BIOS 611 Final Report

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Introduction

In the United States, there are vast racial inequities in outcomes for missing people. Black and Indigenous people, especially women, are more likely to go missing, but they are less likely to receive media attention, spend more time lost, and tend to have worse outcomes when compared to white people. The NamUs (National Missing and Unidentified Persons System) database is a comprehensive, publicly accessible repository that centralizes information on missing persons and unidentified decedents in the United States. It serves as a critical resource for law enforcement, families, and the public, facilitating the collaboration and exchange of information to aid in the resolution of missing persons cases. This project models the relationship between race/ethnicity and time spent in the database to identify inequities in the length of time individuals remain missing.

#Variables

Variable Name	Description
Case.Number	Case number assigned to the missing person
DLC	Date last seen
Last.Name	Last name of the missing person
First.Name	First name of the missing person
Missing.Age	Age at the time of disappearance
City	City where the person was last seen
County	County where the person was last seen
State	State where the person was last seen
Sex	Gender of the missing person
RaceEthnicity	Racial or ethnic background of the missing person
Date.Modified	Date when the information was last modified

Exploratory Analysis

When looking at distribution of entries by race/ethnicity, we see that the vase majority of entries in the database are White / Caucasian. This is roughly proportional the racial distribution of the larger population.

Additionally, we see slightly more females in the database than males.

We see a that the states with the most entries are California, Texas, and Florida. This is also somewhat proportional to the raw population size of each state.

We also see that entries in the database decline as age increases.

Finally, we see that there are no significant patterns when we look at the amount of time in the database by race/ethnicity. However, we do see that entries with no race/ethnicity data at all tend to spend a longer amount of time in the database.

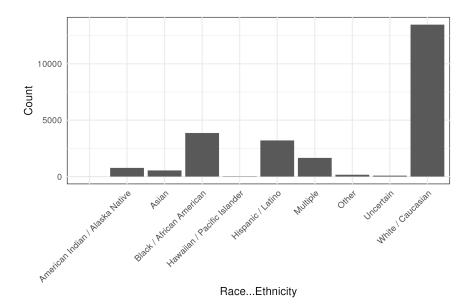


Figure 1: Distribution of entries by race/ethnicity

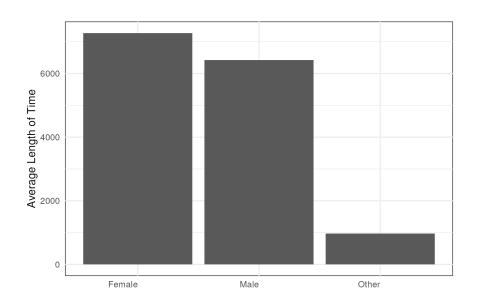


Figure 2: Distribution of entries by sex

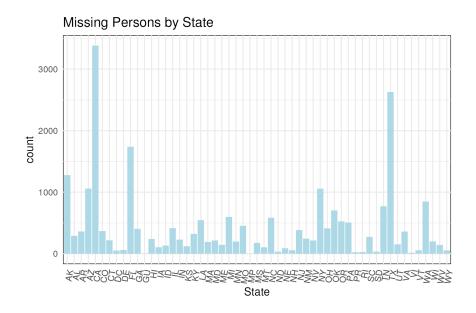


Figure 3: Distribution of entries by state

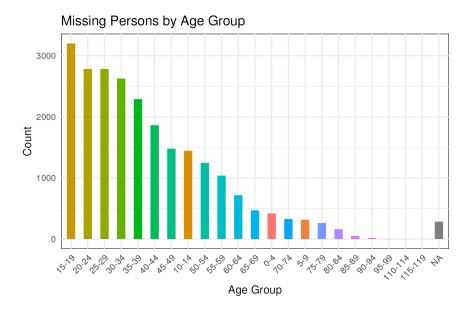


Figure 4: Distribution of entries by age

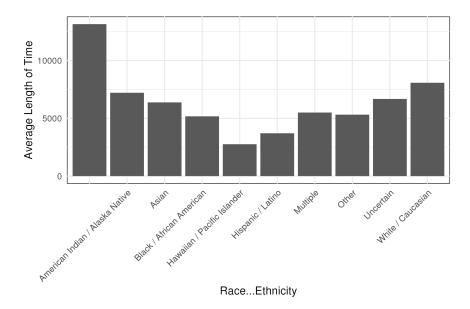


Figure 5: Distribution of time in database by race/ethnicity

Model

I conducted linear regression of Race/Ethnicity with total time in the database. Despite detecting significant differences in multiple racial groups when compared to white people, the model was a generally poor fit, as indicated by the modest multiple R-squared value of 0.09017, implying that only approximately 9% of the variability in the response variable is explained by the predictor variables. The residual standard error of 5323 suggests a considerable amount of unexplained variability in the model's predictions. While the model may capture some significant relationships, its limited explanatory power and high residual error suggest that it may not be a robust fit for this data.

Term	Estimate	Std . Error	Statistic	P.Value
(Intercept)	8077.	45.9	176.	0
RaceEthnicity	5075.	3764.	1.35	1.78e-1
RaceEthnicityAmerican Indian / Alaska Native	-866.	197.	-4.39	1.11e-5
RaceEthnicityAsian	-1701.	232.	-7.32	2.58e-13
RaceEthnicityBlack / African American	-2905.	97.1	-29.9	5.29e-193
RaceEthnicityHawaiian / Pacific Islander	-5321.	990.	-5.38	7.64e-8
RaceEthnicityHispanic / Latino	-4373.	105.	-41.8	0
RaceEthnicityMultiple	-2578.	139.	-18.6	1.02e-76
RaceEthnicityOther	-2761.	418.	-6.60	4.13e-11
RaceEthnicityUncertain	-1402.	586.	-2.39	1.68e-2

I also conducted linear regression of Sex with total time in the database to a similar outcome. Despite detecting significantly less time spent in the database among males when compared to females, the model was a generally poor fit, as indicated by the modest multiple R-squared value of 0.005638, implying that only a small amount of the variability in the response variable is explained by the predictor variables. The residual standard error similarly suggested a considerable amount of unexplained variability in the model's predictions. While the model may capture a significant relationship, its is probably not a great fit for this data.

Term	Estimate	Std.Error	Statistic	P.Value
(Intercept) SexMale SexOther	7268.	58.8	124.	0
	-847.	74.4	-11.4	6.31e-30
	-6302.	2489.	-2.53	1.14e-2

Discussion

While this database is a wealth of information for law enforcement and the families of missing persons and decedents, it is not particularly useful for drawing meaningful statistical inferences. This is because states are not required to use the database and there are no outcomes, meaning that when a person is found, they are removed from the database. In fact, one cannot even FOIA for this information. Therefore, it makes investigations into inequities of outcomes very difficult, if not impossible. This database is not meant for research purposes and drawing conclusions about anything is very difficult.

Future Questions

Additional data will be required to examine inequities in the amount of time that missing persons remain missing. At minimum, a dataset with the outcomes for missing persons and decedents and a well characterized sample strategy (i.e., who is required to report to the database) will be required to make draw important conclusions.