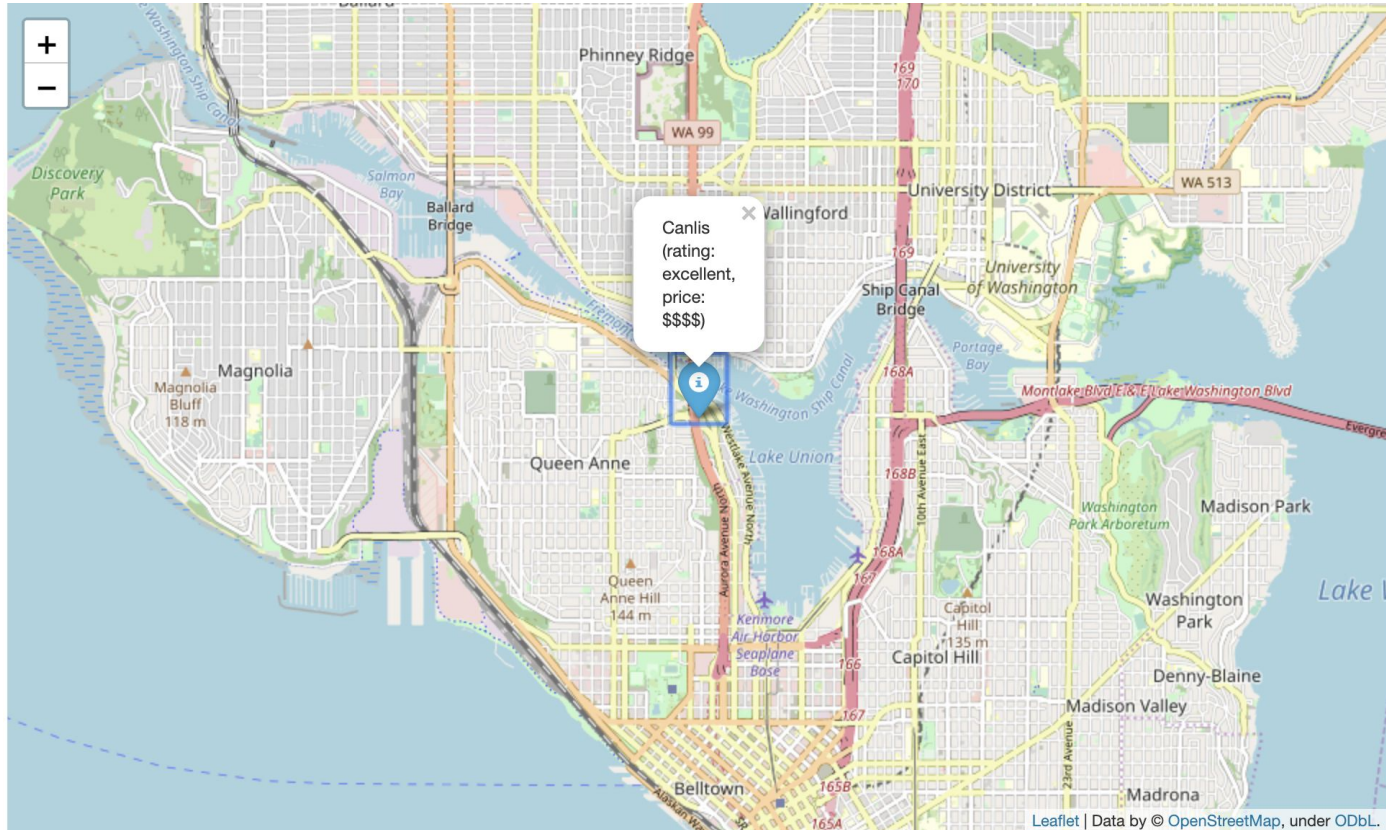
For example, seeing if satisfactory food inspection scores are associated with the marital status of the area surrounding the restaurant.

Libraries - Altair

We need a python library that has mapping features. We'll use [Altair](#) for maps.



Libraries - Folium



Altair v. Folium

Altair will enable:

- Filtering interaction from drop down or “radio button” menus
- Cross-filtering between correlation visuals and the map
- Seamless placement of each restaurant on the basemap, with interactive tooltips

The tradeoffs for taking advantage of Altairs interaction capability:

- Map won't zoom or pan
- May not be able to add streets

Folium will enable:

- Filtering interaction from drop down or “radio button” menus
- Typical “Google Maps” like features

But, it would be harder to:

- Cross-filter with other visuals

Libraries - Pandas

Pandas is a Python package providing fast, flexible, and expressive data structures designed to make working with “relational” or “labeled” data both easy and intuitive. We use Pandas for data analysis.



Libraries - Scikit-learn

Scikit-learn is a simple and efficient tools for data mining and data analysis.

We will use scikit-learn to build the regression model which predicts the overall trend using the demographic data (ACS 5-year estimates).

