

# Exploratory Data Analysis & Data Preparation

This assignment is a chance to explore your data. Be creative with your plots!

## Contents

### Exploring the Data

Answer (at least) the following questions:

- What is your outcome variable(s)? How well does it measure the outcome you are interested in? How does it relate to your expectations?

Our outcome variables are whether or not a resume gets passed to the next round. This directly relates to the outcome we are interested in, which is to determine whether hiring algorithms are biased. This will provide a concrete measure of our expectations.

- What are your key explanatory variables?

Technical skills, years of experience, professional experience, number of past jobs, gender, masculine words, feminine words

In addition, create a table of summary statistics for the variables you are planning to use in your analysis.

You can explore any other questions you see fit for your project.

### Data Visualization

You must include at least 4 visualizations of your data made in Python. You must include your outcome variable in at least two plots and your key explanatory variable in at least two of these plots. You must use visualizations that are *appropriate* for the data type (categorical vs numeric, continuous vs discrete) of your outcome and explanatory variables. For example, you should not use a histogram to plot a categorical variable.

### Data Preparation and Cleaning

Answer the following question:

- What data cleaning did you have to do?

To clean the data, we separated the resume data into more specific column categories. First, we separated the data by "summary", "skill level" and "technical skills". Then, we started extracting certain details such as number of past jobs and number of years of experience.

- How did you prepare the data?

We scraped 600 resumes from [HireItPeople.com](https://HireItPeople.com) under the Java Developers/Architects

Resumes page. Hire IT People provides thousands of resumes within industries for employers to view. We created the resume.csv file by opening each link that holds a resume and then added the text to the csv. Once we had a csv containing the url link and the long resume text, we were able to begin separating the different variables into separate columns.

- Are you deciding to exclude any observations? If so, why?

We might remove some filler information, such as repetitive language and exclude resumes with the “professional summary” section (3 entries)

- Did you have to create any new variables from existing variables? If so, how and why?

We randomly assigned gender (male/female) to each entry. We also added columns for years of experience and number of previous jobs held, which were pulled from the resume descriptions. In addition, we added our own data for awards to relate to gender, specifically for females. We added awards that highlighted their role as a “woman in STEM”.

- How Did We Pick Gender-Coded Words?

We used [Gender Decoder](#) which was inspired by a research paper written by Danielle Gaucher, Justin Friesen, and Aaron C. Kay back in 2011. In this paper, *Evidence That Gendered Wording in Job Advertisements Exists and Sustains Gender Inequality*, “researchers showed job adverts which included different kinds of gender-coded language to men and women and recorded how appealing the jobs seemed and how much the participants felt that they ‘belonged’ in that occupation.”

- What STEM Awards did you use?

We used for female awards:

- Women Tech Award Recipient: *For women in computer science and related fields*
- Women of ENIAC Computer Pioneer Award Recipient: *recipients are individuals who have made significant contributions to the computer industry*
- Zonta Women in STEM Award Recipient: *Honors women between 18-35 for their contributions to STEM.*
- Women of Colour in STEM Award Recipient: *A new award that launched in 2024 to recognize the achievements of women of color in STEM.*
- Grace Murray Hopper Award: *For young computer professionals who made a significant technical or service contribution.*

For male awards:

- ACM Prize in Computing: *Honors early-to-mid career innovators in computing.*
- ACM Software System Award: *For influential software systems with lasting impact.*
- The Java Community Process Award Recipient: *awards for excellence in Java standards development, such as "Member/Participant of the Year" and "Outstanding Spec Lead"*

## Number of past jobs

## Years of experience

## Binary columns of skills

Add gender column based off of gender language decoder

<https://gender-decoder.katmatfield.com/about>

<https://adminvc.ucla.edu/equity/hiring-guide/gender-decoder>

## Example Entry Data

Each data entry comes from a URL like [this](#). It has a summary with past experiences, technical skills, and descriptions of past professional experience.

Below are two examples of a resume data point from our CSV file.

## Codebook

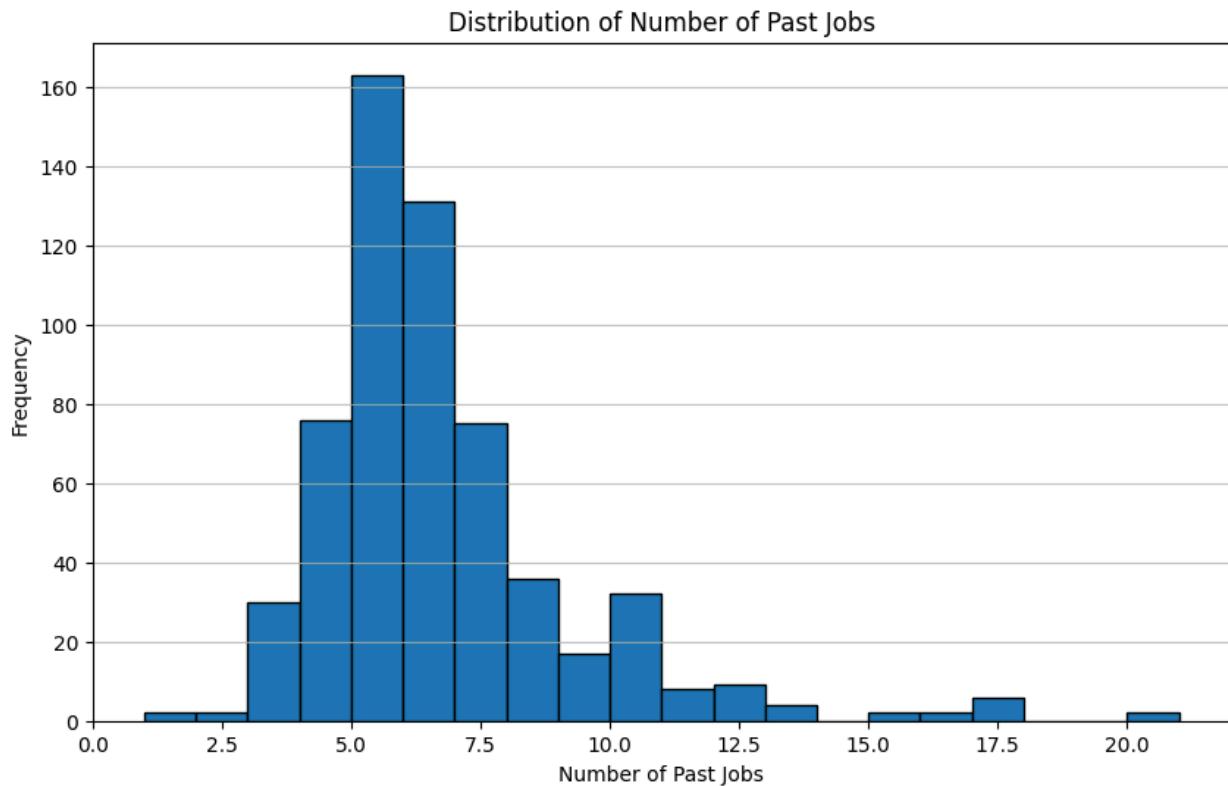
You must add a *codebook* – a description of all variables you are using, including ones you are creating for this project – to the README.md page of the data/ folder of your repo.

## (OPTIONAL) Data Analysis

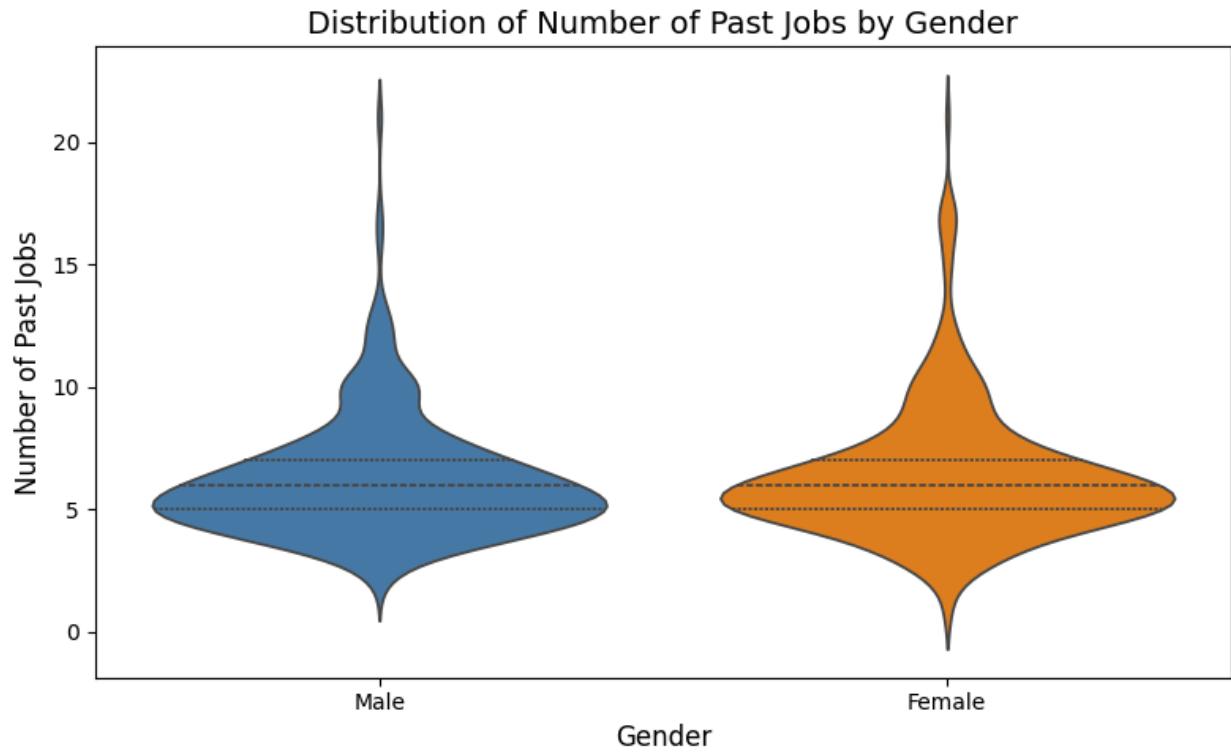
If you would like, you can start to sketch out some data analysis/modeling. This will not be counted for or against you, but I will give you feedback on it (which will be helpful for later).

### Graphs And Analysis:

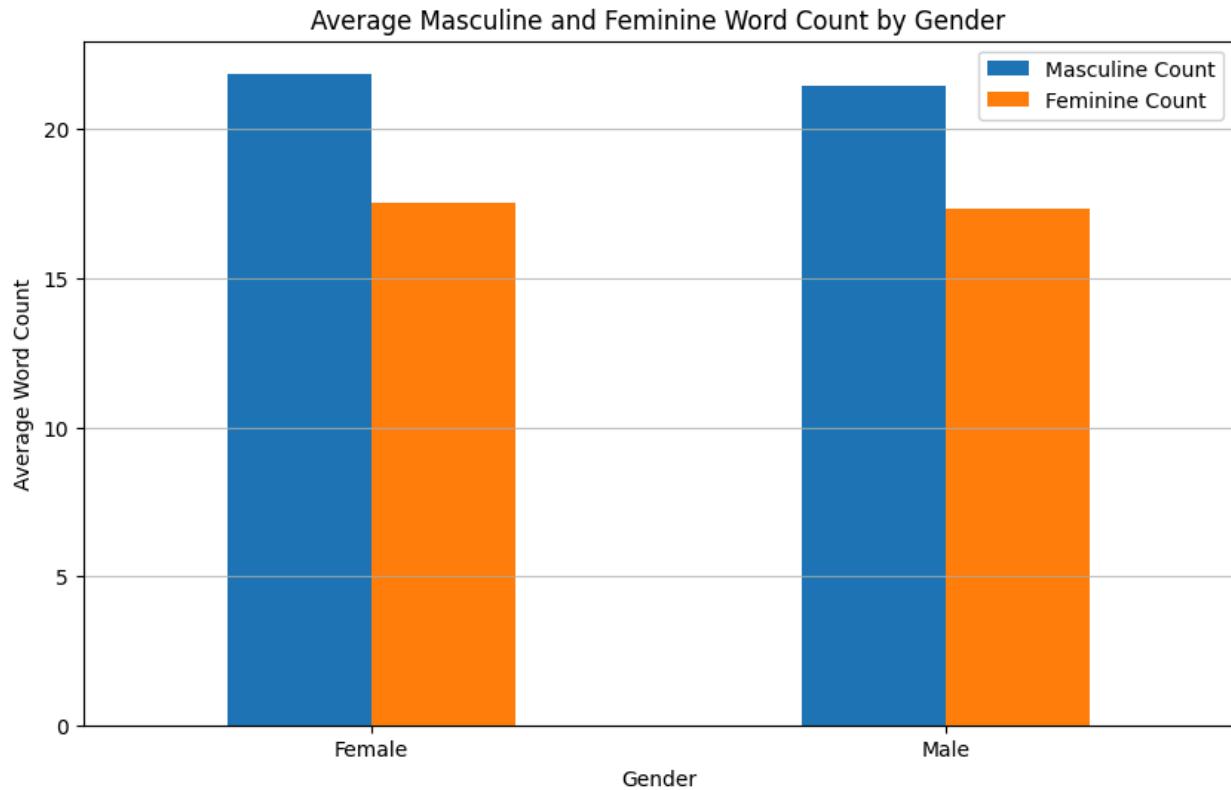
*Below are the graphs that we created to visualize and explore our data.*



This visualization showed the distribution of number of past jobs listed in resumes in the dataset. The distribution is right-skewed and clustered around 5.0 to 7.5 years of experience.



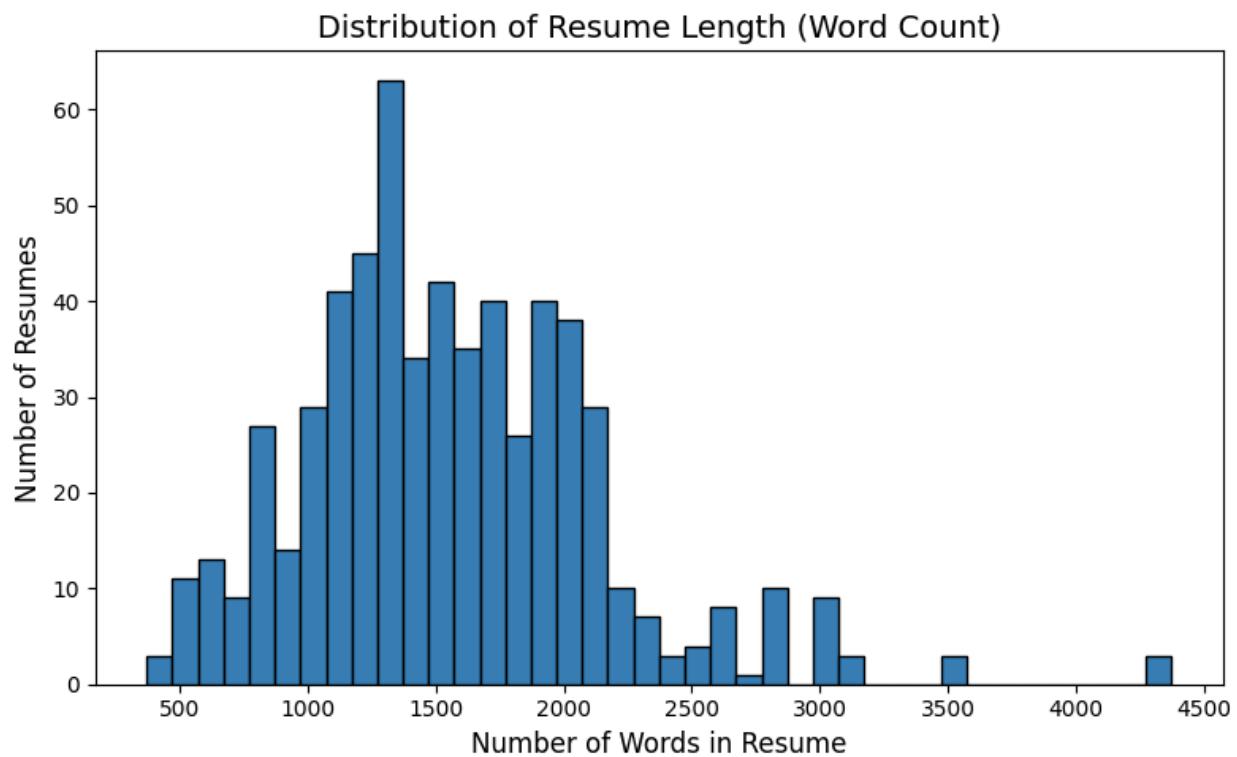
This visualization shows the distribution of the number of past jobs by gender. Our analysis shows that most applicants have held about 5 jobs, and the number of past jobs does not differ significantly by gender.



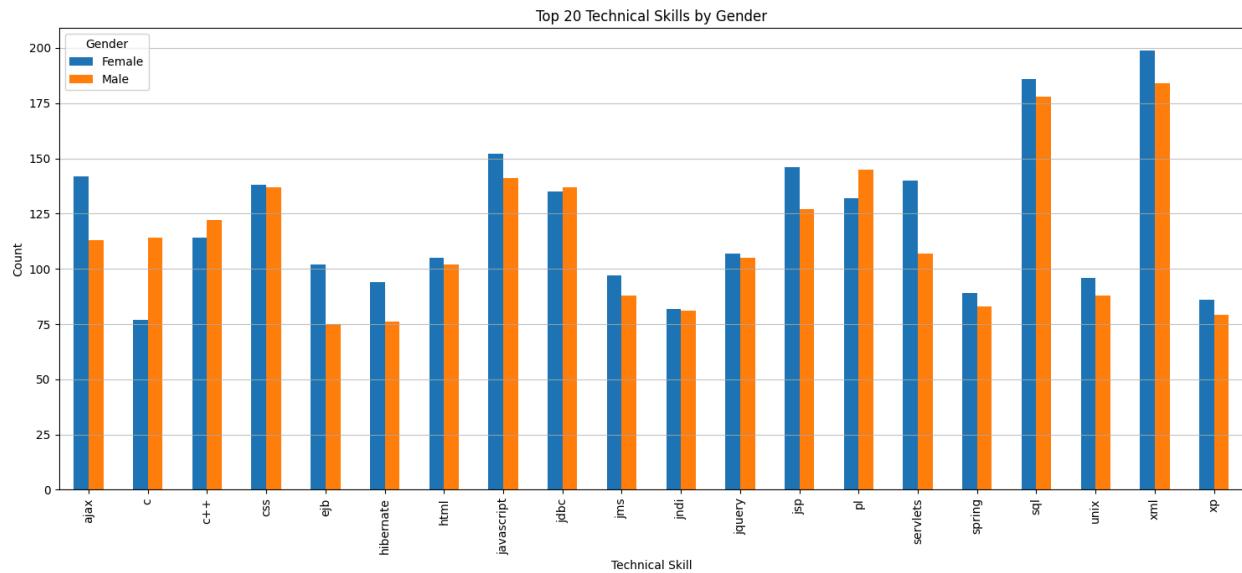
This analysis looked at the average number of masculine and feminine-coded words in resumes by gender. The results of this analysis show an equal number of gender coded words in both gender categories, with masculine-coded words being more prevalent than feminine coded words.



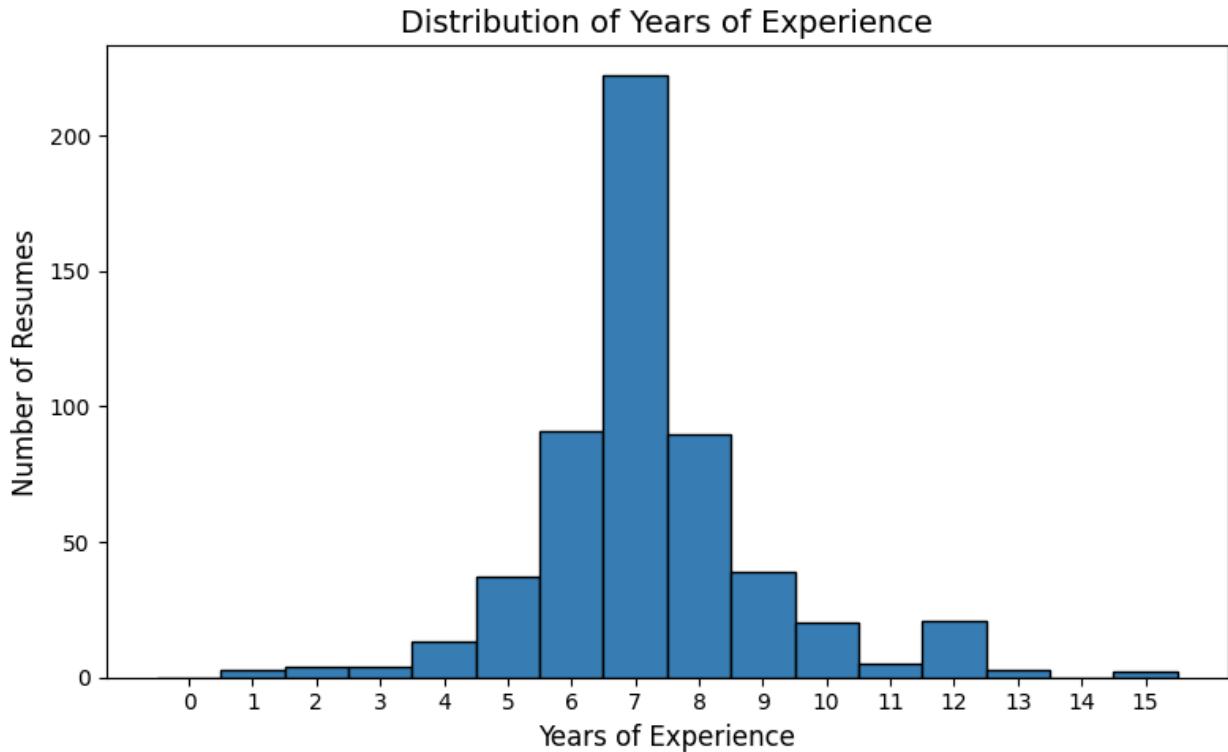
This analysis examined the number of resumes in the dataset by gender. There were 315 female resumes and 282 male resumes included in the dataset.



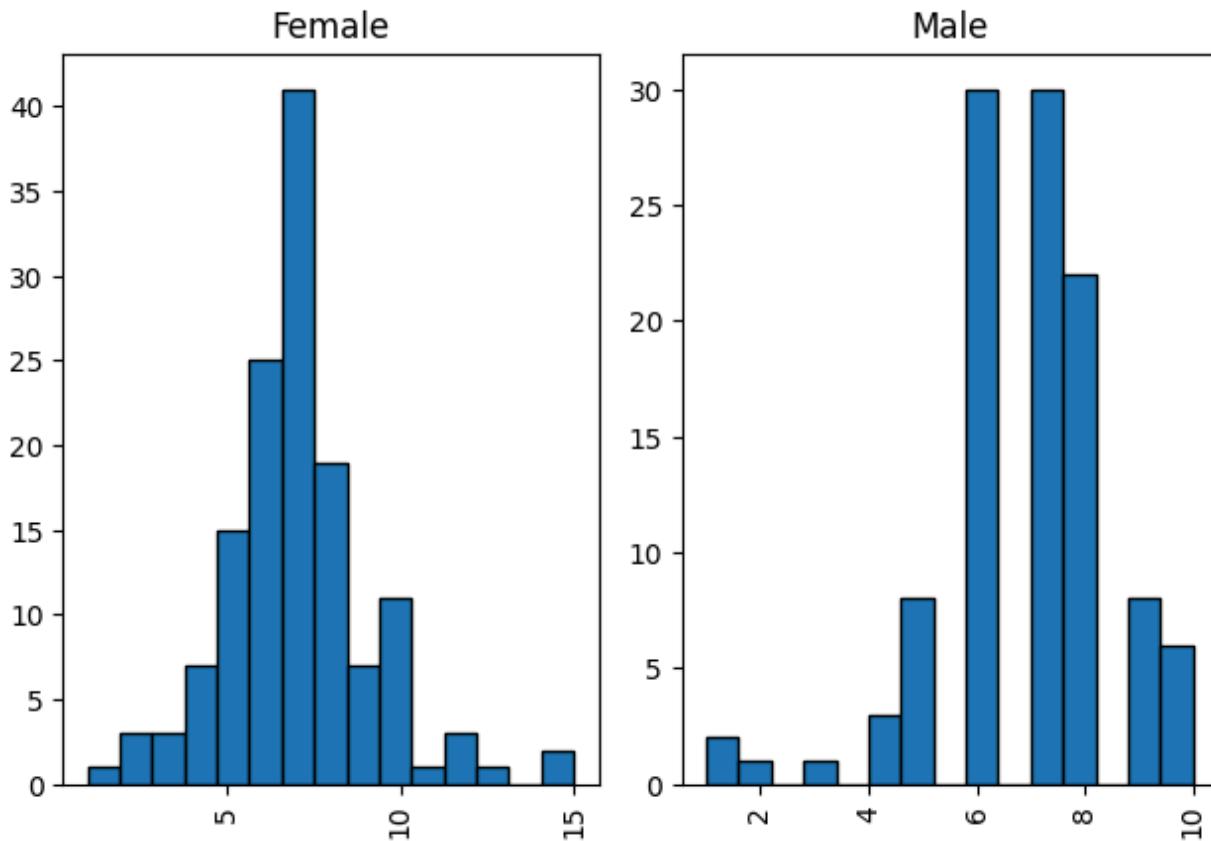
This visualization shows that most resumes have an average length of between 1000 and 2000 words. Very few resumes surpass 3000 words. Less than 10 resumes had below 500 words.



This analysis focused on the top 20 technical skills, with separate bars for female and male counts for each skill. In all skill categories except for C, C++, jdbc, and pl, there were more females than males with the skills.



## Distribution of Years of Experience by Gender



This analysis focused on the distribution of years of experience, with an additional analysis showing the distribution of years of experience by gender. Overall, years of experience were normally distributed in the dataset, with the average number of years of experience for applications being 7.

Years of experience were more normally distributed for women than men, with a slight left skew for women and right skew for men. On average, men had more years of experience than women.

## Formatting

All parts of the Exploratory Data Analysis should be professionally formatted. For example, this means labeling plots and figures, and using data preparation guidelines we went over in class (especially tidy and clean data we went over in Class 3). **The final product should be a .ipynb file. Both your original data file(s) and your cleaned data file(s) should be in your github repo, named accordingly.**

**You should comment your code!**

**You must suppress all warnings and messages.**

**All plots must be professional in appearance, including meaningful axes and legend label and titles.**

**For the Exploratory Data Analysis assignment, you must display your code in the rendered output.**

**Please make sure that your code and plots are accessible. This means including alt-text for all plots and figures (this is in Description[]{} in LaTex), using color-blind friendly color palettes, and using patterns or textures that are noticeable in black-and-white! You will lose points for not doing this. [This site](#) and [this site](#) can help provide guidelines.**

Style and format do count for this assignment, so please take the time to make sure that everything looks good and that your data and code are properly formatted.

## **Repo organization**

You should commit to your repo regularly as you work on your project, and you should keep your repo well organized.

## **Grading**

Submit your Exploratory Data Analysis & Data Preparation assignment by **Wednesday, Oct 15 by 11:59 PM.**

The Exploratory Data Analysis & Data Preparation assignment will be graded as follows:

Total	50 pts
<b>Exploring the data</b>	10 pts
<b>Data preparation and cleaning</b>	10 pts
<b>Codebook</b>	5 pts
<b>Data visualizations</b>	20 pts
<b>Formatting</b>	4 pts
<b>Repo organization</b>	1 pt