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Data 512

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Final report

**Exploring the impact of wildfire smoke on the health care industry of Leavenworth, Kansas**

**Introduction:**

In this work, the author was tasked with conducting a human-centered data analysis project concerned with measuring a specific impact that wildfire smoke has on a particular city, and from there creating a policy recommendation for said city. Specifically, the author was assigned to consider the city of Leavenworth, Kansas and chose to measure the impacts on the healthcare industry. As expected, the main motivation behind this work is to save Leavenworth money and resources through education. To answer the question of why this analysis is important, research [INSERT REF] shows that wildfire smoke is dangerous to inhale over both the short and long term. This means that if the city of Leavenworth is not aware of current smoke exposure trends, they will likely be unable to mitigate incurring extra avoidable health and economic repercussions due to wildfire smoke. Tangentially, this work may be considered interesting from a technical standpoint because it served as an exercise that demonstrates the difficulty in solving real world problems due to reasons such as: Suboptimal datasets, operating on multiple assumptions, and combining datasets for analysis. The rest of this report will go into details about related work, methods used, findings, limitations, and finally the proposed policy recommendation.

**Background/Related work:**

In this work, the first main research question that was asked was a hard requirement by the course staff for the first portion of work named “common analysis”. Specifically, the question was “What are the estimated smoke impacts on your assigned city for the last 60 years?” In this case, the assigned city was Leavenworth, Kansas. After the first smoke estimates were determined through the work done for the “common analysis”, new research questions were chosen by the author to be answered in the second portion of work called the “extended analysis”. The research questions regarding the extended analysis are: “What impacts does wildfire smoke have on the healthcare industry in Leavenworth, Kansas”, “How do specific health conditions relate to emergency department visits”, and “What are the increased risks of emergency department visits for chosen health conditions when wildfire smoke is involved”. Given the research questions, the next paragraphs mention and discuss the use of various different research sources, data sources, and algorithms used to attempt to answer the proposed questions.

In this section, each research or information source used will be discussed. First, in the creation of an annual smoke estimate, we utilized [1] to determine the volume of smoke produced by burning one kilogram of wood, which according to the source was 87.5 meters squared per kilogram. Second, sources [2], [3], and [4] were used as justification to why specific health conditions such as asthma, cardiovascular disease (CVD), chronic obstructive pulmonary disease (COPD), myocardial infarctions (MI), and strokes were considered in this analysis. Next, the results from [5] were used to estimate that 7.7 percent of people in Leavenworth, Kansas have asthma regardless of the year, the results from [6] were used to estimate that 7.53411 percent of people in Leavenworth, Kansas have CVD regardless of the year, the results from [7] were used to estimate that 6.4 percent of the people in Leavenworth, Kansas have COPD regardless of the year, the results from [8] and [9] were used to estimate that 0.241 percent of the people in Leavenworth, Kansas have experienced a heart attack each year, and the results from [9] and [10] were used to estimate that 0.237 percent of the people in Leavenworth, Kansas have experience a stroke each year. After that, sources [11], [12], [13], [14] were used to determine that asthma, CVD, COPD, heart attacks, and strokes account for 1.32 percent, 11.4 percent, 12.44 percent, 0.6 percent, and 0.6 percent of all emergency department (ED) visits in Leavenworth, Kansas regardless of the year. Similarly, the number of ED visits per year was estimated to be equivalent to 40 percent of the population of that year due to [15]. Next, sources [16], [17], [16], [18], and [17] were used to determine that the number of ED visits for asthma, CVD, COPD, heart attacks, and strokes could increase by up to 2.2, 5, 1.39, 42, and 11 percent respectively depending on how severe the annual smoke estimate was. Finally, [19] was used to estimate that each ED visit cost 530 dollars.

This work utilized three different sources of data. First, we used a dataset titled “Combined wildland fire datasets for the United States and certain territories, 1800s-Present”. This dataset was published [here](https://www.sciencebase.gov/catalog/item/61aa537dd34eb622f699df81) on 12/08/2021. In short, the dataset is a combination of many smaller or incomplete United States fire datasets and stores various metadata for historical fires such as: Area, location, date, polygon coordinate geometry, etc. The second source of data utilized the [Air Quality System (AQS) API](https://aqs.epa.gov/aqsweb/documents/data_api.html), which contained “ambient air sample data collected by state, local, tribal and federal air pollution control agencies”. The contents of data called by the API include recorded station data such as: Particulate type, particulate amount measured, date of collection, location of the station, etc. Third, this work utilized United States census data, which can be found [here](https://data.census.gov/). United States census data is quite vast and can include many different schemas which cannot be fully described here. For this work specifically, population estimates, age estimates, healthcare coverage estimates, and employment by industry estimates were used.

In terms of specific algorithms, two main methods were used. First, linear regression [20] was utilized to attempt to see if a custom smoke estimate could be predicted using the number of acres a wildfire burned and the distance the wildfire was from Leavenworth, Kansas. The reason this model was selected is due to course requirements stating a predictive model must be used in the common analysis portion of work. Second, two autoregressive integrated moving average (ARIMA) [21] models were used to make projections of both a custom smoke estimate and the number of healthcare/social/educational worker in Leavenworth to 2050. ARIMA was chosen specifically because it is a simple model that predicts future time series values based on previous values.

**Methodology:**

For the first part of this section, the first portion of work, called the common analysis will be summarized. This section of work was mainly concerned with answering the research question concerned with finding the estimated smoke impact on Leavenworth, Kansas for the last 60 years. To answer that question, multiple different steps were performed in a Jupyter notebook. First, the combined wildfire datasets were read in using the “geojson” python module due to the fact that data was in JavaScript Object Notation (JSON) format. Next, in order to reduce processing time and to eliminate fires that were not relevant to future analyses, the distance between each wildfire and Leavenworth, Kansas was calculated, and fires that occurred greater than 1250 miles away from Leavenworth were not considered for analysis. This calculation was done by getting the average distance between the EPSG:4326 [22] coordinates of Leavenworth and each wildfire polygon coordinate. The decision to calculate distance as an average was made because it is simple to calculate and easy to understand. From there an annual smoke estimate was created using wildfire data and the research from [1]. Specifically, the formula used for calculating the smoke estimate for a specific wildfire is: , where A is the area in square meters burned by the wildfire, V is the volume of smoke in meters squared produced by burning one kilogram of wood, W is the amount of wood in kilograms per square meter of forest, and D is the average distance from the wildfire to Leavenworth. To extend that, the annual smoke estimate was simply a sum over all fires for that year, which can be expressed as: , where n is the number of fires that occurred that year and k is specific fire. The reason this method was used as a wildfire smoke estimate was due to the fact that the only relevant information available regarding smoke production from the fire dataset was the location and areas of the fires. Specifically, we assumed smoke dispersion is inversely proportional to the distance of the wildfire to the fifth power. We used the fifth power to keep the relative scale of the smoke estimate low so that it could be compared to the AQI. Additionally, area, volume, and mass were used due to the fact that [1] had some relevant information about the volume of smoke produced by burning wood. In terms of the annual estimate, the decision to make it a sum over all wildfires for that year was due to the fact that data was sparse, so using daily or weekly averages may have led to the estimate being near zero at all times. After calculating an annual smoke estimate, the next step was to estimate monthly AQI for Leavenworth, Kansas by utilizing the AQS API. Specifically, AQI was used because it was a requirement by the course staff. In terms of how the AQI estimate was derived, this work only considered using PM10 and PM2.5 data for the four nearest AQI measurement stations, as no station directly existed in the city. To get the AQI, the nearest four stations were looked at, their PM10 and PM2.5 values were averaged, and the largest average was chosen as the monthly AQI estimate for Leavenworth. The reason I looked at the nearest four stations was arbitrary, the reason only PM10 and PM2.5 were chosen was because wildfires were a mentioned cause of them according to [22], and the greatest average was chosen because it was simple. Next, a linear regression and ARIMA models were used to predict and forecast the annual smoke estimate to 2049. The reasons these models were specifically used was due to requirement by the course staff and to maintain simplicity. Finally, four visualizations showing the number of fires that occurred every 50 miles from Leavenworth, the annual acres burned by wildfires, the annual projected smoke estimate, and the AQI estimate were created for the sake of understanding what the wildfire trends are.

The extended analysis portion of work required the answering of a new research question, which was “What impacts does wildfire smoke have on the healthcare industry in Leavenworth, Kansas?” To answer this question, the first step performed was loading in US census data using the Python module ‘Pandas’ due to its simplicity. The specific data that was loaded were the population sizes, old age dependency ratios, total number of people employed, total number of healthcare, education, and social workers, and the percentage of the population without insurance for Leavenworth, Kansas. From there, four basic time-series visualizations were created to show various population trends in Leavenworth. These visualizations showed the population over time, the percentage of working-aged people employed in healthcare, education, and social work over time, the old age dependency ratio over time, and the percentage of the population with no health insurance in Leavenworth. Next, in order to be able to properly estimate future healthcare impacts in Leavenworth, the population of Leavenworth was projected to 2050 using a line equation, which again was chosen for simplicity’s sake. From there, the number of residents in Leavenworth with five specific health conditions were estimated using both the projected population counts and the results from [5], [6], [7], [8], [9], and [10]. Specifically, the population was multiplied by a percent factor was described in the literature, and the decision to use a simple multiplier was to maintain simplicity. Then, the number of estimated emergency department visits were calculated for the five specific health conditions using results from [11], [12], [13], [14] and the estimated visits were visualized to better understand trends. After that, the estimated smoke values from the common analysis section of work were combined with the total number of ED visits due to specific health conditions again using simple multipliers from [16], [17], and [18] to get the number of ED visits attributed to specific health conditions when smoke was present. The reason this was done was to help answer the question concerning what the increase in ED visits would be due to wildfire smoke exposure. Finally, visualizations were made concerning the total number of projected ED visits, the extra occurring costs from ED visits, and the number of healthcare workers per projected ED visit to help understand trends, which would be used to make the final policy recommendation. In order to visualize the number of healthcare workers into 2050, another ARIMA [21] model was used for simplicity’s sake.

When it comes to human-centered design, there were two main human-centered considerations that influenced how the project was conducted. First, the main human centered component is reproducibility. Being one of the main points of this work, the course staff were adamant that the project should have complete documentation, data descriptions, license files, etc. With that being said, this work is committed to being reproducible due to the fact that both the Jupyter notebook files are well documented, a complete README.md file exists, and an anaconda python development environment file is provided so that future researchers can quickly emulate my setup. The other main human-centered aspect of this work concerns algorithmic transparency. Specifically, I made sure to use simple models in the work such as linear regression, ARIMA, and line equations to keep analyses as interpretable as possible.

**Findings:**

As previously mentioned, the analyses performed produced many visualizations which can be used to infer general findings that were then used to justify the final policy recommendations made to the city council. Below will discuss each finding and provide a visualization.

The first two findings in this work, which were from the common analysis portion of work, are that both the estimated AQI levels and estimate wildfire smoke exposure levels are increasing in Leavenworth, Kansas. For example, the AQI graph below show how the AQI levels have seen an upward trend from 1990 and that the AQI is starting to shift from the designated ‘Good’ safety level to the ‘Moderate’ safety level. Additionally, the projected smoke estimate graph shows how the smoke levels have drastically increased from 1990 to 2020 and the smoke levels are projected to be greater on average between 2020-2040 than they were between 2000-2020.

A graph showing a safety level

Description automatically generated A graph of smoke

Description automatically generated

The next finding in this work is that the old age dependency ratio, which is “the number of individuals aged 65 and over per 100 people of working age defined as those at ages 20 to 64” [24], has been increasing in Leavenworth, Kansas from 2010 to 2021. After that, using the ARIMA projection along with population data, the next finding was that the population of Leavenworth, Kansas has been rising since at least 2010 and is projected to keep linearly increasing until 2050.

A line graph with numbers and a line

Description automatically generatedA graph with a line

Description automatically generated

From there the total population employed in healthcare, education, or social work was plotted to reveal the trend that the number of healthcare workers is decreasing in recent years.

A graph of a number of people with numbers

Description automatically generated with medium confidence

Next, using the research sources about ED visits related to specific conditions, the number of ED visits that would normally occur in Leavenworth, the increase in ED visits due to wildfire smoke, and the economic cost of an ED visit, the increase in both the number of ED visits and the economic costs attributable to the increased number of ED visits due wildfire smoke was found. Naturally, the visualization conclude that a small increase in costs from ED visits can be attributable to wildfire smoke.

A graph with a line and a line

Description automatically generatedA graph of a graph with numbers and a line

Description automatically generated with medium confidence

After that, another visualization was created to show the percentage of population with no health insurance in Leavenworth, Kansas. Specifically, the finding was that from 2018-2021 the number of people without insurance was increasing.

A line graph with numbers and text

Description automatically generated

Finally, using projected population and health care worker count data, the final finding shows that the ratio of health care, social, and educational workers to the general population is declining in Leavenworth.

A graph of health care workers

Description automatically generated

**Discussion/Implications:**

Asd

**Limitations/assumptions:**

Asd

**Conclusion:**

Asd

**References:**

1. <http://virtual.vtt.fi/virtual/innofirewood/stateoftheart/database/burning/burning.html> – smoke produced by wood
2. <https://www.cdc.gov/asthma/most_recent_national_asthma_data.htm> - why asthma
3. <https://www.cdc.gov/air/wildfire-smoke/default.htm> - why cvd, copd
4. <https://www.epa.gov/wildfire-smoke-course/health-effects-attributed-wildfire-smoke> - why MI, stroke
5. <https://www.cdc.gov/asthma/most_recent_national_asthma_data.htm> - 7.7% asthma
6. <https://professional.heart.org/-/media/PHD-Files-2/Science-News/2/2022-Heart-and-Stroke-Stat-Update/2022-Stat-Update-factsheet-GIobal-Burden-of-Disease.pdf> - 7.5341% CVD
7. <https://www.cdc.gov/copd/basics-about.html> - 6.4% COPD
8. <https://www.cdc.gov/heartdisease/facts.htm> - 0.241% heart attack
9. <https://www.census.gov/library/stories/2022/12/happy-new-year-2023.html> - 2023 pop
10. <https://www.cdc.gov/stroke/facts.htm> - 0.237 stroke
11. <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC8862306/> - 1.32% ED from asthma
12. <https://www.cdc.gov/nchs/data/nhsr/nhsr174.pdf> 11.4% ed from CVD
13. <https://journal.copdfoundation.org/jcopdf/id/1103/Characteristics-of-COPD-Patients-Using-United-States-Emergency-Care-or-Hospitalization> - 12.44% all ed from copd
14. <https://hcup-us.ahrq.gov/reports/statbriefs/sb47.pdf> - 0.6% ED from MI and stroke
15. <https://www.cdc.gov/nchs/products/databriefs/db452.htm/> 40% ED visits
16. <https://www.healio.com/news/pulmonology/20231107/wildfire-pollution-linked-to-cardiopulmonary-ed-visits-among-vulnerable-populations> - 2.2% increased ED asthma, 1.39 copd
17. <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC6015400/> - 5% increase ED CVD, 11% stroke
18. <https://newsroom.heart.org/news/wildfires-may-fuel-heart-health-hazards-smoke-exposure-increases-cardiovascular-risks> - 42% MI ED increase
19. <https://hcup-us.ahrq.gov/reports/statbriefs/sb268-ED-Costs-2017.jsp> 530 per ED visit
20. <https://www.ibm.com/topics/linear-regression> - linear regression
21. <https://www.investopedia.com/terms/a/autoregressive-integrated-moving-average-arima.asp> arima
22. <https://en.wikipedia.org/wiki/EPSG_Geodetic_Parameter_Dataset> - EPSG
23. [Air Quality Index - Washington State Department of Ecology](https://ecology.wa.gov/Research-Data/Monitoring-assessment/Air-Quality-Index) – only PM related to AQI
24. <https://www.oecd-ilibrary.org/social-issues-migration-health/old-age-dependency-ratio/indicator/english_e0255c98-en> old age dep ratio

**Data sources:**

Combined wildland fire datasets:

[Combined wildland fire datasets for the United States and certain territories, 1800s-Present (combined wildland fire polygons) - ScienceBase-Catalog](https://www.sciencebase.gov/catalog/item/61aa537dd34eb622f699df81)

AQS API:

<https://aqs.epa.gov/aqsweb/documents/data_api.html>

US Census Data:

<https://data.census.gov>

<https://data.census.gov/table/ACSST5Y2021.S2407?q=Leavenworth%20city,%20Kansas&t=Class%20of%20Worker>

<https://data.census.gov/table/ACSSE2022.K201803?q=Leavenworth%20city,%20Kansas&t=Disability> -

<https://data.census.gov/table/ACSSE2022.K201801?q=Leavenworth%20city,%20Kansas&t=Health> -

<https://data.census.gov/table/ACSST5Y2021.S0101?q=Leavenworth%20city,%20Kansas&t=Age%20and%20Sex>

<https://data.census.gov/table/ACSST5Y2022.S2701?q=Leavenworth%20city,%20Kansas&t=Health%20Insurance>