**Jodi’s Script!**

Slide 3: Good morning everyone, my name is Jodi Pafford from Aurora, Colorado and I am NOT a Data Scientist, but I am an innovative analyzer who loves jigsaw puzzles, doing homework (NOT hand gesture), and watching it snow outside.

Slide 13: To build a dataset, we used the python library, Beautiful Soup, to scrape indeed.com. Beautiful Soup allowed us to go from the HTML code within the website to a nice clean csv export.

Slide 14: Our dataset was created by searching 5 different job titles: Data Scientist, Data Analyst, Database Administrator, Software Engineer, and Statistician. After completing the webscape, we removed any job postings that were not one of those, such as Machine Learning Engineer or Solutions Analyst.

Slide 15: During the webscrape, we also specifically searched 16 cities across the United States. We ended up with 8,738 unique job postings in our dataset. Of those, we had a total of 1,742 Data Scientist job Postings pulled.

Slide 16: California had the most Data Scientist job postings with 517.

Although Texas had the second highest number of total postings collected, New York, Washington state, and Massachusetts all had more Data Scientist job postings compared to Texas at 170.

Slide 17: Once we had all the data, we had to ensure it was all usable. Indeed.com allows companies to post job postings in any format. This meant that the job posting often included “Equal Opportunity” Language, company information, and varying use of styles making the use of Beautiful Soup Difficult.

Slide 18: To circumvent this, we used BeatifulSoup to extract job postings with bold titles and bulleted items beneath them. We used: Education, Qualifications, Responsibilities, Requirements, and Skills. We then combined all of these into one corpus.

Slide 19: At that time, we were able to use the Python libraries, NLTK and spaCy to remove stop words and eliminate extra items symbols and line breaks. We were finally ready to analyze the data.

Slide 20: NOT SURE HOW TO TIE THIS SLIDE IN WITH MINE. IT REALLY SHOULD GO AFTER THE KNN, RIGHT?? I can talk about it here, but there’d be no explanation as to why I did a Venn Diagram with just these three job titles.

Slide 30: When considering the applying the insights of our findings, it is important to consider how it might be used negatively. One of the ethical considerations of this type of work is to consider that employers may use it to profile candidates. (Need one more very short sentence here)

Slide 31: Additionally, this type of analysis could lead to a person cheating the automated Applicant Tracking System by falsifying information. We know both these things happen in real life, but we feel that this project may contribute to that line of thinking.