Environmental Justice and Climate Change Sentiment

Bootcamp Final Project Bella Mullen

Dataset 1: Environmental Justice Index

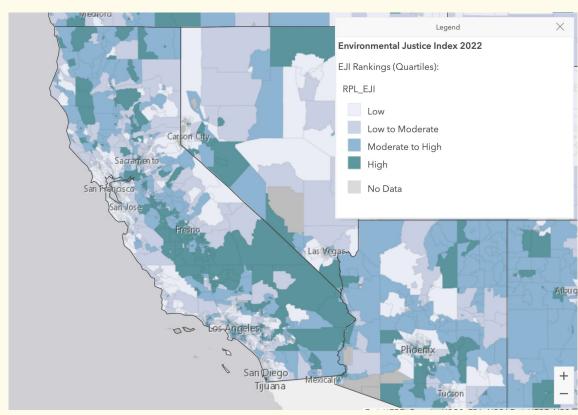
A tool designed to measure the cumulative impacts of environmental burden through the lens of human health

and health equity

This goal is for everyone to enjoy the same degree of protection from environmental and health hazards, and to live, learn, and work in a healthy environment

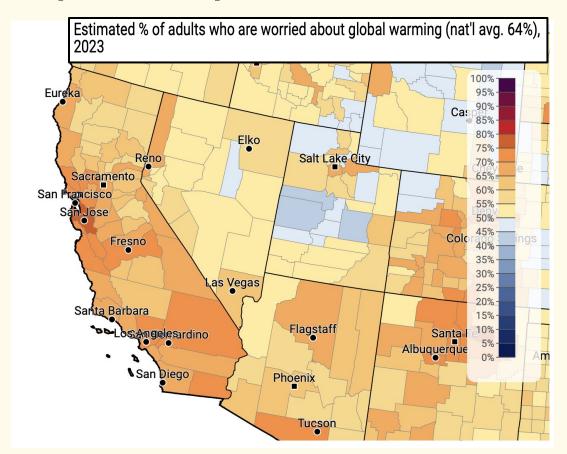
Data from:

- U.S. Census Bureau
- U.S. Environmental Protection Agency
- U.S. Mine Safety and Health Administration
- U.S. Centers for Disease Control and Prevention



Dataset 2: Yale Climate Opinion Maps

- Data through fall 2023
- Provides estimates of U.S. climate change beliefs, risk perceptions, and policy preferences at the state and local levels



```
y = eji_co_merge[['citizens_combined']]

# Splitting the data into training and testing sets with a 80-20 split
# 'random_state=42' ensures reproducibility of the results
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Initializing a Lasso regression model and setting up a parameter grid for alpha values to # Alpha controls the strength of the regularization (higher alpha -> more regularization)
```

Selecting relevant features (independent variables) from the dataset 'eji_co_merge' These
X = eji_co_merge[['E_OZONE', 'E_PM', 'E_DSLPM', 'E_TOTCR', 'E_NPL', 'E_TRI', 'E_TSD', 'E_RMP']

Performing grid search cross-validation to find the best alpha value for the Lasso model reg = GridSearchCV(lasso, parameters)

reg.fit(X_train, y_train)

Extracting the best Lasso model with the optimal alpha value
Using the best parameters
model = reg.best_estimator_

parameters = {'alpha': np.arange(0.01, 1, 0.05)}

lasso = Lasso()

```
# Creating a DataFrame to display the Lasso coefficients corresponding to each feature
lasso coefficients = pd.DataFrame({
    'Feature': column_names,
    'Coefficient': model.coef_
})
print(lasso_coefficients)
                                                                    Coefficient
                                                           Feature
           0
                                           Days above 03 standard
                                                                      -0.000000
                                        Days above PM2.5 standard
                                                                      -0.000000
           2
3
                                 Ambient concentrations of diesel
                                                                       0.000000
                                Probability of contracting cancer
                                                                      -0.000000
                     Proximity to EPA National Priority List site
                                                                       0.000000
            5
                    Proximity to EPA Toxic Release Inventory site
                                                                       0.000000
           6
               Proximity to EPA Treatment, Storage, and Dispos...
                                                                       0.000000
                       Proximity to EPA risk management plan site
                                                                      -0.000000
           8
                                          Proximity to coal mines
                                                                       0.000000
           9
                                          Proximity to lead mines
                                                                       0.000000
           10
                                         Proximity to green space
                                                                       0.188540
           11
                                            Houses built pre-1980
                                                                       0.037684
           12
                                                       Walkability
                                                                       0.000000
           13
                                            Proximity to railroad
                                                                       0.000000
                         Proximity to high-volume road or highway
                                                                       0.000000
           14
           15
                                             Proximity to airport
                                                                      -0.000000
            16
                                                Impacted watershed
                                                                       0.000000
```

Who should act?

• Citizens_combined:

 Estimated percentage who think citizens themselves should be doing more or less to address global warming.

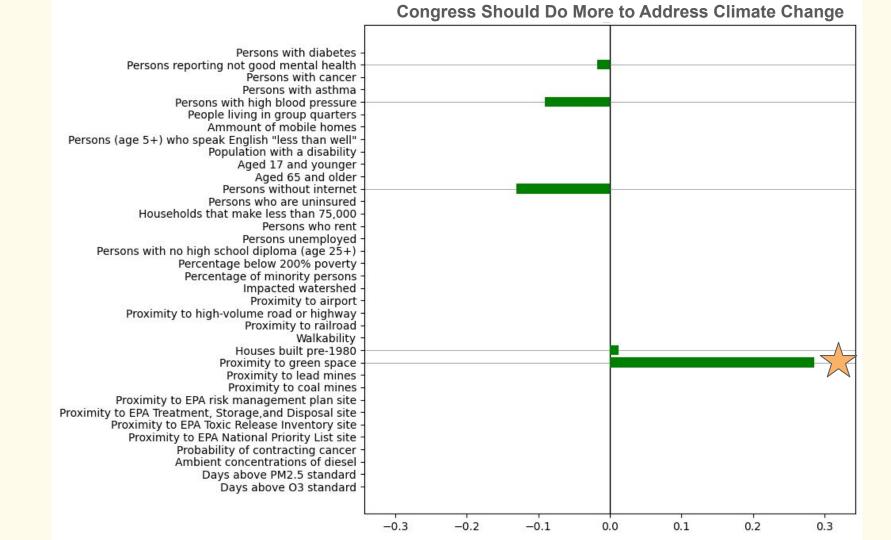
Congress_combined:

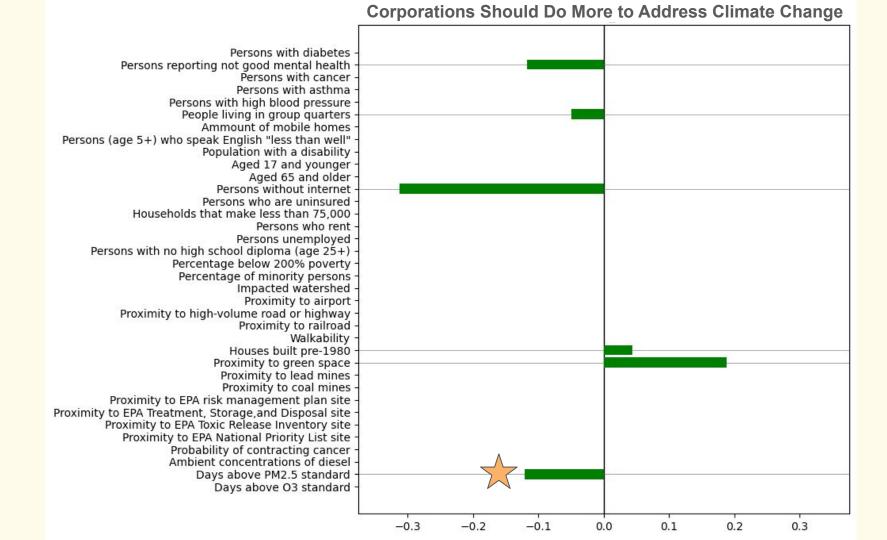
 Estimated percentage who think Congress should be doing more or less to address global warming

Corporations_combined:

 Estimated percentage who think corporations and industry should be doing more or less to address global warming

Citizen	s Feature	Coefficient
	Proximity to green space	0.188540
	Persons without internet	-0.258834
Congre	ss Feature	Coefficient
	Proximity to green space	0.285941
	Persons without internet	-0.130169
Corporat	ons Feature	Coefficient
	Proximity to green space	e 0.187994
	Persons without interne	t -0.313111





Who should act?

Proximity to green spaces

- Might reflect a higher environmental awareness or value placed on nature in these areas
- Might improve mental health overall

Lack of internet access

- Could be due to reduced access to information, awareness, and mobilization tools
- Could be attributed to a lack of other basic needs, which might take priority over climate change action

Who is harmed?

Futuregen_combined:

 Estimated percentage who think global warming will harm future generations

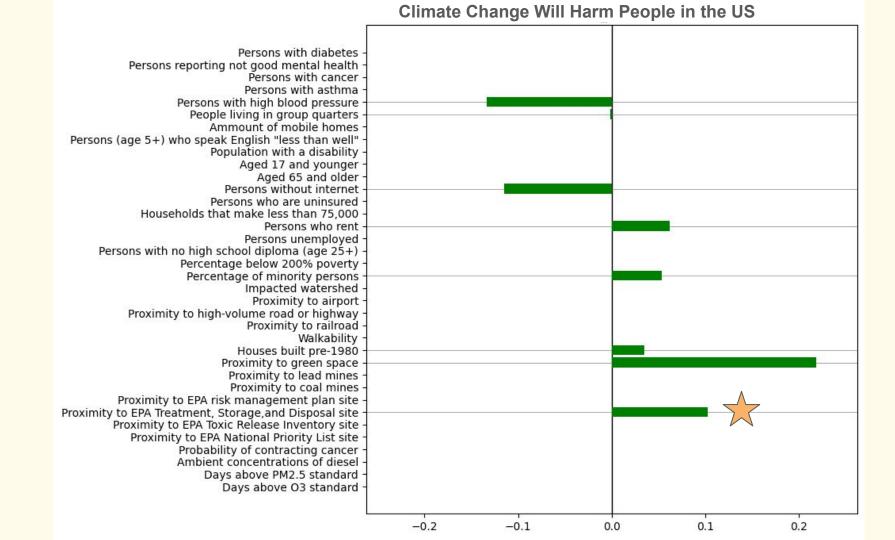
Harmus_combined:

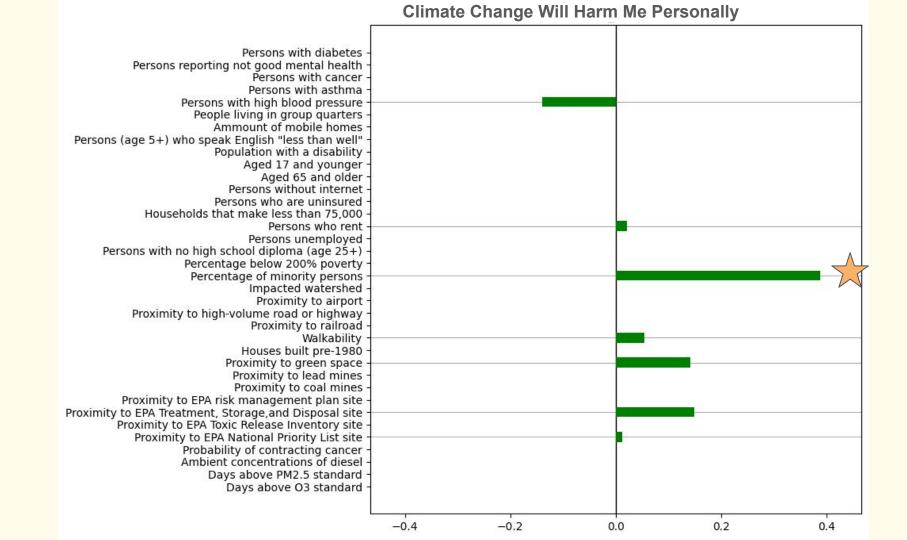
 Estimated percentage who think global warming will harm people in the US

Personal_combined:

 Estimated percentage who think global warming will harm them personally a moderate amount/a great deal

Future Gen Feature Coefficien	t
Proximity to green space 0.265506	3
Persons without internet -0.153935	5
Harm US Feature Coeffic	ient
Proximity to green space 0.218	492
Persons with high blood pressure -0.133	470
Personal Feature Coeffic	ent
Percentage of minority persons 0.388	241
Persons with high blood pressure -0.139	409





Who is harmed?

High blood pressure

 People with poor health may deprioritize global or future environmental risks in favor of personal health concerns

Minority populations

 Might be more likely to believe in personal harm from global warming, possibly due to increased vulnerability to environmental risks or cultural connections to the earth

Key Takeaways:

People's immediate surroundings (proximity to green space and internet access) and personal health challenges heavily impact climate change sentiment in the US.

Proximity to Green Space Increases Climate Awareness

People living near green spaces are consistently more likely to believe in the reality of global warming, feel
they have experienced its effects, discuss it more, and expect greater action from citizens, Congress, and
corporations to address climate change.

Limited Internet Access Reduces Climate Engagement

Areas with higher percentages of people lacking internet access show reduced belief in global warming, less
personal concern, and lower expectations for action from both individuals and institutions, likely due to limited
access to information on climate issues.

Health Issues Lower Climate Concern

 Populations with higher rates of health problems, such as mental health challenges, high blood pressure, and diabetes, tend to engage less with climate change, likely because immediate personal health concerns take priority over broader environmental worries.

Data Cleaning Process:

- Merged data by county
- The EJI data was by census tract so I calculated a weighted average for each county
- Weighted by population
- Could have lost neighborhood-level variation if compared by county

```
# aggregate census tract data into county-level data by calculating weighted averages
#Calculate the weighted average for each theme using E_TOTPOP (population variable) as
def weighted_avg(group, avg_name, weight_name):
    """Calculate weighted average."""
    d = group[avg_name]
    w = group[weight_name]
    return (d * w).sum() / w.sum()
# Group by COUNTY and apply weighted average or sum where appropriate
eji_clean_grouped = eji_clean.groupby('COUNTY').apply(
    lambda x: pd.Series({
        'E_TOTPOP': x['E_TOTPOP'].sum(),
        'E_OZONE': weighted_avg(x, 'E_OZONE', 'E_TOTPOP'),
        'E_PM': weighted_avg(x, 'E_PM', 'E_TOTPOP'),
        'E_DSLPM': weighted_avg(x, 'E_DSLPM', 'E_TOTPOP'),
        'E_TOTCR': weighted_avg(x, 'E_TOTCR', 'E_TOTPOP'),
        'E_NPL': weighted_avg(x, 'E_NPL', 'E_TOTPOP'),
                                 'E_TRI', 'E_TOTPOP'),
        'E_TRI': weighted_avg(x,
        'E_TSD': weighted_avg(x,
                                 'E_TSD', 'E_TOTPOP'),
                                 'E_RMP', 'E_TOTPOP'),
        'E_RMP': weighted_avg(x,
        'E_COAL': weighted_avg(x, 'E_COAL', 'E_TOTPOP'),
        'E_LEAD': weighted_avg(x,
                                  'E_LEAD', 'E_TOTPOP'),
        'E_PARK': weighted_avg(x, 'E_PARK', 'E_TOTPOP'),
        'E_HOUAGE': weighted_avg(x, 'E_HOUAGE', 'E_TOTPOP'),
        'E_WLKIND': weighted_avg(x, 'E_WLKIND', 'E_TOTPOP'),
        'E_RAIL': weighted_avg(x, 'E_RAIL', 'E_TOTPOP'),
        'E_ROAD': weighted_avg(x, 'E_ROAD', 'E_TOTPOP'),
        'E_AIRPRT': weighted_avg(x, 'E_AIRPRT', 'E_TOTPOP'),
        'E_IMPWTR': weighted_avg(x, 'E_IMPWTR', 'E_TOTPOP'),
        'EP_MINRTY': weighted_avg(x, 'EP_MINRTY', 'E_TOTPOP'),
        'EP_POV200': weighted_avg(x, 'EP_POV200', 'E_TOTPOP'),
        'EP_NOHSDP': weighted_avg(x, 'EP_NOHSDP', 'E_TOTPOP'),
        'EP_UNEMP': weighted_avg(x, 'EP_UNEMP', 'E_TOTPOP'),
        'EP_RENTER': weighted_avg(x, 'EP_RENTER', 'E_TOTPOP'),
        'EP_HOUBDN': weighted_avg(x, 'EP_HOUBDN', 'E_TOTPOP'),
        'EP_UNINSUR': weighted_avg(x, 'EP_UNINSUR', 'E_TOTPOP'),
        'EP_NOINT': weighted_avg(x, 'EP_NOINT', 'E_TOTPOP'),
        'EP_AGE65': weighted_avg(x, 'EP_AGE65', 'E_TOTPOP'),
        'EP_AGE17': weighted_avg(x, 'EP_AGE17', 'E_TOTPOP'),
        'EP_DISABL': weighted_avg(x, 'EP_DISABL', 'E_TOTPOP'),
        'EP_LIMENG': weighted_avg(x, 'EP_LIMENG', 'E_TOTPOP'),
        'EP_MOBILE': weighted_avg(x, 'EP_MOBILE', 'E_TOTPOP'),
        'EP_GROUPQ': weighted_avg(x, 'EP_GROUPQ', 'E_TOTPOP'),
        'EP_BPHIGH': weighted_avg(x, 'EP_BPHIGH', 'E_TOTPOP'),
        'EP_ASTHMA': weighted_avg(x, 'EP_ASTHMA', 'E_TOTPOP'),
        'EP_CANCER': weighted_avg(x, 'EP_CANCER', 'E_TOTPOP'),
        'EP_MHLTH': weighted_avg(x, 'EP_MHLTH', 'E_TOTPOP'),
        'EP_DIABETES': weighted_avg(x, 'EP_DIABETES', 'E_TOTPOP')
).reset_index()
# Add the GEOID column, ensuring it matches the county
eii clean grouped['GEOID'] = [eii clean[eii clean['COUNTY'] == county]['GEOID'].iloc[0]
#save back into eji_clean
eji_clean = eji_clean_grouped
```

Voting/ legislation:

• Gwvoteimp_combined:

 Estimated percentage who say a candidate's views on GW are important to their vote

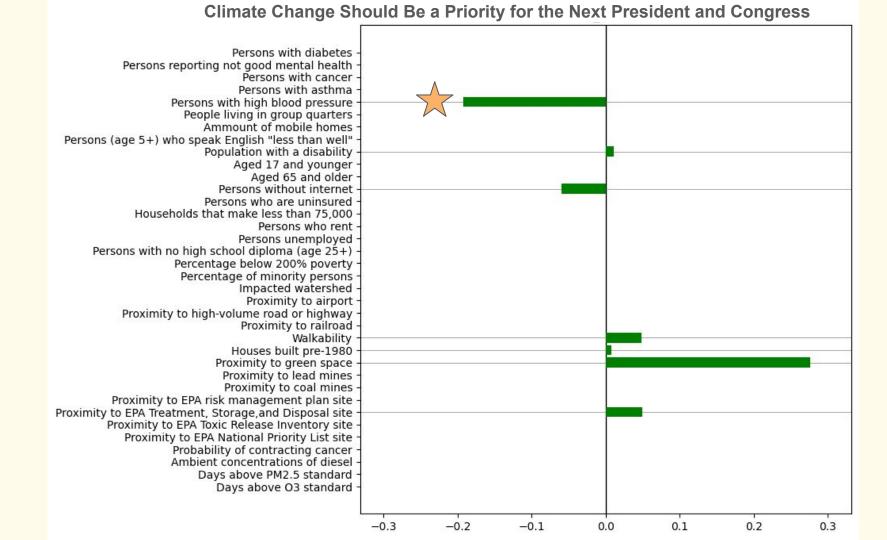
Priority_combined:

 Estimated percentage who say global warming should be a high priority for the next president and Congress

• Regulate_combined:

 Estimated percentage who somewhat/strongly support regulating CO2 as a pollutant

Gwvoteim	o Feature	Coefficient
	Proximity to green space	0.328951
	Persons without internet	-0.096976
Priority	Feature	Coefficient
	Proximity to green space	0.276327
Per	sons with high blood pressur	e -0.192180
Regulate	Feature	Coefficient
	Proximity to green space	0.173203
	Persons without internet	-0.388985



Climate change discussion:

Discuss_combined:

 Estimated percentage who discuss global warming with friends and family

Mediaweekly_combined:

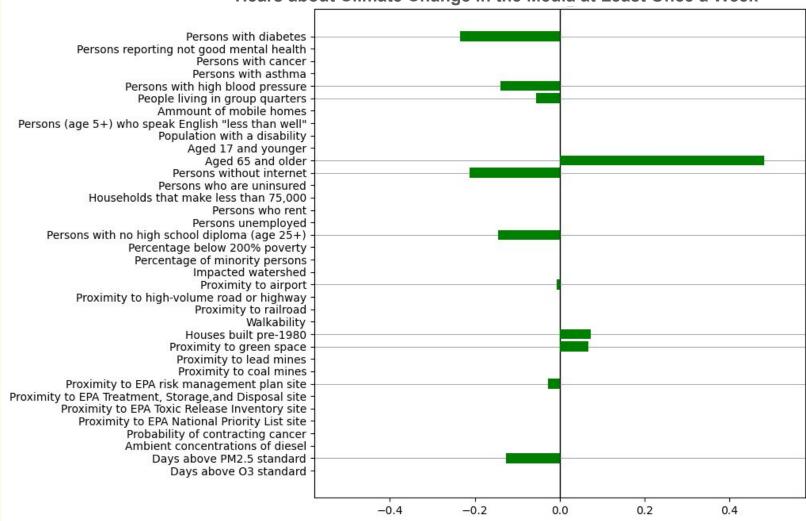
Estimated percentage who hear about global warming in the media

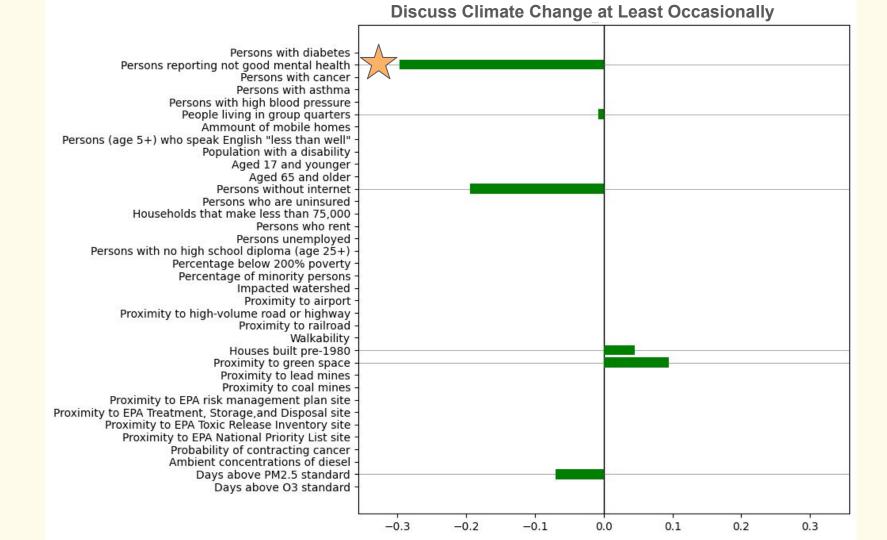
Human_combined:

Estimated percentage who think that global warming is caused mostly by human activities

Dis	scuss	Feature	Coefficient
		Proximity to green space	0.094367
	Pers	ons with not good mental h	nealth -0.297540
	Mediawee	ekly Feature	Coefficient
		Aged 65 and older	0.481906
		Persons with diabete	s -0.235376
	Human	Feature	Coefficient
		Proximity to green space	0.314892
		Persons with asthma	-0.098549

Hears about Climate Change in the Media at Least Once a Week





Climate change discussion:

- Older adults (aged 65 and older)
 - Could be caused by higher traditional media consumption rates and exposure to climate-related content

Climate change concern:

Exp_combined:

 Estimated percentage who responded that they have personally experienced the effects of global warming

Happening_combined:

Estimated percentage who think that global warming is happening

Worried_combined:

 Estimated percentage who are not very/not at all worried about global warming

Proximity to green space 0.241683 Persons without internet -0.192961 Happening Feature Coefficien	
Happening Feature Coefficien	
	nt
Proximity to green space 0.25998	33
Persons without internet -0.22057	75
Worried Feature Coeffici	cient
Proximity to green space 0.278	8782
Persons with high blood pressure -0.129	9625

