In the market for a job post-grad? Our analysis of LinkedIn's most in-demand job skills in the current market may prove to help you out...

#### Outline

- 1. Motivation
- 2. Data
- 3. Data Analysis Plan
- 4. Results
- 5. Next Steps



### Why is this relevant?

- *Context:* We are all players in the job market—most of us are in the thick of the job search right now. We wanted to give all of you a competitive advantage by analyzing the most in-demand skills in said current job market.
- Research Question: What are the most in-demand skills in the current job market based on LinkedIn job postings in 2023?
- *Hypothesis:* The IT, sale, and management skills are the most in demand in the current job market because they have the most frequent appearances in LinkedIn job descriptions and their high associated salary.
- *Model Approach:* We initially planned to analyze skill frequency in job descriptions by counting the occurrences of specific skills using text analysis. This analysis was conducted in two different ways, utilizing word frequency and entity recognition for cross-reference. Additionally, we aimed to apply linear regression to assess the relationship between skill frequency and salary, defining "in-demand" as both frequently requested and high paying. If time permitted, our final objective was to predict salary ranges based on individuals' skill sets using a categorical prediction model, possibly involving DecisionTree algorithms.

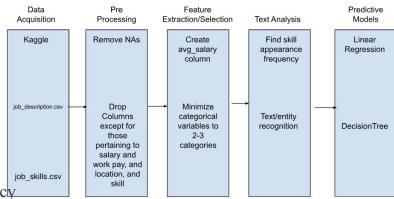
# Data Acquisition & Explanation

- Data set from Kaggle titled "LinkedIn Job Postings 2023"
- The original dataset consisted of 8 CSV files, which were subsequently consolidated into two: "job\_postings" and "job\_skills." The "job\_postings" CSV contained information such as salary details, location, job descriptions from LinkedIn, and a total of 10,956 unique job titles and 13,856 unique job descriptions. Meanwhile, the "job\_skills" CSV contained 15,886 rows with 27 columns, and it had 128,213 missing values. Notably, the "median\_salary" feature had the highest number of missing values, prompting the need to create a new feature from the available data. Additionally, concerns were raised about the "pay\_period" column, where 41% of values were either "YEARLY" or unspecified, with the remaining 59% being null. Furthermore, the "work\_type" column was adjusted by combining "full time" and "contractor" categories into one.
- Included is the data dictionary in which we explored the different variables within the data set.

Feature Name	Description
job_id	Job ID as defined by LinkedIn
description	Job Description
max_salary	Maximum Salary
med_salary	Median Salary
min_salary	Minimum Salary
pay_period	Pay Period
currency	Currency of Salary
compensation_type	Compensation Type
skill_abr	Abbreviation of skill from the skills CSV
work_type	Type of Work
title	Job Title
avg_salary	Average salary based on given maximum and minimum

### Analysis Plan & Justification

- Ensure data consistency, handling exceptions where needed.
- Create an "average salary" column combining min and max salaries to replace median.
- Compile a skills database for machine learning training, matching job skills CSV.
- Train an entity recognition model to extract skills from job descriptions (Model 1).
- Calculate skill frequency across the dataset using recognized skills (Model 2).
- Use linear regression to gauge skill demand based on frequency and wage.
- Predict wages using a DecisionTree model with three randomly generated skills.



## Tricky Analysis Decision!

- Skills (ex. traits, job-skills) versus numeric data (ex. salary) or descriptors (ex. "job description")
  - o Categorical vs. numeric vs. textual variables and showing that relationship
- Lots of missing values; create new feature from remaining values and pre-existing values
- Combine work\_type column with two categories full\_time and contractor into one

# Bias & Uncertainty Validation

- Handling missing data
- Jobs only include *paid* jobs; this may leave out skills geared towards student internships that are unpaid
- Risk of overfitting
- Outliers are present in our results
- Use of synonyms that indicate a similar meaning in regards to skills, jobs, or other relevant variables

### Results & Conclusions

#### Named Entity Recognition

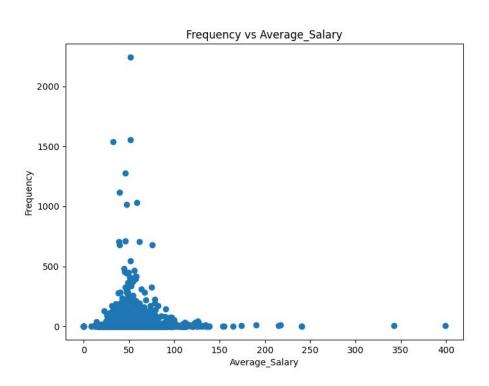
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** support and resolve help desk (centification) issues tickets by gathering information (son axiii) to identify needs security and industry (med skiii) best practice (mod skiii) solutions and deploys pc server hardware and software including operating system (mod skiii) applications and patches – possible nights and weekend work to support and maintain (mod skiii) throughout haco us * supports troubleshoots repairs and maintains (mod skiii) server issues and security for enterprise and large scale (mod skiii) applications * provides actively management (son skiii) assistance to staff in managing (son skiii) user and computer accounts (mod skiii) * troubleshoots reviews and resolves blocked website related to web web management (mod skiii) web filter (mod skiii) management (mod skiii) management (mod skiii) management (mod skiii) monitor systems (mod skiii) networks diagnose and resolve complex software and networking issues a networks support for device device management (mod skiii) installs (mod skiii) and server hardware and components such as disks memory and other components components models (mod skiii) implementing them on production production system (mod skiii) and server hardware and components such as disks memory and other components components models (mod skiii) implementing them on production production system (mod skiii) network network access (mod skiii) infrastructure (son skiii) hardward group architecture account account (mod skiii) and maintains (mod skiii) and maintains (mod skiii) and maintains (mod skiii) and written technical documentation (mod skiii) and maintains (mod skiii) and then device management (mod skiii) and written technical documentation (mod skiii) and maintains (mod skiii) and uthentication (mod skiii) and u
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Word	Frequency
com	2243
problem solve	1557
customer service	1540
job description	1275
microsoft office	1115
project management	1030
write communication	1013
reach	709
organizational skill	708
San	703
computer science	679
time management	677
Act	543
verbal communication skill	481
process improvement	466
detail orient	457
self starter	449
customer experience	442
employee benefit	420
act	416
analytical skill	413
supply chain	408
management system	405
business development	397
decision make	375
diversity and inclusion	366
CRM	350
mental health	337
financial service	336
employee assistance program	331

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## Further Inquiry...



## Next Steps...our recommendation!

#### **Improvements:**

- Removing outliers or influential observations
- Control for other predictors in the linear regression (such as job location, experience level, education level, etc.)

#### **New Questions:**

- Why did frequency and average salary diverge in that pattern?
- How does LinkedIn recognize skills?

#### Become well-equipped with these skills:

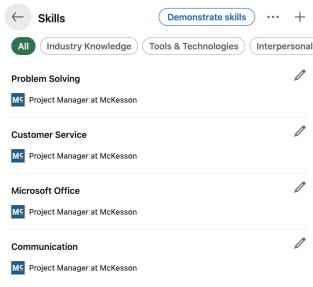
- Problem solving
- Communication
- Organization
- Customer service
- Microsoft Office
- Computer Science



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### References & Resources

Anas Aito. 2022. Skill Ner. Github. <a href="https://github.com/AnasAito/SkillNER">https://github.com/AnasAito/SkillNER</a>

https://github.com/ajzorn/DS4002Project1/tree/main