Gender Distributions of Collection Authors in the Rubenstein Library's Main Entry Card Catalog

These demographics were computed using the Python Gender Guesser package. As part of our exploratory data analysis, we hope to gain a better understanding of *who* is represented in the catalog. We recognize that gender is not considered as strictly binary in current times, but hope to use this analysis to evaluate the identities of the people that past librarians deemed important enough to catalog the work of in a time when gender *was* treated as more black and white than it is now. The intention of this analysis is to explore the extent of gender discrepancies in the card catalog to further study the history of the library's treatment of minority groups.

The authors of collections in the file are typically either a person or an organization. We are evaluating the gender typically associated with only the *people*. The Gender Guesser package classifies genders as one of 6 groups. Male and female result from names that are traditionally associated with one of those genders. Mostly male and mostly female result from names that are less cut and dry in regards to the gender they are associated with. "Andy," or androgynous, means that a name is not traditionally strongly associated with either gender and unknown means that the package was unable to classify a name into any of the other categories. These tended to be non-person organizations or places that would not have a gender and thus were dropped for the visualizations of the results.

We also used the numpy, pandas, and matplotlib packages, which provided us with the tools to manipulate and graph the data easily.

In []:

!pip install gender-guesser

```
In [1]:
        import gender guesser.detector as gender
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        %matplotlib inline
In [2]:
         # Collect first names from dataset
        df = pd.read csv("catalog/csv iterations/all sorted collection.csv")
         # Holds all first names
        first names = []
         # Holds indices of names for use in wordclouds
        indices = []
        for index, row in df.iterrows():
            if row['Coll head'] == 1:
                # if name contains a comma, it's likely a name and not an organization
                if ", " in str(row['Name']):
                    name = str(row['Name'].split(" ")[1])
                     # Find first name after Sir or Mrs.
                    if ("Sir" in name or "Mrs" in name or "Dr." in name or "Capt" in name or "Miss" in name or "Lord"
                        name = str(row['Name'].split(" ")[2])
                     # For initials, try to find whole name
                    if len(name) < 2 and len(row['Name'].split(" ")) > 2 and len(str(row['Name'].split(" ")[2])) > 2:
                        name = str(row['Name'].split(" ")[2])
                     if not "Archive" in name and not "Army" in name and not "University" in name and not "Family" in name
                        name = name.strip(" \r\n,().")
                         # Add valid names to list
                        if not "." in name and len(name) > 3:
                            first names.append(name)
                             indices.append(index)
In [3]:
         # Detect gender for each main entry author
        d = gender.Detector()
        gender_counts = {
            "unknown": 0,
            "andy": 0,
            "male": 0,
            "female": 0,
            "mostly_male": 0,
            "mostly_female": 0
        female_indices = []
        male indices = []
         # Determine gender of each name
        for i in range(len(first names)):
            name = first_names[i]
            gen = d.get_gender(name)
            if gen == "unknown":
                 # Common unclassified name
                if name == "Duff":
                    gen = "male"
                 # Check for trailing e's (common in names from OCR error) and recheck those names
                elif "e" == name[len(name)-1]:
                    name = name[0:len(name)-1]
                    gen = d.get_gender(name)
            if "female" in gen:
                female_indices.append(i)
```

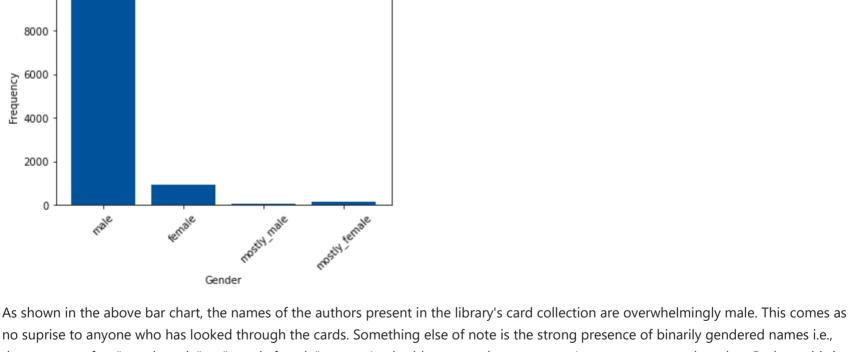
{'unknown': 978, 'andy': 18, 'male': 9522, 'female': 932, 'mostly_male': 67, 'mostly_female': 149}
Most of the 'unknown' gendered names are actually part of non-person organizations and will be dropped for easier visualization, for

example "Britain" is part of the "Name" column, but is a title instead of a person, and thus will not be counted for the distributions.

```
In [4]: # Remove unknown and andy to create bar chart
    known = gender_counts.copy()
    known.pop('unknown')
    known.pop('andy')

Out[4]: 18

In [7]: # Display bar chart of gender frequencies
    plt.bar(*zip(*known.items()), color='#00539B')
    plt.xticks(rotation = 45)
    plt.title("Gender Frequency of Collection Authors in Card Catalog")
    plt.xlabel("Gender")
```



Gender Frequency of Collection Authors in Card Catalog

elif "male" in gen:

print(gender_counts)

plt.ylabel("Frequency")

plt.show()

In [13]:

gender_counts[gen] += 1

male_indices.append(i)

there are very few "mostly male" or "mostly female" names (and seldom an androgynous one), most are one or the other. Perhaps this is indicative of the kinds of names that were given during the time period represented in the cards.

Set up bar chart comparing male-leaning to female-leaning names

The known conv()

```
mf = known.copy()
mf['male'] = mf.get('male') + mf.get('mostly_male')
mf.pop('mostly_male')
mf['female'] = mf.get('female') + mf.get('mostly_female')
mf.pop('mostly_female')

colors = ['#339898', '#E89923']

plt.pie(mf.values(), labels = mf.keys(), autopct='%1.1f%%', startangle = 90, colors=colors)
plt.axis('equal')
plt.title("Gender Ratio of Collection Authors")
plt.show()
Gender Ratio of Collection Authors
female
```

Gender Ratio of Collection Authors
female

10.1%

For the pie chart, we combined "mostly male" names with "male" names and "mostly female" names with female names to more easily visualize the gender frequencies. Androgynous names were dropped for the pie chart's sake, due to the fact they compose only about 0.02% of the names. The chart shows that about nine out of ten of the collection authors present in the main entry file were, in fact, male. This confirms our specualtions that men were more often than women represented in the catalog. In addition, this supports the theory that the "head of the household," likely the husband, would be elevated to the "author" of collections that entail multiple individuals, possibly hiding the presence of women in the collections. From this we can hypothesize other reasons for the discepency, but further research should be done into the history of archival records at Duke. Further research could also be done into the common events described in the manuscripts cataloged in the files (e.g., were there many Civil War male soldier accounts that took presidence over a wife's account of staying home and caring for her children?)