

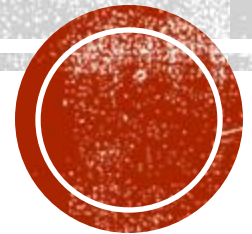
KEY RESUME ATTRIBUTES IMPACTING JOB CALLBACKS

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<https://github.com/anyfruit/Key-Resume-Attributes-Impacting-Job-Callbacks.git>



INTRO



Problem to Solve

- **Objective:** Attributes influential in determining *job application callback rates*
- **Key Question:** Can we predict factors increase the likelihood of receiving callback?



Why This is Important

- Guiding applicants in optimizing their resumes
- Beneficial to help addressing bias in hiring practices
- Improving fairness and transparency in hiring processes



Type of Problem

- Binary classification problem; target variable is either 1 or 0

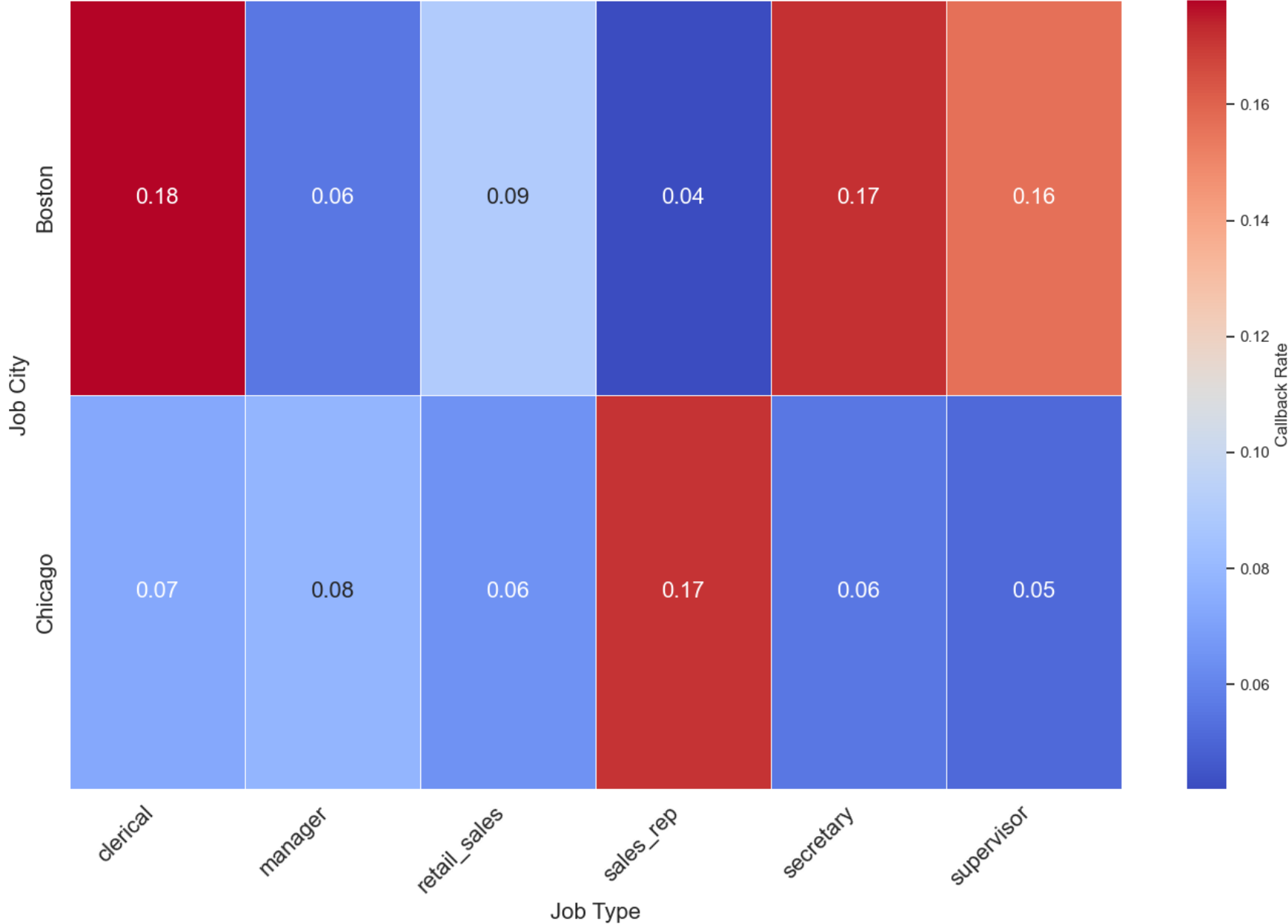


Data Collection

- Kaggle Site: <https://www.kaggle.com/datasets/utkarshx27/which-resume-attributes-drive-job-callbacks>
- Monitored job postings in Boston and Chicago



Callback Rate by Job City and Job Type



EDA

VISUALIZATION 1
(HEATMAP)

- Significant geographic and job-specific disparities
- Boston shows higher callback rates across most jobs

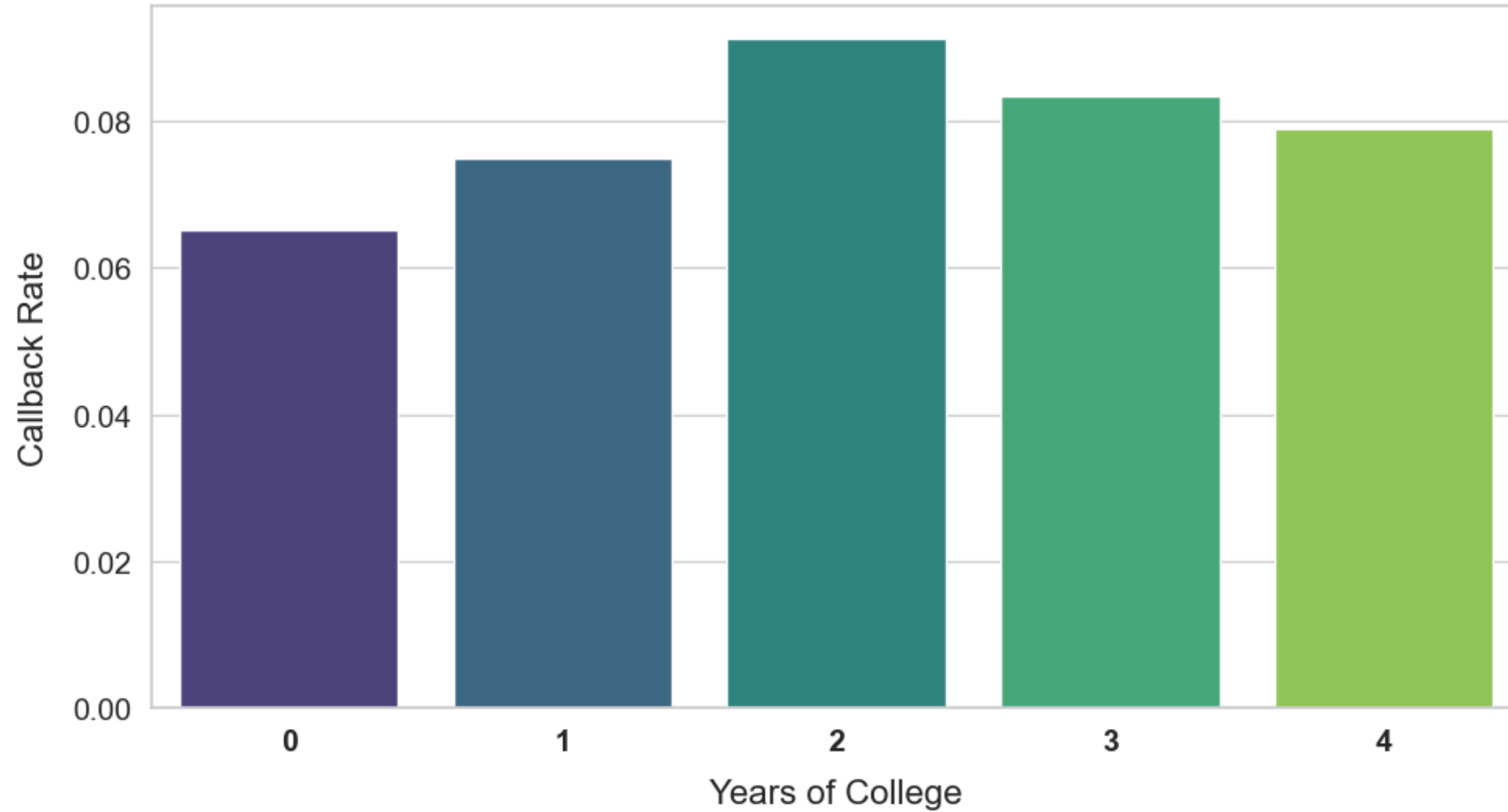


EDA

VISUALIZATION 2 (HISTOGRAM)

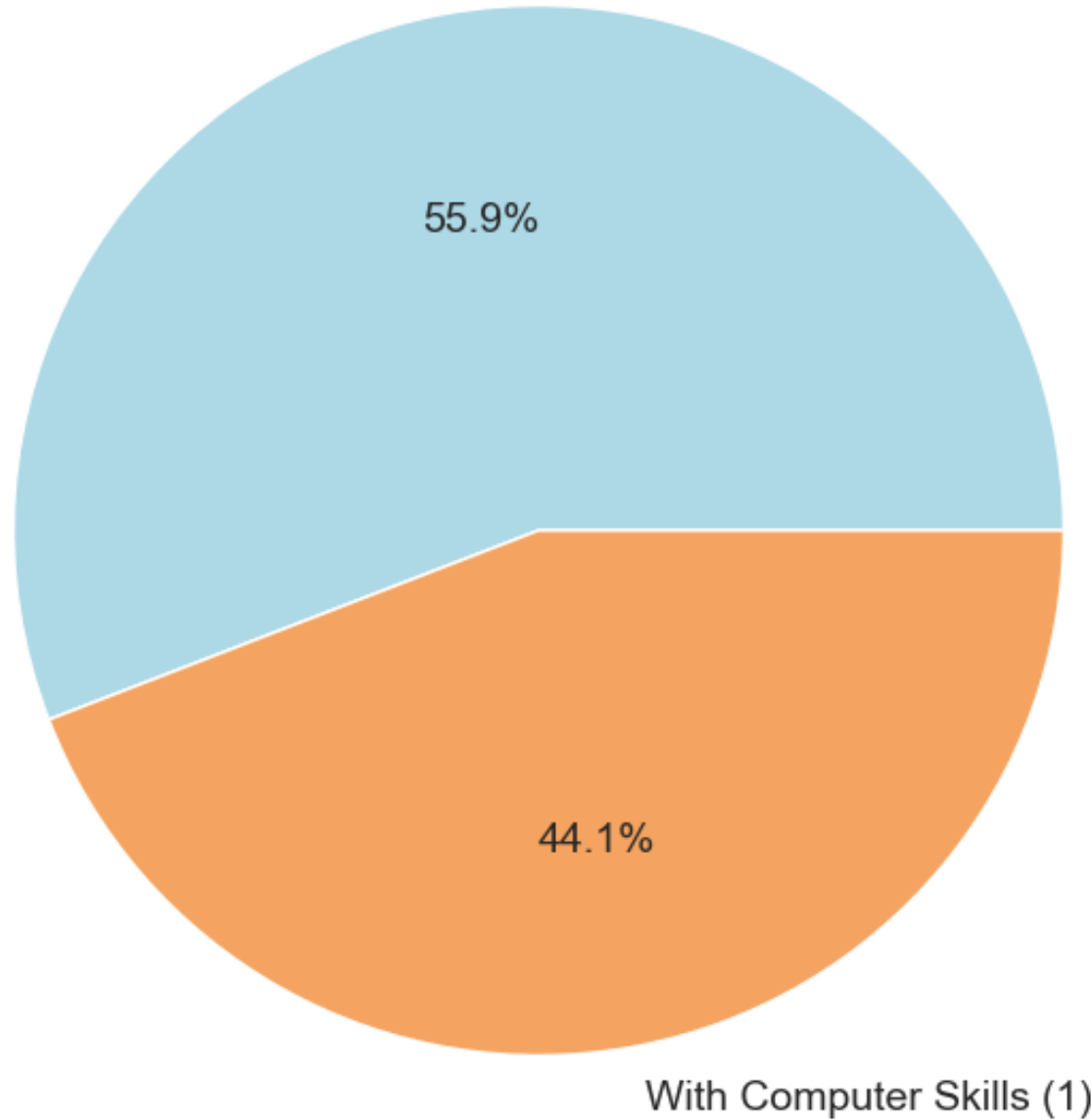
- Unexpected trend with college experience
- Candidates with 2 years of college experience have highest callback rate
- Surprisingly, those with 4 years show lower callback rates

Received Callback Rate by Years of College



Callback Rate Distribution for Computer Skills

No Computer Skills (0)



With Computer Skills (1)

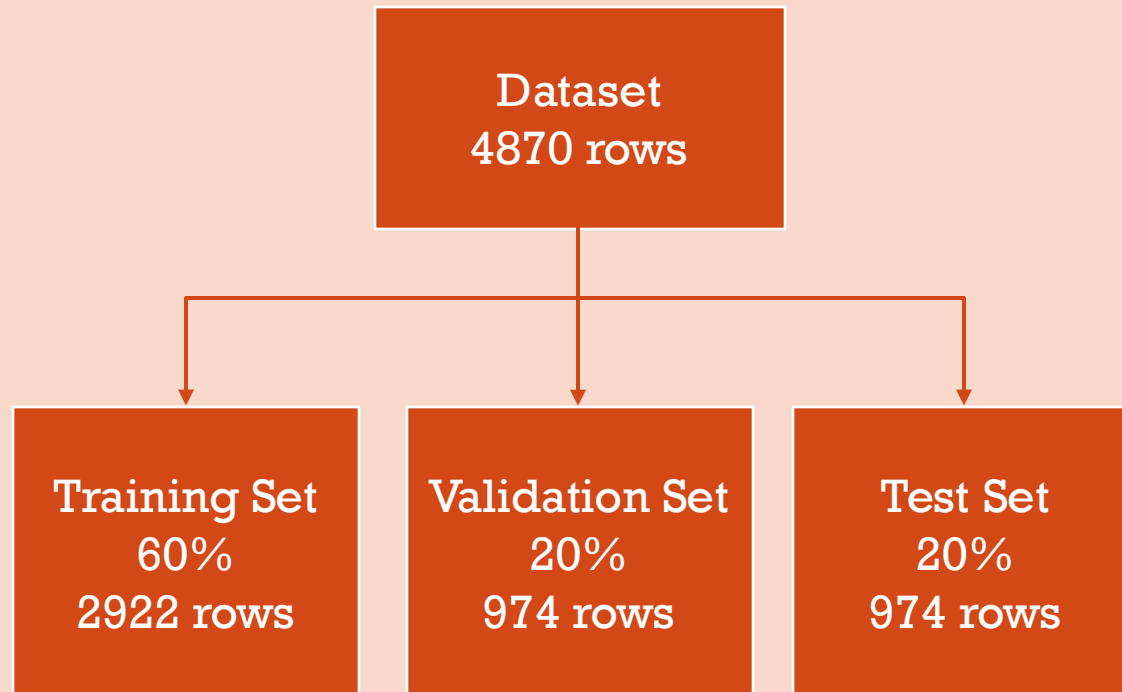
EDA

VISUALIZATION 3 (PIE CHART)

- 55.9% of callbacks are for applicants without listed computer skills
- 44.1% of callbacks are for those with computer skills



SPLITTING



Features excluded:

firstname

*not relevant for predicting callbacks

job_ad_id

*not relevant for predicting callbacks



PREPROCESSING

Handling Missing Data

- Missing values in 3 categorical columns 'UNK', 'Unknown', and NaN were standardized to "unknown" to maintain consistency

Feature Name	Missing Values	Count	%
job_fed_contractor	NA	1768	36.3
job_ownership	Unknown	1992	40.9
job_req_min_experience	NaN	2746	56.4



Renamed Values	
1768	unknown
1992	unknown
2746	unknown



FEATURE ENCODING



Ordinal Encoding

- Applied to features with a natural order
- Example: years_college:
 - 0 ▪ 1 ▪ 2 ▪ 3 ▪ 4



One-Hot Encoding

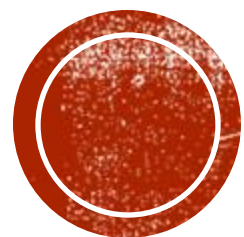
- Used for categorical features with no inherent order
- Example: job_ownership
 - unknown
 - nonprofit
 - private
 - public



PREPROCESSING

- **Before Preprocessing**
 - 27 features
- **After Preprocessing**
 - 60 features (one-hot encoding expansion)
- **Final Training Dataset**
 - Contains 2922 data points with 60 features





THANK YOU FOR LISTENING

