

# Data Exploration Project

# FIT5147 Data Exploration and Visualization Semester 1 2023

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# Introduction

## Problem Description

The global refugee crisis has highlighted the need for countries to provide asylum to refugees, not only for their protection but also to offer them opportunities for employment and resettlement. However, it is important to note that the burden of accepting refugees should not fall solely on one country, as this could lead to an overburdening of resources and negatively impact the well-being of both refugees and host communities.

To gain a better understanding of the refugee situation, this project aims to visualize the flow of refugees in a country and compare it with job demand. By collecting data on the number of refugees entering the country and the availability of job opportunities in different regions or industries, we can analyze the distribution of refugees and the potential economic impact of their resettlement.

#### Questions

- 1. What percentage of the refugee population is employed, and how does this vary by country of origin, age, gender, and level of education?
- 2. What percentage of employment demand from the asylum countries can refugees fulfill?

The first question seeks to understand the employment status of refugees and how it varies based on different factors such as country of origin, age, gender, and education level. This question is important because employment is crucial for refugees to rebuild their lives and become self-sufficient in their host country. It also helps in assessing the economic contribution of refugees and their potential impact on the labor market. By analyzing the percentage of employed refugees, we can identify any gaps in employment opportunities and potential barriers to employment that refugees may face based on their background.

The second question aims to determine the percentage of job demand from asylum countries that refugees can fulfill. This question is essential to assess the potential economic impact of refugee resettlement on the host country. If refugees can fulfill a significant portion of employment demand, it can positively impact the labor market, the overall economy, and reduce any potential burden on the host country's social welfare system. It can also help identify any skill gaps among the refugee population and guide targeted training programs to enhance their employability.

#### Motivation

Gaining insights into the employment status of refugees can help dispel common misconceptions about refugees being a burden on the host country's economy. Instead, evidence-based research can highlight the significant economic contributions that refugees can make to the host country's labor market and economy.

By answering the two questions, we can also promote more inclusive and sustainable refugee resettlement practices, which prioritize the economic and social well-being of both refugees and host communities. Therefore, this research is crucial for policymakers, researchers, and advocates who seek to support refugees' successful integration into their host communities and facilitate their long-term economic self-sufficiency.

# Data Wrangling and Data Checking

## Description of Data Sources

There are three main datasets used along with some supplementary datasets.

The main dataset is the Annual Survey of Refugees 2019 (ASR 2019) dataset which is got UNHCR microdata unit, Central Data Catalog. Though the dataset is named to be of the year 2019 but it is updated every year. The last modification was made on Jun 21, 2022. Thus, the dataset is categorized as a reliable dataset. This dataset has almost 5000 cases and more than 250 variables, but the main objective of this project is to work with unemployment of the refugees. So for this I have selected the variables that are relevant to unemployment. A total of 12 variables are selected, these variables are a code but here the description is given as well, they are listed as follows:

- qn1c is the person married or not?
- qn1d what is the person's age at the last birthday?
- qn1f Male or female?
- qn1h Citizenship of which country?
- qn1jyear what month and year did this person enter in US to stay?
- qn2b what is the highest degree or certificate that this person obtained?
- qn3b what kind of work did this person perform before coming to US?
- qn4b how well does this person speak English?
- qn4c after coming to us did this person have any English instruction?
- qn4e within the past 12 months, has the person attended an English language training?
- ui\_lpr legal permanent residency status
- ui\_work work status

When the dataset is loaded in tableau it looks like this.

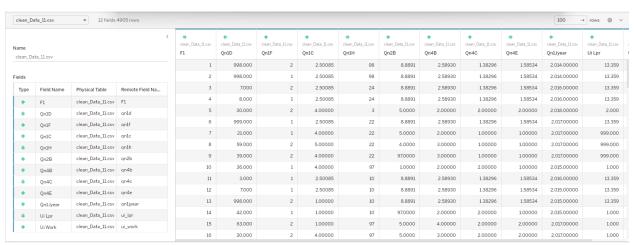


Figure 1

The second Dataset is from the US Bureau of Labor Statistics which contains the labor force of US region from the year 2000 to 2023. The data can be collected by year, quarterly or monthly. As for the thoroughness of the project the monthly data of some variables are collected.

When the data is loaded it looks like this.



Figure 2

A third Dataset is used which gives information of Asylum countries with USA it gives information of Germany, Canada, France and USA.

When the dataset is loaded in the it looks like this

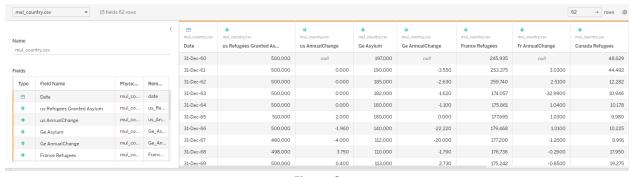


Figure 3

#### Data Cleaning

The ASR dataset has a lot of has more than 250 variables, so after going through all the variables the 12 variables are selected. From those 12 variables a new dataset that is named Clean Dataset is taken. The cleaning process includes thoroughly checking all the variables from the data description to select the important variables of the dataset. The dataset had a lot of missing values and the arbitrary values. The missing values are handled by checking the distribution of the values and filling it with required values. If the missing values are more than 40% of the data, then the column is omitted and if the missing values are less than 40% then it is filled with the majority values and if the missing values are less than 20% then it is filled with the average value. The outliers are omitted as there are values which are way beyond the range of the values. The cleaned data is then imported in Tableau for the visualizations as shown above. The tool used for this Dataset is Python for cleaning.

	Variable	hhid	personid	respondent	cohort	qn1a	numppl	qn1b	qn1c	qn1d	•••	qn34b_05	ui_qn8a_annual	ui_qn10a_annual	ui_cashassist	ui_lfp	ui_emprate	ui,
0	NaN	2	22	0	3	2	5	11	NaN	998.0		NaN	NaN	NaN	2	NaN	NaN	
1	NaN	2	23	0	3	3	5	11	NaN	998.0		NaN	NaN	NaN	2	NaN	NaN	
2	NaN	2	25	0	3	5	5	11	NaN	7.0		NaN	NaN	NaN	2	NaN	NaN	
3	NaN	2	24	0	3	4	5	11	NaN	8.0		NaN	NaN	NaN	2	NaN	NaN	
4	NaN	2	21	1	3	1	5	1	4.0	30.0		NaN	NaN	NaN	2	1.0	1.0	

Figure 4

The second Dataset is the US Bureau of Labor Statistics dataset and this dataset had only few missing values which of the year 2023 and the year 2023 is not included as it had missing values. As the data is for public display and can only be edited by the US Bureau of Labor Statistics and it is being updated so the data is assumed to be correct. Python is also used for cleaning this dataset.

The third dataset was also clean, and it had data from the year 1960 to 2023. It just had a few missing values which are filled with average values of that column because the number of missing values were very less in percentage. Python is used for cleaning the dataset.

# **Data Exploration**

#### **ASR Dataset**

As the main dataset, the ASR dataset is used so the cleaned dataset is imported in the Tableau and some test are carried out. As most of the variables has a range of numbers the arbitrary values are filtered out for a good visualization.

This Shows the total participants of the ASR Dataset which are divided in male and female but note that some participants (2%) actually refused to give away their gender and 1% of participants were confused with their gender. Except this the dataset is evenly distributed between male and female as 50% is female and 49.7% is male in this dataset.

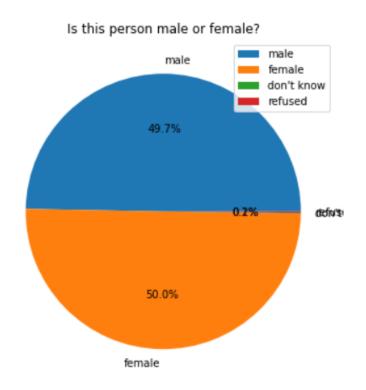
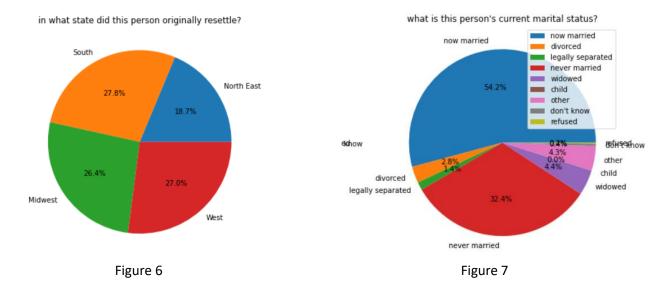


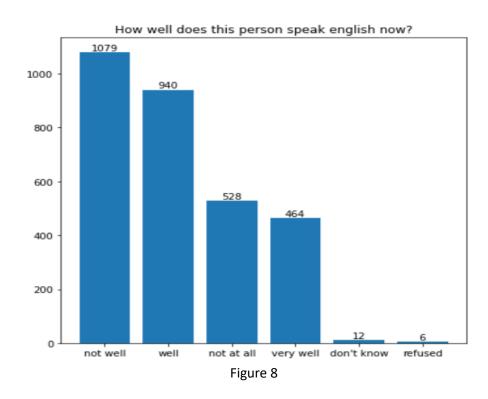
Figure 5



From the above diagram on the left, It can be seen that most refugees settled in the in South, Midwest and West region of US but on Northeast we can not say that they didn't settle there but North East has a bit less settlement compared to the other region of US.

The second diagram on the right shows us that more than half of the participants were recently married, and their partner does not actually stay with them and 32.4% were single. The next largest portion is the widowed portion which comes to 4.4%, very few (0.2%) actually refused to say anything about their marital status and 0.4% did not know about their status while they took shelter.

This graph shows how they spoke English while entering the country. The amount of well spoken participants are only 23% of the population and very well were only 11% of populations, 14% did not even know English while being granted in the asylum country.



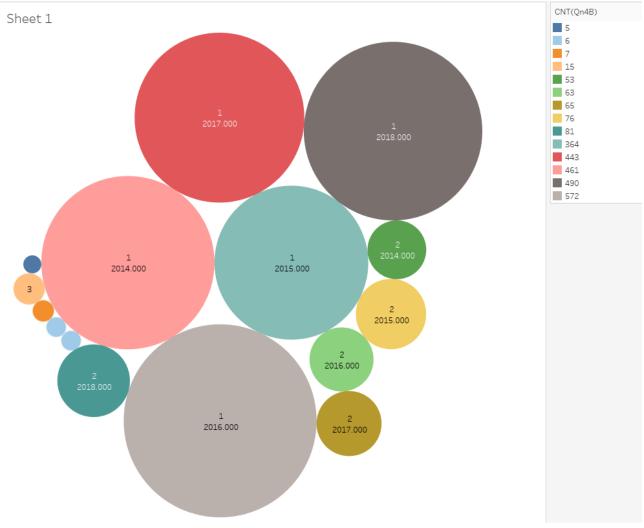


Figure 9

The above visualizations show how many entered in which year with their pr status when the survey was being conducted. 1 denotes they have already adjusted the permanent residency (PR), 2 denotes they are planning to adjust PR status in future, 3 denotes they have not applied to adjust PR yet or may not. The size denotes the number of participants, and the year is displayed in which they were granted asylum.



Figure 10

This image shows the Age (Qn1D), Education Level (Qn2B), Gender (Qn1f), Origin (Qn1H) and PR Status (Ui Lpr) against Work Status (Ui Work). For Work Status working now = 1, not working now but previously worked in us = 2, not working now and never worked in us = 3, not working now and unsure about previous work in = 4. For Gender 1 is Male and 2 is Female. Qn1d denotes the age of the refugee and for Qn2B 1 = none, 2 = primary, 3 = training in refugee camp, 4 = technical school, 5 = secondary, 6 = university, 7 = medical degree. The origin has 10 different war zone countries.

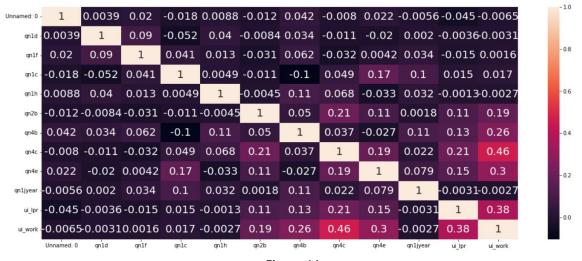


Figure 11

The correlation matrix shows that none of the variables are strongly related to any of the other variables, as 1 means very strong positive correlation and -1 means strong negative correlation with 0 being no relation.

#### US Bureau of Labor Statistics Dataset

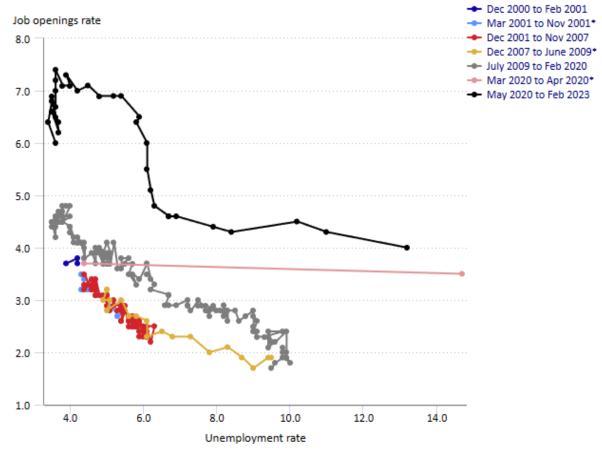
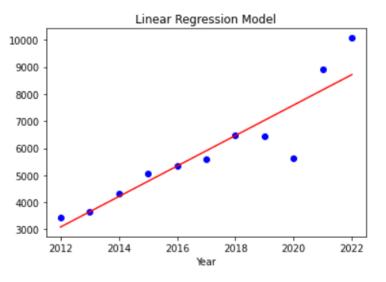


Figure 12

This graph displays the Job Opening Rate vs the Unemployment Rate. It can be observed that the most unemployment rate was from March 2020 to April 2020 with a bit decreased rate for new jobs, which was the start of the pandemic. And only in 2021 the rate was that high unemployment and low job opening rate is seen.



The Linear regression with number of jobs opening against year can be seen with uniform increment but only on the year 2020 the number decreased.

This is Figure 13

# Asylum Countries Dataset

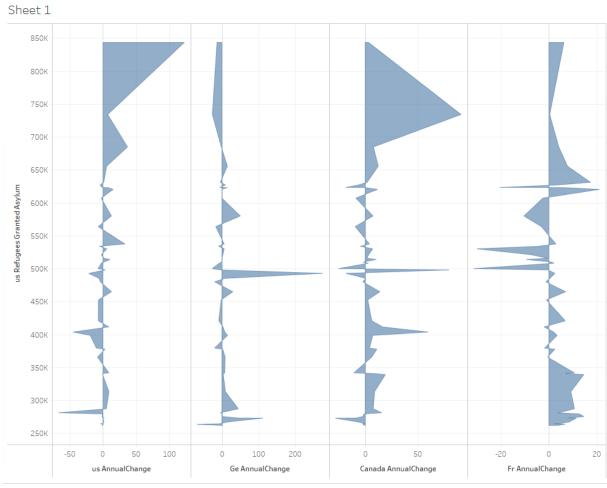


Figure 14

In this Graph the number of US Refugee granted against the US Annual Change is plotted and compared to the Annual Change of Germany Canada and France. We can see that on the high number Canada and US are the ones that were taking in refugees with a mass number. France and Germany we only interested in giving asylum to comparatively less refugees.

# Conclusion

The main two questions were:

- 1. What percentage of the refugee population is employed, and how does this vary by country of origin, age, gender, and level of education?
- 2. What percentage of employment demand from the asylum countries can refugees fulfill?

To find the answer of the first question we can conclude that almost closed to 50% of the refugees are working now as from the Figure 10 we can see that the Ui Work is mostly on level 1 which means they

are currently working while the other 50% are not working now, or never worked and not sure if they will work or not. It can be observed that war prone countries are contributing most of the refugees and almost 60% of the working refugees are of age between 20 to 40, The refugees that are not working are not that educated and are mostly women and elderly people, though it includes some you and male people also. It can also be observed that only women are unsure of they want to work or not. Most of the working refugees, almost 50% are of either primary, secondary or university graduate and they are male. Though the percentage of employed refugees and the unemployed refugees is almost the same, it heavily depends on education and age.

For answer of the second question, a conclusion can not be made as more data is required for that analysis. We could only see that refugees granted asylum and Annual Change of US, the correct number of total refugees and the total number of refugees that are working now is yet to be determined. The job opening rate and the unemployment rate is discovered but that does not say anything about refugees, so those graphs are for refugees as well as citizens. A refugee employment dataset is needed for the analysis of the second question.

#### Reflection

What I learned about the project is that the datasets that I chose do not actually give enough information to come to a rigid conclusion for either of the questions, but it gives insights on the first question only. I need to find out more datasets and combine some of the datasets from the US Bureau of Labor Statistics together to find out the answer to the second question. The website has a lot of different datasets, and those datasets need to be combined to a common pivot point to come to any conclusion.

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