

Is air pollution correlated to inter-state migration in the US?

EDA Final Project

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Abstract

This is my dream abstract. **Keywords: Air Pollution, Migration.**

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Rationale and Research Questions

This project looks at the effects of air pollution on inter-state migration in the United States using the Air Quality Index datasets from EPA and the Population Migration data from the IRS for the period 2013-2020.

Evidence from middle-income countries shows that air pollution has negative impacts on several health and economic outcomes, such as mortality rates, health expenditures, mental health, hours worked, labor productivity and income. Additionally, other studies have shown how migration decisions are affected by air pollution. For instance, Chen, S., Oliva, P., & Zhang, P. (2022) found that a 10 percent increase in air pollution, holding everything else constant, reduces population through net outmigration by about 2.8 percent in a given county in China (see Chen, S., Oliva, P., & Zhang, P. (2022). The effect of air pollution on migration: Evidence from China. *Journal of Development Economics*, 156, 102833. <https://doi.org/10.1016/j.jdeveco.2022.102833>)

Our hypothesis is that there is a positive relationship between high air pollution and migration outflows, that is, the highest the pollution registered by the Air Quality Index (AQI) in a state in a given year, the higher the number of people leaving that state the same year.

Dataset Information

1. Air Quality Datasets

The Air Quality Index dataset provides annual information per county about the maximum values reached by the AQI, the number of days in which this index reached values considered unhealthy, and the number of days with PM2.5 particles recorded. The information is captured by the EPA local air quality stations. These datasets are available at https://aqs.epa.gov/aqswweb/airdata/download_files.html. We create a dataset with the information from 2013 to 2020 and aggregate the information at the state level. Likewise, we calculate the state averages for the variables Unhealthy.Days, Max.AQI, and Days.PM2.5.

2. Inter-state migration datasets

The State-to-State outflows dataset provides annual information at the State level about the number of people whose reported home address changed in their individual income tax returns from one year to the other. These datasets are available at <https://www.irs.gov/statistics/soi-tax-stats-migration-data>. The Inter-state migration datasets do not include the variable “Year”, so we create it from 2013 to 2020.

According to the dictionary for this dataset, the variable “y2_statefips” and the code “96” refer to the total outflows of migrants for each state in a given year. Therefore, we filtered the previous dataset by that value. Additionally, we change this variable’s name to “FIPS_Code”, so later we can merge this dataset with the AQI dataset.

3. Scraping the FIPS codes

The AQI Dataset has the names of each State in the US, but it does not have the code, which is the variable we need to merge this data with the migration one. We scrape the FIPS codes from the webpage (<https://www.bls.gov/respondents/mwr/electronic-data-interchange/appendix-d-usps-state-abbreviations-and-fips-codes.htm>) and we create a data frame. Then, we merge this dataset with the AQI dataset by the variable “State”.

4. Merging the AQI and Migration datasets

We merge and arrange the AQI and the migration datasets by the variables “FIPS_Code” and “Year”. The resulting dataset has information from 2013 to 2020. According to the dictionary for the migration dataset, the variable “n2” refers to the number of individuals who migrated to other states. To facilitate the interpretation, we changed the name of the variable to “Migrants.outflows”. We create a subset of the previous dataset with the variables of interest: FIPS_Code, Year, State, Avg.Unhealthy.Days, Avg.Days.AQI, Avg.Days.PM2.5, and Migrants.outflows. This is the dataset that we will use in our analysis.

The final dataset has 7 variables: FIPS Code, Year, State, Avg.Unhealthy.Days, Avg.Max.AQI, Avg.Days.PM2.5, and number of migrants. In the table 1 we can see the characteristics of the dataset for years 2013 and 2020. Each year has information for the 50 states of the USA (we do not include in the analysis the Distric of Columbia, Puerto Rico and the Virgin islands). For 2013, on average each state showed 0.56 days with air quality classified as “unhealthy”, the maximum value reach by the AQI was 114.89, the number of days with PM2.5 registered were 116.57, and 135,657 people migrated to a different state. For 2020, on average each state showed 0.90 days with air quality classified as “unhealthy”, the maximum value reach by the AQI was 122.81, the number of days with PM2.5 registered were 121.52, and about 138,954 people migrated to a different state.

Table 1: Summary statistics for the final dataset 2013-2020

Variable	Type	Obs	Min	Median	Media	Max
<i>Year=2013</i>						
FIPS Code	int	50				
State	chr	50				
Avg.Unhealthy.Days	num	50	0.00000	0.03348	0.56471	6.13333
Avg.Max.AQI	num	50	73.00	107.22	114.89	240.85
Avg.Days.PM2.5	num	50	17.43	110.33	116.57	309.00
Migrants.outflows	num	50	16540	95343	135657	532619
<i>Year=2020</i>						
FIPS Code	int	50				
State	chr	50				
Avg.Unhealthy.Days	num	50	0.00000	0.09762	0.90793	14.20755
Avg.Max.AQI	num	50	58.75	98.25	122.81	430.35
Avg.Days.PM2.5	num	50	16.11	105.87	121.52	304.39
Migrants.outflows	num	50	16378	92877	138954	684935

Exploratory Analysis

Air Quality Indicators

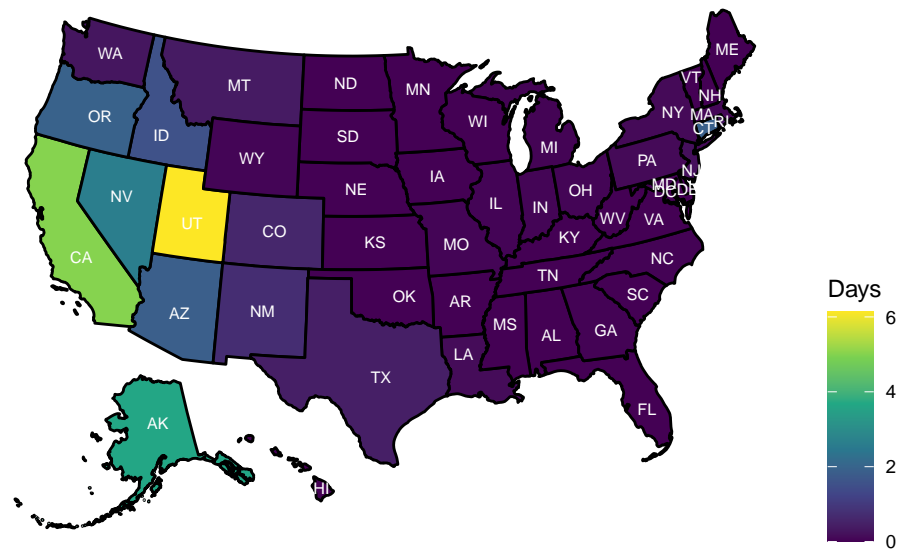


Figure 1: Average Unhealthy Days by State 2013

Migration outflows by State

Air Quality and migration outflows

Analysis

6. Correlations

##

Pearson's product-moment correlation

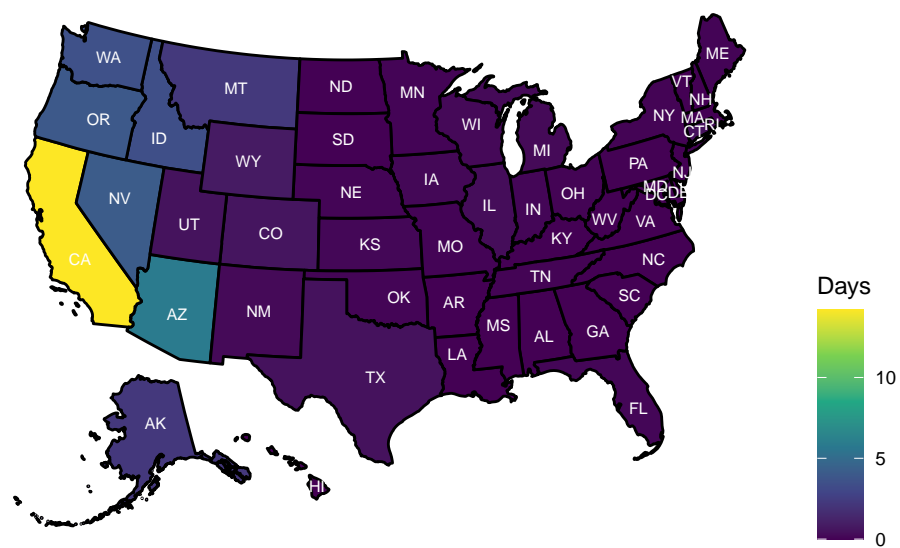


Figure 2: Average Unhealthy Days by State 2020

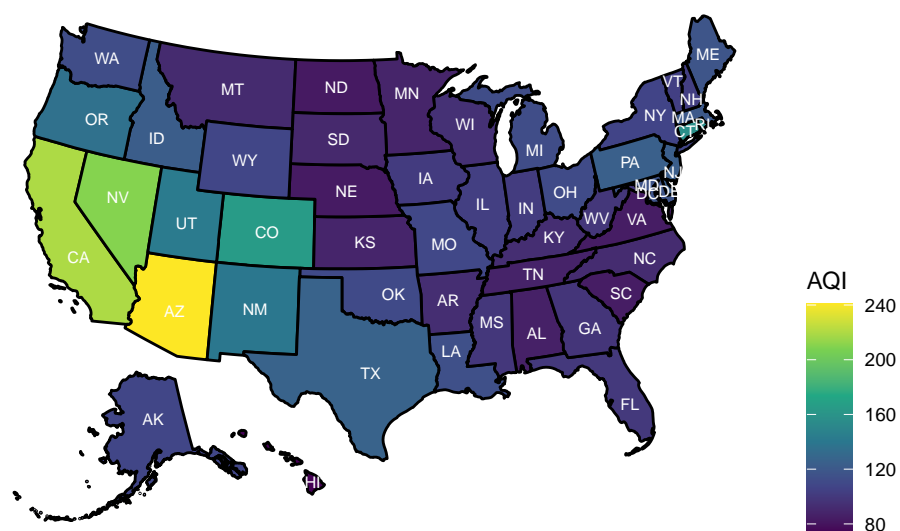


Figure 3: Average Max AQI value by State 2013

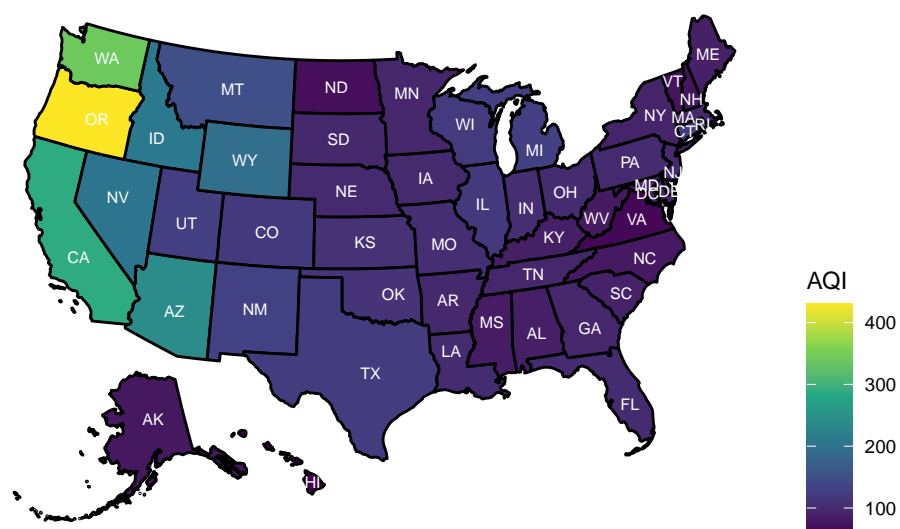


Figure 4: Average Max AQI value by State 2020

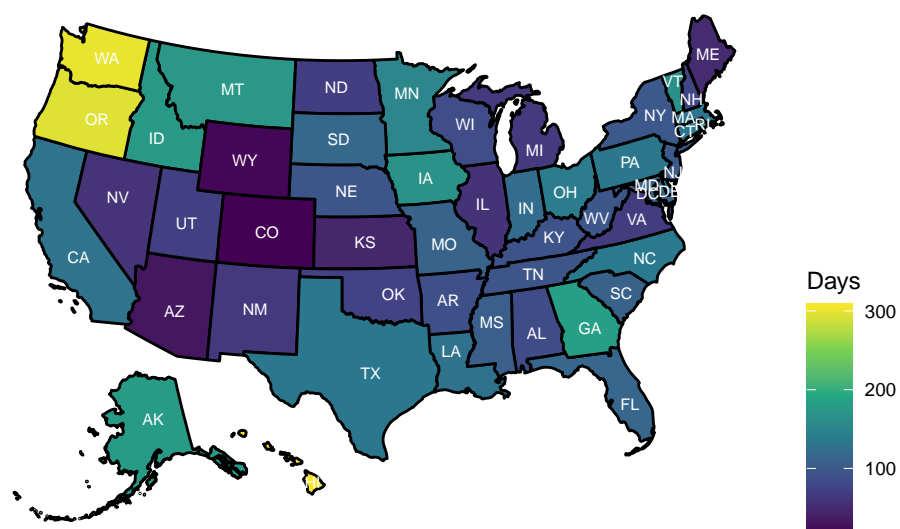


Figure 5: Average PM2.5 by State 2013

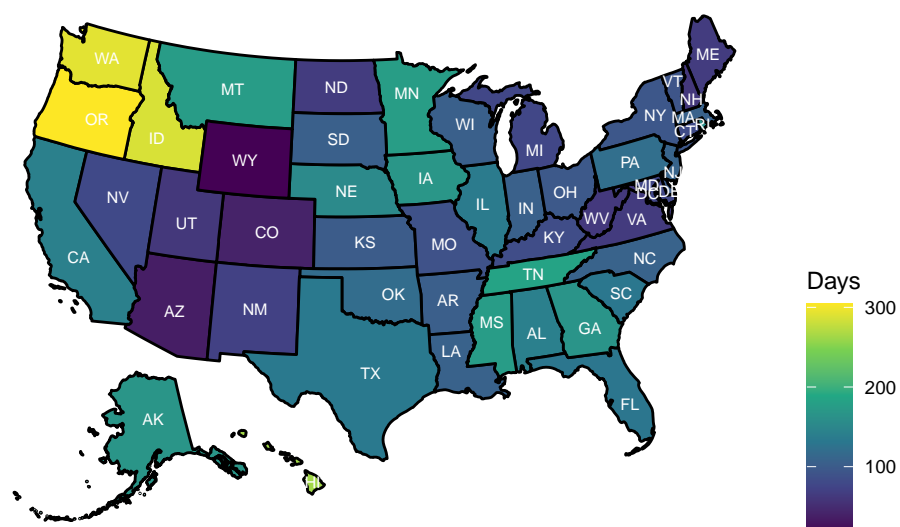


Figure 6: Average PM2.5 by State 2020

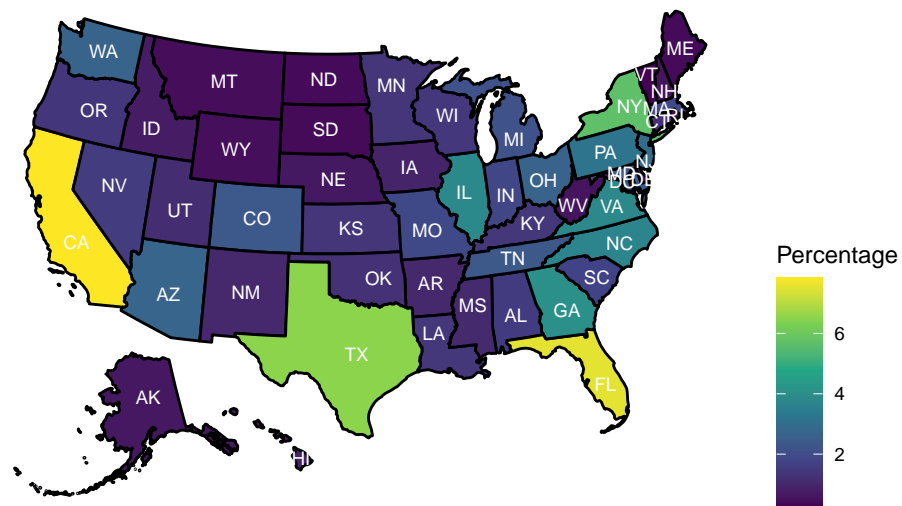


Figure 7: Percentage of migrants' outflows by State in 2013

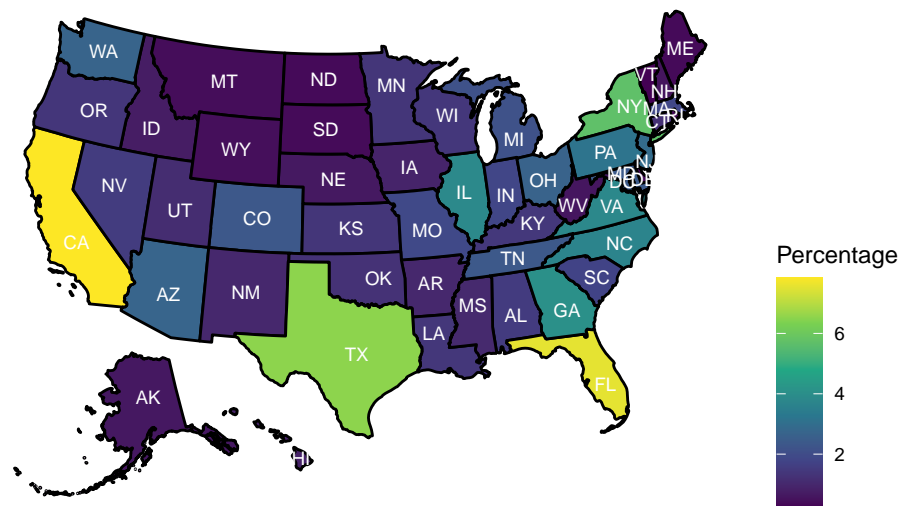


Figure 8: Percentage of migrants' outflows by State in 2020

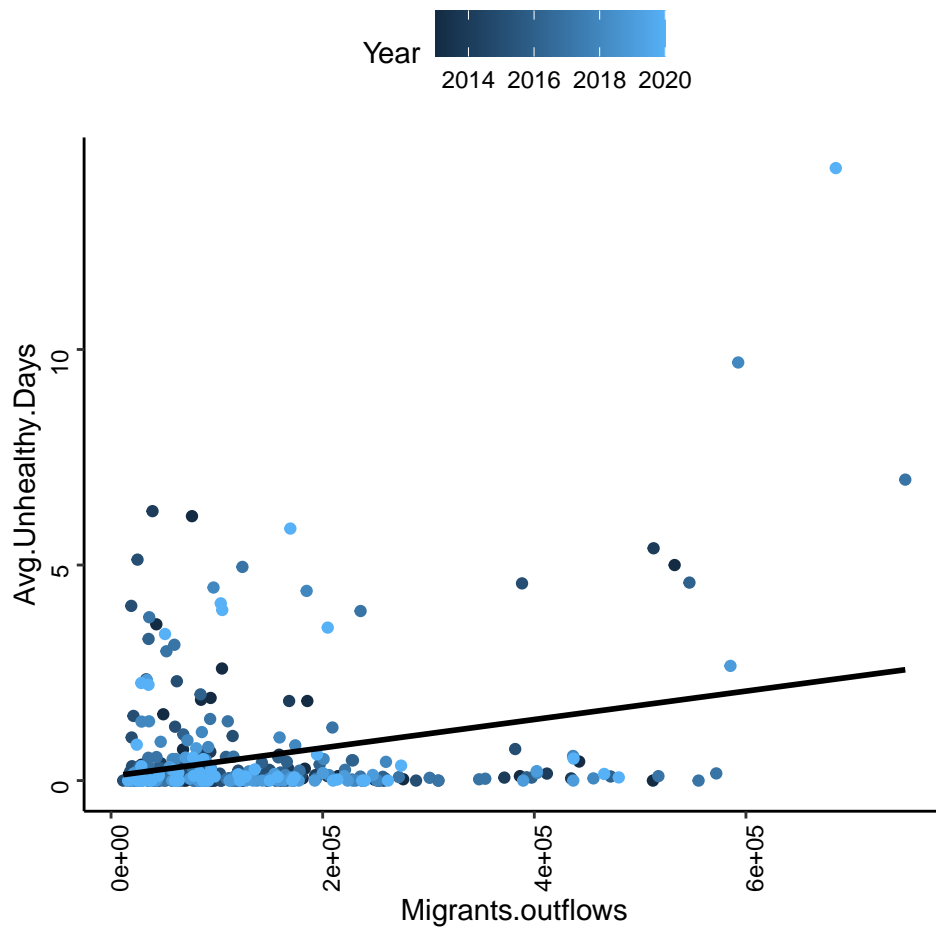


Figure 9: Relationship between Avg. Unhealthy Days and Migration outflows between 2013 and 2020


```

##
## data:  AQI_Mig.outflows_by.state_2013_2020$Avg.Unhealthy.Days and AQI_Mig.outflows_by.state_2013_2020$
## t = 6.0936, df = 398, p-value = 2.609e-09
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.1997911 0.3793075
## sample estimates:
##      cor
## 0.2921202

##
## Pearson's product-moment correlation
##
## data:  AQI_Mig.outflows_by.state_2013_2020$Avg.Max.AQI and AQI_Mig.outflows_by.state_2013_2020$Migran
## t = 5.341, df = 398, p-value = 1.558e-07
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.1647361 0.3478420
## sample estimates:
##      cor
## 0.2586106

##
## Pearson's product-moment correlation
##
## data:  AQI_Mig.outflows_by.state_2013_2020$Avg.Days.PM2.5 and AQI_Mig.outflows_by.state_2013_2020$Mi
## t = -0.4209, df = 398, p-value = 0.6741
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  -0.11889869  0.07711834
## sample estimates:
##      cor
## -0.02109288

```