## A.I. Visualizations for the CDC

The Center for Disease Control and Prevention (CDC) is a federal public health and safety agency of the United States. Among other things, the CDC researches infectious diseases (like COVID-19) and collects public health data. For this assignment, the CDC has contracted *you* to help them design an interactive web feature for the public. The CDC will let the public explore their death dataset, which records cause-of-death of everyone deceased in the United States. The CDC's goal is to raise public awareness of death produced by natural causes or accidents (so the dataset will only include deaths by natural causes or accidents), to help current residents of the country take better precautions about their own health and limit dangerous activities.

Download the data from Github:

https://github.com/saiphcita/Human\_Al\_Interaction/tree/main/Assignment3

## Part 1: Explore & Visualize [50 points]

[Saiph's hint: If you've never visualized in Python, Pandas has some visualization support. <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/visualization.html">https://pandas.pydata.org/pandas-docs/stable/user\_guide/visualization.html</a> You can also use Matplotlib <a href="https://matplotlib.org/gallery/index.html">https://matplotlib.org/gallery/index.html</a>]

The goal of these tasks are to have you experiment with different ways of visualizing the data.

Show your work in code as well as your final visualizations in a notebook. Label each question using markdown in the notebook and include answers to all questions.

**Submit** part 1 as a notebook hw3-part1.ipynb

- A. Create a histogram of **death counts by age. [5 points]**Hint: one axis should be sorted continuous range of age 0 up to the age of the oldest person in the dataset
- B. Create a histogram of deaths caused by **Diabetes by age**. [8 points]

When shown data or an ML model, humans tend to have *confirmation bias*, meaning that they tend to believe that whatever the data or model says *is what they really thought all along*. Ever broke up with a significant other and your friends tell you "I told you so"? This is confirmation bias. With Bayesian reasoning, we can take into account a viewer's *prior reasonable guess* before they see data. This is a good technique to help users reflect on how the data might *conflict* with "what they thought all along".

- C. Write down what you believe (before looking at the data. Just guess!) is the relationship between age and death by heart disease. Do you expect the risk of death by heart disease to be the same across all ages or higher in certain age ranges? Why?

  [2 points]
- D. Write down what you believe (<u>before looking at the data. Just guess!</u>) is the relationship between **age** and death by **Natural Causes**. Do you expect the risk of death by natural causes to be the same across all ages or higher in certain age ranges? Why?

  [2 points]
- E. Create a histogram of **death** by **Natural Causes** by **age**. Create a histogram of **death** by **heart disease** by **age**. [5 points]
- F. Compare your prior guess in C and D to the histograms in E. What did you learn from the histograms? Are there parts of your prior guess that were confirmed by the histograms? Are there parts of your prior guess that were wrong or different than you expected?

  [8 points]

When users see different possibilities separately in a data or Machine Learning systems, there's a bias towards thinking *all possibilities are equally likely*, when really some options are more or less probable in real life. E.g., While a headache could be caused by fall allergies or by brain cancer, the likelihood of fall allergies is far higher in real life than brain cancer.

G. Create a visualization of your choice, where you overlay 4 different causes of death (your pick) by age the same plot. Design this visualization however you wish. Justify your design by writing a few sentences about how your visualization will help users compare the 4

different death risks by age. Talk about encoding choices such as: plot type, use of size, color, and axes labels. [20 points]

## Part 2: Designing Personal Predictions [50 points]

The goal of Part 2 is to start designing an interactive interface, where a user that comes to the CDC visualization can put in their own information, and see the most common causes of death for their attributes (like age, gender, and so on)

Show your work in code as well as your final visualizations in a notebook. Label each question using markdown in the notebook and include answers to all questions.

Hint: To add some minimal interactivity with minimal effort, consider using Jupyter Notebook Widgets: https://ipywidgets.readthedocs.io/en/latest/examples/Widget%20List.html

Hint 2: It's actually rather easy to make it fully interactive. see <a href="https://towardsdatascience.com/interactive-controls-for-jupyter-notebooks-f5c94829aee6">https://towardsdatascience.com/interactive-controls-for-jupyter-notebooks-f5c94829aee6</a>

also, https://ipywidgets.readthedocs.io/en/latest/examples/Using%20Interact.html

## **Submit** part 2 as a notebook hw3-part2.ipynb

- A. Design for personas. For each of the fictional users given, create a single visualization that shows the most likely cause of death for that user. To experiment with design choices, make each user/visualization pair a different visualization that represents different design choices (e.g. you could try a different plot type for some users). [20 points]
  - **Jenny** is a young black female college student. She is 20 and lives a healthy lifestyle. He doesn't smoke or use drugs, but does drink alcohol in social settings.
  - **Marco** is a 72 year old man, immigrated from Korea to the United States in his thirties. His wife passed away from cancer last year.
  - **Eva** is a 36 year old woman with two kids. She is half Cuban and co-parents her kids with a long term romantic partner but does not believe in marriage. She has a strong family history of diabetes.
  - **Elmer** is a 40 year old Boston native, whose family is from Ireland. He was diagnosed with a chronic respiratory disease and has felt depression.

What visualization techniques did you use? [3 points]

B. Given your visualizations in A, what would be *good* questions for a user to ask a personalized visualization from this dataset? What would be some *bad* questions ie. questions that a personalized visualization (with this dataset alone) cannot answer? [7 points]

- C. If users like those in A visit the interactive tool on the CDC website, what information (e.g. age or race) would you have them put in to show the most relevant death visualization and why? [10 points]
- D. For each column in the dataset, describe how you would bin the data for a good user experience and why. For instance, 'age' can be not-binned (exact numbers) or binned into 'child', 'young adult', 'adult', 'elder'... or many other bin choices. [10 points]

**Note**: Some of you have asked if you can submit this as an HTML visualization/interactive page. The answer is a qualified yes. If you do that, please upload a zip file with all your code -- e.g. if you use a Flask backend, please include the Flask python file along with the HTML. Also, use something like `pip freeze` to make a list of all your dependencies.

We will run your server locally. We *must* be able to inspect all your code.